

Principles of Model Checking

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## PRINCIPLES OF

## Model Checking

Christel Baier<br>Joost-Pieter Katoen<br>The MIT Press<br>Cambridge, Massachusetts<br>London, England

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To Michael, Gerda, Inge, and Karl

To Erna, Fons, Joost, and Tom

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## Foreword

Society is increasingly dependent on dedicated computer and software systems to assist us in almost every aspect of daily life. Often we are not even aware that computers and software are involved. Several control functions in modern cars are based on embedded software solutions, e.g., braking, airbags, cruise control, and fuel injection. Mobile phones, communication systems, medical devices, audio and video systems, and consumer electronics in general are containing vast amounts of software. Also transport, production, and control systems are increasingly applying embedded software solutions to gain flexibility and cost-efficiency.

A common pattern is the constantly increasing complexity of systems, a trend which is accelerated by the adaptation of wired and wireless networked solutions: in a modern car the control functions are distributed over several processing units communicating over dedicated networks and buses. Yet computer- and software-based solutions are becoming ubiquitous and are to be found in several safety-critical systems. Therefore a main challenge for the field of computer science is to provide formalisms, techniques, and tools that will enable the efficient design of correct and well-functioning systems despite their complexity.

Over the last two decades or so a very attractive approach toward the correctness of computer-based control systems is that of model checking. Model checking is a formal verification technique which allows for desired behavioral properties of a given system to be verified on the basis of a suitable model of the system through systematic inspection of all states of the model. The attractiveness of model checking comes from the fact that it is completely automatic - i.e., the learning curve for a user is very gentle - and that it offers counterexamples in case a model fails to satisfy a property serving as indispensable debugging information. On top of this, the performance of model-checking tools has long since proved mature as witnessed by a large number of successful industrial applications.

It is my pleasure to recommend the excellent book Principles of Model Checking by Christel Baier and Joost-Pieter Katoen as the definitive textbook on model checking, providing both a comprehensive and a comprehensible account of this important topic. The book contains detailed and complete descriptions of first principles of classical Linear Temporal Logic (LTL) and Computation Tree Logic (CTL) model checking. Also, state-of-the art methods for coping with state-space explosion, including symbolic model checking, abstraction and minimization techniques, and partial order reduction, are fully accounted for. The book also covers model checking of real-time and probabilistic systems, important new directions for model checking in which the authors, being two of the most industrious and creative researchers of today, are playing a central role.

The exceptional pedagogical style of the authors provides careful explanations of constructions and proofs, plus numerous examples and exercises of a theoretical, practical and tool-oriented nature. The book will therefore be the ideal choice as a textbook for both graduate and advanced undergraduate students, as well as for self-study, and should definitely be on the bookshelf of any researcher interested in the topic.

Kim Guldstrand Larsen Professor in Computer Science Aalborg University, Denmark May 2007

## Preface

It is fair to state, that in this digital era correct systems for information processing are more valuable than gold.
(H. Barendregt. The quest for correctness.

In: Images of SMC Research 1996, pages 39-58, 1996.)

This book is on model checking, a prominent formal verification technique for assessing functional properties of information and communication systems. Model checking requires a model of the system under consideration and a desired property and systematically checks whether or not the given model satisfies this property. Typical properties that can be checked are deadlock freedom, invariants, and request-response properties. Model checking is an automated technique to check the absence of errors (i.e., property violations) and alternatively can be considered as an intelligent and effective debugging technique. It is a general approach and is applied in areas like hardware verification and software engineering. Due to unremitting improvements of underlying algorithms and data structures together with hardware technology improvements, model-checking techniques that two decades ago only worked for simple examples are nowadays applicable to more realistic designs. It is fair to say that in the last two decades model checking has developed as a mature and heavily used verification and debugging technique.

## Aims and Scope

This book attempts to introduce model checking from first principles, so to speak, and is intended as a textbook for bachelor and master students, as well as an introductory book for researchers working in other areas of computer science or related fields. The reader is introduced to the material by means of an extensive set of examples, most of which are examples running throughout several chapters. The book provides a complete set of basic results together with all detailed proofs. Each chapter is concluded by a summary,
bibliographic notes, and a series of exercises, of both a theoretical and of a practical nature (i.e., experimenting with actual model checkers).

## Prerequisites

The concepts of model checking have their roots in mathematical foundations such as propositional logic, automata theory and formal languages, data structures, and graph algorithms. It is expected that readers are familiar with the basics of these topics when starting with our book, although an appendix is provided that summarizes the essentials. Knowledge on complexity theory is required for the theoretical complexity considerations of the various model-checking algorithms.

## Content

This book is divided into ten chapters. Chapter 1 motivates and introduces model checking. Chapter 2 presents transition systems as a model for software and hardware systems. Chapter 3 introduces a classification of linear-time properties into safety and liveness, and presents the notion of fairness. Automata-based algorithms for checking (regular) safety and $\omega$-regular properties are presented in Chapter 4. Chapter 5 deals with Linear Temporal Logic (LTL) and shows how the algorithms of Chapter 4 can be used for LTL model checking. Chapter 6 introduces the branching-time temporal logic Computation Tree Logic (CTL), compares this to LTL, and shows how to perform CTL model checking, both explicitly and symbolically. Chapter 7 deals with abstraction mechanisms that are based on trace, bisimulation, and simulation relations. Chapter 8 treats partial-order reduction for LTL and CTL. Chapter 9 is focused on real-time properties and timed automata, and the monograph is concluded with a chapter on the verification of probabilistic models. The appendix summarizes basic results on propositional logic, graphs, language, and complexity theory.

## How to Use This Book

A natural plan for an introductory course into model checking that lasts one semester (two lectures a week) comprises Chapters 1 through 6. A follow-up course of about a semester could cover Chapters 7 through 10, after a short refresher on LTL and CTL model checking.

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## Chapter 1

## System Verification

Our reliance on the functioning of ICT systems (Information and Communication Technology) is growing rapidly. These systems are becoming more and more complex and are massively encroaching on daily life via the Internet and all kinds of embedded systems such as smart cards, hand-held computers, mobile phones, and high-end television sets. In 1995 it was estimated that we are confronted with about 25 ICT devices on a daily basis. Services like electronic banking and teleshopping have become reality. The daily cash flow via the Internet is about $10^{12}$ million US dollars. Roughly $20 \%$ of the product development costs of modern transportation devices such as cars, high-speed trains, and airplanes is devoted to information processing systems. ICT systems are universal and omnipresent. They control the stock exchange market, form the heart of telephone switches, are crucial to Internet technology, and are vital for several kinds of medical systems. Our reliance on embedded systems makes their reliable operation of large social importance. Besides offering a good performance in terms like response times and processing capacity, the absence of annoying errors is one of the major quality indications.

It is all about money. We are annoyed when our mobile phone malfunctions, or when our video recorder reacts unexpectedly and wrongly to our issued commands. These software and hardware errors do not threaten our lives, but may have substantial financial consequences for the manufacturer. Correct ICT systems are essential for the survival of a company. Dramatic examples are known. The bug in Intel's Pentium II floating-point division unit in the early nineties caused a loss of about 475 million US dollars to replace faulty processors, and severely damaged Intel's reputation as a reliable chip manufacturer. The software error in a baggage handling system postponed the opening of Denver's airport for 9 months, at a loss of 1.1 million US dollar per day. Twenty-four hours of failure of


Figure 1.1: The Ariane-5 launch on June 4, 1996; it crashed 36 seconds after the launch due to a conversion of a 64 -bit floating point into a 16 -bit integer value.
the worldwide online ticket reservation system of a large airplane company will cause its bankruptcy because of missed orders.

It is all about safety: errors can be catastrophic too. The fatal defects in the control software of the Ariane-5 missile (Figure 1.1), the Mars Pathfinder, and the airplanes of the Airbus family led to headlines in newspapers all over the world and are notorious by now. Similar software is used for the process control of safety-critical systems such as chemical plants, nuclear power plants, traffic control and alert systems, and storm surge barriers. Clearly, bugs in such software can have disastrous consequences. For example, a software flaw in the control part of the radiation therapy machine Therac- 25 caused the death of six cancer patients between 1985 and 1987 as they were exposed to an overdose of radiation.

The increasing reliance of critical applications on information processing leads us to state:

The reliability of ICT systems is a key issue in the system design process.

The magnitude of ICT systems, as well as their complexity, grows apace. ICT systems are no longer standalone, but are typically embedded in a larger context, connecting and interacting with several other components and systems. They thus become much more vulnerable to errors - the number of defects grows exponentially with the number of interacting system components. In particular, phenomena such as concurrency and nondeterminism that are central to modeling interacting systems turn out to be very hard to handle with standard techniques. Their growing complexity, together with the pressure to drastically reduce system development time ("time-to-market"), makes the delivery of low-defect ICT systems an enormously challenging and complex activity.

## Hard- and Software Verification

System verification techniques are being applied to the design of ICT systems in a more reliable way. Briefly, system verification is used to establish that the design or product under consideration possesses certain properties. The properties to be validated can be quite elementary, e.g., a system should never be able to reach a situation in which no progress can be made (a deadlock scenario), and are mostly obtained from the system's specification. This specification prescribes what the system has to do and what not, and thus constitutes the basis for any verification activity. A defect is found once the system does not fulfill one of the specification's properties. The system is considered to be "correct" whenever it satisfies all properties obtained from its specification. So correctness is always relative to a specification, and is not an absolute property of a system. A schematic view of verification is depicted in Figure 1.2.


Figure 1.2: Schematic view of an a posteriori system verification.

This book deals with a verification technique called model checking that starts from a formal system specification. Before introducing this technique and discussing the role of formal specifications, we briefly review alternative software and hardware verification techniques.

Software Verification Peer reviewing and testing are the major software verification techniques used in practice.

A peer review amounts to a software inspection carried out by a team of software engineers that preferably has not been involved in the development of the software under review. The
uncompiled code is not executed, but analyzed completely statically. Empirical studies indicate that peer review provides an effective technique that catches between $31 \%$ and $93 \%$ of the defects with a median around $60 \%$. While mostly applied in a rather ad hoc manner, more dedicated types of peer review procedures, e.g., those that are focused at specific error-detection goals, are even more effective. Despite its almost complete manual nature, peer review is thus a rather useful technique. It is therefore not surprising that some form of peer review is used in almost $80 \%$ of all software engineering projects. Due to its static nature, experience has shown that subtle errors such as concurrency and algorithm defects are hard to catch using peer review.

Software testing constitutes a significant part of any software engineering project. Between $30 \%$ and $50 \%$ of the total software project costs are devoted to testing. As opposed to peer review, which analyzes code statically without executing it, testing is a dynamic technique that actually runs the software. Testing takes the piece of software under consideration and provides its compiled code with inputs, called tests. Correctness is thus determined by forcing the software to traverse a set of execution paths, sequences of code statements representing a run of the software. Based on the observations during test execution, the actual output of the software is compared to the output as documented in the system specification. Although test generation and test execution can partly be automated, the comparison is usually performed by human beings. The main advantage of testing is that it can be applied to all sorts of software, ranging from application software (e.g., e-business software) to compilers and operating systems. As exhaustive testing of all execution paths is practically infeasible; in practice only a small subset of these paths is treated. Testing can thus never be complete. That is to say, testing can only show the presence of errors, not their absence. Another problem with testing is to determine when to stop. Practically, it is hard, and mostly impossible, to indicate the intensity of testing to reach a certain defect density - the fraction of defects per number of uncommented code lines.

Studies have provided evidence that peer review and testing catch different classes of defects at different stages in the development cycle. They are therefore often used together. To increase the reliability of software, these software verification approaches are complemented with software process improvement techniques, structured design and specification methods (such as the Unified Modeling Language), and the use of version and configuration management control systems. Formal techniques are used, in one form or another, in about $10 \%$ to $15 \%$ of all software projects. These techniques are discussed later in this chapter.

Catching software errors: the sooner the better. It is of great importance to locate software bugs. The slogan is: the sooner the better. The costs of repairing a software flaw during maintenance are roughly 500 times higher than a fix in an early design phase (see Figure 1.3). System verification should thus take place early stage in the design process.


Figure 1.3: Software lifecycle and error introduction, detection, and repair costs [275].

About $50 \%$ of all defects are introduced during programming, the phase in which actual coding takes place. Whereas just $15 \%$ of all errors are detected in the initial design stages, most errors are found during testing. At the start of unit testing, which is oriented to discovering defects in the individual software modules that make up the system, a defect density of about 20 defects per 1000 lines of (uncommented) code is typical. This has been reduced to about 6 defects per 1000 code lines at the start of system testing, where a collection of such modules that constitutes a real product is tested. On launching a new software release, the typical accepted software defect density is about one defect per 1000 lines of code lines ${ }^{1}$.

Errors are typically concentrated in a few software modules - about half of the modules are defect free, and about $80 \%$ of the defects arise in a small fraction (about 20\%) of the modules - and often occur when interfacing modules. The repair of errors that are detected prior to testing can be done rather economically. The repair cost significantly increases from about $\$ 1000$ (per error repair) in unit testing to a maximum of about \$ 12,500 when the defect is demonstrated during system operation only. It is of vital importance to seek techniques that find defects as early as possible in the software design process: the costs to repair them are substantially lower, and their influence on the rest of the design is less substantial.

Hardware Verification Preventing errors in hardware design is vital. Hardware is subject to high fabrication costs; fixing defects after delivery to customers is difficult, and quality expectations are high. Whereas software defects can be repaired by providing

[^0]users with patches or updates - nowadays users even tend to anticipate and accept this hardware bug fixes after delivery to customers are very difficult and mostly require refabrication and redistribution. This has immense economic consequences. The replacement of the faulty Pentium II processors caused Intel a loss of about $\$ 475$ million. Moore's law - the number of logical gates in a circuit doubles every 18 months - has proven to be true in practice and is a major obstacle to producing correct hardware. Empirical studies have indicated that more than $50 \%$ of all ASICs (Application-Specific Integrated Circuits) do not work properly after initial design and fabrication. It is not surprising that chip manufacturers invest a lot in getting their designs right. Hardware verification is a well-established part of the design process. The design effort in a typical hardware design amounts to only $27 \%$ of the total time spent on the chip; the rest is devoted to error detection and prevention.

Hardware verification techniques. Emulation, simulation, and structural analysis are the major techniques used in hardware verification.

Structural analysis comprises several specific techniques such as synthesis, timing analysis, and equivalence checking that are not described in further detail here.

Emulation is a kind of testing. A reconfigurable generic hardware system (the emulator) is configured such that it behaves like the circuit under consideration and is then extensively tested. As with software testing, emulation amounts to providing a set of stimuli to the circuit and comparing the generated output with the expected output as laid down in the chip specification. To fully test the circuit, all possible input combinations in every possible system state should be examined. This is impractical and the number of tests needs to be reduced significantly, yielding potential undiscovered errors.

With simulation, a model of the circuit at hand is constructed and simulated. Models are typically provided using hardware description languages such as Verilog or VHDL that are both standardized by IEEE. Based on stimuli, execution paths of the chip model are examined using a simulator. These stimuli may be provided by a user, or by automated means such as a random generator. A mismatch between the simulator's output and the output described in the specification determines the presence of errors. Simulation is like testing, but is applied to models. It suffers from the same limitations, though: the number of scenarios to be checked in a model to get full confidence goes beyond any reasonable subset of scenarios that can be examined in practice.

Simulation is the most popular hardware verification technique and is used in various design stages, e.g., at register-transfer level, gate and transistor level. Besides these error detection techniques, hardware testing is needed to find fabrication faults resulting from layout defects in the fabrication process.

### 1.1 Model Checking

In software and hardware design of complex systems, more time and effort are spent on verification than on construction. Techniques are sought to reduce and ease the verification efforts while increasing their coverage. Formal methods offer a large potential to obtain an early integration of verification in the design process, to provide more effective verification techniques, and to reduce the verification time.

Let us first briefly discuss the role of formal methods. To put it in a nutshell, formal methods can be considered as "the applied mathematics for modeling and analyzing ICT systems". Their aim is to establish system correctness with mathematical rigor. Their great potential has led to an increasing use by engineers of formal methods for the verification of complex software and hardware systems. Besides, formal methods are one of the "highly recommended" verification techniques for software development of safetycritical systems according to, e.g., the best practices standard of the IEC (International Electrotechnical Commission) and standards of the ESA (European Space Agency). The resulting report of an investigation by the FAA (Federal Aviation Authority) and NASA (National Aeronautics and Space Administration) about the use of formal methods concludes that

Formal methods should be part of the education of every computer scientist and software engineer, just as the appropriate branch of applied maths is a necessary part of the education of all other engineers.

During the last two decades, research in formal methods has led to the development of some very promising verification techniques that facilitate the early detection of defects. These techniques are accompanied by powerful software tools that can be used to automate various verification steps. Investigations have shown that formal verification procedures would have revealed the exposed defects in, e.g., the Ariane-5 missile, Mars Pathfinder, Intel's Pentium II processor, and the Therac- 25 therapy radiation machine.

Model-based verification techniques are based on models describing the possible system behavior in a mathematically precise and unambiguous manner. It turns out that - prior to any form of verification - the accurate modeling of systems often leads to the discovery of incompleteness, ambiguities, and inconsistencies in informal system specifications. Such problems are usually only discovered at a much later stage of the design. The system models are accompanied by algorithms that systematically explore all states of the system model. This provides the basis for a whole range of verification techniques ranging from an exhaustive exploration (model checking) to experiments with a restrictive set of scenarios in the model (simulation), or in reality (testing). Due to unremitting improvements of un-
derlying algorithms and data structures, together with the availability of faster computers and larger computer memories, model-based techniques that a decade ago only worked for very simple examples are nowadays applicable to realistic designs. As the startingpoint of these techniques is a model of the system under consideration, we have as a given fact that

## Any verification using model-based techniques is only as good as the model of the system.

Model checking is a verification technique that explores all possible system states in a brute-force manner. Similar to a computer chess program that checks possible moves, a model checker, the software tool that performs the model checking, examines all possible system scenarios in a systematic manner. In this way, it can be shown that a given system model truly satisfies a certain property. It is a real challenge to examine the largest possible state spaces that can be treated with current means, i.e., processors and memories. State-of-the-art model checkers can handle state spaces of about $10^{8}$ to $10^{9}$ states with explicit state-space enumeration. Using clever algorithms and tailored data structures, larger state spaces ( $10^{20}$ up to even $10^{476}$ states) can be handled for specific problems. Even the subtle errors that remain undiscovered using emulation, testing and simulation can potentially be revealed using model checking.


Figure 1.4: Schematic view of the model-checking approach.

Typical properties that can be checked using model checking are of a qualitative nature: Is the generated result OK?, Can the system reach a deadlock situation, e.g., when two
concurrent programs are waiting for each other and thus halting the entire system? But also timing properties can be checked: Can a deadlock occur within 1 hour after a system reset?, or, Is a response always received within 8 minutes? Model checking requires a precise and unambiguous statement of the properties to be examined. As with making an accurate system model, this step often leads to the discovery of several ambiguities and inconsistencies in the informal documentation. For instance, the formalization of all system properties for a subset of the ISDN user part protocol revealed that $55 \%$ (!) of the original, informal system requirements were inconsistent.

The system model is usually automatically generated from a model description that is specified in some appropriate dialect of programming languages like C or Java or hardware description languages such as Verilog or VHDL. Note that the property specification prescribes what the system should do, and what it should not do, whereas the model description addresses how the system behaves. The model checker examines all relevant system states to check whether they satisfy the desired property. If a state is encountered that violates the property under consideration, the model checker provides a counterexample that indicates how the model could reach the undesired state. The counterexample describes an execution path that leads from the initial system state to a state that violates the property being verified. With the help of a simulator, the user can replay the violating scenario, in this way obtaining useful debugging information, and adapt the model (or the property) accordingly (see Figure 1.4).

Model checking has been successfully applied to several ICT systems and their applications. For instance, deadlocks have been detected in online airline reservation systems, modern ecommerce protocols have been verified, and several studies of international IEEE standards for in-house communication of domestic appliances have led to significant improvements of the system specifications. Five previously undiscovered errors were identified in an execution module of the Deep Space 1 spacecraft controller (see Figure 1.5), in one case identifying a major design flaw. A bug identical to one discovered by model checking escaped testing and caused a deadlock during a flight experiment 96 million km from earth. In the Netherlands, model checking has revealed several serious design flaws in the control software of a storm surge barrier that protects the main port of Rotterdam against flooding.

## Example 1.1. Concurrency and Atomicity

Most errors, such as the ones exposed in the Deep Space-1 spacecraft, are concerned with classical concurrency errors. Unforeseen interleavings between processes may cause undesired events to happen. This is exemplified by analysing the following concurrent program, in which three processes, Inc, Dec, and Reset, cooperate. They operate on the shared integer variable $x$ with arbitrary initial value that can be accessed (i.e., read), and


Figure 1.5: Modules of NASA's Deep Space-1 space-craft (launched in October 1998) have been thoroughly examined using model checking.
modified (i.e., written) by each of the individual processes. The processes are

$$
\begin{aligned}
\text { proc } \operatorname{Inc} & =\text { while true do if } x<200 \text { then } x:=x+1 \text { fi od } \\
\text { proc Dec } & =\text { while true do if } x>0 \text { then } x:=x-1 \mathrm{fi} \text { od } \\
\text { proc Reset } & =\text { while true do if } x=200 \text { then } x:=0 \text { fi od }
\end{aligned}
$$

Process Inc increments $x$ if its value is smaller than 200, Dec decrements $x$ if its value is at least 1, and Reset resets $x$ once it has reached the value 200. They all do so repetitively.

Is the value of $x$ always between (and including) 0 and 200? At first sight this seems to be true. A more thorough inspection, though, reveals that this is not the case. Suppose $x$ equals 200. Process Dec tests the value of $x$, and passes the test, as $x$ exceeds 0 . Then, control is taken over by process Reset. It tests the value of $x$, passes its test, and immediately resets $x$ to zero. Then, control is returned to process Dec and this process decrements $x$ by one, resulting in a negative value for $x$ (viz. -1 ). Intuitively, we tend to interpret the tests on $x$ and the assignments to $x$ as being executed atomically, i.e., as a single step, whereas in reality this is (mostly) not the case.

### 1.2 Characteristics of Model Checking

This book is devoted to the principles of model checking:

> Model checking is an automated technique that, given a finite-state model of a system and a formal property, systematically checks whether this property holds for (a given state in) that model.

The next chapters treat the elementary technical details of model checking. This section describes the process of model checking (how to use it), presents its main advantages and drawbacks, and discusses its role in the system development cycle.

### 1.2.1 The Model-Checking Process

In applying model checking to a design the following different phases can be distinguished:

- Modeling phase:
- model the system under consideration using the model description language of the model checker at hand;
- as a first sanity check and quick assessment of the model perform some simulations;
- formalize the property to be checked using the property specification language.
- Running phase: run the model checker to check the validity of the property in the system model.
- Analysis phase:
- property satisfied? $\rightarrow$ check next property (if any);
- property violated? $\rightarrow$

1. analyze generated counterexample by simulation;
2. refine the model, design, or property;
3. repeat the entire procedure.

- out of memory? $\rightarrow$ try to reduce the model and try again.

In addition to these steps, the entire verification should be planned, administered, and organized. This is called verification organization. We discuss these phases of model checking in somewhat more detail below.

Modeling The prerequisite inputs to model checking are a model of the system under consideration and a formal characterization of the property to be checked.

Models of systems describe the behavior of systems in an accurate and unambiguous way. They are mostly expressed using finite-state automata, consisting of a finite set of states and a set of transitions. States comprise information about the current values of variables, the previously executed statement (e.g., a program counter), and the like. Transitions describe how the system evolves from one state into another. For realistic systems, finite-state automata are described using a model description language such as an appropriate dialect/extension of C, Java, VHDL, or the like. Modeling systems, in particular concurrent ones, at the right abstraction level is rather intricate and is really an art; it is treated in more detail in Chapter 2.

In order to improve the quality of the model, a simulation prior to the model checking can take place. Simulation can be used effectively to get rid of the simpler category of modeling errors. Eliminating these simpler errors before any form of thorough checking takes place may reduce the costly and time-consuming verification effort.

To make a rigorous verification possible, properties should be described in a precise and unambiguous manner. This is typically done using a property specification language. We focus in particular on the use of a temporal logic as a property specification language, a form of modal logic that is appropriate to specify relevant properties of ICT systems. In terms of mathematical logic, one checks that the system description is a model of a temporal logic formula. This explains the term "model checking". Temporal logic is basically an extension of traditional propositional logic with operators that refer to the behavior of systems over time. It allows for the specification of a broad range of relevant system properties such as functional correctness (does the system do what it is supposed to do?), reachability (is it possible to end up in a deadlock state?), safety ("something bad never happens"), liveness ("something good will eventually happen"), fairness (does, under certain conditions, an event occur repeatedly?), and real-time properties (is the system acting in time?).

Although the aforementioned steps are often well understood, in practice it may be a serious problem to judge whether the formalized problem statement (model + properties) is an adequate description of the actual verification problem. This is also known as the validation problem. The complexity of the involved system, as well as the lack of precision
of the informal specification of the system's functionality, may make it hard to answer this question satisfactorily. Verification and validation should not be confused. Verification amounts to check that the design satisfies the requirements that have been identified, i.e., verification is "check that we are building the thing right". In validation, it is checked whether the formal model is consistent with the informal conception of the design, i.e., validation is "check that we are verifying the right thing".

Running the Model Checker The model checker first has to be initialized by appropriately setting the various options and directives that may be used to carry out the exhaustive verification. Subsequently, the actual model checking takes place. This is basically a solely algorithmic approach in which the validity of the property under consideration is checked in all states of the system model.

Analyzing the Results There are basically three possible outcomes: the specified property is either valid in the given model or not, or the model turns out to be too large to fit within the physical limits of the computer memory.

In case the property is valid, the following property can be checked, or, in case all properties have been checked, the model is concluded to possess all desired properties.

Whenever a property is falsified, the negative result may have different causes. There may be a modeling error, i.e., upon studying the error it is discovered that the model does not reflect the design of the system. This implies a correction of the model, and verification has to be restarted with the improved model. This reverification includes the verification of those properties that were checked before on the erroneous model and whose verification may be invalidated by the model correction! If the error analysis shows that there is no undue discrepancy between the design and its model, then either a design error has been exposed, or a property error has taken place. In case of a design error, the verification is concluded with a negative result, and the design (together with its model) has to be improved. It may be the case that upon studying the exposed error it is discovered that the property does not reflect the informal requirement that had to be validated. This implies a modification of the property, and a new verification of the model has to be carried out. As the model is not changed, no reverification of properties that were checked before has to take place. The design is verified if and only if all properties have been checked with respect to a valid model.

Whenever the model is too large to be handled - state spaces of real-life systems may be many orders of magnitude larger than what can be stored by currently available memories - there are various ways to proceed. A possibility is to apply techniques that try to exploit
implicit regularities in the structure of the model. Examples of these techniques are the representation of state spaces using symbolic techniques such as binary decision diagrams or partial order reduction. Alternatively, rigorous abstractions of the complete system model are used. These abstractions should preserve the (non-)validity of the properties that need to be checked. Often, abstractions can be obtained that are sufficiently small with respect to a single property. In that case, different abstractions need to be made for the model at hand. Another way of dealing with state spaces that are too large is to give up the precision of the verification result. The probabilistic verification approaches explore only part of the state space while making a (often negligible) sacrifice in the verification coverage. The most important state-space reduction strategies are discussed in Chapters 7 through 9 of this monograph.

Verification Organization The entire model-checking process should be well organized, well structured, and well planned. Industrial applications of model checking have provided evidence that the use of version and configuration management is of particular relevance. During the verification process, for instance, different model descriptions are made describing different parts of the system, various versions of the verification models are available (e.g., due to abstraction), and plenty of verification parameters (e.g., model-checking options) and results (diagnostic traces, statistics) are available. This information needs to be documented and maintained very carefully in order to manage a practical model-checking process and to allow the reproduction of the experiments that were carried out.

### 1.2.2 Strengths and Weaknesses

The strengths of model checking:

- It is a general verification approach that is applicable to a wide range of applications such as embedded systems, software engineering, and hardware design.
- It supports partial verification, i.e., properties can be checked individually, thus allowing focus on the essential properties first. No complete requirement specification is needed.
- It is not vulnerable to the likelihood that an error is exposed; this contrasts with testing and simulation that are aimed at tracing the most probable defects.
- It provides diagnostic information in case a property is invalidated; this is very useful for debugging purposes.
- It is a potential "push-button" technology; the use of model checking requires neither a high degree of user interaction nor a high degree of expertise.
- It enjoys a rapidly increasing interest by industry; several hardware companies have started their in-house verification labs, job offers with required skills in model checking frequently appear, and commercial model checkers have become available.
- It can be easily integrated in existing development cycles; its learning curve is not very steep, and empirical studies indicate that it may lead to shorter development times.
- It has a sound and mathematical underpinning; it is based on theory of graph algorithms, data structures, and logic.

The weaknesses of model checking:

- It is mainly appropriate to control-intensive applications and less suited for dataintensive applications as data typically ranges over infinite domains.
- Its applicability is subject to decidability issues; for infinite-state systems, or reasoning about abstract data types (which requires undecidable or semi-decidable logics), model checking is in general not effectively computable.
- It verifies a system model, and not the actual system (product or prototype) itself; any obtained result is thus as good as the system model. Complementary techniques, such as testing, are needed to find fabrication faults (for hardware) or coding errors (for software).
- It checks only stated requirements, i.e., there is no guarantee of completeness. The validity of properties that are not checked cannot be judged.
- It suffers from the state-space explosion problem, i.e., the number of states needed to model the system accurately may easily exceed the amount of available computer memory. Despite the development of several very effective methods to combat this problem (see Chapters 7 and 8), models of realistic systems may still be too large to fit in memory.
- Its usage requires some expertise in finding appropriate abstractions to obtain smaller system models and to state properties in the logical formalism used.
- It is not guaranteed to yield correct results: as with any tool, a model checker may contain software defects. ${ }^{2}$

[^1]- It does not allow checking generalizations: in general, checking systems with an arbitrary number of components, or parameterized systems, cannot be treated. Model checking can, however, suggest results for arbitrary parameters that may be verified using proof assistants.

We believe that one can never achieve absolute guaranteed correctness for systems of realistic size. Despite the above limitations we conclude that

> Model checking is an effective technique to expose potential design errors.

Thus, model checking can provide a significant increase in the level of confidence of a system design.

### 1.3 Bibliographic Notes

Model checking. Model checking originates from the independent work of two pairs in the early eighties: Clarke and Emerson [86] and Queille and Sifakis [347]. The term model checking was coined by Clarke and Emerson. The brute-force examination of the entire state space in model checking can be considered as an extension of automated protocol validation techniques by Hajek [182] and West [419, 420]. While these earlier techniques were restricted to checking the absence of deadlocks or livelocks, model checking allows for the examination of broader classes of properties. Introductory papers on model checking can be found in $[94,95,96,293,426]$. The limitations of model checking were discussed by Apt and Kozen [17]. More information on model checking is available in the earlier books by Holzmann [205], McMillan [288], and Kurshan [250] and the more recent works by Clarke, Grumberg, and Peled [92], Huth and Ryan [219], Schneider [365], and Bérard et al. [44]. The model-checking trajectory has recently been described by Ruys and Brinksma [360].

Software verification. Empirical data about software engineering is gathered by the Center for Empirically Based Software Engineering (www.cebase.org); their collected data about software defects has recently been summarized by Boehm and Basili [53]. The different characterizations of verification ("are we building the thing right?") and validation ("are we building the right thing?") originate from Boehm [52]. An overview of software testing is given by Whittaker [421]; books about software testing are by Myers [308] and Beizer [36]. Testing based on formal specifications has been studied extensively in the area of communication protocols. This has led to an international standard for conformance
testing [222]. The use of software verification techniques by the German software industry has been studied by Liggesmeyer et al. [275]. Books by Storey [381] and Leveson [269] describe techniques for developing safety-critical software and discuss the role of formal verification in this context. Rushby [359] addresses the role of formal methods for developing safety-critical software. The book of Peled [327] gives a detailed account of formal techniques for software reliability that includes testing, model checking, and deductive methods.

Model-checking software. Model-checking communication protocols has become popular through the pioneering work of Holzmann [205, 206]. An interesting project at Bell Labs in which a model-checking team and a traditional design team worked on the design of part of the ISDN user part protocol has been reported by Holzmann [207]. In this large case study, 112 serious design flaws were discovered while checking 145 formal properties in about 10,000 verification runs. Errors found by Clarke et al. [89] in the IEEE Futurebus+ standard (checking a model of more than $10^{30}$ states) has led to a substantial revision of the protocol by IEEE. Chan et al. [79] used model checking to verify the control software of a traffic control and alert system for airplanes. Recently, Staunstrup et al. [377] have reported the succesful model checking of a train model consisting of 1421 state machines comprising a state space of $10^{476}$ states. Lowe [278], using model checking, discovered a flaw in the well-known Needham-Schroeder authentication algorithm that remained undetected for over 17 years. The usage of formal methods (that includes model checking) in the software development process of a safety-critical system within a Dutch software house is presented by Tretmans, Wijbrans, and Chaudron [393]. The formal analysis of NASA's Mars Pathfinder and the Deep Space-1 spacecraft are addressed by Havelund, Lowry, and Penix [194], and Holzmann, Najm, and Serhrouchini [210], respectively. The automated generation of abstract models amenable to model checking from programs written in programming languages such as C, C++, or Java has been pursued, for instance, by Godefroid [170], Dwyer, Hatcliff, and coworkers [193], at Microsoft Research by Ball, Podelski, and Rajamani [33] and at NASA Research by Havelund and Pressburger [195].

Model-checking hardware. Applying model checking to hardware originates from Browne et al. [66] analyzing some moderate-size self-timed sequential circuits. Successful applications of (symbolic) model checking to large hardware systems have been first reported by Burch et al. [75] in the early nineties. They analyzed a synchronous pipeline circuit of approximately $10^{20}$ states. Overviews of formal hardware verification techniques can be found in works by Gupta [179], and the books by Yoeli [428] and Kropf [246]. The need for formal verification techniques for hardware verification has been advocated by, among others, Sangiovanni-Vincentelli, McGeer, and Saldanha [362]. The integration of model-checking techniques for error finding in the hardware development process at IBM has been recently described by Schlipf et al. [364] and Abarbanel-Vinov et al. [2]. They conclude that model checking is a powerful extension of the traditional verification pro-
cess, and consider it as complementary to simulation/emulation. The design of a memory bus adapter at IBM showed, e.g., that $24 \%$ of all defects were found with model checking, while $40 \%$ of these errors would most likely not have been found by simulation.

## Chapter 2

## Modelling Concurrent Systems

A prerequisite for model checking is a model of the system under consideration. This chapter introduces transition systems, a (by now) standard class of models to represent hardware and software systems. Different aspects for modeling concurrent systems are treated, ranging from the simple case in which processes run completely autonomously to the more realistic setting where processes communicate in some way. The chapter is concluded by considering the problem of state-space explosion.

### 2.1 Transition Systems

Transition systems are often used in computer science as models to describe the behavior of systems. They are basically directed graphs where nodes represent states, and edges model transitions, i.e., state changes. A state describes some information about a system at a certain moment of its behavior. For instance, a state of a traffic light indicates the current color of the light. Similarly, a state of a sequential computer program indicates the current values of all program variables together with the current value of the program counter that indicates the next program statement to be executed. In a synchronous hardware circuit, a state typically represents the current value of the registers together with the values of the input bits. Transitions specify how the system can evolve from one state to another. In the case of the traffic light a transition may indicate a switch from one color to another, whereas for the sequential program a transition typically corresponds to the execution of a statement and may involve the change of some variables and the program counter. In the case of the synchronous hardware circuit, a transition models the change of the registers and output bits on a new set of inputs.

In the literature, many different types of transition systems have been proposed. We use transition systems with action names for the transitions (state changes) and atomic propositions for the states. Action names will be used for describing communication mechanisms between processes. We use letters at the beginning of the Greek alphabet (such as $\alpha, \beta$, and so on) to denote actions. Atomic propositions are used to formalize temporal characteristics. Atomic propositions intuitively express simple known facts about the states of the system under consideration. They are denoted by arabic letters from the beginning of the alphabet, such as $a, b, c$, and so on. Examples of atomic propositions are " $x$ equals 0 ", or " $x$ is smaller than 200 " for some given integer variable $x$. Other examples are "there is more than a liter of fluid in the tank" or "there are no customers in the shop".

## Definition 2.1. Transition System (TS)

A transition system $T S$ is a tuple $(S, A c t, \rightarrow, I, A P, L)$ where

- $S$ is a set of states,
- Act is a set of actions,
- $\longrightarrow \subseteq S \times$ Act $\times S$ is a transition relation,
- $I \subseteq S$ is a set of initial states,
- $A P$ is a set of atomic propositions, and
- $L: S \rightarrow 2^{A P}$ is a labeling function.
$T S$ is called finite if $S$, Act, and $A P$ are finite.

For convenience, we write $s \xrightarrow{\alpha} s^{\prime}$ instead of $\left(s, \alpha, s^{\prime}\right) \in \longrightarrow$. The intuitive behavior of a transition system can be described as follows. The transition system starts in some initial state $s_{0} \in I$ and evolves according to the transition relation $\longrightarrow$. That is, if $s$ is the current state, then a transition $s \xrightarrow{\alpha} s^{\prime}$ originating from $s$ is selected nondeterministically and taken, i.e., the action $\alpha$ is performed and the transition system evolves from state $s$ into the state $s^{\prime}$. This selection procedure is repeated in state $s^{\prime}$ and finishes once a state is encountered that has no outgoing transitions. (Note that $I$ may be empty; in that case, the transition system has no behavior at all as no initial state can be selected.) It is important to realize that in case a state has more than one outgoing transition, the "next" transition is chosen in a purely nondeterministic fashion. That is, the outcome of this selection process is not known a priori, and, hence, no statement can be made about
the likelihood with which a certain transition is selected. Similarly, when the set of initial states consists of more than one state, the start state is selected nondeterministically.

The labeling function $L$ relates a set $L(s) \in 2^{A P}$ of atomic propositions to any state $s .{ }^{1}$ $L(s)$ intuitively stands for exactly those atomic propositions $a \in A P$ which are satisfied by state $s$. Given that $\Phi$ is a propositional logic formula, then $s$ satisfies the formula $\Phi$ if the evaluation induced by $L(s)$ makes the formula $\Phi$ true; that is:

$$
s \models \Phi \quad \text { iff } \quad L(s) \models \Phi .
$$

(Basic principles of propositional logic are explained in Appendix A.3, see page 915 ff .)

## Example 2.2. Beverage Vending Machine

We consider an (somewhat foolish) example, which has been established as standard in the field of process calculi. The transition system in Figure 2.1 models a preliminary design of a beverage vending machine. The machine can either deliver beer or soda. States are represented by ovals and transitions by labeled edges. State names are depicted inside the ovals. Initial states are indicated by having an incoming arrow without source.


Figure 2.1: A transition system of a simple beverage vending machine.

The state space is $S=\{$ pay, select, soda, beer $\}$. The set of initial states consists of only one state, i.e., $I=\{$ pay $\}$. The (user) action insert_coin denotes the insertion of a coin, while the (machine) actions get_soda and get_beer denote the delivery of soda and beer, respectively. Transitions of which the action label is not of further interest here, e.g., as it denotes some internal activity of the beverage machine, are all denoted by the distinguished action symbol $\tau$. We have:

$$
\text { Act }=\{\text { insert_coin, get_soda, get_beer }, \tau\} .
$$

Some example transitions are:

$$
\text { pay } \xrightarrow{\text { insert_coin }} \text { select and beer } \xrightarrow{\text { get_beer }} \text { pay. }
$$

[^2]It is worthwhile to note that after the insertion of a coin, the vending machine nondeterministically can choose to provide either beer or soda.

The atomic propositions in the transition system depend on the properties under consideration. A simple choice is to let the state names act as atomic propositions, i.e., $L(s)=\{s\}$ for any state $s$. If, however, the only relevant properties do not refer to the selected beverage, as in the property
"The vending machine only delivers a drink after providing a coin",
it suffices to use the two-element set of propositions $A P=\{$ paid, drink $\}$ with labeling function:

$$
L(\text { pay })=\varnothing, \quad L(\text { soda })=L(\text { beer })=\{\text { paid, drink }\}, L(\text { select })=\{\text { paid }\} .
$$

Here, the proposition paid characterizes exactly those states in which the user has already paid but not yet obtained a beverage.

The previous example illustrates a certain arbitrariness concerning the choice of atomic propositions and action names. Even if the formal definition of a transition system requires determining the set of actions Act and the set of propositions AP, the components Act and $A P$ are casually dealt with in the following. Actions are only necessary for modeling communication mechanisms as we will see later on. In cases where action names are irrelevant, e.g., because the transition stands for an internal process activity, we use a special symbol $\tau$ or, in cases where action names are not relevant, even omit the action label. The set of propositions $A P$ is always chosen depending on the characteristics of interest. In depicting transition systems, the set of propositions $A P$ often is not explicitly indicated and it is assumed that $A P \subseteq S$ with labeling function $L(s)=\{s\} \cap A P$.

Crucial for modeling hard- or software systems by transition systems is the nondeterminism, which in this context is by far more than a theoretical concept. Later in this chapter (Section 2.2), we will explain in detail how transition systems can serve as a formal model for parallel systems. We mention here only that nondeterministic choices serve to model the parallel execution of independent activities by interleaving and to model the conflict situations that arise, e.g., if two processes aim to access a shared resource. Essentially, interleaving means the nondeterministic choice of the order in which order the actions of the processes that run in parallel are executed. Besides parallelism, the nondeterminism is also important for abstraction purposes, for underspecification, and to model the interface with an unknown or unpredictable environment (e.g., a human user). An example of the last is provided by the beverage vending machine where the user resolves the nondeterministic choice between the two $\tau$-transitions in state "select" by choosing one of the two available drinks. The notion "underspecification" refers to early design phases where
a coarse model for a system is provided that represents several options for the possible behaviors by nondeterminism. The rough idea is that in further refinement steps the designer realizes one of the nondeterministic alternatives, but skips the others. In this sense, nondeterminism in a transition system can represent implementation freedom.

## Definition 2.3. Direct Predecessors and Successors

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system. For $s \in S$ and $\alpha \in$ Act, the set of direct $\alpha$-successors of $s$ is defined as:

$$
\operatorname{Post}(s, \alpha)=\left\{s^{\prime} \in S \mid s \xrightarrow{\alpha} s^{\prime}\right\}, \quad \operatorname{Post}(s)=\bigcup_{\alpha \in A c t} \operatorname{Post}(s, \alpha) .
$$

The set of $\alpha$-predecessors of $s$ is defined by:

$$
\operatorname{Pre}(s, \alpha)=\left\{s^{\prime} \in S \mid s^{\prime} \xrightarrow{\alpha} s\right\}, \quad \operatorname{Pre}(s)=\bigcup_{\alpha \in A c t} \operatorname{Pre}(s, \alpha) .
$$

Each state $s^{\prime} \in \operatorname{Post}(s, \alpha)$ is called a direct $\alpha$-successor of $s$. Accordingly, each state $s^{\prime} \in \operatorname{Post}(s)$ is called a direct successor of $s$. The notations for the sets of direct successors are expanded to subsets of $S$ in the obvious way (i.e., pointwise extension): for $C \subseteq S$, let

$$
\operatorname{Post}(C, \alpha)=\bigcup_{s \in C} \operatorname{Post}(s, \alpha), \quad \operatorname{Post}(C)=\bigcup_{s \in C} \operatorname{Post}(s)
$$

The notations $\operatorname{Pre}(C, \alpha)$ and $\operatorname{Pre}(C)$ are defined in an analogous way:

$$
\operatorname{Pre}(C, \alpha)=\bigcup_{s \in C} \operatorname{Pre}(s, \alpha), \quad \operatorname{Pre}(C)=\bigcup_{s \in C} \operatorname{Pre}(s) .
$$

Terminal states of a transition system $T S$ are states without any outgoing transitions. Once the system described by TS reaches a terminal state, the complete system comes to a halt.

## Definition 2.4. Terminal State

State $s$ in transition system $T S$ is called terminal if and only if $\operatorname{Post}(s)=\varnothing$.

For a transition system modeling a sequential computer program, terminal states occur as a natural phenomenon representing the termination of the program. Later on, we will
see that for transition systems modeling parallel systems, such terminal states are usually considered to be undesired (see Section 3.1, page 89 ff .).

We mentioned above that nondeterminism is crucial for modeling computer systems. However, it is often useful to consider transition systems where the "observable" behavior is deterministic, according to some notion of observables. There are two general approaches to formalize the visible behavior of a transition system: one relies on the actions, the other on the labels of the states. While the action-based approach assumes that only the executed actions are observable from outside, the state-based approach ignores the actions and relies on the atomic propositions that hold in the current state to be visible. Transition systems that are deterministic in the action-based view have at most one outgoing transition labeled with action $\alpha$ per state, while determinism from the view of state labels means that for any state label $A \in 2^{A P}$ and any state there is at most one outgoing transition leading to a state with label $A$. In both cases, it is required that there be at most one initial state.

## Definition 2.5. Deterministic Transition System

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system.

1. TS is called action-deterministic if $|I| \leqslant 1$ and $|\operatorname{Post}(s, \alpha)| \leqslant 1$ for all states $s$ and actions $\alpha$.
2. $T S$ is called $A P$-deterministic if $|I| \leqslant 1$ and $\left|\operatorname{Post}(s) \cap\left\{s^{\prime} \in S \mid L\left(s^{\prime}\right)=A\right\}\right| \leqslant 1$ for all states $s$ and $A \in 2^{A P}$.

### 2.1.1 Executions

So far, the behavior of a transition system has been described at an intuitive level. This will now be formalized using the notion of executions (also called runs). An execution of a transition system results from the resolution of the possible nondeterminism in the system. An execution thus describes a possible behavior of the transition system. Formally:

## Definition 2.6. Execution Fragment

Let $T S=(S$, Act $, \rightarrow, I, A P, L)$ be a transition system. A finite execution fragment $\varrho$ of $T S$ is an alternating sequence of states and actions ending with a state

$$
\varrho=s_{0} \alpha_{1} s_{1} \alpha_{2} \ldots \alpha_{n} s_{n} \text { such that } s_{i} \xrightarrow{\alpha_{i+1}} s_{i+1} \text { for all } 0 \leqslant i<n,
$$

where $n \geqslant 0$. We refer to $n$ as the length of the execution fragment $\varrho$. An infinite execution fragment $\rho$ of $T S$ is an infinite, alternating sequence of states and actions:

$$
\rho=s_{0} \alpha_{1} s_{1} \alpha_{2} s_{2} \alpha_{3} \ldots \text { such that } s_{i} \xrightarrow{\alpha_{i+1}} s_{i+1} \text { for all } 0 \leqslant i .
$$

Note that the sequence $s$ with $s \in S$ is a legal finite execution fragment of length $n=0$. Each prefix of odd length of an infinite execution fragment is a finite execution fragment. From now on, the term execution fragment will be used to denote either a finite or an infinite execution fragment. Execution fragments $\varrho=s_{0} \alpha_{1} \ldots \alpha_{n} s_{n}$ and $\rho=s_{0} \alpha_{1} s_{1} \alpha_{2} \ldots$ will be written respectively as

$$
\varrho=s_{0} \xrightarrow{\alpha_{1}} \ldots \xrightarrow{\alpha_{n}} s_{n} \quad \text { and } \quad \rho=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} \ldots .
$$

An execution fragment is called maximal when it cannot be prolonged:

## Definition 2.7. Maximal and Initial Execution Fragment

A maximal execution fragment is either a finite execution fragment that ends in a terminal state, or an infinite execution fragment. An execution fragment is called initial if it starts in an initial state, i.e., if $s_{0} \in I$.

## Example 2.8. Executions of the Beverage Vending Machine

Some examples of execution fragments of the beverage vending machine described in Example 2.2 (page 21) are as follows. For simplicity, the action names are abbreviated, e.g., sget is a shorthand for get_soda and coin for insert_coin.

$$
\begin{aligned}
& \rho_{1}=\text { pay } \xrightarrow{\text { coin }} \text { select } \xrightarrow{\tau} \text { soda } \xrightarrow{\text { sget }} \text { pay } \xrightarrow{\text { coin }} \text { select } \xrightarrow{\tau} \text { soda } \xrightarrow{\text { sget }} \ldots \\
& \rho_{2}=\text { select } \xrightarrow{\tau} \text { soda } \xrightarrow{\text { sget }} \text { pay } \xrightarrow{\text { coin }} \text { select } \xrightarrow{\tau} \text { beer } \xrightarrow{\text { bget }} \ldots \\
& \varrho=\text { pay } \xrightarrow{\text { coin }} \text { select } \xrightarrow{\tau} \text { soda } \xrightarrow{\text { sget }} \text { pay } \xrightarrow{\text { coin }} \text { select } \xrightarrow{\tau} \text { soda } .
\end{aligned}
$$

Execution fragments $\rho_{1}$ and $\varrho$ are initial, but $\rho_{2}$ is not. $\varrho$ is not maximal as it does not end in a terminal state. Assuming that $\rho_{1}$ and $\rho_{2}$ are infinite, they are maximal.

## Definition 2.9. Execution

An execution of transition system $T S$ is an initial, maximal execution fragment.

In Example 2.8, $\rho_{1}$ is an execution, while $\rho_{2}$ and $\varrho$ are not. Note that $\rho_{2}$ is maximal but not initial, while $\varrho$ is initial but not maximal.

A state $s$ is called reachable if there is some execution fragment that ends in $s$ and that starts in some initial state.

## Definition 2.10. Reachable States

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system. A state $s \in S$ is called reachable in $T S$ if there exists an initial, finite execution fragment

$$
s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} \ldots \xrightarrow{\alpha_{n}} s_{n}=s
$$

Reach $(T S)$ denotes the set of all reachable states in $T S$.

### 2.1.2 Modeling Hardware and Software Systems

This section illustrates the use of transition systems by elaborating on the modeling of (synchronous) hardware circuits and sequential data-dependent systems - a kind of simple sequential computer programs. For both cases, the basic concept is that states represent possible storage configurations (i.e., evaluations of relevant "variables"), and state changes (i.e., transitions) represent changes of "variables". Here, the term "variable" has to be understood in the broadest sense. For computer programs a variable can be a control variable (like a program counter) or a program variable. For circuits a variable can, e.g, stand for either a register or an input bit.

## Sequential Hardware Circuits

Before presenting a general recipe for modeling sequential hardware circuits as transition systems we consider a simple example to clarify the basic concepts.

## Example 2.11. A Simple Sequential Hardware Circuit

Consider the circuit diagram of the sequential circuit with input variable $x$, output variable $y$, and register $r$ (see left part of Figure 2.2). The control function for output variable $y$ is given by

$$
\lambda_{y}=\neg(x \oplus r)
$$



Figure 2.2: Transition system representation of a simple hardware circuit.
where $\oplus$ stands for exclusive or (XOR, or parity function). The register evaluation changes according to the circuit function

$$
\delta_{r}=x \vee r .
$$

Note that once the register evaluation is $[r=1], r$ keeps that value. Under the initial register evaluation $[r=0]$, the circuit behavior is modeled by the transition system $T S$ with state space

$$
S=\operatorname{Eval}(x, r)
$$

where $\operatorname{Eval}(x, r)$ stands for the set of evaluations of input variable $x$ and register variable $r$. The initial states of $T S$ are $I=\{\langle x=0, r=0\rangle,\langle x=1, r=0\rangle\}$. Note that there are two initial states as we do not make any assumption about the initial value of the input bit $x$.

The set of actions is irrelevant and omitted here. The transitions result directly from the functions $\lambda_{y}$ and $\delta_{r}$. For instance, $\langle x=0, r=1\rangle \rightarrow\langle x=0, r=1\rangle$ if the next input bit equals 0 , and $\langle x=0, r=1\rangle \rightarrow\langle x=1, r=1\rangle$ if the next input bit is 1 .

It remains to consider the labeling $L$. Using the set of atomic propositions $A P=\{x, y, r\}$, then, e.g., the state $\langle x=0, r=1\rangle$ is labeled with $\{r\}$. It is not labeled with $y$ since the circuit function $\neg(x \oplus r)$ results in the value 0 for this state. For state $\langle x=1, r=1\rangle$ we obtain $L(\langle x=1, r=1\rangle)=\{x, r, y\}$, as $\lambda_{y}$ yields the value 1. Accordingly, we obtain: $L(\langle x=0, r=0\rangle)=\{y\}$, and $L(\langle x=1, r=0\rangle)=\{x\}$. The resulting transition system (with this labeling) is depicted in the right part of Figure 2.2.

Alternatively, using the set of propositions $A P^{\prime}=\{x, y\}$ - the register evaluations are assumed to be "invisible" - one obtains:

$$
\begin{aligned}
L^{\prime}(\langle x=0, r=0\rangle) & =\{y\} \\
L^{\prime}(\langle x=1, r=0\rangle) & =\{x\}
\end{aligned} \quad L^{\prime}(\langle x=0, r=1\rangle)=\varnothing \quad L^{\prime}(\langle x=1, r=1\rangle)=\{x, y\}
$$

The propositions in $A P^{\prime}$ suffice to formalize, e.g., the property "the output bit $y$ is set infinitely often". Properties that refer to the register $r$ are not expressible.

The approach taken in this example can be generalized toward arbitrary sequential hardware circuits (without "don't cares") with $n$ input bits $x_{1}, \ldots, x_{n}, m$ output bits $y_{1}, \ldots, y_{m}$, and $k$ registers $r_{1}, \ldots, r_{k}$ as follows. The states of the transition system represent the evaluations of the $n+k$ input and register bits $x_{1}, \ldots, x_{n}, r_{1}, \ldots, r_{k}$. The evaluation of output bits depends on the evaluations of input bits and registers and can be derived from the states. Transitions represent the behavior, whereas it is assumed that the values of input bits are nondeterministically provided (by the circuit environment). Furthermore, we assume a given initial register evaluation

$$
\left[r_{1}=c_{0,1}, \ldots, r_{k}=c_{0, k}\right]
$$

where $c_{0, i}$ denotes the initial value of register $i$ for $0<i \leqslant k$. Alternatively, a set of possible initial register evaluations may be given.

The transition system $T S=(S, A c t, \rightarrow, I, A P, L)$ modeling this sequential hardware circuit has the following components. The state space $S$ is determined by

$$
S=\operatorname{Eval}\left(x_{1}, \ldots, x_{n}, r_{1}, \ldots, r_{k}\right)
$$

Here, $\operatorname{Eval}\left(x_{1}, \ldots, x_{n}, r_{1}, \ldots, r_{k}\right)$ stands for the set of evaluations of input variables $x_{i}$ and registers $r_{j}$ and can be identified with the set $\{0,1\}^{n+k}$. ${ }^{2}$ Initial states are of the form $\left(\ldots, c_{0,1}, \ldots, c_{0, k}\right)$ where the $k$ registers are evaluated with their initial value. The first $n$ components prescribing the values of input bits are arbitrary. Thus, the set of initial states is

$$
I=\left\{\left(a_{1}, \ldots, a_{n}, c_{0,1}, \ldots, c_{0, k}\right) \mid a_{1}, \ldots, a_{n} \in\{0,1\}\right\}
$$

The set Act of actions is irrelevant, and we choose $A c t=\{\tau\}$. For simplicity, let the set of atomic propositions be

$$
A P=\left\{x_{1}, \ldots, x_{n}, y_{1}, \ldots, y_{m}, r_{1}, \ldots, r_{k}\right\}
$$

(In practice, this could be defined as any subset of this $A P$ ). Thus, any register, any input bit, and any output bit can be used as an atomic proposition. The labeling function assigns to any state $s \in \operatorname{Eval}\left(x_{1}, \ldots, x_{n}, r_{1}, \ldots, r_{k}\right)$ exactly those atomic propositions $x_{i}$, $r_{j}$ which are evaluated to 1 under $s$. If for state $s$, output bit $y_{i}$ is evaluated to 1 , then (and only then) the atomic proposition $y_{i}$ is part of $L(s)$. Thus,

$$
\begin{aligned}
L\left(a_{1}, \ldots, a_{n}, c_{1}, \ldots, c_{k}\right)= & \left\{x_{i} \mid a_{i}=1\right\} \cup\left\{r_{j} \mid c_{j}=1\right\} \\
& \cup\left\{y_{i} \mid s \models \lambda_{y_{i}}\left(a_{1}, \ldots, a_{n}, c_{1}, \ldots, c_{k}\right)=1\right\}
\end{aligned}
$$

[^3]where $\lambda_{y_{i}}: S \rightarrow\{0,1\}$ is the switching function corresponding to output bit $y_{i}$ that results from the gates of the circuit.

Transitions exactly represent the behavior. In the following, let $\delta_{r_{j}}$ denote the transition function for register $r_{j}$ resulting from the circuit diagram. Then:

$$
(\underbrace{a_{1}, \ldots, a_{n}}_{\text {input evaluation }}, \underbrace{c_{1}, \ldots, c_{k}}_{\text {register evaluation }}) \xrightarrow{\tau}\left(a_{1}^{\prime}, \ldots, a_{n}^{\prime}, c_{1}^{\prime}, \ldots, c_{k}^{\prime}\right)
$$

if and only if $c_{j}^{\prime}=\delta_{r_{j}}\left(a_{1}, \ldots, a_{n}, c_{1}, \ldots, c_{k}\right)$. Assuming that the evaluation of input bits changes nondeterministically, no restrictions on the bits $a_{1}^{\prime}, \ldots, a_{n}^{\prime}$ are imposed.

It is left to the reader to check that applying this recipe to the example circuit in the left part of Figure 2.2 indeed results in the transition system depicted in the right part of that figure.

## Data-Dependent Systems

The executable actions of a data-dependent system typically result from conditional branching, as in

$$
\text { if } x \% 2=1 \text { then } x:=x+1 \text { else } x:=2 \cdot x \text { fi. }
$$

In principle, when modeling this program fragment as a transition system, the conditions of transitions could be omitted and conditional branchings could be replaced by nondeterminism; but, generally speaking, this results in a very abstract transition system for which only a few relevant properties can be verified. Alternatively, conditional transitions can be used and the resulting graph (labeled with conditions) can be unfolded into a transition system that subsequently can be subject to verification. This unfolding approach is detailed out below. We first illustrate this by means of an example.

## Example 2.12. Beverage Vending Machine Revisited

Consider an extension of the beverage vending machine described earlier in Example 2.2 (page 21) which counts the number of soda and beer bottles and returns inserted coins if the vending machine is empty. For the sake of simplicity, the vending machine is represented by the two locations start and select. The following conditional transitions

$$
\text { start } \xrightarrow{\text { true }: \text { coin }} \text { select } \text { and } \text { start } \simeq \text { true : refill }_{\sim} \text { start }
$$

model the insertion of a coin and refilling the vending machine. Labels of conditional transitions are of the form $g: \alpha$ where $g$ is a Boolean condition (called guard), and $\alpha$ is an action that is possible once $g$ holds. As the condition for both conditional transitions
above always holds, the action coin is always enabled in the starting location. To keep things simple, we assume that by refill both storages are entirely refilled. Conditional transitions

$$
\text { select } \overbrace{}^{\text {nsoda }>0: \text { sget }} \text { start } \text { and } \text { select } \overbrace{}^{\text {nbeer }>0: \text { bget }} \text { start }
$$

model that soda (or beer) can be obtained if there is some soda (or beer) left in the vending machine. The variables nsoda and nbeer record the number of soda and beer bottles in the machine, respectively. Finally, the vending machine automatically switches to the initial start location while returning the inserted coin once there are no bottles left:

$$
\text { select } \xlongequal{\text { nsoda } a=0 \wedge \text { nbeer }=0: \text { ret_coin }} \text { start }
$$

Let the maximum capacity of both bottle repositories be max. The insertion of a coin (by action coin) leaves the number of bottles unchanged. The same applies when a coin is returned (by action ret_coin). The effect of the other actions is as follows:

| Action | Effect |
| :--- | :--- |
| refill | nsoda $:=$ max $;$ nbeer $:=$ max |
| sget | $n$ soda $:=$ nsoda -1 |
| bget | nbeer $:=$ nbeer -1 |

The graph consisting of locations as nodes and conditional transitions as edges is not a transition system, since the edges are provided with conditions. A transition system, however, can be obtained by "unfolding" this graph. For instance, Figure 2.3 on page 31 depicts this unfolded transition system when max equals 2 . The states of the transition system keep track of the current location in the graph described above and of the number of soda- and beer bottles in the vending machine (as indicated by the gray and black dots, respectively, inside the nodes of the graph).

The ideas outlined in the previous example are formalized by using so-called program graphs over a set Var of typed variables such as nsoda and nbeer in the example. Essentially, this means that a standardized type (e.g., boolean, integer, or char) is associated with each variable. The type of variable $x$ is called the domain $\operatorname{dom}(x)$ of $x$. Let $\operatorname{Eval}(\operatorname{Var})$ denote the set of (variable) evaluations that assign values to variables. Cond(Var) is the set of Boolean conditions over Var, i.e., propositional logic formulae whose propositional symbols are of the form " $\bar{x} \in \bar{D}$ " where $\bar{x}=\left(x_{1}, \ldots, x_{n}\right)$ is a tuple consisting of pairwise distinct variables in Var and $\bar{D}$ is a subset of $\operatorname{dom}\left(x_{1}\right) \times \ldots \times \operatorname{dom}\left(x_{n}\right)$. The proposition

$$
\left(-3<x-x^{\prime} \leqslant 5\right) \wedge\left(x \leqslant 2 \cdot x^{\prime}\right) \wedge(y=\text { green })
$$

for instance, is a legal Boolean condition for integer variables $x$ and $x^{\prime}$, and $y$ a variable with, e.g., $\operatorname{dom}(y)=\{$ red, green $\}$. Here and in the sequel, we often use simplified notations


Figure 2.3: Transition system modeling the extended beverage vending machine.
for the propositional symbols such as " $3<x-x^{\prime} \leqslant 5$ " instead of " $\left(x, x^{\prime}\right) \in\{(n, m) \in$ $\left.\mathbb{N}^{2} \mid 3<n-m \leqslant 5\right\} "$.

Initially, we do not restrict the domains. $\operatorname{dom}(x)$ can be an arbitrary, possibly infinite, set. Even if in real computer systems all domains are finite (e.g., the type integer only includes integers $n$ of a finite domain, like $-2^{16}<n<2^{16}$ ), then the logical or algorithmic structure of a program is often based on infinite domains. The decision which restrictions on domains are useful for implementation, e.g., how many bits should be provided for representation of variables of type integer is delayed until a later design stage and is ignored here.

A program graph over a set of typed variables is a digraph whose edges are labeled with conditions on these variables and actions. The effect of the actions is formalized by means of a mapping

$$
\text { Effect }: \operatorname{Act} \times \operatorname{Eval}(\operatorname{Var}) \rightarrow \operatorname{Eval}(\operatorname{Var})
$$

which indicates how the evaluation $\eta$ of variables is changed by performing an action. If, e.g., $\alpha$ denotes the action $x:=y+5$, where $x$ and $y$ are integer variables, and $\eta$ is the evaluation with $\eta(x)=17$ and $\eta(y)=-2$, then

$$
\operatorname{Effect}(\alpha, \eta)(x)=\eta(y)+5=-2+5=3, \quad \text { and } \operatorname{Effect}(\alpha, \eta)(y)=\eta(y)=-2
$$

$\operatorname{Effect}(\alpha, \eta)$ is thus the evaluation that assigns 3 to $x$ and -2 to $y$. The nodes of a program graph are called locations and have a control function since they specify which of the conditional transitions are possible.

## Definition 2.13. Program Graph (PG)

A program graph PG over set Var of typed variables is a tuple (Loc, Act, Effect, $\left.\hookrightarrow, L o c_{0}, g_{0}\right)$ where

- Loc is a set of locations and Act is a set of actions,
- Effect : Act $\times \operatorname{Eval}(\operatorname{Var}) \rightarrow \operatorname{Eval}(\operatorname{Var})$ is the effect function,
- $\hookrightarrow \subseteq \operatorname{Loc} \times \operatorname{Cond}(\operatorname{Var}) \times$ Act $\times$ Loc is the conditional transition relation,
- $L o c_{0} \subseteq L o c$ is a set of initial locations,
- $g_{0} \in \operatorname{Cond}(\operatorname{Var})$ is the initial condition.

The notation $\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}$ is used as shorthand for $\left(\ell, g, \alpha, \ell^{\prime}\right) \in \hookrightarrow$. The condition $g$ is also called the guard of the conditional transition $\ell \stackrel{\text { g: } \alpha}{\longrightarrow} \ell^{\prime}$. If the guard is a tautology (e.g., $g=$ true or $g=(x<1) \vee(x \geqslant 1))$, then we simply write $\ell \stackrel{\alpha}{\hookrightarrow} \ell^{\prime}$.

The behavior in location $\ell \in L o c$ depends on the current variable evaluation $\eta$. A nondeterministic choice is made between all transitions $\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}$ which satisfy condition $g$ in evaluation $\eta$ (i.e., $\eta \models g$ ). The execution of action $\alpha$ changes the evaluation of variables according to $\operatorname{Effect}(\alpha, \cdot)$. Subsequently, the system changes into location $\ell^{\prime}$. If no such transition is possible, the system stops.

## Example 2.14. Beverage Vending Machine

The graph described in Example 2.12 (page 29) is a program graph. The set of variables is

$$
\text { Var }=\{\text { nsoda }, \text { nbeer }\}
$$

where both variables have the domain $\{0,1, \ldots, \max \}$. The set $L o c$ of locations equals $\{$ start, select $\}$ with $\operatorname{Loc}_{0}=\{$ start $\}$, and

$$
\text { Act }=\{\text { bget, sget, coin, ret_coin, refill }\} .
$$

The effect of the actions is defined by:

$$
\begin{array}{ll}
\text { Effect }(\text { coin, } \eta) & =\eta \\
\text { Effect }(\text { ret_coin, } \eta) & =\eta \\
\text { Effect }(\text { sget, } \eta) & =\eta[\text { nsoda }:=\text { nsoda }-1] \\
\text { Effect }(\text { bget }, \eta) & =\eta[\text { nbeer }:=\text { nbeer }-1] \\
\text { Effect }(\text { refill }, \eta) & =[\text { nsoda }:=\text { max }, \text { nbeer }:=\text { max }]
\end{array}
$$

Here, $\eta[n s o d a:=n s o d a-1]$ is a shorthand for evaluation $\eta^{\prime}$ with $\eta^{\prime}(n s o d a)=\eta(n s o d a)-1$ and $\eta^{\prime}(x)=\eta(x)$ for all variables different from nsoda. The initial condition $g_{0}$ states that initially both storages are entirely filled, i.e., $g_{0}=(n s o d a=\max \wedge n b e e r=\max )$.

Each program graph can be interpreted as a transition system. The underlying transition system of a program graph results from unfolding. Its states consist of a control component, i.e., a location $\ell$ of the program graph, together with an evaluation $\eta$ of the variables. States are thus pairs of the form $\langle\ell, \eta\rangle$. Initial states are initial locations that satisfy the initial condition $g_{0}$. To formulate properties of the system described by a program graph, the set AP of propositions is comprised of locations $\ell \in L o c$ (to be able to state at which control location the system currently is), and Boolean conditions for the variables. For example, a proposition like

$$
(x \leqslant 5) \wedge(y \text { is even }) \wedge(\ell \in\{1,2\})
$$

can be formulated with $x, y$ being integer variables and with locations being naturals. The labeling of states is such that $\langle\ell, v\rangle$ is labeled with $\ell$ and with all conditions (over Var) that hold in $\eta$. The transition relation is determined as follows. Whenever $\ell \xrightarrow{g: \alpha} \ell^{\prime}$ is a conditional transition in the program graph, and the guard $g$ holds in the current evaluation $\eta$, then there is a transition from state $\langle\ell, \eta\rangle$ to state $\left\langle\ell^{\prime}, \operatorname{Effect}(\alpha, \eta)\right\rangle$. Note that the transition is not guarded. Formally:

## Definition 2.15. Transition System Semantics of a Program Graph

The transition system $T S(P G)$ of program graph

$$
P G=\left(\text { Loc }, \text { Act }, \text { Effect }, \hookrightarrow, \text { Loc }_{0}, g_{0}\right)
$$

over set Var of variables is the tuple $(S, A c t, \longrightarrow, I, A P, L)$ where

- $S=L o c \times \operatorname{Eval}(\operatorname{Var})$
- $\longrightarrow \subseteq S \times$ Act $\times S$ is defined by the following rule (see remark below):

$$
\frac{\ell \stackrel{g: \alpha}{\hookrightarrow} \ell^{\prime} \wedge \quad \eta \models g}{\langle\ell, \eta\rangle \stackrel{\alpha}{\longrightarrow}\left\langle\ell^{\prime}, \operatorname{Effect}(\alpha, \eta)\right\rangle}
$$

- $I=\left\{\langle\ell, \eta\rangle \mid \ell \in L_{o c}, \eta=g_{0}\right\}$
- $A P=L o c \cup \operatorname{Cond}($ Var $)$
- $L(\langle\ell, \eta\rangle)=\{\ell\} \cup\{g \in \operatorname{Cond}(\operatorname{Var}) \mid \eta \models g\}$.

The definition of $T S(P G)$ determines a very large set of propositions AP. But generally, only a small part of $A P$ is necessary to formulate the relevant system properties. In the following, we exploit the degrees of freedom in choosing the set of propositions of $T S(P G)$ and only use the atomic propositions needed in the context at hand.

## Remark 2.16. Structured Operational Semantics

In Definition 2.15, the transition relation is defined using the so-called SOS-notation (Structured Operational Semantics). This notation will be frequently used in the remainder of this monograph. The notation

$$
\frac{\text { premise }}{\text { conclusion }}
$$

should be read as follows. If the proposition above the "solid line" (i.e., the premise) holds, then the proposition under the fraction bar (i.e., the conclusion) holds as well. Such "if ..., then ..." propositions are also called inference rules or simply rules. If the premise is a tautology, it may be omitted (as well as the "solid line"). In the latter case, the rule is also called an axiom.

Phrases like "The relation $\rightarrow$ is defined by the following (axioms and) rules" have the meaning of an inductive definition where the relation $\rightarrow$ is defined as the smallest relation satisfying the indicated axioms and rules.

### 2.2 Parallelism and Communication

In the previous section, we have introduced the notion of transition systems and have shown how sequential hardware circuits and data-dependent systems (like simple sequential computer programs) can be effectively modeled as transition systems. In reality, however, most hard- and software systems are not sequential but parallel in nature. This section describes several mechanisms to provide operational models for parallel systems by means of transition systems. These mechanisms range from simple mechanisms where no communication between the participating transitions systems takes place, to more advanced (and realistic) schemes in which messages can be transferred, either synchronously (i.e., by means of "handshaking") or asynchronously (i.e., by buffers with a positive capacity). Let us assume that the operational (stepwise) behavior of the processes that run in parallel are given by transition systems $T S_{1}, \ldots, T S_{n}$. The goal is to define an operator $\|$, such that:

$$
T S=T S_{1}\left\|T S_{2}\right\| \ldots \| T S_{n}
$$

is a transition system that specifies the behavior of the parallel composition of transition systems $T S_{1}$ through $T S_{n}$. Here, it is assumed that $\|$ is a commutative and associative operator. The nature of the operator $\|$ will, of course, depend on the kind of communication that is supported. We will for instance see that some notions of parallel composition do not yield an associative operator. In the remainder of this section, several variants of $\|$ will be considered and illustrated by means of examples. Note that the above scheme may be repeated for $T S_{i}$, i.e., $T S_{i}$ may again be a transition system that is composed of several transition systems:

$$
T S_{i}=T S_{i, 1}\left\|T S_{i, 1}\right\| \ldots \| T S_{i, n_{i}}
$$

By using parallel composition in this hierarchical way, complex systems can be described in a rather structured way.

### 2.2.1 Concurrency and Interleaving

A widely adopted paradigm for parallel systems is that of interleaving. In this model, one abstracts from the fact that a system is actually composed of a set of (partly) independent components. That is to say, the global system state - composed of the current individual states of the components - plays a key role in interleaving. Actions of an independent component are merged (also called weaved), or "interleaved", with actions from other components. Thus, concurrency is represented by (pure) interleaving, that is, the nondeterministic choice between activities of the simultaneously acting processes (or components). This perspective is based on the view that only one processor is available on which the actions of the processes are interlocked. The "one-processor view" is only a modeling concept and also applies if the processes run on different processors. Thereby, (at first) no assumptions are made about the order in which the different processes are executed. If there are, e.g., two nonterminating processes $P$ and $Q$, say, acting completely independent of each other, then

$$
\begin{array}{llllllllll}
P & Q & P & Q & P & Q & Q & Q & P & \ldots \\
P & P & Q & P & P & Q & P & P & Q & \ldots \\
P & Q & P & P & Q & P & P & P & Q & \ldots
\end{array}
$$

are three possible sequences in which the steps (i.e., execution of actions) of $P$ and $Q$ can be interlocked. (In Chapter 3, certain restrictions will be discussed to ensure that each participating processor is treated in a somewhat "fair" manner. In particular, execution sequences like $P, P, P, \ldots$, where $Q$ is completely ignored, are ruled out. Unless stated otherwise, we accept all possible interleavings, including the unfair ones.)

The interleaving representation of concurrency is subject to the idea that there is a scheduler which interlocks the steps of concurrently executing processes according to an a priori unknown strategy. This type of representation completely abstracts from the speed of the participating processes and thus models any possible realization by a single-processor machine or by several processors with arbitrary speeds.

## Example 2.17. Two Independent Traffic Lights

Consider the transition systems of two traffic lights for nonintersecting (i.e., parallel) roads. It is assumed that the traffic lights switch completely independent of each other. For example, the traffic lights may be controlled by pedestrians who would like to cross the road. Each traffic light is modeled as a simple transition system with two states, one state


Figure 2.4: Example of interleaving operator for transition systems.
modeling a red light, the other one modeling a green light (see upper part of Figure 2.4). The transition system of the parallel composition of both traffic lights is sketched at the bottom of Figure 2.4 where ||| denotes the interleaving operator. In principle, any form of interlocking of the "actions" of the two traffic lights is possible. For instance, in the initial state where both traffic lights are red, there is a non-deterministic choice between which of the lights turns green. Note that this nondeterminism is descriptive, and does not model a scheduling problem between the traffic lights (although it may seem so).

An important justification for interleaving is the fact that the effect of concurrently executed, independent actions $\alpha$ and $\beta$, say, is identical to the effect when $\alpha$ and $\beta$ are successively executed in arbitrary order. This can symbolically be stated as

$$
\operatorname{Effect}(\alpha\|\| \beta, \eta)=\operatorname{Effect}((\alpha ; \beta)+(\beta ; \alpha), \eta)
$$

where the operator semicolon ; stands for sequential execution, + stands for nondeterministic choice, and ||| for the concurrent execution of independent activities. This fact can be easily understood when the effect is considered from two independent value assignments

$$
\underbrace{x:=x+1}_{=\alpha}\| \| \underbrace{y:=y-2}_{=\beta}
$$

When initially $x=0$ and $y=7$, then $x$ has the value 1 and $y$ the value 5 after executing $\alpha$ and $\beta$, independent of whether the assignments occur concurrently (i.e., simultaneously) or in some arbitrary successive order. This is depicted in terms of transition systems as follows:


Note that the independence of actions is crucial. For dependent actions, the order of actions is typically essential: e.g., the final value of variable $x$ in the parallel program $x:=x+1 \mid \| x:=2 \cdot x$ (with initial value $x=0$, say) depends on the order in which the assignments $x:=x+1$ and $x:=2 \cdot x$ take place.

We are now in a position to formally define the interleaving (denoted ||| ) of transition systems. The transition system $T S_{1} \| \mid T S_{2}$ represents a parallel system resulting from the weaving (or merging) of the actions of the components as described by $T S_{1}$ and $T S_{2}$. It is assumed that no communication and no contentions (on shared variables) occur at all. The ("global") states of $T S_{1}\| \| S_{2}$ are pairs $\left\langle s_{1}, s_{2}\right\rangle$ consisting of "local" states $s_{i}$ of the components $T S_{i}$. The outgoing transitions of the global state $\left\langle s_{1}, s_{2}\right\rangle$ consist of the outgoing transitions of $s_{1}$ together with those of $s_{2}$. Accordingly, whenever the composed system is in state $\left\langle s_{1}, s_{2}\right\rangle$, a nondeterministic choice is made between all outgoing transitions of local state $s_{1}$ and those of local state $s_{2}$.

## Definition 2.18. Interleaving of Transition Systems

Let $T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P_{i}, L_{i}\right) i=1,2$, be two transition systems. The transition system $T S_{1}| | T S_{2}$ is defined by:

$$
T S_{1} \| T S_{2}=\left(S_{1} \times S_{2}, A c t_{1} \cup A c t_{2}, \rightarrow, I_{1} \times I_{2}, A P_{1} \cup A P_{2}, L\right)
$$

where the transition relation $\rightarrow$ is defined by the following rules:

$$
\frac{s_{1} \xrightarrow{\alpha} 1 s_{1}^{\prime}}{\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha}\left\langle s_{1}^{\prime}, s_{2}\right\rangle} \quad \text { and } \frac{s_{2} \xrightarrow{\alpha}_{2} s_{2}^{\prime}}{\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha}\left\langle s_{1}, s_{2}^{\prime}\right\rangle}
$$

and the labeling function is defined by $L\left(\left\langle s_{1}, s_{2}\right\rangle\right)=L\left(s_{1}\right) \cup L\left(s_{2}\right)$.

## Example 2.19.

Consider the two independent traffic lights described in Example 2.17 (page 36). The depicted transition system is actually the transition system

$$
T S=\text { TrLight }_{1}| | \mid T r \text { Light }_{2}
$$

that originates from interleaving.

For program graphs $P G_{1}$ (on $\operatorname{Var}_{1}$ ) and $P G_{2}$ (on $\operatorname{Var}_{2}$ ) without shared variables (i.e., $\operatorname{Var}_{1} \cap \operatorname{Var}_{2}=\varnothing$ ), the interleaving operator, which is applied to the appropriate transition systems, yields a transition system

$$
T S\left(P G_{1}\right) \| T S\left(P G_{2}\right)
$$

that describes the behavior of the simultaneous execution of $P G_{1}$ and $P G_{2}$.

### 2.2.2 Communication via Shared Variables

The interleaving operator ||| can be used to model asynchronous concurrency in which the subprocesses act completely independent of each other, i.e., without any form of message passing or contentions on shared variables. The interleaving operator for transition systems is, however, too simplistic for most parallel systems with concurrent or communicating components. An example of a system whose components have variables in commonshared variables so to speak-will make this clear.

## Example 2.20. The Interleaving Operator for Concurrent Processes

Regard the program graph for the instructions $\alpha$ and $\beta$ of the parallel program

$$
\underbrace{x:=2 \cdot x}_{\text {action } \alpha} \| \underbrace{x:=x+1}_{\text {action } \beta}
$$

where we assume that initially $x=3$. (To simplify the picture, the locations have been skipped.) The transition system $T S\left(P G_{1}\right) \| T S\left(P G_{2}\right)$ contains, e.g., the inconsistent state $\langle x=6, x=4\rangle$ and, thus, does not reflect the intuitive behavior of the parallel execution of $\alpha$ and $\beta$ :


The problem in this example is that the actions $\alpha$ and $\beta$ access the shared variable $x$ and therefore are competing. The interleaving operator for transition systems, however, "blindly" constructs the Cartesian product of the individual state spaces without considering these potential conflicts. Accordingly, it is not identified that the local states $x=6$ and $x=4$ describe exclusive events.

In order to deal with parallel programs with shared variables, an interleaving operator will be defined on the level of program graphs (instead of directly on transition systems). The interleaving of program graphs $P G_{1}$ and $P G_{2}$ is denoted $P G_{1} \| P G_{2}$. The underlying transition system of the resulting program graph $P G_{1} \| \mid P G_{2}$, i.e., $T S\left(P G_{1}| | P G_{2}\right)$ (see Definition 2.15 , page 34 ) faithfully describes a parallel system whose components communicate via shared variables. Note that, in general, $T S\left(P G_{1} \| \mid P G_{2}\right) \neq T S\left(P G_{1}\right)| | \mid T S\left(P G_{2}\right)$.

## Definition 2.21. Interleaving of Program Graphs

Let $P G_{i}=\left(\operatorname{Loc}_{i}, A c t_{i}\right.$, Effect $\left._{i}, \hookrightarrow_{i}, \operatorname{Loc}_{0, i}, g_{0, i}\right)$, for $i=1,2$ be two program graphs over the variables $\operatorname{Var}_{i}$. Program graph $P G_{1} \| \mid P G_{2}$ over $\operatorname{Var}_{1} \cup \operatorname{Var}_{2}$ is defined by

$$
P G_{1} \| \mid P G_{2}=\left(L_{0} \times \text { Loc }_{2}, A c t_{1} \uplus A c t_{2}, \text { Effect }, \hookrightarrow, \operatorname{Loc}_{0,1} \times \operatorname{Loc}_{0,2}, g_{0,1} \wedge g_{0,2}\right)
$$

where $\hookrightarrow$ is defined by the rules:

$$
\frac{\ell_{1} \stackrel{g: \alpha}{\rightrightarrows} \ell_{1}^{\prime}}{\left\langle\ell_{1}, \ell_{2}\right\rangle \stackrel{g: \alpha}{\longleftrightarrow}\left\langle\ell_{1}^{\prime}, \ell_{2}\right\rangle} \quad \text { and } \frac{\ell_{2} \stackrel{g: \alpha}{\rightrightarrows} 2 \ell_{2}^{\prime}}{\left\langle\ell_{1}, \ell_{2}\right\rangle \stackrel{g: \alpha}{\longleftrightarrow}\left\langle\ell_{1}, \ell_{2}^{\prime}\right\rangle}
$$

and $\operatorname{Effect}(\alpha, \eta)=\operatorname{Effect}_{i}(\alpha, \eta)$ if $\alpha \in \operatorname{Act}_{i}$.

The program graphs $P G_{1}$ and $P G_{2}$ have the variables $\operatorname{Var}_{1} \cap \operatorname{Var}_{2}$ in common. These are the shared (sometimes also called "global") variables. The variables in $\operatorname{Var}_{1} \backslash \operatorname{Var}_{2}$ are the local variables of $P G_{1}$, and similarly, those in $\operatorname{Var}_{2} \backslash \operatorname{Var}_{1}$ are the local variables of $P G_{2}$.

## Example 2.22. Interleaving of Program Graphs

Consider the program graphs $P G_{1}$ and $P G_{2}$ that correspond to the assignments $x:=x+1$ and $x:=2 \cdot x$, respectively. The program graph $P G_{1} \| \mid P G_{2}$ is depicted in the bottom left of Figure 2.5. Its underlying transition system $T S\left(P G_{1} \| \mid P G_{2}\right)$ is depicted in the bottom right of that figure where it is assumed that initially $x$ equals 3 . Note that the nondeterminism in the initial state of the transition system does not represent concurrency but just the possible resolution of the contention between the statements $x:=2 \cdot x$ and $x:=x+1$ that both modify the shared variable $x$.

The distinction between local and shared variables has also an impact on the actions of the composed program graph $P G_{1} \| P G_{2}$. Actions that access (i.e., inspect or modify) shared
$P G_{1}$ :

$P G_{2}: \quad \frac{7}{\ell_{2}}$


Figure 2.5: Interleaving of two example program graphs.
variables may be considered as "critical"; otherwise, they are viewed to be noncritical. (For the sake of simplicity, we are a bit conservative here and consider the inspection of shared variables as critical.) The difference between the critical and noncritical actions becomes clear when interpreting the (possible) nondeterminism in the transition system $T S\left(P G_{1} \| \mid P G_{2}\right)$. nondeterminism in a state of this transition system may stand either for
(i) an "internal" nondeterministic choice within program graph $P G_{1}$ or $P G_{2}$,
(ii) the interleaving of noncritical actions of $P G_{1}$ and $P G_{2}$, or
(iii) the resolution of a contention between critical actions of $P G_{1}$ and $P G_{2}$ (concurrency).

In particular, a noncritical action of $P G_{1}$ can be executed in parallel with critical or noncritical actions of $P G_{2}$ as it will only affect its local variables. By symmetry, the same applies to the noncritical actions of $P G_{2}$. Critical actions of $P G_{1}$ and $P G_{2}$, however, cannot be executed simultaneously as the value of the shared variables depends on the order of executing these actions (see Example 2.20). Instead, any global state where critical actions of $P G_{1}$ and $P G_{2}$ are enabled describe a concurrency situation that has to be resolved by an appropriate scheduling strategy. (Simultaneous reading of shared variables could be allowed, however.)

## Remark 2.23. On Atomicity

For modeling a parallel system by means of the interleaving operator for program graphs it is decisive that the actions $\alpha \in$ Act are indivisible. The transition system representation only expresses the effect of the completely executed action $\alpha$. If there is, e.g., an action $\alpha$ with its effect being described by the statement sequence

$$
x:=x+1 ; y:=2 x+1 ; \text { if } x<12 \text { then } z:=(x-z)^{2} * y \mathbf{f i},
$$

then an implementation is assumed which does not interlock the basic substatements $x:=$ $x+1, y:=2 x+1$, the comparison " $x<12$ ", and, possibly, the assignment $z:=(x-z)^{2} * y$ with other concurrent processes. In this case,

$$
\begin{array}{lll}
\operatorname{Effect}(\alpha, \eta)(x) & =\eta(x)+1 \\
\operatorname{Effect}(\alpha, \eta)(y) & =2(\eta(x)+1)+1 & \\
\operatorname{Effect}(\alpha, \eta)(z) & = \begin{cases}(\eta(x)+1-\eta(z))^{2} * 2(\eta(x)+1)+1 & \text { if } \eta(x)+1<12 \\
\eta(z) & \text { otherwise }\end{cases}
\end{array}
$$

Hence, statement sequences of a process can be declared atomic by program graphs when put as a single label to an edge. In program texts such multiple assignments are surrounded by brackets $\langle\ldots\rangle$.


Figure 2.6: Individual program graphs for semaphore-based mutual exclusion.

## Example 2.24. Mutual Exclusion with Semaphores

Consider two simplified processes $P_{i}, i=1,2$ of the form:

```
Pi loop forever
    \vdots}\quad(* noncritical actions *)
    request
    critical section
    release
    \vdots}\quad(* noncritical actions *)
    end loop
```

Processes $P_{1}$ and $P_{2}$ are represented by the program graphs $P G_{1}$ and $P G_{2}$, respectively, that share the binary semaphore $y . y=0$ indicates that the semaphore - the lock to get access to the critical section-is currently possessed by one of the processes. When $y=1$, the semaphore is free. The program graphs $P G_{1}$ and $P G_{2}$ are depicted in Figure 2.6.

For the sake of simplicity, local variables and shared variables different from $y$ are not considered. Also, the activities inside and outside the critical sections are omitted. The locations of $P G_{i}$ are noncrit (representing the noncritical actions), wait (modeling the situation in which $P_{i}$ waits to enter its critical section), and crit ${ }_{i}$ (modeling the critical section). The program graph $P G_{1} \| \mid P G_{2}$ consists of nine locations, including the (undesired) location $\left\langle\right.$ crit $_{1}$, crit $\left._{2}\right\rangle$ that models the situation where both $P_{1}$ and $P_{2}$ are in their critical section, see Figure 2.7.

When unfolding $P G_{1} \| \mid P G_{2}$ into the transition system $T S_{S e m}=T S\left(P G_{1}| | \mid P G_{2}\right)$ (see Figure 2.8 on page 45), it can be easily checked that from the 18 global states in $T S_{S e m}$


Figure 2.7: $P G_{1}| | \mid P G_{2}$ for semaphore-based mutual exclusion.
only the following eight states are reachable:

$$
\begin{array}{ll}
\left\langle\text { noncrit }_{1}, \text { noncrit }_{2}, y=1\right\rangle & \left\langle\text { noncrit }_{1}, \text { wait }_{2}, y=1\right\rangle \\
\left\langle\text { wait }_{1}, \text { noncrit }_{2}, y=1\right\rangle & \left\langle\text { wait }_{1}, \text { wait }_{2}, y=1\right\rangle \\
\left\langle\text { noncrit }_{1}, \text { crit }_{2}, y=0\right\rangle & \left\langle\text { crit }_{1}, \text { noncrit }_{2}, y=0\right\rangle \\
\left\langle\text { wait }_{1}, \text { crit }_{2}, y=0\right\rangle & \left\langle\text { crit }_{1}, \text { wait }_{2}, y=0\right\rangle
\end{array}
$$

States $\left\langle\right.$ noncrit $_{1}$, noncrit $\left.{ }_{2}, y=1\right\rangle$, and $\left\langle\right.$ noncrit $_{1}$, crit $\left.{ }_{2}, y=0\right\rangle$ stand for examples of situations where both $P_{1}$ and $P_{2}$ are able to concurrently execute actions. Note that in Figure $2.8 n$ stands for noncrit, $w$ for wait, and $c$ for crit. The nondeterminism in these states thus stand for interleaving of noncritical actions. State $\left\langle\right.$ crit $_{1}$, wait $\left.{ }_{2}, y=0\right\rangle$, e.g., represents a situation where only $P G_{1}$ is active, whereas $P G_{2}$ is waiting.

From the fact that the global state $\left\langle\right.$ crit $_{1}$, crit $\left._{2}, y=\ldots\right\rangle$ is unreachable in $T S_{S e m}$, it follows that processes $P_{1}$ and $P_{2}$ cannot be simultaneously in their critical section. The parallel system thus satisfies the so-called mutual exclusion property.

In the previous example, the nondeterministic choice in state $\left\langle\right.$ wait $_{1}$, wait $\left.{ }_{2}, y=1\right\rangle$ represents a contention between allowing either $P_{1}$ or $P_{2}$ to enter its critical section. The resolution of this scheduling problem-which process is allowed to enter its critical section next? - is left open, however. In fact, the parallel program of the previous example is "abstract" and does not provide any details on how to resolve this contention. At later design stages, for example, when implementing the semaphore $y$ by means of a queue of waiting processes (or the like), a decision has to be made on how to schedule the processes that are enqueued for acquiring the semaphore. At that stage, a last-in first-out (LIFO),


Figure 2.8: Mutual exclusion with semaphore (transition system representation).
first-in first-out (FIFO), or some other scheduling discipline can be chosen. Alternatively, another (more concrete) mutual exclusion algorithm could be selected that resolves this scheduling issue explicitly. A prominent example of such algorithm has been provided in 1981 by Peterson [332].

## Example 2.25. Peterson's Mutual Exclusion Algorithm

Consider the processes $P_{1}$ and $P_{2}$ with the shared variables $b_{1}, b_{2}$, and $x$. $b_{1}$ and $b_{2}$ are Boolean variables, while $x$ can take either the value 1 or 2 , i.e., $\operatorname{dom}(x)=\{1,2\}$. The scheduling strategy is realized using $x$ as follows. If both processes want to enter the critical section (i.e., they are in location wait $i_{i}$ ), the value of variable $x$ decides which of the two processes may enter its critical section: if $x=i$, then $P_{i}$ may enter its critical section (for $i=1,2$ ). On entering location wait ${ }_{1}$, process $P_{1}$ performs $x:=2$, thus giving privilege to process $P_{2}$ to enter the critical section. The value of $x$ thus indicates which process has its turn to enter the critical section. Symmetrically, $P_{2}$ sets $x$ to 1 when starting to wait. The variables $b_{i}$ provide information about the current location of $P_{i}$. More precisely,

$$
b_{i}=\text { wait }_{i} \vee \text { crit }_{i} .
$$

$b_{i}$ is set when $P_{i}$ starts to wait. In pseudocode, $P_{1}$ performs as follows (the code for process $P_{2}$ is similar):


Figure 2.9: Program graphs for Peterson's mutual exclusion algorithm.

```
P
    \vdots}\quad(*\mathrm{ noncritical actions *)
    \langleb
        (* request *)
    wait until (x=1 \vee \neg\mp@subsup{b}{2}{})
    do critical section od
    b
    \vdots (* noncritical actions *)
    end loop
```

Process $P_{i}$ is represented by program graph $P G_{i}$ over Var $=\left\{x, b_{1}, b_{2}\right\}$ with locations noncrit ${ }_{i}$, wait ${ }_{i}$, and $c r i t_{i}$, see Figure 2.9 above. The reachable part of the underlying transition system $T S_{P e t}=T S\left(P G_{1} \| \mid P G_{2}\right)$ has the form as indicated in Figure 2.10 (page 47), where for convenience $n_{i}, w_{i}, c_{i}$ are used for noncrit $i_{i}$, wait $i_{i}$, and $c^{2} t_{i}$, respectively. The last digit of the depicted states indicates the evaluation of variable $x$. For convenience, the values for $b_{i}$ are not indicated. Its evaluation can directly be deduced from the location of $P G_{i}$. Further, $b_{1}=b_{2}=$ false is assumed as the initial condition.

Each state in $T S_{P e t}$ has the form $\left\langle l o c_{1}, \operatorname{loc}_{2}, x, b_{1}, b_{2}\right\rangle$. As $P G_{i}$ has three possible locations and $b_{i}$ and $x$ each can take two different values, the total number of states of $T S_{P e t}$ is 72 . Only ten of these states are reachable. Since there is no reachable state with $P_{1}$ and $P_{2}$ being in their critical section, it can be concluded that Peterson's algorithm satisfies the mutual exclusion property.

In the above program, the multiple assignments $b_{1}:=\operatorname{true} ; x:=2$ and $b_{2}:=$ true; $x:=1$ are considered as indivisible (i.e., atomic) actions. This is indicated by the brackets <


Figure 2.10: Transition system of Peterson's mutual exclusion algorithm.
and $\rangle$ in the program text, and is also indicated in the program graphs $P G_{1}$ and $P G_{2}$. We like to emphasize that this is not essential, and has only been done to simplify the transition system $T S_{\text {Pet }}$. Mutual exclusion is also ensured when both processes perform the assignments $b_{i}:=$ true and $x:=\ldots$ in this order but in a nonatomic way. Note that, for instance, the order "first $x:=\ldots$, then $b_{i}:=$ true" does not guarantee mutual exclusion. This can be seen as follows. Assume that the location inbetween the assignments $x:=\ldots$ and $b_{i}:=$ true in program graph $P_{i}$ is called req $q_{i}$. The state sequence

$$
\begin{array}{ccccc}
\begin{array}{cccc}
\text { noncrit }
\end{array}, & \text { noncrit }_{2}, & x=1, & b_{1}=\text { false }, & \left.b_{2}=\text { false }\right\rangle \\
\left\langle\text { noncrit }_{1},\right. & \text { req }_{2}, & x=1, & b_{1}=\text { false }, & \left.b_{2}=\text { false }\right\rangle \\
\left\langle\text { req }_{1},\right. & \text { req }_{2}, & x=2, & b_{1}=\text { false, }, & \left.b_{2}=\text { false }\right\rangle \\
\left\langle\text { wait }_{1},\right. & \text { req }_{2}, & x=2, & b_{1}=\text { true, } & \left.b_{2}=\text { false }\right\rangle \\
\left\langle\text { crit }_{1},\right. & \text { req }_{2}, & x=2, & b_{1}=\text { true }, & \left.b_{2}=\text { false }\right\rangle \\
\left\langle\text { crit }_{1},\right. & \text { wait }_{2}, & x=2, & b_{1}=\text { true }, & \left.b_{2}=\text { true }\right\rangle \\
\left\langle\text { crit }_{1},\right. & \text { crit }_{2}, & x=2, & b_{1}=\text { true }, & \left.b_{2}=\text { true }\right\rangle
\end{array}
$$

is an initial execution fragment where $P_{1}$ enters its critical section first (as $b_{2}=$ false) after which $P_{2}$ enters its critical section (as $x=2$ ). As a result, both processes are simultaneously in their critical section and mutual exclusion is violated.

### 2.2.3 Handshaking

So far, two mechanisms for parallel processes have been considered: interleaving and shared-variable programs. In interleaving, processes evolve completely autonomously

- interleaving for $\alpha \notin H$ :

$$
\frac{s_{1} \xrightarrow{\alpha} 1 s_{1}^{\prime}}{\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha}\left\langle s_{1}^{\prime}, s_{2}\right\rangle} \quad \frac{s_{2} \xrightarrow{\alpha}_{2} s_{2}^{\prime}}{\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha}\left\langle s_{1}, s_{2}^{\prime}\right\rangle}
$$

- handshaking for $\alpha \in H$ :

$$
\frac{s_{1} \xrightarrow{\alpha}_{1} s_{1}^{\prime} \stackrel{\wedge}{\wedge} s_{2} \xrightarrow{\alpha}}{2} s_{2}^{\prime}{ }_{\left\langle s_{1}, s_{2}\right\rangle \xrightarrow[\longrightarrow]{\alpha}\left\langle s_{1}^{\prime}, s_{2}^{\prime}\right\rangle}
$$

Figure 2.11: Rules for handshaking.
whereas according to the latter type processes "communicate" via shared variables. In this subsection, we consider a mechanism by which concurrent processes interact via handshaking. The term "handshaking" means that concurrent processes that want to interact have to do this in a synchronous fashion. That is to say, processes can interact only if they are both participating in this interaction at the same time - they "shake hands".

Information that is exchanged during handshaking can be of various nature, ranging from the value of a simple integer, to complex data structures such as arrays or records. In the sequel, we do not dwell upon the content of the exchanged messages. Instead, an abstract view is adopted and only communication (also called synchronization) actions are considered that represent the occurrence of a handshake and not the content.

To do so, a set $H$ of handshake actions is distinguished with $\tau \notin H$. Only if both participating processes are ready to execute the same handshake action, can message passing take place. All actions outside $H$ (i.e., actions in $A c t \backslash H$ ) are independent and therefore can be executed autonomously in an interleaved fashion.

## Definition 2.26. Handshaking (Synchronous Message Passing)

Let $T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P_{i}, L_{i}\right), i=1,2$ be transition systems and $H \subseteq A c t_{1} \cap$ Act $_{2}$ with $\tau \notin H$. The transition system $T S_{1} \|_{H} T S_{2}$ is defined as follows:

$$
T S_{1} \|_{H} T S_{2}=\left(S_{1} \times S_{2}, A c t_{1} \cup A c t_{2}, \rightarrow, I_{1} \times I_{2}, A P_{1} \cup A P_{2}, L\right)
$$

where $L\left(\left\langle s_{1}, s_{2}\right\rangle\right)=L_{1}\left(s_{1}\right) \cup L_{2}\left(s_{2}\right)$, and where the transition relation $\rightarrow$ is defined by the rules shown in Figure 2.11.

Notation: $T S_{1} \| T S_{2}$ abbreviates $T S_{1} \|_{H} T S_{2}$ for $H=$ Act $_{1} \cap$ Act $_{2}$.

## Remark 2.27. Empty Set of Handshake Actions

When the set $H$ of handshake actions is empty, all actions of the participating processes can take place autonomously, i.e., in this special case, handshaking reduces to interleaving

$$
T S_{1}\left\|_{\varnothing} T S_{2}=T S_{1}\right\| T S_{2}
$$

The operator $\|_{H}$ defines the handshaking between two transition systems. Handshaking is commutative, but not associative in general. That is, in general we have

$$
T S_{1}\left\|_{H}\left(T S_{2} \|_{H^{\prime}} T S_{3}\right) \neq\left(T S_{1} \|_{H} T S_{2}\right)\right\|_{H^{\prime}} T S_{3} \quad \text { for } H \neq H^{\prime}
$$

However, for a fixed set $H$ of handshake actions over which all processes synchronize, the operator $\|_{H}$ is associative. Let

$$
T S=T S_{1}\left\|_{H} T S_{2}\right\|_{H} \ldots \|_{H} T S_{n}
$$

denote the parallel composition of transition systems $T S_{1}$ through $T S_{n}$ where $H \subseteq A c t_{1} \cap$ $\ldots \cap A c t_{n}$ is a subset of the set of actions $A c t_{i}$ of all transition systems. This form of multiway handshaking is appropriate to model broadcasting, a communication form in which a process can transmit a datum to several other processes simultaneously.

In many cases, processes communicate in a pairwise fashion over their common actions. Let $T S_{1}\|\ldots\| T S_{n}$ denote the parallel composition of $T S_{1}$ through $T S_{n}$ (with $n>0$ ) where $T S_{i}$ and $T S_{j}(0<i \neq j \leqslant n)$ synchronize over the set of actions $H_{i, j}=A c t_{i} \cap A c t_{j}$ such that $H_{i, j} \cap A c t_{k}=\varnothing$ for $k \notin\{i, j\}$. It is assumed that $\tau \notin H_{i, j}$. The formal definition of $T S_{1}\|\ldots\| T S_{n}$ is analogous to Definition 2.26. The state space of $T S_{1}\|\ldots\| T S_{n}$ results from the Cartesian product of the state spaces of $T S_{i}$. The transition relation $\rightarrow$ is defined by the following rules:

- for $\alpha \in \operatorname{Act}_{i} \backslash\left(\underset{\substack{0<j \leqslant n \\ i \neq j}}{\bigcup} H_{i, j}\right)$ and $0<i \leqslant n$ :

$$
\frac{s_{i} \xrightarrow{\alpha}{ }_{i}^{\prime} s_{i}^{\prime}}{\left\langle s_{1}, \ldots, s_{i}, \ldots, s_{n}\right\rangle \xrightarrow[\longrightarrow]{s_{1}}\left\langle s_{1}, \ldots, s_{i}^{\prime}, \ldots s_{n}\right\rangle}
$$

- for $\alpha \in H_{i, j}$ and $0<i<j \leqslant n$ :

$$
\frac{s_{i} \xrightarrow{\alpha}_{i} s_{i}^{\prime} \wedge s_{j} \xrightarrow{\alpha}{ }_{j} s_{j}^{\prime}}{\left\langle s_{1}, \ldots, s_{i}, \ldots, s_{j}, \ldots, s_{n}\right\rangle \xrightarrow{\alpha}\left\langle s_{1}, \ldots, s_{i}^{\prime}, \ldots, s_{j}^{\prime}, \ldots, s_{n}\right\rangle}
$$

According to the first rule, components can execute actions that are not subject to handshaking in a completely autonomous manner as in interleaving. The second rule states that processes $T S_{i}$ and $T S_{j}(i \neq j)$ have to perform every handshaking action in $A c t_{i} \cap$ Act $_{j}$ together. These rules are in fact just generalizations of those given in Figure 2.11.

## Example 2.28. Mutual Exclusion by Means of an Arbiter

An alternative solution to the mutual exclusion problem between processes $P_{1}$ and $P_{2}$ (as before) is to model the binary semaphore that regulates access to the critical section by a separate parallel process that interacts with $P_{1}$ and $P_{2}$ by means of handshaking. For simplicity, we ignore the waiting phase and assume that $P_{i}$ simply alternates infinitely often between noncritical and critical sections. Assume (much simplified) transition system representations $T S_{1}$ and $T S_{2}$ with just two states: crit $i_{i}$ and noncrit ${ }_{i}$. The new process, named Arbiter, mimics a binary semaphore (see Figure 2.12). $P_{1}$ and $P_{2}$ communicate with the Arbiter via handshaking over the set $H=\{$ request, rel $\}$. Accordingly, the actions request (requesting to access the critical section) and rel (to leave the critical section) have to be executed synchronously with the Arbiter. The complete system

$$
T S_{A r b}=\left(T S_{1} \| T S_{2}\right) \| \text { Arbiter }
$$

guarantees mutual exclusion since there are no states of $T S_{A r b}$ where both $P_{1}$ and $P_{2}$ are in their critical section (see bottom part of Figure 2.12). Note that in the initial state of $T S_{1}| | \mid T S_{2}$, the Arbiter determines which process will enter the critical section next.

## Example 2.29. Booking System

Consider a (strongly simplified) booking system at a cashier of a supermarket. The system consists of three processes: the bar code reader $B C R$, the actual booking program $B P$, and the printer Printer. The bar code reader reads a bar code and communicates the data of the just scanned product to the booking program. On receiving such data, the booking program transmits the price of the article to the printer that prints the article Id together with the price on the receipt. The interactions between the bar code reader and the booking program, and between the booking program and the printer is performed by handshaking. Each process consist of just two states, named 0 and 1 (see Figure 2.13 for the transitions systems of $B C R, B P$, and Printer).

The complete system is given by:

$$
B C R\|B P\| \text { Printer. }
$$

The transition system of the overall system is depicted in Figure 2.14 on page 52. The initial global state of this system is $\langle 0,0,0\rangle$, or in short, 000 . In global state 010 , e.g., the

$\left(T_{1}| | \mid T_{2}\right)|\mid$ Arbiter :


Figure 2.12: Mutual exclusion using handshaking with arbiter process.
nondeterminism stands for the concurrent execution of the actions scanning the bar code and the synchronous transfer of the price to the printer.

## Example 2.30. Railroad Crossing

For a railroad crossing a control system needs to be developed that on receipt of a signal indicating that a train is approaching closes the gates, and only opens these gates after the train has sent a signal indicating that it crossed the road. The requirement that should be met by the control system is that the gates are always closed when the train is crossing the road. The complete system consists of the three components Train, Gate, and Controller:

Train || Gate || Controller.
Figure 2.15 depicts the transition systems of these components from left (modeling the Train) to right (modeling the Gate). For simplicity, it is assumed that all trains pass the relevant track section in the same direction-from left to right. The states of the transition system for the Train have the following intuitive meaning: in state far the train is not close to the crossing, in state near it is approaching the crossing and has just sent a signal to notify this, and in state in it is at the crossing. The states of the Gate have the obvious interpretation. The state changes of the Controller stand for handshaking with


Figure 2.13: The components of the book keeping example.


Figure 2.14: Transition system representation of the booking system.
the trains (via the actions approach and exit) and the Gate (via the actions lower and raise via which the Controller causes the gate to close or to open, respectively).

Figure 2.16 (above) illustrates the transition system of the overall system. A closer inspection of this transition system reveals that the system suffers from a design flaw. This can be seen from the following initial execution fragment:

$$
\langle\text { far }, 0, u p\rangle \xrightarrow{\text { approach }}\langle\text { near }, 1, \text { up }\rangle \xrightarrow{\text { enter }}\langle\text { in, } 1, \text { up }\rangle
$$

in which the gate is about to close, while the train is (already) at the crossing. The nondeterminism in global state $\langle n e a r, 1, u p\rangle$ stands for concurrency: the train approaches the crossing, while the gate is being closed. In fact, the basic concept of the design is correct if and only if closing the gate does not take more time than the train needs to get to the crossing once it signals-"I am approaching". Such real-time constraints cannot be formulated by the concepts introduced so far. The interleaving representation for parallel systems is completely time-abstract. In Chapter 9, concepts and techniques will be introduced to specify and verify such real-time aspects.


Figure 2.15: The components of the railroad crossing.

### 2.2.4 Channel Systems

This section introduces channel systems, parallel systems where processes communicate via so-called channels, i.e., first-in, first-out buffers that may contain messages. We consider channel systems that are closed. That is to say, processes may communicate with other processes in the system (via channels), but not with processes outside the system. Channel systems are popular for describing communication protocols and form the basis of Promela, the input language of the Spin model checker.

Intuitively, a channel system consists of $n$ (data-dependent) processes $P_{1}$ through $P_{n}$. Each $P_{i}$ is specified by a program graph $P G_{i}$ which is extended with communication actions. Transitions of these program graphs are either the usual conditional transitions (labeled with guards and actions) as before, or one of the communication actions with their respective intuitive meaning:

$$
\begin{array}{ll}
c!v & \text { transmit the value } v \text { along channel } c, \\
c ? x & \text { receive a message via channel } c \text { and assign it to variable } x .
\end{array}
$$

When considering channel $c$ as buffer, the communication action $c!v$ puts value $v$ (at the rear of) the buffer whereas $c$ ? $x$ retrieves an element from (the front of) the buffer while assigning it to $x$. It is assumed implicitly that variable $x$ is of the right type, i.e., it has a type that is compatible to that of the messages that are put into channel $c$. Let

$$
\operatorname{Comm}=\{c!v, c ? x \mid c \in \text { Chan, } v \in \operatorname{dom}(c), x \in \operatorname{Var} \text { with } \operatorname{dom}(x) \supseteq \operatorname{dom}(c)\}
$$

denote the set of communication actions where Chan is a finite set of channels with typical element $c$.


Figure 2.16: Transition system for the railroad crossing.

A channel $c$ has a (finite or infinite) capacity indicating the maximum number of messages it can store, and a type (or domain) specifying the type of messages that can be transmitted over $c$. Each channel $c$ has a capacity $\operatorname{cap}(c) \in \mathbb{N} \cup\{\infty\}$, and a domain $\operatorname{dom}(c)$. For a channel $c$ that can only transfer bits, $\operatorname{dom}(c)=\{0,1\}$. If complete texts (of maximum length of 200, say) need to be transmitted over channel $c$, then another type of channel has to be used such that $\operatorname{dom}(c)=\Sigma \leqslant 200$, where $\Sigma$ is the alphabet that forms the basis of the texts, e.g., $\Sigma$ is the set of all letters and special characters used in German texts.

The capacity of a channel defines the size of the corresponding buffer, i.e., the number of messages not yet read that can be stored in the buffer. When $\operatorname{cap}(c) \in \mathbb{N}, c$ is a channel with finite capacity; $\operatorname{cap}(c)=\infty$ indicates that $c$ has an infinite capacity. Note that the special case $\operatorname{cap}(c)=0$ is permitted. In this case, channel $c$ has no buffer. Communication via such a channel $c$ corresponds to handshaking (simultaneous transmission and receipt, i.e., synchronous message passing) plus the exchange of some data. When $\operatorname{cap}(c)>0$, there is a "delay" between the transmission and the receipt of a message, i.e., sending and reading of the same message take place at different moments. This is called asynchronous message passing. Sending and reading a message from a channel with a nonzero capacity can never appear simultaneously. By means of channel systems, both synchronous and asynchronous message passing can thus be modeled.

## Definition 2.31. Channel System

A program graph over (Var, Chan) is a tuple

$$
P G=\left(\text { Loc }, \text { Act }, \text { Effect }, \hookrightarrow, \text { Loc }_{0}, g_{0}\right)
$$

according to Definition 2.13 (page 32) with the only difference that

$$
\hookrightarrow \subseteq \operatorname{Loc} \times(\operatorname{Cond}(\operatorname{Var}) \times(\text { Act } \cup \operatorname{Comm}) \times \text { Loc } .
$$

A channel system CS over (Var, Chan) consists of program graphs $P G_{i}$ over ( $\operatorname{Var}_{i}$, Chan) (for $1 \leqslant i \leqslant n$ ) with Var $=\bigcup_{1 \leqslant i \leqslant n} \operatorname{Var}_{i}$. We denote

$$
C S=\left[P G_{1}|\ldots| P G_{n}\right] .
$$

The transition relation $\hookrightarrow$ of a program graph over (Var, Chan) consists of two types of conditional transitions. As before, conditional transitions $\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}$ are labeled with guards and actions. These conditional transitions can happen whenever the guard holds. Alternatively, conditional transitions may be labeled with communication actions. This yields conditional transitions of type $\ell \stackrel{\text { g:c! } v}{\hookrightarrow} \ell^{\prime}$ (for sending $v$ along $c$ ) and $\ell \stackrel{g: c \text { ? } x}{\hookrightarrow} \ell^{\prime}$ (for receiving a message along $c$ ). When can such conditional transitions happen? Stated
differently, when are these conditional transitions executable? This depends on the current variable evaluation and the capacity and content of the channel $c$. For the sake of simplicity assume in the following that the guard is satisfied.

- Handshaking. If $\operatorname{cap}(c)=0$, then process $P_{i}$ can transmit a value $v$ over channel $c$ by performing

$$
\ell_{i} \stackrel{c!v}{\longrightarrow} \ell_{i}^{\prime}
$$

only if another process $P_{j}$, say, "offers" a complementary receive action, i.e., can perform

$$
\ell_{j} \stackrel{c ? x}{\hookrightarrow} \ell_{j}^{\prime} .
$$

$P_{i}$ and $P_{j}$ should thus be able to perform $c!v\left(\right.$ in $P_{i}$ ) and $c ? x$ (in $P_{j}$ ) simultaneously. Then, message passing can take place between $P_{i}$ and $P_{j}$. The effect of message passing corresponds to the (distributed) assignment $x:=v$.
Note that when handshaking is only used for synchronization purposes and not for data transfer, the name of the channel as well as the value $v$ are not of any relevance.

- Asynchronous message passing. If $\operatorname{cap}(c)>0$, then process $P_{i}$ can perform the conditional transition

$$
\ell_{i} \stackrel{c!v}{\longrightarrow} \ell_{i}^{\prime}
$$

if and only if channel $c$ is not full, i.e., if less than $\operatorname{cap}(c)$ messages are stored in $c$. In this case, $v$ is stored at the rear of the buffer $c$. Channels are thus considered as first-in, first-out buffers. Accordingly, $P_{j}$ may perform

$$
\ell_{j} \stackrel{c ? x}{\hookrightarrow} \ell_{j}^{\prime}
$$

if and only if the buffer of $c$ is not empty. In this case, the first element $v$ of the buffer is extracted and assigned to $x$ (in an atomic manner). This is summarized in Table 2.1.

|  | executable if $\ldots$ | effect |
| :--- | :--- | :--- |
| $c!v$ | $c$ is not "full" | Enqueue $(c, v)$ |
| $c ? x$ | $c$ is not empty | $\langle x:=\operatorname{Front}(c) ; \operatorname{Dequeue}(c)\rangle ;$ |

Table 2.1: Enabledness and effect of communication actions if $\operatorname{cap}(c)>0$.


Figure 2.17: Schematic view of the alternating bit protocol.

Channel systems are often used to model communication protocols. One of the most elementary and well-known protocols is the alternating bit protocol.

## Example 2.32. Alternating Bit Protocol (ABP)

Consider a system essentially consisting of a sender $S$ and a receiver $R$ that communicate with each other over channels $c$ and $d$, see Figure 2.17. It is asssumed that both channels have an unlimited buffer, i.e., $\operatorname{cap}(c)=\operatorname{cap}(d)=\infty$. Channel $c$ is unreliable in the sense that data may get lost when being transmitted from the sender $S$ to channel $c$. Once messages are stored in the buffer of channel $c$, they are neither corrupted nor lost. Channel $d$ is assumed to be perfect. The goal is to design a communication protocol that ensures any distinct transmitted datum by $S$ to be delivered to $R$. To ensure this in the presence of possible message losses, sender $S$ resorts to retransmissions. Messages are transmitted one by one, i.e., $S$ starts sending a new message once the transmission of the previous message has been successful. This is a simple flow control principle, known as "send-and-wait".

We abstract from the real activities of $S$ and $R$ and, instead, concentrate on a simplified representation of the communication structure of the system. $S$ sends the successive messages $m_{0}, m_{1}, \ldots$ together with control bits $b_{0}, b_{1}, \ldots$ over channel $c$ to $R$. Transmitted messages are thus pairs:

$$
\left\langle m_{0}, 0\right\rangle,\left\langle m_{1}, 1\right\rangle,\left\langle m_{2}, 0\right\rangle,\left\langle m_{3}, 1\right\rangle, \ldots
$$

On receipt of $\langle m, b\rangle$ (along channel $c$ ), $R$ sends an acknowledgment (ack) consisting of the control bit $b$ just received. On receipt of ack $\langle b\rangle, S$ transmits a new message $m^{\prime}$ with control bit $\neg b$. If, however, $S$ has to wait "too long" for the ack, it timeouts and retransmits $\langle m, b\rangle$. The program graphs for $S$ and $R$ are sketched in Figure 2.18 and 2.19. For simplicity, the data that is transmitted is indicated by $m$ instead of $m_{i}$.


Figure 2.18: Program graph of ABP sender $S$.


Figure 2.19: Program graph of (left) ABP receiver $R$ and (right) Timer.

Control bit $b$-also called the alternating bit-is thus used to distinguish retransmissions of $m$ from transmissions of subsequent (and previous) messages. Due to the fact that the transmission of a new datum is initiated only when the last datum has been received correctly (and this is acked), a single bit is sufficient for this purpose and notions like, e.g., sequence numbers, are not needed.

The timeout mechanism of $S$ is modeled by a Timer process. $S$ activates this timer on sending a message (along $c$ ), and it stops the timer on receipt of an ack. When raising a timeout, the timer signals to $S$ that a retransmission should be initiated. (Note that due to this way of modeling, so-called premature timeouts may occur, i.e., a timeout may occur whereas an ack is still on its way to $S$.) The communication between the timer and $S$ is modeled by means of handshaking, i.e., by means of channels with capacity 0 .

The complete system can now be represented as the following channel system over Chan $=$ $\left\{c, d, t m r_{\_} o n, t m r_{-} o f f\right.$, timeout $\}$ and $\operatorname{Var}=\left\{x, y, m_{i}\right\}$ :

$$
A B P=[S \mid \text { Timer } \mid R]
$$

The following definition formalizes the successive behavior of a channel system by means of a transition system. The basic concept is similar to the mapping from a program graph onto a transition system. Let $C S=\left[P G_{1}|\ldots| P G_{n}\right]$ be a channel system over (Chan, Var). The (global) states of $T S(C S)$ are tuples of the form

$$
\left\langle\ell_{1}, \ldots, \ell_{n}, \eta, \xi\right\rangle
$$

where $\ell_{i}$ indicates the current location of component $P G_{i}, \eta$ keeps track of the current values of the variables, and $\xi$ records the current content of the various channels (as sequences). Formally, $\eta \in \operatorname{Eval}(\operatorname{Var})$ is an evaluation of the variables (as we have encountered before), and $\xi$ is a channel evaluation, i.e., a mapping from channel $c \in$ Chan onto a sequence $\xi(c) \in \operatorname{dom}(c)^{*}$ such that the length of the sequence cannot exceed the capacity of $c$, i.e., $\operatorname{len}(\xi(c)) \leqslant \operatorname{cap}(c)$ where len $(\cdot)$ denotes the length of a sequence. Eval(Chan) denotes the set of all channel evaluations. For initial states, the control components $\ell_{i} \in L o c_{0, i}$ must be initial and variable evaluation $\eta$ must satisfy the initial condition $g_{0}$. In addition, every channel is initially assumed to be empty, denoted $\varepsilon$.

Before providing the details of the semantics of a transition system, let us introduce some notations. Channel evaluation $\xi(c)=v_{1} v_{2} \ldots v_{k}$ (where $\operatorname{cap}(c) \geqslant k$ ) denotes that $v_{1}$ is at the front of the buffer of $c, v_{2}$ is the second element, etc., and $v_{k}$ is the element at the rear of $c$. len $(\xi(c))=k$ in this case. Let $\xi\left[c:=v_{1}, \ldots, v_{k}\right]$ denote the channel evaluation where sequence $v_{1}, \ldots, v_{k}$ is assigned to $c$ and all other channels are unaffected, i.e.,

$$
\xi\left[c:=v_{1} \ldots v_{k}\right]\left(c^{\prime}\right)= \begin{cases}\xi\left(c^{\prime}\right) & \text { if } c \neq c^{\prime} \\ v_{1} \ldots v_{k} & \text { if } c=c^{\prime}\end{cases}
$$

The channel evaluation $\xi_{0}$ maps any channel to the empty sequence, denoted $\varepsilon$, i.e., $\xi_{0}(c)=$ $\varepsilon$ for any channel $c$. Let len $(\varepsilon)=0$. The set of actions of $T S(C S)$ consists of actions $\alpha \in A c t_{i}$ of component $P G_{i}$ and the distinguished symbol $\tau$ representing all communication actions in which data is exchanged.

## Definition 2.33. Transition System Semantics of a Channel System

Let $C S=\left[P G_{1}|\ldots| P G_{n}\right]$ be a channel system over (Chan, Var) with

$$
P G_{i}=\left(\text { Loc }_{i}, \text { Act }_{i}, \text { Effect }_{i}, \hookrightarrow_{i}, \operatorname{Loc}_{0, i}, g_{0, i}\right), \quad \text { for } 0<i \leqslant n .
$$

The transition system of $C S$, denoted $T S(C S)$, is the tuple ( $S, A c t, \rightarrow, I, A P, L$ ) where:

- $S=\left(\operatorname{Loc}_{1} \times \ldots \times \operatorname{Loc}_{n}\right) \times \operatorname{Eval}(\operatorname{Var}) \times \operatorname{Eval}($ Chan $)$,
- Act $=\biguplus_{0<i \leqslant n} A c t_{i} \uplus\{\tau\}$,
- $\rightarrow$ is defined by the rules of Figure 2.20 (page 61),
- $I=\left\{\left\langle\ell_{1}, \ldots, \ell_{n}, \eta, \xi_{0}\right\rangle \mid \forall 0<i \leqslant n .\left(\ell_{i} \in \operatorname{Loc}_{0, i} \wedge \eta \vDash g_{0, i}\right)\right\}$,
- $A P=\biguplus_{0<i \leqslant n} \operatorname{Loc}_{i} \uplus \operatorname{Cond}(\operatorname{Var})$,
- $L\left(\left\langle\ell_{1}, \ldots, \ell_{n}, \eta, \xi\right\rangle\right)=\left\{\ell_{1}, \ldots, \ell_{n}\right\} \cup\{g \in \operatorname{Cond}(\operatorname{Var}) \mid \eta \models g\}$.

This definition is a formalization of the informal description of the interpretation of a channel system given before. Note that the labeling of the atomic propositions is similar to that for program graphs (see Definition 2.15). For the sake of simplicity, the above definition does not allow for propositions on channels. This could be accommodated by allowing for conditions on channels such as, e.g., "the channel $c$ is empty" or "the channel $c$ is full", and checking these conditions on the channel evaluation $\xi$ in a state.

## Example 2.34. Alternating Bit Protocol (Revisited)

Consider the alternating bit protocol that was modeled as a channel system in Example 2.32. The underlying transition system $T S(A B P)$ has, despite various simplifying assumptions, infinitely many states. This is, e.g., due to the fact that the timer may signal a timeout on each transmission of a datum by $S$ resulting in infinitely many messages in channel $c$.

To clarify the functionality of the alternating bit protocol, consider two execution fragments represented by indicating the states of the various components (sender $S$, receiver $R$, the timer, and the contents of channels $c$ and $d$ ). The first execution fragment shows the loss of a message. Here, $R$ does not execute any action at all as it only acts if channel $c$ contains at least one message:

| sender $S$ | timer | receiver $R$ | channel $c$ | channel $d$ | event |
| :--- | :--- | :--- | :--- | :--- | :--- |
| snd_msg(0) | off | wait $(0)$ | $\varnothing$ | $\varnothing$ |  |
| st_tmr $(0)$ | off | wait $(0)$ | $\varnothing$ | $\varnothing$ | loss of message |
| wait(0) | on | wait $(0)$ | $\varnothing$ | $\varnothing$ |  |
| snd_msg(0) | off | wait $(0)$ | $\varnothing$ | $\varnothing$ | timeout |
| $\vdots$ | $\vdots$ | $\vdots$ | $\vdots$ | $\vdots$ |  |

- interleaving for $\alpha \in A c t_{i}$ :

$$
\frac{\ell_{i} \stackrel{g: \alpha}{\hookrightarrow} \ell_{i}^{\prime} \wedge \eta \models g}{\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{n}, \eta, \xi\right\rangle \xrightarrow{\alpha}\left\langle\ell_{1}, \ldots, \ell_{i}^{\prime}, \ldots, \ell_{n}, \eta^{\prime}, \xi\right\rangle}
$$

where $\eta^{\prime}=\operatorname{Effect}(\alpha, \eta)$.

- asynchronous message passing for $c \in$ Chan, $\operatorname{cap}(c)>0$ :
- receive a value along channel $c$ and assign it to variable $x$ :

$$
\frac{\ell_{i} \stackrel{g: c ? x}{\hookrightarrow} \ell_{i}^{\prime} \wedge \eta \models g \wedge \operatorname{len}(\xi(c))=k>0 \wedge \xi(c)=v_{1} \ldots v_{k}}{\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{n}, \eta, \xi\right\rangle \xrightarrow{\tau}\left\langle\ell_{1}, \ldots, \ell_{i}^{\prime}, \ldots, \ell_{n}, \eta^{\prime}, \xi^{\prime}\right\rangle}
$$

where $\eta^{\prime}=\eta\left[x:=v_{1}\right]$ and $\xi^{\prime}=\xi\left[c:=v_{2} \ldots v_{k}\right]$.

- transmit value $v \in \operatorname{dom}(c)$ over channel $c$ :

$$
\frac{\ell_{i} \stackrel{g: c l v}{\hookrightarrow} \ell_{i}^{\prime} \wedge \eta=g \wedge \operatorname{len}(\xi(c))=k<\operatorname{cap}(c) \wedge \xi(c)=v_{1} \ldots v_{k}}{\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{n}, \eta, \xi\right\rangle \xrightarrow{\tau}\left\langle\ell_{1}, \ldots, \ell_{i}^{\prime}, \ldots, \ell_{n}, \eta, \xi^{\prime}\right\rangle}
$$

where $\xi^{\prime}=\xi\left[c:=v_{1} v_{2} \ldots v_{k} v\right]$.

- synchronous message passing over $c \in$ Chan, $\operatorname{cap}(c)=0$ :

$$
\frac{\ell_{i} \stackrel{g_{1}: c ? x}{\longleftrightarrow} \ell_{i}^{\prime} \wedge \eta \models g_{1} \wedge \eta \models g_{2} \wedge \ell_{j} \stackrel{g_{2}: c!v}{\longleftrightarrow} \ell_{j}^{\prime} \wedge i \neq j}{\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{j}, \ldots, \ell_{n}, \eta, \xi\right\rangle \xrightarrow{\tau}\left\langle\ell_{1}, \ldots, \ell_{i}^{\prime}, \ldots, \ell_{j}^{\prime}, \ldots, \ell_{n}, \eta^{\prime}, \xi\right\rangle}
$$

where $\eta^{\prime}=\eta[x:=v]$.

Figure 2.20: Rules for the transition relation of a channel system.

When the receiver $R$ is in location wait ( 0 ) and receives a message, it anticipates receiving a message with either control bit 0 (as it expects) or with control bit 1 , see left side of Figure 2.19. Symmetrically, in location wait(1) also the possibility is taken into account to receive the (unexpected) ack with control bit 0 . The following execution fragment indicates why this unexpected possibility is essential to take into consideration. In a nutshell, the execution fragment shows that it may happen that $R$ receives $\langle m, 0\rangle$, notifies this by means of sending an ack (with control bit 0 ), and switches to "mode 1 "-waiting to receive a message with control bit 1. In the meanwhile, however, sender $S$ has initiated a retransmission of $\langle m, 0\rangle$ (as it timed out). On receipt of this (unexpected) message, receiver $R$ should act accordingly and ignore the message. This is exactly what happens. Note that if this possibility would not have been taken into consideration in the program graph of $R$, the system would have come to a halt.

| sender $S$ | timer | receiver $R$ | channel $c$ | channel $d$ | event |
| :---: | :---: | :---: | :---: | :---: | :---: |
| snd_msg(0) | off | wait (0) | $\varnothing$ | $\varnothing$ |  |
| st_tmr (0) | off | wait(0) | $\langle m, 0\rangle$ | $\varnothing$ | message with bit 0 transmitted |
| wait(0) | on | wait (0) | $\langle m, 0\rangle$ | $\varnothing$ |  |
| snd_msg(0) | off | wait (0) | $\langle m, 0\rangle$ | $\varnothing$ | timeout |
| st_tmr (0) | off | wait (0) | $\langle m, 0\rangle\langle m, 0\rangle$ | $\varnothing$ | retransmission |
| st_tmr (0) | off | $p r \_m s g(0)$ | $\langle m, 0\rangle$ | $\varnothing$ | receiver reads first message |
| st_tmr (0) | off | snd_ack(0) | $\langle m, 0\rangle$ | $\varnothing$ |  |
| st_tmr (0) | off | wait(1) | $\langle m, 0\rangle$ | 0 | receiver changes into mode-1 |
| st_tmr (0) | off | $p r \_m s g(1)$ | $\varnothing$ | 0 | receiver reads retransmission |
| st_tmr (0) | off | wait(1) | $\varnothing$ | 0 | and ignores it |
| : |  |  |  |  |  |

We conclude this example by pointing out a possible simplification of the program graph of the sender $S$. Since the transmission of acks (over channel $d$ ) is reliable, it is unnecessary (but not wrong) for $S$ to verify the control bit of the ack in location $c h k_{-} a c k(\cdot)$. If $S$ is in location wait(0) and channel $d$ is not empty, then the (first) message in $d$ corresponds to the expected ack 0 , since $R$ acknowledges each message $m$ exactly once regardless of how many times $m$ is received. Therefore, the program graph of $S$ could be simplified such that by the action sequence $d ? x$; timer_off, it moves from location wait (0) to location gen_msg(1). Location chk_ack(0) may thus be omitted. By similar arguments, location $c h k_{-} a c k(1)$ may be omitted. If, however, channel $d$ would be unreliable (like channel $c$ ), these locations are necessary.

## Remark 2.35. Open Channel Systems

The rule for synchronous message passing is subject to the idea that there is a closed channel system that does not communicate with the environment. To model open channel systems, only the rule for handshaking has to be modified. If there is a channel $c$ with $\operatorname{cap}(c)=0$ over which the channel system is communicating with the environment, the rules

$$
\frac{\ell_{i} \stackrel{c!v}{\hookrightarrow} \ell_{i}^{\prime}}{\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{n}, \eta, \xi\right\rangle \stackrel{c!v}{\stackrel{~}{c}}\left\langle\ell_{1}, \ldots, \ell_{i}^{\prime}, \ldots, \ell_{n}, \eta, \xi\right\rangle}
$$

and

$$
\frac{\ell_{i} \stackrel{c ? x}{\hookrightarrow} \ell_{i}^{\prime} \wedge v \in \operatorname{dom}(c)}{\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{n}, \eta, \xi\right\rangle \stackrel{c ? x}{ }\left\langle\ell_{1}, \ldots, \ell_{i}^{\prime}, \ldots, \ell_{n}, \eta[x:=v], \xi\right\rangle}
$$

have to be used. The receipt of value $v$ for variable $x$ along channel $c$ is modeled by means of nondeterminism that is resolved by the environment. That is to say, the environment selects value $v \in \operatorname{dom}(c)$ in a purely nondeterministic way.

### 2.2.5 NanoPromela

The concepts that have been discussed in the previous sections (program graphs, parallel composition, channel systems) provide the mathematical basis for modeling reactive systems. However, for building automated tools for verifying reactive systems, one aims at simpler formalisms to specify the behavior of the system to be analyzed. On the one hand, such specification languages should be simple and easy to understand, such that nonexperts also are able to use them. On the other hand, they should be expressive enough to formalize the stepwise behavior of the processes and their interactions. Furthermore, they have to be equipped with a formal semantics which renders the intuitive meaning of the language commands in an unambiguous manner. In our case, the purpose of the formal semantics is to assign to each program of the specification language a transition system that can serve as a basis for the automated analysis, e.g., simulation or model checking against temporal logical specifications.

In this section, we present the core features of the language Promela, the input language for the prominent model checker SPIN by Holzmann [209]. Promela is short for "process metalanguage". Promela programs $\overline{\mathcal{P}}$ consist of a finite number of processes $\mathcal{P}_{1}, \ldots, \mathcal{P}_{n}$ to be executed concurrently. Promela supports communication over shared variables and message passing along either synchronous or buffered FIFO-channels. The formal semantics of a Promela-program can be provided by means of a channel system, which then can be unfolded into a transition system, as explained in Section 2.2.4. The stepwise behavior of the processes $\mathcal{P}_{i}$ is specified in Promela using a guarded command language
[130, 18] with several features of classical imperative programming languages (variable assignments, conditional and repetitive commands, sequential composition), communication actions where processes may send and receive messages from the channels, and atomic regions that avoid undesired interleavings. Guarded commands have already been used as labels for the edges of program graphs and channel systems. They consist of a condition (guard) and an action. Promela does not use action names, but specifies the effect of actions by statements of the guarded command language.

Syntax of nanoPromela We now explain the syntax and semantics of a fragment of Promela, called nanoPromela, which concentrates on the basic elements of Promela, but abstracts from details like variable declarations and skips several "advanced" concepts like abstract data types (arrays, lists, etc.) or dynamic process creation. A nanoPromela program consists of statements representing the stepwise behavior of the processes $\mathcal{P}_{1}, \ldots$, $\mathcal{P}_{n}$ together with a Boolean condition on the initial values of the program variables. We write nanoPromela programs as:

$$
\overline{\mathcal{P}}=\left[\mathcal{P}_{1}|\ldots| \mathcal{P}_{n}\right] .
$$

The main ingredients of the statements that formalize the stepwise behavior of the processes $\mathcal{P}_{i}$ are the atomic commands skip, variable assignments $x:=$ expr, communication actions $c ? x$ (reading a value for variable $x$ from channel $c$ ) and $c$ !expr (sending the current value of an expression over channel $c$ ), conditional commands (if-then-else) and (while)loops. Instead of the standard if-then-else constructs or whileloops, nanoPromela supports nondeterministic choices and allows specifying a finite number of guarded commands in conditional and repetitive commands.

Variables, Expressions and Boolean Expressions The variables in a nanoPromela program $\overline{\mathcal{P}}$ may be typed (integer, Boolean, char, real, etc.) and either global or local to some process of $\mathcal{P}_{i}$. Similarly, data domains have to be specified for the channels and they have to be declared to be synchronous or fifo-channels of a predefined capacity. We skip the details of variable and channel declarations, as they are irrelevant for the purposes of this chapter. As local variables can be renamed in case they occur in more than one process or as the name of a global variable, we may treat all variables as global variables. Thus, we assume that Var is a set of variables occurring in $\overline{\mathcal{P}}$ and that for any a variable name $x$ the domain (type) of $x$ is given as the set $\operatorname{dom}(x)$. Furthermore, in the declaration part of a Promela program, the type of a channel is specified. We simply write here dom(c) for the type (domain) of channel $c$ and $\operatorname{cap}(c)$ for its capacity. In addition, we assume that the variable declaration of program $\overline{\mathcal{P}}$ contains a Boolean expression that specifies the legal initial values for the variables $x \in \operatorname{Var}$.


Figure 2.21: Syntax of nanoPromela-statements.

The intuitive meaning of an assignment $x:=$ expr is obvious: variable $x$ is assigned the value of the expression expr given the current variable evaluation. The precise syntax of the expressions and Boolean expressions is not of importance here. We may assume that the expressions used in assignments for variable $x$ are built by constants in $\operatorname{dom}(x)$, variables $y$ of the same type as $x$ (or a subtype of $x$ ), and operators on $\operatorname{dom}(x)$, such as Boolean connectives $\wedge, \vee, \neg$, etc. for $\operatorname{dom}(x)=\{0,1\}$ and arithmetic operators,$+ *$, etc. for $\operatorname{dom}(x)=\mathbb{R} .^{3}$ The guards are Boolean expressions that impose conditions on the values of the variables, i.e., we treat the guards as elements of Cond(Var).

Statements The syntax of the statements that specify the behavior of the nanoPromelaprocesses is shown in Figure 2.21 on page 65. Here, $x$ is a variable in Var, expr an expression, and $c$ is a channel of arbitrary capacity. Type consistency of the variable $x$ and the expression expr is required in assignments $x:=$ expr. Similarly, for the messagepassing actions $c ? x$ and $c$ !expr we require that $\operatorname{dom}(c) \subseteq \operatorname{dom}(x)$ and that the type of expr corresponds to the domain of $c$. The $g_{i}$ 's in if-fi- and do-od-statements are guards. As mentioned above, we assume that $g_{i} \in \operatorname{Cond}(\operatorname{Var})$. The body assignments of an atomic region is a nonempty sequential composition of assignments, i.e., assignments has the form

$$
x_{1}:=\operatorname{expr}_{1} ; x_{2}:=\operatorname{expr}_{2} ; \ldots ; x_{m}:=\operatorname{expr}_{m}
$$

where $m \geqslant 1, x_{1}, \ldots, x_{m}$ are variables and expr $_{1}, \ldots$, expr $_{m}$ expressions such that the types of $x_{i}$ and expr ${ }_{i}$ are compatible.

Intuitive Meaning of the Commands Before presenting the formal semantics, let us give some informal explanations on the meaning of the commands. skip stands for a process that terminates in one step, without affecting the values of the variables or contents of the channels. The meaning of assignments is obvious. stmt ${ }_{1}$; stmt ${ }_{2}$ denotes sequential composition, i.e., $\operatorname{stmt}_{1}$ is executed first and after its termination stmt ${ }_{2}$ is executed. The concept of atomic regions is realized in nanoPromela by statements of the

[^4]form atomic\{stmt $\}$. The effect of atomic regions is that the execution of stmt is treated as an atomic step that cannot be interleaved with the activities of other processes. As a side effect atomic regions can also serve as a compactification technique that compresses the state space by ignoring the intermediate configurations that are passed during the executions of the commands inside an atomic region. The assumption that the body of an atomic region consists of a sequence of assignments will simplify the inference rules given below.

Statements build by if-fi or do-od are generalizations of standard if-then-else commands and whileloops. Let us first explain the intuitive meaning of conditional commands. The statement

$$
\text { if }:: g_{1} \Rightarrow \operatorname{stmt}_{1} \ldots:: g_{n} \Rightarrow \operatorname{stmt}_{n} \mathrm{fi}
$$

stands for a nondeterministic choice between the statements stmt ${ }_{i}$ for which the guard $g_{i}$ is satisfied in the current state, i.e., $g_{i}$ holds for the current valuation of the variables. We assume a test-and-set semantics where the evaluation of the guards, the choice between the enabled guarded commands and the execution of the first atomic step of the selected statement, are performed as an atomic unit that cannot be interleaved with the actions of concurrent processes. If none of the guards $g_{1}, \ldots, g_{n}$ is fulfilled in the current state, then the if-fi-command blocks. Blocking has to be seen in the context of the other processes that run in parallel and that might abolish the blocking by changing the values of shared variables such that one or more of the guards may eventually evaluate to true. For instance, the process given by the statement

$$
\text { if }:: y>0 \Rightarrow x:=42 \mathrm{fi}
$$

in a state where $y$ has the value 0 waits until another process assigns a nonzero value to $y$. Standard if-then-else commands, say "if $g$ then $\operatorname{stmt}_{1}$ else stmt ${ }_{2}$ fi", of imperative programming languages can be obtained by:

$$
\text { if }:: g \Rightarrow \operatorname{stmt}_{1}:: \neg g \Rightarrow \operatorname{stmt}_{2} \mathbf{f i},
$$

while statements "if $g$ then $\operatorname{stmt}_{1}$ fi" without an else option are modeled by:

$$
\text { if }:: g \Rightarrow \operatorname{stmt}_{1} \quad: \neg g \Rightarrow \text { skip fi. }
$$

In a similar way, the do-od-command generalizes whileloops. These specify the repetition of the body, as long as one of the guards is fulfilled. That is, statements of the form:

$$
\text { do }:: g_{1} \Rightarrow \operatorname{stmt}_{1} \ldots:: g_{n} \Rightarrow \operatorname{stmt}_{n} \text { od }
$$

stand for the iterative execution of the nondeterministic choice among the guarded commands $g_{i} \Rightarrow \operatorname{stmt}_{i}$ where guard $g_{i}$ holds in the current configuration. Unlike conditional commands, do-od-loops do not block in a state if all guards are violated. Instead, if
$g_{1}, \ldots, g_{n}$ do not hold in the current state, then the whileloop is aborted. In fact, a single guarded loop do $:: g \Rightarrow$ stmt od has the same effect as an ordinary whileloop "while $g$ do stmt od" with body stmt and termination condition $\neg g$. (As opposed to Promela, loops are not terminated by the special command "break".)

## Example 2.36. Peterson's Mutual Exclusion Algorithm

Peterson's mutual exclusion algorithm for two processes (see Example 2.25 on page 45) can be specified in nanoPromela as follows. We deal with two Boolean variables $b_{1}$ and $b_{2}$ and the variable $x$ with domain $\operatorname{dom}(x)=\{1,2\}$ and two Boolean variables crit ${ }_{1}$ and $c^{c} t_{2}$. The activities of the processes inside their noncritical sections are modeled by the action skip. For the critical section, we use the assignment crit ${ }_{i}:=$ true. Initially, we have $b_{1}=b_{2}=c r i t_{1}=c r i t_{2}=$ false, while $x$ is arbitrary. Then the nanoPromela-code of process $\mathcal{P}_{1}$ is given by the statement

```
do :: true => skip;
    atomic{\mp@subsup{b}{1}{}:= true; }x:=2}
    if :: (x=1)\vee\neg\mp@subsup{b}{2}{}=>crit}:=\mathrm{ true fi
    atomic{crit 
```

od

The statement for modeling the second process is similar. The infinite repetition of the three phases "noncritical section", "waiting phase" and "critical section" is modeled by the do-od-loop with the trivial guard true. The request action corresponds to the statement atomic $\left\{b_{1}:=\right.$ true; $\left.x:=2\right\}$ and the release action to the statement atomic\{crit $1:=$ false; $b_{1}:=$ false $\}$. The waiting phase where process $\mathcal{P}_{1}$ has to await until $x=1$ or $b_{2}=$ false is modeled by the if-fi-command.

The use of atomic regions is not necessary, but serves here as a compactification technique. As we mentioned in Example 2.25, the request action can also be split into the two assignments $b_{1}:=$ true and $x:=2$. As long as both assignments are inside an atomic region the order of the assignments $b_{1}:=$ true and $x:=2$ is irrelevant. If, however, we drop the atomic region, then we have to use the order $b_{1}:=$ true; $x:=2$, as otherwise the mutual exclusion property cannot be ensured. That is, the process

$$
\begin{aligned}
\text { do }:: \text { true } \Rightarrow & \text { skip; } \\
& x:=2 ; \\
& b_{1}:=\text { true; } \\
& \text { if }::(x=1) \vee \neg b_{2} \Rightarrow \text { crit }_{1}:=\text { true fi } \\
& \text { atomic }\left\{c r i t_{1}:=\text { false } ; b_{1}:=\text { false }\right\}
\end{aligned}
$$

for $\mathcal{P}_{1}$ together with the symmetric protocol for $\mathcal{P}_{2}$ constitutes a nanoPromela program where the mutual exclusion property "never crit ${ }_{1}=c r i t_{2}=$ true" cannot be guaranteed.

## Example 2.37. Vending Machine

In the above example there are no nondeterministic choices caused by a conditional or repetitive command. For an example where nondeterminism arises through simultaneously enabled guarded commands of a loop, consider the beverage vending machine of Example 2.14 (page 33). The following nanoPromela program describes its behavior:

```
do :: true \(\Rightarrow\)
        skip;
        if \(\quad:: \quad n s o d a>0 \Rightarrow n s o d a:=n s o d a-1\)
        \(::\) nbeer \(>0 \Rightarrow\) nbeer \(:=\) nbeer -1
                        \(:: \quad\) nsoda \(=\) nbeer \(=0 \Rightarrow\) skip
                fi
    \(::\) true \(\Rightarrow\) atomic\{nbeer \(:=\max ; n s o d a:=\max \}\)
od
```

In the starting location, there are two options that are both enabled. The first is the insertion of a coin by the user, modeled by the command skip. The first two options of the if-fi-command represent the cases where the user selects soda or beer, provided some bottles of soda and beer, respectively, are left. The third guarded command in the if-fi substatement applies to the case where neither soda nor beer is available anymore and the machine automatically returns to the initial state. The second alternative that is enabled in the starting location is the refill action whose effect is specified by the atomic region where the variables nbeer and nsoda are reset to max.

Semantics The operational semantics of a nanoPromela-statement with variables and channels from (Var, Chan) is given by a program graph over (Var, Chan). The program graphs $P G_{1}, \ldots, P G_{n}$ for the processes $\mathcal{P}_{1}, \ldots, \mathcal{P}_{n}$ of a nanoPromela program $\overline{\mathcal{P}}=$ $\left[\mathcal{P}_{1}|\ldots| \mathcal{P}_{n}\right]$ constitute a channel system over (Var, Chan). The transition system semantics for channel systems (Definition 2.31 on page 55) then yields a transition system $\operatorname{TS}(\overline{\mathcal{P}})$ that formalizes the stepwise behavior of $\overline{\mathcal{P}}$.

The program graph associated with a nanoPromela-statement stmt formalizes the control flow when executing stmt. That is, the substatements play the role of the locations. For modeling termination, a special location exit is used. Roughly speaking, any guarded command $g \Rightarrow$ stmt corresponds to an edge with the label $g: \alpha$ where $\alpha$ stands for the first action of stmt. For example, for the statement

$$
\begin{aligned}
\text { cond_cmd }=\text { if } & :: x>1 \Rightarrow y:=x+y \\
& :: \text { true } \Rightarrow x:=0 ; y:=x
\end{aligned}
$$

from cond_cmd - viewed as a location of a program graph - there are two edges: one with the guard $x>1$ and action $y:=x+y$ leading to exit, and one with the guard true and action $x:=0$ yielding the location for the statement $y:=x$. Since $y:=x$ is deterministic there is a single edge with guard true, action $y:=x$ leading to location exit.

As another example, consider the statement

$$
\begin{aligned}
\text { loop }= & \text { do } \begin{array}{ll}
: & x>1 \Rightarrow y:=x+y \\
& :: y<x \Rightarrow x:=0 ; y:=x \\
& \text { od }
\end{array} . \quad . \quad \text {. }
\end{aligned}
$$

Here, the repetition semantics of the do-od-loop is modeled by returning the control to stmt whenever the body of the selected alternative has been executed. Thus, from location loop there are three outgoing edges, see Figure 2.22 on page 69. One is labeled with the guard $x>1$ and action $y:=x+y$ and leads back to location loop. The second edge has the guard $y<x$ and action $x:=0$ and leads to the statement $y:=x$; loop. The third edge covers the case where the loop terminates. It has the guard $\neg(x>1) \wedge \neg(y<x)$ and an action without any effect on the variables and leads to location exit.


Figure 2.22: Program graph for a loop

The goal is now to formalize the ideas sketched above. We start with a formal definition of substatements of stmt. Intuitively, these are the potential locations of intermediate states during the execution of stmt.

## Notation 2.38. Substatement

The set of substatements of a nanoPromela-statement stmt is recursively defined. For statements stmt $\in\{$ skip, $x:=\operatorname{expr}, c ? x, c$ !expr $\}$ the set of substatements is sub(stmt) $=$ \{stmt, exit\}. For sequential composition let

$$
\begin{aligned}
& \operatorname{sub}\left(\operatorname{stmt}_{1} ; \operatorname{stmt}_{2}\right)= \\
& \quad\left\{\operatorname{stmt}^{\prime} ; \operatorname{stmt}_{2} \mid \operatorname{stmt}^{\prime} \in \operatorname{sub}\left(\operatorname{stmt}_{1}\right) \backslash\{\text { exit }\}\right\} \cup \operatorname{sub}\left(\operatorname{stmt}_{2}\right) .
\end{aligned}
$$

For conditional commands, the set of substatements is defined as the set consisting of the if-fi-statement itself and substatements of its guarded commands. That is, for cond_cmd
being if $:: g_{1} \Rightarrow \operatorname{stmt}_{1} \ldots:: g_{n} \Rightarrow \operatorname{stmt}_{n}$ fi we have

$$
\operatorname{sub}(\text { cond_cmd })=\{\text { cond_cmd }\} \cup \underset{1 \leqslant i \leqslant n}{\bigcup} \operatorname{sub}\left(\text { stmt }_{i}\right) .
$$

The substatements of a loop loop given by do $:: g_{1} \Rightarrow \operatorname{stmt}_{1} \ldots:: g_{n} \Rightarrow \operatorname{stmt}_{n}$ od are defined similarly, but taking into account that control moves back to loop when guarded commands terminate. That is:

$$
\begin{aligned}
& \operatorname{sub}(\text { loop })= \\
& \quad\{\text { loop, exit }\} \cup \underset{1 \leqslant i \leqslant n}{\bigcup}\left\{\text { stmt }^{\prime} ; \text { loop } \mid \operatorname{stmt}^{\prime} \in \operatorname{sub}\left(\text { stmt }_{i}\right) \backslash\{\text { exit }\}\right\} .
\end{aligned}
$$

For atomic regions let $\operatorname{sub}($ atomic $\{\operatorname{stm} t\})=\{$ atomic $\{\operatorname{stmt}\}$, exit $\}$.

The definition of sub(loop) relies on the observation that the effect of a loop with a single guarded command, say "do $:: g \Rightarrow$ stmt od", corresponds to the effect of

$$
\text { if } g \text { then stmt ; do }:: g \Rightarrow \text { stmt od else skip fi }
$$

An analogous characterization applies to loops with two or more guarded commands. Thus, the definition of the substatements of loop relies on combining the definitions of the sets of substatements for sequential composition and conditional commands.

The formal semantics of nanoPromela program $\overline{\mathcal{P}}=\left[\mathcal{P}_{1}|\ldots| \mathcal{P}_{n}\right]$ where the behavior of the process $\mathcal{P}_{i}$ is specified by a nanoPromela statement is a channel system $\left[P G_{1}|\ldots| P G_{n}\right]$ over (Var, Chan) where Var is the set of variables and Chan the set of channels that are declared in $\overline{\mathcal{P}}$. As mentioned before, a formal syntax for the variable and channel declarations will not be provided, and global and local variables will not be distinguished. We assume that the set Var of typed variables and the set Chan of channels (together with a classification of the channels into synchronous and FIFO-channels of some capacity $\operatorname{cap}(\cdot))$ are given. Hence, local variable and channel declarations for the processes $\mathcal{P}_{i}$ are not considered. It is assumed that they are given by a nanoPromela-statement over some fixed tuple (Var, Chan).

We now provide inference rules for the nanoPromela constructs. The inference rules for the atomic commands (skip, assignment, communication actions) and sequential composition, conditional and repetitive commands give rise to the edges of a "large" program graph where the set of locations agrees with the set of nanoPromela-statements. Thus, the edges have the form

$$
\text { stmt } \stackrel{g: \alpha}{\longrightarrow} \text { stmt }^{\prime} \text { or } \text { stmt } \xrightarrow{g: c o m m} \text { stmt }^{\prime}
$$

where stmt is a nanoPromela statement, stmt ${ }^{\prime}$ a substatement of stmt, and $g$ a guard, $\alpha$ an action, and comm a communication action $c$ ? $x$ or $c$ !expr. The subgraph consisting of
the substatements of $\mathcal{P}_{i}$ then yields the program graph $P G_{i}$ of process $\mathcal{P}_{i}$ as a component of the program $\overline{\mathcal{P}}$.

The semantics of the atomic statement skip is given by a single axiom formalizing that the execution of skip terminates in one step without affecting the variables
skip $\xlongequal{\text { true: } i d}$ exit
where id denotes an action that does not change the values of the variables, i.e., Effect (id, $\eta$ ) $=\eta$ for all variable evaluations $\eta$. Similarly, the execution of a statement consisting of an assignment $x:=$ expr has the trivial guard and terminates in one step:

$$
x:=\text { expr } \quad \text { true }: \operatorname{assign}(x, \text { expr }) \quad \text { exit }
$$

where $\operatorname{assign}(x$, expr $)$ denotes the action that changes the value of $x$ according to the assignment $x:=$ expr and does not affect the other variables, i.e., if $\eta \in E v a l(\operatorname{Var})$ and $y \in \operatorname{Var}$ then Effect $(\operatorname{assign}(x, \operatorname{expr}), \eta)(y)=\eta(y)$ if $y \neq x$ and $\operatorname{Effect}(\operatorname{assign}(x, \operatorname{expr}), \eta)(x)$ is the value of expr when evaluated over $\eta$. For the communication actions $c$ !expr and $c ? x$ the following axioms apply:

$$
\overline{c ? x \stackrel{c ? x}{\longleftrightarrow} \text { exit }} \overline{c!\text { expr } \stackrel{c}{c!e x p r}_{\longrightarrow}^{c} \text { exit }}
$$

The effect of an atomic region atomic $\left\{x_{1}:=\operatorname{expr}_{1} ; \ldots ; x_{m}:=\operatorname{expr}_{m}\right\}$ is the cumulative effect of the assignments $x_{i}:=\operatorname{expr}_{i}$. It can be defined by the rule:

$$
\text { atomic }\left\{x_{1}:=\operatorname{expr}_{1} ; \ldots ; x_{m}:=\operatorname{expr}_{m}\right\} \stackrel{\text { true }: \alpha_{m}}{\leftrightarrows} \text { exit }
$$

where $\alpha_{0}=i d, \alpha_{i}=\operatorname{Effect}\left(\operatorname{assign}\left(x_{i}, \operatorname{expr}_{i}\right), \operatorname{Effect}\left(\alpha_{i-1}, \eta\right)\right)$ for $1 \leqslant i \leqslant m$.
Sequential composition $\operatorname{stmt}_{1} ; \mathrm{stmt}_{2}$ is defined by two rules that distinguish whether or not $\operatorname{stmt}_{1}$ terminates in one step. If the first step of $\operatorname{stmt}_{1}$ leads to a location (statement) different from exit, then the following rule applies:

$$
\frac{\operatorname{stmt}_{1} \stackrel{g: \alpha}{\longleftrightarrow} \operatorname{stmt}_{1}^{\prime} \neq \text { exit }}{\operatorname{stmt}_{1} ; \mathrm{stmt}_{2} \stackrel{g: \alpha}{\longleftrightarrow} \mathrm{stmt}_{1}^{\prime} ; \mathrm{stmt}_{2}}
$$

If the computation of $\operatorname{stmt}_{1}$ terminates in one step by executing action $\alpha$, then control of $\mathrm{stmt}_{1} ; \mathrm{stmt}_{2}$ moves to $\mathrm{stmt}_{2}$ after executing $\alpha$ :

$$
\frac{\operatorname{stmt}_{1} \stackrel{g: \alpha}{\stackrel{2}{\longrightarrow}} \text { exit }}{\text { stmt }_{1} ; \mathrm{stmt}_{2} \stackrel{g: \alpha}{\longleftrightarrow} \text { stmt }_{2}}
$$

The effect of a conditional command cond_cmd $=$ if $:: g_{1} \Rightarrow \operatorname{stmt}_{1} \ldots:: g_{n} \Rightarrow \operatorname{stmt}_{n} \mathrm{f}$ is formalized by means of the following rule:

$$
\frac{\text { stmt }_{i} \stackrel{h: \alpha}{\stackrel{h}{l}} \mathrm{stmt}_{i}^{\prime}}{\text { cond_cmd }<^{g_{i} \wedge h: \alpha} \longrightarrow \mathrm{stmt}_{i}^{\prime}}
$$

This rule relies on the test-and-set semantics where choosing one of the enabled guarded commands and performing its first action are treated as an atomic step. The blocking of cond_cmd when none of its guards is enabled needs no special treatment. The reason is that cond_cmd has no other edges than the ones specified by the rule above. Thus, in a global state $s=\left\langle\ell_{1}, \ldots, \ell_{n}, \eta, \xi\right\rangle$ where the location $\ell_{i}$ of the $i$ th process is $\ell_{i}=$ cond_cmd and all guards $g_{1}, \ldots, g_{n}$ evaluate to false, then there is no action of the $i$ th process that is enabled in $s$. However, actions of other processes might be enabled. Thus, the $i$ th process has to wait until the other processes modify the variables appearing in $g_{1}, \ldots, g_{n}$ such that one or more of the guarded commands $g_{i} \Rightarrow \operatorname{stmt}_{i}$ become enabled.

For loops, say loop $=$ do $:: g_{1} \Rightarrow \operatorname{stmt}_{1} \ldots:: g_{n} \Rightarrow \operatorname{stmt}_{n}$ od, we deal with three rules. The first two rules are similar to the rule for conditional commands, but taking into account that control moves back to loop after the execution of the selected guarded command has been completed. This corresponds to the following rules:


If none of the guards $g_{1}, \ldots, g_{n}$ holds in the current state then the do-od-loop will be aborted. This is formalized by the following axiom:

$$
\text { loop } \xrightarrow{\neg g_{1} \wedge \ldots \wedge \neg g_{n}} \text { exit }
$$

Remark 2.39. Test-and-Set Semantics vs. Two-Step Semantics
The rules for if-fi- and do-od-statements formalize the so-called test-and-set semantics of guarded commands. This means that evaluating guard $g_{i}$ and performing the first step of the selected enabled guarded command $g_{i} \Rightarrow \operatorname{stmt}_{i}$ are performed atomically. In contrast, SPIN's interpretation of Promela relies on a two-step-semantics where the selection of an enabled guarded command and the execution of its first action are split into two steps. The rule for a conditional command is formalized by the axiom

$$
\text { if }:: g_{1} \Rightarrow \operatorname{stmt}_{1} \ldots:: g_{n} \Rightarrow \operatorname{stmt}_{n} \mathrm{fi} \stackrel{g_{i}: \mathrm{id}}{\longrightarrow} \operatorname{stmt}_{i}
$$

where id is an action symbol for an action that does not affect the variables. Similarly, the first two rules for loops have to be replaced for the two-step semantics by the following rule:

$$
\overline{\text { loop } \stackrel{g_{i}: \text { id }}{\longrightarrow} \text { stmt }_{i} \text {; loop }}
$$

The rule for terminating the loop remains unchanged.
As long as we consider the statements in isolation, the test-and-set semantics and the two-step semantics are equal. However, when running several processes in parallel, the interleaving might cause undesired side effects. For example, consider the semaphore-based solution of the mutual exclusion problem, modeled by a nanoPromela program where the behavior of $\mathcal{P}_{i}$ is given by the following nanoPromela-statement:

$$
\begin{aligned}
& \text { do :: true } \Rightarrow \text { skip; } \\
& \text { if }:: y>0 \Rightarrow y:=y-1 \text {; } \\
& \text { crit }_{i}:=\text { true } \\
& \text { fi; } \\
& y:=y+1 \\
& \text { od }
\end{aligned}
$$

The initial value of the semaphore $y$ is zero. Under the two-step semantics the mutual exclusion property is not guaranteed as it allows the processes to verify that the guard $y>0$ of the if-fi-statement holds, without decreasing the value of $y$, and moving control to the assignment $y:=y-1$. But from there the processes can enter their critical sections. However, the protocol works correctly for the test-and-set semantics since then checking $y>0$ and decreasing $y$ is an atomic step that cannot be interleaved by the actions of the other process.

## Remark 2.40. Generalized Guards

So far we required that the guards in conditional or repetitive commands consist of Boolean conditions on the program variables. However, it is also often useful to ask for interaction facilities in the guards, e.g., to specify that a process has to wait for a certain input along a FIFO-channel by means of a conditional command if :: $c$ ? $x \Rightarrow$ stmt fi. The intuitive meaning of the above statement is that the process has to wait until the buffer for $c$ is nonempty. If so, then it first performs the action $c ? x$ and then executes stmt. The use of communication actions in the guards leads to a more general class of program graphs with guards consisting of Boolean conditions on the variables or communication actions. For the case of an asynchronous channel the rules in Figure 2.20 on page 61 then have to be extended by:

$$
\frac{\ell_{i} \xlongequal{c ? x: \alpha} \not \ell_{i}^{\prime} \wedge \operatorname{len}(\xi(c))=k>0 \wedge \xi(c)=v_{1} v_{2} \ldots v_{k}}{\left\langle\ldots, \ell_{i}, \ldots, \eta, \xi\right\rangle \xrightarrow{\tau}\left\langle\ldots, \ell_{i}^{\prime}, \ldots, \eta^{\prime}, \xi^{\prime}\right\rangle}
$$

where $\eta^{\prime}=\operatorname{Effect}\left(\alpha, \eta\left[x:=v_{1}\right]\right)$ and $\xi\left[c:=v_{2} \ldots v_{k}\right]$, and
where $\eta^{\prime}=\operatorname{Effect}(\alpha, \eta)$ and $\xi\left[c:=v_{1} \ldots v_{k} v\right]$.
Another convenient concept is the special guard else which specifies configurations where no other guarded command can be taken. The intuitive semantics of:

$$
\begin{array}{rllll}
\text { if } & : & g_{1} & \Rightarrow & \operatorname{stmt}_{1} \\
& & & \vdots & \\
& :: & g_{n} & \Rightarrow & \operatorname{stmt}_{n} \\
& : & \text { else } & \Rightarrow & \operatorname{stmt}^{\prime} \\
& \text { fi } & & &
\end{array}
$$

is that the else option is enabled if none of the guards $g_{1}, \ldots g_{n}$ evaluates to true. In this case, the execution evolves to a state in which the statement stmt ${ }^{\prime}$ is to be executed. Here, the $g_{i}$ 's can be Boolean expressions on the variables or communication guards. For example,

$$
\text { if }:: d ? x \Rightarrow \text { stmt }:: \text { else } \Rightarrow x:=x+1 \mathrm{fi}
$$

increases $x$ unless a message is obtained from channel $d$. The else option used in loops leads to nonterminating behaviors.

## Remark 2.41. Atomic Regions

The axiom for atomic regions yields that if $s=\left\langle\ell_{1} \ldots, \ell_{n}, \eta, \xi\right\rangle$ is a state in the transition system for the channel system associated with $\overline{\mathcal{P}}=\left[\mathcal{P}_{1}|\ldots| \mathcal{P}_{n}\right]$ and $\ell_{i}=\operatorname{atomic}\left\{x_{1}:=\right.$ $\left.\operatorname{expr}_{1} ; \ldots ; x_{m}:=\operatorname{expr}_{m}\right\} ; \ldots$ then in state $s$ the $i$ th process can perform the sequence of assignments $x_{1}:=\operatorname{expr}_{1} ; \ldots ; x_{m}:=\operatorname{expr}_{m}$ in a single transition. With this semantics we abstract from the intermediate states that are passed when having performed the first $i$ assignments $(1 \leqslant i<m)$ and avoid that other processes can interleave their activities with these assignments.

This concept can be generalized for atomic regions atomic\{stmt\} where the body stmt is an arbitrary statement. The idea is that any terminating execution of stmt is collapsed into a single transition, leading from a state with location stmt to a state with location exit. For this more general approach, the semantic rule for atomic regions operates on execution sequences in the transition system rather than just edges in the program graph. This is not problematic as one could provide the meaning of the statements on the level of transition systems rather than program graph level. However, the semantics is less
obvious for, e.g., infinite executions inside atomic regions, synchronous communication actions inside atomic regions and blocking conditional commands inside atomic regions. One possibility is to insert transitions to a special deadlock state. Another possibility is to work with a semantics that represents also the intermediate steps of an atomic region (but avoids interleaving) and to abort atomic regions as soon as a synchronous communication is required or blocking configurations are reached.

As we mentioned in the beginning of the section, Promela provides many more features than nanoPromela, such as atomic regions with more complex statements than sequences of assignments, arrays and other data types, and dynamic process creation. These concepts will not be explained here and we refer to the literature on the model checker SPIN (see, e.g., [209]).

### 2.2.6 Synchronous Parallelism

When representing asynchronous systems by transition systems, there are no assumptions concerning the relative velocities of the processors on which the components are executed. The residence time of the system in a state and the execution time of the actions are completely ignored. For instance, in the example of the two independent traffic lights (see Example 2.17), no assumption has been made concerning the amount of time a light should stay red or green. The only assumption is that both time periods are finite. The concurrent execution of components is time-abstract.

This is opposed to synchronous systems where components evolve in a lock step fashion. This is a typical computation mechanism in synchronous hardware circuits, for example, where the different components (like adders, inverters, and multiplexers) are connected to a central clock and all perform a (possibly idle) step on each clock pulse. As clock pulses occur periodically with a fixed delay, these pulses may be considered in a discrete manner, and transition systems can be adequately used to describe these synchronous systems. Synchronous composition of two transition systems is defined as follows.

## Definition 2.42. Synchronous Product

Let $T S_{i}=\left(S_{i}, A c t, \rightarrow_{i}, I_{i}, A P_{i}, L_{i}\right), i=1,2$, be transition systems with the same set of actions Act. Further, let

$$
\text { Act } \times \text { Act } \rightarrow \text { Act }, \quad(\alpha, \beta) \rightarrow \alpha * \beta
$$


$T S_{1} \otimes T S_{2}:$


Figure 2.23: Synchronous composition of two hardware circuits.
be a mapping ${ }^{4}$ that assigns to each pair of actions $\alpha, \beta$, the action name $\alpha * \beta$. The synchronous product $T S_{1} \otimes T S_{2}$ is given by:

$$
T S_{1} \otimes T S_{2}=\left(S_{1} \times S_{2}, A c t, \rightarrow, I_{1} \times I_{2}, A P_{1} \cup A P_{2}, L\right)
$$

where the transition relation is defined by the following rule

$$
\frac{s_{1} \xrightarrow{\alpha}_{1} s_{1}^{\prime} \wedge s_{2} \xrightarrow{\beta} 2 s_{2}^{\prime}}{\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha * \beta}\left\langle s_{1}^{\prime}, s_{2}^{\prime}\right\rangle}
$$

and the labeling function is defined by: $L\left(\left\langle s_{1}, s_{2}\right\rangle\right)=L_{1}\left(s_{1}\right) \cup L_{2}\left(s_{2}\right)$.

Action $\alpha * \beta$ denotes the synchronous execution of actions $\alpha$ and $\beta$. Note that compared to the parallel operator $\|$ where components perform actions in common synchronously, and other action autonomously (i.e., asynchronously), in $T S_{1} \otimes T S_{2}$, both transition systems have to perform all steps synchronously. There are no autonomous transitions of either $T S_{1}$ or $T S_{2}$.

[^5]
## Example 2.43. Synchronous Product of Two Circuits

Let $C_{1}$ be a circuit without input variables and with output variable $y$ and register $r$. The control functions for output and register transitions are

$$
\lambda_{y}=r_{1}, \quad \delta_{r_{1}}=\neg r_{1} .
$$

Circuit $C_{2}$ has input variable $x^{\prime}$, output variable $y^{\prime}$, and register variable $r_{2}$ with the control functions

$$
\lambda_{y^{\prime}}=\delta_{r_{2}}=x^{\prime} \vee r_{2} .
$$

The transition system $T S_{C_{1}} \otimes T S_{C_{2}}$ is depicted in Figure 2.23 on page 76. Since action names are omitted for transition systems of circuits, action labels for $T S_{C_{1}} \otimes T S_{C_{2}}$ are irrelevant. $T S_{C_{1}} \otimes T S_{C_{2}}$ is thus the transition system of the circuit with input variable $x^{\prime}$, output variables $y$ and $y^{\prime}$, and registers $r_{1}$ and $r_{2}$, whose control functions are $\lambda_{y}, \lambda_{y^{\prime}}, \delta_{r_{1}}$, and $\delta_{r_{2}}$.

### 2.3 The State-Space Explosion Problem

The previous two sections have shown that various kinds of systems can be modeled using transition systems. This applies to program graphs representing data-dependent systems, and hardware circuits. Different communication mechanisms can be modeled in terms of appropriate operators on transition systems. This section considers the cardinality of the resulting transition systems, i.e., the number of states in these models. Verification techniques are based on systematically analyzing these transition systems. The runtimes of such verification algorithms are mainly determined by the number of states of the transition system to be analyzed. For many practical systems, the state space may be extremely large, and this is a major limitation for state-space search algorithms such as model checking. Chapter 8, Section 6.7, and Chapter 7 introduce a number of techniques to combat this problem.

Program Graph Representation Transition systems generated by means of "unfolding" a program graph may be extremely large, and in some cases - e.g., if there are infinitely many program locations or variables with infinite domains - even have infinitely many states. Consider a program graph over the set of variables Var with $x \in \operatorname{Var}$. Recall that states of the unfolded transition system are of the form $\langle\ell, \eta\rangle$ with location $\ell$ and variable evaluation $\eta$. In case all variables in Var have a finite domain, like bits, or bounded integers, and the number of locations is finite, the number of states in the transition system is

$$
|L o c| \cdot \prod_{x \in \operatorname{Var}}|\operatorname{dom}(x)| .
$$

The number of states thus grows exponentially in the number of variables in the program graph: for $N$ variables with a domain of $k$ possible values, the number of states grows up to $k^{N}$. This exponential growth is also known as the state-space explosion problem.

It is important to realize that for even simple program graphs with just a small number of variables, this bound may already be rather excessive. For instance, a program graph with ten locations, three Boolean variables and five bounded integers (with domain in $\{0, \ldots, 9\}$ ) has $10 \cdot 2^{3} \cdot 10^{5}=8,000,000$ states. If a single bit array of 50 bits is added to this program graph, for example, this bound grows even to $800,000 \cdot 2^{50}$ ! This observation clearly shows why the verification of data-intensive systems (with many variables or complex domains) is extremely hard. Even if there are only a few variables in a program, the state space that must be analyzed may be very large.

If $\operatorname{dom}(x)$ is infinite for some $x \in \operatorname{Var}$, as for reals or integers, the underlying transition system has infinitely many states as there are infinitely many values for $x$. Such program graphs usually yield undecidable verification problems. This is not to say that the verification of all transition systems with an infinite state space is undecidable, however.

It should be remarked that not only the state space of a transition system, but also the number of atomic propositions to represent program graphs (see Definition 2.15, page 34) may in principle be extremely large. Besides, any location, any condition on the variables in the program graph is allowed as an atomic proposition. However, in practice, only a small fragment of the possible atomic propositions is needed. An explicit representation of the labeling function is mostly not necessary, as the truth-values of the atomic formulae are typically derived from the state information. For these reasons, the number of atomic propositions plays only a secondary role.

For sequential hardware circuits (see page 26), states in the transition system are determined by the possible evaluations of the input variables and the registers. The size of the transition system thus grows exponentially in the number of registers and input variables. For $N$ input variables and $K$ registers, the total state space consists of $2^{N+K}$ states.

Parallelism In all variants of parallel operators for transition systems and program graphs, the state space of the complete system is built as the Cartesian product of the local state spaces $S_{i}$ of the components. For example, for state space $S$ of transition system

$$
T S=T S_{1}\| \| \ldots \| T S_{n}
$$

we have $S=S_{1} \times \ldots \times S_{n}$ where $S_{i}$ denotes the state space of transition system $T S_{i}$. The state space of the parallel composition of a system with $n$ states and a system with
$k$ states yields $n \cdot k$ states. The total state space $S$ is thus

$$
\left|S_{1}\right| \cdot \ldots \cdot\left|S_{n}\right|
$$

The number of states in $S$ is growing (at most) exponentially in the number of components: the parallel composition of $N$ components of size $k$ each yields $k^{N}$ states. Even for small parallel systems this may easily run out of control.

Additionally, the variables (and their domains) represented in the transition system essentially influence the size of the state space. If one of the domains is infinite, then the state space is infinitely large. If the domains are finite, then the size of the state space grows exponentially in the number of variables (as we have seen before for program graphs).

The "exponential blowup" in the number of parallel components and the number of variables explains the enormous size of the state space of practically relevant systems. This observation is known under the heading state explosion and is another evidence for the fact that verification problems are particularly space-critical.

Channel Systems For the size of transition systems of channel systems, similar observations can be made as for the representation of program graphs. An important additional component for these systems is the size of the channels, i.e., their capacity. Clearly, if one of these channels has an infinite capacity, this may yield infinitely many states in the transition system. If all channels have finite capacity, however, the number of states is bound in the following way. Let $C S=\left[P G_{1}|\ldots| P G_{n}\right]$ be a channel system over $\operatorname{Var}=\operatorname{Var}_{1} \cup \ldots \cup \operatorname{Var}_{n}$ and channels Chan. The state space of CS is of cardinality

$$
\prod_{i=1}^{n}\left|P G_{i}\right| \cdot \prod_{c \in \text { Chan }}|\operatorname{dom}(c)|^{c p(c)},
$$

which can be rewritten as

$$
\prod_{i=1}^{n}\left(\left|\operatorname{Loc}_{i}\right| \cdot \prod_{x \in \operatorname{Var}_{i}}|\operatorname{dom}(x)|\right) \cdot \prod_{c \in \text { Chan }}|\operatorname{dom}(c)|^{c p(c)} .
$$

For $L$ locations per component, $K$ bit channels of capacity $k$ each, and $M$ variables $x$ with $|\operatorname{dom}(x)| \leqslant m$ totally, the total number of states in the transition system is $L^{n} \cdot m^{M} \cdot 2^{K \cdot k}$. This is typically enormous.

## Example 2.44. State-Space Size of the Alternating Bit Protocol

Consider a variant of the alternating bit protocol (see Example 2.32, page 57) where the
channels $c$ and $d$ have a fixed capacity, 10 say. Recall that along channel $d$, control bits are sent, and along channel $c$, data together with a control bit. Let us assume that data items are also simply bits. The timer has two locations, the sender eight, and the receiver six. As there are no further variables, we obtain that the total number of states is $2 \cdot 8 \cdot 6 \cdot 4^{10} \cdot 2^{10}$, which equals $3 \cdot 2^{35}$. This is around $10^{11}$ states.

### 2.4 Summary

- Transition systems are a fundamental model for modeling software and hardware systems.
- An execution of a transition system is an alternating sequence of states and actions that starts in an initial state and that cannot be prolonged.
- Interleaving amounts to represent the evolvement of "simultaneous" activities of independent concurrent processes by the nondeterministic choice between these activities.
- In case of shared variable communication, parallel composition on the level of transition systems does not faithfully reflect the system's behavior. Instead, composition on program graphs has to be considered.
- Concurrent processes that communicate via handshaking on the set $H$ of actions execute actions outside $H$ autonomously whereas they execute actions in $H$ synchronously.
- In channel systems, concurrent processes communicate via FIFO-buffers (i.e., channels). Handshaking communication is obtained when channels have capacity 0 . For channels with a positive capacity, communication takes place asynchronous-ly-sending and receiving a message takes place at different moments.
- The size of transition system representations grows exponentially in various components, such as the number of variables in a program graph or the number of components in a concurrent system. This is known as the state-space explosion problem.


### 2.5 Bibliographic Notes

Transition systems. Keller was one of the first researchers that explicitly used transition systems [236] for the verification of concurrent programs. Transition systems are used
as semantical models for a broad range of high-level formalisms for concurrent systems, such as process algebras [45, 57, 203, 298, 299], Petri nets [333], and statecharts [189]. The same is true for hardware synthesis and analysis, in which variants of finite-state automata (Mealy and Moore automata) play a central role; these variants can also be described by transition systems [246]. Program graphs and their unfolding into transition systems have been used extensively by Manna and Pnueli in their monograph(s) on temporal logic verification [283].

Synchronization paradigms. Shared variable "communication" dates back to the midsixties and is due to Dijkstra [126]. He also coined the term interleaving in 1971 [128]. Handshaking communication has been the main interaction paradigm in process algebras such as ACP [45], CCS [298, 299], CSP [202, 203], and LOTOS [57]. For a detailed account of process algebra we refer to [46]. The principle of synchronized parallelism has been advocated in Milner's synchronous variant of CCS, SCCS [297], and is used by Arnold to model the interaction between finite transition systems [19]. Synchronous parallelism is also at the heart of Lustre [183], a declarative programming language for reactive systems, and is used in many other hardware-oriented languages.

The interaction between concurrent processes by means of buffers (or channels) has first been considered by Dijkstra [129]. This paradigm has been adopted by specification languages for communication protocols, such as SDL (Specification and Description Language [37]) which is standardized by the ITU. The idea of guarded command languages goes back to Dijkstra [130]. The combination of guarded command languages in combination with channel-based communication is also used in Promela [205], the input language of the model checker SPIN [208]. The recent book by Holzmann [209] gives a detailed account of Promela and SPIN. Structured operational semantics has been introduced by Plotkin [334, 336] in 1981. Its origins are described in [335]. Atomic regions have been first discussed by Lipton [276], Lamport [257] and Owicki [317]. Further details on semantic rules for specification languages for reactive systems can be found, e.g., in [15, 18].

The examples. Most examples that have been provided in this chapter are rather classical. The problem of mutual exclusion was first proposed in 1962 by Dekker together with an algorithm that guarantees two-process mutual exclusion. Dijkstra's solution [126] was the first solution to mutual exclusion for an arbitrary number of processes. He also introduced the concept of semaphores [127] and their use for solving the mutual exclusion problem. A simpler and more elegant solution has been proposed by Lamport in 1977 [256]. This was followed up by Peterson's mutual exclusion protocol [332] in 1981. This algorithm is famous by now due to its beauty and simplicity. For other mutual exclusion algorithms, see e.g. [283, 280]. The alternating bit protocol stems from 1969 and is one of the first flow control protocols [34]. Holzmann gives a historical account of this and related protocols in his first book [205].

### 2.6 Exercises

Exercise 2.1. Consider the following two sequential hardware circuits:


Questions:
(a) Give the transition systems of both hardware circuits.
(b) Determine the reachable part of the transition system of the synchronous product of these transition systems. Assume that the initial values of the registers are $r_{1}=0$ and $r_{2}=1$.

Exercise 2.2. We are given three (primitive) processes $P_{1}, P_{2}$, and $P_{3}$ with shared integer variable $x$. The program of process $P_{i}$ is as follows:

```
Algorithm 1 Process \(P_{i}\)
    for \(k_{i}=1, \ldots, 10\) do
        \(\operatorname{LOAD}(x)\);
        \(\operatorname{INC}(x)\);
        STORE \((x)\);
    od
```

That is, $P_{i}$ executes ten times the assignment $x:=x+1$. The assignment $x:=x+1$ is realized using the three actions $\operatorname{LOAD}(x), \operatorname{INC}(x)$ and $\operatorname{STORE}(x)$. Consider now the parallel program:

```
Algorithm 2 Parallel program \(P\)
    \(x:=0\);
    \(P_{1}\left\|P_{2}\right\| P_{3}\)
```

Question: Does $P$ have an execution that halts with the terminal value $x=2$ ?

Exercise 2.3. Consider the following street junction with the specification of a traffic light as outlined on the right.

(a) Choose appropriate actions and label the transitions of the traffic light transition system accordingly.
(b) Give the transition system representation of a (reasonable) controller $C$ that switches the green signal lamps in the following order: $A_{1}, A_{2}, A_{3}, A_{1}, A_{2}, A_{3}, \ldots$.
(Hint: Choose an appropriate communication mechanism.)
(c) Outline the transition system $A_{1}\left\|A_{2}\right\| A_{3} \| C$.

EXERCISE 2.4. Show that the handshaking operator $\|$ that forces two transition systems to synchronize over their common actions (see Definition 2.26 on page 48 ) is associative. That is, show that

$$
\left(T S_{1} \| T S_{2}\right)\left\|T S_{3}=T S_{1}\right\|\left(T S_{2} \| T S_{3}\right)
$$

where $T S_{1}, T S_{2}, T S_{3}$ are arbitrary transition systems.

EXERCISE 2.5. The following program is a mutual exclusion protocol for two processes due to Pnueli [118]. There is a single shared variable $s$ which is either 0 or 1 , and initially 1 . Besides, each process has a local Boolean variable $y$ that initially equals 0 . The program text for process $P_{i}(i=0,1)$ is as follows:

## 10: loop forever do <br> begin

11: Noncritical section
12: $\left(y_{i}, s\right):=(1, i)$;
13: wait until $\left(\left(y_{1-i}=0\right) \vee(s \neq i)\right)$;
14: Critical section
$15: y_{i}:=0$
end.

Here, the statement $\left(y_{i}, s\right):=(1, i)$; is a multiple assignment in which variable $y_{i}:=1$ and $s:=i$ is a single, atomic step.

Questions:
(a) Define the program graph of a process in Pnueli's algorithm.
(b) Determine the transition system for each process.
(c) Construct their parallel composition.
(d) Check whether the algorithm ensures mutual exclusion.
(e) Check whether the algorithm ensures starvation freedom.

The last two questions may be answered by inspecting the transition system.

EXERCISE 2.6. Consider a stack of nonnegative integers with capacity $n$ (for some fixed $n$ ).
(a) Give a transition system representation of this stack. You may abstract from the values on the stack and use the operations top, pop, and push with their usual meaning.
(b) Sketch a transition system representation of the stack in which the concrete stack content is explicitly represented.

Exercise 2.7. Consider the following generalization of Peterson's mutual exclusion algorithm that is aimed at an arbitrary number $n(n \geqslant 2)$ processes. The basic concept of the algorithm is that each process passes through $n$ "levels" before acquiring access to the critical section. The concurrent processes share the bounded integer arrays $y[0 . . n-1]$ and $p[1 . . n]$ with $y[i] \in\{1, \ldots, n\}$ and $p[i] \in\{0, \ldots, n-1\} . y[j]=i$ means that process $i$ has the lowest priority at level $j$, and $p[i]=j$ expresses that process $i$ is currently at level $j$. Process $i$ starts at level 0 . On requesting access to the critical section, the process passes through levels 1 through $n-1$. Process $i$ waits at level $j$ until either all other processes are at a lower level (i.e., $p[k]<j$ for all $k \neq i$ ) or another process grants process $i$ access to its critical section (i.e., $y[j] \neq i$ ). The behavior of process $i$ is in pseudocode:

```
while true do
    ... noncritical section ...
    forall \(j=1, \ldots, n-1\) do
        \(p[i]:=j ;\)
        \(y[j]:=i\);
        wait until \((y[j] \neq i) \vee\left(\bigwedge_{0<k \leqslant n, k \neq i} p[k]<j\right)\)
    od
    ... critical section ...
    \(p[i]:=0 ;\)
od
```

Questions:
(a) Give the program graph for process $i$.
(b) Determine the number of states (including the unreachable states) in the parallel composition of $n$ processes.
(c) Prove that this algorithm ensures mutual exclusion for $n$ processes.
(d) Prove that it is impossible that all processes are waiting in the for-iteration.
(e) Establish whether it is possible that a process that wants to enter the critical section waits ad infinitum.

ExERCISE 2.8. In channel systems, values can be transferred from one process to another process. As this is somewhat limited, we consider in this exercise an extension that allows for the transfer of expressions. That is to say, the send and receive statements $c!v$ and $c ? x$ (where $x$ and $v$ are of the same type) are generalized into $c!$ expr and $c ? x$, where for simplicity it is assumed that expr is a correctly typed expression (of the same type as $x$ ). Legal expressions are, e.g., $x \wedge(\neg y \vee z)$ for Boolean variables $x, y$, and $z$, and channel $c$ with $\operatorname{dom}(c)=\{0,1\}$. For integers $x, y$, and an integer-channel $c,|2 x+(x-y) \operatorname{div} 17|$ is a legal expression.

Question: Extend the transition system semantics of channel systems such that expressions are allowed in send statements.
(Hint: Use the function $\eta$ such that for expression expr, $\eta($ expr $)$ is the evaluation of expr under the variable valuation $\eta$.)

ExERCISE 2.9. Consider the following mutual exclusion algorithm that uses the shared variables $y_{1}$ and $y_{2}$ (initially both 0 ).

Process $P_{1}$ :
while true do
... noncritical section ...
$y_{1}:=y_{2}+1$;
wait until $\left(y_{2}=0\right) \vee\left(y_{1}<y_{2}\right)$
... critical section ... $y_{1}:=0 ;$
od
Questions:
(a) Give the program graph representations of both processes. (A pictorial representation suffices.)
(b) Give the reachable part of the transition system of $P_{1} \| P_{2}$ where $y_{1} \leqslant 2$ and $y_{2} \leqslant 2$.
(c) Describe an execution that shows that the entire transition system is infinite.
(d) Check whether the algorithm indeed ensures mutual exclusion.
(e) Check whether the algorithm never reaches a state in which both processes are mutually waiting for each other.
(f) Is it possible that a process that wants to enter the critical section has to wait ad infinitum?

EXERCISE 2.10. Consider the following mutual exclusion algorithm that was proposed 1966 [221] as a simplification of Dijkstra's mutual exclusion algorithm in case there are just two processes:

```
1 Boolean array b(0;1) integer k, i, j,
2 comment This is the program for computer i, which may be
    either 0 or 1, computer j =/= i is the other one, 1 or 0;
CO: b(i) := false;
4 C1: if k != i then begin
5 C2: if not b(j) then go to C2;
6 else k := i; go to C1 end;
7 else critical section;
8 b(i) := true;
9 remainder of program;
10 go to C0;
11 end
```

Here C0, C1, and C2 are program labels, and the word "computer" should be interpreted as process.

## Questions:

(a) Give the program graph representations for a single process. (A pictorial representation suffices.)
(b) Give the reachable part of the transition system of $P_{1} \| P_{2}$.
(c) Check whether the algorithm indeed ensures mutual exclusion.

Exercise 2.11. Consider the following two sequential hardware circuits $C_{1}$ and $C_{2}$ :

(a) Give the transition system representation $\operatorname{TS}\left(C_{1}\right)$ of the circuit $C_{1}$.
(b) Let $T S\left(C_{2}\right)$ be the transition system of the circuit $C_{2}$. Outline the transition system $T S\left(C_{1}\right) \otimes$ $T S\left(C_{2}\right)$.

Exercise 2.12. Consider the following leader election algorithm: For $n \in \mathbb{N}, n$ processes $P_{1}, \ldots, P_{n}$ are located in a ring topology where each process is connected by an unidirectional channel to its neighbor in a clockwise manner.

To distinguish the processes, each process is assigned a unique identifier $i d \in\{1, \ldots, n\}$. The aim is to elect the process with the highest identifier as the leader within the ring. Therefore each process executes the following algorithm:

```
send (id); initially set to process' id
while (true) do
    receive ( \(m\) );
    if \((m=i d)\) then stop; process is the leader
    if \((m>i d)\) then send \((m)\); forward identifier
od
```

(a) Model the leader election protocol for $n$ processes as a channel system.
(b) Give an initial execution fragment of $T S\left(\left[P_{1}\left|P_{2}\right| P_{3}\right]\right)$ such that at least one process has executed the send statement within the body of the whileloop. Assume for $0<i \leqslant 3$, that process $P_{i}$ has identifier $i d_{i}=i$.

## Chapter 3

## Linear-Time Properties

For verification purposes, the transition system model of the system under consideration needs to be accompanied with a specification of the property of interest that is to be verified. This chapter introduces some important, though relatively simple, classes of properties. These properties are formally defined and basic model-checking algorithms are presented to check such properties in an automated manner. This chapter focuses on linear-time behavior and establishes relations between the different classes of properties and trace behavior. Elementary forms of fairness are introduced and compared.

### 3.1 Deadlock

Sequential programs that are not subject to divergence (i.e., endless loops) have a terminal state, a state without any outgoing transitions. For parallel systems, however, computations typically do not terminate - consider, for instance, the mutual exclusion programs treated so far. In such systems, terminal states are undesirable and mostly represent a design error. Apart from "trivial" design errors where it has been forgotten to indicate certain activities, in most cases such terminal states indicate a deadlock. A deadlock occurs if the complete system is in a terminal state, although at least one component is in a (local) nonterminal state. The entire system has thus come to a halt, whereas at least one component has the possibility to continue to operate. A typical deadlock scenario occurs when components mutually wait for each other to progress.


Figure 3.1: An example of a deadlock situation.

## Example 3.1. Deadlock for Fault Designed Traffic Lights

Consider the parallel composition of two transition systems

$$
\operatorname{Tr}^{L_{i g h t}^{1}} \|^{\|} \text {TrLight }_{2}
$$

modeling the traffic lights of two intersecting roads. Both traffic lights synchronize by means of the actions $\alpha$ and $\beta$ that indicate the change of light (see Figure 3.1). The apparently trivial error to let both traffic lights start with a red light results in a deadlock. While the first traffic light is waiting to be synchronized on action $\alpha$, the second traffic light is blocked, since it is waiting to be synchronized with action $\beta$.

## Example 3.2. Dining Philosophers

This example, originated by Dijkstra, is one of the most prominent examples in the field of concurrent systems.


Five philosophers are sitting at a round table with a bowl of rice in the middle. For the philosophers (being a little unworldly) life consists of thinking and eating (and waiting, as we will see). To take some rice out of the bowl, a philosopher needs two chopsticks. In between two neighboring philosophers, however, there is only a single chopstick. Thus, at any time only one of two neighboring philosophers can eat. Of course, the use of the chopsticks is exclusive and eating with hands is forbidden.

Note that a deadlock scenario occurs when all philosophers possess a single chopstick. The problem is to design a protocol for the philosophers, such that the complete system is deadlock-free, i.e., at least one philosopher can eat and think infinitely often. Additionally, a fair solution may be required with each philosopher being able to think and eat infinitely often. The latter characteristic is called freedom of individual starvation.

The following obvious design cannot ensure deadlock freedom. Assume the philosophers and the chopsticks are numbered from 0 to 4 . Furthermore, assume all following calculations be "modulo 5 ", e.g., chopstick $i-1$ for $i=0$ denotes chopstick 4 , and so on.

Philosopher $i$ has stick $i$ on his left and stick $i-1$ on his right side. The action request $_{i, i}$ express that stick $i$ is picked up by philosopher $i$. Accordingly, request ${ }_{i-1, i}$ denotes the action by means of which philosopher $i$ picks up the $(i-1)$ th stick. The actions release ${ }_{i, i}$ and release ${ }_{i-1, i}$ have a corresponding meaning.

The behavior of philosopher $i$ (called process $P h i l_{i}$ ) is specified by the transition system depicted in the left part of Figure 3.2. Solid arrows depict the synchronizations with the $i$-th stick, dashed arrows refer to communications with the $i-1$ th stick. The sticks are modeled as independent processes (called Stick ${ }_{i}$ ) with which the philosophers synchronize via actions request and release; see the right part of Figure 3.2 that represents the process of stick $i$. A stick process prevents philosopher $i$ from picking up the $i$ th stick when philosopher $i+1$ is using it.


Figure 3.2: Transition systems for the $i$ th philosopher and the $i$ th stick.

The complete system is of the form:

$$
\text { Phil }_{4} \| \text { Stick }_{3} \| \text { Phil }_{3} \| \text { Stick }_{2} \| \text { Phil }_{2} \| \text { Stick }_{1} \| \text { Phil }_{1} \| \text { Stick }_{0} \| \text { Phil }_{0} \| \text { Stick }_{4}
$$

This (initially obvious) design leads to a deadlock situation, e.g., if all philosophers pick up their left stick at the same time. A corresponding execution leads from the initial state

$$
\left\langle\text { think }_{4}, \text { avail }_{3}, \text { think }_{3}, \text { avail }_{2}, \text { think }_{2}, \text { avail }_{1}, \text { think }_{1}, \text { avail }_{0}, \text { think }_{0}, \text { avail }_{4}\right\rangle
$$

by means of the action sequence request ${ }_{4}$, request $_{3}$, request $_{2}$, request $_{1}$, request $_{0}$ (or any other permutation of these 5 request actions) to the terminal state

$$
\left\langle\text { wait }_{4,0}, \text { occ }_{4,4}, \text { wait }_{3,4}, \text { occ }_{3,3}, \text { wait }_{2,3}, \text { occ }_{2,2}, \text { wait }_{1,2}, \text { occ }_{1,1}, \text { wait }_{0,1}, \text { occ } c_{0,0}\right\rangle .
$$

This terminal state represents a deadlock with each philosopher waiting for the needed stick to be released.

A possible solution to this problem is to make the sticks available for only one philosopher at a time. The corresponding chopstick process is depicted in the right part of Figure 3.3. In state available $_{i, j}$ only philosopher $j$ is allowed to pick up the $i$ th stick. The abovementioned deadlock situation can be avoided by the fact that some sticks (e.g., the first, the third, and the fifth stick) start in state available $_{i, i}$, while the remaining sticks start in state available $_{i, i+1}$. It can be verified that this solution is deadlock- and starvation-free.


Figure 3.3: Improved variant of the $i$ th philosopher and the $i$ th stick.

A further characteristic often required for concurrent systems is robustness against failure of their components. In the case of the dining philosophers, robustness can be formulated in a way that ensures deadlock and starvation freedom even if one of the philosophers is "defective" (i.e., does not leave the think phase anymore). ${ }^{1}$ The above-sketched deadlockand starvation-free solution can be modified to a fault-tolerant solution by changing the transition systems of philosophers and sticks such that philosopher $i+1$ can pick up the $i$ th stick even if philosopher $i$ is thinking (i.e., does not need stick $i$ ) independent of whether stick $i$ is in state available $e_{i, i}$ or available $e_{i, i+1}$. The corresponding is also true when the roles of philosopher $i$ and $i+1$ are reversed. This can be established by adding a single Boolean variable $x_{i}$ to philosopher $i$ (see Figure 3.4). The variable $x_{i}$ informs the neighboring philosophers about the current location of philosopher $i$. In the indicated sketch, $x_{i}$ is a Boolean variable which is true if and only if the $i$ th philosopher is thinking. Stick $i$ is made available to philosopher $i$ if stick $i$ is in location available $e_{i}$ (as before), or if stick $i$ is in location available $e_{i+1}$ while philosopher $i+1$ is thinking.

Note that the above description is at the level of program graphs. The complete system is a channel system with request and release actions standing for handshaking over a channel of capacity 0 .

[^6]

Figure 3.4: Fault-tolerant variant of the dining philosophers.

### 3.2 Linear-Time Behavior

To analyze a computer system represented by a transition system, either an action-based or a state-based approach can be followed. The state-based approach abstracts from actions; instead, only labels in the state sequences are taken into consideration. In contrast, the action-based view abstracts from states and refers only to the action labels of the transitions. (A combined action- and state-based view is possible, but leads to more involved definitions and concepts. For this reason it is common practice to abstract from either action or state labels.) Most of the existing specification formalisms and associated verification methods can be formulated in a corresponding way for both perspectives.

In this chapter, we mainly focus on the state-based approach. Action labels of transitions are only necessary for modeling communication; thus, they are of no relevance in the following chapters. Instead, we use the atomic propositions of the states to formulate system properties. Therefore, the verification algorithms operate on the state graph of a transition system, the digraph originating from a transition system by abstracting from action labels.

### 3.2.1 Paths and State Graph

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system.

## Definition 3.3. State Graph

The state graph of $T S$, notation $G(T S)$, is the digraph $(V, E)$ with vertices $V=S$ and edges $E=\left\{\left(s, s^{\prime}\right) \in S \times S \mid s^{\prime} \in \operatorname{Post}(s)\right\}$.

The state graph of transition system $T S$ has a vertex for each state in $T S$ and an edge between vertices $s$ and $s^{\prime}$ whenever $s^{\prime}$ is a direct successor of $s$ in $T S$ for some action $\alpha$. It is thus simply obtained from TS by omitting all state labels (i.e., the atomic propositions), all transition labels (i.e., the actions), and by ignoring the fact whether a state is initial or not. Moreover, multiple transitions (that have different action labels) between states are represented by a single edge. This seems to suggest that the state labels are no longer of any use; later on, we will see how these state labels will be used to check the validity of properties.

Let Post* $(s)$ denote the states that are reachable in state graph $G(T S)$ from $s$. This notion is generalized toward sets of states in the usual way (i.e., pointwise extension): for $C \subseteq S$ let

$$
\operatorname{Post}^{*}(C)=\bigcup_{s \in C} \operatorname{Post}^{*}(s)
$$

The notations $\operatorname{Pre}^{*}(s)$ and $\operatorname{Pr} e^{*}(C)$ have analogous meaning. The set of states that are reachable from some initial state, notation $\operatorname{Reach}(T S)$, equals $\operatorname{Post}^{*}(I)$.

As explained in Chapter 2, the possible behavior of a transition system is defined by an execution fragment. Recall that an execution fragment is an alternating sequence of states and actions. As we consider a state-based approach, the actions are not of importance and are omitted. The resulting "runs" of a transition system are called paths. The following definitions define path fragments, initial and maximal path fragments, and so on. These notions are easily obtained from the same notions for executions by omitting the actions.

## Definition 3.4. Path Fragment

A finite path fragment $\widehat{\pi}$ of $T S$ is a finite state sequence $s_{0} s_{1} \ldots s_{n}$ such that $s_{i} \in \operatorname{Post}\left(s_{i-1}\right)$ for all $0<i \leqslant n$, where $n \geqslant 0$. An infinite path fragment $\pi$ is an infinite state sequence $s_{0} s_{1} s_{2} \ldots$ such that $s_{i} \in \operatorname{Post}\left(s_{i-1}\right)$ for all $i>0$.

We adopt the following notational conventions for infinite path fragment $\pi=s_{0} s_{1} \ldots$ The initial state of $\pi$ is denoted by $\operatorname{first}(\pi)=s_{0}$. For $j \geqslant 0$, let $\pi[j]=s_{j}$ denote the $j$ th state of
$\pi$ and $\pi[. . j]$ denote the $j$ th prefix of $\pi$, i.e., $\pi[. . j]=s_{0} s_{1} \ldots s_{j}$. Similarly, the $j$ th suffix of $\pi$, notation $\pi[j .$.$] , is defined as \pi[j .]=.s_{j} s_{j+1} \ldots$. These notions are defined analogously for finite paths. Besides, for finite path $\widehat{\pi}=s_{0} s_{1} \ldots s_{n}$, let last $(\widehat{\pi})=s_{n}$ denote the last state of $\widehat{\pi}$, and $\operatorname{len}(\widehat{\pi})=n$ denote the length of $\widehat{\pi}$. For infinite path $\pi$ these notions are defined by $\operatorname{len}(\pi)=\infty$ and last $(\pi)=\perp$, where $\perp$ denotes "undefined".

## Definition 3.5. Maximal and Initial Path Fragment

A maximal path fragment is either a finite path fragment that ends in a terminal state, or an infinite path fragment. A path fragment is called initial if it starts in an initial state, i.e., if $s_{0} \in I$.

A maximal path fragment is a path fragment that cannot be prolonged: either it is infinite or it is finite but ends in a state from which no transition can be taken. Let Paths( $s$ ) denote the set of maximal path fragments $\pi$ with $\operatorname{first}(\pi)=s$, and $\operatorname{Paths}_{f i n}(s)$ denote the set of all finite path fragments $\widehat{\pi}$ with $\operatorname{first}(\widehat{\pi})=s$.

## Definition 3.6. Path

A path of transition system TS is an initial, maximal path fragment. ${ }^{2}$

Let $\operatorname{Paths}(T S)$ denote the set of all paths in $T S$, and $\operatorname{Paths}_{f i n}(T S)$ the set of all initial, finite path fragments of $T S$.

## Example 3.7. Beverage Vending Machine

Consider the beverage vending machine of Example 2.2 on page 21. For convenience, its transition system is repeated in Figure 3.5. As the state labeling is simply $L(s)=\{s\}$ for each state $s$, the names of states may be used in paths (as in this example), as well as atomic propositions (as used later on). Example path fragments of this transition system are

$$
\begin{aligned}
\pi_{1} & =\text { pay select soda pay select soda } \ldots \\
\pi_{2} & =\text { select soda pay select beer } \ldots \\
\widehat{\pi} & =\text { pay select soda pay select soda } .
\end{aligned}
$$

These path fragments result from the execution fragments indicated in Example 2.8 on page 25 . Only $\pi_{1}$ is a path. The infinite path fragment $\pi_{2}$ is maximal but not initial. $\widehat{\pi}$ is initial but not maximal since it is finite while ending in a state that has outgoing

[^7]

Figure 3.5: A transition system of a simple beverage vending machine.
transitions. We have that $\operatorname{last}(\widehat{\pi})=\operatorname{soda}$, first $\left(\pi_{2}\right)=$ select, $\pi_{1}[0]=$ pay, $\pi_{1}[3]=$ pay, $\pi_{1}[. .5]=\widehat{\pi}, \widehat{\pi}[. .2]=\widehat{\pi}[3 .],. \operatorname{len}(\widehat{\pi})=5$, and $\operatorname{len}\left(\pi_{1}\right)=\infty$.

### 3.2.2 Traces

Executions (as introduced in Chapter 2) are alternating sequences consisting of states and actions. Actions are mainly used to model the (possibility of) interaction, be it synchronous or asynchronous communication. In the sequel, interaction is not our prime interest, but instead we focus on the states that are visited during executions. In fact, the states themselves are not "observable", but just their atomic propositions. Thus, rather than having an execution of the form $s_{0} \xrightarrow{\alpha_{0}} s_{1} \xrightarrow{\alpha_{1}} s_{2} \ldots$ we consider sequences of the form $L\left(s_{0}\right) L\left(s_{1}\right) L\left(s_{2}\right) \ldots$ that register the (set of) atomic propositions that are valid along the execution. Such sequences are called traces.

The traces of a transition system are thus words over the alphabet $2^{A P}$. In the following it is assumed that a transition system has no terminal states. In this case, all traces are infinite words. (Recall that the traces of a transition system have been defined as traces induced by its initial maximal path fragments. See also Appendix A.2, page 912). This assumption is made for simplicity and does not impose any serious restriction. First of all, prior to checking any (linear-time) property, a reachability analysis could be carried out to determine the set of terminal states. If indeed some terminal state is encountered, the system contains a deadlock and has to be repaired before any further analysis. Alternatively, each transition system $T S$ (that probably has a terminal state) can be extended such that for each terminal state $s$ in $T S$ there is a new state $s_{\text {stop }}$, transition $s \rightarrow s_{s t o p}$, and $s_{s t o p}$ is equipped with a self-loop, i.e., $s_{\text {stop }} \rightarrow s_{\text {stop }}$. The resulting "equivalent" transition system obviously has no terminal states. ${ }^{3}$

[^8]
## Definition 3.8. Trace and Trace Fragment

Let $T S=(S$, Act, $\rightarrow, I, A P, L)$ be a transition system without terminal states. The trace of the infinite path fragment $\pi=s_{0} s_{1} \ldots$ is defined as $\operatorname{trace}(\pi)=L\left(s_{0}\right) L\left(s_{1}\right) \ldots$. The trace of the finite path fragment $\widehat{\pi}=s_{0} s_{1} \ldots s_{n}$ is defined as $\operatorname{trace}(\widehat{\pi})=L\left(s_{0}\right) L\left(s_{1}\right) \ldots L\left(s_{n}\right)$.

The trace of a path fragment is thus the induced finite or infinite word over the alphabet $2^{A P}$, i.e., the sequence of sets of atomic propositions that are valid in the states of the path. The set of traces of a set $\Pi$ of paths is defined in the usual way:

$$
\operatorname{trace}(\Pi)=\{\operatorname{trace}(\pi) \mid \pi \in \Pi\}
$$

A trace of state $s$ is the trace of an infinite path fragment $\pi$ with $\operatorname{first}(\pi)=s$. Accordingly, a finite trace of $s$ is the trace of a finite path fragment that starts in $s$. Let $\operatorname{Traces}(s)$ denote the set of traces of $s$, and $\operatorname{Traces}(T S)$ the set of traces of the initial states of transition system TS:

$$
\operatorname{Traces}(s)=\operatorname{trace}(\operatorname{Paths}(s)) \quad \text { and } \quad \operatorname{Traces}(T S)=\bigcup_{s \in I} \operatorname{Traces}(s)
$$

In a similar way, the finite traces of a state and of a transition system are defined:

$$
\operatorname{Traces}_{f i n}(s)=\operatorname{trace}^{\left(\operatorname{Path}_{f i n}(s)\right) \quad \text { and } \quad \operatorname{Traces}_{f i n}(T S)=\bigcup_{s \in I} \operatorname{Traces}_{f i n}(s) . . . . . . . . ~}
$$

## Example 3.9. Semaphore-Based Mutual Exclusion

Consider the transition system $T S_{S e m}$ as depicted in Figure 3.6. This two-process mutual exclusion example has been described before in Example 2.24 (page 43).

Assume the available atomic propositions are crit ${ }_{1}$ and $c^{2} t_{2}$, i.e.,

$$
A P=\left\{\text { crit }_{1}, c r i t_{2}\right\} .
$$

The proposition crit ${ }_{1}$ holds in any state of the transition system $T S_{S e m}$ where the first process (called $P_{1}$ ) is in its critical section. Proposition crit ${ }_{2}$ has the same meaning for the second process (i.e., $P_{2}$ ).

Consider the execution in which the processes $P_{1}$ and $P_{2}$ enter their critical sections in an alternating fashion. Besides, they only request to enter the critical section when the

[^9]

Figure 3.6: Transition system of semaphore-based mutual exclusion algorithm.
other process is no longer in its critical section. Situations in which one process is in its critical section whereas the other is moving from the noncritical state to the waiting state are impossible.

The path $\pi$ in the state graph of $T S_{S e m}$ where process $P_{1}$ is the first to enter its critical section is of the form

$$
\left.\begin{array}{rl}
\pi=\left\langle n_{1}, n_{2}, y=1\right\rangle & \rightarrow\left\langle w_{1}, n_{2}, y=1\right\rangle
\end{array} \rightarrow\left\langle c_{1}, n_{2}, y=0\right\rangle \rightarrow 0.1\right\rangle\left\langle n_{1}, w_{2}, y=1\right\rangle \rightarrow\left\langle n_{1}, c_{2}, y=0\right\rangle \rightarrow \ldots .
$$

The trace of this path is the infinite word:

$$
\operatorname{trace}(\pi)=\varnothing \varnothing\left\{\operatorname{crit}_{1}\right\} \varnothing \varnothing\left\{\operatorname{crit}_{2}\right\} \varnothing \varnothing\left\{\text { crit }_{1}\right\} \varnothing \varnothing\left\{\operatorname{crit}_{2}\right\} \ldots
$$

The trace of the finite path fragment

$$
\begin{aligned}
& \widehat{\pi}=\left\langle n_{1}, n_{2}, y=1\right\rangle \rightarrow\left\langle w_{1}, n_{2}, y=1\right\rangle \\
& \rightarrow\left\langle w_{1}, w_{2}, y=1\right\rangle \rightarrow \\
&\left\langle w_{1}, c_{2}, y=0\right\rangle
\end{aligned} \rightarrow\left\langle w_{1}, n_{2}, y=1\right\rangle \rightarrow\left\langle c_{1}, n_{2}, y=0\right\rangle
$$

is $\operatorname{trace}(\widehat{\pi})=\varnothing \varnothing \varnothing\left\{\right.$ crit $\left._{2}\right\} \varnothing\left\{\right.$ crit $\left._{1}\right\}$.

### 3.2.3 Linear-Time Properties

Linear-time properties specify the traces that a transition system should exhibit. Informally speaking, one could say that a linear-time property specifies the admissible (or desired) behavior of the system under consideration. In the following we provide a formal definition of such properties. This definition is rather elementary, and gives a good basic understanding of what a linear-time property is. In Chapter 5, a logical formalism will be introduced that allows for the specification of linear-time properties.

In the following, we assume a fixed set of propositions $A P$. A linear-time (LT) property is a requirement on the traces of a transition system. Such property can be understood as a requirement over all words over $A P$, and is defined as the set of words (over $A P$ ) that are admissible:

## Definition 3.10. LT Property

A linear-time property (LT property) over the set of atomic propositions AP is a subset of $\left(2^{A P}\right)^{\omega}$.

Here, $\left(2^{A P}\right)^{\omega}$ denotes the set of words that arise from the infinite concatenation of words in $2^{A P}$. An LT property is thus a language (set) of infinite words over the alphabet $2^{A P}$. Note that it suffices to consider infinite words only (and not finite words), as transition systems without terminal states are considered. The fulfillment of an LT property by a transition system is defined as follows.

## Definition 3.11. Satisfaction Relation for LT Properties

Let $P$ be an LT property over $A P$ and $T S=(S, A c t, \rightarrow, I, A P, L)$ a transition system without terminal states. Then, $T S=(S, A c t, \rightarrow, I, A P, L)$ satisfies $P$, denoted $T S \vDash P$, iff $\operatorname{Traces}(T S) \subseteq P$. State $s \in S$ satisfies $P$, notation $s \models P$, whenever $\operatorname{Traces}(s) \subseteq P$.

Thus, a transition system satisfies the LT property $P$ if all its traces respect $P$, i.e., if all its behaviors are admissible. A state satisfies $P$ whenever all traces starting in this state fulfill $P$.

## Example 3.12. Traffic Lights

Consider two simplified traffic lights that only have two possible settings: red and green. Let the propositions of interest be

$$
A P=\left\{\text { red }_{1}, \text { green }_{1}, \text { red }_{2}, \text { green }_{2}\right\}
$$



Figure 3.7: Two fully synchronized traffic lights (left and middle) and their parallel composition (right).

We consider two LT properties of these traffic lights and give some example words that are contained by such properties. First, consider the property $P$ that states:
"The first traffic light is infinitely often green".
This LT property corresponds to the set of infinite words of the form $A_{0} A_{1} A_{2} \ldots$ over $2^{A P}$, such that green ${ }_{1} \in A_{i}$ holds for infinitely many $i$. For example, $P$ contains the infinite words

```
{red
\varnothing{\mp@subsup{green_}{1}{}}}\varnothing{\mp@subsup{\mathrm{ green_}}{1}{}}\varnothing{\mp@subsup{\mathrm{ green }}{1}{}}\varnothing{\mp@subsup{\mathrm{ green }}{1}{}}\varnothing\ldots
{red
{\mp@subsup{green }{1}{\prime},\mp@subsup{\mathrm{ green }}{2}{}}{\mp@subsup{\mathrm{ green }}{1}{},\mp@subsup{\mathrm{ green }}{2}{}}{\mp@subsup{\mathrm{ green }}{1}{},\mp@subsup{\mathrm{ green }}{2}{}}{{\mp@subsup{\mathrm{ green }}{1}{},\mp@subsup{\mathrm{ green }}{2}{}}\ldots
```

The infinite word $\left\{\operatorname{red}_{1}\right.$, green $\left._{1}\right\}\left\{\operatorname{red}_{1}\right.$, green $\left._{1}\right\} \varnothing \varnothing \varnothing \varnothing \ldots$ is not in $P$ as it contains only finitely many occurrences of green $_{1}$.

As a second LT property, consider $P^{\prime}$ :
"The traffic lights are never both green simultaneously".

This property is formalized by the set of infinite words of the form $A_{0} A_{1} A_{2} \ldots$ such that either green $_{1} \notin A_{i}$ or green ${ }_{2} \notin A_{i}$, for all $i \geqslant 0$. For example, the following infinite words are in $P^{\prime}$ :

$$
\begin{aligned}
& \left\{\operatorname{red}_{1}, \text { green }_{2}\right\}\left\{\text { green }_{1}, \text { red }_{2}\right\}\left\{\text { red }_{1}, \text { green }_{2}\right\}\left\{\text { green }_{1}, \text { red }_{2}\right\} \ldots, \\
& \varnothing\left\{\text { green }_{1}\right\} \varnothing\left\{\text { green }_{1}\right\} \varnothing\left\{\text { green }_{1}\right\} \varnothing\left\{\text { green }_{1}\right\} \varnothing \ldots \text { and } \\
& \left\{\text { red }_{1}, \text { green }_{1}\right\}\left\{\text { red }_{1}, \text { green }_{1}\right\}\left\{\text { red }_{1}, \text { green }_{1}\right\}\left\{\text { red }_{1}, \text { green }_{1}\right\} \ldots .
\end{aligned}
$$

whereas the infinite word $\left\{\right.$ red $_{1}$ green $\left._{2}\right\}\left\{\right.$ green $_{1}$, green $\left._{2}\right\}, \ldots$ is not in $P^{\prime}$.
The traffic lights depicted in Figure 3.7 are at intersecting roads and their switching is synchronized, i.e., if one light switches from red to green, the other switches from green to
red. In this way, the lights always have complementary colors. Clearly, these traffic lights satisfy both $P$ and $P^{\prime}$. Traffic lights that switch completely autonomously will neither satisfy $P$-there is no guarantee that the first traffic light is green infinitely often-nor $P^{\prime}$.

Often, an LT property does not refer to all atomic propositions occurring in a transition system, but just to a relatively small subset thereof. For a property $P$ over a set of propositions $A P^{\prime} \subseteq A P$, only the labels in $A P^{\prime}$ are relevant. Let $\widehat{\pi}$ be a finite path fragment of TS. We write $\operatorname{trace}_{A P^{\prime}}(\widehat{\pi})$ to denote the finite trace of $\widehat{\pi}$ where only the atomic propositions in $A P^{\prime}$ are considered. Accordingly, $\operatorname{trace}_{A P^{\prime}}(\pi)$ denotes the trace of an infinite path fragment $\pi$ by focusing on propositions in $A P^{\prime}$ Thus, for $\pi=s_{0} s_{1} s_{2} \ldots$, we have

$$
\operatorname{trace}_{A P^{\prime}}(\pi)=L^{\prime}\left(s_{0}\right) L^{\prime}\left(s_{1}\right) \ldots=\left(L\left(s_{0}\right) \cap A P^{\prime}\right)\left(L\left(s_{1}\right) \cap A P^{\prime}\right) \ldots
$$

Let $\operatorname{Traces}_{A P^{\prime}}(T S)$ denote the set of traces $\operatorname{trace}_{A P^{\prime}}(\operatorname{Paths}(T S))$. Whenever the set $A P^{\prime}$ of atomic propositions is clear from the context, the subscript $A P^{\prime}$ is omitted. In the rest of this chapter, the restriction to a relevant subset of atomic propositions is often implicitly made.

## Example 3.13. The Mutual Exclusion Property

In Chapter 2, several mutual exclusion algorithms have been considered. For specifying the mutual exclusion property - always at most one process is in its critical section-it suffices to only consider the atomic propositions crit ${ }_{1}$ and crit ${ }_{2}$. Other atomic propositions are not of any relevance for this property. The formalization of the mutual exclusion property is given by the LT property

$$
P_{\text {mutex }}=\text { set of infinite words } A_{0} A_{1} A_{2} \ldots \text { with }\left\{\text { crit }_{1}, \text { crit }_{2}\right\} \nsubseteq A_{i} \text { for all } 0 \leqslant i
$$

For example, the infinite words

$$
\begin{aligned}
& \left\{\text { crit }_{1}\right\}\left\{\operatorname{crit}_{2}\right\}\left\{\text { crit }_{1}\right\}\left\{\text { crit }_{2}\right\}\left\{\text { crit }_{1}\right\}\left\{\text { crit }_{2}\right\} \ldots, \\
& \left\{\text { crit }_{1}\right\}\left\{\operatorname{crit}_{1}\right\}\left\{\text { crit }_{1}\right\}\left\{\text { crit }_{1}\right\}\left\{\text { crit }_{1}\right\}\left\{\text { crit }_{1}\right\} \ldots, \\
& \varnothing \varnothing \varnothing \varnothing \text { and } \\
& \varnothing \varnothing \varnothing
\end{aligned}
$$

are all contained in $P_{\text {mutex }}$. However, this does not apply to words of the form

$$
\left\{\text { crit }_{1}\right\} \varnothing\left\{\text { crit }_{1}, \text { crit }_{2}\right\} \ldots
$$

The transition system $T S_{A r b}=\left(T S_{1} \| S_{2}\right) \|$ Arbiter described in Example 2.28 (page 50) fulfills the mutex property, i.e.,

$$
T S_{\text {Arb }} \models P_{\text {mutex }}
$$

It is left to the reader to check that the mutex property is also fulfilled by the semaphorebased mutual exclusion algorithm (see Figure 3.6 on page 99) and Peterson's algorithm (see Example 2.25 on page 45).

## Example 3.14. Starvation Freedom

Guaranteeing mutual exclusion is a significant property of mutual exclusion algorithms, but is not the only relevant property. An algorithm that never allows a process to enter its critical section will do, but is certainly not intended. Besides, a property is imposed that requires a process that wants to enter the critical section to be able to eventually do so. This property prevents a process from waiting ad infinitum and is formally specified as the LT property $P_{\text {finwait }}=$ set of infinite words $A_{0} A_{1} A_{2} \ldots$ such that

$$
\forall j . l \text { wait }_{i} \in A_{j} \Rightarrow \exists k \geqslant j . \text { wait }_{i} \in A_{k} \text { for each } i \in\{1,2\} .
$$

Here, we assumed the set of propositions to be:

$$
A P=\left\{\text { wait }_{1}, \text { crit }_{1}, \text { wait }_{2}, \text { crit }_{2}\right\} .
$$

Property $P_{\text {finwait }}$ expresses that each of the two processes enters its critical section eventually if they are waiting. That is, a process has to wait some finite amount before entering the critical section. It does not express that a process that waits often, is often entering the critical section.

Consider the following variant. The LT property $P_{\text {nostarve }}=$ set of infinite words $A_{0} A_{1} A_{2} \ldots$ such that:

$$
\left(\forall k \geqslant 0 . \exists j \geqslant k . \text { wait }_{i} \in A_{j}\right) \Rightarrow\left(\forall k \geqslant 0 . \exists j \geqslant k . \text { crit }_{i} \in A_{j}\right) \quad \text { for each } i \in\{1,2\} .
$$

In abbreviated form we write:

$$
\binom{\infty}{\exists j . \text { wait }_{i} \in A_{j}} \Rightarrow\left(\begin{array}{l}
\infty \\
\exists \\
j . \text { crit }_{i} \in A_{j}
\end{array}\right) \quad \text { for each } i \in\{1,2\}
$$

where $\stackrel{\infty}{\exists}$ stands for "there are infinitely many".
Property $P_{\text {nostarve }}$ expresses that each of the two processes enters its critical section infinitely often if they are waiting infinitely often. This natural requirement is, however, not satisfied for the semaphore-based solution, since

$$
\varnothing\left(\left\{\text { wait }_{2}\right\}\left\{\text { wait }_{1}, \text { wait }_{2}\right\}\left\{\text { crit }_{1}, \text { wait }_{2}\right\}\right)^{\omega}
$$

is a possible trace of the transition system but does not belong to $P_{\text {nostarve }}$. This trace represents an execution in which only the first process enters its critical section infinitely often. In fact, the second process waits infinitely long to enter its critical section.

It is left to the reader to check that the transition system modeling Peterson's algorithm (see Example 2.25, page 45) does indeed satisfy $P_{\text {nostarve }}$.

### 3.2.4 Trace Equivalence and Linear-Time Properties

LT properties specify the (infinite) traces that a transition system should exhibit. If transition systems $T S$ and $T S^{\prime}$ have the same traces, one would expect that they satisfy the same LT properties. Clearly, if $T S \models P$, then all traces of $T S$ are contained in $P$, and when $\operatorname{Traces}(T S)=\operatorname{Traces}\left(T S^{\prime}\right)$, the traces of $T S^{\prime}$ are also contained in $P$. Otherwise, whenever $T S \not \vDash P$, there is a trace in $\operatorname{Traces}(T S)$ that is prohibited by $P$, i.e., not included in the set $P$ of traces. As $\operatorname{Traces}(T S)=\operatorname{Traces}\left(T S^{\prime}\right)$, also $T S^{\prime}$ exhibits this prohibited trace, and thus $T S^{\prime} \notin P$. The precise relationship between trace equivalence, trace inclusion, and the satisfaction of LT properties is the subject of this section.

We start by considering trace inclusion and its importance in concurrent system design. Trace inclusion between transition systems $T S$ and $T S^{\prime}$ requires that all traces exhibited by $T S$ can also be exhibited by $T S^{\prime}$, i.e., $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$. Note that transition system $T S^{\prime}$ may exhibit more traces, i.e., may have some (linear-time) behavior that $T S$ does not have. In stepwise system design, where designs are successively refined, trace inclusion is often viewed as an implementation relation in the sense that

$$
\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right) \text { means } T S \text { "is a correct implementation of" } T S^{\prime}
$$

For example, let $T S^{\prime}$ be a (more abstract) design where parallel composition is modeled by interleaving, and TS its realization where (some of) the interleaving is resolved by means of some scheduling mechanism. TS may thus be viewed as an "implementation" of $T S^{\prime}$, and clearly, $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$.

What does trace inclusion have to do with LT properties? The following theorem shows that trace inclusion is compatible with requirement specifications represented as LT properties.

## Theorem 3.15. Trace Inclusion and LT Properties

Let TS and TS ${ }^{\prime}$ be transition systems without terminal states and with the same set of propositions AP. Then the following statements are equivalent:
(a) $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$
(b) For any LT property $P: T S^{\prime} \models P$ implies $T S \models P$.

Proof: $(\mathrm{a}) \Longrightarrow(\mathrm{b})$ : Assume $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$, and let $P$ be an LT property such that $T S^{\prime} \models P$. From Definition 3.11 it follows that $\operatorname{Traces}\left(T S^{\prime}\right) \subseteq P$. Given $\operatorname{Traces}(T S) \subseteq$
$\operatorname{Traces}\left(T S^{\prime}\right)$, it now follows that $\operatorname{Traces}(T S) \subseteq P$. By Definition 3.11 it follows that $T S \models P$.
$(\mathrm{b}) \Longrightarrow(\mathrm{a})$ : Assume that for all LT properties it holds that: $T S^{\prime} \vDash P$ implies $T S \vDash P$. Let $P=\operatorname{Traces}\left(T S^{\prime}\right)$. Obviously, $T S^{\prime} \models P$, as $\operatorname{Traces}\left(T S^{\prime}\right) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$. By assumption, $T S \models P$. Hence, $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$.

This simple observation plays a decisive role for the design by means of successive refinement. If $T S^{\prime}$ is the transition system representing a preliminary design and $T S$ is a transition system originating from a refinement of $T S^{\prime}$ (i.e., a more detailed design), then it can immediately - without explicit proof-be concluded from the relation $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$ that any LT property that holds in $T S^{\prime}$ also holds for $T S$.

## Example 3.16. Refining the Semaphore-Based Mutual Exclusion Algorithm

Let $T S^{\prime}=T S_{S e m}$, the transition system representing the semaphore-based mutual exclusion algorithm (see Figure 3.6 on page 99) and let TS be the transition system obtained from $T S^{\prime}$ by removing the transition

$$
\left\langle\text { wait }_{1}, \text { wait }_{2}, y=1\right\rangle \rightarrow\left\langle\text { wait }_{1}, \text { crit }_{2}, y=0\right\rangle .
$$

Stated in words, from the situation in which both processes are waiting, it is no longer possible that the second process $\left(P_{2}\right)$ acquires access to the critical section. This thus yields a model that assigns higher priority to process $P_{1}$ than to process $P_{2}$ when both processes are competing to access the critical section. As a transition is removed, it immediately follows that $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$. Consequently, by the fact that $T S^{\prime}$ ensures mutual exclusion, i.e., $T S^{\prime} \models P_{\text {mutex }}$, it follows by Theorem 3.15 that $T S \models P_{\text {mutex }}$.

Transition systems are said to be trace-equivalent if they have the same set of traces:

## Definition 3.17. Trace Equivalence

Transition systems $T S$ and $T S^{\prime}$ are trace-equivalent with respect to the set of propositions $A P$ if $\operatorname{Traces}_{A P}(T S)=\operatorname{Traces}_{A P}\left(T S^{\prime}\right) .{ }^{4}$

Theorem 3.15 implies equivalence of two trace-equivalent transition systems with respect to requirements formulated as LT properties.

[^10]
## Corollary 3.18. Trace Equivalence and LT Properties

Let TS and TS ${ }^{\prime}$ be transition systems without terminal states and with the same set of atomic propositions. Then:

$$
\operatorname{Traces}(T S)=\operatorname{Traces}\left(T S^{\prime}\right) \Longleftrightarrow T S \text { and } T S^{\prime} \text { satisfy the same } L T \text { properties. }
$$

There thus does not exist an LT property that can distinguish between trace-equivalent transition systems. Stated differently, in order to establish that the transition systems TS and $T S^{\prime}$ are not trace-equivalent it suffices to find one LT property that holds for one but not for the other.

## Example 3.19. Two Beverage Vending Machines

Consider the two transition systems in Figure 3.8 that both model a beverage vending


Figure 3.8: Two beverage vending machines.
machine. For simplicity, the observable action labels of transitions have been omitted. Both machines are able to offer soda and beer. The left transition system models a beverage machine that after insertion of a coin nondeterministically chooses to either provide soda or beer. The right one, however, has two selection buttons (one for each beverage), and after insertion of a coin, nondeterministically blocks one of the buttons. In either case, the user has no control over the beverage obtained - the choice of beverage is under full control of the vending machine.

Let $A P=\{$ pay, soda, beer $\}$. Although the two vending machines behave differently, it is not difficult to see that they exhibit the same traces when considering $A P$, as for both machines traces are alternating sequences of pay and either soda or beer. The vending machines are thus trace-equivalent. By Corollary 3.18 both vending machines satisfy exactly the same LT properties. Stated differently, it means that there does not exist an LT property that distinguishes between the two vending machines.

### 3.3 Safety Properties and Invariants

Safety properties are often characterized as "nothing bad should happen". The mutual exclusion property -always at most one process is in its critical section-is a typical safety property. It states that the bad thing (having two or more processes in their critical section simultaneously) never occurs. Another typical safety property is deadlock freedom. For the dining philosophers (see Example 3.2, page 90), for example, such deadlock could be characterized as the situation in which all philosophers are waiting to pick up the second chopstick. This bad (i.e., unwanted) situation should never occur.

### 3.3.1 Invariants

In fact, the above safety properties are of a particular kind: they are invariants. Invariants are LT properties that are given by a condition $\Phi$ for the states and require that $\Phi$ holds for all reachable states.

## Definition 3.20. Invariant

An LT property $P_{\text {inv }}$ over $A P$ is an invariant if there is a propositional logic formula ${ }^{5} \Phi$ over $A P$ such that

$$
P_{i n v}=\left\{A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}\left|\forall j \geqslant 0 . A_{j}\right|=\Phi\right\} .
$$

$\Phi$ is called an invariant condition (or state condition) of $P_{\text {inv }}$.

Note that

$$
\begin{array}{lll}
T S \models P_{\text {inv }} & \text { iff } & \operatorname{trace}(\pi) \in P_{\text {inv }} \text { for all paths } \pi \text { in } T S \\
& \text { iff } & L(s) \models \Phi \text { for all states } s \text { that belong to a path of } T S \\
& \text { iff } & L(s) \models \Phi \text { for all states } s \in \operatorname{Reach}(T S) .
\end{array}
$$

Thus, the notion "invariant" can be explained as follows: the condition $\Phi$ has to be fulfilled by all initial states and satisfaction of $\Phi$ is invariant under all transitions in the reachable fragment of the given transition system. The latter means that if $\Phi$ holds for the source state $s$ of a transition $s \xrightarrow{a} s^{\prime}$, then $\Phi$ holds for the target state $s^{\prime}$ too.

Let us return to the examples of mutual exclusion and deadlock freedom for the dining philosophers. The mutual exclusion property can be described by an invariant using the

[^11]propositional logic formula
$$
\Phi=\neg c^{\prime} t_{1} \vee \neg c r i t_{2} .
$$

For deadlock freedom of the dining philosophers, the invariant ensures that at least one of the philosophers is not waiting to pick up the chopstick. This can be established using the propositional formula:

$$
\Phi=\neg \text { wait }_{0} \vee \neg \text { wait }_{1} \vee \neg \text { wait }_{2} \vee \neg \text { wait }_{3} \vee \neg \text { wait }_{4}
$$

Here, the proposition wait ${ }_{i}$ characterizes the state(s) of philosopher $i$ in which he is waiting for a chopstick.

How do we check whether a transition system satisfies an invariant? As checking an invariant for the propositional formula $\Phi$ amounts to checking the validity of $\Phi$ in every state that is reachable from some initial state, a slight modification of standard graph traversal algorithms like depth-first search (DFS) or breadth-first search (BFS) will do, provided the given transition system $T S$ is finite.

Algorithm 3 on page 109 summarizes the main steps for checking the invariant condition $\Phi$ by means of a forward depth-first search in the state graph $G(T S)$. The notion forward search means that we start from the initial states and investigate all states that are reachable from them. If at least one state $s$ is visited where $\Phi$ does not hold, then the invariance induced by $\Phi$ is violated. In Algorithm 3, $R$ stores all visited states, i.e., if Algorithm 3 terminates, then $R=\operatorname{Reach}(T S)$ contains all reachable states. Furthermore, $U$ is a stack that organizes all states that still have to be visited, provided they are not yet contained in $R$. The operations push, pop, and top are the standard operations on stacks. The symbol $\varepsilon$ is used to denote the empty stack. Alternatively, a backward search could have been applied that starts with all states where $\Phi$ does not hold and calculates (by a DFS or BFS) the set $\bigcup_{s \in S, s \nLeftarrow \Phi} \operatorname{Pre}^{*}(s)$.

Algorithm 3 could be slightly improved by aborting the computation once a state $s$ is encountered that does not fulfill $\Phi$. This state is a "bad" state as it makes the transition system refute the invariant and could be returned as an error indication. Such error indication, however, is not very helpful.

Instead, an initial path fragment $s_{0} s_{1} s_{2} \ldots s_{n}$ in which all states (except the last one) satisfy $\Phi$ and $s_{n} \not \models \Phi$ would be more useful. Such a path fragment indicates a possible behavior of the transition system that violates the invariant. Algorithm 3 can be easily adapted such that a counterexample is provided on encountering a state that violates $\Phi$. To that end we exploit the (depth-first search) stack $U$. When encountering $s_{n}$ that violates $\Phi$, the stack content, read from bottom to top, contains the required initial path fragment. Algorithm 4 on page 110 thus results.

```
Algorithm 3 Naïve invariant checking by forward depth-first search
Input: finite transition system \(T S\) and propositional formula \(\Phi\)
Output: true if \(T S\) satisfies the invariant "always \(\Phi\) ", otherwise false
    set of state \(R:=\varnothing\);
    stack of state \(U:=\varepsilon\);
    bool \(b:=\) true;
    for all \(s \in I\) do
        if \(s \notin R\) then
                visit \((s) \quad\) (* perform a dfs for each unvisited initial state *)
    fi
od
return \(b\)
    procedure visit (state \(s\) )
        push \((s, U)\);
        \(R:=R \cup\{s\} ;\)
    repeat
        \(s^{\prime}:=\operatorname{top}(U)\);
        if \(\operatorname{Post}\left(s^{\prime}\right) \subseteq R\) then
            pop( \(U\) );
            \(b:=b \wedge\left(s^{\prime} \models \Phi\right) ; \quad \quad\left(*\right.\) check validity of \(\Phi\) in \(s^{\prime}\) *)
        else
            let \(s^{\prime \prime} \in \operatorname{Post}\left(s^{\prime}\right) \backslash R\)
            \(\operatorname{push}\left(s^{\prime \prime}, U\right)\);
            \(R:=R \cup\left\{s^{\prime \prime}\right\} ; \quad\) (* state \(s^{\prime \prime}\) is a new reachable state *)
        fi
        until ( \(U=\varepsilon\) )
    endproc
```

```
Algorithm 4 Invariant checking by forward depth-first search
Input: finite transition system \(T S\) and propositional formula \(\Phi\)
Output: "yes" if \(T S \models\) "always \(\Phi\) ", otherwise "no" plus a counterexample
set of states \(R:=\varnothing\);
stack of states \(U:=\varepsilon\);
bool \(b:=\) true;
while ( \(I \backslash R \neq \varnothing \wedge b\) ) do
    let \(s \in I \backslash R\);
        visit(s);
od
if \(b\) then
    return("yes") (* TS \(\models\) "always \(\Phi\) " *)
else
        return(" no", reverse( \(U\) )) (* counterexample arises from the stack content *)
fi
procedure visit (state \(s\) )
    push \((s, U)\); (* push \(s\) on the stack *)
    \(R:=R \cup\{s\} ; \quad\) (* mark \(s\) as reachable \(\left.{ }^{*}\right)\)
    repeat
        \(s^{\prime}:=\operatorname{top}(U)\);
        if \(\operatorname{Post}\left(s^{\prime}\right) \subseteq R\) then
            pop( \(U\) );
            \(b:=b \wedge\left(s^{\prime} \models \Phi\right) ; \quad \quad\left(*\right.\) check validity of \(\Phi\) in \(s^{\prime}\) )
        else
            let \(s^{\prime \prime} \in \operatorname{Post}\left(s^{\prime}\right) \backslash R\)
            push( \(s^{\prime \prime}, U\) );
            \(R:=R \cup\left\{s^{\prime \prime}\right\} ; \quad\) (* state \(s^{\prime \prime}\) is a new reachable state *)
        fi
    until \(((U=\varepsilon) \vee \neg b)\)
endproc
```

The worst-case time complexity of the proposed invariance checking algorithm is dominated by the cost for the DFS that visits all reachable states. The latter is linear in the number of states (nodes of the state graph) and transitions (edges in the state graph), provided we are given a representation of the state graph where the direct successors $s^{\prime} \in \operatorname{Post}(s)$ for any state $s$ can be encountered in time $\Theta(|\operatorname{Post}(s)|)$. This holds for a representation of the sets $\operatorname{Post}(s)$ by adjacency lists. An explicit representation of adjacency lists is not adequate in our context where the state graph of a complex system has to be analyzed. Instead, the adjacency lists are typically given in an implicit way, e.g., by a syntactic description of the concurrent processes, such as program graphs or higherlevel description languages with a program graph semantics such as nanoPromela, see Section 2.2.5, page 63). The direct successors of a state $s$ are then obtained by the axioms and rules for the transition relation for the composite system. Besides the space for the syntactic descriptions of the processes, the space required by Algorithm 4 is dominated by the representation of the set $R$ of visited states (this is typically done by appropriate hash techniques) and stack $U$. Hence, the additional space complexity of invariant checking is linear in the number of reachable states.

## Theorem 3.21. Time Complexity of Invariant Checking

The time complexity of Algorithm 4 is $\mathcal{O}(N *(1+|\Phi|)+M)$ where $N$ denotes the number of reachable states, and $M=\sum_{s \in S}|\operatorname{Post}(s)|$ the number of transitions in the reachable fragment of TS.

Proof: The time complexity of the forward reachability on the state graph $G(T S)$ is $\mathcal{O}(N+M)$. The time needed to check $s \models \Phi$ for some state $s$ is linear in the length of $\Phi .{ }^{6}$ As for each state $s$ it is checked whether $\Phi$ holds, this amounts to a total of $N+M+N *(1+|\Phi|)$ operations.

### 3.3.2 Safety Properties

As we have seen in the previous section, invariants can be viewed as state properties and can be checked by considering the reachable states. Some safety properties, however, may impose requirements on finite path fragments, and cannot be verified by considering the reachable states only. To see this, consider the example of a cash dispenser, also known as an automated teller machine (ATM). A natural requirement is that money can only be withdrawn from the dispenser once a correct personal identifier (PIN) has been provided. This property is not an invariant, since it is not a state property. It is, however, considered

[^12]to be a safety property, as any infinite run violating the requirement has a finite prefix that is "bad", i.e., in which money is withdrawn without issuing a PIN before.

Formally, safety property $P$ is defined as an LT property over $A P$ such that any infinite word $\sigma$ where $P$ does not hold contains a bad prefix. The latter means a finite prefix $\widehat{\sigma}$ where the bad thing has happened, and thus no infinite word that starts with this prefix $\widehat{\sigma}$ fulfills $P$.

## Definition 3.22. Safety Properties, Bad Prefixes

An LT property $P_{\text {safe }}$ over $A P$ is called a safety property if for all words $\sigma \in\left(2^{A P}\right)^{\omega} \backslash P_{\text {safe }}$ there exists a finite prefix $\widehat{\sigma}$ of $\sigma$ such that

$$
P_{\text {safe }} \cap\left\{\sigma^{\prime} \in\left(2^{A P}\right)^{\omega} \mid \widehat{\sigma} \text { is a finite prefix of } \sigma^{\prime}\right\}=\varnothing
$$

Any such finite word $\widehat{\sigma}$ is called a bad prefix for $P_{\text {safe }}$. A minimal bad prefix for $P_{\text {safe }}$ denotes a bad prefix $\widehat{\sigma}$ for $P_{\text {safe }}$ for which no proper prefix of $\widehat{\sigma}$ is a bad prefix for $P_{\text {safe }}$. In other words, minimal bad prefixes are bad prefixes of minimal length. The set of all bad prefixes for $P_{\text {safe }}$ is denoted by $\operatorname{BadPref}\left(P_{\text {safe }}\right)$, the set of all minimal bad prefixes by $\operatorname{MinBadPref}\left(P_{\text {safe }}\right)$.

Let us first observe that any invariant is a safety property. For propositional formula $\Phi$ over $A P$ and its invariant $P_{\text {inv }}$, all finite words of the form

$$
A_{0} A_{1} \ldots A_{n} \in\left(2^{A P}\right)^{+}
$$

with $A_{0} \models \Phi, \ldots, A_{n-1} \models \Phi$ and $A_{n} \not \models \Phi$ constitute the minimal bad prefixes for $P_{\text {inv }}$. The following two examples illustrate that there are safety properties that are not invariants.

## Example 3.23. A Safety Property for a Traffic Light

We consider a specification of a traffic light with the usual three phases "red", "green", and "yellow". The requirement that each red phase should be immediately preceded by a yellow phase is a safety property but not an invariant. This is shown in the following.

Let red, yellow, and green be atomic propositions. Intuitively, they serve to mark the states describing a red (yellow or green) phase. The property "always at least one of the lights is on" is specified by:

$$
\left\{\sigma=A_{0} A_{1} \ldots \mid A_{j} \subseteq A P \wedge A_{j} \neq \varnothing\right\}
$$

The bad prefixes are finite words that contain $\varnothing$. A minimal bad prefix ends with $\varnothing$. The property "it is never the case that two lights are switched on at the same time" is specified
by

$$
\left\{\sigma=A_{0} A_{1} \ldots\left|A_{j} \subseteq A P \wedge\right| A_{j} \mid \leqslant 1\right\}
$$

Bad prefixes for this property are words containing sets such as $\{$ red, green $\},\{$ red, yellow $\}$, and so on. Minimal bad prefixes end with such sets.

Now let $A P^{\prime}=\{$ red, yellow $\}$. The property "a red phase must be preceded immediately by a yellow phase" is specified by the set of infinite words $\sigma=A_{0} A_{1} \ldots$ with $A_{i} \subseteq$ $\{$ red, yellow $\}$ such that for all $i \geqslant 0$ we have that

$$
\text { red } \in A_{i} \text { implies } i>0 \text { and yellow } \in A_{i-1} .
$$

The bad prefixes are finite words that violate this condition. An example of bad prefixes that are minimal is:

$$
\varnothing \varnothing\{\text { red }\} \text { and } \varnothing\{\text { red }\} .
$$

The following bad prefix is not minimal:

$$
\{\text { yellow }\}\{\text { yellow }\}\{\text { red }\}\{\text { red }\} \varnothing\{\text { red }\}
$$

since it has a proper prefix $\{$ yellow $\}\{$ yellow $\}\{$ red $\}\{$ red $\}$ which is also a bad prefix.

The minimal bad prefixes of this safety property are regular in the sense that they constitute a regular language. The finite automaton in Figure 3.9 accepts precisely the minimal bad prefixes for the above safety property. ${ }^{7}$ Here, $\neg$ yellow should be read as either $\varnothing$ or $\{$ red $\}$. Note the other properties given in this example are also regular.


Figure 3.9: A finite automaton for the minimal bad prefixes of a regular safety property.

## Example 3.24. A Safety Property for a Beverage Vending Machine

For a beverage vending machine, a natural requirement is that
"The number of inserted coins is always at least the number of dispensed drinks."

[^13]Using the set of propositions \{ pay, drink $\}$ and the obvious labeling function, this property could be formalized by the set of infinite words $A_{0} A_{1} A_{2} \ldots$ such that for all $i \geqslant 0$ we have

$$
\mid\left\{0 \leqslant j \leqslant i \mid \text { pay } \in A_{j}\right\}|\geqslant|\left\{0 \leqslant j \leqslant i \mid \text { drink } \in A_{j}\right\} \mid
$$

Bad prefixes for this safety property are, for example

$$
\begin{aligned}
& \varnothing\{\text { pay }\}\{\text { drink }\}\{\text { drink }\} \text { and } \\
& \varnothing\{\text { pay }\}\{d r i n k\} \varnothing\{\text { pay }\}\{d r i n k\}\{d r i n k\}
\end{aligned}
$$

It is left to the interested reader to check that both beverage vending machines from Figure 3.8 satisfy the above safety property.

Safety properties are requirements for the finite traces which is formally stated in the following lemma:

## Lemma 3.25. Satisfaction Relation for Safety Properties

For transition system TS without terminal states and safety property $P_{\text {safe }}$ :

$$
T S \models P_{\text {safe }} \text { if and only if } \operatorname{Traces}_{f i n}(T S) \cap \operatorname{BadPref}\left(P_{\text {safe }}\right)=\varnothing .
$$

Proof: "if": By contradiction. Let $\operatorname{Traces}_{\text {fin }}(T S) \cap \operatorname{BadPref}\left(P_{\text {safe }}\right)=\varnothing$ and assume that $T S \not \vDash P_{\text {safe }}$. Then, $\operatorname{trace}(\pi) \notin P_{\text {safe }}$ for some path $\pi$ in TS. Thus, $\operatorname{trace}(\pi)$ starts with a bad prefix $\widehat{\sigma}$ for $P_{\text {safe }}$. But then, $\widehat{\sigma} \in \operatorname{Traces}_{f i n}(T S) \cap \operatorname{BadPref}\left(P_{\text {safe }}\right)$. Contradiction.
"only if": By contradiction. Let $T S \models P_{\text {safe }}$ and assume that $\widehat{\sigma} \in \operatorname{Traces}_{f i n}(T S) \cap$ $\operatorname{BadPref}\left(P_{\text {safe }}\right)$. The finite trace $\widehat{\sigma}=A_{1} \ldots A_{n} \in \operatorname{Traces}_{f i n}(T S)$ can be extended to an infinite trace $\sigma=A_{1} \ldots A_{n} A_{n+1} A_{n+2} \ldots \in \operatorname{Traces}(T S)$. Then, $\sigma \notin P_{\text {safe }}$ and thus, TS $\not \models P_{\text {safe }}$.

We conclude this section with an alternative characterization of safety properties by means of their closure.

## Definition 3.26. Prefix and Closure

For trace $\sigma \in\left(2^{A P}\right)^{\omega}$, let $\operatorname{pref}(\sigma)$ denote the set of finite prefixes of $\sigma$, i.e.,

$$
\operatorname{pref}(\sigma)=\left\{\widehat{\sigma} \in\left(2^{A P}\right)^{*} \mid \widehat{\sigma} \text { is a finite prefix of } \sigma\right\}
$$

that is, if $\sigma=A_{0} A_{1} \ldots$ then $\operatorname{pref}(\sigma)=\left\{\varepsilon, A_{0}, A_{0} A_{1}, A_{0} A_{1} A_{2}, \ldots\right\}$ is an infinite set of finite words. This notion is lifted to sets of traces in the usual way. For property $P$ over AP:

$$
\operatorname{pref}(P)=\bigcup_{\sigma \in P} \operatorname{pref}(\sigma)
$$

The closure of LT property $P$ is defined by

$$
\operatorname{closure}(P)=\left\{\sigma \in\left(2^{A P}\right)^{\omega} \mid \operatorname{pref}(\sigma) \subseteq \operatorname{pref}(P)\right\}
$$

For instance, for infinite trace $\sigma=A B A B A B \ldots$ (where $A, B \subseteq A P$ ) we have $\operatorname{pref}(\sigma)=$ $\{\varepsilon, A, A B, A B A, A B A B, \ldots\}$ which equals the regular language given by the regular expression $(A B)^{*}(A+\varepsilon)$.

The closure of an LT property $P$ is the set of infinite traces whose finite prefixes are also prefixes of $P$. Stated differently, infinite traces in the closure of $P$ do not have a prefix that is not a prefix of $P$ itself. As we will see below, the closure is a key concept in the characterization of safety and liveness properties.

## Lemma 3.27. Alternative Characterization of Safety Properties

Let $P$ be an LT property over AP. Then, $P$ is a safety property iff closure $(P)=P$.

Proof: "if": Let us assume that closure $(P)=P$. To show that $P$ is a safety property, we take an element $\sigma \in\left(2^{A P}\right)^{\omega} \backslash P$ and show that $\sigma$ starts with a bad prefix for $P$. Since $\sigma \notin P=\operatorname{closure}(P)$ there exists a finite prefix $\widehat{\sigma}$ of $\sigma$ with $\widehat{\sigma} \notin \operatorname{pref}(P)$. By definition of $\operatorname{pref}(P)$, none of the words $\sigma^{\prime} \in\left(2^{A P}\right)^{\omega}$ where $\widehat{\sigma} \in \operatorname{pref}\left(\sigma^{\prime}\right)$ belongs to $P$. Hence, $\widehat{\sigma}$ is a bad prefix for $P$, and by definition, $P$ is a safety property.
"only if": Let us assume that $P$ is a safety property. We have to show that $P=$ closure $(P)$. The inclusion $P \subseteq \operatorname{closure}(P)$ holds for all LT properties. It remains to show that closure $(P) \subseteq P$. We do so by contradiction. Let us assume that there is some $\sigma=A_{1} A_{2} \ldots \in \operatorname{closure}(P) \backslash P$. Since $P$ is a safety property and $\sigma \notin P, \sigma$ has a finite prefix

$$
\widehat{\sigma}=A_{1} \ldots A_{n} \in \operatorname{BadPref}(P) .
$$

As $\sigma \in \operatorname{closure}(P)$ we have $\widehat{\sigma} \in \operatorname{pref}(\sigma) \subseteq \operatorname{pref}(P)$. Hence, there exists a word $\sigma^{\prime} \in P$ of the form

$$
\sigma^{\prime}=\underbrace{A_{1} \ldots A_{n}}_{\text {bad prefix }} B_{n+1} B_{n+2} \ldots
$$

This contradicts the fact that $P$ is a safety property.

### 3.3.3 Trace Equivalence and Safety Properties

We have seen before that there is a strong relationship between trace inclusion of transition systems and the satisfaction of LT properties (see Theorem 3.15, page 104):

$$
\begin{array}{rc}
\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right) \quad \text { if and only if } \quad \text { for all LT properties } P: \\
& T S^{\prime} \models P \text { implies } T S \models P
\end{array}
$$

for transition systems $T S$ and $T S^{\prime}$ without terminal states. Note that this result considers all infinite traces. The above thus states a relationship between infinite traces of transition systems and the validity of LT properties. When considering only finite traces instead of infinite ones, a similar connection with the validity of safety properties can be established, as stated by the following theorem.

## Theorem 3.28. Finite Trace Inclusion and Safety Properties

Let TS and TS ${ }^{\prime}$ be transition systems without terminal states and with the same set of propositions AP. Then the following statements are equivalent:
(a) $\operatorname{Traces}_{f i n}(T S) \subseteq \operatorname{Traces}_{f i n}\left(T S^{\prime}\right)$,
(b) For any safety property $P_{\text {safe }}$ : $T S^{\prime} \models P_{\text {safe }}$ implies $T S \models P_{\text {safe }}$.

Proof:
$(\mathrm{a}) \Longrightarrow(\mathrm{b})$ : Let us assume that $\operatorname{Traces}_{f i n}(T S) \subseteq \operatorname{Traces}_{f i n}\left(T S^{\prime}\right)$ and let $P_{\text {safe }}$ be a safety property with $T S^{\prime} \models P_{\text {safe }}$. By Lemma 3.25 , we have $\operatorname{Traces}_{\text {fin }}\left(T S^{\prime}\right) \cap \operatorname{BadPref}\left(P_{\text {safe }}\right)=\varnothing$, and hence, $\operatorname{Traces}_{f i n}(T S) \cap \operatorname{BadPref}\left(P_{\text {safe }}\right)=\varnothing$. Again by Lemma 3.25, we get $T S \models P_{\text {safe }}$.
$(\mathrm{b}) \Longrightarrow(\mathrm{a})$ : Assume that $(\mathrm{b})$ holds. Let $P_{\text {safe }}=\operatorname{closure}\left(\operatorname{Traces}\left(T S^{\prime}\right)\right)$. Then, $P_{\text {safe }}$ is a safety property and we have $T S^{\prime} \models P_{\text {safe }}$ (see Exercise 3.9, page 147). Hence, (b) yields $T S \models P_{\text {safe }}$, i.e.,

$$
\operatorname{Traces}(T S) \subseteq \operatorname{closure}\left(\operatorname{Traces}\left(T S^{\prime}\right)\right)
$$

From this, we may derive

$$
\begin{aligned}
\operatorname{Traces}_{f i n}(T S) & =\operatorname{pref}(\operatorname{Traces}(T S)) \\
& \subseteq \operatorname{pref}\left(\operatorname{closure}\left(\operatorname{Traces}\left(T S^{\prime}\right)\right)\right. \\
& =\operatorname{pref}\left(\operatorname{Traces}\left(T S^{\prime}\right)\right) \\
& =\operatorname{Traces}_{f i n}\left(T S^{\prime}\right)
\end{aligned}
$$

Here we use the property that for any $P$ it holds that $\operatorname{pref}(\operatorname{closure}(P))=\operatorname{pref}(P)$ (see Exercise 3.10, page 147).

Theorem 3.28 is of relevance for the gradual design of concurrent systems. If a preliminary design (i.e., a transition system) $T S^{\prime}$ is refined to a design $T S$ such that

$$
\operatorname{Traces}(T S) \nsubseteq \operatorname{Traces}\left(T S^{\prime}\right)
$$

then the LT properties of $T S^{\prime}$ cannot be carried over to $T S$. However, if the finite traces of $T S$ are finite traces of $T S^{\prime}$ (which is a weaker requirement than full trace inclusion of $T S$ and $T S^{\prime}$ ), i.e.,

$$
\operatorname{Traces}_{f i n}(T S) \subseteq \operatorname{Traces}_{f i n}\left(T S^{\prime}\right)
$$

then all safety properties that have been established for $T S^{\prime}$ also hold for $T S$. Other requirements for $T S$, i.e., LT properties that fall outside the scope of safety properties, need to be checked using different techniques.

## Corollary 3.29. Finite Trace Equivalence and Safety Properties

Let TS and TS' be transition systems without terminal states and with the same set AP of atomic propositions. Then, the following statements are equivalent:
(a) $\operatorname{Traces}_{f i n}(T S)=\operatorname{Traces}_{f i n}\left(T S^{\prime}\right)$,
(b) For any safety property $P_{\text {safe }}$ over $A P: T S \models P_{\text {safe }} \Longleftrightarrow T S^{\prime} \models P_{\text {safe }}$.

A few remarks on the difference between finite trace inclusion and trace inclusion are in order. Since we assume transition systems without terminal states, there is only a slight difference between trace inclusion and finite trace inclusion. For finite transition systems $T S$ and $T S^{\prime}$ without terminal states, trace inclusion and finite trace inclusion coincide. This can be derived from the following theorem.

## Theorem 3.30. Relating Finite Trace and Trace Inclusion

Let TS and TS be transition systems with the same set AP of atomic propositions such that TS has no terminal states and TS' is finite. Then:

$$
\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right) \Longleftrightarrow \operatorname{Traces}_{f i n}(T S) \subseteq \operatorname{Traces}_{f i n}\left(T S^{\prime}\right)
$$

Proof: The implication from left to right follows from the monotonicity of $\operatorname{pref}(\cdot)$ and the fact that $\operatorname{Traces}_{f i n}(T S)=\operatorname{pref}(\operatorname{Traces}(T S))$ for any transition system $T S$.

It remains to consider the proof for the implication $\Longleftarrow$. Let us assume that $\operatorname{Traces}_{f i n}(T S) \subseteq$ $\operatorname{Traces}_{\text {fin }}\left(T S^{\prime}\right)$. As $T S$ has no terminal states, all traces of $T S$ are infinite. Let $A_{0} A_{1} \ldots \in$
$\operatorname{Traces}(T S)$. To prove that $A_{0} A_{1} \ldots \in \operatorname{Traces}\left(T S^{\prime}\right)$ we have to show that there exists a path in $T S^{\prime}$, say $s_{0} s_{1} \ldots$, that generates this trace, i.e., $\operatorname{trace}\left(s_{0} s_{1} \ldots\right)=A_{0} A_{1} \ldots$

Any finite prefix $A_{0} A_{1} \ldots A_{m}$ of the infinite trace $A_{0} A_{1} \ldots$ is in $\operatorname{Traces}_{f i n}(T S)$, and as $\operatorname{Traces}_{f i n}(T S) \subseteq \operatorname{Traces}_{f i n}\left(T S^{\prime}\right)$, also in $\operatorname{Traces}_{f i n}\left(T S^{\prime}\right)$. Thus, for any natural number $m$, there exists a finite path $\pi^{m}=s_{0}^{m} s_{1}^{m} \ldots s_{m}^{m}$ in $T S^{\prime}$ such that

$$
\operatorname{trace}\left(\pi^{m}\right)=L\left(s_{0}^{m}\right) L\left(s_{1}^{m}\right) \ldots L\left(s_{m}^{m}\right)=A_{0} A_{1} \ldots A_{m}
$$

where $L$ denotes the labeling function of $T S^{\prime}$. Thus, $L\left(s_{j}^{m}\right)=A_{j}$ for all $0 \leqslant j \leqslant m$.
Although $A_{0} \ldots A_{m}$ is a prefix of $A_{0} \ldots A_{m+1}$, it is not guaranteed that path $\pi^{m}$ is a prefix of $\pi^{m+1}$. Due to the finiteness of $T S^{\prime}$, however, there is an infinite subsequence $\pi^{m_{0}} \pi^{m_{1}} \pi^{m_{2}} \ldots$ of $\pi^{0} \pi^{1} \pi^{2} \ldots$ such that $\pi^{m_{i}}$ and $\pi^{m_{i+1}}$ agree on the first $i$ states. Thus, $\pi^{m_{0}} \pi^{m_{1}} \pi^{m_{2}} \ldots$ induces an infinite path $\pi$ in $T S^{\prime}$ with the desired property.

This is formally proven using a so-called diagonalization technique. This goes as follows. Let $I_{0}, I_{1}, I_{2}, \ldots$ be an infinite series of infinite sets of indices (i.e., natural numbers) with $I_{n} \subseteq\{m \in \mathbb{N} \mid m \geqslant n\}$ and $s_{0}, s_{1}, \ldots$ be states in $T S^{\prime}$ such that for all natural numbers $n$ it holds that
(1) $n \geqslant 1$ implies $I_{n-1} \supseteq I_{n}$,
(2) $s_{0} s_{1} s_{2} \ldots s_{n}$ is an initial, finite path fragment in $T S^{\prime}$,
(3) for all $m \in I_{n}$ it holds that $s_{0} \ldots s_{n}=s_{0}^{m} \ldots s_{n}^{m}$.

The definition of the sets $I_{n}$ and states $s_{n}$ is by induction on $n$.
Base case ( $n=0$ ): As $\left\{s_{0}^{m} \mid m \in \mathbb{N}\right\}$ is finite (since it is a subset of the finite set of initial states of $T S^{\prime}$ ), there exists an initial state $s_{0}$ in $T S^{\prime}$ and an infinite index set $I_{0}$ such that $s_{0}=s_{0}^{m}$ for all $m \in I_{0}$.

Induction step $n \Longrightarrow n+1$. Assume that the index sets $I_{0}, \ldots, I_{n}$ and states $s_{0}, \ldots, s_{n}$ are defined. Since $T S^{\prime}$ is finite, $\operatorname{Post}\left(s_{n}\right)$ is finite. Furthermore, by the induction hypothesis $s_{n}=s_{n}^{m}$ for all $m \in I_{n}$, and thus

$$
\left\{s_{n+1}^{m} \mid m \in I_{n}, m \geqslant n+1\right\} \subseteq \operatorname{Post}\left(s_{n}\right) .
$$

Since $I_{n}$ is infinite, there exists an infinite subset $I_{n+1} \subseteq\left\{m \in I_{n} \mid m \geqslant n+1\right\}$ and a state $s_{n+1} \in \operatorname{Post}\left(s_{n}\right)$ such that $s_{n+1}^{m}=s_{n+1}$ for all $m \in \bar{I}_{n+1}$. It follows directly that the above properties (1) through (3) are fulfilled.

We now consider the state sequence $s_{0} s_{1} \ldots$ in $T S^{\prime}$. Obviously, this state sequence is a path in $T S^{\prime}$ satisfying trace $\left(s_{0} s_{1} \ldots\right)=A_{0} A_{1} \ldots$ Consequently, $A_{0} A_{1} \ldots \in \operatorname{Traces}\left(T S^{\prime}\right)$.

## Remark 3.31. Image-Finite Transition Systems

The result stated in Theorem 3.30 also holds under slightly weaker conditions: it suffices to require that $T S$ has no terminal states (as in Theorem 3.30) and that $T S^{\prime}$ is $A P$ imagefinite (rather than being finite).

Let $T S^{\prime}=(S, A c t, \rightarrow, I, A P, L)$. Then, $T S^{\prime}$ is called $A P$ image-finite (or briefly imagefinite) if
(i) for all $A \subseteq A P$, the set $\left\{s_{0} \in I \mid L\left(s_{0}\right)=A\right\}$ is finite and
(ii) for all states $s$ in $T S^{\prime}$ and all $A \subseteq A P$, the set of successors $\left\{s^{\prime} \in \operatorname{Post}(s) \mid L\left(s^{\prime}\right)=A\right\}$ is finite.

Thus, any finite transition system is image-finite. Moreover, any transition system that is $A P$-deterministic is image-finite. (Recall that $A P$-determinism requires $\left\{s_{0} \in I \mid L\left(s_{0}\right)=\right.$ $A\}$ and $\left\{s^{\prime} \in \operatorname{Post}(s) \mid L\left(s^{\prime}\right)=A\right\}$ to be either singletons or empty sets; see Definition 2.5 , page 24.)

In fact, a careful inspection of the proof of Theorem 3.30 shows that (i) and (ii) for $T S^{\prime}$ are used in the construction of the index sets $I_{n}$ and states $s_{n}$. Hence, we have $\operatorname{Traces}(T S) \subseteq \operatorname{Traces}\left(T S^{\prime}\right)$ iff $\operatorname{Traces}_{f i n}(T S) \subseteq \operatorname{Traces}_{f i n}\left(T S^{\prime}\right)$, provided $T S$ has no terminal states and $T S^{\prime}$ is image-finite.

Trace and finite trace inclusion, however, coincide neither for infinite transition systems nor for finite ones which have terminal states.

## Example 3.32. Finite vs. Infinite Transition System

Consider the transition systems sketched in Figure 3.10, where $b$ stands for an atomic proposition. Transition system $T S$ (on the left) is finite, whereas $T S^{\prime}$ (depicted on the right) is infinite and not image-finite, because of the infinite branching in the initial state. It is not difficult to observe that

$$
\operatorname{Traces}(T S) \nsubseteq \operatorname{Traces}\left(T S^{\prime}\right) \quad \text { and } \quad \operatorname{Traces}_{f i n}(T S) \subseteq \operatorname{Traces}_{f i n}\left(T S^{\prime}\right)
$$

This stems from the fact that TS can take the self-loop infinitely often and never reaches a $b$-state, whereas $T S^{\prime}$ does not exhibit such behavior. Moreover, any finite trace of $T S$ is
of the form $(\varnothing)^{n}$ for $n \geqslant 0$ and is also a finite trace of $T S^{\prime}$. Consequently, LT properties of $T S^{\prime}$ do not carry over to $T S$ (and those of $T S$ may not hold for $T S^{\prime}$ ). For example, the LT property "eventually $b$ " holds for $T S^{\prime}$, but not for $T S$. Similarly, the LT property "never $b "$ holds for $T S$, but not for $T S^{\prime}$.

Although these transition systems might seem rather artificial, this is not the case: TS could result from an infinite loop in a program, whereas $T S^{\prime}$ could model the semantics of a program fragment that nondeterministically chooses a natural number $k$ and then performs $k$ steps.


Figure 3.10: Distinguishing trace inclusion from finite trace inclusion.

### 3.4 Liveness Properties

Informally speaking, safety properties specify that "something bad never happens". For the mutual exclusion algorithm, the "bad" thing is that more than one process is in its critical section, while for the traffic light the "bad" situation is whenever a red light phase is not preceded by a yellow light phase. An algorithm can easily fulfill a safety property by simply doing nothing as this will never lead to a "bad" situation. As this is usually undesired, safety properties are complemented by properties that require some progress. Such properties are called "liveness" properties (or sometimes "progress" properties). Intuitively, they state that "something good" will happen in the future. Whereas safety properties are violated in finite time, i.e., by a finite system run, liveness properties are violated in infinite time, i.e., by infinite system runs.

### 3.4.1 Liveness Properties

Several (nonequivalent) notions of liveness properties have been defined in the literature. We follow here the approach of Alpern and Schneider [5, 6, 7]. They provided a formal notion of liveness properties which relies on the view that liveness properties do not constrain the finite behaviors, but require a certain condition on the infinite behaviors. A typical example for a liveness property is the requirement that certain events occur infinitely often. In this sense, the "good event" of a liveness property is a condition on the infinite behaviors, while the "bad event" for a safety property occurs in a finite amount of time, if it occurs at all.

In our approach, a liveness property (over AP) is defined as an LT property that does not rule out any prefix. This entails that the set of finite traces of a system are of no use at all to decide whether a liveness property holds or not. Intuitively speaking, it means that any finite prefix can be extended such that the resulting infinite trace satisfies the liveness property under consideration. This is in contrast to safety properties where it suffices to have one finite trace (the "bad prefix") to conclude that a safety property is refuted.

## Definition 3.33. Liveness Property

LT property $P_{\text {live }}$ over AP is a liveness property whenever $\operatorname{pref}\left(P_{\text {live }}\right)=\left(2^{A P}\right)^{*}$.

Thus, a liveness property (over $A P$ ) is an LT property $P$ such that each finite word can be extended to an infinite word that satisfies $P$. Stated differently, $P$ is a liveness property if and only if for all finite words $w \in\left(2^{A P}\right)^{*}$ there exists an infinite word $\sigma \in\left(2^{A P}\right)^{\omega}$ satisfying $w \sigma \in P$.

## Example 3.34. Repeated Eventually and Starvation Freedom

In the context of mutual exclusion algorithms the natural safety property that is required ensures the mutual exclusion property stating that the processes are never simultaneously in their critical sections. (This is even an invariant.) Typical liveness properties that are desired assert that

- (eventually) each process will eventually enter its critical section;
- (repeated eventually) each process will enter its critical section infinitely often;
- (starvation freedom) each waiting process will eventually enter its critical section.

Let's see how these liveness properties are formalized as LT properties and let us check that
they are liveness properties. As in Example 3.14, we will deal with the atomic propositions wait ${ }_{1}$, crit ${ }_{1}$, wait $_{2}$, crit $t_{2}$ where wait ${ }_{i}$ characterizes the states where process $P_{i}$ has requested access to its critical section and is in its waiting phase, while crit ${ }_{i}$ serves as a label for the states where $P_{i}$ has entered its critical section. We now formalize the three properties by LT properties over $A P=\left\{\right.$ wait $_{1}$, crit $_{1}$, wait ${ }_{2}$, crit $\left.{ }_{2}\right\}$. The first property (eventually) consists of all infinite words $A_{0} A_{1} \ldots$ with $A_{j} \subseteq A P$ such that

$$
\left(\exists j \geqslant 0 . \text { crit }_{1} \in A_{j}\right) \wedge\left(\exists j \geqslant 0 . \text { crit }_{2} \in A_{j}\right)
$$

which requires that $P_{1}$ and $P_{2}$ are in their critical sections at least once. The second property (repeated eventually) poses the condition

$$
\left(\forall k \geqslant 0 . \exists j \geqslant k . c r i t_{1} \in A_{j}\right) \wedge\left(\forall k \geqslant 0 . \exists j \geqslant k . \text { crit }_{2} \in A_{j}\right)
$$

stating that $P_{1}$ and $P_{2}$ are infinitely often in their critical sections. This formula is often abbreviated by

$$
\left(\begin{array}{l}
\neq \\
\exists
\end{array} \geqslant 0 . \text { crit }_{1} \in A_{j}\right) \wedge\left(\begin{array}{l}
\nexists \\
\exists
\end{array} \geqslant 0 . \text { crit }_{2} \in A_{j}\right) .
$$

The third property (starvation freedom) requires that

$$
\begin{aligned}
& \forall j \geqslant 0 .\left(\text { wait }_{1} \in A_{j} \Rightarrow\left(\exists k>j . \text { crit }_{1} \in A_{k}\right)\right) \wedge \\
& \forall j \geqslant 0 .\left(\text { wait }_{2} \in A_{j} \Rightarrow\left(\exists k>j . \text { crit }_{2} \in A_{k}\right)\right) .
\end{aligned}
$$

It expresses that each process that is waiting will acquire access to the critical section at some later time point. Note that here we implicitly assume that a process that starts waiting to acquire access to the critical section does not "give up" waiting, i.e., it continues waiting until it is granted access.

All aforementioned properties are liveness properties, as any finite word over $A P$ is a prefix of an infinite word where the corresponding condition holds. For instance, for starvation freedom, a finite trace in which a process is waiting but never acquires access to its critical section can always be extended to an infinite trace that satisfies the starvation freedom property (by, e.g., providing access in an strictly alternating fashion from a certain point on).

### 3.4.2 Safety vs. Liveness Properties

This section studies the relationship between liveness and safety properties. In particular, it provides answers to the following questions:

- Are safety and liveness properties disjoint?, and
- Is any linear-time property a safety or liveness property?

As we will see, the first question will be answered affirmatively while the second question will result in a negative answer. Interestingly enough, though, for any LT property $P$ an equivalent LT property $P^{\prime}$ does exist which is a combination (i.e., intersection) of a safety and a liveness property. All in all, one could say that the identification of safety and liveness properties thus provides an essential characterization of linear-time properties.

The first result states that safety and liveness properties are indeed almost disjoint. More precisely, it states that the only property that is both a safety and a liveness property is nonrestrictive, i.e., allows all possible behaviors. Logically speaking, this is the equivalent of "true".

## Lemma 3.35. Intersection of Safety and Liveness Properties

The only LT property over AP that is both a safety and a liveness property is $\left(2^{A P}\right)^{\omega}$.

Proof: Assume $P$ is a liveness property over $A P$. By definition, $\operatorname{pref}(P)=\left(2^{A P}\right)^{*}$. It follows that closure $(P)=\left(2^{A P}\right)^{\omega}$. If $P$ is a safety property too, $\operatorname{closure}(P)=P$, and hence $P=\left(2^{A P}\right)^{\omega}$.

Recall that the closure of property $P$ (over $A P$ ) is the set of infinite words (over $2^{A P}$ ) for which all prefixes are also prefixes of $P$. In order to show that an LT property can be considered as a conjunction of a liveness and a safety property, the following result is helpful. It states that the closure of the union of two properties equals the union of their closures.

## Lemma 3.36. Distributivity of Union over Closure

For any LT properties $P$ and $P^{\prime}$ :

$$
\operatorname{closure}(P) \cup \operatorname{closure}\left(P^{\prime}\right)=\operatorname{closure}\left(P \cup P^{\prime}\right)
$$

Proof: $\subseteq:$ As $P \subseteq P^{\prime}$ implies closure $(P) \subseteq$ closure $\left(P^{\prime}\right)$, we have $P \subseteq P \cup P^{\prime}$ implies closure $(P) \subseteq \operatorname{closure}\left(P \cup P^{\prime}\right)$. In a similar way it follows that closure $\left(P^{\prime}\right) \subseteq \operatorname{closure}\left(P \cup P^{\prime}\right)$. Thus, closure $(P) \cup \operatorname{closure}\left(P^{\prime}\right) \subseteq \operatorname{closure}\left(P \cup P^{\prime}\right)$.
$\supseteq$ : Let $\sigma \in \operatorname{closure}\left(P \cup P^{\prime}\right)$. By definition of closure, $\operatorname{pref}(\sigma) \subseteq \operatorname{pref}\left(P \cup P^{\prime}\right)$. As $\operatorname{pref}(P \cup$ $\left.P^{\prime}\right)=\operatorname{pref}(P) \cup \operatorname{pref}\left(P^{\prime}\right)$, any finite prefix of $\sigma$ is in $\operatorname{pref}(P)$ or in $\operatorname{pref}\left(P^{\prime}\right)$ (or in both).

As $\sigma \in\left(2^{A P}\right)^{\omega}$, $\sigma$ has infinitely many prefixes. Thus, infinitely many finite prefixes of $\sigma$ belong to $\operatorname{pref}(P)$ or to $\operatorname{pref}\left(P^{\prime}\right)$ (or to both). W.l.o.g., assume $\operatorname{pref}(\sigma) \cap \operatorname{pref}(P)$ to be infinite. Then $\operatorname{pref}(\sigma) \subseteq \operatorname{pref}(P)$, which yields $\sigma \in \operatorname{closure}(P)$, and thus $\sigma \in$ $\operatorname{closure}(P) \cup \operatorname{closure}\left(P^{\prime}\right)$. The fact that $\operatorname{pref}(\sigma) \subseteq \operatorname{pref}(P)$ can be shown by contraposition. Assume $\widehat{\sigma} \in \operatorname{pref}(\sigma) \backslash \operatorname{pref}(P)$. Let $|\widehat{\sigma}|=k$. As $\operatorname{pref}(\sigma) \cap \operatorname{pref}(P)$ is infinite, there exists $\widehat{\sigma^{\prime}} \in \operatorname{pref}(\sigma) \cap \operatorname{pref}(P)$ with length larger than $k$. But then, there exists $\sigma^{\prime} \in P$ with $\widehat{\sigma^{\prime}} \in \operatorname{pref}\left(\sigma^{\prime}\right)$. It then follows that $\widehat{\sigma} \in \operatorname{pref}\left(\widehat{\sigma^{\prime}}\right)$ (as both $\widehat{\sigma}$ and $\widehat{\sigma^{\prime}}$ are prefixes of $\sigma$ ) and as $\operatorname{pref}\left(\widehat{\sigma^{\prime}}\right) \subseteq \operatorname{pref}(P)$, it follows that $\widehat{\sigma} \in \operatorname{pref}(P)$. This contradicts $\widehat{\sigma} \in \operatorname{pref}(\sigma) \backslash \operatorname{pref}(P)$.

Consider the beverage vending machine of Figure 3.5 (on page 97 ), and the following property:
"the machine provides beer infinitely often after initially providing soda three times in a row"

In fact, this property consists of two parts. On the one hand, it requires beer to be provided infinitely often. As any finite trace can be extended to an infinite trace that enjoys this property it is a liveness property. On the other hand, the first three drinks it provides should all be soda. This is a safety property, since any finite trace in which one of the first three drinks provided is beer violates it. The property is thus a combination (in fact, a conjunction) of a safety and a liveness property. The following result shows that every LT property can be decomposed in this way.

## Theorem 3.37. Decomposition Theorem

For any LT property $P$ over AP there exists a safety property $P_{\text {safe }}$ and a liveness property $P_{\text {live }}$ (both over AP) such that

$$
P=P_{\text {safe }} \cap P_{\text {live }}
$$

Proof: Let $P$ be an LT property over $A P$. It is easy to see that $P \subseteq \operatorname{closure}(P)$. Thus: $P=\operatorname{closure}(P) \cap P$, which by set calculus can be rewritten into:

$$
P=\underbrace{\operatorname{closure}(P)}_{=P_{\text {safe }}} \cap \underbrace{\left(P \cup\left(\left(2^{A P}\right)^{\omega} \backslash \operatorname{closure}(P)\right)\right)}_{=P_{\text {live }}}
$$

By definition, $P_{\text {safe }}=\operatorname{closure}(P)$ is a safety property. It remains to prove that $P_{\text {live }}=$ $P \cup\left(\left(2^{A P}\right)^{\omega} \backslash \operatorname{closure}(P)\right)$ is a liveness property. By definition, $P_{\text {live }}$ is a liveness property whenever $\operatorname{pref}\left(P_{\text {live }}\right)=\left(2^{A P}\right)^{*}$. This is equivalent to closure $\left(P_{\text {live }}\right)=\left(2^{A P}\right)^{\omega}$. As for any

LT property $P$, closure $(P) \subseteq\left(2^{A P}\right)^{\omega}$ holds true, it suffices to showing that $\left(2^{A P}\right)^{\omega} \subseteq$ closure $\left(P_{\text {live }}\right)$. This goes as follows:

$$
\begin{array}{rll}
\operatorname{closure}\left(P_{\text {live }}\right) & = & \operatorname{closure}\left(P \cup\left(\left(2^{A P}\right)^{\omega} \backslash \operatorname{closure}(P)\right)\right) \\
\stackrel{\text { Lemma }}{=}{ }^{3.36} & & \operatorname{closure}(P) \cup \operatorname{closure}\left(\left(2^{A P}\right)^{\omega} \backslash \operatorname{closure}(P)\right) \\
& \supseteq & \operatorname{closure}(P) \cup\left(\left(2^{A P}\right)^{\omega} \backslash \operatorname{closure}(P)\right) \\
& =\left(2^{A P}\right)^{\omega}
\end{array}
$$

where in the one-but-last step in the derivation, we exploit the fact that closure $\left(P^{\prime}\right) \supseteq P^{\prime}$ for all LT properties $P^{\prime}$.

The proof of Theorem 3.37 shows that $P_{\text {safe }}=\operatorname{closure}(P)$ is a safety property and $P_{\text {live }}=P \cup\left(\left(2^{A P}\right)^{\omega} \backslash\right.$ closure $\left.(P)\right)$ a liveness property with $P=P_{\text {safe }} \cap P_{\text {live }}$. In fact, this decomposition is the "sharpest" one for $P$ since $P_{\text {safe }}$ is the strongest safety property and $P_{\text {live }}$ the weakest liveness property that can serve for a decomposition of $P$ :

## Lemma 3.38. Sharpest Decomposition

Let $P$ be an LT property and $P=P_{\text {safe }} \cap P_{\text {live }}$ where $P_{\text {safe }}$ is a safety property and $P_{\text {live }}$ a liveness property. We then have

1. closure $(P) \subseteq P_{\text {safe }}$,
2. $P_{\text {live }} \subseteq P \cup\left(\left(2^{A P}\right)^{\omega} \backslash \operatorname{closure}(P)\right)$.

Proof: See Exercise 3.12, page 147.

A summary of the classification of LT properties is depicted as a Venn diagram in Figure 3.11. The circle denotes the set of all LT properties over a given set of atomic propositions.

## Remark 3.39. Topological Characterizations of Safety and Liveness

Let us conclude this section with a remark for readers who are familiar with basic notions of topological spaces. The set $\left(2^{A P}\right)^{\omega}$ can be equipped with the distance function given by $d\left(\sigma_{1}, \sigma_{2}\right)=1 / 2^{n}$ if $\sigma_{1}, \sigma_{2}$ are two distinct infinite words $\sigma_{1}=A_{1} A_{2} \ldots$ and $\sigma_{2}=B_{1} B_{2} \ldots$ and $n$ is the length of the longest common prefix. Moreover, we put $d(\sigma, \sigma)=0$. Then, $d$ is a metric on $\left(2^{A P}\right)^{\omega}$, and hence induces a topology on $\left(2^{A P}\right)^{\omega}$. Under this topology,


Figure 3.11: Classification of linear-time properties.
the safety properties are exactly the closed sets, while the liveness properties agree with the dense sets. In fact, closure $(P)$ is the topological closure of $P$, i.e., the smallest closed set that contains $P$. The result stated in Theorem 3.37 then follows from the well-known fact that any subset of a topological space (of the kind described above) can be written as the intersection of its closure and a dense set.

### 3.5 Fairness

An important aspect of reactive systems is fairness. Fairness assumptions rule out infinite behaviors that are considered unrealistic, and are often necessary to establish liveness properties. We illustrate the concept of fairness by means of a frequently encountered problem in concurrent systems.

## Example 3.40. Process Fairness

Consider $N$ processes $P_{1}, \ldots, P_{N}$ which require a certain service. There is one server process Server that is expected to provide services to these processes. A possible strategy that Server can realize is the following. Check the processes starting with $P_{1}$, then $P_{2}$, and so on, and serve the first thus encountered process that requires service. On finishing serving this process, repeat this selection procedure once again starting with checking $P_{1}$.

Now suppose that $P_{1}$ is continuously requesting service. Then this strategy will result in Server always serving $P_{1}$. Since in this way another process has to wait infinitely long before being served, this is called an unfair strategy. In a fair serving strategy it is required that the server eventually responds to any request by any one of the processes. For instance, a round-robin scheduling strategy where each process is only served for a limited amount of time is a fair strategy: after having served one process, the next (in the round-robin order) is checked and, if needed, served.

When verifying concurrent systems one is often only interested in paths in which enabled transitions (statements) are executed in some "fair" manner. Consider, for instance, a mutual exclusion algorithm for two processes. In order to prove starvation freedom, the situation in which a process that wants to enter its critical section has to wait infinitely long, we want to exclude those paths in which the competitor process is always being selected for execution. This type of fairness is also known as process fairness, since it concerns the fair scheduling of the execution of processes. If we were to consider unfair paths when proving starvation freedom, we would usually fail, since there always exists an unfair strategy according to which some process is always neglected, and thus can never make progress. One might argue that such unfair strategy is unrealistic and should be avoided.

## Example 3.41. Starvation Freedom

Consider the transition systems $T S_{S e m}$ and $T S_{\text {Pet }}$ for the semaphore-based mutual exclusion algorithms (see Example 2.24 on page 43) and Peterson's algorithm. The starvation freedom property
"Once access is requested, a process does not have to wait infinitely long before acquiring access to its critical section"
is violated by transition system $T S_{\text {Sem }}$ while it permits only one of the processes to proceed, while the other process is starving (or only acquiring access to the critical section finitely often). The transition system $T S_{P e t}$ for Peterson's algorithm, however, fulfills this property.

The property
"Each of the processes is infinitely often in its critical section"
is violated by both transition systems as none of them excludes the fact that a process would never (or only finitely often) request to enter the critical section.

Process fairness is a particular form of fairness. In general, fairness assumptions are needed to prove liveness or other properties stating that the system makes some progress ("something good will eventually happen"). This is of vital importance if the transition system to be checked contains nondeterminism. Fairness is then concerned with resolving nondeterminism in such a way that it is not biased to consistently ignore a possible option. In the above example, the scheduling of processes is nondeterministic: the choice of the next process to be executed (if there are at least two processes that can be potentially selected) is arbitrary. Another prominent example where fairness is used to "resolve" nondeterminism is in modeling concurrent processes by means of interleaving. Interleaving is equivalent to modeling the concurrent execution of two independent processes by enumerating all the possible orders in which activities of the processes can be executed (see Chapter 2).

## Example 3.42. Independent Traffic Lights

Consider the transition system

$$
T S=\operatorname{Tr}^{L_{i g h t}^{1}}| || | r \operatorname{Tight}_{2}
$$

for the two independent traffic lights described in Example 2.17 (page 36). The liveness property

> "Both traffic lights are infinitely often green"
is not satisfied, since

$$
\left\{\text { red }_{1}, \operatorname{red}_{2}\right\}\left\{\text { green }_{1}, \text { red }_{2}\right\}\left\{\text { red }_{1}, \operatorname{red}_{2}\right\}\left\{\text { green }_{1}, \operatorname{red}_{2}\right\} \ldots
$$

is a trace of $T S$ where only the first traffic light is infinitely often green.

What is wrong with the above examples? In fact, nothing. Let us explain this. In the traffic light example, the information whether each traffic light switches color infinitely often is lost by means of interleaving. The trace in which only the first traffic light is acting while the second light seems to be completely stopped is formally a trace of the transition system $\operatorname{TrLight} t_{1}| | \operatorname{TrLight}_{2}$. However, it does not represent a realistic behavior as in practice no traffic light is infinitely faster than another.

For the semaphore-based mutual exclusion algorithm, the difficulty is the degree of abstraction. A semaphore is not a willful individual that arbitrarily chooses a process which is authorized to enter its critical section. Instead, the waiting processes are administered in a queue (or another "fair" medium). The required liveness can be proven in one of the following refinement steps, in which the specification of the behavior of the semaphore is sufficiently detailed.

### 3.5.1 Fairness Constraints

The above considerations show that we - to obtain a realistic picture of the behavior of a parallel system modeled by a transition system - need a more alleviated form of satisfaction relation for LT properties, which implies an "adequate" resolution of the nondeterministic decisions in a transition system. In order to rule out the unrealistic computations, fairness constraints are imposed.

In general, a fair execution (or trace) is characterized by the fact that certain fairness constraints are fulfilled. Fairness constraints are used to rule out computations that are considered to be unreasonable for the system under consideration. Fairness constraints come in different flavors:

- Unconditional fairness: e.g., "Every process gets its turn infinitely often."
- Strong fairness: e.g., "Every process that is enabled infinitely often gets its turn infinitely often."
- Weak fairness: e.g., "Every process that is continuously enabled from a certain time instant on gets its turn infinitely often."

Here, the term "is enabled" has to be understood in the sense of "ready to execute (a transition)". Similarly, "gets its turn" stands for the execution of an arbitrary transition. This can, for example, be a noncritical action, acquiring a shared resource, an action in the critical section, or a communication action.

An execution fragment is unconditionally fair with respect to, e.g., "a process enters its critical section" or "a process gets its turn", if these properties hold infinitely often. That is to say, a process enters its critical section infinitely often, or, in the second example, a process gets its turn infinitely often. Note that no condition (such as "a process is enabled") is expressed that constrains the circumstances under which a process gets its turn infinitely often. Unconditional fairness is sometimes referred to as impartiality.

Strong fairness means that if an activity is infinitely often enabled-but not necessarily always, i.e., there may be finite periods during which the activity is not enabled-then it will be executed infinitely often. An execution fragment is strongly fair with respect to activity $\alpha$ if it is not the case that $\alpha$ is infinitely often enabled without being taken beyond a certain point. Strong fairness is sometimes referred to as compassion.

Weak fairness means that if an activity, e.g., a transition in a process or an entire process itself, is continuously enabled-no periods are allowed in which the activity is not
enabled - then it has to be executed infinitely often. An execution fragment is weakly fair with respect to some activity, $\alpha$ say, if it is not the case that $\alpha$ is always enabled beyond some point without being taken beyond this point. Weak fairness is sometimes referred to as justice.

How to express these fairness constraints? There are different ways to formulate fairness requirements. In the sequel, we adopt the action-based view and define strong fairness for (sets of) actions. (In Chapter 5, also state-based notions of fairness will be introduced and the relationship between action-based and state-based fairness is studied in detail.) Let $A$ be a set of actions. The execution fragment $\rho$ is said to be strongly $A$-fair if the actions in $A$ are not continuously ignored under the circumstance that they can be executed infinitely often. $\rho$ is unconditionally $A$-fair if some action in $A$ is infinitely often executed in $\rho$. Weak fairness is defined in a similar way as strong fairness (see below).

In order to formulate these fairness notions formally, the following auxiliary notion is convenient. For state $s$, let $\operatorname{Act}(s)$ denote the set of actions that are executable in state $s$, that is,

$$
\operatorname{Act}(s)=\left\{\alpha \in \operatorname{Act} \mid \exists s^{\prime} \in S . s \xrightarrow{\alpha} s^{\prime}\right\} .
$$

## Definition 3.43. Unconditional, Strong, and Weak Fairness

For transition system $T S=(S, A c t, \rightarrow, I, A P, L)$ without terminal states, $A \subseteq A c t$, and infinite execution fragment $\rho=s_{0} \xrightarrow{\alpha_{0}} s_{1} \xrightarrow{\alpha_{1}} \ldots$ of $T S$ :

1. $\rho$ is unconditionally $A$-fair whenever $\stackrel{\infty}{\exists} j . \alpha_{j} \in A$.
2. $\rho$ is strongly $A$-fair whenever

$$
\left(\begin{array}{l}
\infty \\
\left.\exists j . \operatorname{Act}\left(s_{j}\right) \cap A \neq \varnothing\right)
\end{array} \quad\left(\begin{array}{l}
\infty \\
\exists
\end{array} \cdot \alpha_{j} \in A\right) .\right.
$$

3. $\rho$ is weakly A-fair whenever

$$
\left(\forall j . \operatorname{Act}\left(s_{j}\right) \cap A \neq \varnothing\right) \Longrightarrow\left(\begin{array}{l}
\infty \\
\exists
\end{array} . \alpha_{j} \in A\right)
$$

Here, $\stackrel{\infty}{\exists} j$ stands for "there are infinitely many $j$ " and $\stackrel{\infty}{\forall} j$ for "for nearly all $j$ " in the sense of "for all, except for finitely many $j$ ". The variable $j$, of course, ranges over the natural numbers.

To check whether a run is unconditionally $A$-fair it suffices to consider the actions that occur along the execution, i.e., it is not necessary to check which actions in $A$ are enabled in visited states. However, in order to decide whether a given execution is strongly or weakly $A$-fair, it does not suffice to only consider the actions actually occurring in the execution. Instead, also the enabled actions in all visited states need to be considered. These enabled actions are possible in the visited states, but do not necessarily have to be taken along the considered execution.

## Example 3.44. A Simple Shared-Variable Concurrent Program

Consider the following two processes that run in parallel and share an integer variable $x$ that initially has value 0 :

$$
\begin{aligned}
\text { proc Inc } & =\text { while }\langle x \geqslant 0 \text { do } x:=x+1\rangle \text { od } \\
\text { proc Reset } & =x:=-1
\end{aligned}
$$

The pair of brackets $\langle\ldots\rangle$ embraces an atomic section, i.e., process Inc performs the check whether $x$ is positive and the increment of $x$ (if the guard holds) as one atomic step. Does this parallel program terminate? When no fairness constraints are imposed, it is possible that process Inc is permanently executing, i.e., process Reset never gets its turn, and the assignment $x=-1$ is not executed. In this case, termination is thus not guaranteed, and the property is refuted. If, however, we require unconditional process fairness, then every process gets its turn, and termination is guaranteed.

An important question now is: given a verification problem, which fairness notion to use? Unfortunately, there is no clear answer to this question. Different forms of fairness do exist - the above is just a small, though important, fragment of all possible fairness notions - and there is no single favorite notion. For verification purposes, fairness constraints are crucial, though. Recall that the purpose of fairness constraints is to rule out certain "unreasonable" computations. If the fairness constraint is too strong, relevant computations may not be considered. In case a property is satisfied (for a transition system), it might well be the case that some reasonable computation that is not considered (as it is ruled out by the fairness constraint) refutes this property. On the other hand, if the fairness constraint is too weak, we may fail to prove a certain property as some unreasonable computations (that are not ruled out) refute it.

The relationship between the different fairness notions is as follows. Each unconditionally $A$-fair execution fragment is strongly $A$-fair, and each strongly $A$-fair execution fragment is weakly $A$-fair. In general, the reverse direction does not hold. For instance, an execution fragment that solely visits states in which no $A$-actions are possible is strongly $A$-fair (as the premise of strong $A$-fairness does not hold), but not unconditionally $A$-fair. Besides,
an execution fragment that only visits finitely many states in which some $A$-actions are enabled but never executes an $A$-action is weakly $A$-fair (as the premise of weak $A$-fairness does not hold), but not strongly $A$-fair. Summarizing, we have

$$
\text { unconditional } A \text {-fairness } \Longrightarrow \text { strong } A \text {-fairness } \Longrightarrow \text { weak } A \text {-fairness }
$$

where the reverse implication in general does not hold.

## Example 3.45. Fair Execution Fragments

Consider the transition system $T S_{\text {Sem }}$ for the semaphore-based mutual exclusion solution. We label the transitions with the actions req , enter $_{i}$ (for $i=1,2$ ), and rel in the obvious way, see Figure 3.12.

In the execution fragment

$$
\left\langle n_{1}, n_{2}, y=1\right\rangle \xrightarrow{\text { req }}\left\langle w_{1}, n_{2}, y=1\right\rangle \xrightarrow{\text { enter }_{1}}\left\langle c_{1}, n_{2}, y=0\right\rangle \xrightarrow{\text { rel }}\left\langle n_{1}, n_{2}, y=1\right\rangle \xrightarrow{r e q_{1}} \ldots
$$

only the first process gets its turn. This execution fragment is indicated by the dashed arrows in Figure 3.12. It is not unconditionally fair for the set of actions

$$
A=\left\{\text { enter }_{2}\right\} .
$$

It is, however, strongly $A$-fair, since no state is visited in which the action enter ${ }_{2}$ is executable, and hence the premise of strong fairness is vacuously false. In the alternative execution fragment

$$
\begin{gathered}
\left\langle n_{1}, n_{2}, y=1\right\rangle \xrightarrow{r e q_{2}}\left\langle n_{1}, w_{2}, y=1\right\rangle \xrightarrow{r e q_{1}}\left\langle w_{1}, w_{2}, y=1\right\rangle \xrightarrow{\text { enter }_{1}} \\
\left\langle c_{1}, w_{2}, y=0\right\rangle \xrightarrow{r e l}\left\langle n_{1}, w_{2}, y=1\right\rangle \xrightarrow{r e q_{1}} \ldots
\end{gathered}
$$

the second process requests to enter its critical section but is ignored forever. This execution fragment is indicated by the dotted arrows in Figure 3.12. It is not strongly $A$-fair: although the action enter ${ }_{2}$ is infinitely often enabled (viz. every time when visiting the state $\left\langle w_{1}, w_{2}, y=1\right\rangle$ or $\left\langle n_{1}, w_{2}, y=1\right\rangle$ ), it is never taken. It is, however, weakly $A$-fair, since the action enter ${ }_{2}$ is not continuously enabled-it is not enabled in the state $\left\langle c_{1}, w_{2}, y=0\right\rangle$.

A fairness constraint imposes a requirement on all actions in a set $A$. In order to enable different fairness constraints to be imposed on different, possibly nondisjoint, sets of actions, fairness assumptions are used. A fairness assumption for a transition system may require different notions of fairness with respect to several sets of actions.


Figure 3.12: Two examples of fair execution fragments of the semaphore-based mutual exclusion algorithm.

## Definition 3.46. Fairness Assumption

A fairness assumption for Act is a triple

$$
\mathcal{F}=\left(\mathcal{F}_{\text {ucond }}, \mathcal{F}_{\text {strong }}, \mathcal{F}_{\text {weak }}\right)
$$

with $\mathcal{F}_{\text {ucond }}, \mathcal{F}_{\text {strong }}, \mathcal{F}_{\text {weak }} \subseteq 2^{\text {Act }}$. Execution $\rho$ is $\mathcal{F}$-fair if

- it is unconditionally $A$-fair for all $A \in \mathcal{F}_{\text {ucond }}$,
- it is strongly $A$-fair for all $A \in \mathcal{F}_{\text {strong }}$, and
- it is weakly $A$-fair for all $A \in \mathcal{F}_{\text {weak }}$.

If the set $\mathcal{F}$ is clear from the context, we use the term fair instead of $\mathcal{F}$-fair.

Intuitively speaking, a fairness assumption is a triple of sets of (typically different) action sets, one such set of action sets is treated in a strongly fair manner, one in a weakly fair manner, and one in an unconditionally fair way. This is a rather general definition that allows imposing different fairness constraints on different sets of actions. Quite often, only a single type of fairness constraint suffices. In the sequel, we use the casual notations for these fairness assumptions. For $\mathcal{F} \subseteq 2^{A c t}$, a strong fairness assumption denotes the fairness assumption $(\varnothing, \mathcal{F}, \varnothing)$. Weak, and unconditional fairness assumptions are used in a similar way.

The notion of $\mathcal{F}$-fairness as defined on execution fragments is lifted to traces and paths in the obvious way. An infinite trace $\sigma$ is $\mathcal{F}$-fair if there is an $\mathcal{F}$-fair execution $\rho$ with $\operatorname{trace}(\rho)=\sigma . \mathcal{F}$-fair (infinite) path fragments and $\mathcal{F}$-fair paths are defined analogously.

Let FairPaths $\mathcal{F}_{\mathcal{F}}(s)$ denote the set of $\mathcal{F}$-paths of $s$ (i.e., infinite $\mathcal{F}$-fair path fragments that start in state $s$ ), and FairPaths $\mathcal{F}(T S)$ the set of $\mathcal{F}$-fair paths that start in some initial state of TS. Let FairTraces $\mathcal{F}(s)$ denote the set of $\mathcal{F}$-fair traces of $s$, and FairTraces $\mathcal{F}(T S)$ the set of $\mathcal{F}$-fair traces of the initial states of transition system TS:

$$
\begin{aligned}
\operatorname{FairTraces}_{\mathcal{F}}(s) & =\operatorname{trace}\left(\text { FairPaths }_{\mathcal{F}}(s)\right) \quad \text { and } \\
\operatorname{FairTraces}_{\mathcal{F}}(T S) & =\bigcup_{s \in I} \text { Fair }_{\operatorname{Traces}}^{\mathcal{F}}(s)
\end{aligned}
$$

Note that it does not make much sense to define these notions for finite traces as any finite trace is fair by default.

## Example 3.47. Mutual Exclusion Again

Consider the following fairness requirement for two-process mutual exclusion algorithms:
"process $P_{i}$ acquires access to its critical section infinitely often"
for any $i \in\{1,2\}$. What kind of fairness assumption is appropriate to achieve this? Assume each process $P_{i}$ has three states $n_{i}$ (noncritical), $w_{i}$ (waiting), and $c_{i}$ (critical). As before, the actions $r e q_{i}$, enter ${ }_{i}$, and rel are used to model the request to enter the critical section, the entering itself, and the release of the critical section. The strongfairness assumption

$$
\left\{\left\{\text { enter }_{1}, \text { enter }_{2}\right\}\right\}
$$

ensures that one of the actions enter ${ }_{1}$ or enter 2 , is executed infinitely often. A behavior in which one of the processes gets access to the critical section infinitely often while the other gets access only finitely many times is strongly fair with respect to this assumption. This is, however, not intended. The strong-fairness assumption

$$
\left\{\left\{\text { enter }_{1}\right\},\left\{\text { enter }_{2}\right\}\right\}
$$

indeed realizes the above requirement. This assumption should be viewed as a requirement on how to resolve the contention when both processes are awaiting to get access to the critical section.

Fairness assumptions can be verifiable properties whenever all infinite execution fragments are fair. For example, it can be verified that the transition system for Peterson's algorithm
satisfies the strong-fairness assumption

$$
\mathcal{F}_{\text {strong }}=\left\{\left\{\text { enter }_{1}\right\},\left\{\text { enter }_{2}\right\}\right\} .
$$

But in many cases it is necessary to assume the validity of the fairness conditions to verify liveness properties.

A transition system TS satisfies the LT property $P$ under fairness assumption $\mathcal{F}$ if all $\mathcal{F}$-fair paths fulfill the property $P$. However, no requirements whatsoever are imposed on the unfair paths. This is formalized as follows.

## Definition 3.48. Fair Satisfaction Relation for LT Properties

Let $P$ be an LT property over $A P$ and $\mathcal{F}$ a fairness assumption over Act. Transition system $T S=(S, A c t, \rightarrow, I, A P, L)$ fairly satisfies $P$, notation $T S \models_{\mathcal{F}} P$, if and only if FairTraces $\mathcal{F}(T S) \subseteq P$.

For a transition system that satisfies the fairness assumption $\mathcal{F}$ (i.e., all paths are $\mathcal{F}$-fair), the satisfaction relation $\models$ without fairness assumptions (see Definition 3.11, page 100) corresponds with the fair satisfaction relation $\models_{\mathcal{F}}$. In this case, the fairness assumption does not rule out any trace. However, in case a transition system has traces that are not $\mathcal{F}$-fair, then in general we are confronted with a situation

$$
T S \models_{\mathcal{F}} P \quad \text { whereas } \quad T S \not \vDash P .
$$

By restricting the validity of a property to the set of fair paths, the verification can be restricted to "realistic" executions.

Before turning to some examples, a few words on the relationship between unconditional, strong, and weak fairness are (again) in order. As indicated before, we have that the set of unconditional $A$-fair executions is a subset of all strong $A$-fair executions. In a similar way, the latter set of executions is a subset of all weak $A$-fair executions. Stated differently, unconditional fairness rules out more behaviors than strong fairness, and strong excludes more behaviors than weak fairness. For $\mathcal{F}=\left\{A_{1}, \ldots, A_{k}\right\}$, let fairness assumption $\mathcal{F}_{\text {ucond }}=(\mathcal{F}, \varnothing, \varnothing), \mathcal{F}_{\text {strong }}=(\varnothing, \mathcal{F}, \varnothing)$, and $\mathcal{F}_{\text {weak }}=(\varnothing, \varnothing, \mathcal{F})$. Then for any transition system $T S$ and LT property $P$ it follows that:

$$
T S \models_{\mathcal{F}_{\text {weak }}} P \Rightarrow T S \models_{\mathcal{F}_{\text {strong }}} P \Rightarrow T S \models_{\mathcal{F}_{\text {ucond }}} P .
$$

## Example 3.49. Independent Traffic Lights

Consider again the independent traffic lights. Let action switch2green denote the switching to green. Similarly switch2red denotes the switching to red. The fairness assumption

$$
\mathcal{F}=\left\{\left\{\text { switch } 2 \text { green }_{1}, \text { switch } 2 \text { red }_{1}\right\},\left\{\text { switch } 2 \text { green }_{2},{\text { switch } \left.\left.2 \text { red }_{2}\right\}\right\}}\right.\right.
$$

expresses that both traffic lights infinitely often switch color. In this case, it is irrelevant whether strong, weak, or unconditional fairness is required.

Note that in this example $\mathcal{F}$ is not a verifiable system property (as it is not guaranteed to hold), but a natural property which is satisfied for a practical implementation of the system (with two independent processors). Obviously,
$\operatorname{Tr}^{L i g h t} 1_{1}| | \mid \operatorname{Tr}$ Light $_{2} \models_{\mathcal{F}}$ "each traffic light is green infinitely often"
while the corresponding proposition for the nonfair relation $\models$ is refuted.

## Example 3.50. Fairness for Mutual Exclusion Algorithms

Consider again the semaphore-based mutual exclusion algorithm, and assume the fairness assumption $\mathcal{F}$ consists of

$$
\mathcal{F}_{\text {weak }}=\left\{\left\{\text { req }_{1}\right\},\left\{\text { req }_{2}\right\}\right\} \quad \text { and } \quad \mathcal{F}_{\text {strong }}=\left\{\left\{\text { enter }_{1}\right\},\left\{\text { enter }_{2}\right\}\right\}
$$

and $\mathcal{F}_{\text {ucond }}=\varnothing$. The strong fairness constraint requires each process to enter its critical section infinitely often when it infinitely often gets the opportunity to do so. This does not forbid a process to never leave its noncritical section. To avoid this unrealistic scenario, the weak fairness constraint requires that any process infinitely often requests to enter the critical section. In order to do so, each process has to leave the noncritical section infinitely often. It follows that $T S_{S e m} \models_{\mathcal{F}} P$ where $P$ stands for the property "every process enters its critical section infinitely often".

Weak fairness is sufficient for request actions, as such actions are not critical: if req $q_{i}$ is executable in (global) state $s$, then it is executable in all direct successor states of $s$ that are reached by an action that differs from req $_{i}$.

Peterson's algorithm satisfies the strong fairness property
"Every process that requests access to the critical section will eventually be able to do so".
We can, however, not ensure that a process will ever leave its noncritical section and request the critical section. That is, the property $P$ is refuted. This can be "repaired" by imposing the weak fairness constraint $\mathcal{F}_{\text {weak }}=\left\{\left\{\operatorname{req}_{1}\right\},\left\{\operatorname{req}_{2}\right\}\right\}$. We now have $T S_{\text {Pet }}{\models \mathcal{F}_{\text {weak }}}$.

### 3.5.2 Fairness Strategies

The examples in the previous section indicate that fairness assumptions may be necessary to verify liveness properties of transition system TS. In order to rule out the "unrealistic" computations, fairness assumptions are imposed on the traces of $T S$, and it is checked whether $T S \models_{\mathcal{F}} P$ as opposed to checking $T S \not \models P$ (without fairness). Which fairness assumptions are appropriate to check $P$ ? Many model-checking tools provide the possibility to work with built-in fairness assumptions. Roughly speaking, the intention is to rule out executions that cannot occur in a realistic implementation. But what does that exactly mean? In order to give some insight into this, we consider several fairness assumptions for synchronizing concurrent systems. The aim is to establish a fair communication mechanism between the various processes involved. A rule of thumb is: Strong fairness is needed to obtain an adequate resolution of contentions (between processes), while weak fairness suffices for sets of actions that represent the concurrent execution of independent actions (i.e., interleaving).

For modeling asynchronous concurrency by means of transition systems, the following rule of thumb can be adopted:

$$
\text { concurrency }=\text { interleaving (i.e., nondeterminism) }+ \text { fairness }
$$

Example 3.51. Fair Concurrency with Synchronization
Consider the concurrent transition system:

$$
T S=T S_{1}\left\|T S_{2}\right\| \ldots \| T S_{n}
$$

where $T S_{i}=\left(S_{i}, A_{c t}{ }_{i} \rightarrow_{i}, I_{i}, A P_{i}, L_{i}\right)$, for $1 \leqslant i \leqslant n$, is a transition system without terminal states. Recall that each pair of processes $T S_{i}$ and $T S_{j}$ (for $i \neq j$ ) has to synchronize on their common sets of actions, i.e., $\operatorname{Syn}_{i, j}=A c t_{i} \cap A c t_{j}$. It is assumed that $\operatorname{Syn}_{i, j} \cap$ $A c t_{k}=\varnothing$ for any $k \neq i, j$. For simplicity, it is assumed that TS has no terminal states. (In case there are terminal states, each finite execution is considered to be fair.)

We consider several fairness assumptions on the transition system TS. First, consider the strong fairness assumption

$$
\left\{A c t_{1}, A c t_{2}, \ldots, A c t_{n}\right\}
$$

which ensures that each transition system $T S_{i}$ executes an action infinitely often, provided the composite system $T S$ is infinitely often in a (global) state with a transition being executable in which $T S_{i}$ participates. This fairness assumption, however, cannot ensure that a communication will ever occur-it is possible for each $T S_{i}$ to only execute local actions ad infinitum.

In order to force a synchronization to take place every now and then, the strong fairness assumption

$$
\begin{equation*}
\left\{\{\alpha\} \mid \alpha \in \operatorname{Syn}_{i, j}, 0<i<j \leqslant n\right\} \tag{3.1}
\end{equation*}
$$

could be imposed. It forces every synchronization action to happen infinitely often. Alternatively, a somewhat weaker fairness assumption can be imposed by requiring every pair of processes to synchronize - regardless of the synchronization action-infinitely often. The corresponding strong fairness assumption is

$$
\begin{equation*}
\left\{\operatorname{Syn}_{i, j} \mid 0<i<j \leqslant n\right\} . \tag{3.2}
\end{equation*}
$$

Whereas (3.2) allows processes to always synchronize on the same action, (3.1) does not permit this. The strong fairness assumption:

$$
\left\{\bigcup_{0<i<j \leqslant n} S y n_{i, j}\right\}
$$

goes even one step further as it only requires a synchronization to take place infinitely often, regardless of the process involved. This fairness assumption does not rule out executions in which always the same synchronization takes place or in which always the same pair of processes synchronizes.

Note that all fairness assumptions in this example so far are strong. This requires that infinitely often a synchronization is enabled. As the constituting transition systems $T S_{i}$ may execute internal actions, synchronizations are not continuously enabled, and hence weak fairness is in general inappropriate.

If the internal actions should be fairly considered, too, then we may use, e.g., the strong fairness assumption

$$
\left\{A c t_{1} \backslash \operatorname{Syn}_{1}, \ldots, A c t_{n} \backslash \operatorname{Syn}_{n}\right\} \cup\{\{\alpha\} \mid \alpha \in \operatorname{Syn}\},
$$

where $\operatorname{Syn}_{i}=\bigcup_{j \neq i} \operatorname{Syn}_{i, j}$ denotes the set of all synchronization actions of $T S_{i}$ and Syn $=\bigcup_{i} S y n_{i}$.

Under the assumption that in every (local) state either only internal actions or only synchronization actions are executable, it suffices to impose the weak fairness constraint

$$
\left\{A c t_{1} \backslash S y n_{1}, \ldots, A c t_{n} \backslash S y n_{n}\right\}
$$

Weak fairness is appropriate for the internal actions $\alpha \in A c t_{i} \backslash S y n_{i}$, as the ability to perform an internal action is preserved until it will be executed.

As an example of another form of fairness we consider the following sequential hardware circuit.

## Example 3.52. Circuit Fairness

For sequential circuits we have modeled the environmental behavior, which provides the input bits, by means of nondeterminism. It may be necessary to impose fairness assumptions on the environment in order to be able to verify liveness properties, such as "the values 0 and 1 are output infinitely often". Let us illustrate this by means of a concrete example. Consider a sequential circuit with input variable $x$, output variable $y$, and register $r$. Let the transition function and the output function be defined as

$$
\lambda_{y}=\delta_{r}=x \leftrightarrow \neg r .
$$

That is, the circuit inverts the register and output evaluation if and only if the input bit is set. If $x=0$, then the register evaluation remains unchanged. The value of the register is output. Suppose all transitions leading to a state with a register evaluation of the form $[r=1, \ldots]$ are labeled with the action set. Imposing the unconditional fairness assumption $\{\{$ set $\}\}$ ensures that the values 0 and 1 are output infinitely often.

### 3.5.3 Fairness and Safety

While fairness assumptions may be necessary to verify liveness properties, they are irrelevant for verifying safety properties, provided that they can always be ensured by means of an appropriate scheduling strategy. Such fairness assumptions are called realizable fairness assumptions. A fairness assumption cannot be realized in a transition system whenever there exists a reachable state from where no fair path begins. In this case, it is impossible to design a scheduler that resolves the nondeterminism such that only fair paths remain.

## Example 3.53. A Nonrealizable Fairness Assumption

Consider the transition system depicted in Figure 3.13, and suppose the unconditional fairness assumption $\{\{\alpha\}\}$ is imposed. As the $\alpha$-transition can only be taken once, it is evident that the transition system can never guarantee this form of fairness. As there is a reachable state from which no unconditional fair path exists, this fairness assumption is nonrealizable.

## Definition 3.54. Realizable Fairness Assumption

Let $T S$ be a transition system with the set of actions $A c t$ and $\mathcal{F}$ a fairness assumption for Act. $\mathcal{F}$ is called realizable for $T S$ if for every reachable state $s$ : FairPathsf $(s) \neq \varnothing$.


Figure 3.13: Unconditional fairness.

Stated in words, a fairness assumption is realizable in a transition system $T S$ whenever in any reachable state at least one fair execution is possible. This entails that every initial finite execution fragment of $T S$ can be completed to a fair execution. Note that there is no requirement on the unreachable states.

The following theorem shows the irrelevance of realizable fairness assumptions for the verification of safety properties. The suffix property of fairness assumptions is essential for its proof. This means the following. If

$$
\rho=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} s_{2} \xrightarrow{\alpha_{3}} \ldots
$$

is an (infinite) execution fragment, then $\rho$ is fair if and only if every suffix

$$
s_{j} \xrightarrow{\alpha_{j+1}} s_{j+1} \xrightarrow{\alpha_{j+2}} s_{j+2} \xrightarrow{\alpha_{j+3}} \ldots
$$

of $\rho$ is fair too. Conversely, every fair execution fragment $\rho$ (as above) starting in state $s_{0}$ can be preceded by an arbitrary finite execution fragment

$$
s_{0}^{\prime} \xrightarrow{\beta_{1}} s_{1}^{\prime} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n}^{\prime}=s_{0}
$$

ending in $s_{0}$. Proceeding in $s_{0}$ by execution $\rho$ yields the fair execution fragment:

$$
\underbrace{s_{0}^{\prime} \xrightarrow{\beta_{1}} s_{1}^{\prime} \xrightarrow{\beta_{2}} \ldots \stackrel{\beta_{n}}{\longrightarrow} s_{n}^{\prime}}_{\text {arbitrary starting fragment }}=\underbrace{s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} s_{2} \xrightarrow{\alpha_{3}} \ldots}_{\text {fair continuation }}
$$

## Theorem 3.55. Realizable Fairness is Irrelevant for Safety Properties

Let TS be a transition system with set of propositions AP, $\mathcal{F}$ a realizable fairness assumption for TS, and $P_{\text {safe }}$ a safety property over AP. Then:

$$
T S \models P_{\text {safe }} \quad \text { if and only if } \quad T S \models_{\mathcal{F}} \quad P_{\text {safe }} .
$$

Proof: " $\Rightarrow$ ": Assume $T S \vDash P_{\text {safe }}$. Then, by definition of $\models$ and the fact that the fair traces of $T S$ are a subset of the traces of $T S$, we have

$$
\operatorname{Fair}_{\operatorname{Traces}_{\mathcal{F}}}(T S) \subseteq \operatorname{Traces}(T S) \subseteq P_{\text {safe }}
$$

Thus, by definition of $\models_{\mathcal{F}}$ it follows that $T S \models_{\mathcal{F}} P_{\text {safe }}$.
$" \Leftarrow "$ : Assume $T S \models_{\mathcal{F}} P_{\text {safe }}$. It is to be shown $T S \models P_{\text {safe }}$, i.e., $\operatorname{Traces}(T S) \subseteq P_{\text {safe }}$. This is done by contraposition. Let $\sigma \in \operatorname{Traces}(T S)$ and assume $\sigma \notin P_{\text {safe }}$. As $\sigma \notin P_{\text {safe }}$, there is a bad prefix of $\sigma, \widehat{\sigma}$ say, for $P_{\text {safe }}$. Hence, the set of properties that has $\widehat{\sigma}$ as a prefix, i.e.,

$$
P=\left\{\sigma^{\prime} \in\left(2^{A P}\right)^{\omega} \mid \widehat{\sigma} \in \operatorname{pref}\left(\sigma^{\prime}\right)\right\},
$$

satisfies $P \cap P_{\text {safe }}=\varnothing$. Further, let $\widehat{\pi}=s_{0} s_{1} \ldots s_{n}$ be a finite path fragment of $T S$ with

$$
\operatorname{trace}(\widehat{\pi})=\widehat{\sigma} .
$$

Since $\mathcal{F}$ is a realizable fairness assumption for $T S$ and $s_{n} \in \operatorname{Reach}(T S)$, there is an $\mathcal{F}$-fair path starting in $s_{n}$. Let

$$
s_{n} s_{n+1} s_{n+2} \ldots \in \text { FairPaths }_{\mathcal{F}}\left(s_{n}\right) .
$$

The path $\pi=s_{0} \ldots s_{n} s_{n+1} s_{n+2} \ldots$ is in FairPathsf $\mathcal{F}(T S)$ and thus,

$$
\operatorname{trace}(\pi)=L\left(s_{0}\right) \ldots L\left(s_{n}\right) L\left(s_{n+1}\right) L\left(s_{n+2}\right) \ldots \in \operatorname{FairTraces}_{\mathcal{F}}(T S) \subseteq P_{\text {safe }}
$$

On the other hand, $\widehat{\sigma}=L\left(s_{0}\right) \ldots L\left(s_{n}\right)$ is a prefix of $\operatorname{trace}(\pi)$. Thus, $\operatorname{trace}(\pi) \in P$. This contradicts $P \cap P_{\text {safe }}=\varnothing$.

Theorem 3.55 does not hold if arbitrary (i.e., possibly nonrealizable) fairness assumptions are permitted. This is illustrated by the following example.

## Example 3.56. Nonrealizable Fairness may harm Safety Properties

Consider the transition system $T S$ in Figure 3.14 and suppose the unconditional fairness assumption $\mathcal{F}=\{\{\alpha\}\}$ is imposed. $\mathcal{F}$ is not realizable for $T S$, as the noninitial state (referred to as state $s_{1}$ ), is reachable, but has no $\mathcal{F}$-fair execution. Obviously, TS has only one fair path (namely the path that never leaves the initial state $s_{0}$ ). In contrast, paths of the form $s_{0} \ldots s_{0} s_{1} s_{1} s_{1} \ldots$ are not fair, since $\alpha$ is only executed finitely often. Accordingly, we have that

$$
T S \models_{\mathcal{F}} \text { "never } a \text { " but } T S \not \models \text { "never } a \text { ". }
$$

### 3.6 Summary

- The set of reachable states of a transition system $T S$ can be determined by a search algorithm on the state graph of $T S$.


Figure 3.14: Unconditional fairness may be relevant for safety properties.

- A trace is a sequence of sets (!) of atomic propositions. The traces of a transition system TS are obtained from projecting the paths to the sequence of state labels.
- A linear-time (LT, for short) property is a set of infinite words over the alphabet $2^{A P}$.
- Two transition systems are trace-equivalent (i.e., they exhibit the same traces) if and only if they satisfy the same LT properties.
- An invariant is an LT property that is purely state-based and requires a propositional logic formula $\Phi$ to hold for all reachable states. Invariants can be checked using a depth-first search where the depth-first search stack can be used to provide a counterexample in case an invariant is refuted.
- Safety properties are generalizations of invariants. They constrain the finite behaviors. The formal definition of safety properties can be provided by means of their bad prefixes in the sense that each trace that refutes a safety property has a finite prefix, the bad prefix, that causes this.
- Two transition systems exhibit the same finite traces if and only if they satisfy the same safety properties.
- A liveness property is an LT property if it does not rule out any finite behavior. It constrains infinite behavior.
- Any LT property is equivalent to an LT property that is a conjunction of a safety and a liveness property.
- Fairness assumptions serve to rule out traces that are considered to be unrealistic. They consist of unconditional, strong, and weak fairness constraints, i.e., constraints on the actions that occur along infinite executions.
- Fairness assumptions are often necessary to establish liveness properties, but they are - provided they are realizable - irrelevant for safety properties.


### 3.7 Bibliographic Notes

The dining philosophers example discussed in Example 3.2 has been developed by Dijkstra [128] in the early seventies to illustrate the intricacies of concurrency. Since then it has become one of the standard examples for reasoning about parallel systems.

The depth-first search algorithm that we used as a basis for the invariance checking algorithm goes back to Tarjan [387]. Further details about graph traversal algorithms can be found in any textbook on algorithms and data structures, e.g. [100], or on graph algorithms [188].

Traces. Traces have been introduced by Hoare [202] to describe the linear-time behavior of transition systems and have been used as the initial semantical model for the process algebra CSP. Trace theory has further been developed by, among others, van de Snepscheut [403] and Rem [354] and has successfully been used to design and analyze fine-grained parallel programs that occur in, e.g., asynchronous hardware circuits. Several extensions to traces and their induced equivalences have been proposed, such as failures [65] where a trace is equipped with information about which actions are rejected after execution of such trace. The FDR model checker [356] supports the automated checking of failure-divergence refinement and the checking of safety properties. A comprehensive survey of these refined notions of trace equivalence and trace inclusion has recently been given by Bruda [68].

Safety and liveness. The specification of linear-time properties using sets of infinite sequences of states (and their topological characterizations) goes back to Alpern and Schneider $[5,6,7]$. An earlier approach by Gerth [164] was based on finite sequences. Lamport [257] categorized properties as either safety, liveness, or properties that are neither. Alternative characterizations have been provided by Rem [355] and Gumm [178]. Subclasses of liveness and safety properties in the linear-time framework have been identified by Sistla [371], and Chang, Manna, and Pnueli [80]. Other definitions of liveness properties have been provided by Dederichs and Weber [119] and Naumovich and Clarke [312] (for linear-time properties), and Manolios and Trefler [285, 286] (for branching-time properties). A survey of safety and liveness has been given by Kindler [239].

Fairness. Fairness has implicitly been introduced by Dijkstra [126, 127] by assuming that one should abstract from the speed of processors and that each process gets its turn once it is initiated. Park [321] studied the notion of fairness in providing a semantics to data-flow languages. Weak and strong fairness have been introduced by Lehmann, Pnueli, and Stavi [267] in the context of shared variable concurrent programs. Queille and Sifakis [348] consider fairness for transition systems. An overview of the fairness notions has been provided by Kwiatkowska [252]. An extensive treatment of fairness can be found
in the monograph by Francez [155]. A recent characterization of fairness in terms of topology, language theory, and game theory has been provided by Völzer, Varacca, and Kindler [415].

### 3.8 Exercises

Exercise 3.1. Give the traces on the set of atomic propositions $\{a, b\}$ of the following transition system:


Exercise 3.2. On page 97, a transformation is described of a transition system $T S$ with possible terminal states into an "equivalent" transition system $T S^{*}$ without terminal states. Questions:
(a) Give a formal definition of this transformation $T S \mapsto T S^{*}$
(b) Prove that the transformation preserves trace-equivalence, i.e., show that if $T S_{1}, T S_{2}$ are transition systems (possibly with terminal states) such that $\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)$, then $\operatorname{Traces}\left(T S_{1}^{*}\right)=\operatorname{Traces}\left(T S_{2}^{*}\right) .{ }^{8}$

EXERCISE 3.3. Give an algorithm (in pseudocode) for invariant checking such that in case the invariant is refuted, a minimal counterexample, i.e., a counterexample of minimal length, is provided as an error indication.

Exercise 3.4. Recall the definition of AP-deterministic transition systems (Definition 2.5 on page 24). Let $T S$ and $T S^{\prime}$ be transition systems with the same set of atomic propositions $A P$. Prove the following relationship between trace inclusion and finite trace inclusion:
(a) For $A P$-deterministic $T S$ and $T S^{\prime}$ :

$$
\operatorname{Traces}(T S)=\operatorname{Traces}\left(T S^{\prime}\right) \text { if and only if } \operatorname{Traces}_{f i n}(T S)=\operatorname{Traces}_{f i n}\left(T S^{\prime}\right)
$$

[^14](b) Give concrete examples of $T S$ and $T S^{\prime}$ where at least one of the transition systems is not $A P$-deterministic, but
$$
\operatorname{Traces}(T S) \nsubseteq \operatorname{Traces}\left(T S^{\prime}\right) \text { and } \operatorname{Traces}_{f i n}(T S)=\operatorname{Traces}_{f i n}\left(T S^{\prime}\right)
$$

Exercise 3.5. Consider the set $A P$ of atomic propositions defined by $A P=\{x=0, x>1\}$ and consider a nonterminating sequential computer program $P$ that manipulates the variable $x$. Formulate the following informally stated properties as LT properties:
(a) false
(b) initially $x$ is equal to zero
(c) initially $x$ differs from zero
(d) initially $x$ is equal to zero, but at some point $x$ exceeds one
(e) $x$ exceeds one only finitely many times
(f) $x$ exceeds one infinitely often
(g) the value of $x$ alternates between zero and two
(h) true
(This exercise has been adopted from [355].) Determine which of the provided LT properties are safety properties. Justify your answers.

Exercise 3.6. Consider the set $A P=\{A, B\}$ of atomic propositions. Formulate the following properties as LT properties and characterize each of them as being either an invariance, safety property, or liveness property, or none of these.
(a) $A$ should never occur,
(b) $A$ should occur exactly once,
(c) $A$ and $B$ alternate infinitely often,
(d) $A$ should eventually be followed by $B$.
(This exercise has been inspired by [312].)

Exercise 3.7. Consider the following sequential hardware circuit:


The circuit has input variable $x$, output variable $y$, and registers $r_{1}$ and $r_{2}$ with initial values $r_{1}=0$ and $r_{2}=1$. The set AP of atomic propositions equals $\left\{x, r_{1}, r_{2}, y\right\}$. Besides, consider the following informally formulated LT properties over AP:
$P_{1}$ : Whenever the input $x$ is continuously high (i.e., $x=1$ ), then the output $y$ is infinitely often high.
$P_{2}$ : Whenever currently $r_{2}=0$, then it will never be the case that after the next input, $r_{1}=1$.
$P_{3}$ : It is never the case that two successive outputs are high.
$P_{4}$ : The configuration with $x=1$ and $r_{1}=0$ never occurs.

Questions:
(a) Give for each of these properties an example of an infinite word that belongs to $P_{i}$. Do the same for the property $\left(2^{A P}\right)^{\omega} \backslash P_{i}$, i.e., the complement of $P_{i}$.
(b) Determine which properties are satisfied by the hardware circuit that is given above.
(c) Determine which of the properties are safety properties. Indicate which properties are invariants.
(i) For each safety property $P_{i}$, determine the (regular) language of bad prefixes.
(ii) For each invariant, provide the propositional logic formula that specifies the property that should be fulfilled by each state.

Exercise 3.8. Let LT properties $P$ and $P^{\prime}$ be equivalent, notation $P \cong P^{\prime}$, if and only if $\operatorname{pref}(P)=\operatorname{pref}\left(P^{\prime}\right)$. Prove or disprove: $P \cong P^{\prime}$ if and only if $\operatorname{closure}(P)=\operatorname{closure}\left(P^{\prime}\right)$.

Exercise 3.9. Show that for any transition system TS, the set closure(Traces(TS)) is a safety property such that $T S \models$ closure $(\operatorname{Traces}(T S)$ ).

Exercise 3.10. Let $P$ be an LT property. Prove: $\operatorname{pref}(\operatorname{closure}(P))=\operatorname{pref}(P)$.

Exercise 3.11. Let $P$ and $P^{\prime}$ be liveness properties over $A P$. Prove or disprove the following claims:
(a) $P \cup P^{\prime}$ is a liveness property,
(b) $P \cap P^{\prime}$ is a liveness property.

Answer the same question for $P$ and $P^{\prime}$ being safety properties.

Exercise 3.12. Prove Lemma 3.38 on page 125.

Exercise 3.13. Let $A P=\{a, b\}$ and let $P$ be the LT property of all infinite words $\sigma=$ $A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}$ such that there exists $n \geqslant 0$ with $a \in A_{i}$ for $0 \leqslant i<n,\{a, b\}=A_{n}$ and $b \in A_{j}$ for infinitely many $j \geqslant 0$. Provide a decomposition $P=P_{\text {safe }} \cap P_{\text {live }}$ into a safety and a liveness property.

Exercise 3.14. Let $T S_{S e m}$ and $T S_{P e t}$ be the transition systems for the semaphore-based mutual exclusion algorithm (Example 2.24 on page 43) and Peterson's algorithm (Example 2.25 on page 45), respectively. Let $A P=\left\{\right.$ wait $_{i}$, crit $\left._{i} \mid i=1,2\right\}$. Prove or disprove:

$$
\operatorname{Traces}\left(T S_{S e m}\right)=\operatorname{Traces}\left(T S_{P e t}\right) .
$$

If the property does not hold, provide an example trace of one transition system that is not a trace of the other one.

EXERCISE 3.15. Consider the transition system $T S$ outlined on the right and the sets of actions $B_{1}=\{\alpha\}, B_{2}=\{\alpha, \beta\}$, and $B_{3}=\{\beta\}$. Further, let $E_{b}, E_{a}$ and $E^{\prime}$ be the following LT properties:

- $E_{b}=$ the set of all words $A_{0} A_{1} \cdots \in\left(2^{\{a, b\}}\right)^{\omega}$ with $A_{i} \in\{\{a, b\},\{b\}\}$ for infinitely many $i$ (i.e., infinitely often $b$ ).
- $E_{a}=$ the set of all words $A_{0} A_{1} \cdots \in\left(2^{\{a, b\}}\right)^{\omega}$ with $A_{i} \in\{\{a, b\},\{a\}\}$ for infinitely many $i$ (i.e., infinitely often $a$ ).
- $E^{\prime}=$ set of all words $A_{0} A_{1} \cdots \in\left(2^{\{a, b\}}\right)^{\omega}$ for which there does not exist an $i \in \mathbb{N}$ s.t. $A_{i}=\{a\}, A_{i+1}=$ $\{a, b\}$ and $A_{i+2}=\varnothing$.


Questions:
(a) For which sets of actions $B_{i}(i \in\{1,2,3\})$ and $\operatorname{LT}$ properties $E \in\left\{E_{a}, E_{b}, E^{\prime}\right\}$ it holds that $T S \not \models_{\mathcal{F}_{i}} E$ ? Here, $\mathcal{F}_{i}$ is a strong fairness condition with respect to $B_{i}$ that does not impose any unconditional or weak fairness conditions (i.e., $\left.\mathcal{F}_{i}=\left(\varnothing,\left\{B_{i}\right\}, \varnothing\right)\right)$.
(b) Answer the same question in the case of weak fairness (instead of strong fairness, i.e., $\mathcal{F}_{i}=$ $\left.\left(\varnothing, \varnothing,\left\{B_{i}\right\}\right)\right)$.

Exercise 3.16. Let $T S_{i}$ (for $i=1,2$ ) be the transition system ( $S_{i}, A c t, \rightarrow{ }_{i}, I_{i}, A P_{i}, L_{i}$ ) and $\mathcal{F}=\left(\mathcal{F}_{\text {ucond }}, \mathcal{F}_{\text {strong }}, \mathcal{F}_{\text {weak }}\right)$ be a fairness assumption with $\mathcal{F}_{\text {ucond }}=\varnothing$. Prove or disprove (i.e., give a counterexample for) the following claims:
(a) $\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{1} \| T S_{2}\right)$ where $\operatorname{Syn} \subseteq$ Act
(b) $\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{1} \| \mid T S_{2}\right)$
(c) $\operatorname{Traces}\left(T S_{1} \| T S_{2}\right) \subseteq \operatorname{Traces}\left(T S_{1}\right)$ where $\operatorname{Syn} \subseteq$ Act
(d) $\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right) \Rightarrow \operatorname{FairTraces}_{\mathcal{F}}\left(T S_{1}\right) \subseteq \operatorname{FairTraces}_{\mathcal{F}}\left(T S_{2}\right)$
(e) For liveness property $P$ with $T S_{2} \models_{\mathcal{F}} P$ we have

$$
\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right) \quad \Rightarrow T S_{1} \models_{\mathcal{F}} P
$$

Assume that in items (a) through (c), we have $A P_{2}=\varnothing$ and that $T S_{1} \| T S_{2}$ and $T S_{1} \| \mid T S_{2}$, respectively, have $A P=A P_{1}$ as atomic propositions and $L\left(\left\langle s, s^{\prime}\right\rangle\right)=L_{1}(s)$ as labeling function. In items (d) and (e) you may assume that $A P_{1}=A P_{2}$.

EXERCISE 3.17. Consider the following transition system $T S$ with the set of atomic propositions $\{a\}$ :


Let the fairness assumption

$$
\mathcal{F}=(\varnothing,\{\{\alpha\},\{\beta\}\},\{\{\beta\}\}) .
$$

Determine whether $T S \models_{\mathcal{F}}$ "eventually $a$ ". Justify your answer!

Exercise 3.18. Consider the following transition system TS (without atomic propositions):


Decide which of the following fairness assumptions $\mathcal{F}_{i}$ are realizable for TS. Justify your answers!
(a) $\mathcal{F}_{1}=(\{\{\alpha\}\},\{\{\delta\}\},\{\{\alpha, \beta\}\})$
(b) $\mathcal{F}_{2}=(\{\{\delta, \alpha\}\},\{\{\alpha, \beta\}\},\{\{\delta\}\})$
(c) $\mathcal{F}_{3}=(\{\{\alpha, \delta\},\{\beta\}\},\{\{\alpha, \beta\}\},\{\{\delta\}\})$

Exercise 3.19. Let $A P=\{a, b\}$.
(a) $P_{1}$ denotes the LT property that consists of all infinite words $\sigma=A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}$ such that there exists $n \geqslant 0$ with

$$
\forall j<n . A_{j}=\varnothing \quad \wedge A_{n}=\{a\} \wedge \forall k>n .\left(A_{k}=\{a\} \Rightarrow A_{k+1}=\{b\}\right) .
$$

(i) Give an $\omega$-regular expression for $P_{1}$.
(ii) Apply the decomposition theorem and give expressions for $P_{\text {safe }}$ and $P_{\text {live }}$.
(iii) Justify that $P_{\text {live }}$ is a liveness and that $P_{\text {safe }}$ is a safety property.
(b) Let $P_{2}$ denote the set of traces of the form $\sigma=A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}$ such that

$$
\stackrel{\infty}{\exists} k . A_{k}=\{a, b\} \quad \wedge \quad \exists n \geqslant 0 . \forall k>n .\left(a \in A_{k} \Rightarrow b \in A_{k+1}\right) .
$$

Consider the following transition system TS:


Consider the following fairness assumptions:
(a) $\mathcal{F}_{1}=(\{\{\alpha\}\},\{\{\beta\},\{\delta, \gamma\},\{\eta\}\}, \varnothing)$. Decide whether $T S \models_{\mathcal{F}_{1}} P_{2}$.
(b) $\mathcal{F}_{2}=(\{\{\alpha\}\},\{\{\beta\},\{\gamma\}\},\{\{\eta\}\})$. Decide whether $T S \models_{\mathcal{F}_{2}} P_{2}$.

Justify your answers.

Exercise 3.20. Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system without terminal states and let $A_{1}, \ldots, A_{k}, A_{1}^{\prime}, \ldots, A_{l}^{\prime} \subseteq$ Act.
(a) Let $\mathcal{F}$ be the fairness assumption $\mathcal{F}=\left(\varnothing, \mathcal{F}_{\text {strong }}, \mathcal{F}_{\text {weak }}\right)$ where

$$
\mathcal{F}_{\text {strong }}=\left\{A_{1}, \ldots, A_{k}\right\} \text { and } \mathcal{F}_{\text {weak }}=\left\{A_{1}^{\prime}, \ldots, A_{l}^{\prime}\right\} .
$$

Provide a sketch of a scheduling algorithm that resolves the nondeterminism in $T S$ in an $\mathcal{F}$-fair way.
(b) Let $\mathcal{F}_{u c o n d}=\left\{A_{1}, \ldots, A_{k}\right\}$, viewed as an unconditional fairness assumption for TS. Design a (scheduling) algorithm that checks whether $\mathcal{F}_{\text {ucond }}$ for $T S$ is realizable, and if so, generates an $\mathcal{F}_{\text {ucond }}$-fair execution for $T S$.

## Chapter 4

## Regular Properties

This chapter treats some elementary algorithms to verify important classes of safety properties, liveness properties, and a wide range of other linear-time properties. We first consider regular safety properties, i.e., safety properties whose bad prefixes constitute a regular language, and hence can be recognized by a finite automaton. The algorithm to check a safety property $P_{\text {safe }}$ for a given finite transition system $T S$ relies on a reduction to the invariant-checking problem in a certain product construction of $T S$ with a finite automaton that recognizes the bad prefixes of $P_{\text {safe }}$.

We then generalize this automaton-based verification algorithm to a larger class of lineartime properties, the so-called $\omega$-regular properties. This class of properties covers regular safety properties, but also many other relevant properties such as various liveness properties. $\omega$-regular properties can be represented by so-called Büchi automata, a variant of finite automata that accept infinite (rather than finite) words. Büchi automata will be the key concept to verify $\omega$-regular properties via a reduction to persistence checking. The latter is a variant of invariant checking and aims to show that a certain state-condition holds continuously from some moment on.

### 4.1 Automata on Finite Words

Definition 4.1. Nondeterministic Finite Automaton (NFA)
A nondeterministic finite automaton (NFA) $\mathcal{A}$ is a tuple $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ where

- $Q$ is a finite set of states,
- $\Sigma$ is an alphabet,
- $\delta: Q \times \Sigma \rightarrow 2^{Q}$ is a transition function,
- $Q_{0} \subseteq Q$ is a set of initial states, and
- $F \subseteq Q$ is a set of accept (or: final) states.

The size of $\mathcal{A}$, denoted $|\mathcal{A}|$, is the number of states and transitions in $\mathcal{A}$, i.e.,

$$
|\mathcal{A}|=|Q|+\sum_{q \in Q} \sum_{A \in \Sigma}|\delta(q, A)| .
$$

$\Sigma$ defines the symbols on which the automaton is defined. The (possibly empty) set $Q_{0}$ defines the states in which the automaton may start. The transition function $\delta$ can be identified with the relation $\rightarrow \subseteq Q \times \Sigma \times Q$ given by

$$
q \xrightarrow{A} q^{\prime} \text { iff } q^{\prime} \in \delta(q, A)
$$

Thus, often the notion of transition relation (rather than transition function) is used for $\delta$. Intuitively, $q \xrightarrow{A} q^{\prime}$ denotes that the automaton can move from state $q$ to state $q^{\prime}$ when reading the input symbol $A$.

## Example 4.2. An Example of a Finite-State Automaton

An example of an NFA is depicted in Figure 4.1. Here, $Q=\left\{q_{0}, q_{1}, q_{2}\right\}, \Sigma=\{A, B\}$, $Q_{0}=\left\{q_{0}\right\}, F=\left\{q_{2}\right\}$, and the transition function $\delta$ is defined by

$$
\begin{array}{ll}
\delta\left(q_{0}, A\right)=\left\{q_{0}\right\} & \delta\left(q_{0}, B\right)=\left\{q_{0}, q_{1}\right\} \\
\delta\left(q_{1}, A\right)=\left\{q_{2}\right\} & \delta\left(q_{1}, B\right)=\left\{q_{2}\right\} \\
\delta\left(q_{2}, A\right)=\varnothing & \delta\left(q_{2}, B\right)=\varnothing
\end{array}
$$

This corresponds to the transitions $q_{0} \xrightarrow{A} q_{0}, q_{0} \xrightarrow{B} q_{0}, q_{0} \xrightarrow{B} q_{1}, q_{1} \xrightarrow{A} q_{2}$, and $q_{1} \xrightarrow{B} q_{2}$. The drawing conventions for an NFA are the same as for labeled transition systems. Accept states are distinguished from other states by drawing them with a double circle.

The intuitive operational behavior of an NFA is as follows. The automaton starts in one of the states in $Q_{0}$, and then is fed with an input word $w \in \Sigma^{*}$. The automaton reads this


Figure 4.1: An example of a finite-state automaton.
word character by character from the left to the right. (The reader may assume that the input word is provided on a tape from which the automaton reads the input symbols from the left to the right by moving a cursor from the current position $i$ to the next position $i+1$ when reading the $i$ th input symbol. However, the automaton can neither write on the tape nor move the cursor in another way than one position to the right.) After reading an input symbol, the automaton changes its state according to the transition relation $\delta$. That is, if the current input symbol $A$ is read in the state $q$, the automaton chooses one of the possible transitions $q \xrightarrow{A} q^{\prime}$ (i.e., one state $q^{\prime} \in \delta(q, A)$ ) and moves to $q^{\prime}$ where the next input symbol will be consumed. If $\delta(q, A)$ contains two or more states, then the decision for the next state is made nondeterministically. An NFA cannot perform any transition when its current state $q$ does not have an outgoing transition that is labeled with the current input symbol $A$. In that case, i.e., if $\delta(q, A)=\varnothing$, the automaton is stuck and no further progress can be made. The input word is said to be rejected. When the complete input word has been read, the automaton halts. It accepts whenever the current state is an accept state, and it rejects otherwise.

This intuitive explanation of the possible behaviors of an NFA for a given input word $w=A_{1} \ldots A_{n}$ is formalized by means of runs for $w$ (see Definition 4.3 below). For any input word $w$ there might be several possible behaviors (runs); some of them might be accepting, some of them might be rejecting. Word $w$ is accepted by $\mathcal{A}$ if at least one of its runs is accepting, i.e., succeeds in reading the whole word and ends in a final state. This relies on the typical nondeterministic acceptor criterion which assumes an oracle for resolving the nondeterminism such that, whenever possible, an accepting run will be generated.

## Definition 4.3. Runs, Accepted Language of an NFA

Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NFA and $w=A_{1} \ldots A_{n} \in \Sigma^{*}$ a finite word. A run for $w$ in $\mathcal{A}$ is a finite sequence of states $q_{0} q_{1} \ldots q_{n}$ such that

$$
\text { - } q_{0} \in Q_{0} \text { and }
$$

- $q_{i} \xrightarrow{A_{i+1}} q_{i+1}$ for all $0 \leqslant i<n$.

Run $q_{0} q_{1} \ldots q_{n}$ is called accepting if $q_{n} \in F$. A finite word $w \in \Sigma^{*}$ is called accepted by $\mathcal{A}$ if there exists an accepting run for w. The accepted language of $\mathcal{A}$, denoted $\mathcal{L}(\mathcal{A})$, is the set of finite words in $\Sigma^{*}$ accepted by $\mathcal{A}$, i.e.,

$$
\mathcal{L}(\mathcal{A})=\left\{w \in \Sigma^{*} \mid \text { there exists an accepting run for } w \text { in } \mathcal{A}\right\} .
$$

## Example 4.4. Runs and Accepted Words

Example runs of the automaton $\mathcal{A}$ in Figure 4.1 are $q_{0}$ for the empty word $\varepsilon, q_{0} q_{1}$ for the word consisting of the symbol $B, q_{0} q_{0} q_{0} q_{0}$ for, e.g., the words $A B A$ and $B B A$ and $q_{0} q_{1} q_{2}$ for the words $B A$, and $B B$. Accepting runs are runs that finish in the final state $q_{2}$. For instance, the runs $q_{0} q_{1} q_{2}$ for $B A$ and $B B$ and $q_{0} q_{0} q_{1} q_{2}$ for the words $A B B, A B A$, $B B A$, and $B B B$ are accepting. Thus, these words belong to $\mathcal{L}(\mathcal{A})$. The word $A A A$ is not accepted by $\mathcal{A}$ since it only has single run, namely $q_{0} q_{0} q_{0} q_{0}$, which is not accepting.

The accepted language $\mathcal{L}(\mathcal{A})$ is given by the regular expression $(A+B)^{*} B(A+B)$. Thus, $\mathcal{L}(\mathcal{A})$ is the set of words over $\{A, B\}$ where the last but one symbol is $B$.

The special cases $Q_{0}=\varnothing$ or $F=\varnothing$ are allowed. In both cases, $\mathcal{L}(\mathcal{A})=\varnothing$. If $F=\varnothing$, then there are no accepting runs. If there are no initial states, then there are no runs at all. Intuitively, the automaton rejects any input word immediately.

An equivalent alternative characterization of the accepted language of an NFA $\mathcal{A}$ is as follows. Let $\mathcal{A}$ be an NFA as above. We extend the transition function $\delta$ to the function $\delta^{*}: Q \times \Sigma^{*} \rightarrow 2^{Q}$ as follows: $\delta^{*}(q, \varepsilon)=\{q\}, \delta^{*}(q, A)=\delta(q, A)$, and

$$
\delta^{*}\left(q, A_{1} A_{2} \ldots A_{n}\right)=\bigcup_{p \in \delta\left(q, A_{1}\right)} \delta^{*}\left(p, A_{2} \ldots A_{n}\right)
$$

Stated in words, $\delta^{*}(q, w)$ is the set of states that are reachable from $q$ for the input word w. In particular, $\bigcup_{q_{0} \in Q_{0}} \delta^{*}\left(q_{0}, w\right)$ is the set of all states where a run for $w$ in $\mathcal{A}$ can end. If one of these states is final, then $w$ has an accepting run. Vice versa, if $w \notin \mathcal{L}(\mathcal{A})$, then none of these states is final. Hence, we have the following alternative characterization of the accepted language of an NFA by means of the extended transition function $\delta^{*}$ :

## Lemma 4.5. Alternative Characterization of the Accepted Language

Let $\mathcal{A}$ be an NFA. Then:

$$
\mathcal{L}(\mathcal{A})=\left\{w \in \Sigma^{*} \mid \delta^{*}\left(q_{0}, w\right) \cap F \neq \varnothing \text { for some } q_{0} \in Q_{0}\right\} .
$$

It can be shown that the language accepted by an NFA constitutes a regular language. In fact, there are algorithms that for a given NFA $\mathcal{A}$ generate a regular expression for the language $\mathcal{L}(\mathcal{A})$. Vice versa, for any regular expression $E$, an NFA can be constructed that accepts $\mathcal{L}(E)$. Hence, the class of regular languages agrees with the class of languages accepted by an NFA.

An example of a language that is nonregular (but context-free) is $\left\{A^{n} B^{n} \mid n \geqslant 0\right\}$. There does not exist an NFA that accepts it. The intuitive argument for this is that one needs to be able to count the number of $A$ 's so as to be able to determine the number of $B$ 's that are to follow.

Since NFAs serve to represent (regular) languages we may identify those NFA that accept the same language:

## Definition 4.6. Equivalence of NFAs

Let $\mathcal{A}$ and $\mathcal{A}^{\prime}$ be NFAs with the same alphabet. $\mathcal{A}$ and $\mathcal{A}^{\prime}$ are called equivalent if $\mathcal{L}(\mathcal{A})=\mathcal{L}\left(\mathcal{A}^{\prime}\right)$.

A fundamental issue in automata theory is to decide for a given NFA $\mathcal{A}$ whether its accepted language is empty, i.e., whether $\mathcal{L}(\mathcal{A})=\varnothing$. This is known as the emptiness problem. From the acceptance condition, it follows directly that $\mathcal{L}(\mathcal{A})$ is nonempty if and only if there is at least one run that ends in some final state. Thus, nonemptiness of $\mathcal{L}(\mathcal{A})$ is equivalent to the existence of an accept state $q \in F$ which is reachable from an initial state $q_{0} \in Q_{0}$. This can easily be determined in time $\mathcal{O}(|\mathcal{A}|)$ using a depth-first search traversal that encounters all states that are reachable from the initial states and checks whether one of them is final. For state $q \in Q$, let $\operatorname{Reach}(q)=\bigcup_{w \in \Sigma^{*}} \delta^{*}(q, w)$; that is, $\operatorname{Reach}(q)$ is the set of states $q^{\prime}$ that are reachable via an arbitrary run starting in state $q$.

## Theorem 4.7. Language Emptiness is Equivalent to Reachability

Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NFA. Then, $\mathcal{L}(\mathcal{A}) \neq \varnothing$ if and only if there exists $q_{0} \in Q_{0}$ and $q \in F$ such that $q \in \operatorname{Reach}\left(q_{0}\right)$.

Regular languages exhibit some interesting closure properties, e.g., the union of two regular languages is regular. The same applies to concatenation and Kleene star (finite repetition). This is immediate from the definition of regular languages as those languages that can be generated by regular expressions. They are also closed under intersection and complementation, i.e., if $\mathcal{L}, \mathcal{L}_{1}, \mathcal{L}_{2}$ are regular languages over the alphabet $\Sigma$, then so are $\overline{\mathcal{L}}=\Sigma^{*} \backslash \mathcal{L}$ and $\mathcal{L}_{1} \cap \mathcal{L}_{2}$.

Let us briefly sketch the proofs for this. In both cases, we may proceed on the basis of finite automata and assume a representation of the given regular languages by NFA $\mathcal{A}$, $\mathcal{A}_{1}$, and $\mathcal{A}_{2}$ with the input alphabet $\Sigma$ that accept the regular languages $\mathcal{L}, \mathcal{L}_{1}$, and $\mathcal{L}_{2}$, respectively. Intersection can be realized by a product construction $\mathcal{A}_{1} \otimes \mathcal{A}_{2}$ which can be viewed as a parallel composition with synchronization over all symbols $A \in \Sigma$. In fact, the formal definition of $\otimes$ is roughly the same as the synchronization operator $\|$; see Definition 2.26 on page 48. The idea is simply that for the given input word, we run the two automata in parallel and reject as soon as one automaton cannot read the current input symbol, but accept if the input word is fully consumed and both automata accept (i.e., are in a final state).

## Definition 4.8. Synchronous Product of NFAs

For NFA $\mathcal{A}_{i}=\left(Q_{i}, \Sigma, \delta_{i}, Q_{0, i}, F_{i}\right)$, with $i=1,2$, the product automaton

$$
\mathcal{A}_{1} \otimes \mathcal{A}_{2}=\left(Q_{1} \times Q_{2}, \Sigma, \delta, Q_{0,1} \times Q_{0,2}, F_{1} \times F_{2}\right)
$$

where $\delta$ is defined by

It follows that this product construction of automata corresponds indeed to the intersection of their accepting languages, i.e., $\mathcal{L}\left(\mathcal{A}_{1} \otimes \mathcal{A}_{2}\right)=\mathcal{L}\left(\mathcal{A}_{1}\right) \cap \mathcal{L}\left(\mathcal{A}_{2}\right)$.

Let us now consider the complementation operator. Given an NFA $\mathcal{A}$ with the input alphabet $\Sigma$, we aim to construct an NFA for the complement language $\Sigma^{*} \backslash \mathcal{L}(\mathcal{A})$. The main step to do so is first to construct an equivalent deterministic finite automaton $\mathcal{A}_{\text {det }}$ which can be complemented in a quite simple way.

## Definition 4.9. Deterministic Finite Automaton (DFA)

Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NFA. $\mathcal{A}$ is called deterministic if $\left|Q_{0}\right| \leqslant 1$ and $|\delta(q, A)| \leqslant 1$ for all states $q \in Q$ and all symbols $A \in \Sigma$. We will use the abbreviation DFA for a deterministic finite automaton.

DFA $\mathcal{A}$ is called total if $\left|Q_{0}\right|=1$ and $|\delta(q, A)|=1$ for all $q \in Q$ and all $A \in \Sigma$.

Stated in words, an NFA is deterministic if it has at most a single initial state and if for each symbol $A$ the successor state of each state $q$ is either uniquely defined (if $|\delta(q, A)|=1$ ) or undefined (if $\delta(q, A)=\varnothing$ ). Total DFAs provide unique successor states, and thus, unique runs for each input word. Any DFA can be turned into an equivalent total DFA by simply adding a nonfinal trap state, $q_{\text {trap }}$ say, that is equipped with a self-loop for any symbol $A \in \Sigma$. From any state $q \neq q_{\text {trap }}$, there is a transition to $q_{\text {trap }}$ for any symbol $A$ for which $q$ has no $A$-successor in the given nontotal DFA.

Total DFA are often written in the form $\mathcal{A}=\left(Q, \Sigma, \delta, q_{0}, F\right)$ where $q_{0}$ stands for the unique initial state and $\delta$ is a (total) transition function $\delta: Q \times \Sigma \rightarrow Q$. Also, the extended transition function $\delta^{*}$ of a total DFA can be viewed as a total function $\delta^{*}: Q \times \Sigma^{*} \rightarrow Q$, which for given state $q$ and finite word $w$ returns the unique state $p=\delta^{*}(q, w)$ that is reached from state $q$ for the input word $w$. In particular, the accepted language of a total DFA $\mathcal{A}=\left(Q, \Sigma, \delta, q_{0}, F\right)$ is given by

$$
\mathcal{L}(\mathcal{A})=\left\{w \in \Sigma^{*} \mid \delta^{*}\left(q_{0}, w\right) \in F\right\} .
$$

The observation that total DFAs have exactly one run for each input word allows complementing a total DFA $\mathcal{A}$ by simply declaring all states to be final that are nonfinal in $\mathcal{A}$ and vice versa. Formally, if $\mathcal{A}=\left(Q, \Sigma, \delta, q_{0}, F\right)$ is a total DFA then $\overline{\mathcal{A}}=\left(Q, \Sigma, \delta, q_{0}, Q \backslash F\right)$ is a total DFA with $\mathcal{L}(\overline{\mathcal{A}})=\Sigma^{*} \backslash \mathcal{L}(\mathcal{A})$. Note that the operator $\mathcal{A} \mapsto \overline{\mathcal{A}}$ applied to a nontotal DFA (or NFA with proper nondeterministic choices) fails to provide an automaton for the complement language (why?).

It remains to explain how to construct for a given NFA $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ an equivalent total DFA $\mathcal{A}_{\text {det }}$. This can be done by a powerset construction, also often called a subset construction, since the states of $\mathcal{A}_{\text {det }}$ are the subsets of $Q$. This allows $\mathcal{A}_{\text {det }}$ to simulate $\mathcal{A}$ by moving the prefixes $A_{1} \ldots A_{i}$ of the given input word $w=A_{1} \ldots A_{n} \in \Sigma$ to the set of states that are reachable in $\mathcal{A}$ for $A_{1} \ldots A_{i}$. That is, $\mathcal{A}_{\text {det }}$ starts in $Q_{0}$, the set of initial states in $\mathcal{A}$. If $\mathcal{A}_{\text {det }}$ is in state $Q^{\prime}$ (which is a subset of $\mathcal{A}$ 's state space $Q$ ), then $\mathcal{A}_{\text {det }}$ moves the input symbol $A$ to $Q^{\prime \prime}=\bigcup_{q \in Q^{\prime}} \delta(q, A)$. If the input word has been consumed and $\mathcal{A}_{\text {det }}$ is in a state $Q^{\prime}$ that contains a state in $\mathcal{A}$ 's set of accept states, then $\mathcal{A}_{\text {det }}$ accepts. The latter means that there exists an accepting run in $\mathcal{A}$ for the given input word $w$ that ends in an accept state, and hence, $w \in \mathcal{L}(\mathcal{A})$. The formal definition of $\mathcal{A}_{\text {det }}$ is $\mathcal{A}_{\text {det }}=\left(2^{Q}, \Sigma, \delta_{\text {det }}, Q_{0}, F_{\text {det }}\right)$ where

$$
F_{\text {det }}=\left\{Q^{\prime} \subseteq Q \mid Q^{\prime} \cap F \neq \varnothing\right\}
$$

and where the total transition function $\delta_{d e t}: 2^{Q} \times \Sigma \rightarrow 2^{Q}$ is defined by

$$
\delta_{d e t}\left(Q^{\prime}, A\right)=\bigcup_{q \in Q^{\prime}} \delta(q, A)
$$

Clearly, $\mathcal{A}_{\text {det }}$ is a total DFA and, for all finite words $w \in \Sigma^{*}$, we have

$$
\delta_{d e t}^{*}\left(Q_{0}, w\right)=\bigcup_{q_{0} \in Q_{0}} \delta^{*}\left(q_{0}, w\right)
$$

Thus, by Lemma 4.5, $\mathcal{L}\left(\mathcal{A}_{\text {det }}\right)=\mathcal{L}(\mathcal{A})$.

## Example 4.10. Determinizing a Nondeterministic Finite Automaton

Consider the NFA depicted in Figure 4.1 on page 153. This automaton is not deterministic as on input symbol $B$ in state $q_{0}$ the next state is not uniquely determined. The total DFA that is obtained through the powerset construction is depicted in Figure 4.2.


Figure 4.2: A DFA accepting $\mathcal{L}\left((A+B)^{*} B(A+B)\right)$.

The powerset construction yields a total DFA that is exponentially larger than the original NFA. In fact, although DFAs and NFAs have the same power (both are equivalent formalisms for regular languages), NFAs can be much more efficient. The regular language given by the regular expression $\mathrm{E}_{k}=(A+B)^{*} B(A+B)^{k}$ (where $k$ is a natural number) is accepted by an NFA with $k+2$ states (namely?), but it can be shown that there is no equivalent DFA with less than $2^{k}$ states. The intuitive argument for the latter is that each DFA for $\mathcal{L}\left(\mathrm{E}_{k}\right)$ needs to "remember" the positions of the symbol $B$ among the last $k$ input symbols which yields $\Omega\left(2^{k}\right)$ states.

We finally mention that for any regular language $\mathcal{L}$ there is a unique DFA $\mathcal{A}$ with $\mathcal{L}=\mathcal{L}(\mathcal{A})$ where the number of states is minimal under all DFAs for $\mathcal{L}$. Uniqueness is understood up to isomorphism, i.e., renaming of the states. (This does not hold for NFA. Why?) There is an algorithm to minimize a given DFA with $N$ states into its equivalent minimal DFA which is based on partition refinement and takes $\mathcal{O}(N \cdot \log N)$ time in the worst case. The concepts of this minimization algorithm are outside the scope of this monograph and can be found in any textbook on automata theory. However, in Chapter 7 a very similar partitioning-refinement algorithm will be presented for bisimulation minimization.

### 4.2 Model-Checking Regular Safety Properties

In this section, it will be shown how NFAs can be used to check the validity of an important class of safety properties. The main characteristic of these safety properties is that all their bad prefixes constitute a regular language. The bad prefixes of these so-called regular safety properties can thus be recognized by an NFA. The main result of this section is that checking a regular safety property on a finite transition system can be reduced to invariant checking on the product of $T S$ and an NFA $\mathcal{A}$ for the bad prefixes. Stated differently, if one wants to check whether a regular safety property holds for TS, it suffices to perform a reachability analysis in the product $T S \otimes \mathcal{A}$ to check a corresponding invariant on $T S \otimes \mathcal{A}$.

### 4.2.1 Regular Safety Properties

Recall that safety properties are LT properties, i.e., sets of infinite words over $2^{A P}$, such that every trace that violates a safety property has a bad prefix that causes a refutation (cf. Definition 3.22 on page 112). Bad prefixes are finite, and thus the set of bad prefixes constitutes a language of finite words over the alphabet $\Sigma=2^{A P}$. That is, the input symbols $A \in \Sigma$ of the NFA are now sets of atomic propositions. For instance, if $A P=$ $\{a, b\}$, then $\Sigma=\left\{A_{1}, A_{2}, A_{3}, A_{4}\right\}$ consists of the four input symbols $A_{1}=\{ \}, A_{2}=\{a\}$, $A_{3}=\{b\}$, and $A_{4}=\{a, b\} .{ }^{1}$

## Definition 4.11. Regular Safety Property

Safety property $P_{\text {safe }}$ over AP is called regular if its set of bad prefixes constitutes a regular language over $2^{A P}$.

Every invariant is a regular safety property. If $\Phi$ is the state condition (propositional formula) of the invariant that should be satisfied by all reachable states, then the language of bad prefixes consists of the words $A_{0} A_{1} \ldots A_{n}$ such that $A_{i} \not \vDash \Phi$ for some $0 \leqslant i \leqslant n$. Such languages are regular, since they can be characterized by the (casually written) regular notation

$$
\Phi^{*}(\neg \Phi) \text { true*. }
$$

Here, $\Phi$ stands for the set of all $A \subseteq A P$ with $A \models \Phi, \neg \Phi$ for the set of all $A \subseteq A P$ with $A \not \vDash \Phi$, while true means the set of all subsets $A$ of $A P$. For instance, if $A P=\{a, b\}$ and $\Phi=a \vee \neg b$, then

[^15]- $\Phi$ stands for the regular expression $\}+\{a\}+\{a, b\}$,
- $\neg \Phi$ stands for the regular expression consisting of the symbol $\{b\}$,
- true stands for the regular expression $\}+\{a\}+\{b\}+\{a, b\}$.

The bad prefixes of the invariant over condition $a \vee \neg b$ are given by the regular expression:

$$
\mathrm{E}=\underbrace{(\{ \}+\{a\}+\{a, b\})^{*}}_{\Phi^{*}} \underbrace{\{b\}}_{\neg \Phi} \underbrace{(\{ \}+\{a\}+\{b\}+\{a, b\})^{*}}_{\text {true }^{*}} .
$$

Thus, $\mathcal{L}(\mathrm{E})$ consists of all words $A_{1} \ldots A_{n}$ such that $A_{i}=\{b\}$ for some $1 \leqslant i \leqslant n$. Note that, for $A \subseteq A P=\{a, b\}$, we have $A \not \vDash a \vee \neg b$ if and only if $A=\{b\}$. Hence, $\mathcal{L}(\mathrm{E})$ agrees with the set of bad prefixes for the invariant induced by the condition $\Phi$.


Figure 4.3: NFA accepting all bad prefixes of the invariant over the condition $\Phi$.
In fact, for any invariant $P_{\text {inv }}$, the language of all bad prefixes can be represented by an NFA with two states, as shown in Figure 4.3. Here and in the sequel, we use symbolic notations in the pictures for NFAs over the alphabet $2^{A P}$. We use propositional formulae over $A P$ as labels for the edges. Thus, an edge leading from state $q$ to state $q^{\prime}$ labeled with formula $\Psi$ means that there are transitions $q \xrightarrow{A} q^{\prime}$ for all $A \subseteq A P$ where $A \models \Psi$. E.g., if $A P=\{a, b\}$ and $\Phi=a \vee \neg b$, then Figure 4.3 is a representation for an NFA with two states $q_{0}, q_{1}$ and the transitions

$$
q_{0} \xrightarrow{\{ \}} q_{0}, \quad q_{0} \xrightarrow{\{a\}} q_{0}, \quad q_{0} \xrightarrow{\{a, b\}} q_{0}, \quad q_{0} \xrightarrow{\{b\}} q_{1}
$$

and

$$
q_{1} \xrightarrow{\{ \}} q_{1}, \quad q_{1} \xrightarrow{\{a\}} q_{1}, \quad q_{1} \xrightarrow{\{b\}} q_{1}, \quad q_{1} \xrightarrow{\{a, b\}} q_{1} .
$$

For the invariant over $A P=\{a, b\}$ induced by the condition $\Phi=a \vee \neg b$, the minimal bad prefixes are described by the regular expression $\left(\}+\{a\}+\{a, b\})^{*}\{b\}\right.$. Hence, the minimal bad prefixes constitute a regular language too. An automaton that recognizes all minimal bad prefixes, which are given by the regular expression $\Phi^{*}(\neg \Phi)$, is obtained from Figure 4.3 by omitting the self-loop of state $q_{1}$. In fact, for the definition of regular safety properties it is irrelevant whether the regularity of the set of all bad prefixes or of the set of all minimal bad prefixes is required:

## Lemma 4.12. Criterion for Regularity of Safety Properties

The safety property $P_{\text {safe }}$ is regular if and only if the set of minimal bad prefixes for $P_{\text {safe }}$ is regular.

Proof: "if": Let $\mathcal{A}=\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$ be an NFA for $\operatorname{MinBadPref}\left(P_{\text {safe }}\right)$. Then, an NFA for $\operatorname{BadPref}\left(P_{\text {safe }}\right)$ is obtained by adding self-loops $q \xrightarrow{A} q$ for all states $q \in F$ and all $A \subseteq A P$. It is easy to check that the modified NFA accepts the language consisting of all bad prefixes for $P_{\text {safe }}$. Thus, $\operatorname{BadPref}\left(P_{\text {safe }}\right)$ is regular, which yields the claim.
"only if": Let $\mathcal{A}=\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$ be a DFA for $\operatorname{BadPref}\left(P_{\text {safe }}\right)$. For $\operatorname{MinBadPref}\left(P_{\text {safe }}\right)$, a DFA is obtained by removing all outgoing transitions from the accept states in $\mathcal{A}$. Let $\mathcal{A}^{\prime}$ be the modified DFA and let us check that $\mathcal{L}\left(\mathcal{A}^{\prime}\right)=\operatorname{MinBadPref}\left(P_{\text {safe }}\right)$.

If we are given a word $w=A_{1} \ldots A_{n} \in \mathcal{L}\left(\mathcal{A}^{\prime}\right)$, then $w \in \mathcal{L}(\mathcal{A})$ since the run $q_{0} q_{1} \ldots q_{n}$ in $\mathcal{A}^{\prime}$ for $w$ is also an accepting run in $\mathcal{A}$. Therefore, $w$ is a bad prefix for $P_{\text {safe }}$. Distinguish two cases.

Assume $w$ is not a minimal bad prefix. Then there exists a proper prefix $A_{1} \ldots A_{i}$ of $w$ that is a bad prefix for $P_{\text {safe }}$. Thus, $A_{1} \ldots A_{i} \in \mathcal{L}(\mathcal{A})$. Since $\mathcal{A}$ is deterministic, $q_{0} q_{1} \ldots q_{i}$ is the (unique) run for $A_{1} \ldots A_{i}$ in $\mathcal{A}$ and $q_{i} \in F$. Since $i<n$ and $q_{i}$ has no outgoing transitions in $\mathcal{A}^{\prime}, q_{0} \ldots q_{i} \ldots q_{n}$ cannot be a run for $A_{1} \ldots A_{i} \ldots A_{n}$ in $\mathcal{A}^{\prime}$. This contradicts the assumption and shows that $A_{1} \ldots A_{n}$ is a minimal bad prefix for $P_{\text {safe }}$.

Vice versa, if $w$ is a minimal bad prefix for $P_{\text {safe }}$, then
(1) $A_{1} \ldots A_{n} \in \operatorname{BadPref}\left(P_{\text {safe }}\right)=\mathcal{L}(\mathcal{A})$ and
(2) $A_{1} \ldots A_{i} \notin \operatorname{BadPref}\left(P_{\text {safe }}\right)=\mathcal{L}(\mathcal{A})$ for all $1 \leqslant i<n$.

Let $q_{0} \ldots q_{n}$ be the unique run for $w$ in $\mathcal{A}$. Then, (2) yields $q_{i} \notin F$ for $1 \leqslant i<n$, while $q_{n} \in F$ by (1). Thus, $q_{0} \ldots q_{n}$ is an accepting run for $w$ in $\mathcal{A}^{\prime}$ which yields $w \in \mathcal{L}\left(\mathcal{A}^{\prime}\right)$.

## Example 4.13. Regular Safety Property for Mutual Exclusion Algorithms

Consider a mutual exclusion algorithm such as the semaphore-based one or Peterson's algorithm. The bad prefixes of the safety property $P_{\text {mutex }}$ ("there is always at most one process in its critical section") constitute the language of all finite words $A_{0} A_{1} \ldots A_{n}$ such that

$$
\left\{\text { crit }_{1}, \text { crit }_{2}\right\} \subseteq A_{i}
$$

for some index $i$ with $0 \leqslant i \leqslant n$. If $i=n$ is the smallest such index, i.e., $\left\{\right.$ crit $_{1}$, crit $\left._{2}\right\} \subseteq A_{n}$ and $\left\{\right.$ crit $_{1}$, crit $\left._{2}\right\} \nsubseteq A_{j}$ for $0 \leqslant j<n$, then $A_{0} \ldots A_{n}$ is a minimal bad prefix. The language of all (minimal) bad prefixes is regular. An NFA recognizing all minimal bad prefixes is depicted in Figure 4.4.


Figure 4.4: Minimal bad prefixes that refute the mutual exclusion property.

## Example 4.14. Regular Safety Property for the Traffic Light

Consider a traffic light with three possible colors: red, yellow and green. The property "a red phase must be preceded immediately by a yellow phase" is specified by the set of infinite words $\sigma=A_{0} A_{1} \ldots$ with $A_{i} \subseteq\{$ red, yellow $\}$ such that for all $i \geqslant 0$ we have that

$$
\text { red } \in A_{i} \text { implies } i>0 \text { and yellow } \in A_{i-1} .
$$

The bad prefixes are finite words that violate this condition. Examples of bad prefixes that are minimal are

$$
\}\}\{\text { red }\} \quad \text { and } \quad\}\{\text { red }\} .
$$

In general, the minimal bad prefixes are words of the form $A_{0} A_{1} \ldots A_{n}$ such that $n>0$, red $\in A_{n}$, and yellow $\notin A_{n-1}$. The NFA in Figure 4.5 accepts these minimal bad prefixes. Recall the meaning of the edge labels in the pictorial representations of an NFA over the


Figure 4.5: Minimal bad prefixes in which red is not preceded by yellow.
alphabet $\Sigma=2^{A P}$ where $A P=\{$ yellow,red $\}$. For instance, the edge-label yellow in the self-loop of state $q_{1}$ denotes a formula, namely the positive literal yellow $\in A P$. This stands for all sets $A \subseteq A P=\{$ yellow,red $\}$ where the literal yellow holds, that is, the sets \{yellow \} and $\{$ yellow, red $\}$. Hence, the self-loop of $q_{1}$ in the picture stands for the two
transitions:

$$
q_{1} \xrightarrow{\{\text { yellow }\}} q_{1} \text { and } q_{1} \xrightarrow{\{\text { yellow }, \text { red }\}} q_{1} .
$$

Similarly, the edge label $\neg$ yellow in the transition from $q_{1}$ to $q_{0}$ stands for the negative literal $\neg$ yellow, and thus represents the transitions:

$$
q_{1} \xrightarrow{\{\mathrm{red}\}} q_{0} \quad \text { and } \quad q_{1} \xrightarrow{\{ \}} q_{0} .
$$

In the same way the label red of the transition from $q_{0}$ to $q_{2}$ represents two transitions (with the labels $\{$ red $\}$ and $\{$ red, yellow $\}$ ), while the edge labels $\neg$ red $\wedge$ yellow and $\neg$ red $\wedge \neg$ yellow denote only a single symbol in $2^{A P}$, namely $\{$ yellow $\}$ and $\}$, respectively.

## Example 4.15. A Nonregular Safety Property

Not all safety properties are regular. As an example of a nonregular safety property, consider:
"The number of inserted coins is always at least the number of dispensed drinks."
(See also Example 3.24 on page 113). Let the set of propositions be $\{$ pay, drink $\}$. Minimal bad prefixes for this safety property constitute the language

$$
\left\{\text { pay }^{n} \text { drink }^{n+1} \mid n \geqslant 0\right\}
$$

which is not a regular, but a context-free language. Such safety properties fall outside the scope of the following verification algorithm.

### 4.2.2 Verifying Regular Safety Properties

Let $P_{\text {safe }}$ be a regular safety property over the atomic propositions $A P$ and $\mathcal{A}$ an NFA recognizing the (minimal) bad prefixes of $P_{\text {safe }}$. (Recall that by Lemma 4.12 on page 161 it is irrelevant whether $\mathcal{A}$ accepts all bad prefixes for $P_{\text {safe }}$ or only the minimal ones.) For technical reasons, we assume that $\varepsilon \notin \mathcal{L}(\mathcal{A})$. In fact, this is not a severe restriction since otherwise all finite words over $2^{A P}$ are bad prefixes, and hence, $P_{\text {safe }}=\varnothing$. In this case, $T S \models P_{\text {safe }}$ if and only if $T S$ has no initial state.

Furthermore, let TS be a finite transition system without terminal states with corresponding set of propositions $A P$. In this section, we aim to establish an algorithmic method for verifying whether $T S$ satisfies regular safety property $P_{\text {safe }}$, i.e., to check whether
$T S \models P_{\text {safe }}$ holds. According to Lemma 3.25 on page 114 we have

$$
\begin{gathered}
T S \models P_{\text {safe }} \quad \text { if and only if } \quad \operatorname{Traces}_{f i n}(T S) \cap \operatorname{BadPref}\left(P_{\text {safe }}\right)=\varnothing \\
\\
\text { if and only if } \quad \operatorname{Traces}_{f i n}(T S) \cap \mathcal{L}(\mathcal{A})=\varnothing .
\end{gathered}
$$

Thus, it suffices to check whether $\operatorname{Traces}_{f i n}(T S) \cap \mathcal{L}(\mathcal{A})=\varnothing$ to establish $T S \models P_{\text {safe }}$.
To do so, we adopt a similar strategy as for checking whether two NFAs intersect. Recall that in order to check whether the NFAs $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ do intersect, it suffices to consider their product automaton, so

$$
\mathcal{L}\left(\mathcal{A}_{1}\right) \cap \mathcal{L}\left(\mathcal{A}_{2}\right)=\varnothing \quad \text { if and only if } \quad \mathcal{L}\left(\mathcal{A}_{1} \otimes \mathcal{A}_{2}\right)=\varnothing
$$

The question whether two automata do intersect is thus reduced to a simple reachability problem in the product automaton.

This is now exploited as follows. In order to check whether $\operatorname{Traces}_{\text {fin }}(T S) \cap \mathcal{L}(\mathcal{A})=\varnothing$, we first build a product of transition system $T S$ and NFA $\mathcal{A}$ in the same vein as the synchronous product of NFA. This yields the transition system $T S \otimes \mathcal{A}$. For this transition system, an invariant can be given using a propositional logic formula $\Phi$-derived from the accept states of $\mathcal{A}$ - such that $\operatorname{Traces}_{f i n}(T S) \cap \mathcal{L}(\mathcal{A})=\varnothing$ if and only if $T S \otimes \mathcal{A} \models$ "always $\Phi$ ". In this way, the verification of a regular safety property is reduced to invariant checking. Recall that for checking invariants, Algorithm 4 (see page 110) can be exploited.

We start by formally defining the product between a transition system $T S$ and an NFA $\mathcal{A}$, denoted $T S \otimes \mathcal{A}$. Let $T S=(S, A c t, \rightarrow, I, A P, L)$ and $\mathcal{A}=\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$ with $Q_{0} \cap F=\varnothing$. Recall that the alphabet of $\mathcal{A}$ consists of sets of atomic proposition in TS. Transition system $T S \otimes \mathcal{A}$ has state space $S \times Q$ and a transition relation such that each path fragment $\pi=s_{0} s_{1} \ldots s_{n}$ in $T S$ can be extended to a path fragment

$$
\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots\left\langle s_{n}, q_{n+1}\right\rangle
$$

in $T S \otimes \mathcal{A}$ which has an initial state $q_{0} \in Q_{0}$ for which

$$
q_{0} \xrightarrow{L\left(s_{0}\right)} q_{1} \xrightarrow{L\left(s_{1}\right)} q_{2} \xrightarrow{L\left(s_{2}\right)} \ldots \xrightarrow{L\left(s_{n}\right)} q_{n+1}
$$

is a run-not necessarily accepting-of NFA $\mathcal{A}$ that generates the word

$$
\operatorname{trace}(\pi)=L\left(s_{0}\right) L\left(s_{1}\right) \ldots L\left(s_{n}\right)
$$

Finally, labels of states are state names of $\mathcal{A}$. These considerations lead to the following definition:

## Definition 4.16. Product of Transition System and NFA

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system without terminal states and $\mathcal{A}=$ $\left(Q, \Sigma, \delta, Q_{0}, F\right)$ an NFA with the alphabet $\Sigma=2^{A P}$ and $Q_{0} \cap F=\varnothing$. The product transition system $T S \otimes \mathcal{A}$ is defined as follows:

$$
T S \otimes \mathcal{A}=\left(S^{\prime}, A c t, \rightarrow^{\prime}, I^{\prime}, A P^{\prime}, L^{\prime}\right)
$$

where

- $S^{\prime}=S \times Q$,
- $\rightarrow^{\prime}$ is the smallest relation defined by the rule

$$
\frac{s \xrightarrow{\alpha} t \wedge q \xrightarrow{L(t)} p}{\langle s, q\rangle \xrightarrow{\alpha}\langle t, p\rangle},
$$

- $I^{\prime}=\left\{\left\langle s_{0}, q\right\rangle \mid s_{0} \in I \wedge \exists q_{0} \in Q_{0} . q_{0} \xrightarrow{L\left(s_{0}\right)} q\right\}$,
- $A P^{\prime}=Q$, and
- $L^{\prime}: S \times Q \rightarrow 2^{Q}$ is given by $L^{\prime}(\langle s, q\rangle)=\{q\}$.


## Remark 4.17. Terminal States

For the definition of LT properties (and thus of invariants) we have assumed transition systems to have no terminal states. It is, however, not guaranteed that $T S \otimes \mathcal{A}$ possesses this property, even if $T S$ does. This stems from the fact that in NFA $\mathcal{A}$ there may be a state $q$, say, that has no direct successor states for some set $A$ of atomic propositions, i.e., with $\delta(q, A)=\varnothing$. This technical problem can be treated by either requiring $\delta(q, A) \neq \varnothing$ for all states $q \in Q$ and $A \subseteq A P$ or by extending the notion of invariants to arbitrary transition systems. Note that imposing the requirement $\delta(q, A) \neq \varnothing$ is not a severe restriction, as any NFA can be easily transformed into an equivalent one that satisfies this property by introducing a state $q_{\text {trap }}$ and adding transition $q \xrightarrow{A} q_{\text {trap }}$ to $\mathcal{A}$ whenever $\delta(q, A)=\varnothing$ or $q=q_{\text {trap }}$. We finally remark that for the algorithm for invariant checking, it is not of any relevance whether terminal states exist or not.

## Example 4.18. A Product Automaton

The language of the minimal bad prefixes of the safety property "each red light phase


Figure 4.6: German traffic light (left upper figure), an NFA (right upper figure), and their product (lower figure).
is preceded by a yellow light phase" is accepted by the DFA $\mathcal{A}$ indicated in Example 4.14 (page 162). We consider a German traffic light, which besides the usual possible colors red, green, and yellow, has the possibility to indicate red and yellow simultaneously indicating "green light soon". The transition system GermanTrLight thus has four states with the usual transitions red $\rightarrow$ red+yellow, red+yellow $\rightarrow$ green, green $\rightarrow$ yellow, and yellow $\rightarrow$ red. Let $A P=\{$ red, yellow $\}$ indicating the corresponding light phases. The labeling is defined as follows: $L($ red $)=\{$ red $\}, L($ yellow $)=\{$ yellow $\}, L($ green $)=$ $\varnothing=L($ red + yellow $)$. The product transition system GermanTrLight $\otimes \mathcal{A}$ consists of four reachable states (see Figure 4.6). As action labels are not relevant here, they are omitted.

The following theorem shows that the verification of a regular safety property can be reduced to checking an invariant in the product.

Let $T S$ and $\mathcal{A}$ be as before. Let $P_{\operatorname{inv}(\mathcal{A})}$ be the invariant over $A P^{\prime}=2^{Q}$ which is defined by the propositional formula

$$
\bigwedge_{q \in F} \neg q .
$$

In the sequel, we often write $\neg F$ as shorthand for $\bigwedge_{q \in F} \neg q$. Stated in words, $\neg F$ holds in all nonaccept states.

## Theorem 4.19. Verification of Regular Safety Properties

For transition system TS over AP, NFA $\mathcal{A}$ with alphabet $2^{A P}$ as before, and regular safety property $P_{\text {safe }}$ over AP such that $\mathcal{L}(\mathcal{A})$ equals the set of (minimal) bad prefixes of $P_{\text {safe }}$, the following statements are equivalent:
(a) $T S \models P_{\text {safe }}$
(b) $\operatorname{Traces}_{f i n}(T S) \cap \mathcal{L}(\mathcal{A})=\varnothing$
(c) $T S \otimes \mathcal{A} \models P_{\operatorname{inv}(A)}$

Proof: Let $T S=(S, A c t, \rightarrow, I, A P, L)$ and $\mathcal{A}=\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$.
The equivalence of (a) and (b) follows immediately by Lemma 3.25 (page 114). To establish the equivalence of (a), (b), and (c), we show

$$
(c) \Longrightarrow(a): T S \not \vDash P_{\text {safe }} \quad \text { implies } T S \otimes \mathcal{A} \not \models P_{\operatorname{inv}(A)}
$$

and

$$
(b) \Longrightarrow(c): T S \otimes \mathcal{A} \not \vDash P_{i n v(A)} \quad \text { implies } \quad \operatorname{Traces}_{f i n}(T S) \cap \mathcal{L}(\mathcal{A}) \neq \varnothing
$$

Proof of "(c) $\Longrightarrow$ (a)": If $T S \not \vDash P_{\text {safe }}$, then there is a finite initial path fragment $\widehat{\pi}=$ $s_{0} s_{1} \ldots s_{n}$ in $T S$ with

$$
\operatorname{trace}(\widehat{\pi})=L\left(s_{0}\right) L\left(s_{1}\right) \ldots L\left(s_{n}\right) \in \mathcal{L}(\mathcal{A})
$$

Since $\operatorname{trace}(\widehat{\pi}) \in \mathcal{L}(\mathcal{A})$, there exists an accepting run $q_{0} q_{1} \ldots q_{n+1}$ of $\mathcal{A}$ for $\operatorname{trace}(\widehat{\pi})$. Accordingly

$$
q_{0} \in Q_{0} \text { and } q_{i} \xrightarrow{L\left(s_{i}\right)} q_{i+1} \text { for all } 0 \leqslant i \leqslant n, \text { and } q_{n+1} \in F
$$

Thus, $\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots\left\langle s_{n}, q_{n+1}\right\rangle$ is an initial path fragment in $T S \otimes \mathcal{A}$ with

$$
\left\langle s_{n}, q_{n+1}\right\rangle \nLeftarrow \neg F .
$$

It thus follows that $T S \otimes \mathcal{A} \not \vDash P_{\operatorname{inv}(A)}$.
Proof of " $(\mathrm{b}) \Longrightarrow(\mathrm{c})$ ": Let $T S \otimes \mathcal{A} \not \vDash P_{\operatorname{inv}(A)}$. Then there exists an initial path fragment

$$
\left\langle s_{0}, q_{1}\right\rangle \ldots\left\langle s_{n}, q_{n+1}\right\rangle
$$

in $T S \otimes \mathcal{A}$ with $q_{n+1} \in F$, and $q_{1}, \ldots, q_{n} \notin F$. Besides, $s_{0} s_{1} \ldots s_{n}$ is an initial path fragment in $T S$. Further,

$$
q_{i} \xrightarrow{L\left(s_{i}\right)} q_{i+1} \text { for all } 0 \leqslant i \leqslant n
$$

Since $\left\langle s_{0}, q_{1}\right\rangle$ is an initial state of $T S \otimes \mathcal{A}$, there is an initial state $q_{0}$ in $\mathcal{A}$ such that $q_{0} \xrightarrow{L\left(s_{0}\right)} q_{1}$. Sequence $q_{0} q_{1} \ldots q_{n+1}$ is thus an accepting run for trace $\left(s_{0} s_{1} \ldots s_{n}\right)$. Thus,

$$
\operatorname{trace}\left(s_{0} s_{1} \ldots s_{n}\right) \in \operatorname{Traces}_{f i n}(T S) \cap \mathcal{L}(\mathcal{A})
$$

which yields $\operatorname{Traces}_{f i n}(T S) \cap \mathcal{L}(\mathcal{A}) \neq \varnothing$.

Stated in words, Theorem 4.19 yields that in order to check the transition system TS versus the regular safety property $P_{\text {safe }}$, it suffices to check whether no state $\langle s, q\rangle$ in $T S \otimes \mathcal{A}$ is reachable where the $\mathcal{A}$-component $q$ is an accept state in $\mathcal{A}$. This invariant "visit never an accept state in $\mathcal{A}$ " (formally given by the invariant condition $\Phi=\neg F$ ) can be checked using a depth-first search approach as described in detail in Algorithm 4 (page 110). Note that in case the safety property is refuted, the invariant checking algorithm provides a counterexample. This counterexample is in fact a finite path fragment $\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots\left\langle s_{n}, q_{n+1}\right\rangle$ in the transition system $T S \otimes \mathcal{A}$ that leads to an accept state. The projection to the states in $T S$ yields an initial finite path fragment $s_{0} s_{1} \ldots s_{n}$ in $T S$ where the induced trace $\operatorname{trace}\left(s_{0} s_{1} \ldots s_{n}\right) \in\left(2^{A P}\right)^{*}$ is accepted by $\mathcal{A}$ (since it has an accepting run of the form $\left.q_{0} q_{1} \ldots q_{n+1}\right)$. Thus, $\operatorname{trace}\left(s_{0} s_{1} \ldots s_{n}\right)$ is a bad prefix for $P_{\text {safe }}$. Hence, $s_{0} s_{1} \ldots s_{n}$ yields a useful error indication since trace $(\pi) \notin P_{\text {safe }}$ for all paths $\pi$ in $T S$ that start with the prefix $s_{0} s_{1} \ldots s_{n}$.

## Corollary 4.20.

Let TS, $\mathcal{A}$, and $P_{\text {safe }}$ be as in Theorem 4.19. Then, for each initial path fragment $\left\langle s_{0}, q_{1}\right\rangle \ldots$ $\left\langle s_{n}, q_{n+1}\right\rangle$ of $T S \otimes \mathcal{A}$ :

$$
q_{1}, \ldots, q_{n} \notin F \text { and } q_{n+1} \in F \quad \text { implies } \quad \text { trace }\left(s_{0} s_{1} \ldots s_{n}\right) \in \mathcal{L}(\mathcal{A})
$$

As a result, the skeleton in Algorithm 5 can be used to check a regular safety property against a transition system and to report a counterexample (i.e., finite initial path fragment in TS inducing a bad prefix) as diagnostic feedback if the safety property does not hold for TS.

Example 4.21. Checking a Regular Safety Property for the Traffic Light
Consider again the German traffic light system and the regular safety property $P_{\text {safe }}$

Algorithm 5 Model-checking algorithm for regular safety properties
Input: finite transition system $T S$ and regular safety property $P_{\text {safe }}$
Output: true if $T S \models P_{\text {safe }}$. Otherwise false plus a counterexample for $P_{\text {safe }}$.

```
Let NFA \(\mathcal{A}\) (with accept states \(F\) ) be such that \(\mathcal{L}(\mathcal{A})=\) bad prefixes of \(P_{\text {safe }}\)
Construct the product transition system \(T S \otimes \mathcal{A}\)
Check the invariant \(P_{\operatorname{inv}(\mathcal{A})}\) with proposition \(\neg F=\bigwedge_{q \in F} \neg q\) on \(T S \otimes \mathcal{A}\).
if \(T S \otimes \mathcal{A} \models P_{\text {inv }(\mathcal{A})}\) then
    return true
else
    Determine an initial path fragment \(\left\langle s_{0}, q_{1}\right\rangle \ldots\left\langle s_{n}, q_{n+1}\right\rangle\) of \(T S \otimes \mathcal{A}\) with \(q_{n+1} \in F\)
    return (false, \(s_{0} s_{1} \ldots s_{n}\) )
fi
```

that each red light phase should be immediately preceded by a yellow light phase. The transition system of the traffic light, the NFA accepting the bad prefixes of the safety property, as well as their product automaton, are depicted in Figure 4.6 (page 166). To check the validity of $P_{\text {safe }}$, only the second component of the states $\langle s, q\rangle$ is relevant. The fact that no state of the form $\left\langle\ldots, q_{F}\right\rangle$ is reachable ensures the invariant $\neg q_{F}$ to hold in all reachable states. Thus GermanTrLight $\models P_{\text {safe }}$.

If the traffic light is modified such that the state "red" is the initial state (instead of "green"), then we obtain a transition system that violates $P_{\text {safe }}$. Actually, in this case the invariant $\neg q_{F}$ is already violated in the initial state of the resulting product transition system that has the following form:

$$
\left\langle r e d, \delta\left(q_{0},\{r e d\}\right)\right\rangle=\left\langle r e d, q_{F}\right\rangle
$$

We conclude this part by considering the worst-case time and space complexity of the automata-based algorithm for checking regular safety properties.

## Theorem 4.22. Complexity of Verifying Regular Safety Properties

The time and space complexity of Algorithm 5 is in $\mathcal{O}(|T S| \cdot|\mathcal{A}|)$ where $|T S|$ and $|\mathcal{A}|$ denote the number of states and transitions in TS and $\mathcal{A}$, respectively.

Assuming an generation of the reachable states of $T S$ from a syntactic description of the processes, the above bound also holds if $|T S|$ denotes the size of the reachable fragment of TS.

Proof: Follows directly from the fact that the number of states in the product automaton $T S \otimes \mathcal{A}$ is in $\mathcal{O}(|S| \cdot|Q|)$ (where $S$ and $Q$ denote the state space of $T S$ and $\mathcal{A}$, respectively) and the fact that the time and space complexity of invariant checking is linear in the number of states and transitions of the transition system $T S \otimes \mathcal{A}$. (Thus, we can even establish the bound $\mathcal{O}(|S| \cdot|Q|+|\rightarrow| \cdot|\delta|)$ for the runtime where $|\rightarrow|$ denotes the number of transitions in $T S$ and $|\delta|$ the number of transitions in $\mathcal{A}$.)

### 4.3 Automata on Infinite Words

Finite-state automata accept finite words, i.e., sequences of symbols of finite length, and yield the basis for checking regular safety properties. In this and the following sections, these ideas are generalized toward a more general class of LT properties. These include regular safety, various liveness properties, but also many other properties that are relevant to formalize the requirements for "realistic" systems. The rough idea is to consider variants of NFAs, called nondeterministic Büchi automata (NBAs), which serve as acceptors for languages of infinite words. It will be established that if we are given a nondeterministic Büchi automaton $\mathcal{A}$ that specifies the "bad traces" (i.e., that accepts the complement of the LT property $P$ to be verified), then a graph analysis in the product of the given transition system $T S$ and the automaton $\mathcal{A}$ suffices to either establish or disprove $T S \vDash P$. Whereas for regular safety properties a reduction to invariant checking (i.e., a depth-first search) is possible, the graph algorithms needed here serve to check a so-called persistence property. Such properties state that eventually for ever a certain proposition holds.

We first introduce the " $\omega$-counterpart" to regular languages, both by introducing $\omega$-regular expressions (Section 4.3.1) and nondeterministic Büchi automata (see Section 4.3.2). Variants of nondeterministic Büchi automata will be discussed in Sections 4.3.3 and 4.3.4.

### 4.3.1 $\omega$-Regular Languages and Properties

Infinite words over the alphabet $\Sigma$ are infinite sequences $A_{0} A_{1} A_{2} \ldots$ of symbols $A_{i} \in \Sigma$. $\Sigma^{\omega}$ denotes the set of all infinite words over $\Sigma$. As before, the Greek letter $\sigma$ will be used for infinite words, while $w, v, u$ range over finite words. Any subset of $\Sigma^{\omega}$ is called a language of infinite words, sometimes also called an $\omega$-language. In the sequel, the notion of a language will be used for any subset of $\Sigma^{*} \cup \Sigma^{\omega}$. Languages will be denoted by the symbol $\mathcal{L}$.

To reason about languages of infinite words, the basic operations of regular expressions (union, concatenation, and finite repetition) are extended by infinite repetition, denoted by the Greek letter $\omega .^{2}$ For instance, the infinite repetition of the finite word $A B$ yields the infinite word $A B A B A B A B A B \ldots($ ad infinitum $)$ and is denoted by $(A B)^{\omega}$. For the special case of the empty word, we have $\varepsilon^{\omega}=\varepsilon$. For an infinite word, infinite repetition has no effect, that is, $\sigma^{\omega}=\sigma$ if $\sigma \in \Sigma^{\omega}$. Note that the finite repetition of a word results in a language of finite words, i.e., a subset of $\Sigma^{*}$, whereas infinite repetition of a (finite or infinite) word results in a single word.

Infinite repetition can be lifted to languages as follows. For language $\mathcal{L} \subseteq \Sigma^{*}$, let $\mathcal{L}^{\omega}$ be the set of words in $\Sigma^{*} \cup \Sigma^{\omega}$ that arise from the infinite concatenation of (arbitrary) words in $\Sigma$, i.e.,

$$
\mathcal{L}^{\omega}=\left\{w_{1} w_{2} w_{3} \ldots \mid w_{i} \in \mathcal{L}, i \geqslant 1\right\} .
$$

The result is an $\omega$-language, provided that $\mathcal{L} \subseteq \Sigma^{+}$, i.e., $\mathcal{L}$ does not contain the empty word $\varepsilon$. However, in the sequel, we only need the $\omega$-operator applied to languages of finite words that do not contain the empty word. In this case, i.e., for $\mathcal{L} \subseteq \Sigma^{+}$, we have $\mathcal{L}^{\omega} \subseteq \Sigma^{\omega}$.

In the following definition, the concatenation operator $\mathcal{L}_{1} \cdot \mathcal{L}_{2}$ is used that combines a language $\mathcal{L}_{1}$ of finite words with a language $\mathcal{L}_{2}$ of infinite words. It is defined by $\mathcal{L}_{1} \cdot \mathcal{L}_{2}=$ $\left\{w \sigma \mid w \in \mathcal{L}_{1}, \sigma \in \mathcal{L}_{2}\right\}$.

## Definition 4.23. $\omega$-Regular Expression

An $\omega$-regular expression $G$ over the alphabet $\Sigma$ has the form

$$
\mathrm{G}=\mathrm{E}_{1} \cdot \mathrm{~F}_{1}^{\omega}+\ldots+\mathrm{E}_{n} \cdot \mathrm{~F}_{n}^{\omega}
$$

where $n \geqslant 1$ and $\mathrm{E}_{1}, \ldots, \mathrm{E}_{n}, \mathrm{~F}_{1}, \ldots, \mathrm{~F}_{n}$ are regular expressions over $\Sigma$ such that $\varepsilon \notin \mathcal{L}\left(\mathrm{F}_{i}\right)$, for all $1 \leqslant i \leqslant n$.

The semantics of the $\omega$-regular expression G is a language of infinite words, defined by

$$
\mathcal{L}_{\omega}(\mathrm{G})=\mathcal{L}\left(\mathrm{E}_{1}\right) \cdot \mathcal{L}\left(\mathrm{F}_{1}\right)^{\omega} \cup \ldots \cup \mathcal{L}\left(\mathrm{E}_{n}\right) \cdot \mathcal{L}\left(\mathrm{F}_{n}\right)^{\omega}
$$

where $\mathcal{L}(E) \subseteq \Sigma^{*}$ denotes the language (of finite words) induced by the regular expression E (see page 914).

Two $\omega$-regular expressions $G_{1}$ and $G_{2}$ are equivalent, denoted $G_{1} \equiv G_{2}$, if $\mathcal{L}_{\omega}\left(G_{1}\right)=\mathcal{L}_{\omega}\left(G_{2}\right)$.

[^16]Examples for $\omega$-regular expressions over the alphabet $\Sigma=\{A, B, C\}$ are

$$
(A+B)^{*} A(A A B+C)^{\omega} \quad \text { or } \quad A(B+C)^{*} A^{\omega}+B(A+C)^{\omega} .
$$

If E is a regular expression with $\varepsilon \notin \mathcal{L}(\mathrm{E})$, then also $\mathrm{E}^{\omega}$ can be viewed as an $\omega$-regular expression since it can be identified with $\mathrm{E} . \mathrm{E}^{\omega}$ or $\varepsilon . \mathrm{E}^{\omega}$. Note that we have $\mathcal{L}(\mathrm{E})^{\omega}=$ $\mathcal{L}\left(\mathrm{E} . \mathrm{E}^{\omega}\right)=\mathcal{L}\left(\varepsilon . \mathrm{E}^{\omega}\right)$.

## Definition 4.24. $\omega$-Regular Language

A language $\mathcal{L} \subseteq \Sigma^{\omega}$ is called $\omega$-regular if $\mathcal{L}=\mathcal{L}_{\omega}(\mathrm{G})$ for some $\omega$-regular expression $G$ over $\Sigma$.

For instance, the language consisting of all infinite words over $\{A, B\}$ that contain infinitely many $A$ 's is $\omega$-regular since it is given by the $\omega$-regular expression $\left(B^{*} A\right)^{\omega}$. The language consisting of all infinite words over $\{A, B\}$ that contain only finitely many $A$ 's is $\omega$-regular too. A corresponding $\omega$-regular expression is $(A+B)^{*} B^{\omega}$. The empty set is $\omega$-regular since it is obtained, e.g., by the $\omega$-regular expression $\varnothing^{\omega}$. More generally, if $\mathcal{L} \subseteq \Sigma^{*}$ is regular and $\mathcal{L}^{\prime}$ is $\omega$-regular, then $\mathcal{L}^{\omega}$ and $\mathcal{L} . \mathcal{L}^{\prime}$ are $\omega$-regular.
$\omega$-Regular languages possess several closure properties: they are closed under union, intersection, and complementation. The argument for union is obvious from the definition by $\omega$-regular expressions. The proof for the intersection will be provided later; see Corollary 4.60 on page 198. The more advanced proof for complementation is not provided in this monograph. We refer the interested reader to [174] that covers also other properties of $\omega$-regular languages and various other automata models.

The concepts of $\omega$-regular languages play an important role in verification since most relevant LT properties are $\omega$-regular:

## Definition 4.25. $\omega$-Regular Properties

LT property $P$ over $A P$ is called $\omega$-regular if $P$ is an $\omega$-regular language over the alphabet $2^{A P}$.

For instance, for $A P=\{a, b\}$, the invariant $P_{\text {inv }}$ induced by the proposition $\Phi=a \vee \neg b$ is an $\omega$-regular property since

$$
\begin{aligned}
P_{\text {inv }} & =\left\{A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega} \mid \forall i \geqslant 0 .\left(a \in A_{i} \text { or } b \notin A_{i}\right)\right\} \\
& =\left\{A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega} \mid \forall i \geqslant 0 .\left(A_{i} \in\{\{ \},\{a\},\{a, b\}\}\right\}\right.
\end{aligned}
$$

is given by the $\omega$-regular expression $\mathrm{E}=(\{ \}+\{a\}+\{a, b\})^{\omega}$ over the alphabet $\Sigma=2^{A P}=$ $\{\},\{a\},\{b\},\{a, b\}\}$. In fact, any invariant over $A P$ is $\omega$-regular (the set $A P$ of atomic propositions is arbitrary) as it can be described by the $\omega$-regular expression $\Phi^{\omega}$ where $\Phi$ denotes the underlying propositional formula (that has to hold for all reachable states) and is identified with the regular expression given by the sum of all $A \subseteq A P$ with $A \models \Phi$.

Also, any regular safety property $P_{\text {safe }}$ is an $\omega$-regular property. This follows from the fact that the complement language

$$
\left(2^{A P}\right)^{\omega} \backslash P_{\text {safe }}=\underbrace{\operatorname{BadPref}\left(P_{\text {safe }}\right)}_{\text {regular }} \cdot\left(2^{A P}\right)^{\omega}
$$

is an $\omega$-regular language. The result that $\omega$-regular languages are closed under complementation (stated above, in the end of Section 4.3 .1 on page 172) yields the claim.

## Example 4.26. Mutual Exclusion

Another example of an $\omega$-regular property is the property given by the informal statement "process $\mathcal{P}$ visits its critical section infinitely often" which, for $A P=\{$ wait, crit $\}$, can be formalized by the $\omega$-regular expression:

$$
((\underbrace{\}+\{\text { wait }\}}_{\text {negative literal }- \text { crit }})^{*} \cdot(\underbrace{\{c r i t\}+\{\text { wait, crit }\}}_{\text {positive literal crit }}))^{\omega} .
$$

When allowing a somewhat sloppy notation using propositional formulae, the above expression may be rewritten into $\left((\neg \text { crit })^{*} \text {.crit }\right)^{\omega}$.

Starvation freedom in the sense of "whenever process $\mathcal{P}$ is waiting then it will enter its critical section eventually later" is an $\omega$-regular property as it can be described by

$$
\left((\neg \text { wait })^{*} . \text { wait.true }^{*} . c r i t\right)^{\omega}+\left((\neg \text { wait })^{*} . \text { wait.true }{ }^{*} . c r i t\right)^{*} .(\neg \text { wait })^{\omega}
$$

which is a short form for the $\omega$-regular expression over $A P=\{$ wait, crit $\}$ that results by replacing $\neg$ wait with $\}+\{$ crit $\}$, wait with $\{$ wait $\}+\{$ wait, crit $\}$, true with $\}+\{$ crit $\}+$ $\{$ wait $\}+\{$ wait, crit\}, and crit with $\{$ crit $\}+\{$ wait, crit $\}$. Intuitively, the first summand in the above expression stands for the case where $\mathcal{P}$ requests and enters its critical section infinitely often, while the second summand stands for the case where $\mathcal{P}$ is in its waiting phase only finitely many times.

### 4.3.2 Nondeterministic Büchi Automata

The issue now is to provide a kind of automaton that is suited for accepting $\omega$-regular languages. Finite automata are not adequate for this purpose as they operate on finite
words, while we need an acceptor for infinite words. Automata models that recognize languages of infinite words are called $\omega$-automata. The accepting runs of an $\omega$-automaton have to "check" the entire input word (and not just a finite prefix thereof), and thus have to be infinite. This implies that acceptance criteria for infinite runs are needed.

In this monograph, the simplest variant of $\omega$-automata, called nondeterministic Büchi automata (NBAs), suffices. The syntax of NBAs is exactly the same as for nondeterministic finite automata (NFAs). NBAs and NFAs differ, however, in their semantics: the accepted language of an NFA $\mathcal{A}$ is a language of finite words, i.e., $\mathcal{L}(\mathcal{A}) \subseteq \Sigma^{*}$, whereas the accepted language of NBA $\mathcal{A}$ (denoted $\mathcal{L}_{\omega}(\mathcal{A})$ is an $\omega$-language, i.e., $\mathcal{L}_{\omega}(\mathcal{A}) \subseteq \Sigma^{\omega}$. The intuitive meaning of the acceptance criterion named after Büchi is that the accept set of $\mathcal{A}$ (i.e., the set of accept states in $\mathcal{A}$ ) has to be visited infinitely often. Thus, the accepted language $\mathcal{L}_{\omega}(\mathcal{A})$ consists of all infinite words that have a run in which some accept state is visited infinitely often.

## Definition 4.27. Nondeterministic Büchi Automaton (NBA)

A nondeterministic Büchi automaton (NBA) $\mathcal{A}$ is a tuple $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ where

- $Q$ is a finite set of states,
- $\Sigma$ is an alphabet,
- $\delta: Q \times \Sigma \rightarrow 2^{Q}$ is a transition function,
- $Q_{0} \subseteq Q$ is a set of initial states, and
- $F \subseteq Q$ is a set of accept (or: final) states, called the acceptance set.

A run for $\sigma=A_{0} A_{1} A_{2} \ldots \in \Sigma^{\omega}$ denotes an infinite sequence $q_{0} q_{1} q_{2} \ldots$ of states in $\mathcal{A}$ such that $q_{0} \in Q_{0}$ and $q_{i} \xrightarrow{A_{i}} q_{i+1}$ for $i \geqslant 0$. Run $q_{0} q_{1} q_{2} \ldots$ is accepting if $q_{i} \in F$ for infinitely many indices $i \in \mathbb{N}$. The accepted language of $\mathcal{A}$ is

$$
\mathcal{L}_{\omega}(\mathcal{A})=\left\{\sigma \in \Sigma^{\omega} \mid \text { there exists an accepting run for } \sigma \text { in } \mathcal{A}\right\}
$$

The size of $\mathcal{A}$, denoted $|\mathcal{A}|$, is defined as the number of states and transitions in $\mathcal{A}$.

As for an NFA, we identify the transition function $\delta$ with the induced transition relation $\rightarrow \subseteq Q \times \Sigma \times Q$ which is given by

$$
q \xrightarrow{A} p \text { if and only if } p \in \delta(q, A) .
$$

Since the state space $Q$ of an NBA $\mathcal{A}$ is finite, each run for an infinite word $\sigma \in \Sigma^{\omega}$ is infinite, and hence visits some state $q \in Q$ infinitely often. Acceptance of a run depends on whether or not the set of all states that appear infinitely often in the given run contains an accept state. The definition of an NBA allows for the special case where $F=\varnothing$, which means that there are no accept states. Clearly, in this case, no run is accepting. Thus $\mathcal{L}_{\omega}(\mathcal{A})=\varnothing$ if $F=\varnothing$. There are also no accepting runs whenever, $Q_{0}=\varnothing$ as in this case, no word has a run.

Example 4.28.
Consider the NBA of Figure 4.7 with the alphabet $\Sigma=\{A, B, C\}$. The word $C^{\omega}$ has only


Figure 4.7: An example of an NBA.
one run in $\mathcal{A}$, namely $q_{1} q_{1} q_{1} q_{1} \ldots$, or in short, $q_{1}^{\omega}$. Some other runs are $q_{1} q_{2} q_{3}^{\omega}$ for the word $A B^{\omega},\left(q_{1} q_{1} q_{2} q_{3}\right)^{\omega}$ for the word $(C A B B)^{\omega}$, and $\left(q_{1} q_{2} q_{3}\right)^{n} q_{1}^{\omega}$ for the word $(A B B)^{n} C^{\omega}$ where $n \geqslant 0$.

The runs that go infinitely often through the accept state $q_{3}$ are accepting. For instance, $q_{1} q_{2} q_{3}^{\omega}$ and $\left(q_{1} q_{1} q_{2} q_{3}\right)^{\omega}$ are accepting runs. $q_{1}^{\omega}$ is not an accepting run as it never visits the accept state $q_{3}$, while runs of the form $\left(q_{1} q_{2} q_{3}\right)^{n} q_{1}^{\omega}$ are not accepting as they visit the accept state $q_{3}$ only finitely many times. The language accepted by this NBA is given by the $\omega$-regular expression:

$$
C^{*} A B\left(B^{+}+B C^{*} A B\right)^{\omega}
$$

Later in this chapter (page 198 ff .), NBAs are used for the verification of $\omega$-regular properties - in the same vein as NFAs were exploited for the verification of regular safety properties. In that case, $\Sigma$ is of the form $\Sigma=2^{A P}$. As explained on page 159, propositional logic formulae are used as a shorthand notation for the transitions of such NBAs. For instance, if $A P=\{a, b\}$, then the label $a \vee b$ for an edge from $q$ to $p$ means that there are three transitions from $q$ to $p$ : one for the symbol $\{a\}$, one for the symbol $\{b\}$, and one for the symbol $\{a, b\}$.

## Example 4.29. Infinitely Often Green

Let $A P=\{$ green, red $\}$ or any other set containing the proposition green. The language of words $\sigma=A_{0} A_{1} \ldots \in 2^{A P}$ satisfying the LT property "infinitely often green" is accepted by the NBA $\mathcal{A}$ depicted in Figure 4.8.


Figure 4.8: An NBA accepting "infinitely often green".

The automaton $\mathcal{A}$ is in the accept state $q_{1}$ if and only if the last input set of symbols (i.e., the last set $A_{i}$ ) contains the propositional symbol green. Therefore, $\mathcal{L}_{\omega}(\mathcal{A})$ is exactly the set of all infinite words $A_{0} A_{1} \ldots$ with infinitely many sets $A_{i}$ with green $\in A_{i}$. For example, for the input word

$$
\sigma=\{\text { green }\}\{ \}\{\text { green }\}\{ \}\{\text { green }\}\{ \} \ldots
$$

we obtain the accepting run $q_{0} q_{1} q_{0} q_{1} \ldots$. The same run $q_{0} q_{1} q_{0} q_{1} \ldots$ is obtained for the word

$$
\sigma^{\prime}=(\{\text { green, red }\}\{ \}\{\text { green }\}\{\text { red }\})^{\omega}
$$

or any other word $A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}$ with green $\in A_{2 j}$ and green $\notin A_{2 j+1}$ for all $j \geqslant 0$.

## Example 4.30. Request Response

Many liveness properties are of the form
"Whenever some event $a$ occurs, some event $b$ will eventually occur in the future"

For example, the property "once a request is provided, eventually a response occurs" is of this form. An associated NBA with propositions req and resp is indicated in Figure 4.9. It is assumed that $\{$ req, resp $\} \subseteq A P$, i.e., we assume the NBA to have alphabet $2^{A P}$ with $A P$ containing at least req and resp. It is not difficult to see that this NBA accepts exactly those sequences in which each request is always eventually followed by a response. Note that an infinite trace in which only responses occur, but never a request (or finitely many requests) is also accepting.


Figure 4.9: An NBA accepting "on each request, eventually a response is provided".

## Remark 4.31. NBA and Regular Safety Properties

In Section 4.2, we have seen that there is a strong relationship between bad prefixes of regular safety properties and NFAs. In fact, there is also a strong relationship between NBAs and regular safety properties. This can be seen as follows. Let $P_{\text {safe }}$ be a regular safety property over $A P$ and $\mathcal{A}=\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$ an NFA recognizing the language of all bad prefixes of $P_{\text {safe }}$. Each accept state $q_{F} \in F$ may be assumed to be a trapping state, i.e., $q_{F} \xrightarrow{A} q_{F}$ for all $A \subseteq A P$. This assumption is justified since each extension of a bad prefix is a bad prefix. (As a bad prefix contains a "bad" event that causes the violation of $P_{\text {safe }}$, each extension of this prefix contains this event.)

When interpreting $\mathcal{A}$ as an NBA, it accepts exactly the infinite words $\sigma \in\left(2^{A P}\right)^{\omega}$ that violate $P_{\text {safe }}$, i.e.,

$$
\mathcal{L}_{\omega}(\mathcal{A})=\left(2^{A P}\right)^{\omega} \backslash P_{\text {safe }}
$$

Here, it is important that $\mathcal{A}$ accepts all bad prefixes, and not just the minimal ones (see Exercise 4.18).

If $\mathcal{A}$ is a total deterministic automaton, i.e., in each state there is a single possible transition for each input symbol, then the NBA obtained by

$$
\overline{\mathcal{A}}=\left(Q, 2^{A P}, \delta, Q_{0}, Q \backslash F\right)
$$

accepts the language $\mathcal{L}_{\omega}(\overline{\mathcal{A}})=P_{\text {safe }}$.
This is exemplified by means of a concrete case. Consider again the property "a red light phase should be immediately preceded by a yellow light phase" for a traffic light system. We have seen before (see Example 4.13 on page 161) that the bad prefixes of this safety property constitute a regular language and are accepted by the NFA shown in Figure 4.10. Note that this NFA is total. Applying the procedure described just above to this automaton yields the NBA depicted in Figure 4.11. It is easy to see that the infinite language accepted by this NBA consists exactly of all sequences of the form $\sigma=A_{0} A_{1} A_{2} \ldots$ such that red $\in A_{j}$ implies $j>0$ and yellow $\in A_{j-1}$.

The accepted languages of the NBA examples have so far been $\omega$-regular. It is now shown


Figure 4.10: An NFA for the set of all bad prefixes of $P_{\text {safe }}$.


Figure 4.11: An NBA for the LT property "red should be preceded by yellow".
that this holds for any NBA. Moreover, it will be shown that any $\omega$-regular language can be described by an NBA. Thus, NBAs are as expressive as $\omega$-regular languages. This result is analogous to the fact that NFAs are as expressive as regular languages, and thus may act as an alternative formalism to describe regular languages. In the same spirit, NBAs are an alternative formalism for describing $\omega$-regular languages. This is captured by the following theorem.

## Theorem 4.32. NBAs and $\omega$-Regular Languages

The class of languages accepted by NBAs agrees with the class of $\omega$-regular languages.

The proof of Theorem 4.32 amounts to showing that (1) any $\omega$-regular language is recognized by an NBA (see Corollary 4.38 on page 182) and (2) that the language $\mathcal{L}_{\omega}(\mathcal{A})$ accepted by the NBA $\mathcal{A}$ is $\omega$-regular (see Lemma 4.39 on page 183).

We first consider the statement that $\omega$-regular languages are contained in the class of languages recognized by an NBA. The proof of this fact is divided into the following three steps that rely on operations for NBAs to mimic the building blocks of $\omega$-regular expressions:
(1) For any NBA $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ there exists an NBA accepting $\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right) \cup \mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)$.
(2) For any regular language $\mathcal{L}$ (of finite words) with $\varepsilon \notin \mathcal{L}$ there exists an NBA accepting $\mathcal{L}^{\omega}$.
(3) For regular language $\mathcal{L}$ and NBA $\mathcal{A}^{\prime}$ there exists an NBA accepting $\mathcal{L} . \mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)$.

These three results that are proven below form the basic ingredients to construct an NBA for a given $\omega$-regular expression $G=\mathrm{E}_{1} \cdot \mathrm{~F}_{1}^{\omega}+\ldots+\mathrm{E}_{n} \cdot \mathrm{~F}_{n}^{\omega}$ with $\varepsilon \notin \mathrm{F}_{i}$. This works as follows. As an initial step, (2) is exploited to construct NBA $\mathcal{A}_{1}^{\prime}, \ldots, \mathcal{A}_{n}^{\prime}$ for the expressions $\mathrm{F}_{1}^{\omega}, \ldots, \mathrm{F}_{n}^{\omega}$. Then, (3) is used to construct an NBA for the expressions $\mathrm{E}_{i} . \mathrm{F}_{i}^{\omega}$, for $1 \leqslant i \leqslant n$. Finally, these NBA are combined using (1) to obtain an NBA for G.

Let us start with the union operator on two NBAs. Let $\mathcal{A}_{1}=\left(Q_{1}, \Sigma, \delta_{1}, Q_{0,1}, F_{1}\right)$ and $\mathcal{A}_{2}=\left(Q_{2}, \Sigma, \delta_{2}, Q_{0,2}, F_{2}\right)$ be NBAs over the same alphabet $\Sigma$. Without loss of generality, it may be assumed that the state spaces $Q_{1}$ and $Q_{2}$ of $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ are disjoint, i.e., $Q_{1} \cap Q_{2}=\varnothing$. Let $\mathcal{A}_{1}+\mathcal{A}_{2}$ be the NBA with the joint state spaces of $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$, and with all transitions in $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$. The initial states of $\mathcal{A}$ are the initial states of $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$, and similarly, the accept states of $\mathcal{A}$ are the accept states of $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$. That is,

$$
\mathcal{A}_{1}+\mathcal{A}_{2}=\left(Q_{1} \cup Q_{2}, \Sigma, \delta, Q_{0,1} \cup Q_{0,2}, F_{1} \cup F_{2}\right)
$$

where $\delta(q, A)=\delta_{i}(q, A)$ if $q \in Q_{i}$ for $i=1,2$. Clearly, any accepting run in $\mathcal{A}_{i}$ is also an accepting run in $\mathcal{A}_{1}+\mathcal{A}_{2}$, and vice versa, each accepting run in $\mathcal{A}_{1}+\mathcal{A}_{2}$ is an accepting run in either $\mathcal{A}_{1}$ or $\mathcal{A}_{2}$. This yields $\mathcal{L}_{\omega}\left(\mathcal{A}_{1}+\mathcal{A}_{2}\right)=\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right) \cup \mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)$. We thus obtain:

## Lemma 4.33. Union Operator on NBA

For $N B A \mathcal{A}_{1}$ and $\mathcal{A}_{2}$ (both over the alphabet $\Sigma$ ) there exists an $N B A \mathcal{A}$ such that:

$$
\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right) \cup \mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right) \quad \text { and } \quad|\mathcal{A}|=\mathcal{O}\left(\left|\mathcal{A}_{1}\right|+\left|\mathcal{A}_{2}\right|\right) .
$$

Now consider (2). We will show that for any regular language $\mathcal{L} \subseteq \Sigma^{*}$ there exists an NBA over the alphabet $\Sigma$ that accepts the $\omega$-regular language $\mathcal{L}^{\omega}$. To do so, we start with a representation of $\mathcal{L}$ by an NFA $\mathcal{A}$.

## Lemma 4.34. $\omega$-Operator for NFA

For each NFA $\mathcal{A}$ with $\varepsilon \notin \mathcal{L}(\mathcal{A})$ there exists an NBA $\mathcal{A}^{\prime}$ such that

$$
\mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)=\mathcal{L}(\mathcal{A})^{\omega} \quad \text { and } \quad\left|\mathcal{A}^{\prime}\right|=\mathcal{O}(|\mathcal{A}|)
$$

Proof: Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NFA with $\varepsilon \notin \mathcal{L}(\mathcal{A})$. Without loss of generality, we may assume that all initial states in $\mathcal{A}$ have no incoming transitions and are not accepting.

Any $\mathcal{A}$ that does not possess this property, can be modified into an equivalent NFA as follows. Add a new initial (nonaccept) state $q_{\text {new }}$ to $Q$ with the transitions $q_{\text {new }} \xrightarrow{A} q$ if and only if $q_{0} \xrightarrow{A} q$ for some initial state $q_{0} \in Q_{0}$. All other transitions, as well as the accept states, remain unchanged. The state $q_{\text {new }}$ is the single initial state of the modified NFA, is not accept, and, clearly, has no incoming transitions. This modification neither affects the accepted language nor the asymptotic size of $\mathcal{A}$.

In the sequel, we assume that $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ is an NFA such that the states in $Q_{0}$ do not have any incoming transitions and $Q_{0} \cap F=\varnothing$. We now construct an NBA $\mathcal{A}^{\prime}=\left(Q, \Sigma, \delta^{\prime}, Q_{0}^{\prime}, F^{\prime}\right)$ with $\mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)=\mathcal{L}(\mathcal{A})^{\omega}$. The basic idea of the construction of $\mathcal{A}^{\prime}$ is to add for any transition in $\mathcal{A}$ that leads to an accept state new transitions leading to the initial states of $\mathcal{A}$. Formally, the transition relation $\delta^{\prime}$ in the NBA $\mathcal{A}^{\prime}$ is given by

$$
\delta^{\prime}(q, A)= \begin{cases}\delta(q, A) & \text { if } \delta(q, A) \cap F=\varnothing \\ \delta(q, A) \cup Q_{0} & \text { otherwise }\end{cases}
$$

The initial states in the NBA $\mathcal{A}^{\prime}$ agree with the initial states in $\mathcal{A}$, i.e., $Q_{0}^{\prime}=Q_{0}$. These are also the accept states in $\mathcal{A}^{\prime}$, i.e., $F^{\prime}=Q_{0}$.

Let us check that $\mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)=\mathcal{L}(\mathcal{A})^{\omega}$. This is proven as follows.
$\subseteq:$ Assume that $\sigma \in \mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)$ and let $q_{0} q_{1} q_{2} \ldots$ be an accepting run for $\sigma$ in $\mathcal{A}^{\prime}$. Hence, $q_{i} \in F^{\prime}=Q_{0}$ for infinitely many indices $i$. Let $i_{0}=0<i_{1}<i_{2}<\ldots$ be the strictly increasing sequence of natural numbers with $\left\{q_{i_{0}}, q_{i_{1}}, q_{i_{2}}, \ldots\right\} \subseteq Q_{0}$ and $q_{j} \notin Q_{0}$ for all $j \in \mathbb{N} \backslash\left\{i_{0}, i_{1}, i_{2}, \ldots\right\}$. The word $\sigma$ can be divided into infinitely many nonempty finite subwords $w_{i} \in \Sigma^{*}$ yielding $\sigma=w_{1} w_{2} w_{3} \ldots$ such that $q_{i_{k}} \in \delta^{\prime *}\left(q_{i_{k-1}}, w_{k}\right)$ for all $k \geqslant 1$. (The extension of $\delta^{\prime}$ to a function $\delta^{\prime *}: Q \times \Sigma^{*} \rightarrow 2^{Q}$ is as for an NFA, see page 154.) By definition of $\mathcal{A}^{\prime}$ and since the states $q_{i_{k}} \in Q_{0}$ do not have any predecessor in $\mathcal{A}$, we get $\delta^{*}\left(q_{i_{k-1}}, w_{k}\right) \cap F \neq \varnothing$. This yields $w_{k} \in \mathcal{L}(\mathcal{A})$ for all $k \geqslant 1$, which gives us $\sigma \in \mathcal{L}(\mathcal{A})^{\omega}$.
$\supseteq$ : Let $\sigma=w_{1} w_{2} w_{3} \ldots \in \Sigma^{\omega}$ such that $w_{k} \in \mathcal{L}(\mathcal{A})$ for all $k \geqslant 1$. For each $k$, we choose an accepting run $q_{0}^{k} q_{1}^{k} \ldots q_{n_{k}}^{k}$ for $w_{k}$ in $\mathcal{A}$. Hence, $q_{0}^{k} \in Q_{0}$ and $q_{n_{k}}^{k} \in F$. By definition of $\mathcal{A}^{\prime}$, we have $q_{0}^{k+1} \in \delta^{\prime *}\left(q_{0}^{k}, \mathrm{w}_{k}\right)$ for all $k \geqslant 1$. Thus,

$$
q_{0}^{1} \ldots q_{n_{1}-1}^{1} q_{0}^{2} \ldots q_{n_{2}-1}^{2} q_{0}^{3} \ldots q_{n_{3}-1}^{3} \ldots
$$

is an accepting run for $\sigma$ in $\mathcal{A}^{\prime}$. Hence, $\sigma \in \mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)$.

Example 4.35. $\omega$-Operator for an NFA
Consider the NFA depicted in the left upper part of Figure 4.12. It accepts the language $A^{*} B$. In order to obtain an NBA recognizing $\left(A^{*} B\right)^{\omega}$, we first apply the transformation


Figure 4.12: From an NFA accepting $A^{*} B$ to an NBA accepting $\left(A^{*} B\right)^{\omega}$.
as described in the proof of Lemma 4.34 to remove initial states that have an incoming transition. This yields the NFA depicted in the right upper part of Figure 4.12. This automaton can now be used to apply the construction of the required NBA as detailed in the proof of Lemma 4.34. This yields the NBA depicted in the lower part of Figure 4.12.

It remains to provide a construction for task (3) above. Assume that we have NFA $\mathcal{A}$ for the regular language $\mathcal{L}(\mathcal{A})$ and a given NBA $\mathcal{A}^{\prime}$ at our disposal. The proof of the following lemma will describe a procedure to obtain an NBA for the $\omega$-language $\mathcal{L}(\mathcal{A}) \cdot \mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)$.

## Lemma 4.36. Concatenation of an NFA and an NBA

For NFA $\mathcal{A}$ and NBA $\mathcal{A}^{\prime}$ (both over the alphabet $\Sigma$ ), there exists an $N B A \mathcal{A}^{\prime \prime}$ with

$$
\mathcal{L}_{\omega}\left(\mathcal{A}^{\prime \prime}\right)=\mathcal{L}(\mathcal{A}) \cdot \mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right) \quad \text { and } \quad\left|\mathcal{A}^{\prime \prime}\right|=\mathcal{O}\left(|\mathcal{A}|+\left|\mathcal{A}^{\prime}\right|\right) .
$$

Proof: Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NFA and $\mathcal{A}^{\prime}=\left(Q^{\prime}, \Sigma, \delta^{\prime}, Q_{0}^{\prime}, F^{\prime}\right)$ an NBA with $Q \cap Q^{\prime}=\varnothing$. Let $\mathcal{A}^{\prime \prime}=\left(Q^{\prime \prime}, \Sigma, \delta^{\prime \prime}, Q_{0}^{\prime \prime}, F^{\prime \prime}\right)$ be the following NBA. The state space is $Q^{\prime \prime}=Q \cup Q^{\prime}$. The set of initial and accept states are given by

$$
Q_{0}^{\prime \prime}= \begin{cases}Q_{0} & \text { if } Q_{0} \cap F=\varnothing \\ Q_{0} \cup Q_{0}^{\prime} & \text { otherwise }\end{cases}
$$

and $F^{\prime \prime}=F^{\prime}$ (set of accept state in the NBA $\mathcal{A}^{\prime}$ ). The transition function $\delta^{\prime \prime}$ is given by

$$
\delta^{\prime \prime}(q, A)= \begin{cases}\delta(q, A) & \text { if } q \in Q \text { and } \delta(q, A) \cap F=\varnothing \\ \delta(q, A) \cup Q_{0}^{\prime} & \text { if } q \in Q \text { and } \delta(q, A) \cap F \neq \varnothing \\ \delta^{\prime}(q, A) & \text { if } q \in Q^{\prime}\end{cases}
$$

It is now easy to check that $\mathcal{A}^{\prime \prime}$ fulfills the desired conditions.

Example 4.37. Concatenation of an NFA and an NBA
Consider the NFA $\mathcal{A}$ and the NBA $\mathcal{A}^{\prime}$ depicted in the left and right upper part of Figure 4.13 , respectively. We have $\mathcal{L}(\mathcal{A})=(A B)^{*}$ and $\mathcal{L}\left(\mathcal{A}^{\prime}\right)=(A+B)^{*} B A^{\omega}$. Applying the transformation as described in Lemma 4.36 yields the NBA depicted in the lower part of Figure 4.13. It is not difficult to assess that this NBA accepts indeed the concatenated language $(A B)^{*}(A+B)^{*} B A^{\omega}$.


Figure 4.13: Concatenation of an NFA and an NBA.

By Lemmas 4.33, 4.34, and 4.36 we obtain the first part for the proof of Theorem 4.32:

Corollary 4.38. NBA for $\omega$-Regular Languages
For any $\omega$-regular language $\mathcal{L}$ there exists an NBA $\mathcal{A}$ with $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}$.

The proof of the following lemma shows that the languages accepted by NBA can be described by an $\omega$-regular expression.

## Lemma 4.39. NBAs Accept $\omega$-Regular Languages

For each $N B A \mathcal{A}$, the accepted language $\mathcal{L}_{\omega}(\mathcal{A})$ is $\omega$-regular.

Proof: Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NBA. For states $q, p \in Q$, let $\mathcal{A}_{q p}$ be the NFA $(Q, \Sigma, \delta,\{q\},\{p\})$. Then, $\mathcal{A}_{q p}$ recognizes the regular language consisting of all finite words $w \in \Sigma^{*}$ that have a run in $\mathcal{A}$ leading from $q$ to $p$, that is,

$$
\mathcal{L}_{q p} \stackrel{\text { def }}{=} \mathcal{L}\left(\mathcal{A}_{q p}\right)=\left\{w \in \Sigma^{*} \mid p \in \delta^{*}(q, w)\right\} .
$$

Consider a word $\sigma \in \mathcal{L}_{\omega}(\mathcal{A})$ and an accepting run $q_{0} q_{1} \ldots$ for $\sigma$ in $\mathcal{A}$. Some accept state $q \in F$ appears infinitely often in this run. Hence, we may split $\sigma$ into nonempty finite subwords $w_{0}, w_{1}, w_{2}, w_{3}, \ldots \in \Sigma^{*}$ such that $w_{0} \in \mathcal{L}_{q_{0} q}$ and $w_{k} \in \mathcal{L}_{q q}$ for all $k \geqslant 1$ and

$$
\sigma=\underbrace{W_{0}}_{\in \mathcal{L}_{q_{0} q}} \underbrace{W_{1}}_{\in \mathcal{L}_{q q}} \underbrace{W_{2}}_{\in \mathcal{L}_{q q}} \underbrace{W_{3}}_{\in \mathcal{L}_{q q}} \cdots \cdots
$$

On the other hand, any infinite word $\sigma$ which has the form $\sigma=W_{0} W_{1} W_{2} \ldots$ where the $w_{k}$ 's are nonempty finite words with $w_{0} \in \mathcal{L}_{q_{0} q}$ for some initial state $q_{0} \in Q_{0}$ and $\left\{w_{1}, w_{2}, w_{3}, \ldots\right\} \subseteq \mathcal{L}_{q q}$ for some accept state $q \in F$ has an accepting run in $\mathcal{A}$. This yields

$$
\sigma \in \mathcal{L}_{\omega}(\mathcal{A}) \quad \text { if and only if } \exists q_{0} \in Q_{0} \exists q \in F . \quad \sigma \in \mathcal{L}_{q_{0} q}\left(\mathcal{L}_{q q} \backslash\{\varepsilon\}\right)^{\omega} .
$$

Hence, $\mathcal{L}_{\omega}(\mathcal{A})$ agrees with the language

$$
\bigcup_{q_{0} \in Q_{0}, q \in F} \mathcal{L}_{q_{0} q \cdot} \cdot\left(\mathcal{L}_{q q} \backslash\{\varepsilon\}\right)^{\omega}
$$

which is $\omega$-regular.

## Example 4.40. From NBA to $\omega$-Regular Expression

For the NBA $\mathcal{A}$ shown in Figure 4.7 on page 175, a corresponding $\omega$-regular expression is obtained by

$$
\mathcal{L}_{q_{1} q_{3}} \cdot\left(\mathcal{L}_{q_{3} q_{3}} \backslash\{\varepsilon\}\right)^{\omega}
$$

since $q_{1}$ is the unique initial state and $q_{3}$ the unique accept state in $\mathcal{A}$. The regular language $\mathcal{L}_{q_{3} q_{3}} \backslash\{\varepsilon\}$ can be described by the expression $\left(B^{+}+B C^{*} A B\right)^{+}$, while $\mathcal{L}_{q_{1} q_{3}}$ is given by $\left(C^{*} A B\left(B^{*}+B C^{*} A B\right)^{*} B\right)^{*} C^{*} A B$. Hence, $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}(\mathrm{G})$ where G is the $\omega$-regular expression:

$$
G=\underbrace{\left(C^{*} A B\left(B^{*}+B C^{*} A B\right)^{*} B\right)^{*} C^{*} A B}_{\mathcal{L}_{q_{1} q_{3}}} \underbrace{\left(\left(B^{+}+B C^{*} A B\right)^{+}\right)^{\omega}}_{\left(\mathcal{L}_{q_{3} q_{3}} \backslash\{\varepsilon\}\right)^{\omega}} .
$$

The thus obtained expression $G$ can be simplified to the equivalent expression:

$$
C^{*} A B\left(B^{+}+B C^{*} A B\right)^{\omega}
$$

The above lemma, together with Corollary 4.38, completes the proof of Theorem 4.32 stating the equivalence of the class of languages accepted by NBAs and the class of all $\omega$-regular languages. Thus, NBAs and $\omega$-regular languages are equally expressive.

A fundamental question for any type of automata model is the question whether for a given automaton $\mathcal{A}$ the accepted language is empty. For nondeterministic Büchi automata, an analysis of the underlying directed graph by means of standard graph algorithms is sufficient, as we will show now.

## Lemma 4.41. Criterion for the Nonemptiness of an NBA

Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NBA. Then, the following two statements are equivalent:
(a) $\mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$,
(b) There exists a reachable accept state $q$ that belongs to a cycle in $\mathcal{A}$. Formally,

$$
\exists q_{0} \in Q_{0} \exists q \in F \exists w \in \Sigma^{*} \exists v \in \Sigma^{+} . q \in \delta^{*}\left(q_{0}, w\right) \cap \delta^{*}(q, v) .
$$

Proof: $(\mathrm{a}) \Longrightarrow(\mathrm{b})$ : Let $\sigma=A_{0} A_{1} A_{2} \ldots \in \mathcal{L}_{\omega}(\mathcal{A})$ and let $q_{0} q_{1} q_{2} \ldots$ be an accepting run for $\sigma$ in $\mathcal{A}$. Let $q \in F$ be an accept state with $q=q_{i}$ for infinitely many indices $i$. Let $i$ and $j$ be two indices with $0 \leqslant i<j$ and $q_{i}=q_{j}=q$. We consider the finite words $\mathrm{w}=A_{0} A_{1} \ldots A_{i-1}$ and $v=A_{i} A_{i+1} \ldots A_{j-1}$ and obtain $q=q_{i} \in \delta^{*}\left(q_{0}, w\right)$ and $q=q_{j} \in \delta^{*}\left(q_{i}, v\right)=\delta^{*}(q, v)$. Hence, (b) holds.
$(\mathrm{b}) \Longrightarrow(\mathrm{a})$ : Let $q_{0}, q, w, v$ be as in statement $(\mathrm{b})$. Then, the infinite word $\sigma=w v^{\omega}$ has a run of the form $q_{0} \ldots q \ldots q \ldots q \ldots$ that infinitely often contains $q$. Since $q \in F$ this run is accepting which yields $\sigma \in \mathcal{L}_{\omega}(\mathcal{A})$, and thus, $\mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$.

By the above lemma, the emptiness problem for NBAs can be solved by means of graph algorithms that explore all reachable states and check whether they belong to a cycle. One possibility to do so is to calculate the strongly connected components of the underlying directed graph of $\mathcal{A}$ and to check whether there is at least one nontrivial strongly con-
nected component ${ }^{3}$ that is reachable (from at least one of the initial states) and contains an accept state. Since the strongly connected components of a (finite) directed graph can be computed in time linear in the number of states and edges, the time complexity of this algorithm for the emptiness check of NBA $\mathcal{A}$ is linear in the size of $\mathcal{A}$. An alternative algorithm that also runs in time linear in the size of $\mathcal{A}$, but avoids the explicit computation of the strongly connected components, can be derived from the results stated in Section 4.4.2.

## Theorem 4.42. Checking Emptiness for NBA

The emptiness problem for NBA $\mathcal{A}$ can be solved in time $\mathcal{O}(|\mathcal{A}|)$.

Since NBAs serve as a formalism for $\omega$-regular languages, we may identify two Büchi automata for the same language:

## Definition 4.43. Equivalence of NBA

Let $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ be two NBAs with the same alphabet. $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ are called equivalent, denoted $\mathcal{A}_{1} \equiv \mathcal{A}_{2}$, if $\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right)=\mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)$.

## Example 4.44. Equivalent NBA

As for other finite automata, equivalent NBAs can have a totally different structure. For example, consider the NBA shown in Figure 4.14 over the alphabet $2^{A P}$ where $A P=$ $\{a, b\}$. Both NBAs represent the liveness property "infinitely often $a$ and infinitely often $b$ ", and thus, they are equivalent.

## Remark 4.45. NFA vs. NBA Equivalence

It is interesting to consider more carefully the relationship between the notions of equivalence of NFAs and NBAs. Let $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ be two automata that we can regard as either NFA or as NBA. To distinguish the equivalence symbol $\equiv$ for NFAs from that for NBAs we will write in this example $\equiv_{N F A}$ to denote the equivalence relation for NFA and the symbol $\equiv_{N B A}$ to denote the equivalence relation for NBA, i.e., $\mathcal{A}_{1} \equiv_{N F A} \mathcal{A}_{2}$ iff $\mathcal{L}\left(\mathcal{A}_{1}\right)=\mathcal{L}\left(\mathcal{A}_{2}\right)$ and $\mathcal{A}_{1} \equiv_{N B A} \mathcal{A}_{2}$ iff $\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right)=\mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)$.

1. If $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ accept the same finite words, i.e., $\mathcal{A}_{1} \equiv{ }_{N F A} \mathcal{A}_{2}$, then this does not mean that they also accept the same infinite words. The following two automata examples show this:

[^17]

Figure 4.14: Two equivalent NBA.


We have $\mathcal{L}\left(\mathcal{A}_{1}\right)=\mathcal{L}\left(\mathcal{A}_{2}\right)=\left\{A^{n} \mid n \geqslant 1\right\}$, but $\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right)=\left\{A^{\omega}\right\}$ and $\mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)=\varnothing$. Thus, $\mathcal{A}_{1} \equiv_{N F A} \mathcal{A}_{2}$ but $\mathcal{A}_{1} \not \equiv_{N B A} \mathcal{A}_{2}$.
2. If $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ accept the same infinite words, i.e., $\mathcal{A}_{1} \equiv_{N B A} \mathcal{A}_{2}$, then one might expect that they would also accept the same finite words. This also turns out not to be true. The following example shows this:


We have $\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right)=\mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)=\left\{A^{\omega}\right\}$, but $\mathcal{L}\left(\mathcal{A}_{1}\right)=\left\{A^{2 n} \mid n \geqslant 0\right\}$ and $\mathcal{L}\left(\mathcal{A}_{2}\right)=$ $\left\{A^{2 n+1} \mid n \geqslant 0\right\}$.
3. If $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ are both deterministic (see Definition 4.9 on page 156 ), then $\mathcal{A}_{1} \equiv{ }_{\text {NFA }}$ $\mathcal{A}_{2}$ implies $\mathcal{A}_{1} \equiv_{N B A} \mathcal{A}_{2}$. The reverse is, however, not true, as illustrated by the previous example.

For technical reasons, it is often comfortable to assume for an NBA that for each state $q$ and for each input symbol $A$, there is a possible transition. Such an NBA can be seen to be nonblocking since no matter how the nondeterministic choices are resolved, the automaton cannot fail to consume the current input symbol.

## Definition 4.46. Nonblocking NBA

Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NBA. $\mathcal{A}$ is called nonblocking if $\delta(q, A) \neq \varnothing$ for all states $q$ and all symbols $A \in \Sigma$.

Note that for a given nonblocking NBA $\mathcal{A}$ and input word $\sigma \in \Sigma^{\omega}$, there is at least one (infinite) possibly nonaccepting run for $\sigma$ in $\mathcal{A}$. The following remark demonstrates that it is not a restriction to assume a nonblocking NBA.

## Remark 4.47. Nonblocking NBA

For each NBA $\mathcal{A}$ there exists a nonblocking NBA $\operatorname{trap}(\mathcal{A})$ with $|\operatorname{trap}(\mathcal{A})|=\mathcal{O}(|\mathcal{A}|)$ and $\mathcal{A} \equiv \operatorname{trap}(\mathcal{A})$.

Let us see how such a nonblocking NBA can be derived from $\mathcal{A}$. NBA $\operatorname{trap}(\mathcal{A})$ is obtained from $\mathcal{A}$ by inserting a nonaccept trapping state $q_{\text {trap }}$ equipped with a self-loop for each symbol in the alphabet $\Sigma$. For every symbol $A \in \Sigma$ for which state $q$ in $\mathcal{A}$ does not have an outgoing transition, a transition to $q_{\text {trap }}$ is added. Formally, if $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$, then $\operatorname{trap}(\mathcal{A})=\left(Q^{\prime}, \Sigma, \delta^{\prime}, Q_{0}^{\prime}, F^{\prime}\right)$ as follows. Here, $Q^{\prime}=Q \cup\left\{q_{\text {trap }}\right\}$ where $q_{\text {trap }}$ is a new state (not in $Q$ ) that will be reached in $\mathcal{A}^{\prime}$ whenever $\mathcal{A}$ does not have a corresponding transition. Formally, the transition relation $\delta^{\prime}$ of $\operatorname{trap}(\mathcal{A})$ is defined by:

$$
\delta^{\prime}(q, A)= \begin{cases}\delta(q, A) & \text { if } q \in Q \text { and } \delta(q, A) \neq \varnothing \\ \left\{q_{\text {trap }}\right\} & \text { otherwise }\end{cases}
$$

The initial and accept states are unchanged, i.e., $Q_{0}^{\prime}=Q_{0}$ and $F^{\prime}=F$. By definition, $\operatorname{trap}(\mathcal{A})$ is nonblocking and - since the new trap state is nonaccepting - is equivalent to $\mathcal{A}$.

We conclude this subsection on automata over infinite words with a few more comments on Büchi automata and $\omega$-regular languages. We first study the subclass of deterministic Büchi automata (Section 4.3.3 below) and then in Section 4.3.4 the class of NBA with a more general acceptance condition consisting of several acceptance sets that have to be visited infinitely often.

### 4.3.3 Deterministic Büchi Automata

An important difference between finite-state automata and Büchi automata is the expressive power of deterministic and nondeterministic automata. While for languages of finite words, DFAs and NFAs have the same expressiveness, this does not hold for Büchi automata.

The definition of a deterministic Büchi automaton is the same as for a DFA:

## Definition 4.48. Deterministic Büchi Automaton (DBA)

Let $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)$ be an NBA. $\mathcal{A}$ is called deterministic, if

$$
\left|Q_{0}\right| \leqslant 1 \quad \text { and } \quad|\delta(q, A)| \leqslant 1
$$

for all $q \in Q$ and $A \in \Sigma . \mathcal{A}$ is total if $\left|Q_{0}\right|=1$ and $|\delta(q, A)|=1$ for all $q \in Q$ and $A \in \Sigma$.

Obviously, the behavior of a DBA for a given input word is deterministic: either eventually the DBA will get stuck in some state as it fails to consume the current input symbol or there is a unique (infinite) run for the given input word. Total DBAs rule out the first alternative and ensure the existence of a unique run for every input word $\sigma \in \Sigma^{\omega}$.

## Example 4.49. $D B A$ for $L T$ Properties

Figure 4.15 shows the DBA $\mathcal{A}^{\prime}$ (on the left) and the NBA $\mathcal{A}$ (on the right) over the alphabet $\Sigma=2^{A P}$ where $A P=\{a, b\}$. These automata are equivalent since both represent the LT property "always $b$ and infinitely often $a$ ". Let $\delta$ be the transition function of $\mathcal{A}$ and $\delta^{\prime}$


Figure 4.15: An equivalent $\mathrm{DBA} \mathcal{A}^{\prime}$ (left) and NBA $\mathcal{A}$ (right).
the transition function of $\mathcal{A}^{\prime}$. The NBA $\mathcal{A}$ is not deterministic since for all input symbols containing a $b$, there is the possibility to move to either state $r_{0}$ or $r_{1}$. The DBA $\mathcal{A}^{\prime}$ is
deterministic. Note that both $\mathcal{A}^{\prime}$ and $\mathcal{A}$ are blocking, e.g., any state is blocking on an input symbol containing $\neg b$.

As for deterministic finite automata, the usual notation is $q^{\prime}=\delta(q, A)\left(\right.$ instead of $\left\{q^{\prime}\right\}=$ $\delta(q, A))$ and $\delta(q, A)=\perp$ (undefined), if $\delta(q, A)=\varnothing$. Thus, the transition relation of a DBA is understood as partial function $\delta: Q \times \Sigma \rightarrow Q$. Total DBAs are often written in the form $\left(Q, \Sigma, \delta, q_{0}, F\right)$ where $q_{0}$ is the unique initial state and $\delta$ is viewed as a total function $Q \times \Sigma \rightarrow Q$. Since DBA can always be extended by a nonaccept trapping state without changing the accepting language, it can be assumed without restriction that the transition relation is total. For instance, Figure 4.16 shows an equivalent total DBA for the DBA $\mathcal{A}^{\prime}$ in Figure 4.15 that is obtained by adding such a trapping state.

true

Figure 4.16: A total DBA for "always $b$ and infinitely often $a$ ".

The transition function $\delta$ of a total DBA can be expanded to a total function $\delta^{*}: Q \times \Sigma^{*} \rightarrow$ $Q$ in the obvious way; see also page 157 for the transition function of a total DFA. That is, let $\delta^{*}(q, \varepsilon)=q, \delta^{*}(q, A)=\delta(q, A)$ and

$$
\delta^{*}\left(q, A_{1} A_{2} \ldots A_{n}\right)=\delta^{*}\left(\delta\left(q, A_{1}\right), A_{2} \ldots A_{n}\right)
$$

Then, for every infinite word $\sigma=A_{0} A_{1} A_{2} \ldots \in \Sigma^{\omega}$, the run $q_{0} q_{1} q_{2} \ldots$ in $\mathcal{A}$ belonging to $\sigma$ is given by $q_{i+1}=\delta^{*}\left(q_{0}, A_{0} \ldots A_{i}\right)$ for all $i \geqslant 0$, where $q_{0}$ is the unique initial state of $\mathcal{A}$. In particular, for total $\mathrm{DBA} \mathcal{A}=\left(Q, \Sigma, \delta, q_{0}, F\right)$ the accepted language is given by

$$
\mathcal{L}_{\omega}(\mathcal{A})=\left\{A_{0} A_{1} A_{2} \ldots \in \Sigma^{\omega} \mid \delta^{*}\left(q_{0}, A_{0} \ldots A_{i}\right) \in F \text { for infinitely many } i\right\}
$$

As we have seen before, NFAs are as expressive as deterministic ones. However, NBAs are more expressive than deterministic ones. That is, there do exist NBA for which there
does not exist an equivalent deterministic one. Stated differently, while any $\omega$-language accepted by a DBA is $\omega$-regular, there do exist $\omega$-regular languages for which there does not exist a DBA accepting it. An example of such $\omega$-regular language is the language given by the expression $(A+B)^{*} B^{\omega}$.


Figure 4.17: NBA for the $\omega$-regular expression $(A+B)^{*} B^{\omega}$.

In fact, the language $\mathcal{L}_{\omega}\left((A+B)^{*} B^{\omega}\right)$ is accepted by a rather simple NBA, shown in Figure 4.17. The idea of this NBA is that given an input word $\sigma=w B^{\omega}$ where $w \in\{A, B\}^{*}$ the automaton may stay in $q_{0}$ and guess nondeterministically when the suffix consisting of $B$ 's starts and then moves to the accept state $q_{1}$. This behavior, however, cannot be simulated by a DBA as formally shown in the following theorem.

## Theorem 4.50. NBAs are More Powerful than DBAs

There does not exist a $D B A \mathcal{A}$ such that $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}\left((A+B)^{*} B^{\omega}\right)$.

Proof: By contradiction. Assume that $\mathcal{L}_{\omega}\left((A+B)^{*} B^{\omega}\right)=\mathcal{L}_{\omega}(\mathcal{A})$ for some DBA $\mathcal{A}=$ $\left(Q, \Sigma, \delta, q_{0}, F\right)$ with $\Sigma=\{A, B\}$. Note that since $\mathcal{A}$ is deterministic, $\delta^{*}$ can be considered as a function of type $Q \times \Sigma^{*} \rightarrow Q$.

Since the word $\sigma_{1}=B^{\omega}$ belongs to $\mathcal{L}_{\omega}\left((A+B)^{*} B^{\omega}\right)=\mathcal{L}_{\omega}(\mathcal{A})$, there exists an accepting state $q_{1} \in F$ and a $n_{1} \in \mathbb{N}_{\geqslant 1}$ such that

$$
\begin{equation*}
\delta^{*}\left(q_{0}, B^{n_{1}}\right)=q_{1} \in F . \tag{1}
\end{equation*}
$$

(Since $\mathcal{A}$ is deterministic, $q_{1}$ is uniquely determined.) Now consider the word $\sigma_{2}=$ $B^{n_{1}} A B^{\omega} \in \mathcal{L}_{\omega}\left((A+B)^{*} B^{\omega}\right)=\mathcal{L}_{\omega}(\mathcal{A})$. Since $\sigma_{2}$ is accepted by $\mathcal{A}$, there exists an accepting state $q_{2} \in F$ and $n_{2} \in \mathbb{N}_{\geqslant 1}$, such that

$$
\begin{equation*}
\delta^{*}\left(q_{0}, B^{n_{1}} A B^{n_{2}}\right)=q_{2} \in F \tag{2}
\end{equation*}
$$

The word $B^{n_{1}} A B^{n_{2}} A B^{\omega}$ is in $\mathcal{L}_{\omega}\left((A+B)^{*} B^{\omega}\right)$, and, thus, is accepted by $\mathcal{A}$. So, there is an accepting state $q_{3} \in F$ and $n_{3} \in \mathbb{N}_{\geqslant 1}$ with

$$
\begin{equation*}
\delta^{*}\left(q_{0}, B^{n_{1}} A B^{n_{2}} A B^{n_{3}}\right)=q_{3} \in F . \tag{3}
\end{equation*}
$$

Continuing this process, we obtain a sequence $n_{1}, n_{2}, n_{3}, \ldots$ of natural numbers $\geqslant 1$ and a sequence $q_{1}, q_{2}, q_{3}, \ldots$ of accepting states such that

$$
\delta^{*}\left(q_{0}, B^{n_{1}} A B^{n_{2}} A \ldots B^{n_{i-1}} A B^{n_{i}}\right)=q_{i} \in F, \quad i \geqslant 1 \ldots
$$

Since there are only finitely many states, there exist $i<j$ such that

$$
\delta^{*}\left(q_{0}, B^{n_{1}} A \ldots A B^{n_{i}}\right)=\delta^{*}\left(q_{0}, B^{n_{1}} A \ldots A B^{n_{i}} \ldots A B^{n_{j}}\right)
$$

Thus $\mathcal{A}$ has an accepting run on

$$
B^{n_{1}} A \ldots A B^{n_{i}}\left(A B^{n_{i+1}} A \ldots A B^{n_{j}}\right)^{\omega}
$$

But this word has infinitely many occurrences of $A$, and thus does not belong to $\mathcal{L}_{\omega}((A+$ $B)^{*} B^{\omega}$ ). Contradiction.

## Example 4.51. The Need for Nondeterminism

In Examples 4.29 and 4.30, we provided DBAs for LT properties. To represent liveness properties of the form "eventually forever", the concept of nondeterminism is, however, necessary. Consider the property "eventually forever $a$ ", where $a$ is some atomic proposition. Let $\{a\}=A P$, i.e., $2^{A P}=\{A, B\}$ where $A=\{ \}$ and $B=\{a\}$. Then, the linear-time property "eventually forever $a$ " is given by the $\omega$-regular expression

$$
(A+B)^{*} B^{\omega}=(\{ \}+\{a\})^{*}\{a\}^{\omega} .
$$

By Theorem 4.50, there is no DBA for "eventually forever $a$ ". On the other hand, this property can be described by the NBA $\mathcal{A}$ depicted in Figure 4.18. (Note that state $q_{2}$ could be omitted, as there is no accepting run that starts in $q_{2}$.) Intuitively, $\mathcal{A}$ nondeterministically decides (by means of an omniscient oracle) from which instant the proposition $a$ is continuously true. This behavior cannot be mimicked by a DBA.


Figure 4.18: An NBA accepting "eventually forever $a$ ".

The reader might wonder why the powerset construction known for finite automata (see page 157) fails for Büchi automata. The deterministic automaton $\mathcal{A}_{\text {det }}$ obtained through
the powerset construction allows simulating the given nondeterministic automaton $\mathcal{A}$ by keeping track of the set $Q^{\prime}$ of states that are reachable in $\mathcal{A}$ for any finite prefix of the given input word. (This set $Q^{\prime}$ is a state in $\mathcal{A}_{\text {det }}$.) What is problematic here is the acceptance condition: while for NFAs the information whether an accept state is reachable is sufficient, for infinite words we need one single run that passes an accept state infinitely often. The latter is not equivalent to the requirement that a state $Q^{\prime}$ with $Q^{\prime} \cap F \neq \varnothing$ is visited infinitely often, since there might be infinitely many possibilities (runs) to enter $F$ at different time points, i.e., for different prefixes of the input word. This, in fact, is the case for the NBA in Figure 4.17. For the input word $\sigma=A B A B A \ldots=(A B)^{\omega}$, the automaton in Figure 4.17 can enter the accept state $q_{1}$ after the second, fourth, sixth, etc., symbol by staying in $q_{0}$ for the first $2 n-1$ symbols and moving with the $n$th $B$ to state $q_{1}$ (for $n=1,2, \ldots$ ). Thus, at infinitely many positions there is the possibility to enter $F$, although there is no run that visits $q_{1}$ infinitely often, since whenever $q_{1}$ has been entered the automaton $\mathcal{A}$ rejects when reading the next $\mathcal{A}$. In fact, the powerset construction applied to the NBA $\mathcal{A}$ in Figure 4.17 yields a DBA $\mathcal{A}_{\text {det }}$ with two reachable states (namely $\left\{q_{0}\right\}$ and $\left\{q_{0}, q_{1}\right\}$ ) for the language consisting of all infinite words with infinitely many $B$ 's, but not for the language given by $(A+B)^{*} B^{\omega}$.

Another example that illustrates why the powerset construction fails for Büchi automata is provided in Exercise 4.16 (page 225).

### 4.3.4 Generalized Büchi Automata

In several applications, other $\omega$-automata types are useful as automata models for $\omega$ regular languages. In fact, there are several variants of $\omega$-automata that are equally expressive as nondeterministic Büchi automata, although they use more general acceptance conditions than the Büchi acceptance condition "visit infinitely often the acceptance set $F$ ". For some of these $\omega$-automata types, the deterministic version has the full power of $\omega$-regular languages. These automata types are not relevant for the remaining chapters of this monograph and will not be treated here. ${ }^{4}$

For the purposes of this monograph, it suffices to consider a slight variant of nondeterministic Büchi automata, called generalized nondeterministic Büchi automata, or GNBA for short. The difference between an NBA and a GNBA is that the acceptance condition for a GNBA requires to visit several sets $F_{1}, \ldots, F_{k}$ infinitely often. Formally, the syntax of a GNBA is as for an NBA, except that the acceptance condition is a set $\mathcal{F}$ consisting of finitely many acceptance sets $F_{1}, \ldots, F_{k}$ with $F_{i} \subseteq Q$. That is, if $Q$ is the state space of the automaton then the acceptance condition of a GNBA is an element $\mathcal{F}$ of $2^{2^{Q}}$. Recall

[^18]that for an NBA, it is an element $F \in 2^{Q}$. The accepted language of a GNBA $\mathcal{G}$ consists of all infinite words which have an infinite run in $\mathcal{G}$ that visits all sets $F_{i} \in \mathcal{F}$ infinitely often. Thus, the acceptance criterion in a generalized Büchi automaton can be understood as the conjunction of a number of Büchi acceptance conditions.

## Definition 4.52. Generalized NBA (GNBA)

A generalized NBA is a tuple $\mathcal{G}=\left(Q, \Sigma, \delta, Q_{0}, \mathcal{F}\right)$ where $Q, \Sigma, \delta, Q_{0}$ are defined as for an NBA (see Definition 4.27 on page 174) and $\mathcal{F}$ is a (possibly empty) subset of $2^{Q}$.

The elements $F \in \mathcal{F}$ are called acceptance sets. Runs in a GNBA are defined as for an NBA. That is, a run in $\mathcal{G}$ for the infinite word $A_{0} A_{1} \ldots \in \Sigma^{\omega}$ is an infinite state sequence $q_{0} q_{1} q_{2} \ldots \in Q^{\omega}$ such that $q_{0} \in Q_{0}$ and $q_{i+1} \in \delta\left(q_{i}, A_{i}\right)$ for all $i \geqslant 0$.

The infinite run $q_{0} q_{1} q_{2} \ldots$ is called accepting if

$$
\forall F \in \mathcal{F} .\left(\begin{array}{l}
\infty \\
\exists
\end{array} j \in \mathbb{N} . q_{j} \in F\right) .
$$

The accepted language of $\mathcal{G}$ is:

$$
\mathcal{L}_{\omega}(\mathcal{G})=\left\{\sigma \in \Sigma^{\omega} \mid \text { there exists an accepting run for } \sigma \text { in } \mathcal{G}\right\} .
$$

Equivalence of GNBAs and the size of a GNBA are defined as for NBAs. Thus, GNBA $\mathcal{G}$ and $\mathcal{G}^{\prime}$ are equivalent if $\mathcal{L}_{\omega}(\mathcal{G})=\mathcal{L}_{\omega}\left(\mathcal{G}^{\prime}\right)$. The size of GNBA $\mathcal{G}$, denoted $|\mathcal{G}|$, equals the number of states and transitions in $\mathcal{G}$.

## Example 4.53. GNBA

Figure 4.19 shows a GNBA $\mathcal{G}$ over the alphabet $2^{A P}$ where $A P=\left\{\right.$ crit $_{1}$, crit $\left._{2}\right\}$ with the acceptance sets $F_{1}=\left\{q_{1}\right\}$ and $F_{2}=\left\{q_{2}\right\}$. That is, $\mathcal{F}=\left\{\left\{q_{1}\right\},\left\{q_{2}\right\}\right\}$. The accepted language is the LT property $P_{\text {live }}$ consisting of all infinite words $A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}$ such that the atomic propositions crit ${ }_{1}$ and $c^{2} t_{2}$ hold infinitely often (possibly at different positions), i.e.,

$$
\stackrel{\infty}{\exists} j \geqslant 0 . \text { crit }_{1} \in A_{j} \quad \text { and } \quad \stackrel{\infty}{\exists} j \geqslant 0 . c_{i i t_{2}} \in A_{j} .
$$

Thus, $P_{\text {live }}$ formalizes the property "both processes are infinitely often in their critical section". Let us justify that indeed $\mathcal{L}_{\omega}(\mathcal{G})=P_{\text {live }}$. This goes as follows.
$" \subseteq "$ : Each accepting run has to pass infinitely often through the edges (labeled with crit or crit $)_{2}$ ) leading to the states $q_{1}$ and $q_{2}$. Thus, in every accepted word $\sigma=A_{0} A_{1} A_{2} \ldots \in$


Figure 4.19: GNBA for "infinitely often processes 1 and 2 are in their critical section".
$\mathcal{L}_{\omega}(\mathcal{G})$ the atomic propositions crit ${ }_{1}$ and $c^{\text {cit }}{ }_{2}$ occur infinitely often as elements of the sets $A_{i} \in 2^{A P}$. Thus, $\sigma \in P_{\text {live }}$.
$" \supseteq "$ : Let $\sigma=A_{0} A_{1} A_{2} \ldots \in P_{\text {live }}$. Since both propositions crit ${ }_{1}$ and crit ${ }_{2}$ occur infinitely often in the symbols $A_{i}$, the GNBA $\mathcal{G}$ can behave for the input word $\sigma$ as follows. $\mathcal{G}$ remains in state $q_{0}$ until the first input symbol $A_{i}$ with crit ${ }_{1} \in A_{i}$ appears. The automaton then moves to state $q_{1}$. From there, $\mathcal{G}$ consumes the next input symbol $A_{i+1}$ and returns to $q_{0}$. It then waits in $q_{0}$ until a symbol $A_{j}$ with $c r i t_{2} \in A_{j}$ occurs, in which case the automaton moves to state $q_{2}$ for the symbol $A_{j}$ and returns to $q_{0}$ on the next symbol $A_{j+1}$. Now the whole procedure restarts, i.e., $\mathcal{G}$ stays in $q_{0}$ while reading the symbols $A_{j+1}, \ldots, A_{\ell-1}$ and moves to $q_{1}$ as soon as the current input symbol $A_{\ell}$ contains crit. And so on. In this way, $\mathcal{G}$ generates an accepting run of the form

$$
q_{0}^{k_{1}} q_{1} q_{0}^{k_{2}} q_{2} q_{0}^{k_{3}} q_{1} q_{0}^{k_{4}} q_{2} q_{0}^{k_{5}} \ldots
$$

for the input word $\sigma$. These considerations show that $P_{\text {live }} \subseteq \mathcal{L}_{\omega}(\mathcal{G})$.

## Remark 4.54. No Acceptance Set

The set $\mathcal{F}$ of acceptance sets of a GNBA may be empty. If $\mathcal{F}=\varnothing$ then $\sigma \in \mathcal{L}_{\omega}(\mathcal{G})$ if and only if there exists an infinite run for $\sigma$ in $\mathcal{G}$. We like to stress the difference with NBA with an empty set of accepting states. For an NBA $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, \varnothing\right)$ there are no accepting runs. Therefore, the language $\mathcal{L}_{\omega}(\mathcal{A})$ is empty. Contrary to that, every infinite run of a GNBA $\mathcal{G}=\left(Q, \Sigma, \delta, Q_{0}, \varnothing\right)$ is accepting.

In fact, every GNBA $\mathcal{G}$ is equivalent to a GNBA $\mathcal{G}^{\prime}$ having at least one acceptance set. This is due to the fact that the state space $Q$ can always be added to the set $\mathcal{F}$ of the acceptance sets without affecting the accepted language of the GNBA. Formally, for GNBA $\mathcal{G}=\left(Q, \Sigma, \delta, Q_{0}, \mathcal{F}\right)$ let GNBA $\mathcal{G}^{\prime}=\left(Q, \Sigma, \delta, Q_{0}, \mathcal{F} \cup\{Q\}\right)$. Then it easily follows that: $\mathcal{L}_{\omega}(\mathcal{G})=\mathcal{L}_{\omega}\left(\mathcal{G}^{\prime}\right)$.

## Remark 4.55. Nonblocking GNBA

As for NBAs, each GNBA $\mathcal{G}$ can be replaced with an equivalent GNBA $\mathcal{G}^{\prime}$, in which all possible behaviors for a given infinite input word yield an infinite run. Such a GNBA $\mathcal{G}^{\prime}$ can be constructed by inserting a nonaccept trapping state, as we did for NBA in the remark on page 187.

Obviously, every NBA can be understood as a GNBA with exactly one acceptance set. Conversely, every GNBA can be transformed into an equivalent NBA:

## Theorem 4.56. From GNBA to NBA

For each $G N B A \mathcal{G}$ there exists an NBA $\mathcal{A}$ with $\mathcal{L}_{\omega}(\mathcal{G})=\mathcal{L}_{\omega}(\mathcal{A})$ and $|\mathcal{A}|=\mathcal{O}(|\mathcal{G}| \cdot|\mathcal{F}|)$ where $\mathcal{F}$ denotes the set of acceptance sets in $\mathcal{G}$.

Proof: Let $\mathcal{G}=\left(Q, \Sigma, \delta, Q_{0}, \mathcal{F}\right)$ be a GNBA. According to the remark on page 194, we may assume without loss of generality that $\mathcal{F} \neq \varnothing$. Let $\mathcal{F}=\left\{F_{1}, \ldots, F_{k}\right\}$ where $k \geqslant 1$. The basic idea of the construction of $\mathcal{A}$ is to create $k$ copies of $\mathcal{G}$ such that the acceptance set $F_{i}$ of the $i$ th copy is connected to the corresponding states of the $(i+1)$ th copy. The accepting


Figure 4.20: Idea for transforming a GNBA into an NBA.
condition for $\mathcal{A}$ consists of the requirement that an accepting state of the first copy is visited infinitely often. This ensures that all other accepting sets $F_{i}$ of the $k$ copies are visited infinitely often too, see Figure 4.20 on page 195 . Formally, let $\mathcal{A}=\left(Q^{\prime}, \Sigma, \delta^{\prime}, Q_{0}^{\prime}, F^{\prime}\right)$ where:

$$
\begin{aligned}
& Q^{\prime}=Q \times\{1, \ldots, k\}, \\
& Q_{0}^{\prime}=Q_{0} \times\{1\}=\left\{\left\langle q_{0}, 1\right\rangle \mid q_{0} \in Q_{0}\right\}, \text { and } \\
& F^{\prime}=F_{1} \times\{1\}=\left\{\left\langle q_{F}, 1\right\rangle \mid q_{F} \in F_{1}\right\} .
\end{aligned}
$$

The transition function $\delta^{\prime}$ is given by

$$
\delta^{\prime}(\langle q, i\rangle, A)= \begin{cases}\left\{\left\langle q^{\prime}, i\right\rangle \mid q^{\prime} \in \delta(q, A)\right\} & \text { if } q \notin F_{i} \\ \left\{\left\langle q^{\prime}, i+1\right\rangle \mid q^{\prime} \in \delta(q, A)\right\} & \text { otherwise }\end{cases}
$$

We thereby identify $\langle q, k+1\rangle$ and $\langle q, 1\rangle$. It is not difficult to check that $\mathcal{A}$ can be constructed in time and space $\mathcal{O}(|\mathcal{G}| \cdot|\mathcal{F}|)$ where $|\mathcal{F}|=k$ is the number of acceptance sets in $\mathcal{G}$. The fact $\mathcal{L}_{\omega}(\mathcal{G})=\mathcal{L}_{\omega}(\mathcal{A})$ can be seen as follows.
$\supseteq$ : For a run of $\mathcal{A}$ to be accepting it has to visit some state $\langle q, 1\rangle$ infinitely often, where $q \in F_{1}$. As soon as a run reaches $\langle q, 1\rangle$, the NBA $\mathcal{A}$ moves to the second copy. From the second copy the next copy can be reached by visiting $\left\langle q^{\prime}, 2\right\rangle$ with $q^{\prime} \in F_{2}$. NBA $\mathcal{A}$ can only return to $\langle q, 1\rangle$ if it goes through all $k$ copies. This is only possible if it reaches an accept state in each copy since that is the only opportunity to move to the next copy. So, for a run to visit $\langle q, 1\rangle$ infinitely often it has to visit some accept state in each copy infinitely often.
$\subseteq$ : By a similar reasoning it can be deduced that every word in $\mathcal{L}_{\omega}(\mathcal{G})$ is also accepting in $\mathcal{A}$.

Example 4.5\%. Transformation of a GNBA into an NBA
Consider the GNBA $\mathcal{G}$ described in Example 4.53 on page 193. The construction indicated in the proof of Theorem 4.56 provides an NBA consisting of two copies (as there are two accept sets) of $\mathcal{G}$, see Figure 4.21. For example,

$$
\delta^{\prime}\left(\left\langle q_{0}, 1\right\rangle,\left\{c r i t_{1}\right\}\right)=\left\{\left\langle q_{0}, 1\right\rangle,\left\langle q_{1}, 1\right\rangle\right\}
$$

since $q_{0} \notin F_{1}=\left\{q_{1}\right\}$ and $\delta^{\prime}\left(\left\langle q_{2}, 2\right\rangle, A\right)=\left\{\left\langle q_{0}, 1\right\rangle\right\}$, since $F_{2}=\left\{q_{2}\right\}$. Thereby, $A \subseteq$ $\left\{\right.$ crit $_{1}$, crit $\left._{2}\right\}$ is arbitrary.

Any NBA can be considered as a GNBA by simply replacing the acceptance set $F$ of the NBA with the singleton set $\mathcal{F}=\{F\}$ for the corresponding GNBA. Using this fact, together with the result that NBAs are equally expressive as $\omega$-regular languages (see Theorem 4.32 on page 178), we obtain by Theorem 4.56 :

## Corollary 4.58. GNBA and $\omega$-Regular Languages

The class of languages accepted by GNBAs agrees with the class of $\omega$-regular languages.

As we have seen before, $\omega$-regular languages are closed under union. This is immediate from the definition of $\omega$-regular expressions and can also simply be proven by means of


Figure 4.21: Example for the transformation of a GNBA into an equivalent NBA.

NBA representations (see Lemma 4.33 on page 179). We now use GNBAs to show that $\omega$-regular languages are closed under intersection too.

## Lemma 4.59. Intersection of GNBA

For $G N B A \mathcal{G}_{1}$ and $\mathcal{G}_{2}$ (both over the alphabet $\Sigma$ ), there exists a GNBA $\mathcal{G}$ with

$$
\mathcal{L}_{\omega}(\mathcal{G})=\mathcal{L}_{\omega}\left(\mathcal{G}_{1}\right) \cap \mathcal{L}_{\omega}\left(\mathcal{G}_{2}\right) \quad \text { and } \quad|\mathcal{G}|=\mathcal{O}\left(\left|\mathcal{G}_{1}\right| \cdot\left|\mathcal{G}_{2}\right|\right) .
$$

Proof: Let $\mathcal{G}_{1}=\left(Q_{1}, \Sigma, \delta_{1}, Q_{0,1}, \mathcal{F}_{1}\right)$ and $\mathcal{G}_{2}=\left(Q_{2}, \Sigma, \delta_{2}, Q_{0,2}, \mathcal{F}_{2}\right)$ where without loss of generality $Q_{1} \cap Q_{2}=\varnothing$. Let $\mathcal{G}$ be the GNBA that results from $\mathcal{G}_{1}$ and $\mathcal{G}_{2}$ by a synchronous product construction (as for NFA) and "lifts" the acceptance sets $F \in \mathcal{F}_{1} \cup \mathcal{F}_{2}$ to acceptance sets in $\mathcal{G}$. Formally,

$$
\mathcal{G}=\mathcal{G}_{1} \otimes \mathcal{G}_{2}=\left(Q_{1} \times Q_{2}, \Sigma, \delta, Q_{0,1} \times Q_{0,2}, \mathcal{F}\right)
$$

where the transition relation $\delta$ is defined by the rule

$$
\frac{q_{1} \xrightarrow{A}{ }_{1} q_{1}^{\prime} \wedge q_{2} \xrightarrow{A}{ }_{2} q_{2}^{\prime}}{\left(q_{1}, q_{2}\right) \xrightarrow{A}\left(q_{1}^{\prime}, q_{2}^{\prime}\right)} .
$$

The acceptance condition in $\mathcal{G}$ is given by

$$
\mathcal{F}=\left\{F_{1} \times Q_{2} \mid F_{1} \in \mathcal{F}_{1}\right\} \cup\left\{Q_{1} \times F_{2} \mid F_{2} \in \mathcal{F}_{2}\right\}
$$

It is now easy to verify that $\mathcal{G}$ has the desired properties.

The same result also holds for union, that is, given two GNBAs $\mathcal{G}_{1}$ and $\mathcal{G}_{2}$ with the same alphabet $\Sigma$ there is a GNBA $\mathcal{G}$ with $\mathcal{L}_{\omega}(\mathcal{G})=\mathcal{L}_{\omega}\left(\mathcal{G}_{1}\right) \cup \mathcal{L}_{\omega}\left(\mathcal{G}_{2}\right)$ and $|\mathcal{G}|=\mathcal{O}\left(\left|\mathcal{G}_{1}\right|+\left|\mathcal{G}_{2}\right|\right)$. The argument is the same as for NBA (Lemma 4.33): we simply may take the disjoint union of the two GNBAs and decide nondeterministically which of them is chosen to "scan" the given input word.

Since GNBA yield an alternative characterization of $\omega$-regular languages, we obtain by Lemma 4.59:

## Corollary 4.60. Intersection of $\omega$-Regular Languages

If $\mathcal{L}_{1}$ and $\mathcal{L}_{2}$ are $\omega$-regular languages over the alphabet $\Sigma$, then so is $\mathcal{L}_{1} \cap \mathcal{L}_{2}$.

### 4.4 Model-Checking $\omega$-Regular Properties

The examples provided in Section 4.3.2 (see page 176 ff .) illustrated that NBAs yield a simple formalism for $\omega$-regular properties. We now address the question how the automatabased approach for verifying regular safety properties can be generalized for the verification of $\omega$-regular properties.

The starting point is a finite transition system $T S=(S, A c t, \rightarrow, I, A P, L)$ without terminal states and an $\omega$-regular property $P$. The aim is to check algorithmically whether $T S \models P$. As we did for regular safety properties, the verification algorithm we present now attempts to show that $T S \notin P$ by providing a counterexample, i.e., a path $\pi$ in $T S$ with $\operatorname{trace}(\pi) \notin$ $P$. (If no such path exists, then $P$ holds for $T S$.) For this, we assume that we are given an automata-representation of the "bad traces" by means of an NBA $\mathcal{A}$ for the complement property $\bar{P}=\left(2^{A P}\right)^{\omega} \backslash P$. The goal is then to check whether $\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$. Note that:

$$
\begin{array}{ll} 
& \operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing \\
\text { if and only if } & \operatorname{Traces}(T S) \cap \bar{P} \neq \varnothing \\
\text { if and only if } & \operatorname{Traces}(T S) \cap\left(2^{A P}\right)^{\omega} \backslash P \neq \varnothing \\
\text { if and only if } & \operatorname{Traces}(T S) \nsubseteq P \\
\text { if and only if } & T S \not \models P
\end{array}
$$

The reader should notice the similarities with regular safety property checking. In that case, we started with an NFA for the bad prefixes for the given safety property $P_{\text {safe }}$ (the bad behaviors). This can be seen as an automata-representation of the complement property $\overline{P_{\text {safe }}}$, see Remark 4.31 on page 177 . The goal was then to find a finite, initial path fragment in TS that yields a trace accepted by the NFA, i.e., a bad prefix for $P_{\text {safe }}$.

Let us now address the problem to check whether $\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$. For this, we can follow the same pattern as for regular safety properties and construct the product $T S \otimes \mathcal{A}$ which combines paths in $T S$ with the runs in $\mathcal{A}$. We then perform a graph analysis in $T S \otimes \mathcal{A}$ to check whether there is a path that visits an accept state of $\mathcal{A}$ infinitely often, which then yields a counterexample and proves $T S \nLeftarrow P$. If no such path in the product exists, i.e., if accept states can be visited at most finitely many times on all paths in the product, then all runs for the traces in $T S$ are nonaccepting, and hence $\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A})=\varnothing$ and thus $T S \models P$

In the sequel, we will explain these ideas in more detail. For doing so, we first introduce the notion of a persistence property. This is a simple type of LT property that will serve to formalize the condition stating that accept states are only visited finitely many times. The problem of verifying $\omega$-regular properties is then shown to be reducible to the persistence checking problem. Recall that the problem of verifying regular safety properties is reducible to the invariant checking problem, see Section 4.2.

### 4.4.1 Persistence Properties and Product

Persistence properties are special types of liveness properties that assert that from some moment on a certain state condition $\Phi$ holds continuously. Stated in other words, $\neg \Phi$ is required to hold at most finitely many times. As for invariants, we assume a representation of $\Phi$ by a propositional logic formula over $A P$.

## Definition 4.61. Persistence Property

A persistence property over AP is an LT property $P_{\text {pers }} \subseteq\left(2^{A P}\right)^{\omega}$ "eventually forever $\Phi$ " for some propositional logic formula $\Phi$ over AP. Formally,

$$
P_{\text {pers }}=\left\{A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega} \mid \stackrel{\infty}{\forall} j . A_{j} \models \Phi\right\}
$$

where $\not \forall j$ is short for $\exists i \geqslant 0 . \forall j \geqslant i$. Formula $\Phi$ is called a persistence (or state) condition of $P_{\text {pers }}$.

Intuitively, a persistence property "eventually forever $\Phi$ " ensures the tenacity of the state
property given by the persistence condition $\Phi$. One may say that $\Phi$ is an invariant after a while; i.e., from a certain point on all states satisfy $\Phi$. The formula "eventually forever $\Phi$ " is true for a path if and only if almost all, i.e., all except for finitely many, states satisfy the proposition $\Phi$.

Our goal is now to show that the question whether $\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A})=\varnothing$ holds can be reduced to the question whether a certain persistence property holds in the product of $T S$ and $\mathcal{A}$. The formal definition of the product $T S \otimes \mathcal{A}$ is exactly the same as for an NFA. For completeness, we recall the definition here:

## Definition 4.62. Product of Transition System and NBA

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system without terminal states and $\mathcal{A}=$ $\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$ a nonblocking NBA. Then, $T S \otimes \mathcal{A}$ is the following transition system:

$$
T S \otimes \mathcal{A}=\left(S \times Q, A c t, \rightarrow^{\prime}, I^{\prime}, A P^{\prime}, L^{\prime}\right)
$$

where $\rightarrow^{\prime}$ is the smallest relation defined by the rule

$$
\frac{s \xrightarrow{\alpha} t \wedge q \xrightarrow{L(t)} p}{\langle s, q\rangle \xrightarrow{\alpha}\langle t, p\rangle}
$$

and where

- $I^{\prime}=\left\{\left\langle s_{0}, q\right\rangle \mid s_{0} \in I \wedge \exists q_{0} \in Q_{0} . q_{0} \xrightarrow{L\left(s_{0}\right)} q\right\}$,
- $A P^{\prime}=Q$ and $L^{\prime}: S \times Q \rightarrow 2^{Q}$ is given by $L^{\prime}(\langle s, q\rangle)=\{q\}$.

Furthermore, let $P_{\operatorname{pers}(\mathcal{A})}$ be the persistence property over $A P^{\prime}=Q$ given by
"eventually forever $\neg F$ "
where $\neg F$ denotes the propositional formula $\bigwedge_{q \in Q} \neg q$ over $A P^{\prime}=Q$.

We now turn to the formal proof that the automata-based approach for checking $\omega$-regular properties relies on checking a persistence property for the product transition system:

## Theorem 4.63. Verification of $\omega$-Regular Properties

Let TS be a finite transition system without terminal states over AP and let $P$ be an $\omega$ regular property over AP. Furthermore, let $\mathcal{A}$ be a nonblocking NBA with the alphabet $2^{A P}$ and $\mathcal{L}_{\omega}(\mathcal{A})=\left(2^{A P}\right)^{\omega} \backslash P$. Then, the following statements are equivalent:
(a) $T S \models P$
(b) $\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A})=\varnothing$
(c) $T S \otimes \mathcal{A} \models P_{\operatorname{pers}(\mathcal{A})}$

Proof: Let $T S=(S, A c t, \rightarrow, I, A P, L)$ and $\mathcal{A}=\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$. The equivalence of (a) and (b) was shown on page 198. Let us now check the equivalence of (b) and (c). For this we show

$$
\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing \text { if and only if } T S \otimes \mathcal{A} \not \vDash P_{\operatorname{pers}(\mathcal{A})}
$$

$" \Leftarrow ":$ Assume $T S \otimes \mathcal{A} \not \vDash P_{\operatorname{pers}(\mathcal{A})}$. Let $\pi^{\prime}=\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots$ be a path in $T S \otimes \mathcal{A}$ such that

$$
\pi^{\prime} \not \models P_{\operatorname{pers}(\mathcal{A})}
$$

Then there are infinitely many indices $i$ with $q_{i} \in F$. The projection of $\pi^{\prime}$ to the states in $T S$ yields a path $\pi=s_{0} s_{1} s_{2} \ldots$ in $T S$.

Let $q_{0} \in Q_{0}$ be an initial state of $\mathcal{A}$ such that $q_{0} \xrightarrow{L\left(s_{0}\right)} q_{1}$. Such a state $q_{0}$ exists, since $\left\langle s_{0}, q_{1}\right\rangle$ is an initial state of $T S \otimes \mathcal{A}$. The state sequence $q_{0} q_{1} q_{2} \ldots$ is a run in $\mathcal{A}$ for the word

$$
\operatorname{trace}(\pi)=L\left(s_{0}\right) L\left(s_{1}\right) L\left(s_{2}\right) \ldots \in \operatorname{Traces}(T S)
$$

Since there are infinitely many $i$ with $q_{i} \in F$, the run $q_{0} q_{1} q_{2} \ldots$ is accepting. Hence,

$$
\operatorname{trace}(\pi) \in \mathcal{L}_{\omega}(\mathcal{A})
$$

This yields $\operatorname{trace}(\pi) \in \operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A})$, and thus $\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$.
$" \Rightarrow ":$ Assume that $\operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$. Then there exists a path in $T S$, say $\pi=s_{0} s_{1} s_{2} \ldots$ with

$$
\operatorname{trace}(\pi)=L\left(s_{0}\right) L\left(s_{1}\right) L\left(s_{2}\right) \ldots \in \mathcal{L}_{\omega}(\mathcal{A})
$$

Let $q_{0} q_{1} q_{2} \ldots$ be an accepting run in $\mathcal{A}$ for $\operatorname{trace}(\pi)$. Then

$$
q_{0} \in Q_{0} \quad \text { and } \quad q_{i} \xrightarrow{L\left(s_{i}\right)} q_{i+1} \quad \text { for all } i \geqslant 0
$$

Furthermore, $q_{i} \in F$ for infinitely many indices $i$. Thus, we can combine $\pi$ and the run $q_{0} q_{1} \ldots$ to obtain a path in the product

$$
\pi^{\prime}=\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots \in \operatorname{Paths}(T S \otimes \mathcal{A})
$$



Figure 4.22: A simple traffic light (upper left), an NBA corresponding to $P_{\text {pers }}$ (upper right), and their product (below).

Since $q_{i} \in F$ for infinitely many $i$, we have $\pi^{\prime} \notin P_{\operatorname{pers}(\mathcal{A})}$. This yields $T S \otimes \mathcal{A} \not \vDash P_{\operatorname{pers}(\mathcal{A})}$.

## Example 4.64. Checking a Persistence Property

Consider a simple traffic light as one typically encounters at pedestrian crossings. It only has two possible modes: red or green. Assume that the traffic light is initially red, and alternates between red and green, see the transition system PedTrLight depicted in the upper left part of Figure 4.22. The $\omega$-regular property $P$ to be checked is "infinitely often green". The complement property $\bar{P}$ thus is "eventually always not green". The NBA depicted in the upper right part of Figure 4.22 accepts $\bar{P}$.

To check the validity of $P$, we first construct the product automaton $\operatorname{PedTrLight} \otimes \mathcal{A}$, see the lower part of Figure 4.22. Note that the state $\left\langle s_{1}, q_{1}\right\rangle$ is unreachable and could be omitted. Let $P_{\operatorname{pers}(\mathcal{A})}=$ "eventually forever $\neg q_{1}$ ". From the lower part of Figure 4.22 one can immediately infer that there is no run of the product automaton that goes infinitely often through a state of the form $\left\langle\cdot, q_{1}\right\rangle$. That is, transition system PedTrLight and NBA $\mathcal{A}$ do not have any trace in common. Thus we conclude:

$$
\operatorname{Ped} \operatorname{Tr} \text { Light } \otimes \mathcal{A} \models \text { "eventually forever" } \neg q_{1}
$$

and consequently (as expected):
PedTrLight $\models$ "infinitely often green".

As a slight alternative, we now consider a pedestrian traffic light that may automatically switch off to save energy. Assume, for simplicity, that the light may only switch off when it is red for some undefined amount of time. Clearly, this traffic light cannot guarantee the validity of $\bar{P}=$ "eventually forever $\neg$ green" as it exhibits a run that (possibly after a while) alternates between red and off infinitely often. This can be formally shown as follows. First, we construct the product automaton, see Figure 4.23 (lower part). For instance, the path $\left\langle s_{0}, q_{0}\right\rangle\left(\left\langle s_{2}, q_{1}\right\rangle\left\langle s_{0}, q_{1}\right\rangle\right)^{\omega}$ goes infinitely often through the accept state $q_{1}$ of $\mathcal{A}$ and generates the trace

$$
\{\text { red }\} \varnothing\{\text { red }\} \varnothing\{\text { red }\} \varnothing \ldots \ldots
$$

That is, $\operatorname{Traces}(\operatorname{PedTrLight}) \cap \mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$, and thus

$$
\operatorname{PedTrLight} ' \otimes \mathcal{A} \not \vDash \text { "eventually forever" } \neg q_{1}
$$

and thus

$$
\text { PedTrLight } t^{\prime} \notin \text { "infinitely often green". }
$$

More concretely, the path $\pi=s_{0} s_{1} s_{0} s_{1} \ldots$ generating the trace

$$
\operatorname{trace}(\pi)=\{\operatorname{red}\} \varnothing\{\text { red }\} \varnothing\{\text { red }\} \varnothing \ldots
$$

has an accepting run $q_{0}\left(q_{1}\right)^{\omega}$ in $\mathcal{A}$.

According to Theorem 4.63, the problem of checking an arbitrary $\omega$-regular property can be solved with algorithms that check a simple type of $\omega$-regular liveness property, namely persistence properties. An algorithm for the latter will be provided in the following section where the transition system under consideration results from the product of the original transition system and an NBA for the undesired behaviors.

### 4.4.2 Nested Depth-First Search

The next problem that we need to tackle is how to establish whether for a given finite transition system TS:

$$
T S \not \vDash P_{\text {pers }}
$$

where $P_{\text {pers }}$ is a persistence property. Let $\Phi$ be the underlying propositional formula that specifies the state condition which has to hold "eventually forever".

The following result shows that answering the question "does $T S \not \vDash P_{\text {pers }}$ hold?" amounts to checking whether $T S$ contains a reachable state violating $\Phi$ that is on a cycle in $T S$.


Figure 4.23: A simple traffic light that can switch off (upper part) and its product (lower part).


Figure 4.24: An example of a run violating "eventually always" $\Phi$.

This can be justified intuitively as follows. Suppose $s$ is a state that is reachable from an initial state in $T S$ and $s \not \models \Phi$. As $s$ is reachable, $T S$ has an initial path fragment that ends in $s$. If $s$ is on a cycle, then this path fragment can be continued by an infinite path that is obtained by traversing the cycle containing $s$ infinitely often. In this way, we obtain a path in $T S$ that visits the $\neg \Phi$-state $s$ infinitely often. But then, $T S \not \vDash P_{\text {pers }}$. This is exemplified in Figure 4.24 where a fragment of a transition system is shown; for simplicity the action labels have been omitted. (Note that - in contrast to invariants - a state violating $\Phi$ which is not on a cycle does not cause the violation of $P_{\text {pers }}$.)

The reduction of checking whether $T S \models P_{\text {pers }}$ to a cycle detection problem is formalized by the following theorem.

## Theorem 4.65. Persistence Checking and Cycle Detection

Let TS be a finite transition system without terminal states over AP, $\Phi$ a propositional formula over AP, and $P_{\text {pers }}$ the persistence property "eventually forever $\Phi$ ". Then, the following statements are equivalent:
(a) $T S \not \vDash \not P_{\text {pers }}$,
(b) There exists a reachable $\neg \Phi$-state $s$ which belongs to a cycle. Formally:

$$
\exists s \in \operatorname{Reach}(T S) . s \not \vDash \Phi \wedge s \text { is on a cycle in } G(T S) .^{5}
$$

Before providing the proof, let us first explain how to obtain an error indication whenever $T S \not \vDash P_{\text {pers }}$. Let $\widehat{\pi}=u_{0} u_{1} u_{2} \ldots u_{k}$ be a path in the graph induced by TS, i.e., $G(T S)$, such that $k>0$ and $s=u_{0}=u_{k}$. Assume $s \not \models \Phi$. That is, $\widehat{\pi}$ is a cycle in $G(T S)$ containing

[^19]a state violating $\Phi$. Let $s_{0} s_{1} s_{2} \ldots s_{n}$ be an initial path fragment of $T S$ such that $s_{n}=s$. Then the concatenation of this initial path fragment and the unfolding of the cycle
$$
\pi=s_{0} s_{1} s_{2} \cdots \underbrace{s_{n}}_{=s} u_{1} u_{2} \cdots \underbrace{u_{k}}_{=s} u_{1} u_{2} \cdots \underbrace{u_{k}}_{=s} \cdots
$$
is a path in $T S$. As state $s \not \models \Phi$ is visited by $\pi$ infinitely often, it follows that $\pi$ does not satisfy "eventually always $\Phi$ ". The prefix
$$
s_{0} s_{1} s_{2} \ldots \underbrace{s_{n}}_{=s} u_{1} u_{2} \ldots \underbrace{u_{k}}_{=s}
$$
can be used as diagnostic feedback as it shows that $s$ may be visited infinitely often.

Proof: Let $T S=(S, A c t, \rightarrow, I, A P)$.
(a) $\Longrightarrow(\mathrm{b})$ : Assume $T S \not \vDash P_{\text {pers }}$, i.e., there exists a path $\pi=s_{0} s_{1} s_{2} \ldots$ in $T S$ such that $\operatorname{trace}(\pi) \notin P_{\text {pers }}$. Thus, there are infinitely many indices $i$ such that $s_{i} \not \models \Phi$. Since TS is finite, there is a state $s$ with $s=s_{i} \not \vDash \Phi$ for infinitely many $i$. As $s$ appears on a path starting in an initial state we have $s \in \operatorname{Reach}(T S)$. A cycle $\widehat{\pi}$ is obtained by any fragment $s_{i} s_{i+1} s_{i+2} \ldots s_{i+k}$ of $\pi$ where $s_{i}=s_{i+k}=s$ and $k>0$.
$(\mathrm{b}) \Longrightarrow(\mathrm{a})$ : Let $s$ and $\widehat{\pi}=u_{0} u_{1} \ldots u_{k}$ be as indicated above, i.e., $s \in \operatorname{Reach}(T S)$ and $\widehat{\pi}$ is a cycle in $T S$ with $s=u_{0}=u_{k}$. Since $s \in \operatorname{Reach}(T S)$, there in an initial state $s_{0} \in I$ and a path fragment $s_{0} s_{1} \ldots s_{n}$ with $s_{n}=s$. Then:

$$
\pi=s_{0} s_{1} s_{2} \cdots \underbrace{s_{n}}_{=s} u_{1} u_{2} \cdots \underbrace{u_{k}}_{=s} u_{1} u_{2} \underbrace{u_{k}}_{=s} \cdots
$$

is a path in $T S$. Since $s \not \models \Phi$, it follows that $\pi$ does not satisfy "eventually forever $\Phi$ ", and thus $T S \not \vDash P_{\text {pers }}$.

## Example 4.66. Pedestrian Traffic Lights Revisited

Consider the transition system model of the simple traffic light that one typically encounters at pedestrian crossings (see Figure 4.22) and the persistence property "eventually forever" $\neg q_{1}$ where $q_{1}$ is the accept state of the NBA $\mathcal{A}$. As there is no reachable cycle in the product transition system that contains a state violating $\neg q_{1}$, i.e., a cycle that contains a state labeled with $q_{1}$, it follows that

$$
\operatorname{Ped} \operatorname{Tr} \operatorname{Light} \otimes \mathcal{A} \models \text { "eventually forever" } \neg q_{1} .
$$

For the traffic light that has the possibility to automatically switch off, it can be inferred directly from the product transition system (see the lower part of Figure 4.23) that there
is a reachable state, e.g., $\left\langle s_{2}, q_{1}\right\rangle \not \models \neg q_{1}$, that lies on a cycle. Thus:

$$
\text { PedTrLight' } \otimes \mathcal{A} \not \vDash \text { "eventually forever" } \neg q_{1} .
$$

Thus, persistence checking for a finite transition system requires the same techniques as checking emptiness of an NBA, see page 184. In fact, the algorithm we suggest below can also be used for checking emptiness in an NBA.

A Naive Depth-First Search Theorem 4.65 entails that in order to check the validity of a persistence property, it suffices to check whether there exists a reachable cycle containing a $\neg \Phi$-state. How to check for such reachable cycles? A possibility is to compute the strongly connected components (SCCs, for short) in $G(T S)$ - this can be done in a worst-case time complexity that is linear in the number of states and transitions - and to check whether one such SCC is reachable from an initial state, contains at least one edge, and, moreover, contains a $\neg \Phi$-state. If indeed such a SCC does exist, $P_{\text {pers }}$ is refuted; otherwise the answer is affirmative.

Although this SCC-based technique is optimal with respect to the asymptotic worst-case time complexity, it is more complex and less adequate for an on-the-fly implementation. In that respect, pure cycle-check algorithms are more appropriate. Therefore, in the sequel we will detail the standard DFS-based cycle detection algorithm.

Let us first recall how for a finite directed graph $G$ and node $v$, it can be checked with a DFS-based approach whether $v$ belongs to a cycle. For this, one may simply start a depth-first search in node $v$ and check for any visited node $w$ whether there is an edge from $w$ to $v$. If so, a cycle has been found: it starts in $v$ and follows the path to node $w$ given by the current stack content and then takes the edge from $w$ to $v$. Vice versa, if no such edge is found, then $v$ does not belong to a cycle. To determine whether $G$ has a cycle, a similar technique can be exploited: we perform a depth-first search to visit all nodes in $G$. Moreover, on investigating the edge from $w$ to $v$ it is checked whether $v$ has already been visited and whether $v$ is still on the DFS stack. If so, then a so-called backward edge has been found which closes a cycle. Otherwise, if no backward edge has been found during the DFS in $G$ then $G$ is acyclic. (A detailed description of this DFS-based cycle detection technique can be found in textbooks on algorithms and data structures, e.g., [100].)

We now DFS-based cycle checks (by searching for backward edges) for persistence checking. The naive approach works in two phases, as illustrated in Algorithm 6:

1. In the first step, all states satisfying $\neg \Phi$ that are reachable from some initial state are determined. This is performed by a standard depth-first search.
2. In the second step, for each reachable $\neg \Phi$-state $s$, it is checked whether it belongs to a cycle. This algorithm (called cycle_check, see Algorithm 7) relies on the technique sketched above: we start a depth-first search in $s$ (with initially empty DFS stack $V$ and initially empty set $T$ of visited states) and check for all states reachable from $s$ whether there is an outgoing backward edge leading to $s$.

This yields an algorithm for checking the validity of the persistence property with quadratic worst-case running time. More precisely, its time complexity is in $\mathcal{O}(N \cdot(|\Phi|+N+M))$ where $N$ is the number of reachable states in $T S$ and $M$ the number of transitions between these reachable states. This can be seen as follows. Visiting all states that are reachable from some initial state takes $\mathcal{O}(N+M+N \cdot|\Phi|)$ as a depth-first search over all states suffices and in each reachable state the validity of $\Phi$ is checked (which is assumed to be linear in the size of $\Phi)$. In the worst case, all states refute $\Phi$, and a depth-first search takes place (procedure cycle_check) for all these states. This takes $\mathcal{O}(N \cdot(N+M))$. Together this yields $\mathcal{O}(N \cdot(|\Phi|+N+M))$.

Several simple modifications of the suggested technique are possible to increase efficiency. For instance, cycle_check $\left(s^{\prime}\right)$ can be invoked inside visit $(s)$ immediately before or after $s^{\prime}$ is inserted into $R_{\neg \Phi}$, in which case the whole persistence checking algorithm can abort with the answer "no" if cycle_check ( $s^{\prime}$ ) returns true. However, the quadratic worst-case running time cannot be avoided if the cycle check algorithm for the $\neg \Phi$-states relies on separate depth-first searches. The problem is that certain fragments of TS might be reachable from different $\neg \Phi$-states. These fragments are (re-)explored in the depth-first searches (cycle_check) invoked by several $\neg \Phi$-states. To obtain linear running time, we aim at a cycle detection algorithm that searches for backward edges leading to one of the $\neg \Phi$-states and ensures that any state is visited at most once in the depth-first searches for the cycle detection. This will be explained in the following subsection.

A Nested Depth-First Search Algorithm The rough idea of the linear-time cycle detection-based persistence checking algorithm is to perform two depth-first searches (DFSs) in TS in an interleaved way. The first (outer) depth-first search serves to encounter all reachable $\neg \Phi$-states. The second (inner) depth-first search seeks backward edges leading to a $\neg \Phi$-state. The inner depth-first search is nested in the outer one in the following sense: whenever a $\neg \Phi$-state $s$ has been fully expanded by the outer depth-first search, then the inner depth-first search continues with state $s$ and visits all states $s^{\prime}$ that are reachable from $s$ and that have not yet been visited in the inner depth-first search before. If no backward edge has been found when treating $s$ in the inner DFS, then the outer DFS continues until the next $\neg \Phi$-state $t$ has been fully expanded, in which case the inner DFS proceeds with $t$.

```
Algorithm 6 Naive persistence checking
Input: finite transition system \(T S\) without terminal states, and proposition \(\Phi\)
Output: "yes" if \(T S \models\) "eventually forever \(\Phi\) ", otherwise "no".
```

```
set of states R:= \varnothing; R}\mp@subsup{R}{\neg\Phi}{}:=\varnothing\mathrm{ ;
```

set of states R:= \varnothing; R}\mp@subsup{R}{\neg\Phi}{}:=\varnothing\mathrm{ ;
stack of states U:= ;
stack of states U:= ;
set of states T:= \varnothing;
set of states T:= \varnothing;
stack of states V:= ;
stack of states V:= ;
for all }s\inI\R\mathrm{ do visit(s); od
for all }s\inI\R\mathrm{ do visit(s); od
for all }s\in\mp@subsup{R}{\neg\Phi}{}\mathrm{ do
for all }s\in\mp@subsup{R}{\neg\Phi}{}\mathrm{ do
T:=\varnothing;V := \&; (* initialize set T and stack V *)
T:=\varnothing;V := \&; (* initialize set T and stack V *)
if cycle_check(s) then return "no"
if cycle_check(s) then return "no"
od
od
return "yes" (* none of the }\neg\Phi\mathrm{ -states belongs to a cycle *)
return "yes" (* none of the }\neg\Phi\mathrm{ -states belongs to a cycle *)
procedure visit (state $s$ )
$\operatorname{push}(s, U)$; (* push $s$ on the stack *)
$R:=R \cup\{s\} ; \quad$ (* mark $s$ as reachable *)
repeat
$s^{\prime}:=\operatorname{top}(U)$;
if $\operatorname{Post}\left(s^{\prime}\right) \subseteq R$ then
pop $(U)$;
if $s^{\prime} \not \vDash \Phi$ then $R_{\neg \Phi}:=R_{\neg \Phi} \cup\left\{s^{\prime}\right\} ;$ fi
else
let $s^{\prime \prime} \in \operatorname{Post}\left(s^{\prime}\right) \backslash R$
$\operatorname{push}\left(s^{\prime \prime}, U\right)$;
$R:=R \cup\left\{s^{\prime \prime}\right\} ; \quad$ (* state $s^{\prime \prime}$ is a new reachable state *)
fi
until $(U=\varepsilon)$
endproc

```

\section*{Algorithm 7 Cycle detection}

Input: finite transition system \(T S\) and state \(s\) in \(T S\) with \(s \not \vDash \Phi\)
Output: true if \(s\) lies on a cycle in \(T S\), otherwise false
\begin{tabular}{ll} 
(* \(T\) organizes the set of states that have been visited, \(V\) serves as DFS stack. & \(\left.{ }^{*}\right)\) \\
(* In the standard approach to check whether there is a backward edge to \(s\), & \(\left.{ }^{*}\right)\) \\
\(\left({ }^{*} T\right.\) and \(V\) are initially empty. & \(\left.*^{*}\right)\)
\end{tabular}
procedure boolean cycle_check(state \(s\) )
boolean cycle_found \(:=\) false; (* no cycle found yet *)
\(\operatorname{push}(s, V)\); (* push \(s\) on the stack *)
\(T:=T \cup\{s\} ;\)
repeat
\(s^{\prime}:=\operatorname{top}(V) ; \quad \quad\) (* take top element of \(V^{*}\) )
if \(s \in \operatorname{Post}\left(s^{\prime}\right)\) then
cycle_found \(:=\) true; \(\quad\left({ }^{*}\right.\) if \(s \in \operatorname{Post}\left(s^{\prime}\right)\), a cycle is found \(\left.{ }^{*}\right)\)
push \((s, V)\);
(* push \(s\) on the stack \({ }^{*}\) )
else
if \(\operatorname{Post}\left(s^{\prime}\right) \backslash T \neq \varnothing\) then
let \(s^{\prime \prime} \in \operatorname{Post}\left(s^{\prime}\right) \backslash T\);
push \(\left(s^{\prime \prime}, V\right)\); (* push an unvisited successor of \(s^{\prime *}\) )
\(T:=T \cup\left\{s^{\prime \prime}\right\} ; \quad\) (* and mark it as reachable *)
else
\(\operatorname{pop}(V) ; \quad\left(*\right.\) unsuccessful cycle search for \(s^{\prime}{ }^{*}\) )
fi
fi
until \(((V=\varepsilon) \vee\) cycle_found \()\)
return cycle_found
endproc
```

Algorithm 8 Persistence checking by nested depth-first search
Input: transition system $T S$ without terminal states, and proposition $\Phi$
Output: "yes" if $T S \models$ "eventually forever $\Phi$ ", otherwise "no" plus counterexample
set of states $R:=\varnothing$;
stack of states $U:=\varepsilon$;
set of states $T:=\varnothing$;
stack of states $V:=\varepsilon$;
boolean cycle_found := false;
while ( $I \backslash R \neq \varnothing \wedge \neg$ cycle_found) do

```
```

    let }s\inI\R\mathrm{ ;
    ```
    let }s\inI\R\mathrm{ ;
    reachable_cycle( \(s\) );
od
if \(\neg\) cycle_found then
    return ("yes") (* TS \(=\) "eventually forever \(\Phi\) " *)
else
    return ("no", reverse(V.U)) (* stack contents yield a counterexample *)
fi
procedure reachable_cycle (state \(s\) )
    push \((s, U)\);
    (* push \(s\) on the stack \({ }^{*}\) )
    \(R:=R \cup\{s\} ;\)
    repeat
        \(s^{\prime}:=\operatorname{top}(U)\);
        if \(\operatorname{Post}\left(s^{\prime}\right) \backslash R \neq \varnothing\) then
            let \(s^{\prime \prime} \in \operatorname{Post}\left(s^{\prime}\right) \backslash R\);
            \(\operatorname{push}\left(s^{\prime \prime}, U\right)\); (* push the unvisited successor of \(s^{\prime *}\) )
            \(R:=R \cup\left\{s^{\prime \prime}\right\} ; \quad\) (* and mark it reachable *)
        else
            \(\operatorname{pop}(U)\); (* outer DFS finished for \(\left.s^{\prime *}\right)\)
            if \(s^{\prime} \not \models \Phi\) then
                cycle_found \(:=\operatorname{cycle} \_\)check \(\left(s^{\prime}\right)\); (* proceed with the inner *)
                            (* DFS in state \(s^{\prime *}\) )
            fi
        fi
    until \(((U=\varepsilon) \vee\) cycle_found \() \quad(*\) stop when stack for the outer *)
    (* DFS is empty or cycle found *)
endproc
```

This algorithm is called nested depth-first search. Algorithm 8 shows the pseudocode for the outer depth-first search (called reachable_cycle) which invokes the inner depth-first search (cycle_check, cf. Algorithm 7 on page 210). Later we will explain that it is important that cycle_check $(s)$ is called immediately after the outer depth-first search has visited all successors of $s$. The major difference from the naive approach is that cycle_check $(s)$ reuses the information of previous calls of cycle_check $(\cdot)$ and ignores all states in $T$. When cycle_check $(s)$ is invoked then $T$ consists of all states that have been visited in the inner DFS before, i.e., during the execution of cycle_check $(u)$ called prior to cycle_check $(s)$.

An interesting aspect of this nested depth-first strategy is that once a cycle is determined containing a $\neg \Phi$-state $s$, then a path to $s$ can be computed easily: stack $U$ for the outer DFS contains a path fragment from the initial state $s_{0} \in I$ to $s$ (in reversed order), while stack $V$ for the inner DFS - as maintained by the procedure cycle_check $(\cdot)$ contains a cycle from state $s$ to $s$ (in reversed order). Concatenating these path fragments thus provides an error indication in the same vein as described before in the proof of Theorem 4.65 on page 205.

## Example 4.67. Running the Nested DFS Algorithm

Consider the transition system depicted in Figure 4.25 and assume that $s_{0} \models \Phi, s_{3} \models \Phi$, whereas $s_{1} \not \vDash \Phi$ and $s_{2} \not \vDash \Phi$. Consider the scenario in which $s_{2}$ is considered as a first successor of $s_{0}$, i.e., $s_{2}$ is considered in the outer depth-first search prior to considering state $s_{1}$. This means that the order in which the states are put by Algorithm 8 on the stack $U$ equals $\left\langle s_{1}, s_{3}, s_{2}, s_{0}\right\rangle$ where we write the stack content from the top to the bottom, i.e., $s_{1}$ is the top element (the last element that has been pushed on the stack). Besides, $R=S$. Thus, the outer DFS takes $s_{1}$ as the top element of $U$ and verifies $\operatorname{Post}\left(s_{1}\right) \subseteq R$. As $s_{1} \not \models \Phi$, the invocation cycle_check $\left(s_{1}\right)$ takes place and $s_{1}$ is deleted from stack $U$ for the outer DFS. This yields that $s_{1}$ is put on the stack $V$ for the inner DFS, followed by its only successor $s_{3}$. Then the cycle $s_{1} \rightarrow s_{3} \rightarrow s_{1}$ is detected, cycle_check $\left(s_{1}\right)$ yields true, and as a result, reachable_cycle $\left(s_{0}\right)$ terminates with the conclusion that $P_{\text {pers }}$ on $\Phi$ is refuted. An error indication is obtained by first concatenating the content of $V=\left\langle s_{1}, s_{3}, s_{1}\right\rangle$ and $U=\left\langle s_{3}, s_{2}, s_{0}\right\rangle$ and then reversing the order. This yields the path $s_{0} s_{2} s_{3} s_{1} s_{3} s_{1}$, which is indeed a prefix of a run refuting "eventually always $\Phi$ ".

The soundness of the nested depth-first search is no longer trivial because - by the treatment of $T$ as a global variable - not all states reachable from $s$ are explored in cycle_check $(s)$, but only those states that have not been visited before in the inner depthfirst search. Thus, we could imagine that there is a cycle with $\neg \Phi$-states which will not be found with the nested depth-first search: this cycle might have been explored during cycle_check $(u)$ where $u$ does not belong to this or any other cycle. Then, none of the following procedure calls of $\operatorname{cycle} \_\operatorname{check}(\cdot)$ will find this cycle. The goal is now to show


Figure 4.25: An example of a transition system for nested DFS.
that this cannot happen, i.e., if there is a reachable cycle with $\neg \Phi$-states, then the nested depth-first search will find such a cycle.

In fact, things are more tricky than it seems. In Algorithm 8 it is important that the invocations to the procedure cycle_check $(s)$ for $s \in \operatorname{Reach}(T S)$ and $s \not \vDash \Phi$ occur in appropriate order. Let $s \not \vDash \Phi$. Then cycle_check $(s)$ is only invoked if $\operatorname{Post}(s) \subseteq R$, i.e., when all states reachable from $s$ have been encountered and visited. ${ }^{6}$ So, the invocation cycle_check $(s)$ is made immediately once all states that are reachable from $s$ have been visited and expanded in the outer depth-first search. As a result, Algorithm 8 satisfies the following property: if $s^{\prime}$ is a successor of state $s$ in the visit-order of the outer depth-first search and $s \not \vDash \Phi$ and $s^{\prime} \not \models \Phi$, then the invocation cycle_check $\left(s^{\prime}\right)$ occurs prior to the invocation cycle_check $(s)$.

## Example 4.68. Modifying the Nested DFS

Let us illustrate by an example that the nested depth-first search algorithm would be wrong if the outer and inner DFS are interleaved in an arbitrary way. We consider again the transition system in Figure 4.25 on page 213. We start the outer DFS with the initial state $s_{0}$ and assume that state $s_{2}$ is visited prior to $s_{1}$. Let us see what happens if we do not wait until $s_{2}$ has been fully expanded in the outer DFS and start cycle_check $\left(s_{2}\right)$ immediately. Then, cycle_check $\left(s_{2}\right)$ visits $s_{3}$ and $s_{1}$ and returns false, since there is no backward edge leading to state $s_{2}$. Thus, cycle_check $\left(s_{2}\right)$ yields $T=\left\{s_{2}, s_{3}, s_{1}\right\}$ (and $V=\varepsilon)$. Now the outer DFS proceeds and visits $s_{3}$ and $s_{1}$. It calls cycle_check $\left(s_{1}\right)$ which fails to find the cycle $s_{1} s_{3} s_{1}$. Note that cycle_check $\left(s_{1}\right)$ immediately halts and returns false since $s_{1} \in T=\left\{s_{2}, s_{3}, s_{1}\right\}$. Thus, the nested depth-first search would return the wrong answer "yes".

[^20]
## Theorem 4.69. Correctness of the Nested DFS

Let TS be a finite transition system over AP without terminal states, $\Phi$ a propositional formula over AP, and $P_{\text {pers }}$ the persistence property "eventually forever $\Phi$ ". Then:

Algorithm 8 returns the answer "no" if and only if TS $\not \vDash P_{\text {pers }}$.

## Proof:

$\Rightarrow$ : This follows directly from the fact that the answer "false" is only obtained when an initial path fragment of the form $s_{0} \ldots s \ldots s$ has been encountered for some $s_{0} \in I$ and $s \not \vDash \Phi$. But then, $T S \not \vDash P_{\text {pers }}$.
$\Leftarrow:$ To establish this direction, we first prove that the following condition holds:
$\left.{ }^{*}\right)$ On invoking cycle_check $(s)$, there is no cycle $s_{0}^{\prime} s_{1}^{\prime} \ldots s_{k}^{\prime}$ in $T S$ such that:

$$
\left\{s_{0}^{\prime}, s_{1}^{\prime}, \ldots, s_{k}^{\prime}\right\} \cap T \neq \varnothing \quad \text { and } \quad s \in\left\{s_{0}^{\prime}, \ldots, s_{k}^{\prime}\right\}
$$

So, on invoking cycle_check $(s)$, there is no cycle containing $s$ and a state in $T$. Statement $\left.{ }^{*}\right)$ ensures that in the inner DFS, all already visited states in $T$-the states that were visited during an earlier cycle detection - can be safely ignored during the next cycle search. In other words, if $s$ belongs to a cycle then cycle_check $(s)$ has to find one such cycle. We prove the statement $\left(^{*}\right)$ by contradiction. Consider the invocation cycle_check $(s)$. Assume there exists a cycle $s_{0}^{\prime} s_{1}^{\prime} \ldots s_{k}^{\prime}$ such that

$$
s_{0}^{\prime}=s_{k}^{\prime}=s \quad \text { and } \quad s \not \vDash \Phi \quad \text { and } \quad\left\{s_{0}^{\prime}, s_{1}^{\prime}, \ldots, s_{k}^{\prime}\right\} \cap T \neq \varnothing
$$

Without loss of generality we assume that $s$ is the first state for which this condition holds on invoking cycle_check $(s)$. More precisely, we assume:
$(+)$ For all states $\widehat{s}$ with $\widehat{s} \not \vDash \Phi$ where cycle_check $(\widehat{s})$ has been invoked prior to cycle_check $(s)$, there is no cycle containing $\widehat{s}$ and a state in $T$, on invoking cycle_check $(s)$.

Let $t \in\left\{s_{0}^{\prime}, \ldots, s_{k}^{\prime}\right\} \cap T$, i.e., $t$ is a state on a cycle with state $s$. Since $t \in T$ on the invocation cycle_check $(s)$, there must be a state, $u$ say, for which cycle_check $(\cdot)$ has been invoked earlier and during this search $t$ has been encountered (and added to $T$ ). Thus, we have $u \not \vDash \Phi$ and the following conditions (4.1) and (4.2):

$$
\begin{equation*}
\text { cycle_check }(u) \text { was invoked prior to cycle_check }(s) \tag{4.1}
\end{equation*}
$$

as a result of cycle_check $(u), t$ was visited and added to $T$
Obviously, from (4.2), it follows that state $t$ is reachable from $u$. As the states $s$ and $t$ are on a cycle, and $t$ is reachable from $u$, we have that $s$ is reachable from $u$. This situation is sketched in the following figure:


Now consider the outer depth-first search, i.e., reachable_cycle $(\cdot)$, and discuss the following cases 1 and 2:

1. $u$ has been visited prior to $s$ in the outer DFS, i.e., $s$ has been pushed on stack $U$ after $u$.

Since $s$ is reachable from $u$, state $s$ is visited during the expansion of $u$ in the outer DFS and taken from stack $U$ before $u$. Hence, cycle_check $(s)$ is invoked prior to cycle_check ( $u$ ) which contradicts (4.1).
2. $u$ has been visited after $s$ in the outer DFS, i.e., $s$ has been pushed on stack $U$ before $u$.

By (4.1), $s$ is still in stack $U$ when cycle_check $(u)$ is invoked. This yields that $u$ is reachable from $s$. But as $s$ is reachable from $u$, this means that $s$ and $u$ are on a cycle. This cycle or another cycle with state $u$ would have been encountered during cycle_check $(u)$ because of $(+)$ and Algorithm 8 would have terminated without invoking cycle_check $(s)$.

It remains to discuss the complexity of the nested depth-first search algorithm. Since $T$ is increasing during the execution of the nested depth-first search (i.e., we insert states in $T$, but never take them out) any state in $T S$ is visited at most once in the inner depthfirst search, when ranging over all procedure calls of cycle_check $(\cdot)$. The same holds for the outer depth-first search since it is roughly a standard depth-first search. Thus, each reachable state $s^{\prime}$ in $T S$ is taken

- at most once ${ }^{7}$ as the top element of stack $U$ in the outer depth-first search;
- at most once as the top element of stack $V$ in the inner depth-first search (when ranging over all calls of cycle_check).

In particular, this observation yields the termination of Algorithm 8. The cost caused by the top-element $s^{\prime}$ for the body of the repeat loop in cycle_check (inner depth-first search) and in reachable_cycle (outer depth-first search) is $\mathcal{O}\left(\left|\operatorname{Post}\left(s^{\prime}\right)\right|\right)$. Thus, the worst-case time complexity of Algorithm 8 is linear in the size of the reachable fragment of TS, but has the chance to terminate earlier when a cycle has been found. In the following theorem we also take into account that $\Phi$ might be a complex formula and evaluating the truth value of $\Phi$ for a given state (by means of its label) requires $\mathcal{O}(|\Phi|)$ steps.

## Theorem 4.70. Time Complexity of Persistence Checking

The worst-case time complexity of Algorithm 8 is in $\mathcal{O}((N+M)+N \cdot|\Phi|)$ where $N$ is the number of reachable states, and $M$ the number of transitions between the reachable states.

The space complexity is bounded above by $\mathcal{O}(|S|+|\rightarrow|)$ where $S$ is the state space of $T S$ and $|\rightarrow|$ the number of transitions in TS. This is an adequate bound if we assume a representation of $T S$ by adjacency lists. However, in the context of model checking the starting point is typically not an explicit representation of the composite transition system, but a syntactic description of the concurrent processes, e.g., by high-level modeling languages with a program graph or channel system semantics. Such syntactic descriptions are typically much smaller than the resulting transition system. (Recall the state explosion problem which appears through parallel composition and the unfolding of program graphs into transition systems, see page 77 ff .) In the persistence checking algorithm, we may assume that the elements in $\operatorname{Post}\left(s^{\prime}\right)$ are generated on the fly by means of the semantic rules for the transition relation. Ignoring the space required for the syntactic descriptions of the processes, the additional space requirements for the presented persistence checking algorithm is $\mathcal{O}(N)$ for the sets $T$ and $R$ and the stacks $U$ and $V$, where $N$ is the number of reachable states in TS. In fact, the representation of $T$ and $R$ is the most (space-)critical aspect when implementing the nested depth-first search and running it on large examples. Typically, $T$ and $R$ are organized using appropriate hash techniques. In fact, $T$ and $R$ can even be represented by a single hash table where the entries are pairs $\langle s, b\rangle$ with $b \in\{0,1\}$. The meaning of $\langle s, 0\rangle$ is that $s$ is in $R$, but not in $T$ (i.e., $s$ has been visited in the outer, but not yet in the inner DFS). The pair $\langle s, 1\rangle$ means that $s$ has been visited in both the outer and the inner DFS. The single bit $b$ is sufficient to cover all possible cases, since $T$ is always a subset of $R$.

[^21]Another simple observation can speed up the nested depth-first search in case the persistence property is violated: whenever the inner DFS cycle_check $(s)$ reaches a state $t$ which is on the stack $U$ (the DFS stack for the outer DFS), then the upper part of $U$ yields a path fragment from $t$ to $s$, while the content of the DFS stack $V$ for the inner DFS describes a path fragment from $s$ to $t$. Thus, a cycle has been found which visits $s$ infinitely often and the nested DFS may abort with a counterexample. To support checking whether state $t$ is contained in stack $U$, a further bit can be added to the entries in the hash table for representing $T$ and $R$. I.e., we then have to deal with a hash table for triples $\langle s, b, c\rangle$ with $s \in R$ and $b, c \in\{0,1\}$ depending on whether $s$ is in $T$ (in which case $b=1$ ) and whether $s$ is in $U$ (in which case $c=1$ ). This leads to a slight variant of Algorithm 8 which is often faster and generates smaller counterexamples than the original version.

### 4.5 Summary

- NFAs and DFAs are equivalent automata models for regular languages and can serve to represent the bad prefixes of regular safety properties.
- Checking a regular safety property on a finite transition system is solvable by checking an invariant on a product automaton, and thus amounts to solving a reachability problem.
- $\omega$-Regular languages are languages of infinite words that can be described by $\omega$ regular expressions.
- NBAs are acceptors for infinite words. The syntax is as for NFAs. The accepted language of an NBA is the set of all infinite words that have a run where an accept state is visited infinitely often.
- NBAs can serve to represent $\omega$-regular properties.
- The class of languages that are recognized by NBAs agrees with the class of $\omega$-regular languages.
- DBAs are less powerful than NBAs and fail, for instance, to represent the persistence property "eventually forever $a$ ".
- Generalized NBAs are defined as NBAs, except that they require repeated visits for several acceptance sets. Their expressiveness is the same as for NBAs.
- Checking an $\omega$-regular property $P$ on a finite transition system $T S$ can be reduced to checking the persistence property "eventually forever no accept state" in the product of $T S$ and an NBA for the undesired behaviors (i.e., the complement property $\bar{P}$ ).
- Persistence checking requires checking the existence of a reachable cycle containing a state violating the persistence condition. This is solvable in linear time by a nested depth-first search (or by analyzing the strongly connected components). The same holds for the nonemptiness problem for NBAs which boils down to checking the existence of a reachable cycle containing an accept state.
- The nested depth-first search approach consists of the interleaving of two depth-first searches: one for encountering the reachable states, and one for cycle detection.


### 4.6 Bibliographic Notes

Finite automata and regular languages. The first papers on finite automata were published in the nineteen fifties by Huffman [217], Mealy [291], and Moore [303] who used deterministic finite automata for representing sequential circuits. Regula r expressions and their equivalence to finite automata goes back to Kleene [240]. Rabin and Scott [350] presented various algorithms on finite automata, including the powerset construction. The existence of minimal DFAs relies on results stated by Myhill [309] and Nerode [313]. The $\mathcal{O}(N \log N)$ minimization algorithm we mentioned at the end of Section 4.1 has been suggested by Hopcroft [213]. For other algorithms on finite automata, a detailed description of the techniques sketched here and other aspects of regular languages, we refer to the text books $[272,363,214,383]$ and the literature mentioned therein.

Automata over infinite words. Research on automata over infinite words (and trees) started in the nineteen sixties with the work of Büchi [73], Trakhtenbrot [392], and Rabin [351] on decision problems for mathematical logics. At the same time, Muller [307] studied a special type of deterministic $\omega$-automata (today called Muller automata) in the context of asynchronous circuits. The equivalence of nondeterministic Büchi automata and $\omega$-regular expressions has been shown by McNaughton [290]. He also established a link between NBAs and deterministic Muller automata [307] by introducing another acceptance condition that has been later formalized by Rabin [351]. (The resulting $\omega$-automata type is today called Rabin automata.) An alternative transformation from NBA to deterministic Rabin automata has been presented by Safra [361]. Unlike Büchi automata, the nondeterministic and deterministic versions of Muller and Rabin automata are equally expressive and yield automata-characterizations of $\omega$-regular languages. The same holds for several other types of $\omega$-automata that have been introduced later, e.g., Streett automata [382] or automata with the parity condition [305].

The fact that $\omega$-regular languages are closed under complementation (we stated this result without proof) can be derived easily from deterministic automata representations. The proof of Theorem 4.50 follows the presentation given in the book by Peled [327]. Various
decision problems for $\omega$-automata have been addressed by Landweber [261] and later by Emerson and Lei [143] and Sistla, Vardi, and Wolper [373]. For a survey of automata on infinite words, transformations between the several classes of $\omega$-automata, complementation operators and other algorithms on $\omega$-automata, we refer to the articles by Choueka [81], Kaminsky [229], Staiger [376], and Thomas [390, 391]. An excellent overview of the main concepts of and recent results on $\omega$-automata is provided by the tutorial proceedings [174].

Automata and linear-time properties. The use of Büchi automata for the representation and verification of linear-time properties goes back to Vardi and Wolper [411, 412] who studied the connection of Büchi automata with linear temporal logic. Approaches with similar automata models have been developed independently by Lichtenstein, Pnueli, and Zuck [274] and Kurshan [250]. The verification of (regular) safety properties has been described by Kupferman and Vardi [249]. The notion of persistence property has been introduced by Manna and Pnueli [282] who provided a hierarchy of temporal properties. The nested depth-first algorithm (see Algorithm 8) originates from Courcoubetis et al. [102] and its implementation in the model checker SPIN has been reported by Holzmann, Peled, and Yannakakis [212]. The Mur $\varphi$ verifier developed by Dill [132] focuses on verifying safety properties. Variants of the nested depth-first search have been proposed by several authors, see, e.g., [106, 368, 161, 163]. Approaches that treat generalized Büchi conditions (i.e., conjunctions of Büchi conditions) are discussed in [102, 388, 184, 107]. Further implementation details of the nested depth-first search approach can be found in the book by Holzman [209].

### 4.7 Exercises

Exercise 4.1. Let $A P=\{a, b, c\}$. Consider the following LT properties:
(a) If $a$ becomes valid, afterward $b$ stays valid ad infinitum or until $c$ holds.
(b) Between two neighboring occurrences of $a, b$ always holds.
(c) Between two neighboring occurrences of $a, b$ occurs more often than $c$.
(d) $a \wedge \neg b$ and $b \wedge \neg a$ are valid in alternation or until $c$ becomes valid.

For each property $P_{i}(1 \leqslant i \leqslant 4)$, decide if it is a regular safety property (justify your answers) and if so, define the NFA $\mathcal{A}_{i}$ with $\mathcal{L}\left(\mathcal{A}_{i}\right)=\operatorname{BadPref}\left(P_{i}\right)$. (Hint: You may use propositional formulae over the set AP as transition labels.)

Exercise 4.2. Let $n \geqslant 1$. Consider the language $\mathcal{L}_{n} \subseteq \Sigma^{*}$ over the alphabet $\Sigma=\{A, B\}$ that consists of all finite words where the symbol $B$ is on position $n$ from the right, i.e., $\mathcal{L}$ contains exactly the words $A_{1} A_{2} \ldots A_{k} \in\{A, B\}^{*}$ where $k \geqslant n$ and $A_{k-n+1}=B$. For instance, the word $A B B A A B A B$ is in $\mathcal{L}_{3}$.
(a) Construct an NFA $\mathcal{A}_{n}$ with at most $n+1$ states such that $\mathcal{L}\left(\mathcal{A}_{n}\right)=\mathcal{L}_{n}$.
(b) Determinize this NFA $\mathcal{A}_{n}$ using the powerset construction algorithm.

Exercise 4.3. Consider the transition system $T S_{S e m}$ for the two-process mutual exclusion with a semaphore (see Example 2.24 on page 43) and $T S_{P e t}$ for Peterson's algorithm (see Example 2.25 on page 45).
(a) Let $P_{\text {safe }}$ be the regular safety property "process 1 never enters its critical section from its noncritical section (i.e., process 1 must be in its waiting location before entering the critical section)" and $A P=\left\{\right.$ wait $_{1}$, crit $\left._{1}\right\}$.
(i) Depict an NFA for the minimal bad prefixes for $P_{\text {safe }}$.
(ii) Apply the algorithm in Section 4.2 to verify $T S_{\text {Sem }} \models P_{\text {safe }}$.
(b) Let $P_{\text {safe }}$ be the safety property "process 1 never enters its critical section from a state where $x=2$ " and $A P=\left\{\right.$ crit $\left._{1}, x=2\right\}$.
(i) Depict an NFA for the minimal bad prefixes for $P_{\text {safe }}$.
(ii) Apply the algorithm in Section 4.2 to verify $T S_{\text {Pet }} \not \vDash P_{\text {safe }}$. Which counterexample is returned by the algorithm?

Exercise 4.4. Let $P_{\text {safe }}$ be a safety property. Prove or disprove the following statements:
(a) If $\mathcal{L}$ is a regular language with $\operatorname{MinBadPref}\left(P_{\text {safe }}\right) \subseteq \mathcal{L} \subseteq \operatorname{BadPref}\left(P_{\text {safe }}\right)$, then $P_{\text {safe }}$ is regular.
(b) If $P_{\text {safe }}$ is regular, then any $\mathcal{L}$ for which $\operatorname{MinBadPref}\left(P_{\text {safe }}\right) \subseteq \mathcal{L} \subseteq \operatorname{BadPref}\left(P_{\text {safe }}\right)$ is regular.

Exercise 4.5. Let $A P=\{a, b, c\}$. Consider the following NFA $\mathcal{A}$ (over the alphabet $2^{A P}$ ) and the following transition system TS:
$\mathcal{A}$ :


TS :


Construct the product $T S \otimes \mathcal{A}$ of the transition system and the NFA.

Exercise 4.6. Consider the following transition system TS

and the regular safety property

$$
P_{\text {safe }}=\begin{aligned}
& \text { "always if } a \text { is valid and } b \wedge \neg c \text { was valid somewhere before, } \\
& \text { then } a \text { and } b \text { do not hold thereafter at least until } c \text { holds" }
\end{aligned}
$$

As an example, it holds:

$$
\begin{array}{ll}
\{b\} \varnothing\{a, b\}\{a, b, c\} & \in \operatorname{pref}\left(P_{\text {safe }}\right) \\
\{a, b\}\{a, b\} \varnothing\{b, c\} & \in \operatorname{pref}\left(P_{\text {safe }}\right) \\
\{b\}\{a, c\}\{a\}\{a, b, c\} & \in \operatorname{BadPref}\left(P_{\text {safe }}\right) \\
\{b\}\{a, c\}\{a, c\}\{a\} & \in \operatorname{BadPref}\left(P_{\text {safe }}\right)
\end{array}
$$

Questions:
(a) Define an NFA $\mathcal{A}$ such that $\mathcal{L}(\mathcal{A})=\operatorname{MinBadPref}\left(P_{\text {safe }}\right)$.
(b) Decide whether $T S \models P_{\text {safe }}$ using the $T S \otimes \mathcal{A}$ construction.

Provide a counterexample if $T S \not \vDash P_{\text {safe }}$.

Exercise 4.7. Prove or disprove the following equivalences for $\omega$-regular expressions:
(a) $\left(\mathrm{E}_{1}+\mathrm{E}_{2}\right) \cdot \mathrm{F}^{\omega} \equiv \mathrm{E}_{1} \cdot \mathrm{~F}^{\omega}+\mathrm{E}_{2} \cdot \mathrm{~F}^{\omega}$
(b) $\mathrm{E} \cdot\left(\mathrm{F}_{1}+\mathrm{F}_{2}\right)^{\omega} \equiv \mathrm{E} \cdot \mathrm{F}_{1}^{\omega}+\mathrm{E} \cdot \mathrm{F}_{2}^{\omega}$
(c) E. $\left(\text { F.F } F^{*}\right)^{\omega} \equiv$ E.F. ${ }^{\omega}$
(d) $\left(E^{*} \cdot F\right)^{\omega} \equiv E^{*} \cdot F^{\omega}$
where $E, E_{1}, E_{2}, F, F_{1}, F_{2}$ are arbitrary regular expressions with $\varepsilon \notin \mathcal{L}(F) \cup \mathcal{L}\left(F_{1}\right) \cup \mathcal{L}\left(F_{2}\right)$.

Exercise 4.8. Generalized $\omega$-regular expressions are built from the symbols $\underline{\varnothing}$ (to denote the empty language), $\underline{\varepsilon}$ (to denote the language $\{\varepsilon\}$ consisting of the empty word), the symbols $\underline{A}$ for $A \in \Sigma$ (for the singleton sets $\{A\}$ ) and the language operators " + " (union), "." (concatenation), " $*$ " (Kleene star, finite repetition), and " $\omega$ "(infinite repetition). The semantics of a generalized $\omega$-regular expression G is a language $\mathcal{L}_{g}(\mathrm{G}) \subseteq \Sigma^{*} \cup \Sigma^{\omega}$, which is defined by

- $\mathcal{L}_{g}(\underline{\varnothing})=\varnothing, \quad \mathcal{L}_{g}(\underline{\varepsilon})=\{\varepsilon\}, \quad \mathcal{L}_{g}(\underline{A})=\{A\}$,
- $\mathcal{L}_{g}\left(\mathrm{G}_{1}+\mathrm{G}_{2}\right)=\mathcal{L}_{g}\left(\mathrm{G}_{1}\right) \cup \mathcal{L}_{g}\left(\mathrm{G}_{2}\right)$ and $\mathcal{L}_{g}\left(\mathrm{G}_{1} \cdot \mathrm{G}_{2}\right)=\mathcal{L}_{g}\left(\mathrm{G}_{1}\right) \cdot \mathcal{L}_{g}\left(\mathrm{G}_{2}\right)$,
- $\mathcal{L}_{g}\left(\mathrm{G}^{*}\right)=\mathcal{L}_{g}(\mathrm{G})^{*}$, and $\mathcal{L}_{g}\left(\mathrm{G}^{\omega}\right)=\mathcal{L}_{g}(\mathrm{G})^{\omega}$.

Two generalized $\omega$-regular expressions G and $\mathrm{G}^{\prime}$ are called equivalent iff $\mathcal{L}_{g}(\mathrm{G})=\mathcal{L}_{g}\left(\mathrm{G}^{\prime}\right)$.
Show that for each generalized $\omega$-regular expression $G$ there exists an equivalent generalized $\omega$ regular expression $\mathrm{G}^{\prime}$ of the form

$$
\mathrm{G}^{\prime}=\mathrm{E}+\mathrm{E}_{1} \cdot \mathrm{~F}_{1}^{\omega}+\ldots \mathrm{E}_{n} \cdot \mathrm{~F}_{n}^{\omega}
$$

where $\mathrm{E}, \mathrm{E}_{1}, \ldots, \mathrm{E}_{n}, \mathrm{~F}_{1}, \ldots, \mathrm{~F}_{n}$ are regular expressions and $\varepsilon \notin \mathcal{L}\left(\mathrm{F}_{i}\right), i=1, \ldots, n$.

Exercise 4.9. Let $\Sigma=\{A, B\}$. Construct an NBA $\mathcal{A}$ that accepts the set of infinite words $\sigma$ over $\Sigma$ such that $A$ occurs infinitely many times in $\sigma$ and between any two successive $A$ 's an odd number of $B$ 's occur.

Exercise 4.10. Let $\Sigma=\{A, B, C\}$ be an alphabet.
(a) Construct an NBA $\mathcal{A}$ that accepts exactly the infinite words $\sigma$ over $\Sigma$ such that $A$ occurs infinitely many times in $\sigma$ and between any two successive $A$ 's an odd number of $B$ 's or an odd number of C's occur. Moreover, between any two successive $A$ 's either only $B$ 's or only C's are allowed. That is, the accepted words should have the form

$$
{ }_{w} A v_{1} A v_{2} A v_{3} \ldots
$$

where $w \in\{B, C\}^{*}, v_{i} \in\left\{B^{2 k+1} \mid k \geqslant 0\right\} \cup\left\{C^{2 k+1} \mid k \geqslant 0\right\}$ for all $i>0$. Give also an $\omega$-regular expression for this language.
(b) Repeat the previous exercise such that any accepting word contains only finitely many C's.
(c) Change your automaton from part (a) such that between any two successive $A$ 's an odd number of symbols from the set $\{B, C\}$ may occur.
(d) Same exercise as in (c), except that now an odd number of $B$ 's and an odd number of $C$ 's must occur between any two successive $A$ symbols.

ExERCISE 4.11. Depict an NBA for the language described by the $\omega$-regular expression

$$
(A B+C)^{*}((A A+B) C)^{\omega}+\left(A^{*} C\right)^{\omega} .
$$

Exercise 4.12. Consider the following NBA $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ over the alphabet $\{A, B, C\}$ :

$$
\mathcal{A}_{2}:
$$



Find $\omega$-regular expressions for the languages accepted by $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$.

Exercise 4.13. Consider the NFA $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ :


Construct an NBA for the language $\mathcal{L}\left(\mathcal{A}_{1}\right) \cdot \mathcal{L}\left(\mathcal{A}_{2}\right)^{\omega}$.

Exercise 4.14. Let $A P=\{a, b\}$. Give an NBA for the LT property consisting of the infinite words $A_{0} A_{1} A_{2} \ldots\left(2^{A P}\right)^{\omega}$ such that

$$
\stackrel{\infty}{\exists} j \geqslant 0 .\left(a \in A_{j} \wedge b \in A_{j}\right) \quad \text { and } \quad \exists j \geqslant 0 .\left(a \in A_{j} \wedge b \notin A_{j}\right) .
$$

Provide an $\omega$-regular expression for $\mathcal{L}_{\omega}(\mathcal{A})$.

ExERCISE 4.15. Let $A P=\{a, b, c\}$. Depict an NBA for the LT property consisting of the infinite words $A_{0} A_{1} A_{2} \ldots\left(2^{A P}\right)^{\omega}$ such that

$$
\forall j \geqslant 0 . A_{2 j} \models(a \vee(b \wedge c))
$$

Recall that $A \models(a \vee(b \wedge c))$ means $a \in A$ or $\{b, c\} \subseteq A$, i.e., $A \in\{\{a\},\{b, c\},\{a, b, c\}\}$.

Exercise 4.16. Consider NBA $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ depicted in Figure 4.26. Show that the powerset construction applied to $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ (viewed as NFA) yields the same deterministic automaton, while $\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right) \neq \mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)$. (This exercise is taken from [408].)


Figure 4.26: NBA $\mathcal{A}_{1}$ (a) and $\mathcal{A}_{2}(\mathrm{~b})$.

Exercise 4.17. Consider the following NBA $\mathcal{A}$ with the alphabet $\Sigma=2^{A P}$ where $A P=$ $\left\{a_{1}, \ldots, a_{n}\right\}$ for $n>0$.

(a) Determine the accepted language $\mathcal{L}_{\omega}(\mathcal{A})$.
(b) Show that there is no NBA $\mathcal{A}^{\prime}$ with $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}\left(\mathcal{A}^{\prime}\right)$ and less than $n$ states.
(This exercise is inspired by [149].)

Exercise 4.18. Provide an example for a regular safety property $P_{\text {safe }}$ over $A P$ and an NFA $\mathcal{A}$ for its minimal bad prefixes such that

$$
\mathcal{L}_{\omega}(\mathcal{A}) \neq\left(2^{A P}\right)^{\omega} \backslash P_{\text {safe }}
$$

when $\mathcal{A}$ is viewed as an NBA.

Exercise 4.19. Provide an example for a liveness property that is not $\omega$-regular. Justify your answer.

EXERCISE 4.20. Is there a DBA that accepts the language described by the $\omega$-regular expression $(A+B)^{*}(A B+B A)^{\omega}$ ? Justify your answer.

Exercise 4.21. Provide an example for an $\omega$-regular language $\mathcal{L}=\mathcal{L}_{k}$ that is recognizable for a DBA such that the following two conditions are satisfied:
(a) There exists an NBA $\mathcal{A}$ with $|\mathcal{A}|=\mathcal{O}(k)$ and $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}$.
(b) Each DBA $\mathcal{A}^{\prime}$ for $\mathcal{L}$ is of the size $\left|\mathcal{A}^{\prime}\right|=\Omega\left(2^{k}\right)$.

Hint: There is a simple answer to this question that uses the result that the regular language for the expression $(A+B)^{*} B(A+B)^{k}$ is recognizable by an NFA of size $\mathcal{O}(k)$, while any DFA has $\Omega\left(2^{k}\right)$ states.

EXERCISE 4.22. Show that the class of languages that are accepted by DBAs is not closed under complementation.

Exercise 4.23. Show that the class of languages that are accepted by DBAs is closed under union. To do so, prove the following stronger statement:

Let $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ be two DBAs both over the alphabet $\Sigma$. Show that there exists a DBA $\mathcal{A}$ with $|\mathcal{A}|=\mathcal{O}\left(\left|\mathcal{A}_{1}\right| \cdot\left|\mathcal{A}_{2}\right|\right)$ and $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right) \cup \mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)$.

Exercise 4.24.
Consider the GNBA outlined on the right with acceptance sets $F_{1}=\left\{q_{1}\right\}$ and $F_{2}=\left\{q_{2}\right\}$. Construct an equivalent NBA.


Exercise 4.25. Provide NBA $\mathcal{A}_{1}$ and $\mathcal{A}_{2}$ for the languages given by the expressions $(A C+B)^{*} B^{\omega}$ and $\left(B^{*} A C\right)^{\omega}$ and apply the product construction to obtain a GNBA $\mathcal{G}$ with $\mathcal{L}_{\omega}(\mathcal{G})=\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right) \cap$ $\mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)$. Justify that $\mathcal{L}_{\omega}(\mathcal{G})=\varnothing$.

Exercise 4.26. A nondeterministic Muller automaton is a quintuple $\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, \mathcal{F}\right)$ where $Q, \Sigma, \delta, Q_{0}$ are as for $N B A$ and $\mathcal{F} \subseteq 2^{Q}$. For an infinite run $\rho$ of $\mathcal{A}$, let $\lim (\rho):=$ $\{q \in Q \mid \stackrel{\infty}{\exists} i \geq 0 . \rho[i]=q\}$. Let $\alpha \in \Sigma^{\omega}$.
$\mathcal{A}$ accepts $\alpha \Longleftrightarrow$ ex. infinite run $\rho$ of $\mathcal{A}$ on $\alpha$ s.t. $\lim (\rho) \in \mathcal{F}$
(a) Consider the following Muller automaton $\mathcal{A}$ with $\mathcal{F}=\left\{\left\{q_{2}, q_{3}\right\},\left\{q_{1}, q_{3}\right\},\left\{q_{0}, q_{2}\right\}\right\}$ :


Define the language accepted by $\mathcal{A}$ by means of an $\omega$-regular expression.
(b) Show that every GNBA $\mathcal{G}$ can be transformed into a nondeterministic Muller automaton $\mathcal{A}$ such that $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}(\mathcal{G})$ by defining the corresponding transformation.

EXERCISE 4.27. Consider the transition systems $T S_{S e m}$ and $T S_{P e t}$ for mutual exclusion with a semaphore and the Peterson algorithm, respectively. Let $P_{\text {live }}$ be the following $\omega$-regular property over $A P=\left\{\right.$ wait $_{1}$, crit $\left._{1}\right\}:$
"whenever process 1 is in its waiting location then it will eventually enter its critical section"
(a) Depict an NBA for $P_{\text {live }}$ and an NBA $\overline{\mathcal{A}}$ for the complement property $\bar{P}_{\text {live }}=\left(2^{A P}\right)^{\omega} \backslash P_{\text {live }}$.
(b) Show that $T S_{S e m} \not \vDash P_{\text {live }}$ by applying the techniques explained in Section 4.4:
(i) Depict the reachable fragment of the product $T S_{S e m} \otimes \overline{\mathcal{A}}$
(ii) Sketch the main steps of the nested depth-first search applied to $T S_{S e m} \otimes \overline{\mathcal{A}}$ for the persistence property "eventually forever $\neg F$ " where $F$ is the acceptance set of $\overline{\mathcal{A}}$. Which counterexample is returned by Algorithm ?8
(c) Apply now the same techniques (product construction, nested DFS) to show that $T S_{P e t} \models$ $P_{\text {live }}$.

Exercise 4.28. The nested depth-first search approach can also be reformulated for an emptiness check for NBA. The path fragment returned by Algorithm 8 in case of a negative answer then yields a prefix of an accepting run.

Consider the automaton shown in Exercise 4.24 as an NBA, i.e., the acceptance set is $F=\left\{q_{1}, q_{2}\right\}$. Apply the nested depth-first search approach to verify that $\mathcal{L}_{\omega}(\mathcal{A}) \neq \varnothing$.

## Chapter 5

## Linear Temporal Logic

This chapter introduces (propositional) linear temporal logic (LTL), a logical formalism that is suited for specifying LT properties. The syntax and semantics of linear temporal logic are defined. Various examples are provided that show how linear temporal logic can be used to specify important system properties. The second part of the chapter is concerned with a model-checking algorithm-based on Büchi automata-for LTL. This algorithm can be used to answer the question: given a transition system TS and LTLformula $\varphi$, how to check whether $\varphi$ holds in $T S$ ?

### 5.1 Linear Temporal Logic

For reactive systems, correctness depends on the executions of the system - not only on the input and output of a computation - and on fairness issues. Temporal logic is a formalism par excellence for treating these aspects. Temporal logic extends propositional or predicate logic by modalities that permit to referral to the infinite behavior of a reactive system. They provide a very intuitive but mathematically precise notation for expressing properties about the relation between the state labels in executions, i.e., LT properties. Temporal logics and related modal logics have been studied in ancient times in different areas such as philosophy. Their application to verifying complex computer systems was proposed by Pnueli in the late seventies.

In this monograph, we will focus our attention on propositional temporal logics, i.e., extensions of propositional logic by temporal modalities. These logics should be distinguished from first- (or higher-) order temporal logics that impose temporal modalities on top of
predicate logic. Throughout this monograph we assume some familiarity with the basic principles of propositional logic. A brief introduction and summary of our notations can be found in Appendix A.3. The elementary temporal modalities that are present in most temporal logics include the operators:
$\diamond$ "eventually" (eventually in the future)
$\square$ "always" (now and forever in the future)
The underlying nature of time in temporal logics can be either linear or branching. In the linear view, at each moment in time there is a single successor moment, whereas in the branching view it has a branching, tree-like structure, where time may split into alternative courses. This chapter considers LTL (Linear Temporal Logic), a temporal logic that is based on a linear-time perspective. Chapter 6 introduces CTL (Computation Tree Logic), a logic that that is based on a branching-time view. Several model-checking tools use LTL (or a slight variant thereof) as a property specification language. The model checker SPIN is a prominent example of such an automated verification tool. One of the main advantages of LTL is that imposing fairness assumptions (such as strong and weak fairness) does not require the use of any new machinery: the typical fairness assumptions can all be specified in LTL. Verifying LTL-formulae under fairness constraints can be done using the algorithm for LTL. This does not apply to CTL.

Before introducing LTL in more detail, a short comment on the adjective "temporal" is in order to avoid any possible confusion. Although the term temporal suggests a relationship with the real-time behavior of a reactive system, this is only true in an abstract sense. A temporal logic allows for the specification of the relative order of events. Some examples are "the car stops once the driver pushes the brake", or "the message is received after it has been sent". It does however not support any means to refer to the precise timing of events. The fact that there is a minimal delay of at least $3 \mu s$ between braking and the actual halting of the car cannot be specified. In terms of transition systems, neither the duration of taking a transition nor state residence times can be specified using the elementary modalities of temporal logics. Instead, these modalities do allow for specifying the order in which state labels occur during an execution, or to assess that certain state labels occur infinitely often in a (or all) system execution. One might thus say that the modalities in temporal logic are time-abstract.

As will be discussed in this chapter, LTL may be used to express the timing for the class of synchronous systems in which all components proceed in a lock-step fashion. In this setting, a transition corresponds to the advance of a single time-unit. The underlying time domain is thus discrete, i.e., the present moment refers to the current state and the next moment corresponds to the immediate successor state. Stated differently, the system behavior is assumed to be observable at the time points $0,1,2, \ldots$. The treatment
of real-time constraints in asynchronous systems by means of a continuous-time domain will be discussed in Chapter 9 where a timed version of CTL, called Timed CTL, will be introduced. Table 5.1 summarizes the distinguishing features of the main temporal logics considered in this monograph.

| logic | linear-time <br> (path-based) | branching-time <br> (state-based) | real-time requirements <br> (continuous-time domain) |
| :--- | :---: | :---: | :---: |
| LTL | $\sqrt{ }$ |  |  |
| CTL |  | $\sqrt{ }$ |  |
| Timed CTL |  | $\sqrt{ }$ | $\sqrt{ }$ |

Table 5.1: Classification of the temporal logics in this monograph.

### 5.1.1 Syntax

This subsection describes the syntactic rules according to which formulae in LTL can be constructed. The basic ingredients of LTL-formulae are atomic propositions (state labels $a \in A P$ ), the Boolean connectors like conjunction $\wedge$, and negation $\neg$, and two basic temporal modalities $\bigcirc$ (pronounced "next") and U (pronounced "until"). The atomic proposition $a \in A P$ stands for the state label $a$ in a transition system. Typically, the atoms are assertions about the values of control variables (e.g., locations in program graphs) or the values of program variables such as " $x>5$ " or " $x \leqslant y$ ". The $\bigcirc$-modality is a unary prefix operator and requires a single LTL formula as argument. Formula $\bigcirc \varphi$ holds at the current moment, if $\varphi$ holds in the next "step". The U-modality is a binary infix operator and requires two LTL formulae as argument. Formula $\varphi_{1} \cup \varphi_{2}$ holds at the current moment, if there is some future moment for which $\varphi_{2}$ holds and $\varphi_{1}$ holds at all moments until that future moment.

## Definition 5.1. Syntax of LTL

LTL formulae over the set AP of atomic proposition are formed according to the following grammar: ${ }^{1}$

$$
\varphi::=\operatorname{true}|a| \varphi_{1} \wedge \varphi_{2}|\neg \varphi| \bigcirc \varphi \mid \varphi_{1} \cup \varphi_{2}
$$

where $a \in A P$.

[^22]We mostly abstain from explicitly indicating the set $A P$ of propositions as this follows either from the context or can be defined as the set of atomic propositions occurring in the LTL formula at hand.

The precedence order on the operators is as follows. The unary operators bind stronger than the binary ones. $\neg$ and $\bigcirc$ bind equally strong. The temporal operator $U$ takes precedence over $\wedge, \vee$, and $\rightarrow$. Parentheses are omitted whenever appropriate, e.g., we write $\neg \varphi_{1} \mathrm{U} \bigcirc \varphi_{2}$ instead of $\left(\neg \varphi_{1}\right) \mathrm{U}\left(\bigcirc \varphi_{2}\right)$. Operator U is right-associative, e.g., $\varphi_{1} \mathrm{U} \varphi_{2} \mathrm{U} \varphi_{3}$ stands for $\varphi_{1} \mathrm{U}\left(\varphi_{2} \mathrm{U} \varphi_{3}\right)$.

Using the Boolean connectives $\wedge$ and $\neg$, the full power of propositional logic is obtained. Other Boolean connectives such as disjunction $\vee$, implication $\rightarrow$, equivalence $\leftrightarrow$, and the parity (or: exclusive or) operator $\oplus$ can be derived as follows:

$$
\begin{aligned}
\varphi_{1} \vee \varphi_{2} & \stackrel{\text { def }}{=} \neg\left(\neg \varphi_{1} \wedge \neg \varphi_{2}\right) \\
\varphi_{1} \rightarrow \varphi_{2} & \stackrel{\text { def }}{=} \neg \varphi_{1} \vee \varphi_{2} \\
\varphi_{1} \leftrightarrow \varphi_{2} & \stackrel{\text { def }}{=}\left(\varphi_{1} \rightarrow \varphi_{2}\right) \wedge\left(\varphi_{2} \rightarrow \varphi_{1}\right) \\
\varphi_{1} \oplus \varphi_{2} & \stackrel{\text { def }}{=}\left(\varphi_{1} \wedge \neg \varphi_{2}\right) \vee\left(\varphi_{2} \wedge \neg \varphi_{1}\right) \\
& \vdots
\end{aligned}
$$

The until operator allows to derive the temporal modalities $\diamond$ ("eventually", sometimes in the future) and $\square$ ("always", from now on forever) as follows:

$$
\diamond \varphi \stackrel{\text { def }}{=} \operatorname{true} \cup \varphi \quad \square \varphi \stackrel{\text { def }}{=} \neg \diamond \neg \varphi
$$

As a result, the following intuitive meaning of $\diamond$ and $\square$ is obtained. $\Delta \varphi$ ensures that $\varphi$ will be true eventually in the future. $\square \varphi$ is satisfied if and only if it is not the case that eventually $\neg \varphi$ holds. This is equivalent to the fact that $\varphi$ holds from now on forever.

Figure 5.1 sketches the intuitive meaning of temporal modalities for the simple case in which the arguments of the modalities are just atomic propositions from $\{a, b\}$. On the left-hand side, some LTL formulae are indicated, whereas on the right hand side sequences of states (i.e., paths) are depicted.

By combining the temporal modalities $\diamond$ and $\square$, new temporal modalities are obtained. For instance, $\square \diamond a$ ("always eventually $a$ ") describes the (path) property stating that at any moment $j$ there is a moment $i \geqslant j$ at which an $a$-state is visited. This thus amounts to assert that an $a$-state is visited infinitely often. The dual modality $\diamond \square a$ expresses that from some moment $j$ on, only $a$-states are visited. So:

$$
\begin{array}{ll}
\square \diamond \varphi \quad \text { "infinitely often } \varphi \text { " } \\
\diamond \square \varphi \quad \text { "eventually forever } \varphi \text { " }
\end{array}
$$



Figure 5.1: Intuitive semantics of temporal modalities.

Before proceeding with the formal semantics of LTL, we present some examples.

## Example 5.2. Properties for the Mutual Exclusion Problem

Consider the mutual exclusion problem for two concurrent processes $P_{1}$ and $P_{2}$, say. Process $P_{i}$ is modeled by three locations: (1) the noncritical section, (2) the waiting phase which is entered when the process intends to enter the critical section, and (3) the critical section. Let the propositions wait $i_{i}$ and crit ${ }_{i}$ denote that process $P_{i}$ is in its waiting phase and critical section, respectively.

The safety property stating that $P_{1}$ and $P_{2}$ never simultaneously have access to their critical sections can be described by the LTL-formula:

$$
\square\left(\neg c^{\prime} i t_{1} \vee \neg c_{i} t_{2}\right) .
$$

This formula expresses that always ( $\square$ ) at least one of the two processes is not in its critical section ( $\neg$ crit $i_{i}$ ).

The liveness requirement stating that each process $P_{i}$ is infinitely often in its critical
section is described by the LTL formula:

$$
\left(\square \diamond c_{i} t_{1}\right) \wedge\left(\square \diamond c_{1} t_{2}\right)
$$

The weakened form that every waiting process will eventually enter its critical section (i.e., starvation freedom) can-by using the additional proposition wait ${ }_{i}$-be formulated as follows:

$$
\left(\square \diamond \text { wait }_{1} \rightarrow \square \diamond \text { crit }_{1}\right) \wedge\left(\square \diamond \text { wait }_{2} \rightarrow \square \diamond \text { crit }_{2}\right) .
$$

These formulae only refer to the locations (i.e., values of the program counters) by the atomic propositions wait $t_{i}$ and crit ${ }_{i}$. Propositions can however also refer to program variables. For instance, for the solution of the mutual exclusion problem using a binary semaphore $y$, the formula:

$$
\square\left((y=0) \rightarrow \operatorname{crit}_{1} \vee \text { crit }_{2}\right)
$$

states that whenever the semaphore $y$ has the value 0 , one of the processes is in its critical section.

Example 5.3. Properties for the dining philosophers
For the dining philosophers (see Example 3.2 on page 90) deadlock freedom can be described by the LTL formula

$$
\square \neg\left(\bigwedge_{0 \leqslant i<n} \text { wait }_{i} \wedge \bigwedge_{0 \leqslant i<n} \text { occupied }_{i}\right) .
$$

We assume here that there are $n$ philosophers and chop sticks, indexed from 0 to $n-1$. The atom wait ${ }_{i}$ means that philosopher $i$ waits for one of the sticks on his left or right, but keeps the other one in his hand. Similarly, occupied ${ }_{i}$ indicates that stick $i$ is in use.

## Example 5.4. Properties for a Traffic Light

For a traffic light with the phases "green", "red" and "yellow", the liveness property $\square \diamond$ green expresses that the traffic light is infinitely often green. A specification of the traffic light cycles and their chronological order can be provided by means of a conjunction of LTL-formulae stating the predecessor phase of any phase. For instance, the requirement "once red, the light cannot become green immediately" can be expressed by the LTL formula

$$
\square(\text { red } \rightarrow \neg \bigcirc \text { green }) .
$$

The requirement "once red, the light always becomes green eventually after being yellow for some time" is expressed by

$$
\square(\text { red } \rightarrow \bigcirc(\text { red } \cup(\text { yellow } \wedge \bigcirc(\text { yellow } \cup \text { green }))))
$$

A progress property like "every request will eventually lead to a response" can be described by the following formula of the type

$$
\square(\text { request } \rightarrow \diamond \text { response }) .
$$

## Remark 5.5. Length of a Formula

Let $|\varphi|$ denote the length of LTL formula $\varphi$ in terms of the number of operators in $\varphi$. This can easily be defined by induction on the structure of $\varphi$. For instance, the length of the formula true and $a \in A P$ is 0 . Formulae $\bigcirc a \vee b$ and $a \vee \neg b$ have length 2, and $(\bigcirc a) \cup(a \wedge \neg b)$ has length 4. Throughout this monograph, mostly the asymptotic size $\Theta(|\varphi|)$ is needed. For this purpose, it is irrelevant whether or not the derived Boolean operators $\vee, \rightarrow$, and so on, and the derived temporal modalities $\diamond$ and $\square$ are taken into account in determining the length.

### 5.1.2 Semantics

LTL formulae stand for properties of paths (or in fact their trace). This means that a path can either fulfill an LTL-formula or not. To precisely formulate when a path satisfies an LTL formula, we proceed as follows. First, the semantics of LTL formula $\varphi$ is defined as a language $\operatorname{Words}(\varphi)$ that contains all infinite words over the alphabet $2^{A P}$ that satisfy $\varphi$. That is, to every LTL formula a single LT property is associated. Then, the semantics is extended to an interpretation over paths and states of a transition system.

## Definition 5.6. Semantics of LTL (Interpretation over Words)

Let $\varphi$ be an LTL formula over AP. The LT property induced by $\varphi$ is

$$
\operatorname{Words}(\varphi)=\left\{\sigma \in\left(2^{A P}\right)^{\omega} \mid \sigma \models \varphi\right\}
$$

where the satisfaction relation $\models \subseteq\left(2^{A P}\right)^{\omega} \times$ LTL is the smallest relation with the properties in Figure 5.2.

Here, for $\sigma=A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}, \sigma[j \ldots]=A_{j} A_{j+1} A_{j+2} \ldots$ is the suffix of $\sigma$ starting in the $(j+1)$ st symbol $A_{j}$.

Note that in the definition of the semantics of LTL-formulae the word fragment $\sigma[j \ldots]$ cannot be replaced with $A_{j}$. For the formula $\bigcirc(a \cup b)$, e.g., the suffix $A_{1} A_{2} A_{3} \ldots$ has to

$$
\begin{array}{rlll}
\sigma & \models \text { true } \\
\sigma & \models a & \text { iff } & \left.a \in A_{0} \quad \text { (i.e., } A_{0} \models a\right) \\
\sigma & \models \varphi_{1} \wedge \varphi_{2} & \text { iff } & \sigma \models \varphi_{1} \text { and } \sigma \models \varphi_{2} \\
\sigma & \models \neg \varphi & \text { iff } & \sigma \not \models \varphi \\
\sigma & \models \bigcirc \varphi & \text { iff } & \sigma[1 \ldots]=A_{1} A_{2} A_{3} \ldots \models \varphi \\
\sigma & \models \varphi_{1} \cup \varphi_{2} & \text { iff } \quad \exists j \geqslant 0 . \sigma[j \ldots] \models \varphi_{2} \text { and } \sigma[i \ldots] \models \varphi_{1}, \text { for all } 0 \leqslant i<j
\end{array}
$$

Figure 5.2: LTL semantics (satisfaction relation $\models$ ) for infinite words over $2^{A P}$.
be regarded in order to be able to refer to the truth-value of the subformula $a \cup b$ in the "next step".

For the derived operators $\diamond$ and $\square$ the expected result is:

$$
\begin{aligned}
\sigma & \models \diamond \varphi \quad \text { iff } \quad \exists j \geqslant 0 . \sigma[j \ldots] \models \varphi \\
\sigma & \models \square \varphi \text { iff } \quad \forall j \geqslant 0 . \sigma[j \ldots] \models \varphi .
\end{aligned}
$$

The statement for $\diamond$ is immediate from the definition of $\diamond$ and the semantics of $U$. The statement for $\square$follows from:

$$
\begin{array}{lll}
\sigma \models \square \varphi=\neg \diamond \neg \varphi & \text { iff } \quad \neg \exists j \geqslant 0 . \sigma[j \ldots] \vDash \neg \varphi \\
& \text { iff } \quad \neg \exists j \geqslant 0 . \sigma[j \ldots] \nLeftarrow \varphi \\
& \text { iff } \quad \forall j \geqslant 0 . \sigma[j \ldots] \vDash \varphi .
\end{array}
$$

The semantics of the combinations of $\square$ and $\diamond$ can now be derived:

$$
\begin{aligned}
\sigma & \models \square \Delta \varphi \text { iff } \quad \stackrel{\infty}{\exists} . \sigma[j \ldots] \models \varphi \\
\sigma & \models \diamond \square \varphi \text { iff } \quad \forall j . \sigma[j \ldots] \models \varphi .
\end{aligned}
$$

Here, $\stackrel{\infty}{\exists} j$ means $\forall i \geqslant 0$. $\exists j \geqslant i$, "for infinitely many $j \in \mathbb{N}$ ", while $\forall j$ stands for $\exists i \geqslant 0 . \forall j \geqslant i$, "for almost all $j \in \mathbb{N}$ ". Let us verify the first statement. The argument for the second statement is similar.

$$
\begin{array}{rll}
\sigma \models \square \diamond \varphi & \text { iff } \quad \forall i \geqslant 0 . \sigma[i \ldots] \models \diamond \varphi \\
& \text { iff } \quad \forall i \geqslant 0 . \exists j \geqslant i . \sigma[j \ldots] \models \varphi \\
& \text { iff } \quad \exists j . \sigma[j \ldots] \models \varphi .
\end{array}
$$

As a subsequent step, we determine the semantics of LTL-formulae with respect to a
transition system. According to the satisfaction relation for LT properties (see Definition 3.11 on page 100), the LTL formula $\varphi$ holds in state $s$ if all paths starting in $s$ satisfy $\varphi$. The transition system $T S$ satisfies $\varphi$ if $T S$ satisfies the LT property $\operatorname{Words}(\varphi)$, i.e., if all initial paths of $T S$-paths starting in an initial state $s_{0} \in I$-satisfy $\varphi$.

Recall that we may assume without loss of generality that transition system $T S$ has no terminal states (if it has such states, a trap state can be introduced. Thus, we may assume that all paths and traces are infinite. This assumption is made for the sake of simplicity; it is also possible to define the semantics of LTL for finite paths. Note that for the semantics it is irrelevant whether or not $T S$ is finite. Only for the model-checking algorithm later on in this chapter, is the finiteness of $T S$ required.

As for the LT properties, when defining $T S \models \varphi$ for transition system $T S$ over $A P^{\prime}$, it is assumed that $\varphi$ is an LTL-formula with atomic propositions in $A P=A P^{\prime}$. (Here, one could be more liberal and allow for $A P \subseteq A P^{\prime}$.)

## Definition 5.7. Semantics of LTL over Paths and States

Let $T S=(S$, Act $, \rightarrow, I, A P, L)$ be a transition system without terminal states, and let $\varphi$ be an LTL-formula over $A P$.

- For infinite path fragment $\pi$ of $T S$, the satisfaction relation is defined by

$$
\pi \models \varphi \quad \text { iff } \quad \operatorname{trace}(\pi) \models \varphi
$$

- For state $s \in S$, the satisfaction relation $\models$ is defined by

$$
s \models \varphi \quad \text { iff } \quad(\forall \pi \in \operatorname{Paths}(s) . \pi \models \varphi)
$$

- TS satisfies $\varphi$, denoted $T S \models \varphi$, if $\operatorname{Traces}(T S) \subseteq \operatorname{Words}(\varphi)$.

From this definition, it immediately follows that

```
    \(T S \models \varphi\)
iff
    \(\operatorname{Traces}(T S) \subseteq \operatorname{Words}(\varphi)\)
iff (* Definition of \(\models\) for LT properties *)
    \(T S ~ \models \operatorname{Words}(\varphi)\)
iff
    iff \(\pi \models \varphi\) for all \(\pi \in \operatorname{Paths}(T S)\)
iff
    \(s_{0} \models \varphi\) for all \(s_{0} \in I\).
        (* Definition \(5.7^{*}\) )
                                    (* Definition of \(\left.\operatorname{Words}(\varphi)^{*}\right)\)
                                    (* Definition 5.7 of \(\models\) for states *)
```

Thus, $T S \models \varphi$ if and only if $s_{0} \models \varphi$ for all initial states $s_{0}$ of $T S$.


Figure 5.3: Example for semantics of LTL.

## Example 5.8. Semantics of LTL

Consider the transition system TS depicted in Figure 5.3 with the set of propositions $A P=\{a, b\}$. For example, we have that $T S \models \square a$, since all states are labeled with $a$, and hence, all traces of $T S$ are words of the form $A_{0} A_{1} A_{2} \ldots$ with $a \in A_{i}$ for all $i \geqslant 0$. Thus, $s_{i} \models \square a$ for $i=1,2,3$. Moreover:

$$
\begin{gathered}
s_{1} \models \bigcirc(a \wedge b) \text { since } s_{2} \models a \wedge b \text { and } s_{2} \text { is the only successor of } s_{1} \\
s_{2} \not \models \bigcirc(a \wedge b) \text { and } s_{3} \not \models \bigcirc(a \wedge b) \text { as } s_{3} \in \operatorname{Post}\left(s_{2}\right), s_{3} \in \operatorname{Post}\left(s_{3}\right) \text { and } s_{3} \not \models a \wedge b
\end{gathered}
$$

This yields $T S \not \models \bigcirc(a \wedge b)$ as $s_{3}$ is an initial state for which $s_{3} \not \models \bigcirc(a \wedge b)$. As another example:

$$
T S \models \square(\neg b \rightarrow \square(a \wedge \neg b)),
$$

since $s_{3}$ is the only $\neg b$ state, $s_{3}$ cannot be left anymore, and $a \wedge \neg b$ in $s_{3}$ is true. However,

$$
T S \not \vDash b \cup(a \wedge \neg b),
$$

since the initial path $\left(s_{1} s_{2}\right)^{\omega}$ does not visit a state for which $a \wedge \neg b$ holds. Note that the initial path $\left(s_{1} s_{2}\right)^{*} s_{3}^{\omega}$ satisfies $b \cup(a \wedge \neg b)$.

## Remark 5.9. Semantics of Negation

For paths, it holds $\pi \models \varphi$ if and only if $\pi \not \vDash \neg \varphi$. This is due to the fact that

$$
\operatorname{Words}(\neg \varphi)=\left(2^{A P}\right)^{\omega} \backslash \operatorname{Words}(\varphi)
$$

However, the statements $T S \not \vDash \varphi$ and $T S \models \neg \varphi$ are not equivalent in general. Instead, we have $T S \models \neg \varphi$ implies $T S \not \models \varphi$. Note that

$$
\begin{aligned}
T S \notin \varphi & \text { iff } \operatorname{Traces}(T S) \nsubseteq \operatorname{Words}(\varphi) \\
& \text { iff } \operatorname{Traces}(T S) \backslash \operatorname{Words}(\varphi) \neq \varnothing \\
& \text { iff } \operatorname{Traces}(T S) \cap \operatorname{Words}(\neg \varphi) \neq \varnothing
\end{aligned}
$$

Thus, it is possible that a transition system (or a state) satisfies neither $\varphi$ nor $\neg \varphi$. This is caused by the fact that there might be paths $\pi_{1}$ and $\pi_{2}$ in $T S$ such that $\pi_{1} \models \varphi$ and $\pi_{2} \models \neg \varphi$ (and therefore $\pi_{2} \not \vDash \varphi$ ). In this case, $T S \not \vDash \varphi$ and $T S \not \vDash \neg \varphi$ holds.

To illustrate this effect, consider the transition system depicted in Figure 5.4. Let $A P=$ $\{a\}$. It follows that $T S \not \vDash \diamond a$, since the initial path $s_{0}\left(s_{2}\right)^{\omega} \not \vDash \diamond a$. On the other hand, $T S \not \vDash \neg \diamond a$ also holds, since the initial path $s_{0}\left(s_{1}\right)^{\omega} \models \diamond a$, and thus, $s_{0}\left(s_{1}\right)^{\omega} \not \vDash \neg \diamond a$.


Figure 5.4: A transition system for which $T S \not \vDash \diamond a$ and $T S \not \vDash \neg \diamond a$.

### 5.1.3 Specifying Properties



Figure 5.5: Transition system of semaphore-based mutual exclusion algorithm.

## Example 5.10. Semaphore-Based Mutual Exclusion Revisited

Consider the transition system $T S_{S e m}$ depicted in Figure 5.5 which represents a semaphorebased solution to the mutual exclusion problem; see also Example 3.9 on page 98. Each
state of the form $\left\langle c_{1}, \cdot, \cdot\right\rangle$ is labeled with proposition $c r i t_{1}$ and each state of the form $\left\langle\cdot, c_{2}, \cdot\right\rangle$ is labeled with crit ${ }_{2}$. It follows that

$$
T S_{S e m} \vDash \square\left(\neg \text { crit }_{1} \vee \neg \text { crit }_{2}\right) \quad \text { and } \quad T S_{S e m} \vDash \square \diamond \text { crit }_{1} \vee \square \diamond \text { crit }_{2}
$$

where the first LTL-formula stands for the mutual exclusion property and the second LTL-formula for the fact that at least one of the two processes enters its critical section infinitely often. However,

$$
T S_{S e m} \not \vDash \square \diamond \text { crit }_{1} \wedge \square \diamond \text { crit }_{2}
$$

since - in the absence of any fairness assumption-it is not ensured that process $P_{1}$ is enabled infinitely often. It may not be able to acquire access to its critical section once. (A similar argument applies to process $P_{2}$.) The same argument applies to show that

$$
T S_{S e m} \not \vDash \square \diamond \text { wait }_{1} \rightarrow \square \diamond \text { crit }_{1}
$$

as in principle process $P_{1}$ may not get its turn once it starts to wait.

## Example 5.11. Modulo 4 Counter

A modulo 4 counter can be represented by a sequential circuit $C$, which outputs 1 in every fourth cycle, otherwise $0 . C$ has no input bits, one output bit $y$, and two registers $r_{1}$ and $r_{2}$. The register evaluation $\left[r_{1}=c_{1}, r_{2}=c_{2}\right.$ ] can be identified with the number $i=2 \cdot r_{1}+r_{2}$. In every cycle, the value of $i$ is increased by 1 (modulo 4 ). We construct $C$ in a way such that the output bit $y$ is set exactly for $i=0$ (hence, $r_{1}=r_{2}=0$ ). The transition relation and output function are given by

$$
\delta_{r_{1}}=r_{1} \oplus r_{2}, \quad \delta_{r_{2}}=\neg r_{1}, \quad \lambda_{y}=\neg r_{1} \wedge \neg r_{2} .
$$

Figure 5.6 illustrates the diagram (on the left) and the transition system $T S_{C}$ (on the right). Let $A P=\left\{r_{1}, r_{2}, y\right\}$. The following statement can be directly inferred from $T S_{C}$ :

$$
\begin{aligned}
& T S_{C} \models \square\left(y \leftrightarrow \neg r_{1} \wedge \neg r_{2}\right) \\
& T S_{C} \vDash \square\left(r_{1} \rightarrow(\bigcirc y \vee \bigcirc \bigcirc y)\right) \\
& T S_{C} \vDash \square(y \rightarrow(\bigcirc \neg y \wedge \bigcirc \bigcirc \neg y))
\end{aligned}
$$

If it is assumed that only the output variable $y$ (and not the register evaluations) can be perceived by an observer, then an appropriate choice for $A P$ is $A P=\{y\}$. The property that at least during every four cycles the output 1 is obtained holds for $T S_{C}$, i.e., we have

$$
T S_{C} \models \square(y \vee \bigcirc y \vee \bigcirc \bigcirc y \vee \bigcirc \bigcirc \bigcirc y)
$$



Figure 5.6: A modulo 4 counter.

The fact that these outputs are produced in a periodic manner where every fourth cycle yields the output 1 is expressed as

$$
T S_{C} \vDash \square(y \longrightarrow(\bigcirc \neg y \wedge \bigcirc \bigcirc \neg y \wedge \bigcirc \bigcirc \bigcirc \neg y))
$$

## Example 5.12. A Communication Channel

Consider an unidirectional channel between two communicating processes, a sender $S$ and a receiver $R$. Sender $S$ is equipped with an output buffer $S$.out and recipient $R$ with an input buffer $R$.in. If sender $S$ sends a message $m$ to $R$ it inserts the message into its output buffer $S$.out. The output buffer $S$.out and the input buffer $R$.in are connected via an unidirectional channel. The receiver $R$ receives messages by deleting messages from its input buffer $R$.in. The capacity of the buffers is not of importance here.

A schematic view of the system under consideration is:


In the following LTL-specifications, we use the atoms " $m \in S . o u t$ " and " $m \in R$. in" where $m$ is an arbitrary message. We formalize the following informal requirements by LTL formulae:

- "Whenever message $m$ is in the out-buffer of $S$, then $m$ will eventually be consumed by the receiver."

$$
\square(m \in S . \text { out } \longrightarrow \diamond(m \in R . \text { in }))
$$

The above property is still satisfied for paths $s_{1} s_{2} s_{3} \ldots$ where $s_{1} \models m \in S$.out, $s_{2} \models m \notin S$.out, $s_{2} \models m \notin R$.in, and $s_{3} \models m \in R$.in. However, such paths stand for a mysterious behavior where message $m$ in the output buffer for $S$ (state $s_{1}$ ) gets lost (state $s_{2}$ ), but still arrives in the input buffer of $R$ (state $s_{3}$ ). In fact, such a behavior is impossible for a reliable FIFO channel which satisfies the following stronger condition

$$
\square(m \in S . \text { out } \longrightarrow(m \in S . \text { out } \cup m \in R . \text { in }))
$$

stating that message $m$ stays in $S$.out until the receiver $R$ consumes $m$. Since writing and reading in a FIFO channel cannot happen at the same moment, we can even use the formula

$$
\square(m \in S . \text { out } \longrightarrow \bigcirc(m \in S . \text { out } \cup m \in R . \text { in }))
$$

- If we assume that no message occurs twice in $S$.out then the asynchronous behavior of a FIFO channel ensures that the property "message $m$ cannot be in both buffers at the same time". This is formalized by the LTL formula:

$$
\square \neg(m \in S . \text { out } \wedge m \in R . \mathrm{in})
$$

- The characteristic of FIFO-channels is that they are "order-preserving" according to the "first in, first out principle" stating that if message $m$ is offered first by $S$ to its output buffer $S$.out and subsequently $m^{\prime}$, then $m$ will be received by $R$ before $m^{\prime}$ :

$$
\begin{aligned}
& \square\left(m \in S . \text { out } \wedge \neg m^{\prime} \in \text { S.out } \wedge \diamond\left(m^{\prime} \in \text { S.out }\right)\right. \\
& \left.\longrightarrow \diamond\left(m \in R . \text { in } \wedge \neg m^{\prime} \in R . \text { in } \wedge \diamond\left(m^{\prime} \in \text { R.in }\right)\right)\right)
\end{aligned}
$$

Note that in the premise the conjunct $\neg m^{\prime} \in S$.out is needed in order to specify that $m^{\prime}$ is put in S.out after $m . \diamond\left(m^{\prime} \in S\right.$.out $)$ on its own does not exclude that $m^{\prime}$ is already in the sender's buffer when message $m$ is in S.out.

The above formulae refer to fixed messages $m$ and $m^{\prime}$. In order to state the above properties for all messages, we have to take the conjunction over all messages $m, m^{\prime}$. As long as the message alphabet is finite we still obtain an LTL formula.

## Example 5.13. Dynamic Leader Election

(This example has been taken from [69].) In current distributed systems several services are offered by some dedicated process(es) in the system. Consider, for example, address assignment and registration, query coordination in a distributed database system,
clock distribution, token regeneration after token loss in a token ring network, initiation of topology updates in a mobile network, load balancing, and so forth. Usually many processes in the system are potentially capable of providing these services. However, for consistency reasons it is usually the case that at any time only one process is allowed to actually provide a given service. This process - called the "leader" - is in fact elected. Sometimes it suffices to elect an arbitrary process, but for other services it is important to elect the process with the best capabilities for performing that service. Here we abstract from specific capabilities and use ranking on the basis of process identities. The idea is therefore that the higher the process' identity, the better its capabilities.

Assume we have a finite number $N>0$ of processes connected via some communication means. The communication between processes is asynchronous, as in the previous example. Pictorially,


Each process has a unique identity, and it is assumed that a total ordering exists on these identities. Processes behave dynamically in the sense that they are initially inactive, i.e., not participating in the election, and may become active, i.e., participating in the election, at arbitrary moments. In order to have some progress we assume that a process cannot be inactive indefinitely; that is, each process becomes active at some time. (This corresponds to a fairness condition.) Once a process participates it continues to do so, i.e., it does not become inactive anymore. For a given set of active processes a leader will be elected; if an inactive process becomes active, a new election takes place if this process has a higher identity than the current leader.

To give an idea of using LTL as specification formalism we formulate several properties by LTL formulae. We will use $i, j$ as process identities. Let the set of atomic propositions be $\left\{\right.$ leader $_{i}$, active $\left.\mid 1 \leqslant i, j \leqslant N\right\}$, where leader ${ }_{i}$ means that process $i$ is a leader, active ${ }_{i}$ means that process $i$ is active. An inactive process cannot be a leader.

- The property "There is always one leader" can be formalized by

$$
\square\left(\bigvee_{1 \leqslant i \leqslant N} \text { leader }_{i} \wedge \bigwedge_{\substack{1 \leqslant j \leqslant N \\ j \neq i}} \neg \text { leader }_{j}\right)
$$

Although this formula expresses the informally stated property, it will not be satisfied by any realistic protocol. One reason is that processes may be initially inactive, and thus no leader is guaranteed to exist initially. Besides, in a distributed system with asynchronous communication, switching from one leader to another can hardly be made atomic. So, it is more realistic to allow the temporary absence of a leader. As a first attempt to do so, one could modify the above formula into

$$
\varphi=\square \diamond\left(\bigvee_{1 \leqslant i \leqslant N} \operatorname{leader}_{i} \wedge \bigwedge_{\substack{1 \leqslant j \leqslant N \\ j \neq i}} \neg \text { leader }_{j}\right)
$$

Problematic, though, is that this allows there to be more than one leader at a time temporarily - it is only stated that infinitely often there should be exactly one leader, but no statement is made about the moments at which this is not the case. For consistency reasons this is not desired. We therefore replace the above formula $\varphi$ with $\varphi_{1} \wedge \varphi_{2}$ where $\varphi_{1}$ and $\varphi_{2}$ correspond to the following two properties.

- "There must always be at most one leader":

$$
\varphi_{1}=\square \bigwedge_{1 \leqslant i \leqslant N}\left(\text { leader }_{i} \rightarrow \bigwedge_{\substack{1 \leqslant j \leqslant N \\ j \neq i}} \neg \text { leader }_{j}\right)
$$

- "There will be enough leaders in due time":

$$
\varphi_{2}=\square \diamond \bigvee_{1 \leqslant i \leqslant N} \text { leader }_{i}
$$

$\varphi_{2}$ does not imply that there will be infinitely many leaders. It only states that there are infinitely many states at which a leader exists. This requirement classifies a leader election protocol that never elects a leader to be wrong. In fact, such a protocol would fulfill the previous requirement, but is not desired for obvious reasons.

- "In the presence of an active process with a higher identity the leader will resign at some time":

$$
\square\left(\bigwedge_{\substack{1 \leqslant i, j \leqslant N \\ i<j}}\left(\left(\text { leader }_{i} \wedge \neg \text { leader }_{j} \wedge \text { active }_{j}\right) \rightarrow \diamond \neg \text { leader }_{i}\right)\right)
$$

For reasons of efficiency it is assumed not to be desirable that a leader eventually resigns in the presence of an inactive process that may participate at some unknown time in the future. Therefore we require $j$ to be an active process.

- "A new leader will be an improvement over the previous one". This property requires that successive leaders have an increasing identity. In particular, a process that resigns once will not become a leader anymore.

$$
\square\left(\bigwedge_{1 \leqslant i, j \leqslant N}\left(\text { leader }_{i} \wedge \neg \bigcirc \text { leader }_{i} \wedge \bigcirc \diamond \text { leader }_{j}\right) \rightarrow(i<j)\right)
$$

Here, we use " $i<j$ " as an atomic proposition that compares the identifiers of processes $P_{i}$ and $P_{j}$ and evaluates to true if and only if the process identifier of $P_{i}$ is smaller than that of $P_{j}$. Assuming that the identity of $P_{i}$ is $i($ for $i=1, \ldots, N)$, the above property can also be specified by the LTL formula

$$
\square \neg\left(\bigwedge_{\substack{1 \leqslant i, j \leqslant N \\ i \geqslant j}}\left(\text { leader }_{i} \wedge \neg \bigcirc \text { leader }_{i} \wedge \bigcirc \diamond \text { leader }_{j}\right)\right)
$$

## Example 5.14. Specifying the Input/Output Behavior of Sequential Programs

The typical requirements on sequential programs such as partial correctness and termination can "in principle" be represented in LTL. Let us briefly describe what termination and partial correctness mean. Assume that a sequential program Prog computes a function of the type $f: \operatorname{Inp} \rightarrow$ Outp, i.e., Prog takes as input a value $i \in \operatorname{Inp}$ and terminates either by reporting an output value $o \in$ Outp or does not terminate. Prog is called terminating if the computation of Prog halts for each input value $i \in \operatorname{Inp}$. Prog is partially correct if for any input value $i \in \operatorname{Inp}$, whenever Prog terminates then the output value $o$ equals $f(i)$. How can termination and partial correctness be expressed by means of LTL formulae?

Termination can be specified by a formula of the form init $\rightarrow \diamond$ halt where init is the labeling for the initial states and halt is an atomic proposition characterizing exactly those states that stand for termination. (Without loss of generality it can be assumed that transition systems have no terminal states, i.e., this means terminating states either are equipped with a self-loop, or have a transition leading to a trap-state with a self-loop and no other outgoing transitions.)

Partial correctness can be represented by a formula of the form

$$
\square(\text { halt } \longrightarrow \diamond(y=f(x)))
$$

where $y$ is the output variable and $x$ the input variable which is assumed not to change during program execution. Additional initial conditions such as expressed by the formula init can be added as premise as follows:

$$
\text { init } \longrightarrow \square(\text { halt } \longrightarrow \diamond(y=f(x)))
$$

It should be stressed that this is an extremely simplified representation. In practice, predicate logic concepts are needed to precisely formulate partial correctness. And even in cases where propositional logic formulae of the above form can be used to exactly describe termination and partial correctness, the algorithmic proof of the LTL formulae is very difficult or even impossible. (Recall the undecidability of the halting problem.)

## Remark 5.15. Specifying Timed Properties with LTL for Synchronous Systems

For synchronous systems, LTL can be used as a formalism to specify "real-time" properties that refer to a discrete time scale. Recall that in synchronous systems, the involved processes proceed in a lock step fashion, i.e., at each discrete time instance each process performs a (sometimes idle) step. In this kind of system, the next-step operator $\bigcirc$ has a "timed" interpretation: $\bigcirc \varphi$ states that "at the next time instant $\varphi$ holds". By putting applications of $\bigcirc$ in sequence, we obtain, e.g.:

$$
\bigcirc^{k} \varphi \stackrel{\text { def }}{=} \underbrace{\bigcirc \ldots \bigcirc}_{k \text {-times }} \varphi \quad \text { " } \varphi \text { holds after (exactly) } k \text { time instants". }
$$

Assertions like " $\varphi$ will hold within at most $k$ time instants" are obtained by

$$
\diamond^{\leqslant k} \varphi=\bigvee_{0 \leqslant i \leqslant k} \bigcirc^{i} \varphi
$$

Statements like " $\varphi$ holds now and will hold during the next $k$ instants" can be represented as follows:

$$
\square^{\leqslant k} \varphi=\neg \diamond \leqslant k \neg \varphi=\neg \bigvee_{0 \leqslant i \leqslant k} \bigcirc^{i} \neg \varphi
$$

For the modulo 4 counter of Example 5.11 (page 240) we in fact already implicitly used LTL-formulae as real-time specifications. For example, the formula expressing that once the output is $y=1$, the next three steps the output is $y=0$ :

$$
\square(y \longrightarrow(\bigcirc \neg y \wedge \bigcirc \bigcirc \neg y \wedge \bigcirc \bigcirc \bigcirc \neg y))
$$

can be abbreviated as $\square(y \longrightarrow \bigcirc \square \leqslant 2 \neg y)$.
It should, however, be noted that the temporal interpretation of the next-step operator is only appropriate for synchronous systems. Every transition in these systems represents the cumulative effect of the actions possible within a single time instant. For asynchronous systems (for which the transition system representation is time-abstract), the next-step operator cannot be interpreted as a real-time modality. In fact, for asynchronous systems the next-step operator should be used with care. The phase changes of a traffic light, for example, can be described by

$$
\varphi=\square(\text { green } \rightarrow \bigcirc \text { yellow }) \wedge \square(\text { yellow } \rightarrow \bigcirc \text { red }) \wedge \ldots
$$

For the interleaving of two independent traffic lights (Example 2.17 on page 36) and the formulae $\varphi_{1}, \varphi_{2}$ (where the indexed atomic propositions green $_{i}$, yellow $_{i}$, etc., are used),

$$
\operatorname{Tr}^{L i g h t} t_{1}| | \operatorname{TrLight}_{2} \mid \vDash \varphi_{1} \wedge \varphi_{2}
$$

This stems from the fact that, e.g., the first traffic light does not change its location when the second traffic light changes its phase. To avoid this problem, the until operator can be used instead, e.g.,

$$
\varphi^{\prime}=\square(\text { green } \rightarrow(\text { green } \cup \text { yellow })) \wedge \square(\text { yellow } \rightarrow(\text { yellow } \cup \text { red })) \wedge \ldots
$$

This differs from the synchronous product operator $\otimes$, where

$$
\operatorname{TrLight}_{1} \otimes \operatorname{TrLight}_{2} \models \varphi .
$$

## Remark 5.16. Other Notations and Variants of LTL

Many variants and notations have been introduced for LTL. Alternative notations for the temporal modalities are $X$ for $\bigcirc$ (ne $X \mathrm{t}$ ), $F$ for $\diamond$ (Finally), and $G$ for $\square$ (Globally). All operators from LTL refer to the future (including the current state). Consequently, operators are known as future operators. LTL can, however, also be extended with past operators. This can be useful for specifying some properties more easily (and succinctly) in terms of the past than in terms of the future. For instance, $\square^{-1} a$ ("always in the past") means that $a$ is valid now and in any state in the past. $\diamond^{-1} a$ ("sometime in the past") means that either $a$ is valid in the current state or in some state in the past and $\bigcirc^{-1} a$ means that $a$ holds in the previous state, provided such state exists. For example, the property "every red light phase is preceded by a yellow one" can be described by

$$
\square\left(\text { red } \rightarrow \bigcirc^{-1} \text { yellow }\right) .
$$

The main reason for introducing past operators is to simplify the specification of several properties. The expressive power of the logic is, however, not affected by the addition of past operators when a discrete notion of time is taken (as we do). Thus, for any property which contains one or more past operators, an LTL-formula with only future temporal operators exists expressing the same thing. More information is described in Section 5.4.

### 5.1.4 Equivalence of LTL Formulae

For any type of logic, a clear separation between syntax and semantics is an essential aspect. On the other hand, two formulae are intuitively identified whenever they have the

| duality law | idempotency law |
| :---: | :---: |
| $\neg \bigcirc \varphi \equiv \bigcirc \neg \varphi$ | $\diamond \diamond \varphi \equiv \diamond \varphi$ |
| $\neg \nabla \varphi \equiv \square \neg \varphi$ | $\square \square \varphi \equiv \square \varphi$ |
| $\neg \square \varphi \equiv \diamond \neg \varphi$ | $\varphi \mathrm{U}(\varphi \mathrm{U} \psi) \equiv \varphi \mathrm{U} \psi$ |
|  | $(\varphi \cup \psi) \cup \psi \equiv \varphi \mathrm{U} \psi$ |
| absorption law | expansion law |
| $\diamond \square \diamond \varphi \equiv \square \diamond \varphi$ | $\varphi \mathrm{U} \psi \equiv \psi \vee(\varphi \wedge \bigcirc(\varphi \cup \psi))$ |
| $\square \diamond \square \varphi \equiv \diamond \square \varphi$ | $\diamond \psi \equiv \psi \vee \bigcirc\rangle \psi$ |
|  | $\square \psi \equiv \psi \wedge \bigcirc \square \psi$ |
| distributive law |  |
| $\bigcirc(\varphi \cup \psi) \equiv$ | ) $\mathrm{U}(\bigcirc \psi)$ |
| $\diamond(\varphi \vee \psi) \equiv$ | $\Delta \psi$ |
| $\square(\varphi \wedge \psi) \equiv$ | $\square \psi$ |

Figure 5.7: Some equivalence rules for LTL.
same truth-value under all interpretations. For example, it seems useless to distinguish between $\neg \neg a$ and $a$, although both formulae are syntactically different.

## Definition 5.17. Equivalence of LTL Formulae

LTL formulae $\varphi_{1}, \varphi_{2}$ are equivalent, denoted $\varphi_{1} \equiv \varphi_{2}$, if $\operatorname{Words}\left(\varphi_{1}\right)=\operatorname{Words}\left(\varphi_{2}\right)$.

As LTL subsumes propositional logic, equivalences of propositional logic also hold for LTL, e.g., $\neg \neg \varphi \equiv \varphi$ and $\varphi \wedge \varphi \equiv \varphi$. In addition, there exist a number of equivalence rules for temporal modalities. They include the equivalence laws indicated in Figure 5.7. We explain some of these equivalence laws. The duality rule $\neg \bigcirc \varphi \equiv \bigcirc \neg \varphi$ shows that the next-step operator $\bigcirc$ is dual to itself. It results from the observation that

|  | $A_{0} A_{1} A_{2} \ldots \models \neg \bigcirc \varphi$ |
| :--- | :--- |
| iff | $A_{0} A_{1} A_{2} \ldots \not \models \bigcirc \varphi$ |
| iff | $A_{1} A_{2} \ldots \not \models \varphi$ |
| iff | $A_{1} A_{2} \ldots \models \neg \varphi$ |
| iff | $A_{0} A_{1} A_{2} \ldots \models \bigcirc \neg \varphi$. |

The first absorption law is explained by the fact that "infinitely often $\varphi$ " is equal to "from
a certain point of time on, $\varphi$ is true infinitely often".
The distributive laws for $\diamond$ and disjunction, or $\square$ and conjunction, respectively, are dual to each other. They can be regarded as the temporal logic analogon to the distributive laws for $\exists$ and $\vee$ or $\forall$ and $\wedge$ in predicate logic. It should be noted, however, that $\diamond$ does not distribute over conjunction (as existential quantification), and $\square$ does not distribute over disjunction (like universal quantification):

$$
\diamond(a \wedge b) \not \equiv \diamond a \wedge \diamond b \quad \text { and } \quad \square(a \vee b) \not \equiv \square a \vee \square b
$$

The formula $\diamond(a \wedge b)$ ensures that a state will be reached for which $a$ and $b$ hold, while $\diamond a \wedge \diamond b$ ensures that eventually an $a$-state and eventually a $b$-state will be reached. According to the latter formula, $a$ and $b$ need not to be satisfied at the same time. Figure 5.8 depicts a transition system that satisfies $\diamond a \wedge \diamond b$, but not $\diamond(a \wedge b)$.


Figure 5.8: $T S \not \vDash \diamond(a \wedge b)$ and $T S \models \diamond a \wedge \diamond b$.
The expansion laws play an important role. They describe the temporal modalities $\mathrm{U}, \diamond$, and $\square$ by means of a recursive equivalence. These equivalences all have the same global structure: they assert something about the current state, and about the direct successor state. The assertion about the current state is done without the need to use temporal modalities whereas the assertion about the next state is done using the $\bigcirc$ operator. The expansion law for until, for instance, can be considered as follows. Formula $\phi \mathrm{U} \psi$ is a solution of the equivalence

$$
\kappa \equiv \underset{\text { current state }}{\psi \vee} \underset{\nearrow}{\vee}(\varphi \wedge \underset{\text { first suffix }}{\bigcirc} \kappa \kappa)
$$

Let us explain the expansion law for the until operator in more detail. Let $A_{0} A_{1} A_{2} \ldots$ be an infinite word over the alphabet $2^{A P}$, such that $A_{0} A_{1} A_{2} \ldots \models \varphi \cup \psi$. By the definition of "until", there exists a $k \geqslant 0$, such that

$$
A_{i} A_{i+1} A_{i+2} \ldots \vDash \varphi, \quad \text { for all } 0 \leqslant i<k \quad \text { and } \quad A_{k} A_{k+1} A_{k+2} \ldots=\psi
$$

Distinguish between $k=0$ and $k>0$. If $k=0$, then $A_{0} A_{1} A_{2} \ldots \models \psi$ and thus $A_{0} A_{1} A_{2} \ldots \models \psi \vee \ldots$. If $k>0$, then

$$
A_{0} A_{1} A_{2} \ldots \models \varphi \quad \text { and } \quad A_{1} A_{2} \ldots \models \varphi \cup \psi
$$

From this it immediately follows that

$$
A_{0} A_{1} A_{2} \ldots \models \varphi \wedge \bigcirc(\varphi \cup \psi)
$$

Gathering the results for $k=0$ and $k>0$ yields

$$
A_{0} A_{1} A_{2} \ldots \models \psi \vee(\varphi \wedge \bigcirc(\varphi \cup \psi))
$$

For the reverse direction, a similar argument can be provided.

The expansion law for $\Delta \psi$ is a special case of the expansion law for until:

$$
\diamond \psi=\operatorname{true} \cup \psi \equiv \psi \vee \underbrace{(\operatorname{true} \wedge \bigcirc(\operatorname{true} \cup \psi))}_{\equiv \bigcirc(\operatorname{true} \cup \psi)=\bigcirc \diamond \psi} \equiv \psi \vee \bigcirc \diamond \psi
$$

The expansion law for $\square \psi$ now results from the duality of $\diamond$ and $\square$, the duality of $\vee$ and $\wedge$ (i.e., de Morgan's law) and from the duality law for $\bigcirc$

$$
\begin{array}{lr}
=\square \psi & (* \text { definition of } \square *) \\
\equiv \neg \diamond \neg \psi & \left(* \text { expansion law for } \diamond^{*}\right) \\
\equiv \neg(\neg \psi \vee \bigcirc \diamond \neg \psi) & \left(* \text { de Morgan's law }{ }^{*}\right) \\
\equiv \neg \neg \psi \wedge \neg \bigcirc \diamond \neg \psi & \left(* \text { self-duality of } \bigcirc^{*}\right) \\
\left.\equiv \begin{array}{l} 
\\
\equiv \\
\equiv
\end{array}\right) .\left(* \text { definition of } \square{ }^{*}\right)
\end{array}
$$

It is important to realize that none of the indicated expansion laws represents a complete recursive characterization of the temporal operator at hand. For example, the formulae $\varphi=$ false and $\varphi=\square a$ both satisfy the recursive "equation" $\varphi \equiv a \wedge \bigcirc \varphi$ since false $\equiv$ $a \wedge \bigcirc$ false and $\square a \equiv a \wedge \bigcirc \square a$. However, $\varphi \cup \psi$ and $\diamond \varphi$ are the least solutions of the expansion law indicated for "until" and "eventually", respectively. Similarly, $\square \varphi$ is the greatest solution of the expansion law for "always". The precise meaning of these statements will be explained by considering the until operator as example:

## Lemma 5.18. Until is the Least Solution of the Expansion Law

For LTL formulae $\varphi$ and $\psi$, $\operatorname{Words}(\varphi \cup \psi)$ is the least LT property $P \subseteq\left(2^{A P}\right)^{\omega}$ such that:

$$
\begin{equation*}
\operatorname{Words}(\psi) \cup\left\{A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi) \mid A_{1} A_{2} \ldots \in P\right\} \subseteq P \tag{*}
\end{equation*}
$$

Moreover, Words $(\varphi \cup \psi)$ agrees with the set
$\operatorname{Words}(\psi) \cup\left\{A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi) \mid A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi \cup \psi)\right\}$.

The formulation "least LT property satisfying condition (*)" means that the following conditions hold:
(1) $P=\operatorname{Words}(\varphi \cup \psi)$ satisfies $\left(^{*}\right)$.
(2) $\operatorname{Words}(\varphi \cup \psi) \subseteq P$ for all LT properties $P$ satisfying condition (*).

Proof: Condition (1) follows immediately from the expansion law $\varphi \mathrm{U} \psi \equiv \psi \vee(\varphi \wedge$ $\bigcirc(\varphi \cup \psi))$. In fact, the expansion law even yields that $\subseteq$ in $\left(^{*}\right)$ can be replaced by equality, i.e., $\operatorname{Words}(\varphi \cup \psi)$ agrees with

$$
\operatorname{Words}(\psi) \cup\left\{A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi) \mid A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi \cup \psi)\right\}
$$

To prove condition (2), $P$ is assumed to be an LT property that satisfies $\left(^{*}\right)$. We show that $\operatorname{Words}(\varphi \cup \psi) \subseteq P$. Since $P$ fulfills $\left(^{*}\right)$, we have:
(i) $\operatorname{Words}(\psi) \subseteq P$,
(ii) If $B_{0} B_{1} B_{2} \ldots \in \operatorname{Words}(\varphi)$ and $B_{1} B_{2} \ldots \in P$ then $B_{0} B_{1} B_{2} \ldots \in P$.

Let $A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi \cup \psi)$. Then, there exists an index $k \geqslant 0$ such that
(iii) $A_{i} A_{i+1} A_{i+2} \ldots \in \operatorname{Words}(\varphi)$, for all $0 \leqslant i<k$,
(iv) $A_{k} A_{k+1} A_{k+2} \ldots \in \operatorname{Words}(\psi)$.

We now derive

$$
\begin{array}{lll} 
& A_{k} A_{k+1} A_{k+2} A_{k+3} \ldots \in P & \text { due to (iv) and (i) } \\
\Longrightarrow & A_{k-1} A_{k} A_{k+1} A_{k+2} \ldots \in P & \text { due to (ii) and (iii) } \\
\Longrightarrow & A_{k-2} A_{k-1} A_{k} A_{k+1} \ldots \in P & \text { due to (ii) and (iii) } \\
\vdots & & \\
\Longrightarrow & A_{0} A_{1} A_{2} A_{3} \ldots \in P & \text { due to (ii) and (iii). }
\end{array}
$$

### 5.1.5 Weak Until, Release, and Positive Normal Form

Any LTL formula can be transformed into a canonical form, the so-called positive normal form (PNF). This canonical form is characterized by the fact that negations only occur adjacent to atomic propositions. PNF formulae in propositional logic are constructed from the constants true and false, the literals $a$ and $\neg a$, and the operators $\wedge$ and $\vee$. For instance, $\neg a \wedge((\neg b \wedge c) \vee \neg a)$ is in PNF, while $\neg(a \wedge \neg b)$ is not. The well-known disjunctive and conjunctive normal forms are special cases of the PNF for propositional logic.

In order to transform any LTL formula into PNF, for each operator a dual operator needs to be incorporated into the syntax of PNF formulae. The propositional logical primitives of the positive normal form for LTL are the constant true and its dual constant false $=\neg$ true, as well as conjunction $\wedge$ and its dual, $\vee$. De Morgan's rules $\neg(\varphi \vee \psi) \equiv \neg \varphi \wedge \neg \psi$ and $\neg(\varphi \wedge \psi) \equiv \neg \varphi \vee \neg \psi$ yield the duality of conjunction and disjunction. According to the duality rule $\neg \bigcirc \varphi \equiv \bigcirc \neg \varphi$, the next-step operator is a dual of itself. Therefore, no extra operator is necessary for $\bigcirc$. Now consider the until operator. First we observe that

$$
\neg(\varphi \cup \psi) \equiv((\varphi \wedge \neg \psi) \cup(\neg \varphi \wedge \neg \psi)) \vee \square(\varphi \wedge \neg \psi) .
$$

The first disjunct on the right-hand side asserts that $\varphi$ stops to hold "too early", i.e., before $\psi$ becomes valid. The second disjunct states that $\varphi$ always holds but never $\psi$. Clearly, in both cases, $\neg(\varphi \cup \psi)$ holds.

This observation provides the motivation to introduce the operator W (called weak until or unless) as the dual of $\mathbf{U}$. It is defined by:

$$
\varphi \mathrm{W} \psi \stackrel{\text { def }}{=}(\varphi \mathrm{U} \psi) \vee \square \varphi
$$

The semantics of $\varphi \mathrm{W} \psi$ is similar to that of $\varphi \mathrm{U} \psi$, except that $\varphi \mathrm{U} \psi$ requires a state to be reached for which $\psi$ holds, whereas this is not required for $\varphi \mathrm{W} \psi$. Until U and weak until W are dual in the following sense:

$$
\begin{aligned}
\neg(\varphi \mathrm{U} \psi) & \equiv(\varphi \wedge \neg \psi) \mathrm{W}(\neg \varphi \wedge \neg \psi) \\
\neg(\varphi \mathrm{W} \psi) & \equiv(\varphi \wedge \neg \psi) \cup(\neg \varphi \wedge \neg \psi)
\end{aligned}
$$

The reason that weak until is not a standard operator for LTL is that $U$ and $W$ have the same expressiveness, since

$$
\begin{aligned}
\square \psi & \equiv \psi \mathrm{W} \text { false, } \\
\varphi \mathrm{U} \psi & \equiv(\varphi \mathrm{~W} \psi) \wedge \underbrace{\diamond \psi}_{\equiv \neg \square \neg \psi}
\end{aligned}
$$

That is to say, weak until is relevant only if restrictions are imposed on the occurrence of negation (as in PNF). It is interesting to observe that $W$ and $U$ satisfy the same expansion law:

$$
\varphi \mathrm{W} \psi \equiv \psi \vee(\varphi \wedge \bigcirc(\varphi \mathrm{~W} \psi))
$$

For $\psi=$ false this results in the expansion law for $\square \varphi$ :

$$
\square \varphi=\varphi \mathrm{W} \text { false } \equiv \text { false } \vee(\varphi \wedge \bigcirc(\varphi \mathrm{W} \text { false })) \equiv \varphi \wedge \bigcirc \square \varphi
$$

The semantic difference between U and W is shown by the fact that $\varphi \mathrm{W} \psi$ is the greatest solution of

$$
\kappa \equiv \psi \vee(\varphi \wedge \bigcirc \kappa)
$$

This result is proven below. Recall $\varphi \mathrm{U} \psi$ is the least solution of this equivalence, see Lemma 5.18 on page 251.

## Lemma 5.19. Weak-Until is the Greatest Solution of the Expansion Law

For LTL formulae $\varphi$ and $\psi$, $\operatorname{Words}(\varphi \mathrm{W} \psi)$ is the greatest LT property $P \subseteq\left(2^{A P}\right)^{\omega}$ such that:

$$
\begin{equation*}
\operatorname{Words}(\psi) \cup\left\{A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi) \mid A_{1} A_{2} \ldots \in P\right\} \supseteq P \tag{*}
\end{equation*}
$$

Moreover, Words $(\varphi \mathrm{W} \psi)$ agrees with the $L T$ property

$$
\operatorname{Words}(\psi) \cup\left\{A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi) \mid A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi \operatorname{W} \psi)\right\}
$$

The formulation "greatest LT property with the indicated condition $\left(^{*}\right)$ is to be understood in the following sense:
(1) $P \supseteq \operatorname{Words}(\varphi \mathrm{~W} \psi)$ satisfies $\left({ }^{*}\right)$.
(2) $\operatorname{Words}(\varphi \mathrm{W} \psi) \supseteq P$ for all LT properties $P$ satisfying condition (*).

Proof: The fact that condition (1) is satisfied, even with equality rather than $\supseteq$, i.e., Words $(\varphi \mathrm{W} \psi)$ agrees with

$$
\operatorname{Words}(\psi) \cup\left\{A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi) \mid A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi \operatorname{W} \psi)\right\}
$$ is an immediate conclusion from the expansion law $\varphi \mathrm{W} \psi \equiv \psi \vee(\varphi \wedge \bigcirc(\varphi \mathrm{W} \psi))$.

For proving condition (2), $P$ is assumed to be an LT property satisfying $\left(^{*}\right)$. In particular, for all words $B_{0} B_{1} B_{2} B_{3} \ldots \in\left(2^{A P}\right)^{\omega} \backslash \operatorname{Words}(\psi)$ we have:
(i) If $B_{0} B_{1} B_{2} B_{3} \ldots \notin \operatorname{Words}(\varphi)$, then $B_{0} B_{1} B_{2} B_{3} \ldots \notin P$.
(ii) If $B_{1} B_{2} B_{3} \ldots \notin P$, then $B_{0} B_{1} B_{2} B_{3} \ldots \notin P$.

Now we demonstrate that

$$
\left(2^{A P}\right)^{\omega} \backslash \operatorname{Words}(\varphi \mathrm{W} \psi) \subseteq\left(2^{A P}\right)^{\omega} \backslash P
$$

Let $A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega} \backslash \operatorname{Words}(\varphi \mathrm{W} \psi)$. Then $A_{0} A_{1} A_{2} \ldots \not \vDash \varphi \mathrm{~W} \psi$ and thus

$$
A_{0} A_{1} A_{2} \ldots \models \neg(\varphi \mathrm{~W} \psi) \equiv(\varphi \wedge \neg \psi) \cup(\neg \varphi \wedge \neg \psi)
$$

Thus there exists $k \geqslant 0$, such that:
(iii) $A_{i} A_{i+1} A_{i+2} \ldots \vDash \varphi \wedge \neg \psi, \quad$ for all $0 \leqslant i<k$,
(iv) $A_{k} A_{k+1} A_{k+2} \ldots \models \neg \varphi \wedge \neg \psi$.

In particular, none of the words $A_{i} A_{i+1} A_{i+2} \ldots$ belongs to $\operatorname{Words}(\psi)$ for $0 \leqslant i \leqslant k$. We obtain:

$$
\begin{array}{rll} 
& A_{k} A_{k+1} A_{k+2} A_{k+3} \ldots \notin P & \text { due to (i) and (iv) } \\
\Longrightarrow & A_{k-1} A_{k} A_{k+1} A_{k+2} \ldots \notin P & \text { due to (ii) and (iii) } \\
\Longrightarrow & A_{k-2} A_{k-1} A_{k} A_{k+1} \ldots \notin P & \text { due to (ii) and (iii) } \\
& \vdots &
\end{array}
$$

Thus, $\left(2^{A P}\right)^{\omega} \backslash \operatorname{Words}(\varphi \mathrm{W} \psi) \subseteq\left(2^{A P}\right)^{\omega} \backslash P$, or equivalently, $\operatorname{Words}(\varphi \mathrm{W} \psi) \supseteq P$.

We are now ready to introduce the positive normal form for LTL which permits negation only on the level of literals and - to ensure the full expressiveness of LTL - uses the dual Boolean connectors $\wedge$ and $\vee$, the self-dual next-step operator $\bigcirc$, and $U$ and $W$ :

## Definition 5.20. Positive Normal Form for LTL (Weak-Until PNF)

For $a \in A P$, the set of LTL formulae in weak-until positive normal form (weak-until PNF, for short, or simply PNF) is given by:

$$
\varphi::=\text { true } \mid \text { false }|a| \neg a\left|\varphi_{1} \wedge \varphi_{2}\right| \varphi_{1} \vee \varphi_{2}|\bigcirc \varphi| \varphi_{1} \cup \varphi_{2} \mid \varphi_{1} \mathrm{~W} \varphi_{2}
$$

Due to the law $\square \varphi \equiv \varphi \mathrm{W}$ false, $\square$ can also be considered as permitted operator of the $W$-positive normal form. As before, $\Delta \varphi=\operatorname{true} U \varphi$. An example LTL formula in PNF is

$$
\diamond(a \cup \square b) \vee(a \wedge \neg c) \mathrm{W}(\diamond(\neg a \cup b))
$$

The LTL formulae $\neg(a \cup b)$ and $c \vee \neg(a \wedge \diamond b)$ are not in weak-until PNF.

The previous considerations were aimed to rewrite any LTL formula into weak-until PNF. This is done by successively "pushing" negations "inside" the formula at hand. This is facilitated by the following transformations:

$$
\begin{array}{ll}
\neg \text { true } & \rightsquigarrow \text { false } \\
\neg \text { false } & \rightsquigarrow \text { true } \\
\neg \neg \varphi & \rightsquigarrow \varphi \\
\neg(\varphi \wedge \psi) & \rightsquigarrow \neg \varphi \vee \neg \psi \\
\neg \bigcirc \varphi & \rightsquigarrow \bigcirc \neg \varphi \\
\neg(\varphi \cup \psi) & \rightsquigarrow(\varphi \wedge \neg \psi) \mathrm{W}(\neg \varphi \wedge \neg \psi) .
\end{array}
$$

These rewrite rules are lifted to the derived operators as follows:

$$
\neg(\varphi \vee \psi) \rightsquigarrow \neg \varphi \wedge \neg \psi \quad \text { and } \quad \neg \diamond \varphi \rightsquigarrow \square \neg \varphi \quad \text { and } \quad \neg \square \varphi \rightsquigarrow \diamond \neg \varphi \text {. }
$$

## Example 5.21. Positive Normal Form

Consider the LTL formula $\neg \square((a \cup b) \vee \bigcirc c)$. This formula is not in PNF, but can be transformed into an equivalent LTL formula in weak-until PNF as follows:

$$
\begin{aligned}
& \neg \square((a \cup b) \vee \bigcirc c) \\
\equiv & \diamond \neg((a \cup b) \vee \bigcirc c) \\
\equiv & \diamond(\neg(a \cup b) \wedge \neg \bigcirc c) \\
\equiv & \diamond((a \wedge \neg b) \mathrm{W}(\neg a \wedge \neg b) \wedge \bigcirc \neg c)
\end{aligned}
$$

## Theorem 5.22. Existence of Equivalent Weak-Until PNF Formulae

For each LTL formula there exists an equivalent LTL formula in weak-until PNF.

The main drawback of the rewrite rules introduced above is that the length of the resulting LTL formula (in weak-until PNF) may be exponential in the length of the original nonPNF LTL formula. This is due to the rewrite rule for until where $\varphi$ and $\psi$ are duplicated. Although this can be slightly improved by adopting the law:

$$
\neg(\varphi \cup \psi) \equiv(\neg \psi) \mathrm{W}(\neg \varphi \wedge \neg \psi)
$$

an exponential growth in length is not avoided here too due to the duplication of $\psi$ in the right-hand side.

To avoid this exponential blowup in transforming an LTL formula in PNF, another temporal modality is used that is dual to the until operator: the so-called binary release-operator, denoted R. It is defined by

$$
\varphi \mathrm{R} \psi \stackrel{\text { def }}{=} \neg(\neg \varphi \mathrm{U} \neg \psi) .
$$

Its intuitive interpretation is as follows. Formula $\varphi \mathrm{R} \psi$ holds for a word if $\psi$ always holds, a requirement that is released as soon as $\varphi$ becomes valid. Thus, the formula false $\mathrm{R} \varphi$ is valid if $\varphi$ always holds, since the release condition (false) is a contradiction. Formally, for a given word $\sigma=A_{0} A_{1} \ldots \in\left(2^{A P}\right)^{\omega}$ :

```
    \(\sigma \models \varphi \mathrm{R} \psi\)
iff (* definition of \(\left.\mathrm{R}{ }^{*}\right)\)
    \(\neg \exists j \geqslant 0 .(\sigma[j ..] \models \neg \psi \wedge \forall i<j . \sigma[i ..] \models \neg \varphi)\)
iff (* semantics of negation \(\left.{ }^{*}\right)\)
    \(\neg \exists j \geqslant 0 .(\sigma[j ..] \neq \psi \wedge \forall i<j . \sigma[i ..] \not \vDash \varphi)\)
iff (* duality of \(\exists\) and \(\forall *)\)
    \(\forall j \geqslant 0 . \neg(\sigma[j ..] \not \vDash \psi \wedge \forall i<j . \sigma[i ..] \not \vDash \varphi)\)
iff (* de Morgan's law *)
    \(\forall j \geqslant 0 .(\neg(\sigma[j ..] \not \vDash \psi) \vee \neg \forall i<j . \sigma[i ..] \not \vDash \varphi)\)
iff (* semantics of negation \(\left.{ }^{*}\right)\)
    \(\forall j \geqslant 0 .(\sigma[j ..] \models \psi \vee \exists i<j . \sigma[i ..] \models \varphi)\)
iff
    \(\forall j \geqslant 0 . \sigma[j ..] \models \psi \quad\) or \(\quad \exists i \geqslant 0 .(\sigma[i ..] \models \varphi) \wedge \forall k \leqslant i . \sigma[k ..] \models \psi)\).
```

The always operator is obtained from the release operator by:
$\square \varphi \equiv$ false $\mathrm{R} \varphi$.

The weak-until and the until operator are obtained by:

$$
\varphi \mathrm{W} \psi \equiv(\neg \varphi \vee \psi) \mathrm{R}(\varphi \vee \psi), \quad \varphi \cup \psi \equiv \neg(\neg \varphi \mathrm{R} \neg \psi)
$$

Vice versa, $\varphi \mathrm{R} \psi \equiv(\neg \varphi \wedge \psi) \mathrm{W}(\varphi \wedge \psi)$. The expansion law (see Exercise 5.8) for release reads as follows:

$$
\varphi \mathrm{R} \psi \equiv \psi \wedge(\varphi \vee \bigcirc(\varphi \mathrm{R} \psi))
$$

We now revisit the notion of PNF, which is defined using the R -operator rather than W :

## Definition 5.23. Positive Normal Form (release PNF)

For $a \in A P$, LTL formulae in release positive normal form (release PNF, or simply PNF) are given by

$$
\varphi::=\text { true } \mid \text { false }|a| \neg a\left|\varphi_{1} \wedge \varphi_{2}\right| \varphi_{1} \vee \varphi_{2}|\bigcirc \varphi| \varphi_{1} \cup \varphi_{2} \mid \varphi_{1} \mathrm{R} \varphi_{2}
$$

The following transformation rules push negations inside and serve to transform a given LTL formula into an equivalent LTL formula in PNF:

$$
\begin{array}{ll}
\neg \text { true } & \rightsquigarrow \text { false } \\
\neg \neg \varphi & \rightsquigarrow \varphi \\
\neg(\varphi \wedge \psi) & \rightsquigarrow \neg \varphi \vee \neg \psi \\
\neg \bigcirc \varphi & \rightsquigarrow \bigcirc \neg \varphi \\
\neg(\varphi \cup \psi) & \rightsquigarrow \neg \varphi \mathrm{R} \neg \psi
\end{array}
$$

In each rewrite rule the size of the resulting formula increases at most by an additive constant.

## Theorem 5.24. Existence of Equivalent Release PNF Formulae

For any LTL formula $\varphi$ there exists an equivalent LTL formula $\varphi^{\prime}$ in release PNF with $\left|\varphi^{\prime}\right|=\mathcal{O}(|\varphi|)$.

### 5.1.6 Fairness in LTL

In Chapter 3, we have seen that typically some fairness constraints are needed to verify liveness properties. Three types of fairness constraints (for sets of actions) have been distinguished, namely unconditional, strong, and weak fairness. Accordingly, the satisfaction
relation for LT properties (denoted $\models$ ) has been adapted to the fairness assumption $\mathcal{F}$ (denoted $\models_{\mathcal{F}}$ ), where $\mathcal{F}$ is a triple of (sets of) fairness constraints. This entails that only fair paths are considered while determining the satisfaction of a property. In this section, this approach is adopted in the context of LTL. That is to say, rather than determining for transition system $T S$ and LTL formula $\varphi$ whether $T S \models \varphi$, we focus on the fair executions of TS. The main difference with the action-based fairness assumptions (and constraints) is that we now focus on state-based fairness.

## Definition 5.25. LTL Fairness Constraints and Assumptions

Let $\Phi$ and $\Psi$ be propositional logic formulae over $A P$.

1. An unconditional LTL fairness constraint is an LTL formula of the form

$$
\text { ufair }=\square \diamond \Psi
$$

2. A strong LTL fairness condition is an LTL formula of the form

$$
\text { sfair }=\square \diamond \Phi \longrightarrow \square \diamond \Psi .
$$

3. A weak LTL fairness constraint is an LTL formula of the form

$$
\text { wfair }=\diamond \square \Phi \longrightarrow \square \diamond \Psi
$$

An LTL fairness assumption is a conjunction of LTL fairness constraints (of any arbitrary type).

For instance, a strong LTL fairness assumption denotes a conjunction of strong LTL fairness constraints, i.e., a formula of the form

$$
\text { sfair }=\bigwedge_{0<i \leqslant k}\left(\square \diamond \Phi_{i} \longrightarrow \square \diamond \Psi_{i}\right)
$$

for propositional logical formulae $\Phi_{i}$ and $\Psi_{i}$ over $A P$. Weak and unconditional LTL fairness assumptions are defined in a similar way.

In their most general form, LTL fairness assumptions are (as in Definition 3.46, page 133) a conjunction of unconditional, strong, and weak fairness assumptions:

$$
\text { fair }=\text { ufair } \wedge \text { sfair } \wedge \text { wfair. }
$$

where ufair, sfair, and wfair are unconditional, strong, and weak LTL fairness assumptions, respectively. As in the case of action-based fairness assumptions, the rule of thumb for imposing fairness assumptions is: strong (or unconditional) fairness assumptions are useful for solving contentions, and weak fairness is often sufficient for resolving the nondeterminism that results from interleaving.

In the sequel, we adopt the same notations as for action-based fairness assumptions. Let FairPaths ( $s$ ) denote the set of all fair paths starting in $s$ and FairTraces(s) the set of all traces induced by fair paths starting in $s$. Formally, for fixed formula fair,

$$
\begin{aligned}
& \operatorname{FairPaths}(s)=\{\pi \in \operatorname{Paths}(s) \mid \pi \models \text { fair }\} \\
& \text { FairTraces }(s)=\{\operatorname{trace}(\pi) \mid \pi \in \operatorname{FairPaths}(s)\} .
\end{aligned}
$$

These notions can be lifted to transition systems in the obvious way yielding FairPaths( $T S$ ) and FairTraces $(T S)$. To identify the fairness assumption fair, we may write FairPathsfair $(\cdot)$ or FairTracesfair $(\cdot)$.

## Definition 5.26. Satisfaction Relation for LTL with Fairness

For state $s$ in transition system $T S$ (over $A P$ ) without terminal states, LTL formula $\varphi$, and LTL fairness assumption fair let

$$
\begin{aligned}
& s \models_{\text {fair }} \varphi \quad \text { iff } \quad \forall \pi \in \operatorname{FairPaths}(s) . \pi \models \varphi \quad \text { and } \\
& T S \models_{\text {fair }} \varphi \text { iff } \forall s_{0} \in I . s_{0} \models_{\text {fair }} \varphi \text {. }
\end{aligned}
$$

TS satisfies $\varphi$ under the LTL fairness assumption fair if $\varphi$ holds for all fair paths that originate from some initial state.

## Example 5.27. Mutual Exclusion with Randomized Arbiter (Fairness)

Consider the following approach to two-process mutual exclusion. A randomized arbiter, see the program graphs in Figure 5.9, decides which process is acquiring access to the critical section. It does so by tossing coins. We abstract from the probabilities, and model the coin tossing by a nondeterministic choice between the alternatives "heads" and "tails". It is assumed that the two contending processes communicate with the arbiter via the actions enter ${ }_{1}$ and enter ${ }_{2}$. The critical section is released by synchronizing over


Figure 5.9: Mutual exclusion with a randomized arbiter.
the action release. For the sake of simplicity, we refrain from indicating which process is releasing the critical section.

The property "process $P_{1}$ is in its critical section infinitely often" cannot be established, since, for instance, the underlying transition system representation does not exclude an execution in which only the second process may perform an action while $P_{1}$ is entirely ignored. Thus:

$$
T S_{1} \| \text { Arbiter } \| T S_{2} \not \vDash \square \diamond \text { crit }_{1}
$$

If a coin is assumed to be fair enough such that both events "heads" and "tails" occur with positive probability, the unrealistic case of one of the two alternatives never happening can be ignored by means of the unconditional LTL fairness assumption:

$$
\text { fair }=\square \diamond \text { heads } \wedge \square \diamond \text { tails. }
$$

It is not difficult to check that now:

$$
T S_{1} \| \text { Arbiter } \| T S_{2} \models_{\text {fair }} \square \diamond \text { crit }_{1} \wedge \square \diamond \text { crit }_{2}
$$

## Example 5.28. Communication Protocol (Fairness)

Consider the alternating bit protocol as described in Example 2.32 (page 57). For the sake of convenience, the program graph of the sender of the alternating bit protocol is repeated in Figure 5.10. The liveness property $\square \diamond$ start states that the protocol returns infinitely often to its initial state. In this initial state the action $\operatorname{snd\_ msg}(0)$ is enabled. It follows that

$$
A B P \not \models \square \diamond \text { start }
$$



Figure 5.10: Program graph of ABP sender $S$.
since the unrealistic scenario in which (after some finite time) each message with alternating bit 1 is lost cannot be excluded. This corresponds to the path

$$
\ldots \ldots s_{i} \xrightarrow{\text { lost }} s_{i+1} \xrightarrow{\text { tmr_on }} s_{i+2} \xrightarrow{\text { timeout }} s_{i+3} \xrightarrow{\text { lost }} \ldots \ldots
$$

Suppose we impose the strong LTL fairness assumption

$$
\text { sfair }=\bigwedge_{b=0,1} \bigwedge_{\substack{k \\ k<\operatorname{cap}(c)}}(\square \diamond(\operatorname{send}(b) \wedge|c|=k) \rightarrow \square \diamond|c|=k+1)
$$

Here, $|c|=n$ stands for the atomic proposition that holds in the states $\langle\ell, \eta, \xi\rangle$ in which channel $c$ contains exactly $n$ elements, i.e., the length of the word $\xi(c)$ equals $n$. Thus, sfair describes (from the state-based point of view) that the loss of a transmitted message is not continuously possible. We now obtain

$$
A B P \models_{\text {sfair }} \square \diamond \text { start. }
$$

Note that it is essential to impose a strong fairness assumption on $\operatorname{send}(b)$; as this action is not continuously enabled, a weak fairness assumption is insufficient.

In Section 3.5 (page 126 ff .), fairness was introduced using sets of actions; e.g., an execution is unconditionally $A$-fair for a set of actions $A$, whenever each action $\alpha \in A$ occurs infinitely often. LTL-fairness, however, is defined on atomic propositions, i.e., from a state-based perspective. Is there any relationship between these two - at first sight, rather differentapproaches toward fairness?

The advantage of the action-based formulation of fairness assumptions is that many useful (and realizable) fairness assumptions can easily be expressed. Using the state-based
perspective, this may be less intuitive. For instance, the enabling of a process (or, more generally, of a certain action) is not necessarily a property that can be determined from the (atomic propositions in a) state. This discrepancy is, however, not always present.


Figure 5.11: Semaphore-based mutual exclusion algorithm.

## Example 5.29. State-Based vs. Action-Based Fairness

To exemplify this, consider the semaphore-based two-process mutual exclusion protocol (see Figure 5.11) together with the action-based strong-fairness assumption

$$
\mathcal{F}_{\text {strong }}=\left\{\left\{\text { enter }_{1}\right\},\left\{\text { enter }_{2}\right\}\right\} .
$$

Let us try to state the same constraint by means of a (state-based) LTL fairness assumption. Observe first that the action enter ${ }_{1}$ is executable if and only if process $P_{1}$ is in the local state wait ${ }_{1}$ and process $P_{2}$ is not in its critical section. Besides, on executing action enter ${ }_{1}$, process $P_{1}$ moves to its critical section. Thus, strong fairness for $\left\{e^{e n t e r} r_{1}\right\}$ can be described by the LTL fairness assumption:

$$
\text { sfair }_{1}=\square \diamond\left(\text { wait }_{1} \wedge \neg \text { crit }_{2}\right) \rightarrow \square \diamond \text { crit }_{1} .
$$

The assumption sfair $r_{2}$ is defined analogously. It now follows that sfair $=s f a i r_{1} \wedge s f a i r_{2}$ describes $\mathcal{F}_{\text {strong }}$.
$\mathcal{F}_{\text {strong }}$ requires each process to enter its critical section infinitely often when it infinitely often gets the opportunity to do so. This does not forbid a process to never leave its noncritical section. To avoid this unrealistic scenario, the weak fairness constraint

$$
\mathcal{F}_{\text {weak }}=\left\{\left\{r e q_{1}\right\},\left\{r e q_{2}\right\}\right\}
$$

requires that any process infinitely often requests to enter the critical section when it continuously is able to do so. This (action-based) weak fairness constraint can be formulated as (state-based) LTL fairness assumption in a similar way as above. Observe that the request action of $P_{i}$ is executable if and only if process $P_{i}$ is in the local state noncrit ${ }_{i}$. Weak fairness for $\left\{r e q_{i}\right\}$ thus correponds to the LTL fairness assumption:

$$
\text { wfair }_{i}=\diamond \square \text { noncrit }_{i} \rightarrow \square \diamond \text { wait }_{i} .
$$

Let fair $=$ sfair $\wedge$ wfair where wfair $=$ wfair $_{1} \wedge$ wfair $_{2}$. It then follows that

$$
T S_{S e m} \models_{\text {fair }} \square \diamond \text { crit }_{1} \wedge \square \diamond \text { crit }_{2} .
$$

In many cases, it is possible to replace action-based fairness by state-based LTL fairness assumptions. This, however, requires the possibility to deduce from the state label the possible enabled actions and the last executed action. It turns out that action-based fairness assumptions can always be "translated" into analogous LTL fairness assumptions. This works as follows. The intuition is to make a copy of each noninitial state $s$ such that it is recorded which action was executed to enter state $s$. Such copy is made for every possible action via which the state can be entered. The copied state $\langle s, \alpha\rangle$ indicates that state $s$ has been reached by performing $\alpha$ as last action.

Formally, this works as follows. For transition system $T S=(S, A c t, \rightarrow, I, A P, L)$ let

$$
T S^{\prime}=\left(S^{\prime}, A c t^{\prime}, \rightarrow^{\prime}, I^{\prime}, A P^{\prime}, L^{\prime}\right)
$$

where Act $t^{\prime}=\operatorname{Act} \uplus\{$ begin $\}, I^{\prime}=I \times\{$ begin $\}$ and $S^{\prime}=I^{\prime} \cup(S \times$ Act $)$. The transition relation in $T S^{\prime}$ is defined by the rules:

$$
\frac{s \xrightarrow{\alpha} s^{\prime}}{\langle s, \beta\rangle \xrightarrow{\alpha \prime}\left\langle s^{\prime}, \alpha\right\rangle} \quad \text { and } \quad \frac{s_{0} \xrightarrow{\alpha} s s_{0} \in I}{\left\langle s_{0}, \text { begin }\right\rangle \xrightarrow{\alpha \prime}\langle s, \alpha\rangle}
$$

The state labeling is defined as follows. Let

$$
A P^{\prime}=A P \cup\{\operatorname{enabled}(\alpha), \operatorname{taken}(\alpha) \mid \alpha \in \operatorname{Act}\}
$$

with the labeling function

$$
L^{\prime}(\langle s, \alpha\rangle)=L(s) \cup\{\operatorname{taken}(\alpha)\} \cup\{\operatorname{enabled}(\beta) \mid \beta \in \operatorname{Act}(s)\}
$$

for $\langle s, \alpha\rangle \in S \times$ Act and

$$
L^{\prime}\left(\left\langle s_{0}, \text { begin }\right\rangle\right)=L\left(s_{0}\right) \cup\left\{\operatorname{enabled}(\beta) \mid \beta \in \operatorname{Act}\left(s_{0}\right)\right\} .
$$

It can easily be established that

$$
\operatorname{Traces}_{A P}(T S)=\operatorname{Traces}_{A P}\left(T S^{\prime}\right)
$$

Strong fairness for a set of actions $A \subseteq$ Act can now be described by the strong LTL fairness assumption:

$$
\text { sfair }_{A}=\square \diamond \operatorname{enabled}(A) \rightarrow \square \diamond \operatorname{taken}(A)
$$

where

$$
\operatorname{enabled}(A)=\bigvee_{\alpha \in A} \operatorname{enabled}(\alpha) \quad \text { and } \quad \operatorname{taken}(A)=\bigvee_{\alpha \in A} \operatorname{taken}(\alpha)
$$

Unconditional and weak action-based fairness assumptions for $T S$ can be transformed into LTL fairness assumptions for $T S^{\prime}$ in a similar way. For action-based fairness assumption $\mathcal{F}$ for $T S$ and fair the corresponding LTL fairness assumption for $T S^{\prime}$, it follows that the set of fair traces coincides:

$$
\begin{aligned}
& \left\{\operatorname{trace}_{A P}(\pi) \mid \pi \in \operatorname{Paths}(T S), \pi \text { is } \mathcal{F} \text {-fair }\right\} \\
= & \left\{\operatorname{trace}_{A P}\left(\pi^{\prime}\right) \mid \pi^{\prime} \in \operatorname{Paths}\left(T S^{\prime}\right), \pi^{\prime} \models \text { fair }\right\} .
\end{aligned}
$$

 sitions in $A P$ are considered. In particular, for every LT property $P$ over $A P$,

$$
T S \models_{\mathcal{F}} P \text { iff } T S^{\prime} \models_{\text {fair }} P .
$$

Conversely, a (state-based) LTL fairness assumption cannot always be represented as action-based fairness assumption. This fact follows from the fact that strong or weak LTL fairness assumptions need not be realizable, while any action-based strong or weak fairness assumptions can be realized by a scheduler. In this sense, state-based LTL fairness assumptions are more general than action-based fairness assumptions.

The following theorem shows that the satisfaction relation $\models_{\text {fair }}$ as defined in Definition 5.26 has a strong relation to the usual satisfaction relation $\models$.

Theorem 5.30. Reduction of $\models_{\text {fair }}$ to $\models$
For transition system TS without terminal states, LTL formula $\varphi$, and LTL fairness assumption fair:

$$
T S \models_{\text {fair }} \varphi \quad \text { if and only if } \quad T S \models(\text { fair } \rightarrow \varphi) .
$$

Proof: $\Rightarrow$ : Assume $T S \models_{\text {fair }} \varphi$. Then, for any path $\pi \in \operatorname{Paths}(T S)$ either $\pi \models$ fair $\wedge \varphi$ or $\pi \models \neg$ fair. Thus, $\pi \models($ fair $\rightarrow \varphi)$, and consequently TS $\models($ fair $\rightarrow \varphi)$.
$\Leftarrow$ : by a similar reasoning.

## Example 5.31. On-The-Fly Garbage Collection

An important characteristic of pointer-manipulating programs is that certain parts of the dynamically changing data structure, such as a list or a tree, may become inaccessible. That is to say, this part of the data structure is not "reachable" by dereferencing one of the program variables. In order to be able to recycle these inaccessible memory cells, so-called garbage collecting algorithms are employed. These algorithms are key parts of operating systems, and play a major role in compilers. In this example, we focus on on-the-fly garbage collection algorithms. These algorithms attempt to identify and recycle inaccessible memory cells concurrently with the running programs that may manipulate the dynamically changing data structure. Typical requirements for such garbage collection algorithms are

Safety: An accessible (i.e., reachable) memory cell is never collected.
Liveness: Any unreachable storage cell is eventually collected.

The memory cells. The memory is considered to consist of a fixed number $N$ of memory cells. The number of memory cells is static, i.e., the dynamic allocation and deallocation of cells (as in the C-statement malloc) is not considered. The memory cells are organized in a directed graph, whose structure may change during the computation; this happens, e.g., when carrying out an assignment $v:=w$ where $v$ and $w$ point to a memory cell. For the sake of simplicity, it is assumed that each cell ( $=$ vertex) has at most one pointer (= edge) to another cell. Let son (i) denote the immediate successor of cell $i$. Cells that do not have an outgoing reference are equipped with a self-reference. Thus, each cell has exactly one outgoing edge in the graph representation. Vertices are numbered. There is a fixed set of root vertices. Root cells are never considered as garbage. A memory cell is reachable if it is reachable in the graph from a root vertex. Cells that are not reachable from a root vertex are garbage cells. The memory is partitioned into three fragments: the accessible cells (i.e., cells that are reachable by the running processes), the free cells (i.e., cells that are unused and that can be assigned to running processes), and the unreachable cells. As for the garbage collection algorithm, it is only of importance which cells are unreachable. The distinction between free and accessible cells is of no importance, and will be neglected in the remainder.

Modeling the garbage collector. The entire system is modeled as the parallel composition of the garbage collector (process Collector), and a process that models the behaviour of the running processes manipulating the shared data structure. Thus:

## Mutator || Collector

For simplicity, we abstain from a detailed transition system representation and describe the mutator and the collector by pseudocode algorithms. As the garbage collecting algorithm should work correctly for any arbitrary set of concurrently running processes, the mutator is modeled by means of a single process that can nondeterministically choose two arbitrary reachable cells $i$ and $k$ and change the current reference $i \rightarrow j$ into $i \rightarrow k$. This corresponds to the assignment "son $(i):=k$ ". Note that if $i$ was the only cell pointing to $j$, then after the assignment $\operatorname{son}(i):=k$, the memory cell $j$ has become garbage.

```
Algorithm 9 Mutator and garbage collector (naive version)
                                    (* mutator *)
    while true do
        Nondeterministically choose two reachable nodes \(i\) and \(k\);
        \(\operatorname{son}(i):=k\)
    od
                                    (* garbage collector \({ }^{*}\) )
    while true do
    label all reachable cells; (* e.g., by depth-first search *)
    for all cells \(i\) do
        if \(i\) is not labeled then collect \(i\)
        od
        add the collected cells to the list of free memory cells
    od
```

Now let us consider the garbage collector. A naive technique could be based on a DFS or BFS which labels cells and, subsequently, collects all unlabeled cells as garbage (see Algorithm 9). The collected cells are added to the list of free memory cells, and thus are turned into reachable cells. This simple idea works correctly when the actions of the mutator are not interlocked with those of the garbage collector. As an on-the-fly algorithm, however, this naive idea fails due to the possible interleavings between the mutator and the collector. This is illustrated in Figure 5.12 where six memory configurations are depicted. Circles denote memory cells and edges represent the pointer structure. Nonshaded cells are not visited, gray ones are visited, and cells with a thick border have been completely processed. Due to the changes made by the mutator during the garbage collection phase, one of the memory cells remains unlabeled although it is accessible. The collector would now collect the unlabeled but accessible cell.


Figure 5.12: Failure of the naive garbage collector.

An alternative is to use a labeling algorithm that successively passes through all cells and labels their sons if the appropriate cell is labeled itself. Initially, only root cells are labeled. This technique is iterated as long as the number of labeled cells does not change anymore. The auxiliary variables $M$ and $M_{\text {old }}$ are used to this purpose. $M$ counts the number of labeled cells in the current stage of labeling. $M_{\text {old }}$ stands for the number of labeled cells in the previous stage of labeling. Additionally, the mutator supports the labeling process by labeling cell $k$ as soon as a reference is redirected to $k$ (see Algorithm 10). It is not difficult to show that the participation of the mutator is necessary to obtain a correct garbage collection algorithm. If it does not participate, there can always be found some interference pattern between the mutator and the collector such that reachable cells are made inaccessible by the mutator, causing a flaw in the garbage collection. Figure 5.13 demonstrates the functioning of the collector for four linearly linked reachable cells, where it is assumed that the labeling steps of the collector are not interrupted by the mutator.

The shading represents the labeling by the collector. The thick border of a node indicates that this node was considered in the for loop. Thus, in the last step, all four nodes were processed in the for loop. Therefore, the result of the first iteration is $M_{\text {old }}=1$ (number of root cells) and $M=3$, since exactly three cells were labeled. The labels arising at the end of the subsequent iterations are indicated in Figure 5.14.

Figure 5.15 shows an example that indicates that the mutator has to label cells, as otherwise it cannot be ensured that the collector identifies all reachable cells. The leftmost figure indicates the starting memory configuration. The situation after the collector has labeled the root cell and has processed the top left node in the FOR loop is depicted in

```
Algorithm 10 Ben Ari's on-the-fly garbage collection algorithm
                                    (* mutator \({ }^{*}\) )
    while true do
        let \(i\) and \(k\) be reachable memory cells;
        label \(k\);
        \(\operatorname{son}(i):=k\)
    od
        (* garbage collector \({ }^{*}\) )
    while true do
        label all root cells;
        \(M:=\) number of root cells;
        repeat
            \(M_{\text {old }}:=M\);
            for all node \(i\) do
                if \(i\) is labeled then
                if \(\operatorname{son}(i)\) is unlabeled then label \(\operatorname{son}(i) ; M:=M+1 ; \mathbf{f i}\)
                fi
            od
        until \(M=M_{\text {old }}\);
    for all cell \(i\) do
        if \(i\) is labeled then delete label for cell \(i\)
        else collect cell \(i\)
        (* cell \(i\) is garbage \({ }^{*}\) )
        fi
    od
        add the collected cells to the list of free cells
    od
```



Figure 5.13: Example of Ben Ari's on-the-fly garbage collector (one iteration).


Figure 5.14: Example of Ben-Ari's on-the-fly garbage collector (all iterations).
the next figure. The third and fourth figures show a possible modification of the pointer structure by the mutator. The collector then would investigate the upper cell on the right and label it. Since this cell does not have an unlabeled successor, the first round of the collector is finished. As the obtained memory configuration is symmetric to the initial one, the whole procedure might be repeated, which leads to the initial configuration of Figure 5.15. In the second round, the collector also just labels the two upper cells. Since the number of labeled cells has not increased and assuming that the mutator does not label the lower cell, this cell would be wrongly regarded as garbage (and thus be collected).

The examples illustrate the functioning of the algorithm but are by no means meant as proof of correctness. Due to the enormous size of the state space and the extremely high degree of nondeterminism, the analysis of the garbage collection algorithm is extremely difficult. Note that the states of the transition system, which results from the parallel composition of mutator and collector, consist of control components plus a representation of the current labels and of another component that gives information about the current


Figure 5.15: Labeling memory cells by the mutator is necessary.
memory configuration. For $N$ nodes, there exist $2^{N}$ possible labels and $N^{N}$ possible graphs. The high degree of nondeterminism results from the mutator that can change an arbitrary pointer. It is not useful to restrict the mutator (and thus reducing the degree of nondeterminism), since this would impose restrictions on the applicability of the garbage collecting algorithm.

To conclude, we will formalize the requirements on the concurrent garbage collection algorithm. To that end, let atomic proposition collect $(i)$ hold for cell $i$ in a state if and only if cell $i$ has been collected in that state. Similarly, accessible $(i)$ holds in those states in which cell $i$ is reachable from a root cell. Given these atomic propositions, the safety property "an accessible memory cell is never collected" can be specified as

$$
\bigwedge_{0<i \leqslant N} \square(\operatorname{collect}(i) \rightarrow \neg \operatorname{accessible}(i)) .
$$

The liveness property "any unreachable storage cell is eventually collected" is described as LTL formula

$$
\bigwedge_{0<i \leqslant N} \square(\neg \operatorname{accessible}(i) \rightarrow \diamond \operatorname{collect}(i))
$$

It turns out that Ben Ari's concurrent garbage collection algorithm satisfies the safety property, but refutes the liveness property. The latter happens, e.g., in the pathological case where only the mutator acts infinitely often. To rule out this unrealistic behavior, weak process fairness can be imposed.

### 5.2 Automata-Based LTL Model Checking

In this section we address the model-checking problem for LTL. The starting point is a finite transition system $T S$ and an LTL formula $\varphi$ that formalizes a requirement on $T S$. The problem is to check whether $T S \models \varphi$. If $\varphi$ is refuted, an error trace needs to be provided for debugging purposes. The considerations in Section 2.1 show that transition
systems are typically huge. Therefore, a manual proof of $T S \models \varphi$ is extremely difficult. Instead, verification tools are desirable, which enable a fully automated analysis of the transition system.

In general, not just a single requirement but several requirements are relevant. These requirements can be represented by separate formulae, such as $\varphi_{1}, \ldots, \varphi_{k}$, and be combined into $\varphi_{1} \wedge \ldots \wedge \varphi_{k}$ to obtain a specification of all requirements. Alternatively, the requirements $\varphi_{i}$ can be treated separately. This is often more efficient than considering them together. Moreover, the decomposition of the entire requirement specification into several requirements is advised if errors are to be expected or if the validity of $\varphi_{i}$ is known by a prior analysis.

An LTL model-checking algorithm is a decision procedure which for a transition system $T S$ and LTL formula $\varphi$ returns the answers "yes" if $T S \models \varphi$, and "no" (plus a counterexample) if $T S \nLeftarrow \varphi$. The counterexample consists of an appropriate finite prefix of an infinite path in $T S$ where $\varphi$ does not hold. Theorem 5.30 shows that special measures to handle fairness assumptions are unnecessary as fairness assumptions can be encoded in the LTL formula to be checked. (For reasons of efficiency, however, it is advised to use special algorithms to treat fairness assumptions.)

Throughout this section, TS is assumed to be finite and to have no terminal states. The model-checking algorithm presented in the following is based on the automata-based approach as originally suggested by Vardi and Wolper (1986). This approach is based on the fact that each LTL formula $\varphi$ can be represented by a nondeterministic Büchi automaton (NBA). The basic idea is to try to disprove $T S \models \varphi$ by "looking" for a path $\pi$ in $T S$ with $\pi \models \neg \varphi$. If such a path is found, a prefix of $\pi$ is returned as error trace. If no such path is encountered, it is concluded that $T S \models \varphi$.

The essential steps of the model-checking algorithm as summarized in Algorithm 11 and Figure 5.16, rely on the following observations:

$$
\begin{array}{lll}
T S \models \varphi & \text { iff } & \operatorname{Traces}(T S) \subseteq \operatorname{Words}(\varphi) \\
& \text { iff } & \operatorname{Traces}(T S) \cap\left(\left(2^{A P}\right)^{\omega} \backslash \operatorname{Words}(\varphi)\right)=\varnothing \\
& \text { iff } & \operatorname{Traces}(T S) \cap \operatorname{Words}(\neg \varphi)=\varnothing
\end{array}
$$

Hence, for NBA $\mathcal{A}$ with $\mathcal{L}_{\omega}(\mathcal{A})=\operatorname{Words}(\neg \varphi)$ we have

$$
T S \models \varphi \quad \text { if and only if } \quad \operatorname{Traces}(T S) \cap \mathcal{L}_{\omega}(\mathcal{A})=\varnothing .
$$

Thus, to check whether $\varphi$ holds for $T S$ one first constructs an NBA for the negation of the input formula $\varphi$ (representing the "bad behaviors") and then applies the techniques explained in Chapter 4 for the intersection problem.

```
Algorithm 11 Automaton-based LTL model checking
Input: finite transition system TS and LTL formula \(\varphi\) (both over AP)
Output:"yes" if \(T S=\varphi\); otherwise, "no" plus a counterexample
    Construct an NBA \(\mathcal{A}_{\neg \varphi}\) such that \(\mathcal{L}_{\omega}\left(\mathcal{A}_{\neg \varphi}\right)=\operatorname{Words}(\neg \varphi)\)
    Construct the product transition system \(T S \otimes \mathcal{A}\)
    if there exists a path \(\pi\) in \(T S \otimes \mathcal{A}\) satisfying the accepting condition of \(\mathcal{A}\) then
        return "no" and an expressive prefix of \(\pi\)
    else
        return "yes"
    fi
```

It remains to explain how a given LTL formula can be represented by an NBA and how such an NBA can be constructed algorithmically. First observe that for the LTL formula $\varphi$, the LTL semantics provided in Definition 5.6 on page 235 yields a language $\operatorname{Words}(\varphi) \subseteq$ $\left(2^{A P}\right)^{\omega}$. Thus, the alphabet of NBA for LTL formulae is $\Sigma=2^{A P}$. The next step is to show that $\operatorname{Words}(\varphi)$ is $\omega$-regular, and hence, representable by a nondeterministic Büchi automaton.

## Example 5.32. NBA for LTL Formulae

Before treating the details of transforming an LTL formula into an NBA, we provide some examples. As in Chapter 4, propositional logic formulae are used (instead of the set notations) for a symbolic representation of the edges of an NBA. These formulae are built by the symbols $a \in A P$, the constant true, and the Boolean connectors, and thus they can be interpreted over sets of atomic propositions, i.e., the elements $A \in \Sigma=2^{A P}$. For instance, if $A P=\{a, b\}$ then $q \xrightarrow{a \vee b} q^{\prime}$ is a short notation for the three transitions:

$$
q \xrightarrow{\{a\}} q^{\prime}, q \xrightarrow{\{b\}} q^{\prime}, \text { and } q \xrightarrow{\{a, b\}} q^{\prime} .
$$

The language of all words $\sigma=A_{0} A_{1} \ldots \in 2^{A P}$ satisfying the LTL formula $\square \diamond$ green ("infinitely often green") is accepted by the NBA $\mathcal{A}$ shown in Figure 5.17. Here, $A P$ is a set of atomic propositions containing green. Note that $\mathcal{A}$ is in the accept state $q_{1}$ if and only if the last consumed symbol (the last set $A_{i}$ of the input word $A_{0} A_{1} A_{2} \ldots \in$ $\left.\left(2^{A P}\right)^{\omega}\right)$ contains the propositional symbol green. Therefore, the accepted language $\mathcal{L}_{\omega}(\mathcal{A})$ is exactly the set of all infinite words $A_{0} A_{1} A_{2} \ldots$ with infinitely many indices $i$ where green $\in A_{i}$. Thus,

$$
\mathcal{L}_{\omega}(\mathcal{A})=\operatorname{Words}(\square \diamond \text { green }) .
$$

The accepting run that generates the word $\sigma=\{$ green $\} \varnothing\{$ green $\} \varnothing \ldots$ is $\left(q_{0} q_{1}\right)^{\omega}$.


Figure 5.16: Overview of LTL model checking.


Figure 5.17: NBA for $\square \diamond$ green.

As a second example consider the liveness property: "whenever event $a$ occurs, event $b$ will eventually occur". For example, the property given by the LTL formula $\square$ (request $\rightarrow$ $\diamond$ response) is of this form. An associated NBA over the alphabet $2^{\{a, b\}}$ where $a=$ request and $b=$ response is shown in Figure 5.18.

The automata in Figures 5.17 and 5.18 are deterministic, i.e., they have exactly one run for each input word. To represent temporal properties like "eventually forever (from some moment on)", the concept of nondeterminism is, however, necessary.

The NBA $\mathcal{A}$ shown in Figure 5.19 accepts the language $\operatorname{Words}(\diamond \square a)$. Here, $A P \supseteq\{a\}$ and $\Sigma=2^{A P}$; see also Example 4.51 (page 191). Intuitively, the NBA $\mathcal{A}$ nondeterministically decides (by means of an omniscient oracle) when $a$ continuously holds. Note that state $q_{2}$ may be omitted, as there is no accepting run beginning in $q_{2}$. (The reader should bear in mind that DBA and NBA are not equally expressive; see Section 4.3.3 on page 188.)


Figure 5.18: NBA for $\square(a \rightarrow \diamond b)$.


Figure 5.19: NBA for $\diamond \square a$.

A key ingredient to the model-checking algorithm for LTL is the construction of an NBA $\mathcal{A}$ satisfying

$$
\mathcal{L}_{\omega}(\mathcal{A})=\operatorname{Words}(\varphi)
$$

for the LTL formula $\varphi$. In order to do so, first a generalized NBA is constructed for $\varphi$, which subsequently is transformed into an equivalent NBA. For the latter step we employ the recipe as provided in Theorem 4.56 on page 195. For the sake of convenience we recall the definition of generalized NBA; see also Definition 4.52 on page 193.

## Definition 5.33. Generalized NBA (GNBA)

A generalized NBA is a tuple $\mathcal{G}=\left(Q, \Sigma, \delta, Q_{0}, \mathcal{F}\right)$ where $Q, \Sigma, \delta, Q_{0}$ are defined as for NBA (i.e., $Q$ is a finite state space, $\Sigma$ an alphabet, $Q_{0} \subseteq Q$ the set of initial states, and $\delta: Q \times \Sigma \rightarrow 2^{Q}$ the transition relation) and $\mathcal{F}$ is a (possibly empty) subset of $2^{Q}$. The elements of $\mathcal{F}$ are called acceptance sets. The accepted language $\mathcal{L}_{\omega}(\mathcal{G})$ consists of all infinite words in $\left(2^{A P}\right)^{\omega}$ that have at least one infinite run $q_{0} q_{1} q_{2} \ldots$ in $\mathcal{G}$ such that for each acceptance set $F \in \mathcal{F}$ there are infinitely many indices $i$ with $q_{i} \in F$.

A GNBA for which $\mathcal{F}$ is a singleton set can be regarded as an NBA. If the set $\mathcal{F}$ of acceptance sets in $\mathcal{G}$ is empty, the language $\mathcal{L}_{\omega}(\mathcal{G})$ consists of all infinite words that have an infinite run in $\mathcal{G}$. Hence, if $\mathcal{F}=\varnothing$, then $\mathcal{G}$ can be viewed as an NBA for which all states are accepting.

Let us consider how to construct a GNBA over the alphabet $2^{A P}$ for a given LTL formula $\varphi$ (over $A P$ ), i.e., a GNBA $\mathcal{G}_{\varphi}$ with $\mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)=\operatorname{Words}(\varphi)$. Assume $\varphi$ only contains the operators $\wedge, \neg, \bigcirc$ and $\mathbb{U}$, i.e., the derived operators $\vee, \rightarrow, \diamond, \square, W$, and so on are assumed to be expressed in terms of the basic operators. Since the special case $\varphi=$ true is trivial, it may be assumed that $\varphi \neq$ true.

The basic idea for the construction of $\mathcal{G}_{\varphi}$ is as follows. Let $\sigma=A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi)$. The sets $A_{i} \subseteq A P$ are expanded by subformulae $\psi$ of $\varphi$ such that an infinite word $\bar{\sigma}=$ $B_{0} B_{1} B_{2} \ldots$ with the following property arises:

$$
\psi \in B_{i} \quad \text { if and only if } \quad \underbrace{A_{i} A_{i+1} A_{i+2} \ldots}_{\sigma^{i}}=\psi .
$$

For technical reasons, the subformulae $\psi$ of $\varphi$ are considered as well as their negation $\neg \psi$. Consider, e.g.,

$$
\varphi=a \cup(\neg a \wedge b) \quad \text { and } \quad \sigma=\{a\}\{a, b\}\{b\} \ldots
$$

In this case, $B_{i}$ is a subset of the set of formulae

$$
\underbrace{\{a, b, \neg a, \neg a \wedge b, \varphi\}}_{\text {subformulae of } \varphi} \cup \underbrace{\{\neg b, \neg(\neg a \wedge b), \neg \varphi\}}_{\text {their negation }} .
$$

The set $A_{0}=\{a\}$ is extended with the formulae $\neg b, \neg(\neg a \wedge b)$, and $\varphi$, since all these formulae hold in $\sigma^{0}=\sigma$, and all other subformulae in the above set are refuted by $\sigma$. We thus obtain

$$
B_{0}=\{a, \neg b, \neg(\neg a \wedge b), \varphi\}
$$

The set $A_{1}=\{a, b\}$ is extended with $\neg(\neg a \wedge b)$ and $\varphi$, as these are the only subformulae in the above set that hold in $\sigma^{1}=\{a, b\}\{b\} \ldots$. The $A_{2}=\{b\}$ is extended with $\neg a$, $\neg a \wedge b$ and $\varphi$ as they hold in $\sigma^{2}=\{b\} \ldots$. This yields a word of the form:

$$
\bar{\sigma}=\{a, \neg b, \neg(\neg a \wedge b), \varphi\}\{a, b, \neg(\neg a \wedge b), \varphi\}\{\neg a, b, \neg a \wedge b, \varphi\} \ldots
$$

As $\sigma$ is infinite, this procedure is of course not effective. The example is just meant to explain the intuition behind the construction of the states in the GNBA.

The GNBA $\mathcal{G}_{\varphi}$ is constructed such that the sets $B_{i}$ constitute its states. Moreover, the construction ensures that $\bar{\sigma}=B_{0} B_{1} B_{2} \ldots$ is a run for $\sigma=A_{0} A_{1} A_{2} \ldots$ in $\mathcal{G}_{\varphi}$. The accepting conditions for $\mathcal{G}_{\varphi}$ are chosen such that the run $\bar{\sigma}$ is accepting if and only if $\sigma \models \varphi$. Thus, we have to encode the meaning of the logical operators into the states, transitions, and acceptance sets of $\mathcal{G}_{\varphi}$. The meaning of propositional logic operators $\wedge$, $\neg$, and the constant true will be encoded in the states by requiring consistent formula sets $B_{i}$. The semantics of the next-step operator relies on a nonlocal condition and will be encoded in the transition relation. The meaning of the until operator is split according to the expansion law into local conditions (encoded in the states) and a next-step condition (encoded in the transitions). Since the expansion law does not provide a full characterization of until, a further condition is imposed which expresses the fact that the meaning of until yields the least solution of the expansion law (see Lemma 5.18 on page 251). This will be encoded by the acceptance sets of $\mathcal{G}_{\varphi}$.

As explained above, the formula sets are subsets of subformulae of $\varphi$ and their negation.

## Definition 5.34. Closure of $\varphi$

The closure of LTL formula $\varphi$ is the set closure $(\varphi)$ consisting of all subformulae $\psi$ of $\varphi$ and their negation $\neg \psi$ (where $\psi$ and $\neg \neg \psi$ are identified).

For instance, for $\varphi=a \cup(\neg a \wedge b)$, the set closure $(\varphi)$ consists of the formulae

$$
a, b, \neg a, \neg b, \neg a \wedge b, \neg(\neg a \wedge b), \varphi, \neg \varphi .
$$

It is not difficult to assess that $|\operatorname{closure}(\varphi)| \in \mathcal{O}(|\varphi|)$. A set of formulae $B \subseteq \operatorname{closure}(\varphi)$ is called elementary if $B$ is the set of all formulae $\psi \in \operatorname{closure}(\varphi)$ with $\pi \models \psi$ for a path $\pi$. For this, $B$ should not contain propositional logic contradictions and it must be locally consistent with respect to the until operator. Since for any path $\pi$ and formula $\psi$, either $\pi \models \psi$ or $\pi \models \neg \psi$, it is additionally required that elementary sets of formulae are maximal.
The precise definition of these three conditions is provided in Figure 5.20 on page 277.

## Definition 5.35. Elementary Sets of Formulae

$B \subseteq \operatorname{closure}(\varphi)$ is elementary if it is consistent with respect to propositional logic, maximal, and locally consistent with respect to the until operator.

The requirements for local consistency result from the expansion law

$$
\varphi_{1} \cup \varphi_{2} \equiv \varphi_{2} \vee\left(\varphi_{1} \wedge \bigcirc\left(\varphi_{1} \cup \varphi_{2}\right)\right)
$$

Due to the required maximality and propositional logic consistency, we have

$$
\psi \in \text { Bif and only if } \quad \neg \psi \notin B
$$

for all elementary sets $B$ and subformulae $\psi$ of $\varphi$. Further, due to maximality and local consistency, we have

$$
\varphi_{1}, \varphi_{2} \notin B \quad \text { implies } \quad \varphi_{1} \cup \varphi_{2} \notin B .
$$

Hence, if $\varphi_{1}, \varphi_{2} \notin B$ then $\left\{\neg \varphi_{1}, \neg \varphi_{2}, \neg\left(\varphi_{1} \cup \varphi_{2}\right)\right\} \subseteq B$; here, it is assumed that $\varphi_{1} \cup \varphi_{2}$ is a subformula of $\varphi$.

## Example 5.36. Elementary Sets of Formulae

Let $\varphi=a \cup(\neg a \wedge b)$. The set $B=\{a, b, \varphi\} \subseteq \operatorname{closure}(\varphi)$ is consistent with respect to propositional logic and locally consistent with respect to the until operator. It is, however, not maximal, since for $\neg a \wedge b \in \operatorname{closure}(\varphi)$ :

$$
\neg a \wedge b \notin B \quad \text { and } \quad \neg(\neg a \wedge b) \notin B
$$

1. $B$ is consistent with respect to propositional logic, i.e., for all $\varphi_{1} \wedge \varphi_{2}, \psi \in \operatorname{closure}(\varphi):$

- $\varphi_{1} \wedge \varphi_{2} \in B \Leftrightarrow \varphi_{1} \in B$ and $\varphi_{2} \in B$
- $\psi \in B \Rightarrow \neg \psi \notin B$
- true $\in \operatorname{closure}(\varphi) \Rightarrow$ true $\in B$.

2. $B$ is locally consistent with respect to the until operator, i.e., for all $\varphi_{1} \cup \varphi_{2} \in \operatorname{closure}(\varphi)$ :

- $\varphi_{2} \in B \Rightarrow \varphi_{1} \cup \varphi_{2} \in B$
- $\varphi_{1} \cup \varphi_{2} \in B$ and $\varphi_{2} \notin B \Rightarrow \varphi_{1} \in B$.

3. $B$ is maximal, i.e., for all $\psi \in \operatorname{closure}(\varphi)$ :

- $\psi \notin B \Rightarrow \neg \psi \in B$.

Figure 5.20: Properties of elementary sets of formulae.
The set of formulae $\{a, b, \neg a \wedge b, \varphi\}$ contains the propositional logic "contradiction" $a$ and $\neg a \wedge b$ and therefore is not elementary. The set

$$
\{\neg a, \neg b, \neg(\neg a \wedge b), \varphi\}
$$

is consistent with respect to propositional logic but contains a local inconsistency with respect to the until operator U , since $a \cup(\neg a \wedge b) \in B$ and $\neg a \wedge b \notin B$, but $a \notin B$. This means that

$$
\pi \models \neg a, \quad \pi \models \neg(\neg a \wedge b), \quad \text { and } \quad \pi \models \varphi
$$

are impossible for any path $\pi$.

The following sets are elementary:

$$
\begin{aligned}
& B_{1}=\{\quad a, \quad b, \quad \neg(\neg a \wedge b), \quad \varphi\}, \\
& B_{2}=\{\quad a, \quad b, \quad \neg(\neg a \wedge b), \quad \neg \varphi\}, \\
& B_{3}=\{a, \neg b, \neg(\neg a \wedge b), \quad \varphi\}, \\
& B_{4}=\{\quad a, \neg b, \neg(\neg a \wedge b), \neg \varphi\}, \\
& B_{5}=\{\neg a, \neg b, \neg(\neg a \wedge b), \neg \varphi\}, \\
& B_{6}=\{\neg a, \quad b, \quad \neg a \wedge b, \quad \varphi\} .
\end{aligned}
$$

The proof of the following theorem shows how to construct for an arbitrary LTL formula $\varphi$ a GNBA $\mathcal{G}_{\varphi}$ with $\mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)=\operatorname{Words}(\varphi)$. This construction is one of the initial steps of the

LTL model checking algorithm; see Figure 5.16 (page 273). Subsequently, the resulting GNBA $\mathcal{G}_{\varphi}$ is transformed into an NBA $\mathcal{A}_{\varphi}$ by means of the technique indicated in the proof of Theorem 4.56 (page 195).

## Theorem 5.37. GNBA for LTL Formula

For any LTL formula $\varphi$ (over AP) there exists a GNBA $\mathcal{G}_{\varphi}$ over the alphabet $2^{A P}$ such that
(a) $\operatorname{Words}(\varphi)=\mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)$.
(b) $\mathcal{G}_{\varphi}$ can be constructed in time and space $2^{\mathcal{O}(|\varphi|)}$.
(c) The number of accepting sets of $\mathcal{G}_{\varphi}$ is bounded above by $\mathcal{O}(|\varphi|)$.

Proof: Let $\varphi$ be an LTL formula over $A P$. Let $\mathcal{G}_{\varphi}=\left(Q, 2^{A P}, \delta, Q_{0}, \mathcal{F}\right)$ where

- $Q$ is the set of all elementary sets of formulae $B \subseteq \operatorname{closure}(\varphi)$,
- $Q_{0}=\{B \in Q \mid \varphi \in B\}$,
- $\mathcal{F}=\left\{F_{\varphi_{1} \cup \varphi_{2}} \mid \varphi_{1} \cup \varphi_{2} \in \operatorname{closure}(\varphi)\right\}$ where

$$
F_{\varphi_{1} \cup \varphi_{2}}=\left\{B \in Q \mid \varphi_{1} \cup \varphi_{2} \notin B \text { or } \varphi_{2} \in B\right\}
$$

The transition relation $\delta: Q \times 2^{A P} \rightarrow 2^{Q}$ is given by:

- If $A \neq B \cap A P$, then $\delta(B, A)=\varnothing$.
- If $A=B \cap A P$, then $\delta(B, A)$ is the set of all elementary sets of formulae $B^{\prime}$ satisfying
(i) for every $\bigcirc \psi \in \operatorname{closure}(\varphi): \bigcirc \psi \in B \Leftrightarrow \psi \in B^{\prime}$, and
(ii) for every $\varphi_{1} \cup \varphi_{2} \in \operatorname{closure}(\varphi)$ :

$$
\varphi_{1} \cup \varphi_{2} \in B \quad \Leftrightarrow \quad\left(\varphi_{2} \in B \vee\left(\varphi_{1} \in B \wedge \varphi_{1} \cup \varphi_{2} \in B^{\prime}\right)\right)
$$

The constraints (i) and (ii) reflect the semantics of the next-step and the until operator, respectively. Rule (ii) is justified by the expansion law:

$$
\varphi_{1} \cup \varphi_{2} \equiv \varphi_{2} \vee\left(\varphi_{1} \wedge \bigcirc\left(\varphi_{1} \cup \varphi_{2}\right)\right)
$$

To model the semantics of $U$, an acceptance set $F_{\psi}$ is introduced for every subformula $\psi=\varphi_{1} \mathrm{U} \varphi_{2}$ of $\varphi$. The underlying idea is to ensure that in every run $B_{0} B_{1} B_{2} \ldots$ for which $\psi \in B_{0}$, we have $\varphi_{2} \in B_{j}$ (for some $j \geqslant 0$ ) and $\varphi_{1} \in B_{i}$ for all $i<j$. The requirement that a word $\sigma$ satisfies $\varphi_{1} \cup \varphi_{2}$ only if $\varphi_{2}$ will actually eventually become true is ensured by the accepting set $F_{\varphi_{1} \cup \varphi_{2}}$.

Let us first consider the claim (b). States in the GNBA $\mathcal{G}_{\varphi}$ are elementary sets of formulae in $\operatorname{closure}(\varphi)$. Let $\operatorname{subf}(\varphi)$ denote the set of all subformulae of $\varphi$. The number of states in $\mathcal{G}_{\varphi}$ is bounded by $2^{|\operatorname{subf}(\varphi)|}$, the number of possible formula sets. (The elementary formula sets $B$ can be represented by bit vectors containing a single bit per subformula $\psi$ of $\varphi$ which indicates whether $\psi$ or $\neg \psi$ belongs to $B$.) As $|\operatorname{subf}(\varphi)| \leqslant 2 \cdot|\varphi|$, the number of states in the GNBA $\mathcal{G}_{\varphi}$ is bounded by $2^{\mathcal{O}(|\varphi|)}$. Claim (c) follows directly from the fact that the number of accept sets is equal to the number of until-subformulae in $\varphi$.

It remains to show that (a) $\mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)=\operatorname{Words}(\varphi)$. We prove set inclusion in both directions.
$\supseteq$ : Let $\sigma=A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\varphi)$. Then, $\sigma \in\left(2^{A P}\right)^{\omega}$ and $\sigma \models \varphi$. The elementary set $B_{i}$ of formulae is defined as follows:

$$
\begin{equation*}
B_{i}=\left\{\psi \in \operatorname{closure}(\varphi) \mid A_{i} A_{i+1} \ldots \models \psi\right\} \tag{5.1}
\end{equation*}
$$

Obviously, $B_{i}$ is an elementary set of formulae, i.e., $B_{i} \in Q$. We now prove that $B_{0} B_{1} B_{2} \ldots$ is an accepting run for $\sigma$. Observe that $B_{i+1} \in \delta\left(B_{i}, A_{i}\right)$ for all $i \geqslant 0$, since for all $i$ :

- $A_{i}=B_{i} \cap A P$
- for $\bigcirc \psi \in \operatorname{closure}(\varphi)$ :

$$
\begin{array}{lr} 
& \bigcirc \psi \in B_{i} \\
\text { iff } & \\
& A_{i} A_{i+1} \ldots \models \bigcirc \psi
\end{array} \quad\left(* \text { equation (5.1) for } B_{i}^{*}\right)
$$

- for $\varphi_{1} \mathrm{U} \varphi_{2} \in \operatorname{closure}(\varphi)$ :

$$
\begin{array}{lc}
\text { iff } & \begin{array}{l}
\varphi_{1} \cup \varphi_{2} \in B_{i} \\
A_{i} A_{i+1} \ldots \vDash \varphi_{1} \cup \varphi_{2}
\end{array} \\
\text { iff } & \left(* \text { equation (5.1) for } B_{i}{ }^{*}\right) \\
& \begin{array}{l}
A_{i} A_{i+1} \ldots=\varphi_{2} \text { or } \\
A_{i} A_{i+1} \ldots \vDash \varphi_{1} \text { and }\left(A_{i+1} A_{i+2} \ldots \vDash \varphi_{1} \cup \varphi_{2}\right.
\end{array} \\
\text { iff } & \left(* \text { semantics of until }{ }^{*}\right) \\
& \left.\quad \text { equation (5.1) for } B_{i} \text { and } B_{i+1}{ }^{*}\right) \\
\varphi_{2} \in B_{i} \text { or }\left(\varphi_{1} \in B_{i} \text { and } \varphi_{1} \cup \varphi_{2} \in B_{i+1}\right)
\end{array}
$$

This shows that $B_{0} B_{1} B_{2} \ldots$ is a run of $\mathcal{G}_{\varphi}$. It remains to prove that this run is accepting, i.e., for each subformula $\varphi_{1, j} \cup \varphi_{2, j}$ in closure $(\varphi), B_{i} \in F_{j}$ for infinitely many $i$. By contraposition. Assume there are finitely many $i$ such that $B_{i} \in F_{j}$. We have:

$$
B_{i} \notin F_{j}=F_{\varphi_{1, j}} \cup \varphi_{2, j} \Rightarrow \varphi_{1, j} \cup \varphi_{2, j} \in B_{i} \text { and } \varphi_{2, j} \notin B_{i} .
$$

As $B_{i}=\left\{\psi \in \operatorname{closure}(\varphi) \mid A_{i} A_{i+1} \ldots \models \psi\right\}$, it follows that if $B_{i} \notin F_{j}$, then:

$$
A_{i} A_{i+1} \ldots \models \varphi_{1, j} \cup \varphi_{2, j} \quad \text { and } \quad A_{i} A_{i+1} \ldots \not \vDash \varphi_{2, j} .
$$

Thus, $A_{k} A_{k+1} \ldots \vDash \varphi_{2, j}$ for some $k>i$. By definition of the formula sets $B_{i}$, it then follows that $\varphi_{2, j} \in B_{k}$, and by definition of $F_{j}, B_{k} \in F_{j}$. Thus, $B_{i} \in F_{j}$ for finitely many $i$, then $B_{k} \in F_{j}$ for infinitely many $k$. Contradiction.

Thus, $B_{0} B_{1} B_{2} \ldots$ is an accepting run of $\mathcal{G}_{\varphi}$, and hence, $A_{0} A_{1} A_{2} \ldots \in \mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)$.
$\subseteq$ Let $\sigma=A_{0} A_{1} A_{2} \ldots \in \mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)$, i.e., there is an accepting run $B_{0} B_{1} B_{2} \ldots$, say, for $\sigma$ in $\mathcal{G}_{\varphi}$. Since

$$
\delta(B, A)=\varnothing \quad \text { for all pairs }(B, A) \text { with } A \neq B \cap A P
$$

it follows that $A_{i}=B_{i} \cap A P$ for $i \geqslant 0$. Thus

$$
\sigma=\left(B_{0} \cap A P\right)\left(B_{1} \cap A P\right)\left(B_{2} \cap A P\right) \ldots
$$

Our proof obligation now becomes $\left(B_{0} \cap A P\right)\left(B_{1} \cap A P\right)\left(B_{2} \cap A P\right) \ldots \vDash \varphi$. We prove the following more general proposition:

For $B_{0} B_{1} B_{2} \ldots$ a sequence with $B_{i} \in Q$ satisfying
(i) for all $i \geqslant 0: B_{i+1} \in \delta\left(B_{i}, A_{i}\right)$, and
(ii) for all $F \in \mathcal{F}: \exists j \geqslant 0 . B_{j} \in F$,
we have for all $\psi \in \operatorname{closure}(\varphi)$ :

$$
\psi \in B_{0} \quad \Leftrightarrow \quad A_{0} A_{1} A_{2} \ldots \models \psi
$$

The proof of this claim is by structural induction on the structure of $\psi$.
Base case: The statement for $\psi=$ true or $\psi=a$ with $a \in A P$ follows directly from equation (5.1) and the definition of closure.

Induction step: Based on the induction hypothesis that the claim holds for $\psi^{\prime}, \varphi_{1}, \varphi_{2} \in$ $\operatorname{closure}(\varphi)$, it is proven that for the formulae

$$
\psi=\bigcirc \psi^{\prime}, \psi=\neg \psi^{\prime}, \psi=\varphi_{1} \wedge \varphi_{2} \text { and } \psi=\varphi_{1} \cup \varphi_{2}
$$

the claim also holds. We provide the detailed proof for $\psi=\varphi_{1} \cup \varphi_{2}$. Let $A_{0} A_{1} A_{2} \ldots \in$ $\left(2^{A P}\right)^{\omega}$ and $B_{0} B_{1} B_{2} \ldots \in Q^{\omega}$ satisfying the constraints (i) and (ii). It is now shown that:

$$
\psi \in B_{0} \quad \text { iff } \quad A_{0} A_{1} A_{2} \ldots \models \psi
$$

This goes as follows.
$\Leftarrow$ : Assume $A_{0} A_{1} A_{2} \ldots \models \psi$ where $\psi=\varphi_{1} \cup \varphi_{2}$. Then, there exists $j \geqslant 0$ such that

$$
A_{j} A_{j+1} \ldots \vDash \varphi_{2} \quad \text { and } \quad A_{i} A_{i+1} \ldots \vDash \varphi_{1} \text { for } 0 \leqslant i<j
$$

From the induction hypothesis (applied to $\varphi_{1}$ and $\varphi_{2}$ ) it follows that

$$
\varphi_{2} \in B_{j} \quad \text { and } \quad \varphi_{1} \in B_{i} \quad \text { for } 0 \leqslant i<j
$$

By induction on $j$ we obtain: $\varphi_{1} \cup \varphi_{2} \in B_{j}, B_{j-1} \ldots, B_{0}$.
$\Rightarrow$ : Assume $\varphi_{1} \cup \varphi_{2} \in B_{0}$. Since $B_{0}$ is elementary, $\varphi_{1} \in B_{0}$ or $\varphi_{2} \in B_{0}$. Distinguish between $\varphi_{2} \in B_{0}$ and $\varphi_{2} \notin B_{0}$. If $\varphi_{2} \in B_{0}$, it follows from the induction hypothesis $A_{0} A_{1} \ldots \vDash \varphi_{2}$, and thus $A_{0} A_{1} \ldots \vDash \varphi_{1} \cup \varphi_{2}$. This remains the case $\varphi_{2} \notin B_{0}$. Then $\varphi_{1} \in B_{0}$ and $\varphi_{1} \cup \varphi_{2} \in B_{0}$. Assume $\varphi_{2} \notin B_{j}$ for all $j \geqslant 0$. From the definition of the transition relation $\delta$, we obtain using an inductive argument (successively applied to $\varphi_{1} \in B_{j}, \varphi_{2} \notin B_{j}$ and $\varphi_{1} \cup \varphi_{2} \in B_{j}$ for $\left.j \geqslant 0\right)$ :

$$
\varphi_{1} \in B_{j} \quad \text { and } \quad \varphi_{1} \cup \varphi_{2} \in B_{j} \quad \text { for all } j \geqslant 0
$$



Figure 5.21: A generalised Büchi automaton for the LTL formula $\bigcirc a$.

As $B_{0} B_{1} B_{2} \ldots$ satisfies constraint (ii), it follows that

$$
B_{j} \in F_{\varphi_{1} \cup \varphi_{2}} \text { for infinitely many } j \geqslant 0
$$

On the other hand, we have

$$
\underbrace{\varphi_{2} \notin B_{j} \quad \text { and } \quad \varphi_{1} \cup \varphi_{2} \in B_{j}}_{\text {iff } \quad B_{j} \notin F_{\varphi_{1} \cup \varphi_{2}}}
$$

for all $j$. Contradiction! Thus, $\varphi_{2} \in B_{j}$ for some $j \geqslant 0$. Without loss of generality, assume $\varphi_{2} \notin B_{0}, \ldots, B_{j-1}$, i.e., let $j$ be the smallest index such that $\varphi_{2} \in B_{j}$. The induction hypothesis for $0 \leqslant i<j$ yields

$$
\varphi_{1} \in B_{i} \quad \text { and } \quad \varphi_{1} \cup \varphi_{2} \in B_{i} \quad \text { for all } 0 \leqslant i<j
$$

From the induction hypothesis applied to $\varphi_{1}$ and $\varphi_{2}$ it follows that

$$
A_{j} A_{j+1} \ldots \models \varphi_{2} \quad \text { and } \quad A_{i} A_{i+1} \ldots \models \varphi_{1} \text { for } 0 \leqslant i<j
$$

We conclude that $A_{0} A_{1} A_{2} \ldots \vDash \varphi_{1} \cup \varphi_{2}$.

Example 5.38. Construction of a GNBA (Next Step)
Consider $\varphi=\bigcirc a$. The GNBA $\mathcal{G}_{\varphi}$ (see Figure 5.21) is obtained as indicated in the proof of Theorem 5.37, . The states of the automaton are the elementary sets of formulae contained in

$$
\operatorname{closure}(\varphi)=\{a, \bigcirc a, \neg a, \neg \bigcirc a\}
$$

The state space $Q$ consists of the following elementary sets

$$
\begin{array}{ll}
B_{1}=\{a, \bigcirc a\}, & B_{2}=\{a, \neg \bigcirc a\}, \\
B_{3}=\{\neg a, \bigcirc a\}, & B_{4}=\{\neg a, \neg \bigcirc a\} .
\end{array}
$$

The initial states of $\mathcal{G}_{\varphi}$ are the elementary sets $B \in Q$ with $\varphi=\bigcirc a \in B$. Thus $Q_{0}=$ $\left\{B_{1}, B_{3}\right\}$. Let us justify some of the transitions. For state $B_{1}, B_{1} \cap\{a\}=\{a\}$, so $\delta\left(B_{1}, \varnothing\right)=\varnothing$. In addition, $\delta\left(B_{1},\{a\}\right)=\left\{B_{1}, B_{2}\right\}$ since $\bigcirc a \in B_{1}$ and $B_{1}$ and $B_{2}$ are the only states that contain $a$. As $B_{2} \cap\{a\}=\{a\}$, we get $\delta\left(B_{2}, \varnothing\right)=\varnothing$. Moreover, $\delta\left(B_{2},\{a\}\right)=\left\{B_{3}, B_{4}\right\}$. This follows from the fact that for $\neg \bigcirc \psi \in \operatorname{closure}(\varphi)$, and any direct successor $B^{\prime}$ of $B$ we have

$$
\neq \bigcirc \psi \in B \text { if and only if } \psi \notin B^{\prime}
$$

(This follows by the definition of $\delta$, local consistency and maximality.) Since $\neg \bigcirc a \in B_{2}$, and $B_{3}$ and $B_{4}$ are the only states that do not contain $a$, we have $\delta\left(B_{2},\{a\}\right)=\left\{B_{3}, B_{4}\right\}$. Hence, $\delta\left(B_{4},\{a\}\right)=\varnothing$ since $B_{4} \cap\{a\}=\varnothing \neq\{a\}$. Using a similar reasoning as above, we obtain $\delta\left(B_{4}, \varnothing\right)=\left\{B_{3}, B_{4}\right\}$. The outgoing transitions of $B_{3}$ are determined analogously. The set $\mathcal{F}$ is empty as $\varphi=\bigcirc a$ does not contain an until operator. Since $\mathcal{F}=\varnothing$, every infinite run in the GNBA $\mathcal{G}_{\varphi}$ is accepting. As each infinite run is either of the form $B_{1} B_{1} \ldots, B_{1} B_{2} \ldots, B_{3} B_{1} \ldots$, or $B_{3} B_{2} \ldots$, and $\varphi=\bigcirc a \in B_{1}, B_{2}$, it follows that indeed all runs satisfy $\bigcirc a$.

Example 5.39. Construction of a GNBA (Until)
Consider $\varphi=a \cup b$ and let $A P=\{a, b\}$. Then

$$
\operatorname{closure}(\varphi)=\{a, b, \neg a, \neg b, a \cup b, \neg(a \cup b)\}
$$

The construction in the proof of Theorem 5.37 yields the GNBA $\mathcal{G}_{\varphi}$ illustrated in Figure 5.22. In order not to blur the figure, transition labels have been omitted. (The label of transition $B \rightarrow B^{\prime}$ equals the propositional logic formula characterising the set $B \cap A P$.) The states correspond to the elementary sets of closure $(\varphi)$

$$
\begin{aligned}
& B_{1}=\left\{\begin{array}{lll}
\{a, & b, & \varphi
\end{array}\right\}, \\
& B_{2}=\{\neg a, \\
& b, \\
& B_{3}=\left\{\begin{array}{l}
\{a, \\
\neg b,
\end{array}\right) \\
& B_{4}=\{\neg a, \\
& B_{5}=\{b, \\
& \neg a, \\
& \neg b, \\
& \neg \varphi\}, \\
& \{\varphi\} .
\end{aligned}
$$

The initial states are the sets $B_{i} \in Q$ with $\varphi \in B_{i}$; thus, $Q_{0}=\left\{B_{1}, B_{2}, B_{3}\right\}$. The set $\mathcal{F}=\left\{F_{\varphi}\right\}$ of the accepting sets is a singleton, since $\varphi$ contains a single until operator.

The set $F_{\varphi}$ is given by

$$
F_{\varphi}=\{B \in Q \mid \varphi \notin B \vee b \in B\}=\left\{B_{1}, B_{2}, B_{4}, B_{5}\right\} .
$$

Since the accepting set is a singleton set, the GNBA $\mathcal{G}_{\varphi}$ can be understood as an NBA with the accepting set $F_{\varphi}$.

Let us justify some of the transitions. We have $B_{1} \cap A P=\{a, b\}$ and $a \cup b \in B_{1}$. As:

$$
b \in B_{1} \vee\left(a \in B_{1} \wedge a \cup b \in B^{\prime}\right)
$$

holds for any elementary set $B^{\prime} \subseteq$ closure $(a \cup b)$, we have that $\delta\left(B_{1},\{a, b\}\right)$ contains all states. These are the only outgoing transitions of state $B_{1}$. A similar reasoning applies to $\delta\left(B_{2},\{b\}\right)$. Consider state $B_{3}$. Then $B_{3} \cap A P=\{a\}$. The states $B^{\prime}$ that are possible direct successors of $B_{3}$ under $\{a\}$ are those satisfying

$$
a \in B_{3} \wedge a \cup b \in B^{\prime} .
$$

Thus, $\delta\left(B_{3},\{a\}\right)=\left\{B_{1}, B_{2}, B_{3}\right\}$. Finally, consider state $B_{5}$. This state only has successors for input symbol $B_{5} \cap A P=\{a\}$. For any $\varphi_{1} \cup \varphi_{2} \in \operatorname{closure}(\varphi)$ it can be shown that for any successor $B^{\prime}$ of $B$ :

$$
\varphi_{1} \cup \varphi_{2} \notin B \quad \text { iff } \quad \varphi_{2} \notin B \wedge\left(\varphi_{1} \notin B \vee \varphi_{1} \cup \varphi_{2} \notin B^{\prime}\right)
$$

(The proof of this fact is left as an exercise.) Applying this to the state $B_{5}$ yields that all states not containing $\varphi$ are possible successors of $B_{5}$. For example, the run $B_{3} B_{3} B_{1} B_{4}^{\omega}$ is accepting. This run corresponds to the word $\{a\}\{a\}\{a, b\} \varnothing^{\omega}$ which indeed satisfies $a \cup b$. The word $\{a\}^{\omega}$ does not satisfy $a \cup b$. It has exactly one run in $\mathcal{G}_{\varphi}$, namely $B_{3}^{\omega}$. As $B_{3}$ is not a final state, this run is not accepting, i.e., $\{a\}^{\omega} \notin \mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)$.

## Remark 5.40. Simplified Representation of the Automata States

Any state of the GNBA for an LTL formula $\varphi$ contains either $\psi$ or its negation $\neg \psi$ for every subformula $\psi$ of $\varphi$. This is somewhat redundant. It suffices to represent state $B \in \operatorname{closure}(\varphi)$ by the propositional symbols $a \in B \cap A P$, and the formulae $\bigcirc \psi$ or $\varphi_{1} \cup \varphi_{2} \in B$.

Having constructed a GNBA $\mathcal{G}_{\varphi}$ for a given LTL formula $\varphi$, an NBA for $\varphi$ can be obtained by the transformation "GNBA $\rightsquigarrow$ NBA" described in Theorem 4.56 on page 195. Recall that this transformation for GNBA with two or more acceptance sets generates a copy of $\mathcal{G}_{\varphi}$ for each acceptance set of $\mathcal{G}_{\varphi}$. In our case, the number of copies that we need is given by the number of until subformulae of $\varphi$. We obtain the following result:


Figure 5.22: A generalised Büchi automaton for $a \cup b$.

## Theorem 5.41. Constructing an NBA for an LTL Formula

For any LTL formula $\varphi$ (over AP) there exists an NBA $\mathcal{A}_{\varphi}$ with $\operatorname{Words}(\varphi)=\mathcal{L}_{\omega}\left(\mathcal{A}_{\varphi}\right)$ which can be constructed in time and space $2^{\mathcal{O}(|\varphi|)}$.

Proof: From Theorem 5.37 (page 278), it follows that a GNBA $\mathcal{G}_{\varphi}$ can be constructed which has at most $2^{|\varphi|}$ states. As the number of accepting states in $\mathcal{G}_{\varphi}$ equals the number of until-subformulas in $\varphi$, GNBA $\mathcal{G}_{\varphi}$ has at most $|\varphi|$ accepting states. Transforming the GNBA into an equivalent NBA (as described in the proof of Theorem 4.56, page 195), yields an NBA with at most $|\varphi|$ copies of the state space of $\mathcal{G}_{\varphi}$. Thus, the number of states in the NBA is at most $2^{|\varphi|} \cdot|\varphi|=2^{|\varphi|+\log |\varphi|}$ states. This yields the claim.

There are various algorithms in the literature for associating an automaton for infinite words to an LTL formula. The presented algorithm is one of the conceptually simplest algorithms, but often yields unnecessarily large GNBAs. For example, for the LTL formulae $\bigcirc a$ and $a \cup b$, an NBA with two states suffices. (It is left to the reader to provide these NBAs.) Several optimizations are possible to improve the size of the resulting GNBA, but the exponential blowup cannot be avoided. This is formally stated in the following theorem:

## Theorem 5.42. Lower Bound for NBA from LTL Formulae

There exists a family of LTL formulae $\varphi_{n}$ with $\left|\varphi_{n}\right|=\mathcal{O}(\operatorname{poly}(n))$ such that every NBA for $\varphi_{n}$ has at least $2^{n}$ states.

Proof: Let $A P$ be an arbitrary nonempty set of atomic propositions, that is, $\left|2^{A P}\right| \geqslant 2$. Consider the family of languages:

$$
\mathcal{L}_{n}=\left\{A_{1} \ldots A_{n} A_{1} \ldots A_{n} \sigma \mid A_{i} \subseteq A P \wedge \sigma \in\left(2^{A P}\right)^{\omega}\right\}, \quad \text { for } n \geqslant 0
$$

It is not difficult to check that $\mathcal{L}_{n}=\operatorname{Words}\left(\varphi_{n}\right)$ where

$$
\varphi_{n}=\bigwedge_{a \in A P} \bigwedge_{0 \leqslant i<n}\left(\bigcirc^{i} a \longleftrightarrow \bigcirc^{n+i} a\right)
$$

Here, $\bigcirc^{j}$ stands for the $j$-fold application of the next-step operator $\bigcirc$, i.e., $\bigcirc^{1} \varphi=\bigcirc \varphi$ and $\bigcirc^{n+1} \varphi=\bigcirc \bigcirc{ }^{n} \varphi$. It follows that $\varphi_{n}$ is an LTL formula of polynomial length. More precisely, $\left|\varphi_{n}\right| \in \mathcal{O}(|A P| \cdot n)$.

However, any NBA $\mathcal{A}$ with $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{n}$ has at least $2^{n}$ states. Essentially, this is justified by the following consideration. Since the words

$$
A_{1} \ldots A_{n} A_{1} \ldots A_{n} \varnothing \varnothing \varnothing \ldots
$$

are accepted by $\mathcal{A}, \mathcal{A}$ contains for every word $A_{1} \ldots A_{n}$ of length $n$, a state $q\left(A_{1} \ldots A_{n}\right)$, which can be reached from an initial state by consuming the prefix $A_{1} \ldots A_{n}$. Starting from $q\left(A_{1} \ldots A_{n}\right)$ it is possible to visit an accept state infinitely often by accepting the suffix $A_{1} \ldots A_{n} \varnothing \varnothing \varnothing \ldots$. If $A_{1} \ldots A_{n} \neq A_{1}^{\prime} \ldots A_{n}^{\prime}$ then

$$
A_{1} \ldots A_{n} A_{1}^{\prime} \ldots A_{n}^{\prime} \varnothing \varnothing \varnothing \ldots \notin \mathcal{L}_{n}=\mathcal{L}_{\omega}(\mathcal{A})
$$

Therefore, the states $q\left(A_{1} \ldots A_{n}\right)$ are all pairwise different. As there are $\left|2^{A P}\right|$ possible combinations for $A_{1} \ldots A_{n}, \mathcal{A}$ has at least $\left(\left|2^{A P}\right|\right)^{n} \geqslant 2^{n}$ states.

## Remark 5.43. Büchi Automata are More Expressive Than LTL

The results so far show that for every LTL formula $\varphi$ an NBA can be constructed that accepts exactly the infinite sequences satisfying $\varphi$. We state without proof that the reverse, however, is not true. It can be shown that for, e.g., the LT property

$$
P=\left\{A_{0} A_{1} A_{2} \ldots \in\left(2^{\{a\}}\right)^{\omega} \mid a \in A_{2 i} \text { for } i \geqslant 0\right\}
$$

which requires $a$ to hold in every even position, there is no LTL formula $\varphi$ with $\operatorname{Words}(\varphi)=$ $P$. On the other hand, there exists an NBA $\mathcal{A}$ with $\mathcal{L}_{\omega}(\mathcal{A})=P$. (It is left to the reader to provide such an NBA.)

### 5.2.1 Complexity of the LTL Model-Checking Problem

Let us summarize the results of the previous sections in order to provide an overview of LTL model checking. Subsequently, we discuss the complexity of the LTL model-checking problem.

As explained before, the essential idea behind the automata-based model-checking algorithm for LTL is based upon the following relations:

$$
\begin{aligned}
T S \models \varphi & \text { iff } \operatorname{Traces}(T S) \subseteq \operatorname{Words}(\varphi) \\
& \text { iff } \operatorname{Traces}(T S) \subseteq\left(2^{A P}\right)^{\omega} \backslash \operatorname{Words}(\neg \varphi) \\
& \text { iff } \operatorname{Traces}(T S) \cap \underbrace{\operatorname{Words}(\neg \varphi)}_{\mathcal{L}_{\omega}\left(\mathcal{A}_{\neg \varphi}\right)}=\varnothing \\
& \text { iff } T S \otimes \mathcal{A}_{\neg \varphi} \models \diamond \square \neg F .
\end{aligned}
$$

Here, NBA $\mathcal{A}_{\neg \varphi}$ accepts $\operatorname{Words}(\neg \varphi)$ and $F$ is its set of accept states. The algorithm to transform an LTL formula $\varphi$ into an NBA may give rise to an NBA whose state space size is exponential in the length of $\varphi$. The NBA $\mathcal{A}_{\neg \varphi}$ can thus be constructed in exponential time:

$$
\mathcal{O}\left(2^{|\varphi|} \cdot|\varphi|\right)=\mathcal{O}\left(2^{|\varphi|+\log |\varphi|}\right)
$$

This complexity bound, together with the fact that the state space of $\mathcal{A}$ is exponential in $|\varphi|$, yields an upper bound for the time- and space-complexity of LTL model checking (see Algorithm 11, page 272):

$$
\mathcal{O}\left(|T S| \cdot 2^{|\varphi|}\right)
$$

## Remark 5.44. LTL Model Checking with Fairness

As a consequence of Theorem 5.30 (see page 264), the model-checking problem for LTL with fairness assumptions can be reduced to the model-checking problem for plain LTL. So, in order to check the formula $\varphi$ under fairness assumption fair, it suffices to verify the formula fair $\rightarrow \varphi$ with an LTL model-checking algorithm. This approach, however, has as its main drawback that the length $\mid$ fair $\mid$ can have an exponential influence on the run-time of the algorithm. This is due to the construction of an NBA for the negated formula, i.e., $\neg($ fair $\rightarrow \varphi)$, whose size is exponential in $\mid \neg($ fair $\rightarrow \varphi)|=|$ fair $|+|\varphi|$. To avoid this additional exponential blowup, a modified persistence check (see Algorithm 8 on page 211) can be exploited to analyze the product transition system $T S \otimes \mathcal{A}_{\neg \varphi}$ (instead of $\left.T S \otimes \mathcal{A}_{\neg(f a i r \rightarrow \varphi)}\right)$. This can be done using standard graph algorithms. The reader is referred to Exercise 5.22 (on page 308) for more details.

An interesting aspect of the LTL model-checking algorithm is that it can be executed on-the-fly, i.e., while constructing the NBA $\mathcal{A}_{\neg \varphi}$. This may avoid the need for constructing the entire automaton $\mathcal{A}_{\neg \varphi}$. This on-the-fly procedure works as follows. Suppose we are given a high-level description of the transition system $T S$, e.g., by means of a syntactic description of the concurrent processes (as in Spin's input language Promela). The generation of the reachable states of $T S$ can proceed in parallel with the construction of the relevant fragment of $\mathcal{A}_{\neg \varphi}$. Simultaneously, the reachable fragment of the product transition system $T S \otimes \mathcal{A}_{\neg \varphi}$ is constructed in a DFS-manner. (This, in fact, yields the outermost DFS in the nested DFS for checking persistence in $T S \otimes \mathcal{A}_{\neg \varphi}$.) So the entire LTL model-checking procedure can be interleaved with the generation of the relevant fragments of $T S$ and $\mathcal{A}_{\neg \varphi}$. In this way, the product transition system $T S \otimes \mathcal{A}_{\neg \varphi}$ is constructed "on demand", so to speak. A new vertex is only considered if no accepting cycle has been encountered yet in the partially constructed product transition system $T S \otimes \mathcal{A}_{\neg \varphi}$. When generating the successors of a state in $\mathcal{A}_{\neg \varphi}$, it suffices to only consider the successors matching the current state $T S$ (rather than all possible successors). It is thus possible that an accepting cycle is found, i.e., a violation of $\varphi$ (with corresponding counterexample), without the need for generating the entire automaton $\mathcal{A}_{\neg \varphi}$.

This on-the-fly generation of $\operatorname{Reach}(T S), \mathcal{A}_{\neg \varphi}$, and $T S \otimes \mathcal{A}_{\neg \varphi}$ is adopted in practical LTL model checkers (such as Spin) and for many examples yields an efficient verification procedure. From a theoretical point of view, though, the LTL model-checking problem remains computationally hard and is "probably" not efficiently solvable. It is shown in the sequel of this section that the LTL model-checking problem is PSPACE-complete. We assume some familiarity with basic notions of complexity theory and the complexity classes coNP and PSPACE; see, e.g., the textbooks [160, 320] and the Appendix.

Before proving the PSPACE-hardness of the LTL model-checking problem, a weaker result is provided. To that end, let us recall the so-called Hamiltonian path problem. Consider a finite directed graph $G=(V, E)$ with set $V$ of vertices and set $E \subseteq V \times V$ of edges. The Hamiltonian path problem is a decision problem that yields an affirmative answer when $G$ has a Hamiltonian path, i.e., a path which passes through every vertex in $V$ exactly once.

The next ingredient needed for the following result is the complement of the LTL modelchecking problem. This decision problem takes as input a finite transition system TS and an LTL formula $\varphi$ and asks whether $T S \not \vDash \varphi$. Thus, whenever the LTL model-checking problem provides an affirmative answer, its complement yields "no", and whenever the result of the LTL model-checking problem is negative, its complement yields "yes".

## Lemma 5.45.

The Hamiltonian path problem is polynomially reducible to the complement of the LTL model-checking problem.

Proof: To establish a polynomial reduction from the Hamiltonian path problem to the complement of the LTL model-checking problem, a mapping is needed from instances of the Hamiltonian path problem onto instances for the complement model-checking problem. That is, we need to map a finite directed graph $G$ onto a pair $(T S, \varphi)$ where $T S$ is a finite transition system and $\varphi$ is an LTL formula, such that
(i) $G$ has a Hamiltonian path if and only if $T S \not \vDash \varphi$, and
(ii) the transformation $G \rightsquigarrow(T S, \varphi)$ can be performed in polynomial time.

The basic concept of the mapping is as follows. Essentially, $T S$ corresponds to $G$, i.e., the states of $T S$ are the vertices in $G$ and the transitions in $T S$ correspond to the edges in $G$. For technical reasons, $G$ is slightly modified to ensure that $T S$ has no terminal states. More precisely, the state graph of TS arises from $G$ by inserting a new vertex $b$ to $G$, which is reachable from every vertex $v$ via an edge and which is only equipped with a self-loop $b \rightarrow b$ (i.e., $b$ has no further outgoing edges). Formally, for $G=(V, E)$ the associated transition system is

$$
T S=(V \uplus\{b\},\{\tau\}, \rightarrow, V, V, L)
$$

where $L(v)=\{v\}$ for any vertex $v \in V$ and $L(b)=\varnothing$. The atomic propositions for $T S$ are thus obtained by taking the identities of the vertices in $G$. The transition relation $\rightarrow$ is defined as follows:

$$
\frac{(v, w) \in E}{v \xrightarrow{\tau} w} \quad \text { and } \quad \frac{v \in V \cup\{b\}}{v \xrightarrow{\tau} b}
$$

This completes the mapping of directed graph $G$ onto transition system $T S$. It remains to formalize the negation of the existence (i.e., the absence) of a Hamiltonian path by means of an LTL formula. Let

$$
\varphi=\neg \bigwedge_{v \in V}(\diamond v \wedge \square(v \rightarrow \bigcirc \square \neg v))
$$

Stated in words, $\varphi$ asserts that it is not the case that each vertex $v \in V$ is eventually visited and never visited again. Note that the conjunction does not quantify over $b \notin V$.

It is evident that $T S$ and $\varphi$ can be constructed in polynomial time for a given directed graph $G$. As a last step, we need to show that $G$ has a Hamiltonian path if and only if $T S \not \vDash \varphi$.
$\Leftarrow$ : Assume $T S \not \vDash \varphi$. Then there exists a path $\pi$ in $T S$ such that

$$
\pi \vDash \bigwedge_{v \in V}(\Delta v \wedge \square(v \rightarrow \bigcirc \square \neg v))
$$

As $\pi \models \bigwedge_{v \in V} \diamond v$, each vertex $v \in V$ occurs at least once in the path $\pi$. Since

$$
\pi \models \bigwedge_{v \in V} \square(v \rightarrow \bigcirc \square \neg v)
$$

there is no vertex that occurs more than once in $\pi$. Thus, $\pi$ is of the form $v_{1} \ldots v_{n} b b b \ldots$ where $V=\left\{v_{1}, \ldots, v_{n}\right\}$ and $|V|=n$. In particular, $v_{1} \ldots v_{n}$ is a Hamiltonian path in $G$.
$\Rightarrow$ : Each Hamiltonian path $v_{1} \ldots v_{n}$ in $G$ can be extended to a path $\pi=v_{1} \ldots v_{n} b b b \ldots$ in TS. Thus, $\pi \not \vDash \varphi$ and, hence, $T S \not \vDash \varphi$.

Since the Hamiltonian path problem is known to be NP-complete, it follows from the previous lemma that the LTL model-checking problem is coNP-hard. The following complexity result states that the LTL model-checking problem is PSPACE-hard.

## Theorem 5.46. Lower Bound for LTL Model Checking

The LTL model-checking problem is PSPACE-hard.

Proof: Let TS be a finite transition system and $\varphi$ an LTL formula. As a first observation, note that it suffices to show the PSPACE-hardness of the existential variant of the LTL model-checking problem. This existential decision problem takes as input TS and $\varphi$ and yields "yes" if $\pi \models \varphi$ for some (initial, infinite) path $\pi$ in $T S$, and "no" otherwise. Given that

$$
\begin{array}{lll}
T S \models \varphi & \text { if and only if } \pi \models \varphi \text { for all paths } \pi \\
& \text { if and only if } \operatorname{not}(\pi \models \neg \varphi \text { for some path } \pi)
\end{array}
$$

it follows that the output for the instance $(T S, \varphi)$ of the LTL model-checking problem yields "yes" if and only if the output for the instance $(T S, \neg \varphi)$ of the existential variant of the LTL model-checking problem yields "no". Thus, the PSPACE-hardness of the existential variant of the LTL model-checking problem yields the PSPACE-hardness of the complement of the existential variant of the LTL model checking problem which again induces the PSPACE-hardness of the LTL model-checking problem. Recall that PSPACE $=$ coPSPACE, thus, the complement of any PSPACE-hard problem is PSPACEhard.

In the remainder of the proof we concentrate on showing the PSPACE-hardness of the existential LTL model-checking problem. This is established by providing a polynomial reduction from any decision problem $K \in$ PSPACE to the existential LTL model-checking problem. Let $\mathcal{M}$ be a polynomial space-bounded deterministic Turing machine that accepts exactly the words $w \in K$. The goal is now to transform $(\mathcal{M}, w)$ by a deterministic
polynomial-time bounded procedure into a pair $(T S, \varphi)$ such that $\mathcal{M}$ accepts $w$ if and only if $T S$ contains a path $\pi$ with $\pi \models \varphi$.

The following proof, adopted from [372], has some similarities with Cook's theorem stating the NP-completeness of the satisfiability problem for propositional logic. The rough idea is to encode the initial configurations, possible transitions, and accepting configurations of a given Turing machine $\mathcal{M}$ by LTL formulae. In the sequel, let $\mathcal{M}$ be a deterministic single-tape Turing machine with state-space $Q$, starting state $q_{0} \in Q$, the set $F$ of accept states, the tape alphabet $\Sigma$, and the transition function $\delta: Q \times \Sigma \rightarrow Q \times \Sigma \times\{L, R, N\}$. The intuitive meaning of $\delta$ is as follows. Suppose $\delta(q, A)=(p, B, L)$. Then, whenever the current state is $q$ and the current symbol in cell $i$ under the cursor is $A$, then $\mathcal{M}$ changes to state $p$, overwrites the symbol $A$ by $B$ in cell $i$, and moves the cursor one position to the left, i.e., it positions the cursor under cell $i-1$. For $R$ or $N$, the cursor moves one position to the right, or does not move at all, respectively. (For our purposes, there is no need to distinguish between the input and the tape alphabet.)

For technical reasons, it is required that the accept states are absorbing, i.e., $\delta(q, A)=$ $(q, A, N)$ for all $q \in F$. Moreover, it is assumed that $\mathcal{M}$ is polynomially space-bounded, i.e., there is a polynomial $P$ such that the computation for an input word $A_{1} \ldots A_{n} \in \Sigma^{*}$ of length $n$ visits at most the first $P(n)$ cells on the tape. Without loss of generality, the coefficients of $P$ are assumed to be natural numbers and $P(n) \geqslant n$.

To encode $\mathcal{M}$ 's possible behaviors by LTL formulae for an input word of length $n$, the Turing machine is transformed into a transition system $T S=\operatorname{TS}(\mathcal{M}, n)$ with the following state space:

$$
S=\{0,1, \ldots, P(n)\} \cup\{(q, A, i) \mid q \in Q \cup\{*\}, A \in \Sigma, 0<i \leqslant P(n)\}
$$

The structure of $T S$ is shown in Figure 5.23. TS consists of $P(n)$ copies of "diamonds". The $i$ th copy starts in state $i-1$, ends in state $i$, and contains the states $(q, A, i)$ where $q \in Q \cup\{*\}$ and $A \in \Sigma$ "between" state $i-1$ and state $i$. Intuitively, the first component $q \in Q \cup\{*\}$ of state ( $q, A, i$ ) in the $i$ th copy indicates whether the cursor points to cell $i$ (in which case $q \in Q$ is the current state) or to some other tape cell (in which case $q=*$ ). The symbol $A \in \Sigma$ in state ( $q, A, i$ ) stands for the current symbol in cell $i$. The path fragments from state 0 to state $P(n)$ serve to represent the possible configurations of $\mathcal{M}$. More precisely, the configuration in which the current content of the tape is $A_{1} A_{2} \ldots A_{P(n)}$, the current state is $q$, and the cursor points to cell $i$ is encoded by the path fragment:

$$
0\left(*, A_{1}, 1\right) 1\left(*, A_{2}, 2\right) 2 \ldots i-1\left(q, A_{i}, i\right) i\left(*, A_{i+1}, i+1\right) i+1 \ldots P(n)
$$

Accordingly, the computation of $\mathcal{M}$ for an input word of length $n$ can be described by the concatenation of such path fragments.


Figure 5.23: Transition system $\operatorname{TS}(\mathcal{M}, n)$ for Turing machine $\mathcal{M}$ and input length $n$.

We use the state identities as atomic propositions. In addition, proposition begin is used to identify state 0 , while proposition end is used for state $P(n)$. That is, $A P=S \cup$ \{begin, end $\}$ with the obvious labeling function. Let $\Phi_{Q}$ denote the disjunction over all atoms ( $q, A, i$ ) where $q \in Q$ (i.e., $q \neq *$ ) and $A \in \Sigma, 0<i \leqslant P(n)$. The LTL formulae:

$$
\begin{aligned}
& \varphi_{\text {Conf }}=\square\left(\text { begin } \longrightarrow \varphi_{\text {Conf }}^{1} \wedge \varphi_{\text {Conf }}^{2}\right) \text { where } \\
& \varphi_{\text {Conf }}^{1}=\bigvee_{1 \leqslant i \leqslant P(n)} \bigcirc^{2 i-1} \Phi_{Q} \\
& \varphi_{\text {Conf }}^{2}=\bigwedge_{1 \leqslant i \leqslant P(n)}\left(\bigcirc^{2 i-1} \Phi_{Q} \longrightarrow \bigwedge_{\substack{1 \leqslant j \leqslant P(n) \\
j \neq i}} \bigcirc^{2 j-1} \neg \Phi_{Q}\right)
\end{aligned}
$$

characterize any path $\pi$ in $T S$ such that all path fragments of $\pi$ that lead from 0 via $1, \ldots, P(n-1)$ to $P(n)$ encode a configuration of $\mathcal{M}$. Note that $\varphi_{\text {Conf }}^{1} \wedge \varphi_{\text {Conf }}^{2}$ ensures that the cursor points exactly to one of the positions $1, \ldots, P(n)$. Thus, any path $\pi$ in $T S$ such that $\pi \models \varphi_{\text {Conf }}$ can be viewed as a sequence of configurations in $\mathcal{M}$. However, this sequence need not be a computation of $\mathcal{M}$, since the configurations might not be consecutive according to $\mathcal{M}$ 's operational behavior. For this, we need additional constraints that formalize $\mathcal{M}$ 's stepwise behavior.

The transition relation $\delta$ of $\mathcal{M}$ can be encoded by an LTL formula $\varphi_{\delta}$ that arises through a conjunction of formulae $\varphi_{q, A}$ describing the semantics of $\delta(q, A)$. Here, $q$ ranges over all states in $Q$ and $A$ over the tape-symbols in $\Sigma$. For instance, if $\delta(q, A)=(p, B, L)$ then

$$
\varphi_{q, A}=\square \bigwedge_{1 \leqslant i \leqslant P(n)}\left(\bigcirc^{2 i-1}(q, A, i) \longrightarrow \psi_{(q, A, i, p, B, L)}\right)
$$

where $\psi_{(q, A, i, p, B, L)}$ is defined as


Here, $C$ denotes the disjunction of the atoms $(r, C, j)$ where $r \in Q \cup\{*\}$ and $1 \leqslant j \leqslant P(n)$, and $p$ for the disjunction of all atoms $(p, D, j)$ where $D \in \Sigma$ and $1 \leqslant j \leqslant P(n)$.

The starting configuration of $\mathcal{M}$ for a given input word $w=A_{1} \ldots A_{n} \in \Sigma^{*}$ is given by the formula

$$
\varphi_{\text {start }}^{w}=\bigcirc q_{0} \wedge \bigwedge_{1 \leqslant i \leqslant n} \bigcirc^{2 i-1} A_{i} \wedge \bigwedge_{n<i \leqslant P(n)} \bigcirc^{2 i-1} \sqcup
$$

The first conjunct $\bigcirc q_{0}$ asserts that $\mathcal{M}$ starts in its starting state; the other conjuncts assert that $A_{1} \ldots A_{n} \sqcup^{P(n)-n}$ is the content of the tape where $\sqcup$ denotes the blank symbol. The accepting configurations are formalized by the formula

$$
\varphi_{\mathrm{accept}}=\diamond \bigvee_{q \in F} q
$$

For a given input word $w=A_{1} \ldots A_{n}$ of length $n$ for $\mathcal{M}$, let

$$
\varphi_{w}=\varphi_{\text {start }}^{w} \wedge \varphi_{\text {Conf }} \wedge \varphi_{\delta} \wedge \varphi_{\text {accept }} .
$$

Note that the length of $\varphi_{w}$ is polynomial in the size of $\mathcal{M}$ and the length $n$ of $w$. Thus, $T S=T S(\mathcal{M}, n)$ and $\varphi_{w}$ can be constructed from $(\mathcal{M}, w)$ in polynomial time. Moreover, it also follows that there exists a path $\pi \models \varphi_{w}$ in $T S$ if and only if $\mathcal{M}$ accepts the input word $w$.

For real applications, the aforementioned theoretical complexity result is less dramatic than it seems at first sight, since the complexity is linear in the size of the transition system and exponential in formula length. In practice, typical requirement specifications yield short LTL formulae. In fact, the exponential growth in the length of the formula is not decisive for the practical application of LTL model checking. Instead, the linear dependency on the size of the transition system is the critical factor. As has been discussed at the end of Chapter 2 (page 77), the size of transition systems may be huge even for relatively simple systems - the state-space explosion problem. In later chapters of this monograph, various techniques will be treated to combat this state-space explosion problem.

We mentioned before that the LTL model-checking problem is PSPACE-complete. The previous result showed PSPACE-hardness. It remains to show that it belongs to the complexity class PSPACE.

To prove that the LTL model-checking problem is in PSPACE, we resort (again) to the existential variant of the LTL model-checking problem. This is justified by the fact that PSPACE - as any other deterministic complexity class - is closed under complementation. That is, the LTL model-checking problem is in PSPACE iff its complement is in PSPACE. This is equivalent to the statement that the existential variant of the LTL model-checking problem is solvable by a polynomial space-bounded algorithm. By Savitch's theorem (PSPACE agrees with NPSPACE), it suffices to provide a nondeterministic polynomial space-bounded algorithm that solves the existential LTL model-checking problem.

## Lemma 5.47.

The existential LTL model-checking problem is solvable by a nondeterministic space-bounded algorithm.

Proof: In the sequel, let $\varphi$ be an LTL formula and $T S=(S, A c t, \rightarrow, I, A P, L)$ a finite transition system. The goal is to check nondeterministically whether $T S$ has a path $\pi$ with $\pi \vDash \varphi$, while the memory requirements are bounded by $\mathcal{O}(\operatorname{poly}(\operatorname{size}(T S),|\varphi|))$. The techniques discussed in Section 5.2 (page 270 ff ) suggest to build an NBA $\mathcal{A}_{\varphi}$ for $\varphi$, construct the product transition system $T S \otimes \mathcal{A}_{\varphi}$ and check whether this product contains a reachable cycle containing an accept state of $\mathcal{A}_{\varphi}$. We now modify this approach to obtain an NPSPACE algorithm. Instead of an NBA for $\varphi$, we deal here with the GNBA $\mathcal{G}_{\varphi}$ for $\varphi$. Recall that states in this automaton are elementary subsets of the closure of $\varphi$ (see Definition 5.34 on page 276). The goal is to guess nondeterministically a finite path $u_{0} u_{1} \ldots u_{n-1} v_{0} v_{1} \ldots v_{m-1}$ in $T S \otimes \mathcal{G}_{\varphi}$ and to check whether the components of $\mathcal{G}_{\varphi}$ in the infinite path

$$
u_{0} u_{1} \ldots u_{n-1}\left(v_{0} v_{1} \ldots v_{m-1}\right)^{\omega}
$$

constitute an accepting run in $\mathcal{G}_{\varphi}$. This, of course, requires that $v_{0}$ is a successor of $v_{m-1}$. The states $u_{i}, v_{j}$ in the infinite path in $T S \otimes \mathcal{G}_{\varphi}$ are pairs consisting of a state in $T S$ and an elementary set of formulae. For the lengths of the prefix $u_{0} \ldots u_{n-1}$ and the cycle $v_{0} v_{1} \ldots v_{m-1} v_{m}$ we can deal with $n \leqslant k$ and $m \leqslant k \cdot|\varphi|$ where $k$ is the number of reachable states in $T S \otimes \mathcal{G}_{\varphi}$. (Note that $|\varphi|$ is an upper bound for the number of acceptance sets in $\mathcal{G}_{v}$ arphi.) An upper bound for the value $k$ is given by

$$
K=N_{T S} \cdot 2^{N_{\varphi}}
$$

where $N_{T S}$ denotes the number of states in $T S$ and $N_{\varphi}=|\operatorname{closure}(\varphi)|$. Note that

$$
K=\mathcal{O}(\operatorname{size}(T S) \cdot \exp (|\varphi|))
$$

The algorithm now works as follows. We first nondeterministically choose two natural numbers $n$, $m$ with $n \leqslant K$ and $m \leqslant K \cdot|\varphi|$ (by guessing $\lceil\log K\rceil=\mathcal{O}(\log (\operatorname{size}(T S)) \cdot|\varphi|)$
bits for $n$ and $\lceil\log K\rceil+\lceil\log |\varphi|\rceil=\mathcal{O}(\log (\operatorname{size}(T S)) \cdot|\varphi|)$ bits for $m)$. Then the algorithm guesses nondeterministically a sequence $u_{0} \ldots u_{n-1}, u_{n} \ldots u_{n+m}$ where the $u_{i}$ 's are pairs $\left\langle s_{i}, B_{i}\right\rangle$ consisting of a state $s_{i}$ in $T S$ and a subset $B_{i}$ of closure $(\varphi)$. For each such state $u_{i}=\left\langle s_{i}, B_{i}\right\rangle$, the algorithm checks whether

1. $s_{i}$ is a successor of $s_{i-1}$, provided that $i \geqslant 1$,
2. $B_{i}$ is elementary,
3. $B_{i} \cap A P=L\left(s_{i}\right)$,
4. $B_{i} \in \delta\left(B_{i-1}, L\left(s_{i}\right)\right)$, provided that $i \geqslant 1$.

For $i=0$, the algorithm checks whether $s_{0} \in I$ is an initial state of $T S$ and whether $B_{0} \in$ $\delta\left(B, L\left(s_{0}\right)\right)$ for some elementary set $B$ which contains $\varphi$. Here, $\delta$ denotes the transition relation of the GNBA $\mathcal{G}_{\varphi}$. (Recall that the sets $B$ where $\varphi \in B$ are the initial states in the GNBA for $\varphi$.) Conditions 1-4 are local and simple to check. If one of these four conditions is violated for some $i$, the algorithm rejects and halts. Otherwise $u_{0} \ldots u_{n+m}$ is a finite path in $T S \otimes \mathcal{G}_{\varphi}$. We finally check whether $u_{n}$ agrees with the last state $u_{n+m}$. Again, the algorithm rejects and halts if this condition does not hold. Otherwise, $u_{0} \ldots u_{n-1}\left(u_{n} \ldots u_{n+m-1}\right)^{\omega}$ is an infinite path in $T S \otimes \mathcal{G}_{\varphi}$, and it finally amounts to check whether the acceptance condition of $\mathcal{G}_{\varphi}$ is fulfilled. This means we have to verify that whenever $\psi_{1} \cup \psi_{2} \in B_{i}$ for some $i \in\{n, \ldots, n+m-1\}$, then there is some $j \in\{n, \ldots, n+$ $m-1\}$ such that $\psi_{2} \in B_{j}$. If this condition holds, then the algorithm terminates with the answer "yes".

This algorithm is correct since if $T S$ has a path where $\varphi$ holds, then there is a computation of the above algorithm that returns "yes". Otherwise, i.e., if TS does not have a path where $\varphi$ holds, then all computations of the algorithm are rejecting.

It remains to explain how the sketched algorithm can be realized such that the memory requirements are polynomial in the size of $T S$ and length of $\varphi$. Although the length $n+m$ of the path $u_{0} \ldots u_{n+m}$ might be exponentially in the length of $\varphi$ (note that, e.g., $n=K$ is possible and that $K$ grows exponentially in the length of $\varphi$ ), this procedure can be realized with only polynomial space requirements. This is due to the observation that for checking the above conditions 1-4 for $u_{i}=\left\langle s_{i}, B_{i}\right\rangle$, we only need state $u_{i-1}=\left\langle s_{i-1}, B_{i-1}\right\rangle$. Thus, there is no need to store all states $u_{j}$ for $0 \leqslant j \leqslant n+m$. Instead, the actual and previous states are sufficient. Moreover, to verify that $\mathcal{G}_{\varphi}$ 's acceptance condition holds, we only need to remember the subformulae $\psi_{1} \cup \psi_{2}$ of $\varphi$ that are contained in some of the sets $B_{n}, \ldots B_{n+m-1}$, and the subformulae $\psi_{2}$ that appear on the right hand side of an until subformula of $\varphi$ and are contained in some of the sets $B_{n}, \ldots B_{n+m-1}$. This additional
information requires $\mathcal{O}(|\varphi|)$ space. Thus, the above nondeterministic algorithm can be realized in a polynomially space-bounded way.

By Lemma 5.47 and Theorem 5.46 we get:

## Theorem 5.48.

The LTL model-checking problem is PSPACE-complete.

### 5.2.2 LTL Satisfiability and Validity Checking

The last section of this chapter considers the satisfiability problem and the validity problem for LTL. The satisfiability problem is: for a given LTL formula $\varphi$, does there exist a model for which $\varphi$ holds? That is, do we have $\operatorname{Words}(\varphi) \neq \varnothing$ ? Satisfiability can be solved by constructing an NBA $\mathcal{A}_{\varphi}$ for LTL formula $\varphi$. In this way, the existence of an infinite word $\sigma \in \operatorname{Words}(\varphi)=\mathcal{L}_{\omega}\left(\mathcal{A}_{\varphi}\right)$ can be established. The emptiness problem for NBA $\mathcal{A}$, i.e., whether $\mathcal{L}_{\omega}(\mathcal{A})=\varnothing$ or not, can be solved by means of a technique similar to persistence checking, see Algorithm 12. In addition to an affirmative response, a prefix of a word $\sigma \in \mathcal{L}_{\omega}(\mathcal{A})=\operatorname{Words}(\varphi)$ can be provided similar to a counterexample for model checking LTL.

```
Algorithm 12 Satisfiability checking for LTL
Input: LTL formula \(\varphi\) over AP
Output: "yes" if \(\varphi\) is satisfiable. Otherwise "no".
```

Construct an NBA $\mathcal{A}=\left(Q, 2^{A P}, \delta, Q_{0}, F\right)$ with $\mathcal{L}_{\omega}(\mathcal{A})=\operatorname{Words}(\varphi)$
(* Check whether $\mathcal{L}_{\omega}(\mathcal{A})=\varnothing$. $\left.{ }^{*}\right)$
Perform a nested DFS to determine whether there exists a state $q \in F$ reachable from $q_{0} \in Q_{0}$ and that lies on a cycle

If so, then return "yes". Otherwise, "no".

Given that satisfiability for LTL can be tackled using similar techniques as for model checking LTL, let us now consider the validity problem. Formula $\varphi$ is valid whenever $\varphi$ holds under all interpretations, i.e., $\varphi \equiv$ true. For LTL formula $\varphi$ over AP we have $\varphi$ is valid if and only if $\operatorname{Words}(\varphi)=\left(2^{A P}\right)^{\omega}$. The validity of $\varphi$ can be established by using the
observation that $\varphi$ is valid if and only if $\neg \varphi$ is not satisfiable. Hence, to algorithmically check whether $\varphi$ is obtained, one constructs an NBA for $\neg \varphi$ and applies the satisfiablity algorithm (see Algorithm 12) to $\neg \varphi$.

The outlined LTL satisfiability algorithm has a runtime that is exponential in the length of $\varphi$. The following result shows that an essentially more efficient technique cannot be achieved as both the validity and satisfiability problems are PSPACE-hard. In fact, both problems are even PSPACE-complete. Membership to PSPACE for the LTL satisfiability problem can be shown by providing a nondeterministic polynomially space-bounded algorithm that guesses a finite run in the GNBA $\mathcal{G}_{\varphi}$ for the given formula $\varphi$ and checks whether this finite run is a prefix of an accepting run in $\mathcal{G}_{\varphi}$. The details are skipped here since they are very similar to the algorithm we provided for the existential LTL model checking problem (see Lemma 5.47 on page 294). The fact that the LTL validity problem belongs to PSPACE can be derived from the observations that $\varphi$ is valid iff $\neg \varphi$ is not satisfiable and that PSPACE is closed under complementation. We now focus on the proof for the PSPACE-hardness.

## Theorem 5.49. LTL Satisfiability and Validity (Lower Bound)

The satisfiability and validity problems for LTL are PSPACE-hard.

Proof: Since satisfiability and validity are complementary in the sense that $\varphi$ is satisfiable if and only if $\neg \varphi$ is not valid, and vice versa, it suffices to show the PSPACE-hardness of the satisfiability problem. This is done by providing a polynomial reduction from the existential variant of the LTL model-checking problem (see the proof of Theorem 5.46 on page 290) to the satisfiability problem for LTL.

Let $T S=(S$, Act, $\rightarrow, I, A P, L)$ be a finite transition system and $\varphi$ an LTL formula over $A P$. The goal is to construct an LTL formula $\psi$ such that $\psi$ is satisfiable if and only if there is a path $\pi$ in $T S$ with $\pi \models \varphi$. Besides, $\psi$ should be constructed in polynomial time.

The atomic propositions in $\psi$ are elements in $A P^{\prime}=A P \uplus S$. For any state $s \in S$ let

$$
\Phi_{s}=\bigwedge_{a \in L(s)} a \wedge \bigwedge_{a \notin L(s)} \neg a
$$

The formula $\Phi_{s}$ can be viewed as a characteristic formula for the labeling of $s$, since $s^{\prime} \models \Phi_{s}$ if and only if $L(s)=L\left(s^{\prime}\right)$, for any $s^{\prime} \in S$. However, for the LTL satisfiability problem, there is no fixed transition system and $\Phi_{s}$ can hold also for other states (in another transition system). For $s \in S$ and $T \subseteq S$, let $\Psi_{T}=\bigvee_{t \in T} t$ be the characteristic formula for the set $T$. Let

$$
\psi_{s}=s \rightarrow\left(\Phi_{s} \wedge \bigcirc \Psi_{\operatorname{Post}(s)}\right)
$$

assert that in state $s$, the labels $L(s)$ hold, and that transitions exist to any of its immediate successors Post(s). Let

$$
\Xi=\bigvee_{s \in S}\left(s \wedge \bigwedge_{t \in S \backslash\{s\}} \neg t\right)
$$

Stated in words, $\Xi$ asserts that exactly one of the atomic propositions $s \in A P^{\prime}$ holds. These definitions constitute the ingredients for the definition of $\psi$. For set $I$ of initial states, let

$$
\psi=\Psi_{I} \wedge \square \Xi \wedge \square \Psi_{S} \wedge \bigwedge_{s \in S} \square \psi_{s} \wedge \varphi
$$

It follows directly that $\psi$ can be derived from $T S$ and $\varphi$ in polynomial time. It remains to show that

$$
\exists \pi \in \operatorname{Paths}(T S) . \pi \models \varphi \quad \text { if and only if } \quad \psi \text { is satisfiable. }
$$

$\Rightarrow$ : Let $\pi=s_{0} s_{1} s_{2} \ldots$ be an initial, infinite path in $T S$ with $\pi \models \varphi$. Now consider $\pi$ as a path in the transition system $T S^{\prime}$ that agrees with $T S$ but uses the extended labeling function $L^{\prime}(s)=L(s) \cup\{s\}$. Then, $\pi \vDash \Psi_{I}$, since $s_{0} \in I$. In addition, $\pi \vDash \square \Xi$ and $\pi \vDash \square \psi_{s}$ since $\Xi$ and $\psi_{s}$ hold in all states of $T S^{\prime}$. Thus, $\pi \models \psi$. Hence, $\pi$ is a witness for the satisfiability of $\psi$.
$\Leftarrow$ : Assume $\psi$ is satisfiable. Let $A_{0} A_{1} A_{2} \ldots$ be an infinite word over the alphabet $2^{A P^{\prime}}$ with $A_{0} A_{1} A_{2} \ldots \in \operatorname{Words}(\psi)$. Since

$$
A_{0} A_{1} A_{2} \ldots=\square \Xi
$$

there is a unique state sequence $\pi=s_{0} s_{1} s_{2} \ldots$ in $T S$ with $s_{i} \in A_{i}$ for all $i \geqslant 0$. Since $A_{0} A_{1} A_{2} \ldots \vDash \Psi_{I}$, we get $s_{0} \in I$. As

$$
A_{0} A_{1} A_{2} \ldots \models \square \bigwedge_{s \in S} \psi_{s}
$$

we have $A_{i} \cap A P=L\left(s_{i}\right)$ and $s_{i+1} \in \operatorname{Post}\left(s_{i}\right)$ for all $i \geqslant 0$. This yields that $\pi$ is a path in TS and $\pi \models \varphi$.

### 5.3 Summary

- Linear Temporal Logic (LTL) is a logic for formalizing path-based properties.
- LTL formulae can be transformed algorithmically into nondeterministic Büchi automata (NBA). This transformation can cause an exponential blowup.
- The presented algorithm for the construction of an NBA for a given LTL formula $\varphi$ relies on first constructing a GNBA for $\varphi$, which is then transformed into an equivalent NBA.
- The GNBA for $\varphi$ encodes the semantics of propositional logic and the semantics of the next-step operator in its transitions. Based on the expansion law, the meaning of until is split into local requirements (encoded by the states of a GNBA), next-step requirements (encoded by the transitions of the GNBA), and a fairness condition (encoded by the acceptance sets of the GNBA).
- LTL formulae describe $\omega$-regular LT properties, but do not have the same expressiveness as $\omega$-regular languages.
- The LTL model-checking problem can be solved by a nested depth-first search in the product of the given transition system and an NBA for the negated formula.
- The time complexity of the automata-based model-checking algorithm for LTL is linear in the size of the transition system and exponential in the length of the formula.
- Fairness assumptions can be described by LTL formulae. The model-checking problem for LTL with fairness assumptions is reducible to the standard LTL modelchecking problem.
- The LTL model-checking problem is PSPACE-complete.
- Satisfiability and validity of LTL formulae can be solved via checking emptiness of nondeterministic Büchi automata. The emptiness check can be determined by a nested DFS that checks the existence of a reachable cycle containing an accept state. Both problems are PSPACE-complete.


### 5.4 Bibliographic Notes

Linear temporal logic. Based on prior work on modal logics and temporal modalities [270, 345, 244, 230], Pnueli introduced (linear) temporal logics for reasoning about reactive systems in the late seventies in his seminal paper [337]. Since then, a variety of variants and extensions of LTL have been investigated, such as Lamport's Temporal Logic of Actions (TLA) [260] and LTL with past operators [159, 274, 262]. LTL forms the basis for the recently standardized industrial property specification language PSL [136]. The past extension of LTL does not change the expressiveness of LTL, but can be helpful for specification convenience and modular reasoning. For several properties, the use of past operators may lead to (exponentially) more succinct formulae than in ordinary LTL. To
cover the full class of $\omega$-regular LT properties, Vardi and Wolper introduced an extension of LTL by automata formulae [424, 425, 411, 412].

LTL model checking. Vardi and Wolper also developed the automata-based model checking algorithm for LTL presented in this chapter. The presented algorithm to construct an NBA from a given LTL formula is, in our opinion, the simplest and most intuitive one. In the meantime, various alternative techniques have been developed that generate more compact NBAs or that attempt to minimize a given NBA, see e.g. [166, 110, 148, 375, 162, 149, 167, 157, 389, 369]. Alternative LTL model-checking algorithms that do not use Büchi automata, but a so-called tableau for the LTL formula, were presented by Lichtenstein and Pnueli [273] and Clarke, Grumberg, and Hamaguchi [88]. The results about the complexity of LTL model checking and the satisfiability problem are due to Sistla and Clarke [372].

There is a variety of surveys and textbooks; see, e.g., $[245,138,173,283,158,284,92,219$, 379, 365], where several other aspects of LTL and related logics, such as deductive proof systems, alternative model-checking algorithms, or more details about the expressiveness, are treated.

Examples. The garbage collection algorithm presented in Example 5.31 is due to Ben-Ari [41]. Several leader election protocols that fit into the shape of Example 5.13 have been suggested; see, e.g., [280].

LTL model checkers. SPIN is the most well-known LTL model checker and has been developed by Holzmann [209]. Transition systems are described in the modeling language Promela, and LTL formulae are checked using the algorithm advocated by Gerth et al. [166]. LTL model checking using a tableau construction is supported by NuSMV [83].

### 5.5 Exercises

Exercise 5.1. Consider the following transition system over the set of atomic propositions $\{a, b\}$ :


Indicate for each of the following LTL formulae the set of states for which these formulae are
fulfilled:
(a) $\bigcirc a$
(d) $\square \diamond a$
(b) $\bigcirc \bigcirc \bigcirc$
(e) $\square(b \cup a)$
(c) $\square b$
(f) $\diamond(a \cup b)$

Exercise 5.2. Consider the transition system $T S$ over the set of atomic propositions $A P=$ $\{a, b, c\}$ :


Decide for each of the LTL formulae $\varphi_{i}$ below, whether $T S \vDash \varphi_{i}$ holds. Justify your answers! If $T S \not \vDash \varphi_{i}$, provide a path $\pi \in \operatorname{Paths}(T S)$ such that $\pi \not \vDash \varphi_{i}$.

$$
\begin{aligned}
& \varphi_{1}=\Delta \square c \\
& \varphi_{2}=\square \Delta c \\
& \varphi_{3}=\bigcirc \neg c \rightarrow \bigcirc \bigcirc c \\
& \varphi_{4}=\square a \\
& \varphi_{5}=a \cup \square(b \vee c) \\
& \varphi_{6}=(\bigcirc \bigcirc b) \cup(b \vee c)
\end{aligned}
$$

Exercise 5.3. Consider the sequential circuit in Figure 5.24 and let $A P=\left\{x, y, r_{1}, r_{2}\right\}$. Provide LTL formulae for the following properties:


Figure 5.24: Circuit for Exercise 5.3.
(a) "It is impossible that the circuit outputs two successive 1s."
(b) "Whenever the input bit is 1 , in at most two steps the output bit will be 1."
(c) "Whenever the input bit is 1 , the register bits do not change in the next step."
(d) "Register $r_{1}$ has infinitely often the value 1."

Determine which of these properties are satisfied for the initial register evaluation where $r_{1}=0$ and $r_{2}=0$ ? Justify your answers.

Exercise 5.4. Suppose we have two users, Peter and Betsy, and a single printer device Printer. Both users perform several tasks, and every now and then they want to print their results on the Printer. Since there is only a single printer, only one user can print a job at a time. Suppose we have the following atomic propositions for Peter at our disposal:

- Peter.request $::=$ indicates that Peter requests usage of the printer;
- Peter.use $::=$ indicates that Peter uses the printer;
- Peter.release $::=$ indicates that Peter releases the printer.

For Betsy, similar predicates are defined. Specify in LTL the following properties:
(a) Mutual exclusion, i.e., only one user at a time can use the printer.
(b) Finite time of usage, i.e., a user can print only for a finite amount of time.
(c) Absence of individual starvation, i.e., if a user wants to print something, he/she eventually is able to do so.
(d) Absence of blocking, i.e., a user can always request to use the printer
(e) Alternating access, i.e., users must strictly alternate in printing.

Exercise 5.5. Consider an elevator system that services $N>0$ floors numbered 0 through $N-1$. There is an elevator door at each floor with a call-button and an indicator light that signals whether or not the elevator has been called. For simplicity consider $N=4$. Present a set of atomic propositions - try to minimize the number of propositions - that are needed to describe the following properties of the elevator system as LTL formulae and give the corresponding LTL formulae:
(a) The doors are "safe", i.e., a floor door is never open if the elevator is not present at the given floor.
(b) A requested floor will be served sometime.
(c) Again and again the elevator returns to floor 0 .
(d) When the top floor is requested, the elevator serves it immediately and does not stop on the way there.

Exercise 5.6. Which of the following equivalences are correct? Prove the equivalence or provide a counterexample that illustrates that the formula on the left and the formula on the right are not equivalent.
(a) $\square \varphi \rightarrow \diamond \psi \equiv \varphi \mathrm{U}(\psi \vee \neg \varphi)$
(b) $\diamond \square \varphi \rightarrow \square \diamond \psi \equiv \square(\varphi \cup(\psi \vee \neg \varphi))$
(c) $\square \square(\varphi \vee \neg \psi) \equiv \neg \diamond(\neg \varphi \wedge \psi)$
(d) $\diamond(\varphi \wedge \psi) \equiv \diamond \varphi \wedge \diamond \psi$
(e) $\square \varphi \wedge \bigcirc \diamond \varphi \equiv \square \varphi$
(f) $\diamond \varphi \wedge \bigcirc \square \varphi \equiv \diamond \varphi$
(g) $\square \diamond \varphi \rightarrow \square \diamond \psi \equiv \square(\varphi \rightarrow \diamond \psi)$
(h) $\neg\left(\varphi_{1} \cup \varphi_{2}\right) \equiv \neg \varphi_{2} \mathrm{~W}\left(\neg \varphi_{1} \wedge \neg \varphi_{2}\right)$
(i) $\bigcirc \diamond \varphi_{1} \equiv \diamond \bigcirc \varphi_{2}$
(j) $\left(\diamond \square \varphi_{1}\right) \wedge\left(\diamond \square \varphi_{2}\right) \equiv \diamond\left(\square \varphi_{1} \wedge \square \varphi_{2}\right)$
(k) $\left(\varphi_{1} \cup \varphi_{2}\right) \cup \varphi_{2} \equiv \varphi_{1} \cup \varphi_{2}$

Exercise 5.7. Let $\varphi$ and $\psi$ be LTL formulae. Consider the following new operators:
(a) "At next" $\varphi \mathrm{N} \psi$ : at the next time where $\psi$ holds, $\varphi$ also holds.
(b) "While" $\varphi \mathrm{W} \psi: \varphi$ holds as least as long as $\psi$ does.
(c) "Before" $\varphi \mathrm{B} \psi$ : if $\psi$ holds sometime, $\varphi$ does so before.

Make the definitions of these informally explained operators precise by providing LTL formulae that formalize their intuitive meanings.

EXERCISE 5.8. We consider the release operator R which was defined by $\varphi \mathrm{R} \psi \stackrel{\text { def }}{=} \neg(\neg \varphi \mathrm{U} \neg \psi)$; see Section 5.1.5 on page 252 ff .
(a) Prove the expansion law $\varphi_{1} \mathrm{R} \varphi_{2} \equiv \varphi_{2} \wedge\left(\varphi_{1} \vee \bigcirc\left(\varphi_{1} \wedge \varphi_{2}\right)\right)$.
(b) Prove that $\varphi \mathrm{R} \psi \equiv(\neg \varphi \wedge \psi) \mathrm{W}(\varphi \wedge \psi)$.
(c) Prove that $\varphi_{1} W \varphi_{2} \equiv\left(\neg \varphi_{1} \vee \varphi_{2}\right) \mathrm{R}\left(\varphi_{1} \vee \varphi_{2}\right)$.
(d) Prove that $\varphi_{1} \cup \varphi_{2} \equiv \neg\left(\neg \varphi_{1} \mathrm{R} \neg \varphi_{2}\right)$.

Exercise 5.9. Consider the LTL formula

$$
\varphi=\neg((\square a) \rightarrow((a \wedge \neg c) \cup \neg(\bigcirc b))) \wedge \neg(\neg a \vee \bigcirc \diamond c) .
$$

Transform $\varphi$ into an equivalent LTL formula in PNF
(a) using the weak-until operator W ,
(b) using the release operator R .

Exercise 5.10. Provide an example for a sequence $\left(\varphi_{n}\right)$ of LTL formulae such that the LTL formula $\psi_{n}$ is in weak-until $\operatorname{PNF}, \varphi_{n} \equiv \psi_{n}$, and $\psi_{n}$ is exponentially longer than $\varphi_{n}$. Use the transformation rules in Section 5.1.5


Figure 5.25: Transition system for Exercise 5.11.

Exercise 5.11. Consider the transition system $T S$ in Figure 5.25 with the set $A P=\{a, b, c\}$ of atomic propositions. Note that this is a single transition system with two initial states. Consider the LTL fairness assumption

$$
\text { fair }=(\square \diamond(a \wedge b) \rightarrow \square \diamond \neg c) \wedge(\diamond \square(a \wedge b) \rightarrow \square \diamond \neg b) \text {. }
$$

Questions:
(a) Determine the fair paths in TS, i.e., the initial, infinite paths satisfying fair
(b) For each of the following LTL formulae:

$$
\begin{aligned}
\varphi_{1} & =\diamond \square a \\
\varphi_{2} & =\bigcirc \neg a \longrightarrow \Delta \square a \\
\varphi_{3} & =\square a \\
\varphi_{4} & =b \cup \square \neg b \\
\varphi_{5} & =b \mathrm{~W} \square \neg b \\
\varphi_{6} & =\bigcirc \bigcirc b \cup \square \neg b
\end{aligned}
$$

determine whether $T S \models_{\text {fair }} \varphi_{i}$. In case $T S \not \models_{\text {fair }} \varphi_{i}$, indicate a path $\pi \in \operatorname{Paths}(T S)$ for which $\pi \not \vDash \varphi_{i}$.

Exercise 5.12. Let $\varphi=(a \rightarrow \bigcirc \neg b) \mathrm{W}(a \wedge b)$ and $P=\operatorname{Words}(\varphi)$ where $A P=\{a, b\}$.
(a) Show that $P$ is a safety property.
(b) Define an NFA $\mathcal{A}$ with $\mathcal{L}(\mathcal{A})=\operatorname{BadPref}(P)$.
(c) Now consider $P^{\prime}=\operatorname{Words}((a \rightarrow \bigcirc \neg b) \cup(a \wedge b))$. Decompose $P^{\prime}$ into a safety property $P_{\text {safe }}$ and a liveness property $P_{\text {live }}$ such that

$$
P^{\prime}=P_{\text {safe }} \cap P_{\text {live }} .
$$

Show that $P_{\text {safe }}$ is a safety and that $P_{\text {live }}$ is a liveness property.


Figure 5.26: Transition system for Exercise 5.14.
Exercise 5.13. Provide an NBA for each of the following LTL formulae:

$$
\square(a \vee \neg \bigcirc b) \quad \text { and } \quad \diamond a \vee \square \diamond(a \leftrightarrow b) \quad \text { and } \quad \bigcirc \bigcirc(a \vee \diamond \square b) .
$$

Exercise 5.14. Consider the transition system TS in Figure 5.26 with the atomic propositions $\{a, b\}$. Sketch the main steps of the LTL model-checking algorithm applied to $T S$ and the LTL formulae

$$
\varphi_{1}=\square \diamond a \rightarrow \square \diamond b \quad \text { and } \quad \varphi_{2}=\diamond(a \wedge \bigcirc a)
$$

To that end, carry out the following steps:
(a) Depict an NBA $A_{i}$ for $\neg \varphi_{i}$.
(b) Depict the reachable fragment of the product transition system $T S \otimes \mathcal{A}_{i}$.
(c) Explain the main steps of the nested DFS in $T S \otimes \mathcal{A}_{i}$ by illustrating the order in which the states are visited during the "outer" and "inner" DFS.


Figure 5.27: GNBA for Exercise 5.15.
(d) If $T S \not \vDash \varphi_{i}$, provide the counterexample resulting from the nested DFS.

Exercise 5.15. Consider the GNBA $\mathcal{G}$ in Figure 5.27 with the alphabet $\Sigma=2^{\{a, b\}}$ and the set $\mathcal{F}=\left\{\left\{q_{1}, q_{3}\right\},\left\{q_{2}\right\}\right\}$ of accepting sets.
(a) Provide an LTL formula $\varphi$ with $\operatorname{Words}(\varphi)=\mathcal{L}_{\omega}(\mathcal{G})$. Justify your answer.
(b) Depict the NBA $\mathcal{A}$ with $\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}(\mathcal{G})$.

Exercise 5.16. Depict a GNBA $\mathcal{G}$ over the alphabet $\Sigma=2^{\{a, b, c\}}$ such that

$$
\mathcal{L}_{\omega}(\mathcal{G})=\operatorname{Words}((\square \diamond a \rightarrow \square \diamond b) \wedge \neg a \wedge(\neg a \mathrm{~W} c)) .
$$

Exercise 5.17. Let $\psi=\square(a \leftrightarrow \bigcirc \neg a)$ and $A P=\{a\}$.
(a) Show that $\psi$ can be transformed into the following equivalent basic LTL formula

$$
\varphi=\neg[\operatorname{true} \cup(\neg(a \wedge \bigcirc \neg a) \wedge \neg(\neg a \wedge \neg \bigcirc \neg a))] .
$$

The basic LTL syntax is given by the following context-free grammar:

$$
\varphi::=\operatorname{true}|a| \varphi_{1} \wedge \varphi_{2}|\neg \varphi| \bigcirc \varphi \mid \varphi_{1} \cup \varphi_{2} .
$$

(b) Compute all elementary sets with respect to closure( $\varphi$ ) (Hint: There are six elementary sets.)
(c) Construct the GNBA $\mathcal{G}_{\varphi}$ with $\mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)=\operatorname{Words}(\varphi)$. To that end:
(i) Define its set of initial states and its acceptance component.
(ii) For each elementary set $B$, define $\delta(B, B \cap A P)$.

Exercise 5.18. Let $A P=\{a\}$ and $\varphi=(a \wedge \bigcirc a) \cup \neg a$ an LTL formula over $A P$.
(a) Compute all elementary sets with respect to $\varphi$.
(Hint: There are five elementary sets.).
(b) Construct the GNBA $\mathcal{G}_{\varphi}$ such that $\mathcal{L}_{\omega}\left(\mathcal{G}_{\varphi}\right)=\operatorname{Words}(\varphi)$.

Exercise 5.19. Consider the formula $\varphi=a \cup(\neg a \wedge b)$ and let $\mathcal{G}$ be the GNBA for $\varphi$ that is obtained through the construction explained in the proof of Theorem 4.56. What are the initial states in $\mathcal{G}$ ? What are the accept states in $\mathcal{G}$ ? Provide an accepting run for the word $\{a\}\{a\}\{a, b\}\{b\}^{\omega}$. Explain why there are no accepting runs for the words $\{a\}^{\omega}$ and $\{a\}\{a\}\{a, b\}^{\omega}$. (Hint: The answers to these questions can be given without depicting $\mathcal{G}$.)

Exercise 5.20.
We consider the LTL formula $\varphi=\square(a \rightarrow(\neg b \cup(a \wedge b)))$ over the set $A P=\{a, b\}$ of atomic propositions and we want to check $T S \models \varphi$ for $T S$ outlined on the right.
(a) To check $T S \models \varphi$, convert $\neg \varphi$ into an equivalent LTL formula $\psi$ which is constructed according to the following grammar:

$$
\Phi::=\text { true } \mid \text { false }|a| b|\Phi \wedge \Phi| \neg \Phi|\bigcirc \Phi| \Phi \cup \Phi .
$$



Then construct closure $(\psi)$.
(b) Give the elementary sets w.r.t. closure $(\psi)$ !
(c) Construct the GNBA $\mathcal{G}_{\psi}$.
(d) Construct an NBA $\mathcal{A}_{\neg \varphi}$ directly from $\neg \varphi$, i.e., without relying on $\mathcal{G}_{\psi}$. (Hint: Four states suffice.)
(e) Construct $T S \otimes \mathcal{A}_{\neg \varphi}$.
(f) Use the nested DFS algorithm to check $T S \models \varphi$. Therefore, sketch the algorithm's main steps and interpret its outcome!

Exercise 5.21. The construction of a $\mathcal{G}$ from a given LTL formula in the proof of Theorem 4.56 assumes an LTL formula that only uses the basic temporal modalities $\bigcirc$ and U . The derived operators $\diamond, \square, W$ and R can be treated by syntactic replacements of their definitions. Alternatively, and more efficient, is to treat them as basic modalities and to allow for formulae $\Delta \psi$, $\square \psi, \varphi_{1} \mathrm{~W} \varphi_{2}$ and $\varphi_{1} \mathrm{R} \varphi_{2}$ as elements of elementary sets of formulae and redefine the components of the constructed GNBA.

Explain which modifications are necessary for such a "direct" treatment of $\diamond$ (eventually), (always), W (weak-until), and R (release). That is, which additional conditions do the elementary sets, the transition function $\delta$, and the set $\mathcal{F}$ of accepting sets have to fulfill?

Exercise 5.22. Let $T S=\left(S, A c t, \rightarrow, S_{0}, A P, L\right)$ be a finite transition system without terminal states and let wfair $=\diamond \square b_{1} \rightarrow \square \diamond b_{2}$ be a weak LTL fairness assumption with $b_{1}, b_{2} \in A P$. Explain how the nested DFS can be modified to check directly whether $T S \models_{\text {wfair }} \diamond \square a$ (where $a \in A P$ ), that is, without using the transformation $T S \vDash \models_{\text {wfair }} \diamond \square a$ iff $T S \models($ wfair $\rightarrow \Delta \square a)$.

Exercise 5.23. Which of the following LTL formulae $\varphi_{i}$ are representable by a deterministic Büchi automaton?

$$
\varphi_{1}=\square(a \rightarrow \Delta b), \quad \varphi_{2}=\neg \varphi_{1} .
$$

Explain your answer.

Exercise 5.24. Check for the following LTL formula whether they are (i) satisfiable, and/or (ii) valid:
(a) $\bigcirc \bigcirc a \Rightarrow \bigcirc a$
(b) $\bigcirc(a \vee \diamond a) \Rightarrow \diamond a$
(c) $\square a \Rightarrow \neg \bigcirc(\neg a \wedge \square \neg a)$
(d) $(\square a) \cup(\diamond b) \Rightarrow \square(a \cup \diamond b)$
(e) $\Delta b \Rightarrow(a \cup b)$

## Practical Exercises

EXERCISE 5.25. Consider an arbitrary, but finite, number of identical processes ${ }^{2}$, that execute in parallel. Each process consists of a noncritical part and a critical part, usually called the critical section. In this exercise we are concerned with the verification of a mutual exclusion protocol, that is, a protocol that should ensure that at any moment of time at most one process (among the $N$ processes in our configuration) is in its critical section. There are many different mutual exclusion protocols developed in the literature. In this exercise we are concerned with Szymanski's protocol [384]. Assume there are $N$ processes for some fixed $N>0$. There is a global variable, referred to as flag, which is an array of length $N$, such that flag[ $i]$ is a value between 0 and 4 (for $0 \leqslant i<N$ ). The idea is that flag $[i]$ indicates the status of process $i$. The protocol executed by process $i$ looks as follows:

## 10: loop forever do <br> begin

11: Noncritical section
12: flag $[i]:=1$;

[^23]```
    13: wait until (flag[0] \(<3\) and flag[1] \(<3\) and \(\ldots\) and flag \([N-1]<3\) )
    14: flag \([i]:=3\);
    15: if \((\) flag \([0]=1\) or \(\operatorname{flag}[1]=1\) or \(\ldots\) or \(\operatorname{flag}[N-1]=1)\)
    then begin
            16: flag[i] := 2 ;
            17: wait until \((\) flag \([0]=4\) or \(\operatorname{flag}[1]=4\) or \(\ldots\) or \(\operatorname{flag}[N-1]=4)\)
    end
    18: flag \([i]:=4\);
    19: wait until (flag[0] \(<2\) and flag[1] \(<2\) and \(\ldots\) and \(\operatorname{flag}[i-1]<2\) )
    110: Critical section
    111: wait until \((f \operatorname{fag}[i+1] \in\{0,1,4\})\) and \(\ldots\) and \((f l a g[N-1] \in\{0,1,4\})\)
    112: flag \([i]:=0\);
end.
```

Before doing any of the exercises listed below, try first to informally understand what the protocol is doing and why it could be correct in the sense that mutual exclusion is ensured. If you are convinced of the fact that the correctness of this protocol is not easy to see - otherwise please inform me - then start with the following questions.

1. Model Szymanski's protocol in Promela. Assume that all tests on the global variable flag (such as the one in statement 13) are atomic. Look carefully at the indices of the variable flag used in the tests. Make the protocol description modular such that the number of processes can be changed easily.
2. Check for several values of $N(N \geqslant 2)$ that the protocol indeed ensures mutual exclusion. Report your results for $N$ equal to 4 .
3. The code that a process has to go through before reaching the critical section can be divided into several segments. We refer to statement 14 as the doorway, to segments 15,16 , and 17 , as the waiting room and to segments 18 through 112 (which contains the critical section) as the inner sanctum. You are requested to check the following basic claims using assertions. Give for each case the changes to your original Promela specification for Szymanski's protocol and present the verification results. In case of negative results, simulate the counterexample by means of guided simulation.
(a) Whenever some process is in the inner sanctum, the doorway is locked, that is, no process is at location 14.
(b) If a process $i$ is at 110,111 or 112 , then it has the least index of all the processes in the waiting room and the inner sanctum.
(c) If some process is at 112 , then all processes in the waiting room and in the inner sanctum must have flag value 4 .

Exercise 5.26. We assume $N$ processes in a ring topology, connected by unbounded queues. A process can only send messages in a clockwise manner. Initially, each process has a unique identifier ident (which is assumed to be a natural number). A process can be either active or
relaying. Initially a process is active. In Peterson's leader election algorithm (1982) each process in the ring carries out the following task:

```
active:
\(d:=\) ident;
do forever
begin
    /* start phase */
    send(d);
    receive(e);
    if \(e=i d e n t\) then announce elected;
    if \(d>e\) then \(\operatorname{send}(d)\) else \(\operatorname{send}(e)\);
    receive(f);
    if \(f=\) ident then announce elected;
    if \(e \geqslant \max (d, f)\) then \(d:=e\) else goto relay;
end
relay:
do forever
begin
    receive(d);
    if \(d=i d e n t\) then announce elected;
    send (d)
end
```

Solve the following questions concerning the leader election protocol:

1. Model Peterson's leader election protocol in Promela (avoid invalid end states).
2. Verify the following properties:
(a) There is always at most one leader.
(b) Eventually always a leader will be elected.
(c) The elected leader will be the process with the highest number.
(d) The maximum total amount of messages sent in order to elect a leader is at most $2 N\left\lfloor\log _{2} N\right\rfloor+N$.

Exercise 5.27. This exercise deals with a simple fault-tolerant communication protocol in which processes can fail. A failed process is still able to communicate, i.e., it is able to send and receive messages, but the content of its transmitted messages is unreliable. More precisely, a failed process can send messages with arbitrary content. A failed process is therefore also called unreliable.

We are given $N$ reliable processes (i.e., processes that have not failed and that are working as expected) and $K$ unreliable processes, where $N$ is larger than $3 \cdot K$ and $K$ is at least 0 . There is no way, a priori, to distinguish the reliable and the unreliable processes. All processes communicate
by means of exchanging messages. Each process has a local variable, which initially has a value, 0 or 1 . The following informally described protocol is aimed to be followed by the reliable processes, such that at the end of round $K+1$ we have

- eventually every reliable process has the same value in its local variable, and
- if all reliable processes have the same initial value, then their final value is the same as their common initial value.

The difficulty of this protocol is to establish these constraints in the presence of the $K$ unreliable processes!

Informal description of the protocol The following protocol is due to Berman and Garay [47]. Let the processes be numbered 1 through $N+K$. Processes communicate with each other in "rounds". Each round consists of two phases of message transmissions: In round $i, i>0$, in the first phase, every process sends its value to all processes (including itself); in the second phase, process $i$ sends the majority value it received in the first phase (for majority to be well-defined we assume that $N+K$ is odd) to all processes. If a process receives $N$, or more, instances of the same value in its first phase of the round, it sets its local variable to this value; otherwise, it sets its local variable to the value received (from process $i$ ) in the second phase of this round.

1. Model this protocol in Promela. Make the protocol description modular such that the number of reliable and unreliable processes can be changed easily. As the state space of your protocol model could be very large, instantiate your model with a single unreliable process and four reliable processes.
First hint: One of the main causes for the large state space is the model for the unreliable process, so try to keep this model as simple as possible. This can be achieved by, for instance, assuming that an unreliable process can only transmit arbitrary 0 or 1 values (and not any other value) and that a process always starts with a fixed initial value (and not with a randomly selected one). In addition, use atomic broadcast for message transmissions.
Second hint: It might be convenient (though not necessary) to use a matrix of size ( $N+K$ ). $(N+K)$ of channels for the communication structure. As Promela does not support multidimensional arrays, you could use the following construct (where $M$ equals $N+K$ ):
```
typedef Arraychan {
chan ch[M] = [1] of {bit}; /* M channels of size 1 */
}
Arraychan A[M]; /* a matrix A of MxM channels of size 1 */
```

Statement A[i].ch[j]!0 denotes an output of value 0 over the channel directed process from i to j . Similarly, statement $\mathrm{A}[\mathrm{i}] . \mathrm{ch}[\mathrm{j}]$ ?b denotes the receipt of a bit value stored in variable $b$ via the channel directed from process $i$ to $j$.
2. Formalize the two constraints of the protocol in LTL and convert these into never claims.
3. Check the two temporal logic properties by Spin and hand in the verification output generated by Spin.
4. Show that the requirement $N>3 \cdot K$ is essential, by, for instance, changing the configuration of your system such that $N \leqslant 3 \cdot K$ and checking that for this configuration the first aforementioned constraint is violated. Create the shortest counterexample (select the shortest trail in the advanced verification options) and perform a guided simulation of this undesired scenario. Hand in the counterexample you found and give an explanation of it.

## Chapter 6

## Computation Tree Logic

This chapter introduces Computation Tree Logic (CTL), a prominent branching temporal logic for specifying system properties. In particular, the syntax and semantics of CTL are presented, a comparison to Linear Temporal Logic (LTL) is provided, and the issue of fairness in CTL is treated. CTL model checking is explained in detail. First, the core recursive algorithm is presented that is based on a bottom-up traversal of the parse tree of the formula at hand. The foundations of this algorithm are discussed and the adaptations needed to incorporate fairness are detailed. This is followed by an algorithm for the generation of counterexamples. The chapter is concluded by presenting a model-checking algorithm for CTL*, a branching-time logic that subsumes both LTL and CTL.

### 6.1 Introduction

Pnueli [337] has introduced linear temporal logic for the specification and verification of reactive systems. LTL is called linear, because the qualitative notion of time is path-based and viewed to be linear: at each moment of time there is only one possible successor state and thus each time moment has a unique possible future. Technically speaking, this follows from the fact that the interpretation of LTL formulae is defined in terms of paths, i.e., sequences of states.

Paths themselves, though, are obtained from a transition system that might be branching: a state may have several, distinct direct successor states, and thus several computations may start in a state. The interpretation of LTL-formulae in a state requires that a formula $\varphi$ holds in state $s$ if all possible computations that start in $s$ satisfy $\varphi$. The universal
quantification over all computations that is implicit in the LTL semantics can also be made explicit in the formula, e.g.:

$$
s \models \forall \varphi \text { if and only if } \pi \models \varphi \text { for all paths } \pi \text { starting in } s
$$

In LTL, we thus can state properties over all possible computations that start in a state, but not easily about some of such computations. To some extent this may be overcome by exploiting the duality between universal and existential quantification. For instance, to check whether there exists some computation starting in $s$ that satisfies $\varphi$ we may check whether $s \models \forall \neg \varphi$; if this formula is not satisfied by $s$, then there must be a computation that meets $\varphi$, otherwise they should all refute $\varphi$.

For more complicated properties, like "for every computation it is always possible to return to the initial state", this is, however, not possible. A naive attempt would be to require $\square \diamond$ start for every computation, i.e., $s \models \forall \square \diamond$ start, where the proposition start uniquely identifies the initial state. This is, however, too strong as it requires a computation to always return to the initial state, not just possibly. Other attempts to specify the intended property also fail, and it even turns out to be the case that the property cannot be specified in LTL.

To overcome these problems, in the early eighties another strand of temporal logics for specification and verification purposes was introduced by Clarke and Emerson [86]. The semantics of this kind of temporal logic is not based on a linear notion of time - an infinite sequence of states - but on a branching notion of time - an infinite tree of states. Branching time refers to the fact that at each moment there may be several different possible futures. Each moment of time may thus split into several possible futures. Due to this branching notion of time, this class of temporal logic is known as branching temporal logic. The semantics of a branching temporal logic is defined in terms of an infinite, directed tree of states rather than an infinite sequence. Each traversal of the tree starting in its root represents a single path. The tree itself thus represents all possible paths, and is directly obtained from a transition system by "unfolding" at the state of interest. The tree rooted at state $s$ thus represents all possible infinite computations in the transition system that start in $s$. Figure 6.1 depicts a transition system and its unfolding. (For convenience, each node in the tree consists of a pair indicating the state and the level of the node in the tree.)

The temporal operators in branching temporal logic allow the expression of properties of some or all computations that start in a state. To that end, it supports an existential path quantifier (denoted $\exists$ ) and a universal path quantifier (denoted $\forall$ ). For instance, the property $\exists \diamond \Phi$ denotes that there exists a computation along which $\diamond \Phi$ holds. That is, it states that there is at least one possible computation in which a state that satisfies $\Phi$

(a)

(b)

Figure 6.1: (a) A transition system and (b) a prefix of its infinite computation tree
is eventually reached. This does not, however, exclude the fact that there can also be computations for which this property does not hold, for instance, computations for which $\Phi$ is always refuted. The property $\forall \diamond \Phi$, in contrast, states that all computations satisfy the property $\diamond \Phi$. More complicated properties can be expressed by nesting universal and existential path quantifiers. For instance, the aforementioned property "for every computation it is always possible to return to the initial state" can be faithfully expressed by $\forall \square \exists \diamond$ start: in any state $(\square)$ of any possible computation $(\forall)$, there is a possibility ( $\exists$ ) to eventually return to the start state ( $\rangle$ start).

This chapter considers Computation Tree Logic (CTL), a temporal logic based on propositional logic with a discrete notion of time, and only future modalities. CTL is an important branching temporal logic that is sufficiently expressive for the formulation of an important set of system properties. It was originally used by Clarke and Emerson [86] and (in a slightly different form) by Queille and Sifakis [347] for model checking. More importantly, it is a logic for which efficient and - as we will see - rather simple model-checking algorithms do exist.

Anticipatory to the results presented in this chapter, we summarize the major aspects of the linear-vs-branching-time debate and provide arguments that justify the treatment of model checking based on linear or branching time logics:

- The expressiveness of many linear and branching temporal logics is incomparable. This means that some properties that are expressible in a linear temporal logic cannot be expressed in certain branching temporal logics, and vice versa.

| Aspect | Linear time | Branching time |
| :---: | :---: | :---: |
| "behavior" <br> in a state $s$ | path-based: <br> trace $(s)$ | state-based: <br> computation tree of $s$ |
| temporal <br> logic | LTL: path formulae $\varphi$ <br> $s \models \varphi$ iff <br> $\forall \pi \in \operatorname{Paths}(s) \cdot \pi \models \varphi$ | CTL: state formulae <br> existential path quantification $\exists \varphi$ <br> universal path quantification: $\forall \varphi$ |
| complexity of the <br> model checking <br> problems | $\operatorname{PSPACE-complete}$$\mathcal{O}(\|T S\| \cdot \exp (\|\varphi\|))$ | $\mathcal{O}(\|T S\| \cdot\|\Phi\|)$ |

Table 6.1: Linear-time vs. branching-time in a nutshell.

- The model-checking algorithms for linear and branching temporal logics are quite different. This results, for instance, in significantly different time and space complexity results.
- The notion of fairness can be treated in linear temporal logic without the need for any additional machinery since fairness assumptions can be expressed in the logic. For various branching temporal logics this is not the case.
- The equivalences and preorders between transition systems that "correspond" to linear temporal logic are based on traces, i.e., trace inclusion and equality, whereas for branching temporal logic such relations are based on simulation and bisimulation relations (see Chapter 7).

Table 6.1 summarizes the main differences between the linear-time and branching-time perspective in a succinct way.

### 6.2 Computation Tree Logic

This section presents the syntax and the semantics of CTL. The following sections will discuss the relation and differences between CTL and LTL, present a model-checking algorithm for CTL, and introduce some extensions of CTL.

### 6.2.1 Syntax

CTL has a two-stage syntax where formulae in CTL are classified into state and path formulae. The former are assertions about the atomic propositions in the states and their branching structure, while path formulae express temporal properties of paths. Compared to LTL formulae, path formulae in CTL are simpler: as in LTL they are built by the next-step and until operators, but they must not be combined with Boolean connectives and no nesting of temporal modalities is allowed.

## Definition 6.1. Syntax of CTL

CTL state formulae over the set AP of atomic proposition are formed according to the following grammar:

$$
\Phi::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\neg \Phi| \exists \varphi \mid \forall \varphi
$$

where $a \in A P$ and $\varphi$ is a path formula. CTL path formulae are formed according to the following grammar:

$$
\varphi::=\bigcirc \Phi \mid \Phi_{1} \cup \Phi_{2}
$$

where $\Phi, \Phi_{1}$ and $\Phi_{2}$ are state formulae.

Greek capital letters will denote CTL state formulae (CTL formulae, for short), whereas lowercase Greek letters will denote CTL path formulae.

CTL distinguishes between state formulae and path formulae. Intuitively, state formulae express a property of a state, while path formulae express a property of a path, i.e., an infinite sequence of states. The temporal operators $\bigcirc$ and $U$ have the same meaning as in LTL and are path operators. Formula $\bigcirc \Phi$ holds for a path if $\Phi$ holds at the next state in the path, and $\Phi \cup \Psi$ holds for a path if there is some state along the path for which $\Psi$ holds, and $\Phi$ holds in all states prior to that state. Path formulae can be turned into state formulae by prefixing them with either the path quantifier $\exists$ (pronounced "for some path") or the path quantifier $\forall$ (pronounced "for all paths"). Note that the linear
temporal operators $\bigcirc$ and $U$ are required to be immediately preceded by $\exists$ or $\forall$ to obtain a legal state formula. Formula $\exists \varphi$ holds in a state if there exists some path satisfying $\varphi$ that starts in that state. Dually, $\forall \varphi$ holds in a state if all paths that start in that state satisfy $\varphi$.

## Example 6.2. Legal CTL Formulae

Let $A P=\{x=1, x<2, x \geqslant 3\}$ be a set of atomic propositions. Examples of syntactically correct CTL formulae are

$$
\exists \bigcirc(x=1), \forall \bigcirc(x=1), \text { and } x<2 \vee x=1
$$

and $\exists((x<2) \cup(x \geqslant 3))$ and $\forall($ true $U(x<2))$. Some examples of formulae that are syntactically incorrect are

$$
\exists(x=1 \wedge \forall \bigcirc(x \geqslant 3)) \text { and } \exists \bigcirc(\operatorname{true} U(x=1))
$$

The first is not a CTL formula since $x=1 \wedge \forall \bigcirc(x \geqslant 3)$ is not a path formula and thus must not be preceded by $\exists$ The second formula is not a CTL formula since true $\mathrm{U}(x=1)$ is a path formula rather than a state formula, and thus cannot be preceded by $\bigcirc$. Note that

$$
\exists \bigcirc(x=1 \wedge \forall \bigcirc(x \geqslant 3)) \text { and } \exists \bigcirc \forall(\operatorname{true} \cup(x=1))
$$

are, however, syntactically correct CTL formulae.

The Boolean operators true, false, $\wedge, \rightarrow$ and $\Leftrightarrow$ are defined in the usual way. The temporal modalities "eventually", "always", and "weak until" can be derived-similarly as for LTL—as follows:

$$
\begin{array}{ll}
\text { eventually: } & \exists \diamond \Phi=\exists(\text { true U } \Phi) \\
& \forall \diamond \Phi=\forall(\text { true } U \Phi) \\
\text { always: } & \exists \square \Phi=\neg \forall \diamond \neg \Phi \\
& \forall \square \Phi=\neg \exists \diamond \neg \Phi
\end{array}
$$

$\exists \diamond \Phi$ is pronounced " $\Phi$ holds potentially" and $\forall \diamond \Phi$ is pronounced " $\Phi$ is inevitable". $\exists \square \Phi$ is pronounced "potentially always $\Phi$ ", $\forall \square \Phi$ is pronounced "invariantly $\Phi$ ", and $\forall \bigcirc \Phi$ is pronounced "for all paths next $\Phi$ ".

Note that "always" $\Phi$ cannot be obtained from the "equation" $\square \Phi=\neg \diamond \neg \Phi$ (as in LTL), since propositional logic operators cannot be applied to path formulae. In particular, $\exists \neg \checkmark \neg \Phi$ is not a CTL formula. Instead, we exploit the duality of existential and universal quantification:

- there exists a path with the property $E$ if and only if the state property "not all paths violate property $E$ " is satisfied, and
- all paths satisfy property $E$ if and only if the state property "there is a path without property $E$ " is violated.

Accordingly, $\exists \square \Phi$ is not defined as $\exists \neg \diamond \neg \Phi$, but rather as $\neg \forall \diamond \neg \Phi$.

## Example 6.3. CTL Formulae

To give a feeling about how simple properties can be formalized in CTL we treat some intuitive examples. The mutual exclusion property can be described in CTL by the formula

$$
\forall \square\left(\neg c r i t_{1} \vee \neg c r i t_{2}\right) .
$$

CTL formulae of the form $\forall \square \forall \diamond \Phi$ express that $\Phi$ is infinitely often true on all paths. (This fact will be formally proven later; see Remark 6.8 on page 326 .) The CTL formula

$$
\left(\forall \square \forall \diamond c_{i} t_{1}\right) \wedge\left(\forall \square \forall \diamond c_{i} t_{2}\right)
$$

thus requires each process to have access to the critical section infinitely often. In case of a traffic light, the safety property "each red light phase is preceded by a yellow light phase" can be formulated in CTL by

$$
\forall \square(\text { yellow } \vee \forall \bigcirc \neg \text { red })
$$

depending on the precise meaning of a "phase". The liveness property "the traffic light is infinitely often green" can be formulated as

$$
\forall \square \forall \diamond \text { green . }
$$

Progress properties such as "every request will eventually be granted" can be described by

$$
\forall \square(\text { request } \longrightarrow \forall \diamond \text { response }) .
$$

Finally, the CTL formula

$$
\forall \square \exists \diamond \text { start }
$$

expresses that in every reachable system state it is possible to return (via 0 or more transitions) to (one of) the starting state(s).

### 6.2.2 Semantics

CTL formulae are interpreted over the states and paths of a transition system TS. Formally, given a transition system TS, the semantics of CTL formulae is defined by two satisfaction relations (both denoted by $\models_{T S}$, or briefly $\vDash$ ): one for the state formulae and one for the path formulae. For the state formulae, $\models$ is a relation between the states in $T S$ and state formulae. We write $s \models \Phi$ rather than $(s, \Phi) \in \models$. The intended interpretation is: $s \models \Phi$ if and only if state formula $\Phi$ holds in state $s$. For the path formulae, $\models$ is a relation between maximal path fragments in $T S$ and path formulae. We write $\pi \models \Phi$ rather than $(\pi, \Phi) \in \models$. The intended interpretation is: $\pi \models \varphi$ if and only if path $\pi$ satisfies path formula $\varphi$.

## Definition 6.4. Satisfaction Relation for CTL

Let $a \in A P$ be an atomic proposition, $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system without terminal states, state $s \in S, \Phi, \Psi$ be CTL state formulae, and $\varphi$ be a CTL path formula. The satisfaction relation $\models$ is defined for state formulae by

$$
\begin{array}{lll}
s \models a & \text { iff } & a \in L(s) \\
s \models \neg \Phi & \text { iff } & \text { not } s \models \Phi \\
s \models \Phi \wedge \Psi & \text { iff } & (s \models \Phi) \text { and }(s \models \Psi) \\
s \models \exists \varphi & \text { iff } & \pi \models \varphi \text { for some } \pi \in \operatorname{Paths}(s) \\
s \models \forall \varphi & \text { iff } & \pi \models \varphi \text { for all } \pi \in \operatorname{Paths}(s)
\end{array}
$$

For path $\pi$, the satisfaction relation $\models$ for path formulae is defined by

$$
\begin{array}{lll}
\pi & =\bigcirc \Phi & \text { iff }
\end{array} \quad \pi[1] \models \Phi
$$

where for path $\pi=s_{0} s_{1} s_{2} \ldots$ and integer $i \geqslant 0, \pi[i]$ denotes the $(i+1)$ th state of $\pi$, i.e., $\pi[i]=s_{i}$.

The interpretations for atomic propositions, negation, and conjunction are as usual, where it should be noted that in CTL they are interpreted over states, whereas in LTL they are interpreted over paths. state formula $\exists \varphi$ is valid in state $s$ if and only if there exists some path starting in $s$ that satisfies $\varphi$. In contrast, $\forall \varphi$ is valid in state $s$ if and only if all paths starting in $s$ satisfy $\varphi$. The semantics of the path formulae is identical (although formulated slightly more simply) to that for LTL. ${ }^{1}$ For instance, $\exists \bigcirc \Phi$ is valid in state $s$ if

[^24]and only if there exists some path $\pi$ starting in $s$ such that in the next state of this path, state $\pi[1]$, the property $\Phi$ holds. This is equivalent to the existence of a direct successor $s^{\prime}$ of $s$ such that $s^{\prime} \models \Phi . \forall(\Phi \cup \Psi)$ is valid in state $s$ if and only if every path starting in $s$ has an initial finite prefix (possibly only containing $s$ ) such that $\Psi$ holds in the last state of this prefix and $\Phi$ holds in all other states along the prefix. $\exists(\Phi \cup \Psi)$ is valid in $s$ if and only if there exists a path starting in $s$ that satisfies $\Phi \cup \Psi$. As for LTL, the semantics of CTL here is nonstrict in the sense that the path formula $\Phi \cup \Psi$ is valid if the initial state of the path satisfies $\Psi$.

## Definition 6.5. CTL Semantics for Transition Systems

Given a transition system $T S$ as before, the satisfaction set $\operatorname{Sat}_{T S}(\Phi)$, or briefly $\operatorname{Sat}(\Phi)$, for CTL-state formula $\Phi$ is defined by:

$$
\operatorname{Sat}(\Phi)=\{s \in S|s|=\Phi\}
$$

The transition system $T S$ satisfies CTL formula $\Phi$ if and only if $\Phi$ holds in all initial states of TS:

$$
T S \models \Phi \quad \text { if and only if } \quad \forall s_{0} \in I . s_{0} \models \Phi .
$$

This is equivalent to $I \subseteq \operatorname{Sat}(\Phi)$.

The semantics of the derived path operators "always" and "eventually" is similar to that in LTL. For path fragment $\pi=s_{0} s_{1} s_{2} \ldots$ :

$$
\pi \models \diamond \Phi \quad \text { if and only if } \quad s_{j} \models \Phi \text { for some } j \geqslant 0
$$

From this it can be derived that:

$$
\begin{aligned}
& s \models \exists \square \Phi \quad \text { iff } \quad \exists \pi \in \operatorname{Paths}(s) . \pi[j] \models \Phi \text { for all } j \geqslant 0, \\
& s \models \forall \square \Phi \quad \text { iff } \quad \forall \pi \in \operatorname{Paths}(s) . \pi[j] \models \Phi \text { for all } j \geqslant 0 .
\end{aligned}
$$

Therefore, $\square \Phi$ can be understood as CTL path formula with the semantics:

$$
\pi=s_{0} s_{1} s_{2} \ldots \models \square \Phi \quad \text { if and only if } \quad s_{j} \models \Phi \text { for all } j \geqslant 0
$$

In particular, $\forall \square \Phi$ corresponds to the invariant over the invariant condition $\Phi$.
In a similar way, one can derive that $\exists \square \Phi$ is valid in state $s$ if and only if there exists some path starting at $s$ such that for each state on this path the formula $\Phi$ holds. The formula $\exists \diamond \Phi$ is valid in state $s$ if and only if $\Phi$ holds eventually along some path that starts in $s$, and $\forall \diamond \Phi$ is valid if and only if this property holds for all paths that start in $s$. A schematic overview of the validity of $\exists \square, \exists \diamond, \forall \diamond$, and $\forall \square$ is given in Figure 6.2, where


Figure 6.2: Visualization of semantics of some basic CTL formulae.


Figure 6.3: A transition system of the TMR system.

| Property | Formalization in CTL |
| :--- | :--- |
| Possibly the system never goes down | $\exists \square \neg$ down |
| Invariantly the system never goes down | $\forall \square \neg$ down |
| It is always possible to start as new | $\forall \square \exists \diamond$ up $_{3}$ |
| The system always eventually goes down <br> and is operational until going down | $\forall\left(\left(u p_{3} \vee u p_{2}\right) \cup\right.$ down $)$ |

Table 6.2: Some properties for the TMR system and their formalization in CTL.
black-colored states satisfy the proposition black, gray states are labeled with gray, and all other states are labeled neither with black nor with gray.

## Example 6.6. A Triple Modular Redundant System

Consider a triple modular redundant (TMR) system with three processors and a single voter. As each component of this system can fail, the reliability is increased by letting all processors execute the same program. The voter takes a majority vote of the outputs of the three processors. If a single processor fails, the system can still produce reliable outputs. Each component can be repaired. It is assumed that only one component at a time can fail and only one at a time can be repaired. On failure of the voter, the entire system fails. On repair of the voter, it is assumed that the system starts as being new, i.e., with three processors and a voter. The transition system of this TMR is depicted in Figure 6.3. States are of the form $s_{i, j}$ where $i$ denotes the number of processors that is currently up $(0<i \leqslant 3)$ and $j$ the number of operational voters $(j=0,1)$. We consider the TMR system to be operational if at least two processors are functioning properly. Some interesting properties of this system and their formulation in CTL are listed in Table 6.2 on page 323. We consider each of the formulae in isolation:

- state formula $\exists \square \neg$ down holds in state $s_{3,1}$, as there is a path, e.g., $\left(s_{3,1} s_{2,1}\right)^{\omega}$, starting in that state and that never reaches the down state, i.e., $\left(s_{3,1} s_{2,1}\right)^{\omega} \vDash$ $\square \neg$ down.
- Formula $\forall \square \neg$ down, however, does not hold in state $s_{3,1}$, as there is a path starting from that state, such as $\left(s_{3,1}\right)^{+} s_{0,0} \ldots$, that satisfies $\neg \square \neg$ down, or, equivalently, $\diamond$ down.
- Formula $\forall \square \exists \diamond u p_{3}$ holds in state $s_{3,1}$, as in any state of any of its paths it is possible to return to the initial state, e.g., by first moving to state $s_{0,0}$ and then to $s_{3,1}$. For instance, the path $s_{3,1}\left(s_{2,1}\right)^{\omega} \models \square \exists \diamond u p_{3}$ since $s_{2,1} \models \exists \diamond u p_{3,1}$. This property should not be confused with the CTL formula $\forall \diamond u p_{3}$, which expresses that each path eventually will visit the initial state. (Note that this formula is trivially valid for state $s_{3,1}$ as it satisfies up3.)
- The last property of Table 6.2 does not hold in state $s_{3,1}$ as there exists a path, such as $s_{3,1} s_{2,1} s_{1,1} s_{0,0} \ldots$, for which the path formula $\left(u p_{3} \vee u p_{2}\right) \mathrm{U}$ down does not hold. The formula is refuted since the path visits state $s_{1,1}$, a state that satisfies neither down nor up $p_{3}$ nor $u p_{2}$.


## Example 6.7. CTL Semantics

Consider the transition system depicted at the top (a) of Figure 6.4. Just below the transition system the validity of several CTL formulae is indicated for each state. (For simplicity, the initial states are not indicated.) A state is colored black if the formula is valid in that state; otherwise, it is white. The following formulae are considered:

- The formula $\exists \bigcirc a$ is valid for all states since all states have some direct successor state that satisfies $a$.
- $\forall \bigcirc a$ is not valid for state $s_{0}$, since a possible path starting at $s_{0}$ goes directly to state $s_{2}$ for which $a$ does not hold. Since the other states have only direct successors for which $a$ holds, $\forall \bigcirc a$ is valid for all other states.
- For all states except state $s_{2}$, it is possible to have a computation that leads to state $s_{3}$ (such as $s_{0} s_{1} s_{3}^{\omega}$ when starting in $s_{0}$ ) for which $a$ is globally valid. Therefore, $\exists \square a$ is valid in these states. Since $a \notin L\left(s_{2}\right)$ there is no path starting at $s_{2}$ for which $a$ is globally valid.
- $\forall \square a$ is only valid for $s_{3}$ since its only path, $s_{3}^{\omega}$, always visits a state in which $a$ holds. For all other states it is possible to have a path which contains $s_{2}$ that does not satisfy $a$. So for these states $\forall \square a$ is not valid.


Figure 6.4: Interpretation of several CTL formulae.

- $\exists \diamond(\exists \square a)$ is valid for all states since from each state another state (either $s_{0}, s_{1}$, or $s_{3}$ ) can be eventually reached from which some computation can start along which $a$ is globally valid.
- $\forall(a \cup b)$ is not valid in $s_{3}$ since its only computation $\left(s_{3}^{\omega}\right)$ never reaches a state for which $b$ holds. In state $s_{0}$ proposition $a$ holds until $b$ holds, and in states $s_{1}$ and $s_{2}$ proposition $b$ holds immediately. So, for these states the formula is true.
- Finally, $\exists(a \cup(\neg a \wedge \forall(\neg a \cup b)))$ is not valid in $s_{3}$, since from $s_{3}$ a $b$-state can never be reached. For the states $s_{0}$ and $s_{1}$ the formula is valid, since state $s_{2}$ can be reached from these states via an $a$-path; $\neg a$ is valid in $s_{2}$, and from $s_{2}$ all possible paths satisfy $\neg a \cup b$, since $s_{2}$ is a $b$-state. For instance, for state $s_{0}$ the path $\left(s_{0} s_{2} s_{1}\right)^{\omega}$ satisfies $a \cup(\neg a \wedge \forall(\neg a \cup b))$ since $a \in L\left(s_{0}\right), a \notin L\left(s_{2}\right)$, and $b \in L\left(s_{1}\right)$. For state $s_{2}$ the property is valid since $a$ is invalid in $s_{2}$ and for all paths starting at $s_{2}$ the first state is a $b$-state.


## Remark 6.8. Infinitely Often

For a better comprehension of the semantics of CTL, let us prove that:
$s \models \forall \square \forall \diamond a \quad$ if and only if $\forall \pi \in \operatorname{Paths}(s) . \pi[i] \models a$ for infinitely many $i$.
$\Rightarrow$ : Let $s$ be a state, such that $s \models \forall \square \forall \diamond a$. The proof obligation is to show that every infinite path fragment $\pi$ starting in $s$ passes through an $a$-state infinitely often. Let $\pi=s_{0} s_{1} s_{2} \ldots \in \operatorname{Paths}(s)$ and $j \geqslant 0$. We demonstrate that there exists an index $i \geqslant j$ with $s_{i} \models a$. Since $s \models \forall \square \forall \diamond a$, we have

$$
\pi \models \square \forall \diamond a
$$

In particular, $s_{j} \models \forall \diamond$. From $\pi[j .]=.s_{j} s_{j+1} \ldots \in \operatorname{Paths}\left(s_{j}\right)$ it follows that

$$
s_{j} s_{j+1} s_{j+2} \ldots \models \diamond a
$$

Thus, there exists an index $i \geqslant j$ with $s_{i} \models a$. As this reasoning applies to any index $j$, path $\pi$ visits an $a$-state infinitely often.
$\Leftarrow$ : Let $s$ be a state such that every infinite path fragment starting in $s$ visits infinitely many $a$-states. Let $\pi=s_{0} s_{1} s_{2} \ldots \in \operatorname{Paths}(s)$. To show that $s \models \forall \square \forall \Delta a$, it has to be proven that $\pi \models \square \forall \diamond a$, i.e.:

$$
s_{j} \models \forall \diamond a, \quad \text { for any } j \geqslant 0
$$

Let $j \geqslant 0$ and $\pi^{\prime}=s_{j}^{\prime} s_{j+1}^{\prime} s_{j+2}^{\prime} \ldots \in \operatorname{Paths}\left(s_{j}\right)$. To show that $s_{j} \models \forall \diamond a$, it suffices to prove that $\pi^{\prime}$ visits at least one $a$-state. It is not difficult to infer that

$$
\pi^{\prime \prime}=\underbrace{s_{0} s_{1} s_{2} \ldots s_{j}}_{\text {prefix of } \pi} \underbrace{s_{j+1}^{\prime} s_{j+2}^{\prime} \ldots}_{\pi^{\prime} \in \operatorname{Paths}\left(s_{j}\right)} \in \operatorname{Paths}(s)
$$

By assumption, $\pi^{\prime \prime}$ visits infinitely many $a$-states. In particular, there is an $i>j$ such that $s_{i}^{\prime} \models a$. It now follows that

$$
\pi^{\prime}=s_{j} s_{j+1}^{\prime} \ldots s_{i-1}^{\prime} s_{i}^{\prime} s_{i+1}^{\prime}, \ldots \models \diamond a
$$

and, as this holds for any path $\pi^{\prime} \in \operatorname{Paths}\left(s_{j}\right)$, thus $s_{j} \models \forall \diamond a$. This yields $\pi \models \square \forall \diamond a$ for all paths $\pi \in \operatorname{Paths}(s)$. Thus, we have $s \models \forall \square \forall \diamond a$.

## Remark 6.9. Weak Until

As for LTL (see Section 5.1.5), a slight variant of the until operator can be defined, viz. the weak-until operator, denoted W . The intuition behind this operator is that path $\pi$ satisfies $\Phi W \Psi$, for state formulae $\Phi$ and $\Psi$, if either $\Phi \cup \Psi$ or $\square \Phi$ holds. That is, the difference between until and weak until is that the latter does not require a $\Psi$-state to be reached eventually.

The weak-until operator in CTL cannot be defined directly starting from the LTL definition

$$
\varphi \mathrm{W} \psi=\varphi \mathrm{U} \psi \vee \square \varphi,
$$

since $\exists(\varphi \cup \psi \vee \square \varphi)$ is not a syntactically correct CTL formula. However, using the LTL equivalence law $\varphi \mathrm{W} \psi \equiv \neg((\varphi \wedge \neg \psi) \mathrm{U}(\neg \varphi \wedge \neg \psi))$ and the duality between universal and existential quantification, the weak-until operator can be defined in CTL by

$$
\begin{aligned}
& \exists(\Phi \mathrm{W} \Psi)=\neg \forall((\Phi \wedge \neg \Psi) \mathrm{U}(\neg \Phi \wedge \neg \Psi)), \\
& \forall(\Phi \mathrm{W} \Psi)=\neg \exists((\Phi \wedge \neg \Psi) \mathrm{U}(\neg \Phi \wedge \neg \Psi)) .
\end{aligned}
$$

Let us now check the semantics of $\exists \mathrm{W}$. From the above-defined duality, it follows that $s \vDash \exists(\Phi \mathrm{~W} \Psi)$ if and only if there exists a path $\pi=s_{0} s_{1} s_{2} \ldots$ that starts in $s$ (i.e., $\left.s_{0}=s\right)$ such that

$$
\pi \not \vDash(\Phi \wedge \neg \Psi) \cup(\neg \Phi \wedge \neg \Psi)
$$

Such path exists if and only if

- either $s_{j} \models \Phi \wedge \neg \Psi$ for all $j \geqslant 0$, i.e., $\pi \models \square(\Phi \wedge \neg \Psi)$, or
- there exists an index $j$ such that

$$
\begin{aligned}
& -s_{j} \not \models \Phi \wedge \neg \Psi \text { and } s_{j} \not \models \neg \Phi \wedge \neg \Psi \text {, i.e., } s_{j} \models \Psi \text {, and } \\
& -s_{i} \models \Phi \wedge \neg \Psi \text { for all } 0 \leqslant i<j \text {. }
\end{aligned}
$$

This is equivalent to $\pi \models \Phi \cup \Psi$.

Gathering these results yields

$$
\begin{aligned}
\pi \models \Phi \mathrm{W} \Psi \quad & \text { if and only if } \quad \pi \models \Phi \cup \Psi \text { or } \pi \models \square(\Phi \wedge \neg \Psi), \\
& \text { if and only if } \quad \pi \models \Phi \cup \Psi \text { or } \pi \models \square \Phi .
\end{aligned}
$$

Thus, the CTL formula $\exists(\Phi \mathrm{W} \Psi)$ is equivalent to $\exists(\Phi \cup \Psi) \vee \exists \square \Phi$. In the same way, one can check that the meaning of $\forall(\Phi \mathrm{W} \Psi)$ is as expected, i.e., $s \models \forall(\Phi \mathrm{~W} \Psi)$ if and only if all paths starting in $s$ fulfill $\Phi W \Psi$ according to the LTL semantics of $W$.

## Remark 6.10. The Semantics of Negation

For state $s$, we have $s \not \vDash \Phi$ if and only if $s \models \neg \Phi$. This, however, does not hold in general for transition systems. That is to say, it is possible that the statements $T S \not \vDash \Phi$ and $T S \not \vDash \neg \Phi$ both hold. This stems from the fact that there might be two initial states, $s_{0}$ and $s_{0}^{\prime}$, say, such that $s_{0} \models \Phi$ and $s_{0}^{\prime} \not \models \Phi$. Furthermore:
$T S \not \vDash \neg \exists \varphi$ iff there exists a path $\pi \in \operatorname{Paths}(T S)$ with $\pi \models \varphi$.
This-at first glance surprising - equivalence is justified by the fact that the interpretation of CTL state formulae over transition systems is based on a universal quantification over the initial states. The statement $T S \not \vDash \neg \exists \varphi$ thus holds if and only if there exists an initial state $s_{0} \in I$ with $s_{0} \not \models \neg \exists \varphi$, i.e., $s_{0} \models \exists \varphi$. On the other hand, $T S \models \exists \varphi$ requires that $s_{0} \models \exists \varphi$ for all $s_{0} \in I$. Consider the following transition system:


It follows that $s_{0} \models \exists \square a$, whereas $s_{0}^{\prime} \not \models \exists \square a$. Accordingly, $T S \not \vDash \neg \exists \square a$ and $T S \not \vDash \exists \square a$.

The semantics of CTL has been defined for a transition system without terminal states. This has the (technically) pleasant effect that all paths are infinite and simplifies the
definition of $\models$ for paths. In the following remark it is shown how to adapt the path semantics in case transition systems are considered with terminal states, i.e., when finite paths are possible.

## Remark 6.11. CTL Semantics for Transition Systems with Terminal States

For finite maximal path fragment $\pi=s_{0} s_{1} s_{2} \ldots s_{n}$ of length $n$, i.e., $s_{n}$ is a terminal state, let

$$
\begin{array}{rlll}
\pi & \models \bigcirc \Phi & \text { iff } & n>0 \text { and } s_{1} \models \Phi \\
\pi & \models \Phi \cup \Psi \text { iff } & \text { there exists an index } j \in \mathbb{N} \text { with } j \leqslant n, \text { and } \\
& s_{i} \models \Phi, \text { for } i=0,1, \ldots, j-1, \text { and } s_{j} \models \Psi .
\end{array}
$$

Then, $s \models \forall \bigcirc$ false if and only if $s$ is a terminal state. For the derived operators $\diamond$ and $\square$ we obtain

$$
\begin{aligned}
\pi & \models \diamond \Phi \quad \text { iff there exists an index } j \leqslant n \text { with } s_{j} \models \Phi \\
\pi & \models \square \Phi \quad \text { iff } \quad \text { for all } j \in \mathbb{N} \text { with } j \leqslant n \text { we have } s_{j} \models \Phi
\end{aligned}
$$

### 6.2.3 Equivalence of CTL Formulae

CTL formulae $\Phi$ and $\Psi$ are called equivalent whenever they are semantically identical, i.e., when for any state $s$ it holds that ${ }^{2} s \models \Phi$ if and only if $s \models \Psi$.

## Definition 6.12. Equivalence of CTL Formulae

CTL formulae $\Phi$ and $\Psi$ (over $A P$ ) are called equivalent, denoted $\Phi \equiv \Psi$, if $\operatorname{Sat}(\Phi)=$ $\operatorname{Sat}(\Psi)$ for all transition systems $T S$ over $A P$.

Accordingly, $\Phi \equiv \Psi$ if and only if for any transition system $T S$ we have:

$$
T S \models \Phi \quad \text { if and only if } \quad T S \models \Psi
$$

Besides the standard equivalence laws for the propositional logic fragment of CTL, there exist a number of equivalence rules for temporal modalities in CTL. An important set of equivalence laws is indicated in Figure 6.5. To understand the expansion laws, let us reconsider the expansion law for the until operator in LTL:

$$
\varphi \cup \psi \equiv \psi \vee(\varphi \wedge \bigcirc(\varphi \cup \psi))
$$

[^25]duality laws for path quantifiers
\[

$$
\begin{array}{rlrl}
\forall \bigcirc \Phi & \equiv \neg \exists \bigcirc \neg \Phi & \exists \bigcirc \Phi \equiv \neg \forall \bigcirc \neg \Phi \\
\forall \diamond \Phi & \equiv \neg \exists \square \neg \Phi & \exists \diamond \Phi \equiv \neg \forall \square \neg \Phi \\
\forall(\Phi \cup \Psi) & \equiv \neg \exists(\neg \Psi \cup(\neg \Phi \wedge \neg \Psi)) \wedge \neg \exists \square \neg \Psi \\
& \equiv \neg \exists((\Phi \wedge \neg \Psi) \cup(\neg \Phi \wedge \neg \Psi)) \wedge \neg \exists \square(\Phi \wedge \neg \Psi) \\
& \equiv \neg \exists((\Phi \wedge \neg \Psi) \mathrm{W}(\neg \Phi \wedge \neg \Psi))
\end{array}
$$
\]

expansion laws

$$
\begin{aligned}
\forall(\Phi \cup \Psi) & \equiv \Psi \vee(\Phi \wedge \forall \bigcirc \forall(\Phi \cup \Psi)) \\
\forall \diamond \Phi & \equiv \Phi \vee \forall \bigcirc \forall \diamond \Phi \\
\forall \square \Phi & \equiv \Phi \wedge \forall \bigcirc \forall \square \Phi \\
\exists(\Phi \cup \Psi) & \equiv \Psi \vee(\Phi \wedge \exists \bigcirc \exists(\Phi \cup \Psi)) \\
\exists \diamond \Phi & \equiv \Phi \vee \exists \bigcirc \exists \diamond \Phi \\
\exists \square \Phi & \equiv \Phi \wedge \exists \bigcirc \exists \square \Phi
\end{aligned}
$$

distributive laws

$$
\begin{aligned}
& \forall \square(\Phi \wedge \Psi) \equiv \forall \square \Phi \wedge \forall \square \Psi \\
& \exists \diamond(\Phi \vee \Psi) \equiv \exists \diamond \Phi \vee \exists \diamond \Psi
\end{aligned}
$$

Figure 6.5: Some equivalence rules for CTL.

In CTL, similar expansion laws for $\exists(\Phi \cup \Psi)$ and $\forall(\Phi \cup \Psi)$ exist. For instance, we have that $\exists(\Phi \cup \Psi)$ is equivalent to the fact that the current state either satisfies $\Psi$, or it satisfies $\Phi$, and for some direct successor state, $\exists(\Phi \cup \Psi)$ holds. The expansion laws for $\exists \diamond \Phi$ and $\exists \square \Phi$ can be simply derived from the expansion laws for $\exists \mathrm{U}$. The basic idea behind these laws - as for LTL-is to express the validity of a formula by a statement about the current state (without the need to use temporal operators) and a statement about the direct successors of this state (using either $\exists \bigcirc$ or $\forall \bigcirc$ depending on whether an existential or a universally quantified formula is treated). For instance, $\exists \square \Phi$ is valid in state $s$ if $\Phi$ is valid in $s$ (a statement about the current state) and $\Phi$ holds for all states along some path starting at $s$ (a statement about the successor states).

Not any law in LTL can be easily lifted to CTL. Consider, for example, the following statement:

$$
\diamond(\varphi \vee \psi) \equiv \diamond \varphi \vee \diamond \psi
$$

which is valid for any path. The same is true for:

$$
\exists \diamond(\Phi \vee \Psi) \equiv \exists \diamond \Phi \vee \exists \diamond \Psi
$$

This can be seen as follows.
$\Leftarrow$ : Assume that $s \vDash \exists \diamond \Phi \vee \exists \diamond \Psi$. Then, without loss of generality, we may assume that $s \models \exists \diamond \Phi$. This means that there is some state $s^{\prime}$ (possibly $s=s^{\prime}$ ), reachable from state $s$, such that $s^{\prime} \models \Phi$. But then $s^{\prime} \models \Phi \vee \Psi$. This means that there exists a reachable state from $s$ which satisfies $\Phi \vee \Psi$. By the semantics of CTL it now follows that $s \vDash \exists \diamond(\Phi \vee \Psi)$.
$\Rightarrow$ : Let $s$ be an arbitrary state such that $s \nexists \exists \diamond(\Phi \vee \Psi)$. Then there exists a state $s^{\prime}$ (possibly $s=s^{\prime}$ ) such that $s^{\prime} \models \Phi \vee \Psi$. Without loss of generality we may assume that $s^{\prime} \models \Phi$. But then we can conclude that $s \models \exists \diamond \Phi$, as $s^{\prime}$ is reachable from $s$. Therefore we also have $s \models \exists \diamond \Phi \vee \exists \diamond \Psi$.

However, $\forall \diamond(\Phi \vee \Psi) \not \equiv \forall \diamond \Phi \vee \forall \diamond \Psi$ since $\forall \diamond(\Phi \vee \Psi) \Rightarrow \forall \diamond \Phi \vee \forall \diamond \Psi$ is invalid as shown by the following transition system:


For each path that starts in state $s$ we have that $\diamond(a \vee b)$ holds, so $s \models \forall \diamond(a \vee b)$. This follows directly from the fact that each path visits either state $s^{\prime}$ or state $s^{\prime \prime}$ eventually, and $s^{\prime} \models a \vee b$ and the same applies to $s^{\prime \prime}$. However, state $s$ does not satisfy $\forall \diamond a \vee \forall \diamond b$. For instance, path $s\left(s^{\prime \prime}\right)^{\omega} \models \diamond a$ but $s\left(s^{\prime \prime}\right)^{\omega} \not \vDash \diamond b$. Thus, $s \not \vDash \forall \diamond b$. By a similar reasoning applied to path $s\left(s^{\prime}\right)^{\omega}$ it follows that $s \not \models \forall \diamond a$. Thus, $s \not \vDash \forall \diamond a \vee \forall \diamond b$. Stated in words, not all computations that start in state $s$ eventually reach an $a$-state nor do they all eventually reach a $b$-state.

### 6.2.4 Normal Forms for CTL

The duality law for $\forall \bigcirc \Phi$ shows that $\forall \bigcirc$ can be treated as a derived operator of $\exists \bigcirc$. That is to say, the basic operators $\exists \bigcirc, \exists \mathrm{U}$, and $\forall \mathrm{U}$ would have been sufficient to define the syntax of CTL. The following theorem demonstrates that we can even omit the universal path quantifier and define all temporal modalities in CTL using the basic operators $\exists \bigcirc$, $\exists \mathrm{U}$, and $\exists \square$.

## Definition 6.13. Existential Normal Form (for CTL)

For $a \in A P$, the set of CTL state formulae in existential normal form (ENF, for short) is given by

$$
\Phi \quad::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\neg \Phi| \exists \bigcirc \Phi\left|\exists\left(\Phi_{1} \cup \Phi_{2}\right)\right| \exists \square \Phi .
$$

## Theorem 6.14. Existential Normal Form for CTL

For each CTL formula there exists an equivalent CTL formula in ENF.

Proof: The following duality laws allow elimination of the universal path quantifier and thus provide a translation of CTL formulae into equivalent ENF formulae:

$$
\begin{aligned}
\forall \bigcirc \Phi & \equiv \neg \exists \bigcirc \neg \Phi \\
\forall(\Phi \cup \Psi) & \equiv \neg \exists(\neg \Psi \cup(\neg \Phi \wedge \neg \Psi)) \wedge \neg \exists \square \neg \Psi
\end{aligned}
$$

Recall that the basis syntax of CTL only uses $\exists \bigcirc, \exists U$ and $\forall \bigcirc$ and $\forall U$. Thus, the two rules used in the proof of Theorem 6.14 allow the removal of all universal quantifiers from a given CTL formula. However, when implementing the translation from CTL formulae to ENF formulae one might use analogous rules for the derived operators, such as

$$
\begin{aligned}
\forall \diamond \Phi & \equiv \neg \exists \square \neg \Phi, \\
\forall \square \Phi & \equiv \neg \exists \diamond \neg \Phi=\neg \exists(\operatorname{true} U \Phi) .
\end{aligned}
$$

Since the rewrite rule for $\forall U$ triples the occurrences of the right formula $\Psi$, the translation from CTL to ENF can cause an exponential blowup.

Another normal form of importance is the positive normal form. A CTL formula is said to be in positive normal form (PNF, for short) whenever negations only occur adjacent to atomic propositions. That is, e.g., $\neg \forall(a \cup \neg b)$ is not in PNF, whereas $\exists(\neg a \wedge \neg b \cup a)$ is in PNF. To ensure that every CTL formula is equivalent to a formula in PNF, for each operator a dual operator is necessary. We have that conjunction and disjunction are dual, and that $\bigcirc$ is dual to itself. As for LTL, we adopt the weak until operator $W$ as a dual operator of U .

## Definition 6.15. Positive Normal Form (for CTL)

The set of CTL state formulae in positive normal form (PNF, for short) is given by

$$
\Phi \quad::=\text { true } \mid \text { false }|a| \neg a\left|\Phi_{1} \wedge \Phi_{2}\right| \Phi_{1} \vee \Phi_{2}|\exists \varphi| \forall \varphi
$$

where $a \in A P$ and the path formulae are given by

$$
\varphi::=\bigcirc \Phi\left|\Phi_{1} \cup \Phi_{2}\right| \Phi_{1} \mathbb{W} \Phi_{2}
$$

Theorem 6.16. Existence of Equivalent PNF Formulae

[^26]Proof: Any CTL formula can be transformed into PNF by successively "pushing" negations "inside" the formula at hand. This is facilitated by the following equivalence laws:

$$
\begin{array}{ll}
\neg \text { true } & \equiv \text { false } \\
\neg \neg \Phi & \equiv \Phi \\
\neg(\Phi \wedge \Psi) & \equiv \neg \Phi \vee \neg \Psi \\
\neg \forall \bigcirc \Phi & \equiv \exists \bigcirc \neg \Phi \\
\neg \exists \bigcirc \Phi & \equiv \forall \bigcirc \neg \Phi \\
\neg \forall(\Phi \cup \Psi) & \equiv \exists((\Phi \wedge \neg \Psi) \mathrm{W}(\neg \Phi \wedge \neg \Psi)) \\
\neg \exists(\Phi \cup \Psi) & \equiv \forall((\Phi \wedge \neg \Psi) \mathrm{W}(\neg \Phi \wedge \neg \Psi)) .
\end{array}
$$

Due to the fact that in the rules for $\forall U$ and $\exists U$ the number of occurrences of $\Psi$ (and $\Phi$ ) is doubled, the length of an equivalent CTL formula may be exponentially longer than the original CTL formula. ${ }^{3}$ The same phenomenon appeared in defining the PNF for LTL when using the weak-until operator. As for LTL, this exponential blowup can be avoided by using the release operator, which in CTL can be defined by: $\exists(\Phi \mathrm{R} \Psi)=\neg \forall((\neg \Phi) \cup(\neg \Psi))$ and $\forall(\Phi \mathrm{R} \Psi)=\neg \exists((\neg \Phi) \mathrm{U}(\neg \Psi))$.

### 6.3 Expressiveness of CTL vs. LTL

Although many relevant properties of reactive systems can be specified in LTL and CTL, the logics CTL and LTL are incomparable according to their expressiveness. More precisely, there are properties that one can express in CTL, but that cannot be expressed in LTL, and vice versa.

Let us first define what it means for CTL and LTL formulae to be equivalent. Intuitively speaking, equivalent means "express the same thing". More precisely:

## Definition 6.17. Equivalence between CTL- and LTL Formulae

CTL formula $\Phi$ and LTL formula $\varphi$ (both over AP) are equivalent, denoted $\Phi \equiv \varphi$, if for any transition system $T S$ over AP:

$$
T S \models \Phi \quad \text { if and only if } \quad T S \models \varphi .
$$

[^27]The LTL formula $\varphi$ holds in state $s$ of a transition system whenever all paths starting in $s$ satisfy $\varphi$. Given this (semantical) universal quantification over paths, it seems natural that, e.g., the LTL formula $\diamond a$ is equivalent to the CTL formula $\forall \diamond a$. This seems to suggest that for a given CTL formula, an equivalent LTL formula is obtained by simply omitting all universal path quantifiers (as these are implicit in LTL). The following result by Clarke and Draghicescu [85] (for which the proof is omitted) shows that dropping all (universal and existential) quantifiers is a safe way to generate an equivalent LTL formula, provided there are equivalent LTL formulae:

## Theorem 6.18. Criterion for Transforming CTL Formulae into Equivalent LTL Formulae

Let $\Phi$ be a CTL formula, and $\varphi$ the LTL formula that is obtained by eliminating all path quantifiers in $\Phi$. Then:

$$
\Phi \equiv \varphi \text { or there does not exist any LTL formula that is equivalent to } \Phi .
$$

For the following CTL formulae, an equivalent LTL formula is obtained by simply omitting all path quantifiers: $a, \forall \bigcirc a, \forall(a \cup b), \forall \diamond a, \forall \square a$, and $\forall \square \forall \diamond a$. The fact that the CTL formula $\forall \square \forall \diamond a$ is equivalent to the LTL formula $\square \diamond a$ has been established earlier in Remark 6.8 (326). However, $\forall \diamond \forall \square a$ and $\diamond \square a$ are not equivalent. The LTL formula $\diamond \square a$ ensures that $a$ will eventually forever (i.e., continuously from some point on) hold. The semantics of $\forall \diamond \forall \square a$ is different, however. The CTL formula $\forall \diamond \forall \square a$ asserts that on any computation, eventually some state, $s$ say, is reached such that $s \models \forall \square a$. Note that

$$
s \models \forall \diamond \underbrace{\forall \square a}_{\Phi}
$$

if and only if for any path $\pi=s_{0} s_{1} s_{2} \ldots \in \operatorname{Paths}(s), s_{j} \models \Phi$ for some $j$. For $\Phi=\forall \square a$, this entails that for any such path $\pi$ there is some state $s_{j}$ such that all reachable states from $s_{j}$ satisfy the atomic proposition $a$.

## Lemma 6.19. Persistence

The CTL formula $\forall \diamond \forall \square a$ and the LTL formula $\diamond \square a$ are not equivalent.

Proof: Consider the following transition system TS over $A P=\{a\}$ :


The initial state $s_{0}$ satisfies the LTL formula $\diamond \square a$, since each path starting in $s_{0}$ eventually remains forever in one of the two states $s_{0}$ or $s_{2}$, which are both labeled with $a$. The CTL formula $\forall \diamond \forall \square a$, however, does not hold in $s_{0}$, since we have $s_{0}^{\omega} \not \models \diamond \forall \square a$ (as $s_{0} \not \models \forall \square a$ ). This is due to the fact that the path $s_{0}^{*} s_{1} s_{2}^{\omega}$ passes through the $\neg a$-state $s_{1}$. Thus, $s_{0}^{\omega}$ is a path starting in $s_{0}$ which will never reach a state satisfying $\forall \square a$, i.e., $s_{0}^{\omega} \not \vDash \diamond \forall \square a$. Accordingly, it follows that

$$
s_{0} \not \vDash \forall \forall \diamond \forall \square a
$$

Given that the CTL formulae $\forall \diamond \forall \square a$ and the LTL formula $\diamond \square a$ are not equivalent and the fact that $\diamond \square a$ is obtained from $\forall \diamond \forall \square a$ by eliminating the universal path quantifiers, it follows from Theorem 6.18 that there does not exist an LTL formula that is equivalent to $\forall \diamond \forall \square a$. In a similar way, it can be shown that the CTL formulae $\forall \diamond(a \wedge \forall \bigcirc a)$ and $\diamond(a \wedge \bigcirc a)$ are not equivalent, and thus, the requirement $\forall \diamond(a \wedge \forall \bigcirc a)$ cannot be expressed in LTL.

## Lemma 6.20. Eventually an a-State with only direct a-Successors

The CTL formula $\forall \diamond(a \wedge \forall \bigcirc a)$ and the LTL formula $\diamond(a \wedge \bigcirc a)$ are not equivalent.

Proof: Consider the transition system depicted in Figure 6.6. All paths that start in the initial state $s_{0}$ have either as prefix the path fragment $s_{0} s_{1}$ or $s_{0} s_{3} s_{4}$. Clearly, all such paths satisfy the LTL formula $\diamond(a \wedge \bigcirc a)$, and so, $s_{0} \models \diamond(a \wedge \bigcirc a)$. On the other hand, however, $s_{0} \not \models \forall \diamond(a \wedge \forall \bigcirc a)$ as the path $s_{0} s_{1}\left(s_{2}\right)^{\omega}$ does not satisfy $\diamond(a \wedge \forall \bigcirc a)$. This follows from the fact that state $s_{0}$ has the non- $a$-state $s_{3}$ as direct successor, i.e., $s_{0} \not \models a \wedge \forall \bigcirc a$.

These examples show that certain requirements that can be expressed in CTL, cannot be expressed in LTL. The following theorem provides, in addition, some examples of LTL formulae for which no equivalent CTL formula exists. This establishes that the expressiveness of the temporal logics LTL and CTL are incomparable.


Figure 6.6: Transition system for $\forall \diamond(a \wedge \forall \bigcirc a)$.


Figure 6.7: The base transition systems: $T S_{0}$ (left) and $T S_{0}^{\prime}$ (right).

Theorem 6.21. Incomparable Expressiveness of CTL and LTL
(a) There exist LTL formulae for which no equivalent CTL formula exists. This holds for, for instance

$$
\diamond \square a \quad \text { or } \quad \diamond(a \wedge \bigcirc a)
$$

(b) There exist CTL formulae for which no equivalent LTL formula exists. This holds for, for instance

$$
\forall \diamond \forall \square a \quad \text { and } \quad \forall \diamond(a \wedge \forall \bigcirc a) \quad \text { and } \quad \forall \square \exists \diamond a .
$$

Proof:
(a) Consider the formula $\diamond \square a$. The proof for $\diamond(a \wedge \bigcirc a)$ goes along similar lines and is omitted here. Consider the two series of transition systems $T S_{0}, T S_{1}, T S_{2}, \ldots$ and $T S_{0}^{\prime}, T S_{1}^{\prime}, T S_{2}^{\prime}, \ldots$ that are inductively defined as follows (see Figures 6.7 and 6.8).


Figure 6.8: Inductive construction of $T S_{n}$ (upper) and $T S_{n}^{\prime}$ (lower).

For all transition systems, $A P=\{a\}$, and the action labels are not important. Let, for $n \geqslant 0$,

$$
T S_{n}=\left(S_{n},\{\tau\}, \rightarrow_{n},\left\{s_{n}\right\},\{a\}, L_{n}\right)
$$

and

$$
T S_{n}=\left(S_{n}^{\prime},\{\tau\}, \rightarrow_{n}^{\prime},\left\{s_{n}^{\prime}\right\},\{a\}, L_{n}^{\prime}\right)
$$

where $S_{0}=\left\{s_{0}, t_{0}\right\}, S_{0}^{\prime}=\left\{s_{0}^{\prime}, t_{0}^{\prime}\right\}$, and for $n>0$ :

$$
S_{n}=S_{n-1}^{\prime} \cup\left\{s_{n}, t_{n}\right\} \quad \text { and } \quad S_{n}^{\prime}=S_{n-1}^{\prime} \cup\left\{s_{n}^{\prime}, t_{n}^{\prime}\right\}
$$

The labeling functions are defined such that all states $t_{i}$ are labeled with $\{a\}$ and all states $s_{i}$ are labeled with $\varnothing$. Thus, $L_{0}\left(s_{0}\right)=\varnothing$ and $L_{0}\left(t_{0}\right)=\{a\}$, and for $n>0$, the labels of all states in $T S_{n-1}^{\prime}$ remain the same and are extended with

$$
L_{n}\left(s_{n}\right)=L_{n}^{\prime}\left(s_{n}^{\prime}\right)=\varnothing \quad \text { and } \quad L_{n}\left(t_{n}\right)=L_{n}^{\prime}\left(t_{n}^{\prime}\right)=\{a\}
$$

Finally, the transition relations $\rightarrow_{n}$ and $\rightarrow_{n}^{\prime}$ contain $\rightarrow_{n-1}^{\prime}\left(\right.$ where $\rightarrow_{-1}=\varnothing$ ), as well as the transitions

$$
\begin{array}{llll}
T S_{n}: & s_{n} \rightarrow_{n} t_{n}, & t_{n} \rightarrow_{n} t_{n}, & t_{n} \rightarrow_{n} s_{n-1}^{\prime}, \\
T S_{n}^{\prime}: & s_{n}^{\prime} \rightarrow_{n} \rightarrow_{n} t_{n}^{\prime}, & t_{n}^{\prime} \rightarrow_{n}^{\prime} t_{n}^{\prime}, & t_{n}^{\prime} \rightarrow_{n}^{\prime} s_{n-1}^{\prime}
\end{array}
$$

where the action labels are omitted for simplicity.
Thus, the only difference between $T S_{n}$ and $T S_{n}^{\prime}$ is the fact that $T S_{n}$ includes the edge $t_{n} \rightarrow s_{n}$, whereas this edge is absent in $T S_{n}^{\prime}$. Some example instantiations of $T S_{n}$ and $T S_{n}^{\prime}$ are indicated in Figure 6.9.


Figure 6.9: The transition systems $T S_{1}$ (upper) and $T S_{1}^{\prime}$ (lower).

It follows from the construction of $T S_{n}$ and $T S_{n}^{\prime}$ that

$$
T S_{n} \not \models \diamond \square a \quad \text { and } \quad T S_{n}^{\prime} \models \diamond \square a \quad \text { for all } n \geqslant 0
$$

This can be proven as follows. $T S_{n}$ contains the initial path $\left(s_{n} t_{n}\right)^{\omega}$ that visits $s_{n}$ and $t_{n}$ in an alternating fashion. We have that

$$
\operatorname{trace}\left(\left(s_{n} t_{n}\right)^{\omega}\right)=\varnothing\{a\} \varnothing\{a\} \varnothing \ldots \quad \text { thus } \quad \operatorname{trace}\left(\left(s_{n} t_{n}\right)^{\omega}\right) \not \models \diamond \square a .
$$

As the considered path is initial, it follows that $T S_{n} \not \vDash \diamond \square a$. On the other hand, as $T S_{n}^{\prime}$ has no opportunity to infinitely often return to an $\neg a$-state, each initial path in $T S_{n}^{\prime}$ is of the following form:

$$
\pi_{i}^{\prime}=s_{n}^{\prime} t_{n}^{\prime} \ldots s_{i}^{\prime}\left(t_{i}^{\prime}\right)^{\omega}
$$

for some $i$. Due to the fact that

$$
\operatorname{trace}\left(\pi_{i}^{\prime}\right)=\varnothing\{a\} \varnothing \ldots \varnothing(\{a\})^{\omega}
$$

it follows that $\operatorname{trace}\left(\pi_{i}^{\prime}\right) \models \diamond \square a$. As this applies to any initial path of $T S_{n}^{\prime}$, we have $T S_{n}^{\prime} \models \diamond \square a$
By means of induction on $n$, it can be proven that $T S_{n}$ and $T S_{n}^{\prime}$ cannot be distinguished by any CTL formula of length at most $n$. That is to say, for all $n \geqslant 0$,
$\forall \mathrm{CTL}$ formula $\Phi$ with $|\Phi| \leqslant n: T S_{n} \models \Phi \quad$ if and only if $\quad T S_{n}^{\prime} \models \Phi$.
(The proof of this fact is left to the interested reader.)


Figure 6.10: Two transition systems for $\forall \square \exists \diamond a$.

The final step of the proof is now as follows. Assume there is a CTL formula $\Phi$ that is equivalent to $\diamond \square a$. Let $n=|\Phi|$ be the length of the formula $\Phi$ and

$$
T S=T S_{n}, \quad T S^{\prime}=T S_{n}^{\prime}
$$

On the one hand, it follows from $T S \not \models \diamond \square a$ and $T S^{\prime} \models \diamond \square a$ that

$$
T S \not \vDash \Phi \text { and } T S^{\prime} \models \Phi .
$$

On the other hand, it results from the fact that $T S_{n}$ and $T S_{n}^{\prime \prime}$ cannot be distinguished by a CTL formula (of length at most $n$ ) that $\Phi$ has the same truth-value under $T S$ and under $T S^{\prime}$. This yields a contradiction.
(b) We concentrate on the proof that there does not exist an equivalent formulation of the CTL formula $\forall \square \exists \diamond a$ in LTL. The fact that there do not exist equivalent LTL formulations of the CTL formulae $\forall \diamond \forall \square a$ and $\forall \diamond(a \wedge \forall \bigcirc a)$ follows directly from Lemmas 6.19 and 6.20 respectively, and Theorem 6.18. The proof for $\forall \square \exists \diamond a$ is by contraposition. Let $\varphi$ be an LTL formula that is equivalent to $\forall \square \exists \diamond a$. Consider the transition system $T S$ depicted in Figure 6.10(a). As $T S \models \forall \square \exists \diamond a$, and $\varphi \equiv \forall \diamond \exists \diamond a$, it follows that $T S \models \varphi$, and therefore

$$
\operatorname{Traces}(T S) \subseteq \operatorname{Words}(\varphi)
$$

Since $\pi=s^{\omega}$ is a path in $T S$, it follows that

$$
\operatorname{trace}(\pi)=\varnothing \varnothing \varnothing \varnothing \ldots \in \operatorname{Traces}(T S) \subseteq \operatorname{Words}(\varphi)
$$

Now consider the transition system $T S^{\prime}$ depicted in Figure 6.10(b). Note that $T S^{\prime \prime}$ is part of transition system $T S$. The paths starting in $s$ in $T S^{\prime}$ are also paths starting from $s$ in $T S$, so we have $s \models \varphi$ in $T S^{\prime}$ and thus $T S^{\prime} \models \varphi$. However, $s \not \vDash \forall \square \exists \diamond a$, since $\exists \diamond a$ is never valid along the only path $s^{\omega}$, and thus $T S^{\prime} \not \vDash \forall \square \exists \diamond a$. This is a contradiction.

Note that the CTL formula $\forall \square \exists \diamond a$ is of significant practical use, since it expresses the fact that it is possible to reach a state for which $a$ holds irrespective of the current state. If $a$ characterizes a state where a certain error is repaired, the formula expresses the fact that it is always possible to recover from that error.

### 6.4 CTL Model Checking

This section is concerned with CTL model checking, which means a decision algorithm that checks whether $T S \models \Phi$ for a given transition system $T S$ and CTL formula $\Phi$. Throughout this section, it is assumed that $T S$ is finite, and has no terminal states. We will see that CTL-model checking can be performed by a recursive procedure that calculates the satisfaction set for all subformulae of $\Phi$ and finally checks whether all initial states belong to the satisfaction set of $\Phi$.

Throughout this section, we consider CTL formulae in ENF, i.e., CTL formulae built by the basic modalities $\exists \bigcirc, \exists \mathrm{U}$, and $\exists \square$. This requires algorithms that generate $\operatorname{Sat}(\exists \bigcirc \Phi)$, $\operatorname{Sat}(\exists(\Phi \cup \Psi)$ ), and $\operatorname{Sat}(\exists \square \Phi)$ when $\operatorname{Sat}(\Phi)$ and $\operatorname{Sat}(\Psi)$ are given. Although each CTL formula can be transformed algorithmically into an equivalent ENF formula, for an implementation of the CTL model-checking algorithm it is recommended to use similar techniques to handle universal quantification, i.e., to design algorithms that generate $\operatorname{Sat}(\forall \bigcirc \Phi)$, $\operatorname{Sat}(\forall(\Phi \cup \Psi))$, and $\operatorname{Sat}(\forall \square \Phi)$ directly from $\operatorname{Sat}(\Phi)$ and $\operatorname{Sat}(\Psi)$. (The same holds for other derived operators such as W or R.)

### 6.4.1 Basic Algorithm

The model-checking problem for CTL is to verify for a given transition system $T S$ and CTL formula $\Phi$ whether $T S \models \Phi$. That is, we need to establish whether the formula $\Phi$ is valid in each initial state $s$ of $T S$. The basic procedure for CTL model checking is rather straightforward:

- the set $\operatorname{Sat}(\Phi)$ of all states satisfying $\Phi$ is computed recursively, and
- it follows that $T S \models \Phi$ if and only if $I \subseteq \operatorname{Sat}(\Phi)$
where $I$ is the set of initial states of $T S$. Note that by computing $\operatorname{Sat}(\Phi)$ a more general problem than just checking whether $T S \models \Phi$ is solved. In fact, it checks for any state $s$ in $S$ whether $s \neq \Phi$, and not just for the initial states. This is sometimes referred
to as a global model-checking procedure. The basic idea of the algorithm is sketched in Algorithm 13 where $\operatorname{Sub}(\Phi)$ is the set of subformulae of $\Phi$. In the sequel of this section it is assumed that TS is finite and has no terminal states.

```
Algorithm 13 CTL model checking (basic idea)
Input: finite transition system \(T S\) and CTL formula \(\Phi\) (both over AP)
Output: \(T S \models \Phi\)
```

```
                                    (* compute the sets \(\left.\operatorname{Sat}(\Phi)=\{s \in S \mid s \models \Phi\}^{*}\right)\)
```

                                    (* compute the sets \(\left.\operatorname{Sat}(\Phi)=\{s \in S \mid s \models \Phi\}^{*}\right)\)
    for all $i \leqslant|\Phi|$ do
for all $i \leqslant|\Phi|$ do
for all $\Psi \in \operatorname{Sub}(\Phi)$ with $|\Psi|=i$ do
for all $\Psi \in \operatorname{Sub}(\Phi)$ with $|\Psi|=i$ do
compute $\operatorname{Sat}(\Psi)$ from $\operatorname{Sat}\left(\Psi^{\prime}\right) \quad$ (* for maximal genuine $\left.\Psi^{\prime} \in \operatorname{Sub}(\Psi)^{*}\right)$
compute $\operatorname{Sat}(\Psi)$ from $\operatorname{Sat}\left(\Psi^{\prime}\right) \quad$ (* for maximal genuine $\left.\Psi^{\prime} \in \operatorname{Sub}(\Psi)^{*}\right)$
od
od
od
od
return $I \subseteq \operatorname{Sat}(\Phi)$

```
return \(I \subseteq \operatorname{Sat}(\Phi)\)
```

The recursive computation of $\operatorname{Sat}(\Phi)$ basically boils down to a bottom-up traversal of the parse tree of the CTL state formula $\Phi$. The nodes of the parse tree represent the subformulae of $\Phi$. The leaves stand for the constant true or an atomic proposition $a \in A P$. All inner nodes are labeled with an operator. For ENF formulae the labels of the inner nodes are $\neg, \wedge, \exists \bigcirc, \exists \mathrm{U}$, or $\exists \square$.

For each node of the parse tree, i.e., for each subformula $\Psi$ of $\Phi$, the set $\operatorname{Sat}(\Psi)$ of states is computed for which $\Psi$ holds. This computation is carried out in a bottom-up fashion, starting from the leaves of the parse tree and finishing at the root of the tree, the (unique) node in the parse tree that corresponds to $\Phi$. At an intermediate node, the results of the computations of its children are used and combined in an appropriate way to establish the states of its associated subformula. The type of computation at such node, $v$ say, depends on the operator (e.g., $\wedge, \exists \bigcirc$, or $\exists \mathrm{U}$ ) that is at the "top level" of the subformula treated. The children of node $v$ stand for the maximal genuine subformulae of the formula $\Psi_{v}$ that is represented by $v$. As soon as $\operatorname{Sat}(\Psi)$ is computed, subformula $\Psi$ is (theoretically) replaced by a new atomic proposition $a_{\Psi}$, and the labeling function $L$ is adjusted as follows: $a_{\Psi}$ is added to $L(s)$ if and only if $s \in \operatorname{Sat}(\Psi)$. Once the bottom-up computations continue with the father, $w$ say, of node $v, \Psi=\Psi_{v}$ is a maximal genuine subformula of $\Psi_{w}$, and all states that are labeled with $a_{\Psi}$ are known to satisfy $\Psi$. In fact, one might say that $\Psi$ is replaced in the formula by the atomic proposition $a_{\Psi}$. This technique will be of importance when treating the model checking of CTL with fairness.

## Example 6.22.

Consider the following state formula over $A P=\{a, b, c\}$ :

$$
\Phi=\underbrace{\exists \bigcirc a}_{\Psi} \wedge \underbrace{\exists(b \cup \underbrace{\exists \square \neg c)}_{\Psi^{\prime \prime}}}_{\Psi^{\prime}}
$$

The indicated formulae $\Psi$ and $\Psi^{\prime}$ are the maximal proper subformulae of $\Phi$, while $\Psi^{\prime \prime}$ is a maximal proper subformula of $\Psi$. The syntax tree for $\Phi$ is of the following form:


The satisfaction sets for the leaves result directly from the labeling function $L$. The treatment of subformula $\neg c$ only needs the satisfaction set for $S a t(c)$ to be complemented. Using $\operatorname{Sat}(\neg c)$, the set $\operatorname{Sat}(\exists \square \neg c)$ can be computed. The subformula $\Psi^{\prime \prime}$ can now be replaced by the fresh atomic proposition $a_{3}$ where $a_{3} \in L(s)$ if and only if $s \in \operatorname{Sat}(\exists \square \neg c)$. The computation now continues with determining $\operatorname{Sat}\left(\exists\left(b \cup a_{3}\right)\right)$. In a similar way, $\operatorname{Sat}(\exists \bigcirc a)$ can be computed by means of $\operatorname{Sat}(a)$.

Once the subformulae $\Psi$ and $\Psi^{\prime}$ are treated, they can be replaced by the atomic propositions $a_{1}, a_{2}$, respectively, such that

$$
a_{1} \in L(s) \quad \text { iff } \quad s \models \exists \bigcirc a \quad \text { and } \quad a_{2} \in L(s) \quad \text { iff } \quad s \models \exists\left(b \cup a_{3}\right) .
$$

The formula that is to be treated for the root node simply thus is: $\Phi^{\prime}=a_{1} \wedge a_{2}$. $\operatorname{Sat}\left(\Phi^{\prime}\right)$ results from intersecting $\operatorname{Sat}\left(a_{1}\right)=\operatorname{Sat}(\Psi)$ and $\operatorname{Sat}\left(a_{2}\right)=\operatorname{Sat}\left(\Psi^{\prime}\right)$. Note that $a_{1}, a_{2}$, and $a_{3}$ are fresh atomic propositions, i.e., $\left\{a_{1}, a_{2}, a_{3}\right\} \cap A P=\varnothing$. The above procedure thus is considered over $A P^{\prime}=A P \cup\left\{a_{1}, a_{2}, a_{3}\right\}$.

Theorem 6.23. Characterization of Sat $(\cdot)$ for CTL formulae in ENF
Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system without terminal states. For all $C T L$ formulae $\Phi, \Psi$ over AP it holds that
(a) $\operatorname{Sat}($ true $)=S$,
(b) $\operatorname{Sat}(a)=\{s \in S \mid a \in L(s)\}$, for any $a \in A P$,
(c) $\operatorname{Sat}(\Phi \wedge \Psi)=\operatorname{Sat}(\Phi) \cap \operatorname{Sat}(\Psi)$,
(d) $\operatorname{Sat}(\neg \Phi)=S \backslash \operatorname{Sat}(\Phi)$,
(e) $\operatorname{Sat}(\exists \bigcirc \Phi)=\{s \in S \mid \operatorname{Post}(s) \cap \operatorname{Sat}(\Phi) \neq \varnothing\}$,
(f) $\operatorname{Sat}(\exists(\Phi \cup \Psi))$ is the smallest subset $T$ of $S$, such that
(1) $\operatorname{Sat}(\Psi) \subseteq T$ and (2) $s \in \operatorname{Sat}(\Phi)$ and $\operatorname{Post}(s) \cap T \neq \varnothing$ implies $s \in T$,
(g) $\operatorname{Sat}(\exists \square \Phi)$ is the largest subset $T$ of $S$, such that
(3) $T \subseteq \operatorname{Sat}(\Phi)$ and (4) $s \in T$ implies $\operatorname{Post}(s) \cap T \neq \varnothing$.

In the clauses (f) and (g), the terms "smallest" and "largest" should be interpreted with respect to the partial order induced by set inclusion.

Proof: The validity of the claims indicated in (a) through (e) is straightforward. We only prove the propositions (f) and (g):

Proof of (f): The proof of this claim consists of two parts:
(i) Show that $T=\operatorname{Sat}(\exists(\Phi \cup \Psi))$ satisfies (1) and (2). From the expansion law

$$
\exists(\Phi \cup \Psi) \equiv \Psi \vee(\Phi \wedge \exists \bigcirc \exists(\Phi \cup \Psi))
$$

it directly follows that $T$ satisfies the properties (1) and (2).
(ii) Show that for any $T$ satisfying properties (1) and (2) we have

$$
\operatorname{Sat}(\exists(\Phi \cup \Psi)) \subseteq T
$$

This is proven as follows. Let $s \in \operatorname{Sat}(\exists(\Phi \cup \Psi))$. Distinguish between $s \in \operatorname{Sat}(\Psi)$ and $s \notin \operatorname{Sat}(\Psi)$. If $s \in \operatorname{Sat}(\Psi)$, it follows from (1) that $s \in T$. In case $s \notin \operatorname{Sat}(\Psi)$, there exists a path $\pi=s_{0} s_{1} s_{2} \ldots$ starting in $s=s_{0}$, such that $\pi \models \Phi \cup \Psi$. Let $n>0$, such that $s_{i}=\Phi, 0 \leqslant i<n$, and $s_{n} \neq \Psi$. Then:

- $s_{n} \in \operatorname{Sat}(\Psi) \subseteq T$,
- $s_{n-1} \in T$, since $s_{n} \in \operatorname{Post}\left(s_{n-1}\right) \cap T$ and $s_{n-1} \in \operatorname{Sat}(\Phi)$,
- $s_{n-2} \in T$, since $s_{n-1} \in \operatorname{Post}\left(s_{n-2}\right) \cap T$ and $s_{n-2} \in \operatorname{Sat}(\Phi)$,
- .......,
- $s_{1} \in T$, since $s_{2} \in \operatorname{Post}\left(s_{1}\right) \cap T$ and $s_{1} \in \operatorname{Sat}(\Phi)$, and finally
- $s_{0} \in T$, since $s_{1} \in \operatorname{Post}\left(s_{0}\right) \cap T$ and $s_{0} \in \operatorname{Sat}(\Phi)$.

It thus follows that $s=s_{0} \in T$.

Proof of $(\mathrm{g})$ : The proof of this claim consists of two parts:
(i) Show that $T=\operatorname{Sat}(\exists \square \Phi)$ satisfies (3) and (4). From the expansion law

$$
\exists \square \Phi \equiv \Phi \wedge \exists \bigcirc \exists \square \Phi
$$

it directly follows that $T$ satisfies the properties (3) and (4).
(ii) Show that for any $T$ satisfying properties (3) and (4)

$$
T \subseteq \operatorname{Sat}(\exists \square \Phi)
$$

This proof goes as follows. Let $T \subseteq S$ satisfy (3) and (4) and $s \in T$. Let $\pi=$ $s_{0} s_{1} s_{2} \ldots$ be a path starting in $s=s_{0}$. (As $T S$ has no terminal states, such a path exists.) Then we derive

- $s_{0}=s$.
- Since $s_{0} \in T$, there exists a state $s_{1} \in \operatorname{Post}\left(s_{0}\right) \cap T$.
- Since $s_{1} \in T$, there exists a state $s_{2} \in \operatorname{Post}\left(s_{1}\right) \cap T$.
- .......

Here, property (4) is exploited in every step. From property (3), it follows that

$$
s_{i} \in T \subseteq \operatorname{Sat}(\Phi), \quad i \geqslant 0
$$

Thus, $\pi=s_{0} s_{1} s_{2} \ldots$ satisfies $\square \Phi$. It follows that

$$
s \in \operatorname{Sat}(\exists \square \Phi) .
$$

As this reasoning applies to any $s \in T$, it follows that $T \subseteq \operatorname{Sat}(\exists \square \Phi)$.

Remark 6.24. Alternative Formulation of $\operatorname{Sat}(\exists(\Phi \cup \Psi)$ and $\operatorname{Sat}(\exists \square \Phi)$
The characterizations of the sets $\operatorname{Sat}(\exists(\Phi \cup \Psi))$ and $\operatorname{Sat}(\exists \square \Phi)$ indicated in Theorem 6.23
are based upon the fixed-point equation induced by the expansion laws for $\exists(\Phi \cup \Psi)$ and$\Phi$, respectively. Consider, for instance, the expansion law

$$
\exists(\Phi \cup \Psi) \equiv \Psi \quad \vee \quad(\Phi \wedge \exists \bigcirc \exists(\Phi \cup \Psi))
$$

The recursive nature of this law suggests to considering the CTL formula $\exists(\Phi \cup \Psi)$ as a fixed point of the logical equation

$$
F \equiv \Psi \vee(\Phi \wedge \exists \bigcirc F)
$$

By the expansion law $F=\exists(\Phi \cup \Psi)$ is a solution, but there are also other solutions that are not equivalent to $\exists(\Phi \cup \Psi)$, such as $F=\exists(\Phi W \Psi)$ (see Remark 6.25). However, a unique characterization of $\exists(\Phi \cup \Psi)$ is obtained by the fact that $\exists(\Phi \cup \Psi)$ is the least solution of $F \equiv \Psi \vee(\Phi \wedge \exists \bigcirc F)$. Using a set-theoretical counterpart by means of $\operatorname{Sat}(\cdot)$, we obtain the following equivalent formulation of constraint (f) in Theorem 6.23:
(f $\left.\mathrm{f}^{\prime}\right) \operatorname{Sat}(\exists(\Phi \cup \Psi))$ is the smallest set $T \subseteq S$ satisfying

$$
\operatorname{Sat}(\Psi) \cup\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T \neq \varnothing\} \subseteq T
$$

In fact, " $\subseteq$ " may be replaced by " $=$ ".
In a similar way, $\exists \square \Phi$ can be considered as the greatest fixed point of the logical equation

$$
F=\Phi \wedge \exists \bigcirc F
$$

Using a set-theoretical counterpart of this equation we obtain the following equivalent formulation of constraint (g) in Theorem 6.23:
$\left(\mathrm{g}^{\prime}\right) \operatorname{Sat}(\exists \square \Phi)$ is the largest set $T \subseteq S$ satisfying

$$
T \subseteq\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T \neq \varnothing\}
$$

Also in this characterization " $\subseteq$ " may be replaced by " $=$ ".

Characterizations of the satisfaction sets for universally quantified CTL formulae can be obtained using the result in Theorem 6.23. This yields
(h) $\operatorname{Sat}(\forall \bigcirc \Phi)=\{s \in S \mid \operatorname{Post}(s) \subseteq \operatorname{Sat}(\Phi)\}$.
(i) $\operatorname{Sat}(\forall(\Phi \cup \Psi))$ is the smallest set $T \subseteq S$ satisfying

$$
\operatorname{Sat}(\Psi) \cup\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \subseteq T\} \subseteq T
$$

(j) $\operatorname{Sat}(\forall \square \Phi)$ is the largest set $T \subseteq S$ satisfying

$$
T \subseteq\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \subseteq T\}
$$

## Remark 6.25. Weak Until

The weak-until operator satisfies the same expansion laws as the until operator.

$$
\begin{aligned}
& \exists(\Phi W \Psi) \equiv \Psi \vee(\Phi \wedge \exists \bigcirc \exists(\Phi \mathrm{W} \Psi)), \\
& \forall(\Phi \mathrm{W} \Psi) \equiv \Psi \vee(\Psi \wedge \forall \bigcirc \forall(\Phi W \Psi))
\end{aligned}
$$

The difference, however, is that the weak-until operator represents the largest solution (i.e., fixed point) of the expansion law, whereas the until operator denotes the smallest solution. The satisfaction sets for weak until are characterized as follows:
(k) $\operatorname{Sat}(\exists(\Phi \mathrm{W} \Psi))$ is the largest set $T \subseteq S$ satisfying

$$
T \subseteq \operatorname{Sat}(\Psi) \cup\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T \neq \varnothing\}
$$

(l) $\operatorname{Sat}(\forall(\Phi \mathrm{W} \Psi))$ is the largest set $T \subseteq S$ satisfying

$$
T \subseteq \operatorname{Sat}(\Psi) \cup\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \subseteq T\}
$$

Without loss of generality, we may assume that the CTL formula $\Phi$ to be verified is in ENF (see Theorem 6.14, page 332). That is, the model-checking algorithm is supposed to be preceded by transforming the CTL formula at hand into ENF. The characterizations of the satisfaction sets indicated in Theorem 6.23 are exploited to compute the sets $\operatorname{Sat}(\cdot)$. The essential steps for computing the satisfaction sets are summarized in Algorithm 14 on page 348 .

### 6.4.2 The Until and Existential Always Operator

To treat constrained reachability properties, given by CTL formulae of the form $\exists(\Phi \cup \Psi)$, the characterization in Theorem 6.23 is exploited. Recall that $\operatorname{Sat}(\exists(\Phi \cup \Psi))$ is character-

```
Algorithm 14 Computation of the satisfaction sets
```

Input: finite transition system $T S$ with state set $S$ and CTL formula $\Phi$ in ENF
Output: $\operatorname{Sat}(\Phi)=\{s \in S \mid s \models \Phi\}$
(* recursive computation of the sets $\operatorname{Sat}(\Psi)$ for all subformulae $\Psi$ of $\Phi^{*}$ )

```
switch(\Phi):
```

| true | return $S$; |
| :---: | :---: |
| $a$ | return $\{s \in S \mid a \in L(s)\}$; |
| $\Phi_{1} \wedge \Phi_{2}$ | return $\operatorname{Sat}\left(\Phi_{1}\right) \cap \operatorname{Sat}\left(\Phi_{2}\right)$; |
| $\neg \Psi$ | return $S \backslash \operatorname{Sat}(\Psi)$; |
| $\exists \bigcirc \Psi$ | return $\{s \in S \mid \operatorname{Post}(s) \cap \operatorname{Sat}(\Psi) \neq \varnothing\} ;$ |
| $\exists\left(\Phi_{1} \cup \Phi_{2}\right)$ | $\begin{aligned} & T:=\operatorname{Sat}\left(\Phi_{2}\right) ;\left({ }^{*} \text { compute the smallest fixed point }{ }^{*}\right) \\ & \text { while }\left\{s \in \operatorname{Sat}\left(\Phi_{1}\right) \backslash T \mid \operatorname{Post}(s) \cap T \neq \varnothing\right\} \neq \varnothing \text { do } \\ & \quad \text { let } s \in\left\{s \in \operatorname{Sat}\left(\Phi_{1}\right) \backslash T \mid \operatorname{Post}(s) \cap T \neq \varnothing\right\} ; \\ & \quad T:=T \cup\{s\} ; \end{aligned}$ |
|  | od; <br> return $T$; |
| $\exists \square \Phi$ | $\begin{aligned} & T:=\operatorname{Sat}(\Phi) ; \quad\left(* \text { compute the greatest fixed point }{ }^{*}\right) \\ & \text { while }\{s \in T \mid \operatorname{Post}(s) \cap T=\varnothing\} \neq \varnothing \text { do } \\ & \quad \text { let } s \in\{s \in T \mid \operatorname{Post}(s) \cap T=\varnothing\} \\ & \quad T:=T \backslash\{s\} ; \end{aligned}$ |
|  | od; return $T$; |

ized as the smallest set $T \subseteq S$, where $S$ is the set of states in the transition system under consideration, such that
(1) $\operatorname{Sat}(\Psi) \subseteq T$ and $(2)(s \in \operatorname{Sat}(\Phi)$ and $\operatorname{Post}(s) \cap T \neq \varnothing) \Rightarrow s \in T$.

This characterization suggests adopting the following iterative procedure to compute $\operatorname{Sat}(\exists(\Phi \cup \Psi))$ :

$$
T_{0}=\operatorname{Sat}(\Psi) \quad \text { and } \quad T_{i+1}=T_{i} \cup\left\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T_{i} \neq \varnothing\right\}
$$

Intuitively speaking, the set $T_{i}$ contains all states that can reach a $\Psi$-state in at most $i$ steps via a $\Phi$-path. This is to be understood as follows. From the fact that $\operatorname{Sat}(\exists(\Phi \cup \Psi))$ satisfies the conditions (1) and (2) in (6.1), it can be demonstrated by induction on $j$ that

$$
T_{0} \subseteq T_{1} \subseteq T_{2} \subseteq \ldots \subseteq T_{j} \subseteq T_{j+1} \subseteq \ldots \subseteq \operatorname{Sat}(\exists(\Phi \cup \Psi))
$$

Since we assume a finite transition system $T S$, there exists a $j \geqslant 0$ such that

$$
T_{j}=T_{j+1}=T_{j+2}=\ldots
$$

Therefore

$$
T_{j}=T_{j} \cup\left\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T_{j} \neq \varnothing\right\}
$$

and, hence:

$$
\left\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T_{j} \neq \varnothing\right\} \subseteq T_{j} .
$$

Hence, $T_{j}$ satisfies property (2). Further,

$$
\operatorname{Sat}(\Psi)=T_{0} \subseteq T_{j} .
$$

These considerations show that $T_{j}$ possesses the properties (1) and (2). Since $\operatorname{Sat}(\exists(\Phi \cup \Psi))$ is the smallest set of states satisfying the properties (1) and (2), it follows that

$$
\operatorname{Sat}(\exists(\Phi \cup \Psi)) \subseteq T_{j}
$$

and thus $\operatorname{Sat}(\exists(\Phi \cup \Psi))=T_{j}$. Hence, for any $j \geqslant 0$ we have

$$
T_{0} \varsubsetneqq T_{1} \varsubsetneqq T_{2} \varsubsetneqq \ldots \varsubsetneqq T_{j}=T_{j+1}=\ldots=\operatorname{Sat}(\exists(\Phi \cup \Psi)) .
$$

Algorithm 15 (see page 351) shows a more detailed version of the backward search indicated earlier in Algorithm 14 (see page 348). As each $\Psi$-state obviously satisfies $\exists(\Phi \cup \Psi)$, all states in $\operatorname{Sat}(\Psi)$ are initially considered to satisfy $\exists(\Phi \cup \Psi))$. This conforms to the initialization to the variable $E$. An iterative procedure is subsequently started that can be considered to systematically check the state space in a "backward" manner. In each iteration, all $\Phi$-states are determined that can move by a single transition to (one of) the states of which we already know to satisfy $\exists(\Phi \cup \Psi)$ ). Thus, in the $i$ th iteration of the procedure, all $\Phi$-states are considered that can move to a $\Psi$-state in at most $i$ steps. This corresponds to the set $T_{i}$. Termination of the algorithm intuitively follows, as the number of states in the transition system is finite. Note that the algorithm assumes a transition system representation (or its state graph) by means of "inverse" adjacency lists, i.e., list representations for the sets of predecessors $\operatorname{Pre}\left(s^{\prime}\right)=\left\{s \in S \mid s^{\prime} \in \operatorname{Post}(s)\right\}$.

## Example 6.26.

Consider the transition system depicted in Figure 6.11, and suppose we are interested in checking the formula $\exists \diamond \Phi$ with $\Phi=((a=c) \wedge(a \neq b))$. Recall that $\exists \diamond \Phi=\exists(\operatorname{true} \cup \Phi)$. To check $\exists \diamond \Phi$ we invoke Algorithm 14 (see page 348). This algorithm recursively computes $\operatorname{Sat}($ true ) and $\operatorname{Sat}((a=c) \wedge(a \neq b))$. This corresponds to the situation depicted in Figure 6.12(a), where all states in the set $T$ are colored black, and white otherwise. In the first iteration, we select and delete $s_{5}$ from $E$, but as $\operatorname{Pre}\left(s_{5}\right)=\varnothing, T$ remains unaffected. On considering $s_{4} \in E, \operatorname{Pre}\left(s_{4}\right)=\left\{s_{6}\right\}$ is added to $T$ (and $E$ ), see Figure 6.12(b). During the next iteration, the only predecessor of $s_{6}$ is added, yielding the snapshot in Figure 6.12(c). After the fourth iteration, the algorithm terminates as there are no new predecessors of $\Phi$-states encountered, i.e., $E=\varnothing$ (see Figure 6.12(d)).


Figure 6.11: An example of a transition system.


Figure 6.12: Example of backward search for $\exists(\operatorname{true} \cup(a=c) \wedge(a \neq b))$.

```
Algorithm 15 Enumerative backward search for computing \(\operatorname{Sat}(\exists(\Phi \cup \Psi))\)
Input: finite transition system \(T S\) with state set \(S\) and CTL formula \(\exists(\Phi \cup \Psi)\)
Output: \(\operatorname{Sat}(\exists(\Phi \cup \Psi))=\{s \in S \mid s \models \exists(\Phi \cup \Psi)\}\)
```

```
\(E:=\operatorname{Sat}(\Psi) ; \quad \quad\left(* E\right.\) administers the states \(s\) with \(\left.s \vDash \exists(\Phi \cup \Psi)^{*}\right)\)
```

$E:=\operatorname{Sat}(\Psi) ; \quad \quad\left(* E\right.$ administers the states $s$ with $\left.s \vDash \exists(\Phi \cup \Psi)^{*}\right)$
$T:=E ; \quad\left(* T\right.$ contains the already visited states $s$ with $\left.s \models \exists(\Phi \cup \Psi){ }^{*}\right)$
$T:=E ; \quad\left(* T\right.$ contains the already visited states $s$ with $\left.s \models \exists(\Phi \cup \Psi){ }^{*}\right)$
while $E \neq \varnothing$ do
while $E \neq \varnothing$ do
let $s^{\prime} \in E$;
let $s^{\prime} \in E$;
$E:=E \backslash\left\{s^{\prime}\right\} ;$
$E:=E \backslash\left\{s^{\prime}\right\} ;$
for all $s \in \operatorname{Pre}\left(s^{\prime}\right)$ do
for all $s \in \operatorname{Pre}\left(s^{\prime}\right)$ do
if $s \in \operatorname{Sat}(\Phi) \backslash T$ then $E:=E \cup\{s\} ; T:=T \cup\{s\}$ fi
if $s \in \operatorname{Sat}(\Phi) \backslash T$ then $E:=E \cup\{s\} ; T:=T \cup\{s\}$ fi
od
od
od
od
return $T$

```
return \(T\)
```

Let us now consider the computation of $\operatorname{Sat}(\exists \square \Phi)$ for the transition system $T S$. As for the until-operator, the algorithm for $\exists \square \Phi$ is based on the characterization in Theorem 6.23, i.e., $\operatorname{Sat}(\exists \square \Phi)$ is the largest set $T \subseteq S$ satisfying

$$
T \subseteq \operatorname{Sat}(\Phi) \quad \text { and } \quad(s \in T \text { implies } T \cap \operatorname{Post}(s) \neq \varnothing) .
$$

The basic idea is to compute $\operatorname{Sat}(\exists \square \Phi)$ by means of the iteration

$$
T_{0}=\operatorname{Sat}(\Phi) \quad \text { and } \quad T_{i+1}=T_{i} \cap\left\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T_{i} \neq \varnothing\right\}
$$

Then, for all $j \geqslant 0$, it holds that

$$
T_{0} \supsetneqq T_{1} \supsetneqq T_{2} \supsetneqq \ldots \supsetneqq T_{j}=T_{j+1}=\ldots=T=\operatorname{Sat}(\exists \square \Phi) .
$$

The above iteration can be realized by means of a backward search starting with

$$
T=\operatorname{Sat}(\Phi) \quad \text { and } \quad E=S \backslash \operatorname{Sat}(\Phi) .
$$

Here $T$ equals $T_{0}$ and $E$ contains all states that refute $\exists \square \Phi$. During the backward search, states are iteratively removed from $T$, for which it has been established that they refute$\Phi$. This applies to any $s \in T$ satisfying

$$
\operatorname{Post}(s) \cap T=\varnothing
$$

Although $s \models \Phi$ (as it is in $T$ ), all its successors refute $\exists \square \Phi$ (as they are not in $T$ ), and therefore $s$ refutes $\exists \square \Phi$. Once such states are encountered, they are inserted in $E$ to enable the possible removal of other states in $T$.

```
Algorithm 16 Enumerative backward search to compute \(\operatorname{Sat}(\exists \square \Phi)\)
Input: finite transition system \(T S\) with state set \(S\) and CTL formula \(\exists \square \Phi\)
Output: \(\operatorname{Sat}(\exists \square \Phi)=\{s \in S \mid s \models \exists \square \Phi\}\)
\(E:=S \backslash \operatorname{Sat}(\Phi) ; \quad \quad{ }^{*} E\) contains any not visited \(s^{\prime}\) with \(\left.s^{\prime} \not \vDash \exists \square \Phi^{*}\right)\)
\(T:=\operatorname{Sat}(\Phi) ; \quad \quad\left({ }^{*} T\right.\) contains any \(s\) for which \(s \models \exists \square \Phi\) has not yet been disproven \(\left.{ }^{*}\right)\)
for all \(s \in \operatorname{Sat}(\Phi)\) do \(\operatorname{count}[s]:=|\operatorname{Post}(s)|\); od (* initialize array count *)
while \(E \neq \varnothing\) do
        let \(s^{\prime} \in E\);
                                    (* loop invariant: count \(\left.[s]=|\operatorname{Post}(s) \cap(T \cup E)|^{*}\right)\)
                                    (* \(\left.s^{\prime} \notin \Phi^{*}\right)\)
        \(E:=E \backslash\left\{s^{\prime}\right\} ;\)
        for all \(s \in \operatorname{Pre}\left(s^{\prime}\right)\) do
                                    (* update counters count \([s]\) for all predecessors \(s\) of \(s^{\prime *}\) )
                if \(s \in T\) then
            count \([s]:=\operatorname{count}[s]-1\);
            if count \([s]=0\) then
                \(T:=T \backslash\{s\} ; \quad\left(* s\right.\) does not have any successor in \(\left.T^{*}\right)\)
                \(E:=E \cup\{s\} ;\)
            fi
                fi
        od
od
return \(T\)
```

The resulting computational procedure is detailed in Algorithm 16 on page 352. In order to support the test whether $\operatorname{Post}(s) \cap T=\varnothing$, a counter count $[s]$ is exploited that keeps track of the number of direct successors of $s$ in $T \cup E$ :

$$
\operatorname{count}[s]=|\operatorname{Post}(s) \cap(T \cup E)| .
$$

Once count $[s]=0$, we have that $\operatorname{Post}(s) \cap(T \cup E)=\varnothing$, and hence $\operatorname{Post}(s) \cap T=\varnothing$. Thus, state $s$ is not in $\operatorname{Sat}(\exists \square \Phi)$ and therefore can be safely removed from $T$. On termination, $E=\varnothing$, and thus count $[s]=|\operatorname{Post}(s) \cap T|$. It follows that any state $s \in \operatorname{Sat}(\Phi)$ for which count $[s]>0$ satisfies the CTL formula $\exists \square \Phi$.

It is left to the interested reader to consider how the outlined approach can be modified to compute $\operatorname{Sat}(\exists(\Phi \mathrm{W} \Psi))$.

## Example 6.27.

Consider the transition system depicted in Figure 6.11 and the formula $\exists \square b$. Initially,

$$
T_{0}=\left\{s_{0}, s_{1}, s_{2}, s_{4}\right\} \quad \text { and } \quad E=\left\{s_{3}, s_{5}, s_{6}, s_{7}\right\} \quad \text { and } \quad \text { count }=[1,1,2,1,2,2,1,2] .
$$

Suppose state $s_{3} \in E$ is selected in the first iteration. As $s_{1} \in \operatorname{Pre}\left(s_{3}\right)$ and $s_{1} \in T$, count $\left[s_{1}\right]:=0$. Accordingly, $s_{1}$ is deleted from $T$ and added to $E$. This yields

$$
T_{1}=\left\{s_{0}, s_{2}, s_{4}\right\} \quad \text { and } \quad E=\left\{s_{1}, s_{5}, s_{6}, s_{7}\right\} \quad \text { and } \quad \text { count }=[1,0,2,1,2,2,1,2] .
$$

We also have $s_{7} \in \operatorname{Pre}\left(s_{3}\right)$, but as $s_{7} \notin T$, this affects neither $\operatorname{count}\left[s_{7}\right], T$ nor $E$.
Suppose $s_{6}$ and $s_{7}$ are selected in the second and third iteration, respectively. None of these states has a predecessor in $T_{1}$, so we obtain

$$
T_{3}=\left\{s_{0}, s_{2}, s_{4}\right\} \quad \text { and } \quad E=\left\{s_{1}, s_{5}\right\} \quad \text { and } \quad \text { count }=[1,0,2,1,2,2,1,2] .
$$

Now select $s_{1} \in E$ in the next iteration. As $\operatorname{Pre}\left(s_{1}\right) \cap T_{3}=\left\{s_{2}, s_{4}\right\}$, the counters for $s_{2}$ and $s_{4}$ are decremented. This yields

$$
T_{4}=\left\{s_{0}, s_{2}, s_{4}\right\} \quad \text { and } \quad E=\left\{s_{5}\right\} \quad \text { and } \quad \text { count }=[1,0,1,1,1,2,1,2] .
$$

As $\operatorname{Pre}\left(s_{5}\right)=\varnothing, T$ and count are unaffected in the subsequent iteration. As $E=\varnothing$, the algorithm terminates and returns $T=\left\{s_{0}, s_{2}, s_{4}\right\}$ as the final outcome.

This section is concluded by outlining an alternative algorithm for computing $\operatorname{Sat}(\exists \square \Phi)$. Since the computation of the satisfaction sets takes place by means of a bottom-up traversal through the parse tree of the formula at hand, it is assumed that $\operatorname{Sat}(\Phi)$ is at our disposal. A possibility to compute $\operatorname{Sat}(\exists \square \Phi)$ is to only consider the $\Phi$-states of transition system $T S$ and ignore all $\neg \Phi$-states. The justification of this modification to $T S$ is that all removed states will not satisfy $\exists \square \Phi$ (as they violate $\Phi$ ) and therefore can be safely removed. For $T S=(S$, Act, $\rightarrow, I, A P, L)$ let $T S[\Phi]=\left(S^{\prime}, A c t, \rightarrow^{\prime}, I^{\prime}, A P, L^{\prime}\right)$ with $S^{\prime}=\operatorname{Sat}(\Phi)$, $\rightarrow^{\prime}=\rightarrow \cap\left(S^{\prime} \times\right.$ Act $\left.\times S^{\prime}\right), I^{\prime}=I \cap S^{\prime}$ and $L^{\prime}(s)=L(s)$ for all $s \in S^{\prime}$. Then, all nontrivial strongly connected components (SCCs) ${ }^{4}$ in the state graph induced by $T S[\Phi]$ are computed. All states in each such SCC $C$ satisfy $\exists \square \Phi$, as any state in $C$ is reachable from any other state in $C$, and-by construction-all states in $C$ satisfy $\Phi$. Finally, all states in $T S[\Phi]$ are computed that can reach such SCC. If state $s \in S^{\prime}$ and there exists such a path, then - by construction of $T S[\Phi]$ — the property $\exists \square \Phi$ is satisfied by $s$; otherwise, it is not. This can be done by a backward search. The worst-case time complexity of this

[^28]

Figure 6.13: Computing $\operatorname{Sat}(\exists \square b)$ using strongly connected components in $T S[\Phi]$.
alternative algorithm is the same as for Algorithm 16. This approach is illustrated by the following example.

Example 6.28. Alternative Algorithm for $\exists \square \Phi$
Consider the transition system of Figure 6.11 and CTL formula $\exists \square b$. The modified transition system $T S[b]$ consists of the four states that are labeled with $b$, see the gray states in Figure 6.13(a). The only nontrivial SCC of this structure is indicated by the black states, see Figure 6.13(b). As there is only a single $b$-state (that is not in the SCC) that can reach the nontrivial SCC, this state satisfies $\exists \square b$, and the computation finishes, see Figure 6.13(c).

## Theorem 6.29.

For state $s$ in transition system TS and CTL formula $\Phi$ :
$s \models \exists \square \Phi$ iff $s \models \Phi$ and there is a nontrivial SCC in $T S[\Phi]$ reachable from $s$.

Proof: $\Rightarrow$ : Suppose $s \vDash \exists \square \Phi$. Clearly, $s$ is a state in $T S[\Phi]$. Let $\pi$ be a path in $T S$ starting in $s$ such that $\pi \models \square \Phi$. As TS is finite, $\pi$ has a suffix $\rho=s_{1} s_{2} \ldots s_{k}$ for $k>1$, representing a cycle that is traversed infinitely often. As $\pi$ is also a path in $T S[\Phi]$, the
states $s_{1}$ through $s_{k}$ are all in $T S[\Phi]$. Since $\pi$ is traversed infinitely often, it represents a cycle, and thus any pair of states $s_{i}$ and $s_{j}$ is mutually reachable. Stated differently, $\left\{s_{1}, \ldots, s_{k}\right\}$ is either an SCC or contained in some SCC in $T S[\Phi]$. As $\pi$ is a path starting in $s$, these states are reachable from $s$.
$\Leftarrow$ : Suppose $s$ is a state in $T S[\Phi]$ and there exists an SCC in $T S[\Phi]$ reachable from $s$. Let $s^{\prime}$ be a state in such SCC. As the SCC is nontrivial, $s^{\prime}$ is reachable from itself by a path of length at least one. Repeating this cycle infinitely often yields an infinite path $\pi$ with $\pi \models \square \Phi$. The path from $s$ to $s^{\prime}$ followed by $\pi[1 .$.$] now satisfies \square \Phi$ and starts in $s$. Thus, $s \vDash \exists \square \Phi$.

### 6.4.3 Time and Space Complexity

The time complexity of the CTL model-checking algorithm is determined as follows. Let $T S$ be a finite transition system with $N$ states and $K$ transitions. Under the assumption that the sets of predecessors $\operatorname{Pre}(\cdot)$ are represented as linked lists, the time complexity of Algorithms 15 and 16 lies in $\mathcal{O}(N+K)$. Given that the computation of the satisfaction sets $\operatorname{Sat}(\Phi)$ is a bottom-up traversal over the parse tree of $\Phi$ and thus linear in $|\Phi|$, the time complexity of Algorithm 14 (see page 348) is in

$$
\mathcal{O}((N+K) \cdot|\Phi|)
$$

When the initial states of $T S$ are administrated in, e.g., a linked list, checking whether $I \subseteq \operatorname{Sat}(\Phi)$ can be done in $\mathcal{O}\left(N_{0}\right)$ where $N_{0}$ is the cardinality of the set $I$. Recall that the CTL model-checking algorithm requires the CTL formula to be checked to be in existential normal form. As the transformation of any CTL formula into an equivalent CTL formula in ENF may yield an exponential blowup of the formula, it is recommended to treat the modalities such as $\forall \mathrm{U}, \forall \diamond, \forall \square, \exists \diamond, \forall \mathrm{W}$, and $\exists \mathrm{W}$, analogous to the introduced approaches for $\exists U$ and $\exists \square$, by exploiting the characterizations of $\operatorname{Sat}(\forall \square \Phi), \operatorname{Sat}(\exists \diamond \Phi)$, and so on. The resulting algorithms are all linear in $N$ and $K$. Thus, we obtain the following theorem:

## Theorem 6.30. Time Complexity of CTL Model Checking

For transition system TS with $N$ states and $K$ transitions, and CTL formula $\Phi$, the CTL model-checking problem $T S=\Phi$ can be determined in time $\mathcal{O}((N+K) \cdot|\Phi|)$.

Let us compare this complexity bound with that for LTL model checking. Recall that LTL model checking is exponential in the size of the formula. Although the difference in time complexity with respect to the length of the formula seems drastic (exponential for LTL vs.
linear for CTL), this should not be interpreted as "CTL model checking is more efficient than LTL model checking". From Theorem 6.18 it follows that whenever $\varphi \equiv \Phi$ for LTL formula $\varphi$ and CTL formula $\Phi$, then $\varphi$ is obtained by removing all path quantifiers in $\Phi$. Thus, CTL formulae are at least as long as their equivalent counterparts in LTL (if these exist). In fact, if $\mathrm{P} \neq \mathrm{NP}$, then there exist LTL formulae $\varphi_{n}$ with $\left|\varphi_{n}\right| \in \mathcal{O}(\operatorname{poly}(n))$ for which CTL-equivalent formulae do exist, but not of polynomial length. LTL formulae may be exponentially shorter than any of their equivalent formulation in CTL. The latter effect is illustrated in the following example. From Lemma 5.45, it follows that the Hamiltonian path problem for digraphs with $n$ nodes can be encoded in LTL formula $\varphi_{n}$ whose length is polynomial in $n$. More precisely, given a digraph $G$ with nodeset $\{1, \ldots, n\}$ there is an LTL formula $\varphi_{n}$ such that (1) $G$ has a Hamiltonian path if and only if $\neg \varphi_{n}$ does not hold for the transition system associated with $G$ and (2) $\neg \varphi_{n}$ has an equivalent CTL formula.

To express the existence of a Hamiltonian path in CTL requires a CTL formula of exponential length, unless $\mathrm{P}=$ NP: if it is assumed that $\neg \varphi_{n} \equiv \neg \Phi_{n}$ for CTL formulae $\Phi_{n}$ of polynomial length, then the Hamiltonian path problem could be solved by CTL model checking, and thus in polynomial time. Since, however, the Hamiltonian path problem is NP-complete, this is only possible for the case of PTIME $=$ NP.

## Example 6.31. The Hamiltonian Path Problem

Consider the NP-complete problem of finding a Hamiltonian path in an arbitrary, connected, directed graph $G=(V, E)$ where $V$ denotes the set of vertices and $E \subseteq V \times V$, the set of edges. Let $V=\left\{v_{1}, \ldots, v_{n}\right\}$. A Hamiltonian path is a (finite) path through the graph $G$ which visits each state exactly once. Starting from graph $G$, a transition system $T S(G)$ is derived as well as a CTL formula $\Phi_{n}$, such that

$$
G \text { contains a Hamiltonian path if and only if } T S \not \vDash \neg \Phi_{n} \text {. }
$$

The transition system $T S$ is defined as follows

$$
T S=(V \cup\{b\},\{\tau\}, \rightarrow, V, V, L)
$$

with $L\left(v_{i}\right)=\left\{v_{i}\right\}, L(b)=\varnothing$. The transition relation $\rightarrow$ is defined by

$$
\frac{\left(v_{i}, v_{j}\right) \in E}{v_{i} \xrightarrow{\tau} v_{j}} \quad \text { and } \quad \frac{v_{i} \in V \cup\{b\}}{v_{i} \xrightarrow{\tau} b} .
$$

All vertices of $G$ are states in the transition system $T S$. A new state $b$ is introduced that is a direct successor of any state (including $b$ ). Figure 6.14 shows (a) an example of a directed graph $G$ and (b) its transition system $T S(G)$. The sole purpose of the new state $b$ is to ensure that the transition system has no terminal states. Note that $T S(G)$ is defined as in the proof for the existence of a polynomial reduction of the Hamiltonian


Figure 6.14: Example of encoding the Hamiltonian path problem as a transition system.
path problem onto the complement of the LTL model-checking problem; see Lemma 5.45 on page 288.

It remains to give a recipe for the construction of $\Phi_{n}$, a CTL formula that captures the existence of a Hamiltonian path. Let $\Phi_{n}$ be defined as follows:

$$
\Phi_{n}=\bigvee_{\substack{\left(i_{1}, \ldots, i_{n}\right) \\ \text { permutation of }(1, \ldots, n)}}^{\bigvee} \Psi\left(v_{i_{1}}, v_{i_{2}}, \ldots, v_{i_{n}}\right)
$$

such that $\Psi\left(v_{i_{1}}, \ldots, v_{i_{n}}\right)$ is a CTL formula that is satisfied if and only if $v_{i_{1}}, v_{i_{2}}, \ldots, v_{i_{n}}$ is a Hamiltonian path in $G$. The formulae $\Psi\left(v_{i_{1}}, \ldots, v_{i_{n}}\right)$ are inductively defined as follows:

$$
\begin{aligned}
\Psi\left(v_{i}\right) & =v_{i} \\
\Psi\left(v_{i_{1}}, v_{i_{2}}, \ldots, v_{i_{n}}\right) & =v_{i_{1}} \wedge \exists \bigcirc \Psi\left(v_{i_{2}}, \ldots, v_{i_{n}}\right) \quad \text { if } n>1 .
\end{aligned}
$$

Stated in words, $\Psi\left(v_{i_{1}}, \ldots, v_{i_{n}}\right)$ holds if there exists a path on which $v_{i_{1}}, v_{i_{2}}, v_{i_{3}}$ successively hold. Since each state has a transition to the $b$ state, the trace $\left\{v_{i_{1}}\right\}\left\{v_{i_{2}}\right\} \ldots\left\{v_{i_{n}}\right\}$ can be extended with $\{b\}^{\omega}$. An example of an instantiation of $\Psi_{n}$ is

$$
\Phi_{2}=\left(v_{1} \wedge \exists \bigcirc v_{2}\right) \vee\left(v_{2} \wedge \exists \bigcirc v_{1}\right)
$$

and

$$
\begin{array}{rlll}
\Phi_{3}= & \left(v_{1} \wedge \exists \bigcirc\left(v_{2} \wedge \exists \bigcirc v_{3}\right)\right) & \vee & \left(v_{1} \wedge \exists \bigcirc\left(v_{3} \wedge \exists \bigcirc v_{2}\right)\right) \\
& \vee\left(v_{2} \wedge \exists \bigcirc\left(v_{1} \wedge \exists \bigcirc v_{3}\right)\right) & \vee & \left(v_{2} \wedge \exists \bigcirc\left(v_{3} \wedge \exists \bigcirc v_{1}\right)\right) \\
& \vee\left(v_{3} \wedge \exists \bigcirc\left(v_{1} \wedge \exists \bigcirc v_{2}\right)\right) & \vee & \left(v_{3} \wedge \exists \bigcirc\left(v_{2} \wedge \exists \bigcirc v_{1}\right)\right)
\end{array}
$$

It is not difficult to infer that

$$
\operatorname{Sat}\left(\Psi\left(v_{i_{1}}, \ldots, v_{i_{n}}\right)\right)= \begin{cases}\left\{v_{i_{1}}\right\} & \text { if } v_{i_{1}}, \ldots, v_{i_{n}} \text { is a Hamiltonian path in } G \\ \varnothing & \text { otherwise }\end{cases}
$$

Thus:

$$
T S \not \models \neg \Phi_{n}
$$

iff there is an initial state $s$ of $T S$ for which $s \not \vDash \neg \Phi_{n}$
iff there is an initial state $s$ of $T S$ for which $s \models \Phi_{n}$
iff $\quad \exists v$ in $G$ and a permutation $i_{1}, \ldots, i_{n}$ of $1, \ldots, n$ with $v \in \operatorname{Sat}\left(\Psi\left(v_{i_{1}}, \ldots, v_{i_{n}}\right)\right)$,
iff $\quad \exists v$ in $G$ and a permutation $i_{1}, \ldots, i_{n}$ of $1, \ldots, n$, such that $v=v_{i_{1}}$ and $v_{i_{1}}, \ldots, v_{i_{n}}$ is a Hamiltonian path in $G$
iff $\quad G$ has a Hamiltonian path.
Thus, $G$ contains a Hamiltonian path if and only if $T S \not \vDash \neg \Phi_{n}$.

By the explicit enumeration of all possible permutations we obtain a formula with a length that is exponential in the number of vertices in the graph. This does not prove that there does not exist an equivalent, but shorter, CTL formula which describes the Hamiltonian path problem. Actually, shorter formalizations in CTL cannot be expected, since the CTL model-checking problem is polynomially solvable whereas the Hamiltonian path problem is NP-complete.

### 6.5 Fairness in CTL

Recall that fairness assumptions (see Section 3.5) are used to rule out certain computations that are considered to be unrealistic for the system under consideration. These unreasonable computations that ignore certain transition alternatives forever are the "unfair" ones; the remaining computations are thus the "fair" ones. As there are different notions of fairness, various distinct forms of fairness can be imposed: unconditional, strong, and weak fairness constraints.

An important distinction between LTL and CTL is that fairness assumptions can be incorporated into LTL without any specific changes while a special treatment of fairness is required for CTL. That is to say, fairness assumptions can be added as premise to the LTL formula to be verified. The LTL model-checking problem where only fair paths are considered (i.e., considering the fair satisfaction relation $\models_{\text {fair }}$ ) can thus be reduced to the traditional LTL model-checking problem, i.e., with respect to the common satisfaction relation $\models$. In this case, the LTL formula $\varphi$ to be verified is replaced by fair $\rightarrow \varphi$ :

$$
T S \models_{\text {fair }} \varphi \quad \text { if and only if } \quad T S \models(\text { fair } \rightarrow \varphi) .{ }^{5}
$$

[^29]For more details we refer to Section 5.1.6 (see page 257).
An analogous approach is not possible for CTL. This stems from the fact that most fairness constraints cannot be encoded in a CTL formula. An indication of this is the fact that persistence properties $\diamond \square a$ are inherent in, e.g., strong fairness conditions $\square \diamond b \rightarrow$ $\square \diamond c \equiv \diamond \square \neg b \vee \square \diamond c$ and cannot be expressed in CTL. To be a bit more precise: fairness constraints operate on the path level and replace the standard meaning "for all paths" of universal quantification with "for all fair paths" and existential quantification "there exists a path" with "there exists a fair path". Thus, if fair expresses the fairness condition on the path level, then CTL formulae that encode the intuitive meaning of $\forall(f a i r \rightarrow \varphi)$ and $\exists(\operatorname{fair} \wedge \varphi)$ would be needed. However, these are not legal CTL formulae since (1) the Boolean connectives $\rightarrow$ and $\wedge$ are not allowed on the level of CTL path formulae and (2) fairness conditions cannot be described by CTL path formulae.

Therefore, an alternative approach is taken to treat fairness in CTL. In order to deal with fairness constraints in CTL, the semantics of CTL is slightly modified such that the state formulae $\forall \varphi$ and $\exists \varphi$ are interpreted over all fair paths rather than over all possible paths. A fair path is a path that satisfies a set of fairness constraints. It is assumed that the given fixed fairness constraint is described by a formula as in LTL.

## Definition 6.32. CTL Fairness Assumptions

A strong CTL fairness constraint (over $A P$ ) is a term of the form

$$
\text { sfair }=\bigwedge_{1 \leqslant i \leqslant k}\left(\square \diamond \Phi_{i} \rightarrow \square \diamond \Psi_{i}\right)
$$

where $\Phi_{i}$ and $\Psi_{i}$ (for $1 \leqslant i \leqslant k$ ) are CTL formulae over AP. Weak and unconditional CTL fairness constraints are defined analogously by conjunctions of terms of the form $\left(\diamond \square \Phi_{i} \rightarrow \square \diamond \Psi_{i}\right)$ and $\square \diamond \Psi_{i}$, respectively. A CTL fairness assumption is a conjunction of strong, weak, and unconditional CTL fairness constraints.

Note that CTL fairness assumptions are not CTL path formulae, but they can be viewed as LTL formulae using CTL state formulae instead of atomic propositions. For instance, imposing the strong fairness constraint $\bigwedge_{1 \leqslant i \leqslant k}\left(\square \diamond \Phi_{i} \rightarrow \square \diamond \Psi_{i}\right)$ on paths means that a path must either have only finitely many states satisfying $\Phi_{i}$ or infinitely many states satisfying $\Psi_{i}$, for any $1 \leqslant i \leqslant k$ (or both).

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system without terminal states, $\pi$ be an infinite path fragment in TS, and fair a fixed CTL fairness assumption. $\pi \models$ fair denotes

[^30]that $\pi$ satisfies the formula fair, where $\models$ should be read as the LTL semantics. Consider, for example, strong fairness. For an infinite path $\pi=s_{0} s_{1} s_{2} \ldots$, we have
$$
\pi \models \bigwedge_{1 \leqslant i \leqslant k}\left(\square \diamond \Phi_{i} \rightarrow \square \diamond \Psi_{i}\right)
$$
if and only if for every $i \in\{1, \ldots, k\}$ either $s_{j} \models \Phi_{i}$ for finitely many indices $j$, or $s_{j} \models \Psi_{i}$ for infinitely many $j$ (or both). Here, the statement $s_{j} \models \Phi_{i}$ should be interpreted according to the CTL semantics, that is, the semantics without taking any fairness into account.

The semantics of CTL under fairness assumption fair is identical to the semantics given earlier (see Definition 6.4), except that the path quantifications range over all fair paths rather than over all paths. The fair paths starting in state $s$ are defined as

$$
\operatorname{FairPaths}(s)=\{\pi \in \operatorname{Paths}(s) \mid \pi \models \text { fair }\}
$$

Let FairPaths $(T S)$ denote the set of all fair paths in $T S$, i.e.:

$$
\operatorname{FairPaths}(T S)=\bigcup_{s_{0} \in I} \operatorname{FairPaths}\left(s_{0}\right)
$$

The fair interpretation of CTL is defined in terms of the satisfaction relation $\models_{\text {fair }}$. We have $s \models_{\text {fair }} \Phi$ if and only if $\Phi$ is valid in state $s$ under the fairness assumption fair.

## Definition 6.33. Satisfaction Relation for CTL with Fairness

Let $T S=(S, A c t, \rightarrow, I, A P, L)$ be a transition system without terminal states and $s \in$ $S$. The satisfaction relation $\models_{\text {fair }}$ for CTL fairness assumption fair is defined for state formulae by

$$
\begin{array}{lll}
s \models_{\text {fair }} a & \text { iff } & a \in L(s) \\
s \models_{\text {fair }} \neg \Phi & \text { iff } & \text { not } s \models_{\text {fair }} \Phi \\
s \models_{\text {fair }} \Phi \wedge \Psi & \text { iff } & \left(s \models_{\text {fair }} \Phi\right) \text { and }\left(s \models_{\text {fair }} \Psi\right) \\
s \models_{\text {fair }} \exists \varphi & \text { iff } & \pi \models_{\text {fair }} \varphi \text { for some } \pi \in \operatorname{FairPaths(s)~} \\
s \models_{\text {fair }} \forall \varphi & \text { iff } & \pi \models_{\text {fair }} \varphi \text { for all } \pi \in \operatorname{FairPaths(s)}
\end{array}
$$

Here, $a \in A P, \Phi, \Psi$ are CTL state formulae, and $\varphi$ a CTL path formula. For path $\pi$, the satisfaction relation $\models_{\text {fair }}$ for path formulae is defined as in Definition 6.4:

$$
\begin{array}{lll}
\pi \models_{\text {fair }} \bigcirc \Phi & \text { iff } & \pi[1] \models_{\text {fair }} \Phi \\
\pi \models_{\text {fair }} \Phi \cup \Psi & \text { iff } & \exists j \geqslant 0 .\left(\pi[j] \models_{\text {fair }} \Psi \wedge\left(\forall 0 \leqslant k<j . \pi[k] \models_{\text {fair }} \Phi\right)\right)
\end{array}
$$

where for path $\pi=s_{0} s_{1} s_{2} \ldots$ and integer $i \geqslant 0, \pi[i]$ denotes the ( $i+1$ )-th state of $\pi$, i.e., $\pi[i]=s_{i}$.

Whereas for LTL, fairness constraints can be specified as part of the formula to be checked, for CTL similar constraints are imposed on the underlying model of the system under consideration, i.e., the transition system.

## Definition 6.34. CTL Semantics with Fairness for Transition Systems

For CTL-state formula $\Phi$, and CTL fairness assumption fair, the satisfaction set $\operatorname{Sat}_{\text {fair }}(\Phi)$ is defined by

$$
\operatorname{Sat}_{\text {fair }}(\Phi)=\left\{s \in S \mid s \models_{\text {fair }} \Phi\right\}
$$

The transition system $T S$ satisfies CTL formula $\Phi$ under fairness assumption fair if and only if $\Phi$ holds in all initial states of $T S$ :

$$
T S \models_{\text {fair }} \Phi \text { if and only if } \forall s_{0} \in I . s_{0} \models_{\text {fair }} \Phi .
$$

This is equivalent to $I \subseteq \operatorname{Sat}_{\text {fair }}(\Phi)$.

## Example 6.35. CTL Fairness Assumption

Consider the transition system TS depicted in Figure 6.15 and suppose we are interested in establishing whether or not $T S \vDash \forall \square(a \Rightarrow \forall \diamond b)$. This formula is invalid since the path $s_{0} s_{1}\left(s_{2} s_{4}\right)^{\omega}$ never goes through a $b$-state. The reason that this property is not valid is as follows. At state $s_{2}$ there is a nondeterministic choice between moving either to state $s_{3}$ or to $s_{4}$. By continuously ignoring the possibility of going to $s_{3}$ we obtain a computation for which $\forall \square(a \Rightarrow \forall \diamond b)$ is invalid, hence:

$$
T S \not \vDash \forall \square(a \Rightarrow \forall \Delta b)
$$

Usually, though, the intuition is that if there is infinitely often a choice of moving to $s_{3}$, then $s_{3}$ should be visited in some fair way.

Let us impose the unconditional fairness assumption:

$$
\text { fair } \equiv \square \diamond a \wedge \square \diamond b
$$

We now check $\forall \square(a \Rightarrow \forall \diamond b)$ under fair, i.e., consider the verification problem $T S \models_{\text {fair }}$ $\forall \square(a \Rightarrow \forall \diamond b)$. Due to the fairness assumption, paths like $s_{0} s_{1}\left(s_{2} s_{4}\right)^{\omega}$ are excluded, since $s_{3}$ is never visited along this path. It can now easily be established that

$$
T S \models_{\text {fair }} \forall \square(a \Rightarrow \forall \Delta b)
$$



Figure 6.15: An example of a transition system.

## Example 6.36. Mutual Exclusion

Consider the semaphore-based solution to the two-process mutual exclusion problem. The transition system of this concurrent program is denoted $T S_{\text {sem }}$. The CTL formula

$$
\Phi=\left(\forall \square \forall \diamond \text { crit }_{1}\right) \wedge\left(\forall \square \forall \diamond \text { crit }_{2}\right)
$$

describes the liveness property that both processes infinitely often have access to the critical section. It follows that $T S_{\text {sem }} \not \vDash \Phi$. We first impose the weak fairness assumption

$$
\text { wfair }=\left(\diamond \square \text { noncrit }_{1} \rightarrow \square \diamond \text { wait }_{1}\right) \wedge\left(\diamond \square \text { noncrit }_{2} \rightarrow \square \diamond \text { wait }_{2}\right)
$$

and the strong fairness assumption

$$
\text { sfair }=\left(\square \diamond \text { wait }_{1} \rightarrow \square \diamond \text { crit }_{1}\right) \wedge\left(\square \diamond \text { wait }_{1} \rightarrow \square \diamond \text { crit }_{1}\right)
$$

It then follows that $T S_{\text {sem }} \models_{\text {fair }} \Phi$ where fair $=$ wfair $\wedge$ sfair.
As a second example, consider the arbiter-based solution to the two-process mutual exclusion problem, see Example 5.27 on page 259. The decision as to which process will acquire access to the critical section is determined by coin flipping by the arbiter. Consider the unconditional fairness assumption

$$
\text { ufair }=\square \diamond \text { head } \wedge \square \diamond \text { tail. }
$$

This fairness assumption can be considered to require a fair coin such that the events "heads" and "tails" occur infinitely often with probability 1. Apparently, it follows that

$$
T S_{1} \| \text { Arbiter } \| T S_{2} \not \vDash \Phi, \quad \text { and } \quad T S_{1} \| \text { Arbiter } \| T S_{2} \models_{\text {ufair }} \Phi .
$$

As explained before, fairness is treated in CTL by considering a fair satisfaction relation, denoted $\models_{\text {fair }}$ where fair is the fairness assumption considered. In the remainder of this section, algorithms will be provided in order to check whether

$$
T S \models_{\text {fair }} \Phi
$$

for CTL formula $\Phi$ and CTL-fairness assumption fair. As before, TS is supposed to be finite and have no terminal states and $\Phi$ is a CTL formula in ENF. Note that the assumption that $\Phi$ is in ENF does not impose any restriction as any CTL formula can be transformed into an equivalent (with respect to $\models_{\text {fair }}$ ) CTL formula in ENF. This can be established in the same way as Theorem 6.14 (page 332).

The basic idea is to exploit the CTL model-checking algorithms (without fairness) to compute $\operatorname{Sat}_{\text {fair }}(\Phi)=\left\{s \in S \mid s \models_{\text {fair }} \Phi\right\}$. Suppose fair is the strong CTL fairness constraint:

$$
\text { fair }=\bigwedge_{0<i \leqslant k}\left(\square \diamond \Phi_{i} \rightarrow \square \diamond \Psi_{i}\right)
$$

where $\Phi_{i}$ and $\Psi_{i}$ are CTL formulae over $A P$. Recall that $\Phi_{i}$ and $\Psi_{i}$ are interpreted according to the standard CTL semantics, i.e., without taking any fairness assumptions into account. By applying the CTL model-checking algorithm, first the sets $\operatorname{Sat}\left(\Phi_{i}\right)$ and $\operatorname{Sat}\left(\Psi_{i}\right)$ are determined. The formulae $\Phi_{i}$ and $\Psi_{i}$ can thus be replaced by (fresh) atomic propositions $a_{i}$ and $b_{i}$, say. It thus suffices to consider strong fairness assumptions of the form

$$
\text { fair }=\bigwedge_{0<i \leqslant k}\left(\square \diamond a_{i} \rightarrow \square \diamond b_{i}\right)
$$

Once the fairness assumption is simplified, the sets $\operatorname{Sat}_{\text {fair }}(\Psi)$ are determined for all subformulae $\Psi$ of $\Phi$ using the standard CTL model-checking algorithm (i.e., without fairness), together with an algorithm to compute $\operatorname{Sat}_{f a i r}(\exists \square a)$ for $a \in A P$. The outcome of the model-checking procedure is "yes" if $I \subseteq \operatorname{Sat}_{\text {fair }}(\Phi)$, and "no" otherwise.

The essential ideas are outlined in Algorithm 17 (page 364).
The subformulae of $\Phi$ are treated as in the CTL model-checking routine. Based on the syntax tree of $\Phi$, a bottom-up computation is initiated. It is essential that during the computation of $\operatorname{Sat}_{\text {fair }}(\Psi)$, the maximal genuine subformulae of $\Psi$ have been already processed and replaced by atomic propositions. For the propositional logic fragment,

```
Algorithm 17 CTL model checking with fairness (basic idea)
Input: finite transition system TS, CTL formula \(\Phi\) in ENF, and CTL fairness assumption fair over
    \(k\) CTL state formulae \(\Phi_{i}\) and \(\Psi_{i}\)
Output: \(T S \models_{\text {fair }} \Phi\)
    for all \(0<i \leqslant k\) do
        determine \(\operatorname{Sat}\left(\Phi_{i}\right)\) and \(\operatorname{Sat}\left(\Psi_{i}\right)\)
        if \(s \in \operatorname{Sat}\left(\Phi_{i}\right)\) then \(L(s):=L(s) \cup\left\{a_{i}\right\} ;\) fi
        if \(s \in \operatorname{Sat}\left(\Psi_{i}\right)\) then \(L(s):=L(s) \cup\left\{b_{i}\right\} ;\) fi
    od
    compute \(\operatorname{Sat}_{\text {fair }}(\exists \square\) true \()=\{s \in S \mid \operatorname{FairPaths}(s) \neq \varnothing\} ;\)
    forall \(s \in \operatorname{Sat}_{\text {fair }}(\exists \square\) true \()\) do \(L(s):=L(s) \cup\left\{a_{\text {fair }} ;\right\}\) od
    for all \(i \leqslant|\Phi|\) do
        for all \(\Psi \in \operatorname{Sub}(\Phi)\) with \(|\Psi|=i\) do
            switch \((\Psi)\) :
\begin{tabular}{lll} 
true & \(:\) & \(\operatorname{Sat}_{\text {fair }}(\Psi):=S ;\) \\
\(a\) & \(:\) & \(\operatorname{Sat}_{\text {fair }}(\Psi):=\{s \in S \mid a \in L(s)\} ;\) \\
\(a \wedge a^{\prime}\) & \(:\) & \(\operatorname{Sat}_{\text {fair }}(\Psi):=\left\{s \in S \mid a, a^{\prime} \in L(s)\right\} ;\) \\
\(\neg a\) & \(:\) & \(\operatorname{Sat}_{\text {fair }}(\Psi):=\{s \in S \mid a \notin L(s)\} ;\) \\
\(\exists \bigcirc a a\) & \(:\) & \(\operatorname{Sat}_{\text {fair }}(\Psi):=\operatorname{Sat}\left(\exists \bigcirc\left(a \wedge a_{\text {fair }}\right)\right) ;\) \\
\(\exists\left(a \cup a^{\prime}\right)\) & \(:\) & \(\operatorname{Sat}_{\text {fair }}(\Psi):=\operatorname{Sat}\left(\exists\left(a \cup\left(a^{\prime} \wedge a_{\text {fair }}\right)\right)\right) ;\) \\
\(\exists \square a\) & \(:\) & compute \(\operatorname{Sat} t_{\text {fair }}(\exists \square a)\)
\end{tabular}
            end switch
            replace all occurrences of \(\Psi\) in \(\Phi\) by the atomic proposition \(a_{\Psi}\);
            forall \(s \in \operatorname{Sat}_{f a i r}(\Psi)\) do \(L(s):=L(s) \cup\left\{a_{\Psi}\right\}\) od
            od
    od
    return \(I \subseteq \operatorname{Sat}_{\text {fair }}(\Phi)\)
```

the approach is straightforward:

$$
\begin{array}{ll}
\operatorname{Sat}_{\text {fair }}(\text { true }) & =S \\
\operatorname{Sat}_{\text {fair }}(a) & =\{s \in S \mid a \in L(s)\} \\
\operatorname{Sat}_{\text {fair }}(\neg a) & =S \backslash \operatorname{Sat}_{\text {fair }}(a) \\
\operatorname{Sat}_{\text {fair }}\left(a \wedge a^{\prime}\right) & =\operatorname{Sat}_{\text {fair }}(a) \cap \operatorname{Sat}_{\text {fair }}\left(a^{\prime}\right)
\end{array}
$$

For all nodes of the syntax tree that are labeled with either $\exists \bigcirc$ or $\exists \mathrm{U}$ (i.e, nodes that represent a subformula of the form $\Psi=\exists \bigcirc a$ or of the form $\left.\Psi=\exists\left(a \cup a^{\prime}\right)\right)$, the following observation is used. For any infinite path fragment $\pi$ in $T S, \pi$ is fair if and only if one (or all) suffix(es) of $\pi$ is (are) fair:

$$
\pi \models \text { fair } \quad \text { iff } \quad \pi[j . .] \models \text { fair for some } j \geqslant 0 \quad \text { iff } \quad \pi[j . .] \models \text { fair for all } j \geqslant 0 \text {. }
$$

The following two lemmas provide the ingredients for checking subformulae of the "type" $\exists \bigcirc$ and $\exists \mathrm{U}$.

## Lemma 6.37. Next Step for Fair Satisfaction Relation

$s \models_{\text {fair }} \exists \bigcirc a$ if and only if $\exists s^{\prime} \in \operatorname{Post}(s)$ with $s^{\prime} \models a$ and FairPaths $\left(s^{\prime}\right) \neq \varnothing$.

Proof: $\Rightarrow$ : Assume $s \models_{\text {fair }} \exists \bigcirc a$. Then there exists a fair path $\pi=s_{0} s_{1} s_{2} s_{3} \ldots \in$ $\operatorname{Paths}(s)$ with $s_{1} \vDash a$. Since $\pi$ is fair, the path $\pi[1 .]=.s_{1} s_{2} s_{3} \ldots$ is fair too. Thus, $s^{\prime}=s_{1} \in \operatorname{Post}(s)$ satisfies the indicated property.
$\Leftarrow$ : Assume $s^{\prime} \models a$ and $\operatorname{FairPaths}\left(s^{\prime}\right) \neq \varnothing$ for some $s^{\prime} \in \operatorname{Post}(s)$. Thus there exists a fair path

$$
\pi^{\prime}=s^{\prime} s_{1}^{\prime} s_{2}^{\prime} s_{3}^{\prime} \ldots
$$

starting in $s^{\prime}$. Therefore, $\pi=s s^{\prime} s_{1}^{\prime} s_{2}^{\prime} s_{3}^{\prime} \ldots$ is a fair path starting $s$ such that $\pi \models \bigcirc a$. Thus, $s \models_{\text {fair }} \exists \bigcirc a$.

Using analogous arguments we obtain:

## Lemma 6.38. Until for Fair Satisfaction Relation

$s \models_{\text {fair }} \exists\left(a_{1} \cup a_{2}\right)$ if and only if there exists a finite path fragment

$$
s_{0} s_{1} s_{2} s_{3} \ldots s_{n} \in \operatorname{Path}_{f i n}(s) \quad \text { with } n \geqslant 0
$$

such that $s_{i} \models a_{1}$ for $0 \leqslant i<n, s_{n} \models a_{2}$, and $\operatorname{FairPaths}\left(s_{n}\right) \neq \varnothing$.

The results of the previous two lemmas lead to the following approach. As a first step, the set

$$
\operatorname{Sat}_{\text {fair }}(\exists \square \operatorname{true})=\{s \in S \mid \text { FairPaths }(s) \neq \varnothing\}
$$

is computed. (The algorithmic way to do so is explained later in this chapter.) That is, all states are determined for which it is guaranteed that at least one fair path emanates. Once such a state is visited, it is thus guaranteed that a fair continuation is possible. The labels of the thus computed states are extended with the new atomic proposition $a_{\text {fair }}$, i.e.:

$$
a_{\text {fair }} \in L(s) \quad \text { if and only if } \quad s \in \operatorname{Sat}_{\text {fair }}(\exists \square \text { true })
$$

The sets $\operatorname{Sat}_{\text {fair }}(\exists \bigcirc a)$ and $\operatorname{Sat}_{\text {fair }}\left(\exists\left(a \cup a^{\prime}\right)\right)$ result from

$$
\begin{array}{ll}
\operatorname{Sat}_{\text {fair }}(\exists \bigcirc a) & =\operatorname{Sat}\left(\exists \bigcirc\left(a \wedge a_{\text {fair }}\right)\right), \\
\operatorname{Sat}_{\text {fair }}\left(\exists\left(a \cup a^{\prime}\right)\right) & =\operatorname{Sat}\left(\exists\left(a \cup\left(a^{\prime} \wedge a_{\text {fair }}\right)\right)\right) .
\end{array}
$$

As a result, these satisfaction sets can be computed using an ordinary CTL model checker. These considerations lead to Algorithm 17 on page 364 and provide the basis for the following result:

## Theorem 6.39. Model-Checking CTL with Fairness

The model-checking problem for CTL with fairness can be reduced to

- the model-checking problem for CTL (without fairness), and
- the problem of computing $\operatorname{Sat}_{\text {fair }}(\exists \square a)$ for the atomic proposition a.

Note that the set $\operatorname{Sat}_{\text {fair }}(\exists \square$ true $)$ corresponds to $\operatorname{Sat}_{\text {fair }}(\exists \square a)$ whenever every state is labeled with $a$. Thus, an algorithm to compute $\operatorname{Sat}_{f a i r}(\exists \square a)$ can also be used to compute $\operatorname{Sat}_{\text {fair }}(\exists \square$ true).

In the following, we explain how to compute the satisfaction set $\operatorname{Sat}_{\text {fair }}(\exists \square a)$ for $a \in A P$ in case fair is a strong CTL fairness assumption. Weak fairness assumptions can be treated in an analogous way. As we will see, unconditional fairness assumptions are a special case. Arbitrary fairness assumptions with unconditional, strong, and weak fairness constraints can be treated by using algorithms in which the corresponding techniques are appropriately combined.

Consider the strong CTL fairness assumption over the atomic propositions $a_{i}$ and $b_{i}(0<$ $i \leqslant k)$ :

$$
\text { sfair }=\bigwedge_{0<i \leqslant k}\left(\square \diamond a_{i} \rightarrow \square \diamond b_{i}\right) .
$$

The following lemma provides a graph-theoretical characterization of the fair satisfaction set $\operatorname{Sat}_{\text {sfair }}(\exists \square a)$ where $a$ is an atomic proposition.

## Lemma 6.40. Characterization of Sat $\operatorname{Sfair}(\exists \square a)$

$s \models_{\text {sfair }} \exists \square a$ if and only if there exists a finite path fragment $s_{0} s_{1} \ldots s_{n}$ and a cycle $s_{0}^{\prime} s_{1}^{\prime} \ldots s_{r}^{\prime}$ such that
(1) $s_{0}=s \quad$ and $\quad s_{n}=s_{0}^{\prime}=s_{r}^{\prime}$,
(2) $s_{i} \models a$, for any $0 \leqslant i \leqslant n$, and $s_{j}^{\prime} \models a$, for any $0<j \leqslant r$, and
(3) $\operatorname{Sat}\left(a_{i}\right) \cap\left\{s_{1}^{\prime}, \ldots, s_{r}^{\prime}\right\}=\varnothing$ or $\operatorname{Sat}\left(b_{i}\right) \cap\left\{s_{1}^{\prime}, \ldots, s_{r}^{\prime}\right\} \neq \varnothing$ for all $1 \leqslant i \leqslant k$.

Proof: $\Leftarrow$ : Assume there exists a finite path fragment $s_{0} s_{1} \ldots s_{n}$ and a cycle $s_{0}^{\prime} s_{1}^{\prime} \ldots s_{r}^{\prime}$ such that the conditions (1) through (3) hold for state $s$. Consider the infinite path fragment $\pi=s_{0} s_{1} \ldots s_{n} s_{1}^{\prime} \ldots s_{r}^{\prime} s_{1}^{\prime} \ldots s_{r}^{\prime} \ldots \in \operatorname{Paths}(s)$ which is obtained by traversing the cycle $s_{0}^{\prime} s_{1}^{\prime} \ldots s_{r}^{\prime}$ infinitely often. From constraints (1) and (3), it follows that $\pi$ is fair, i.e., $\pi \models$ sfair. From (2), it follows that $\pi \models \square a$. Thus, $s \models_{\text {sfair }} \exists \square a$.
$\Rightarrow$ : Assume $s \models_{\text {sfair }} \exists \square a$. Since $s \models_{\text {sfair }} \exists \square a$, there exists an infinite path fragment $\pi=s_{0} s_{1} s_{2} \ldots$ with $\pi \models \square a$ and $\pi \models$ sfair. Let $i \in\{1, \ldots, k\}$. Distinguish between two cases:

1. $\pi \models \square \diamond a_{i}$. Since $\pi \models$ sfair, there is a state $s^{\prime} \in \operatorname{Sat}\left(b_{i}\right)$ which is infinitely often visited in $\pi$. Let $n \in \mathbb{N}$ such that $s_{n}=s^{\prime}$ and $s^{\prime} \notin\left\{s_{0}, s_{1}, \ldots, s_{n-1}\right\}$. That is, $s_{n}$ is the first occurrence of $s^{\prime}$ in $\pi$. Further, let $r>n$ such that $s_{n}=s_{r}$. Then $s_{n} s_{n+1} \ldots s_{r-1} s_{r}=s_{n}$ is a cycle with

$$
s_{n} \in \operatorname{Sat}\left(b_{i}\right) \cap\left\{s_{n}, s_{n+1}, \ldots, s_{r-1}, s_{r}\right\} .
$$

2. $\pi \not \vDash \square \diamond a_{i}$. Then there exists an index $n$ such that $s_{n}, s_{n+1}, s_{n+2} \ldots \notin \operatorname{Sat}\left(a_{i}\right)$. Assume without loss of generality that $s_{n}$ occurs infinitely often in $\pi$. Since there are finitely many states, there exists an $r>n$ such that $s_{n}=s_{r}$. Thus, $s_{n} s_{n+1} \ldots s_{r-1} s_{r}=$ $s_{n}$ is a cycle with $\operatorname{Sat}\left(a_{i}\right) \cap\left\{s_{n}, s_{n+1}, \ldots, s_{r-1}, s_{r}\right\}=\varnothing$.

From both cases it follows that there exists a finite path fragment $s_{0} s_{1} \ldots s_{n}$ and a cycle $s_{0}^{\prime} \ldots s_{r}^{\prime}$ satisfying the constraints (1) through (3).

This characterization is used to compute $\operatorname{Sat}_{\text {sfair }}(\exists \square a)$ in the following way. Consider the directed graph $G[a]=\left(S, E_{a}\right)$ whose set of edges $E_{a}$ is defined as

$$
\left(s, s^{\prime}\right) \in E_{a} \quad \text { if and only if } \quad s^{\prime} \in \operatorname{Post}(s) \wedge s \models a \wedge s^{\prime} \models a
$$

Stated in words, $G[a]$ is obtained from the state graph $G_{T S}$ by eliminating all edges $\left(s, s^{\prime}\right)$ for which either $s \not \vDash a$ or $s^{\prime} \not \vDash a$. (Thus, $G[a]$ is the state graph of the transition system $T S[a]$.) Apparently, each infinite path in $G[a]$ (that starts in $s \in I$ ) corresponds to an infinite path in the transition system $T S$ satisfying $\square a$. Conversely, each infinite path fragment $\pi$ of $T S$ with $\pi \models \square a$ is a path in $G[a]$. In particular,

$$
s \models_{\text {sfair }} \exists \square a
$$

if and only if there exists a nontrivial, strongly connected set of nodes $D$ in $G[a]$ that is reachable from $s$, such that for all $1 \leqslant i \leqslant k$ :

$$
D \cap \operatorname{Sat}\left(a_{i}\right)=\varnothing \quad \text { or } \quad D \cap \operatorname{Sat}\left(b_{i}\right) \neq \varnothing .
$$

Let $T$ be the union of all nontrivial strongly connected components $C$ in the graph $G[a]$ such that sfair is realizable in $C$, i.e., there exists a nontrivial strongly connected subset $D$ of $C$ satisfying $D \cap \operatorname{Sat}\left(a_{i}\right)=\varnothing$ or $D \cap \operatorname{Sat}\left(b_{i}\right) \neq \varnothing$, for all $1 \leqslant i \leqslant k$. It then follows that

$$
\operatorname{Sat}_{\text {sfair }}(\exists \square a)=\left\{s \in S \mid \operatorname{Reach}_{G[a]}(s) \cap T \neq \varnothing\right\}
$$

Here, $\operatorname{Reach}_{G[a]}(s)$ is the set of states that is reachable from $s$ in $G[a]$. Note that as SCC $D$ is contained in SCC $C, D$ can be reached from any state in $C$; reachability of $C$ is thus of relevance.

The key part of the model-checking algorithm is now the computation of the set $T$. This amounts to investigating the SCCs $C$ of $G[a]$ and checking for which SCC sfair is realizable. As the computation of $T$ for the general case is slightly involved, we first consider two special (and simpler) cases: $a_{i}=$ true for all $i$, i.e., unconditional fairness, and the case for which $k=1$, i.e., a single strong fairness constraint.

In the sequel, each set $C$ of nodes in $G[a]$ is identified with the subgraph $\left(C, E_{C}\right)$ where $E_{C}$ results from the edge relation in $G[a]$ by removing all edges with starting or target vertex not in $C$.

Unconditional Fairness Let $a_{i}=$ true for all $1 \leqslant i \leqslant k$. In this case

$$
\text { sfair } \equiv \bigwedge_{1 \leqslant i \leqslant k} \square \diamond b_{i}
$$

Obviously, $s \neq_{\text {sfair }} \exists \square a$ if and only if there exists a nontrivial SCC $C$ in $G[a]$ that is reachable from $s$, such that $C$ contains at least one $b_{i}$-state for any $i$. In that case, $\square \diamond b_{i}$ is realizable in SCC $C$, for any $i$. Let $T$ be the set union of all nontrivial SCCs $C$ of $G[a]$ satisfying $C \cap \operatorname{Sat}\left(b_{i}\right) \neq \varnothing$ for all $1 \leqslant i \leqslant k$. It now follows that

$$
s \models_{\text {sfair }} \exists \square a \quad \text { if and only if } \quad \operatorname{Reach}_{G[a]}(s) \cap T \neq \varnothing
$$

The following example illustrates this.

## Example 6.41.

Consider the transition systems $T S$ (upper) and $T S^{\prime}$ (lower) in which it is implicitly assumed that all states are labeled with the atomic proposition $a$ :

(Due to this assumption, $T S[a]=T S$ and $T S^{\prime}[a]=T S^{\prime}$.) Consider the unconditional fairness assumption:

$$
\text { sfair }=\square \diamond b_{1} \wedge \square \diamond b_{2}
$$

The transition system $T S$ contains the reachable nontrivial SCC $C=\left\{s_{2}, s_{3}, s_{4}\right\}$ such that $C \cap \operatorname{Sat}\left(b_{1}\right) \neq \varnothing$ and $C \cap \operatorname{Sat}\left(b_{2}\right) \neq \varnothing$. We thus have $s_{0} \models_{\text {sfair }} \exists \square a$, and hence $T S \models_{\text {sfair }} \exists \square a$. On the other hand, $T S^{\prime}$ contains two non-trivial SCCs, but none that contains a $b_{1}$-state and a $b_{2}$-state. Therefore $s_{0}^{\prime} \not \models_{\text {sfair }} \exists \square a$ and hence $T S^{\prime} \not \vDash_{\text {sfair }} \exists \square a$.

Strong Fairness with a Single Fairness Constraint Assume $k=1$. i.e., sfair is assumed to be of the form

$$
\text { sfair }=\square \diamond a_{1} \rightarrow \square \diamond b_{1}
$$

We have that $s \models_{\text {sfair }} \exists \square a$ if and only if there exists a nontrivial strongly connected component $C$ of $G[a]$ with $C \subseteq \operatorname{Reach}_{G[a]}(s)$ such that at least one of the following two conditions (1) or (2) holds:
(1) $C \cap \operatorname{Sat}\left(b_{1}\right) \neq \varnothing$, or
(2) $D \cap \operatorname{Sat}\left(a_{1}\right)=\varnothing$ for some nontrivial SCC $D$ of $C$.

```
Algorithm 18 Computation of \(\operatorname{Sat}_{\text {sfair }}(\exists \square a)\)
Input: A finite \(T S\) without terminal states, \(a \in A P\) and fair \(=\bigwedge_{0<i \leqslant k}\) sfair \(_{i}\) with sfair \({ }_{i}=\square \diamond a_{i} \rightarrow\)
    \(\diamond b_{i}\)
Output: \(\left\{s \in S \mid s \models_{\text {fair }} \exists \square a\right\}\)
    compute the SCCs of the state graph \(G[a]\) of \(T S[a]\);
    \(T:=\varnothing\);
    for all nontrivial SCCs \(C\) in \(G[a]\) do
                            (* check whether the fairness assumption sfair can be realized in \(C^{*}\) )
        if CheckFair \(\left(C, k\right.\), sfair \(_{1}, \ldots\), sfair \(\left._{k}\right)\) then
        \(T:=T \cup C ;\)
        fi
od
return \(\left\{s \in S \mid \operatorname{Reach}_{G[a]}(s) \cap T \neq \varnothing\right\} \quad\) (* e.g., backwards reachability *)
```

Intuitively, in case (1) $C$ stands for a cyclic set of states that realizes $\square \diamond b_{1}$, while $D$ in (2) represents a cyclic set of states for which $\neg a_{1}$ continuously holds. The SCCs $D$ of $C$ that do not contain any $a_{1}$-state can easily be computed by determining the nontrivial SCCs in the graph that is obtained from $C$ by eliminating all $a_{1}$-states. (Stated differently, $C$ realizes the unconditional fairness constraint $\square \diamond b_{1}$, while $D$ realizes the unconditional fairness constraint $\square \diamond$ true in the transition system induced by $C$ after eliminating all $a_{1^{-}}$ states. This characterization is helpful to understand the general algorithm below.) Every path that has a suffix consisting of the infinite repetition of a cycle that runs through all states of $C$ (case (1)) or that eventually reaches $D$ and stays there forever satisfies the fairness assumption sfair. Accordingly, we may define $T$ as the union of all nontrivial SCCs $C$ of $G[a]$ satisfying the above constraints (1) and (2). Then, $s \models_{s f a i r ~} \exists \square a$ if and only if $\operatorname{Reach}_{G[a]}(s) \cap T \neq \varnothing$.

Strong Fairness with $k>1$ Fairness Constraints In this case, sfair is defined for $k>1$ by

$$
\text { sfair }=\bigwedge_{1 \leqslant i \leqslant k} \text { sfair }_{i} \quad \text { with } \quad \text { sfair }_{i}=\square \diamond a_{i} \rightarrow \square \diamond b_{i} .
$$

As for the other cases, the initial step is to determine $G[a]$ and its set of SCCs. We have to compute all nontrivial SCCs $C$ such that for some cyclic subset $D$ of $C$ all fairness constraints $\square \diamond a_{i} \rightarrow \square \diamond b_{i}$ are realizable in $D$, i.e., for $1 \leqslant i \leqslant k$ :

$$
D \cap \operatorname{Sat}\left(a_{i}\right)=\varnothing \quad \text { or } \quad D \cap \operatorname{Sat}\left(b_{i}\right) \neq \varnothing .
$$

Algorithm 18 on page 370 provides the schema for the computation of $\operatorname{Sat}_{\text {sfair }}(\exists \square a)$. It uses the recursive procedure CheckFair (see Algorithm 19, page 372) to check the real-
izability of sfair in SCC $C$ of $G[a]$. The inputs of CheckFair are a cyclic strongly connected subgraph $D$ of $G[a]$, index $j$, and $j$ strong fairness assumptions sfair $i_{1}, \ldots$, sfair $_{i_{j}}$. CheckFair ( $D, j$, sfair $_{i_{1}}, \ldots$, sfair $_{i_{j}}$ ) returns true if $\bigwedge_{1 \leqslant \ell \leqslant j} s f a i r_{i_{\ell}}$ is realizable in $D$. Thus, the invocation CheckFair $\left(C, k\right.$, sfair $_{1}, \ldots$, sfair $\left._{k}\right)$ yields an affirmative answer whether $C \subseteq T$ by returning true if and only if the fairness assumption sfair $=\bigwedge_{1 \leqslant i \leqslant k} s f a i r_{i}$ is realizable in $C$.

The idea of Algorithm 19, CheckFair $\left(C, k\right.$, sfair $_{1}, \ldots$, sfair $\left._{k}\right)$, is as follows. First, it is checked whether $C \cap \operatorname{Sat}\left(b_{i}\right) \neq \varnothing$ for each fairness constraint sfair ${ }_{i}=\square \diamond a_{i} \rightarrow \square \diamond b_{i}$.

- If yes, then sfair $=\bigwedge_{0<i \leqslant k} s f a i r_{i}$ is realizable in $C$.
- Otherwise, there exists an index $j \in\{1, \ldots, k\}$ such that $C \cap \operatorname{Sat}\left(b_{j}\right)=\varnothing$. The goal is then to realize the condition

$$
\bigwedge_{\substack{0<i \leqslant k \\ i \neq j}} \operatorname{sfair}_{i} \wedge \square \neg a_{j}
$$

in $C$. For this, we investigate the subgraph $C\left[\neg a_{j}\right]$ of $C$ that arises by removing all states where $a_{j}$ holds and their incoming and outgoing edges. The goal is to check the existence of a cyclic subgraph of $C\left[\neg a_{j}\right]$ such that the remaining $k-1$ fairness constraints sfair $_{1}, \ldots$, sfair $_{j-1}$, sfair $_{j+1}, \ldots$, sfair $_{k}$ are realizable in this subgraph. This is done by analyzing the nontrivial SCCs $D$ of $C\left[\neg a_{j}\right]$.

- If there is no nontrivial SCC $D$ of $C\left[\neg a_{j}\right]$, then sfair is not realizable in $C$.
- Otherwise, invoke CheckFair $(D, k-1, \ldots)$ for each of these nontrivial SCCs $D$ of $C\left[\neg a_{j}\right]$ to check whether the remaining $k-1$ fairness constraints are realizable in $D$.

The main steps of this procedure are summarized in Algorithm 19 on page 372. Note that in each recursive call one of the fairness constraints is removed, and thus the recursion depth is at most $k$, in which case $k=0$ and the condition " $\forall i \in\{1, \ldots, k\} . C \cap \operatorname{Sat}\left(b_{j}\right) \neq \varnothing$ " of the first ifstatement is obviously fulfilled.

The cost function for $\operatorname{CheckFair}\left(C, k\right.$, sfair $_{1}, \ldots$, sfair $\left._{k}\right)$, is given by the recurrence equation:

$$
T(n, k)=\mathcal{O}(n)+\max \left\{\sum_{1 \leqslant \ell \leqslant r} T\left(n_{\ell}, k-1\right) \mid n_{1}, \ldots, n_{r} \geqslant 1, n_{1}+\ldots+n_{r} \leqslant n\right\}
$$

where $n$ is the size (number of vertices and edges) of the subgraph $C$ of $G[a]$. The values $n_{1}, \ldots, n_{r}$ denote the sizes of the nontrivial strongly connected components $D_{1}, \ldots, D_{r}$ of $C\left[\neg a_{j}\right]$. It is easy to see that the solution of this recurrence is $\mathcal{O}(n \cdot k)$. This yields:

```
Algorithm 19 Recursive algorithm \(\operatorname{CheckFair}\left(C, k\right.\), sfair \(_{1}, \ldots\), sfair \(\left._{k}\right)\)
Input: nontrivial SCC \(C\) in \(G[a]\), and strong fairness constraints sfair \({ }_{i}=\square \diamond a_{i} \rightarrow \square \diamond b_{i}, i=\)
    \(1, \ldots, k\).
Output: true if \(\bigwedge_{1 \leqslant i \leqslant k}\) sfair \(_{i}\) is realizable in \(C\). Otherwise false.
```

```
if \(\forall i \in\{1, \ldots, k\} . C \cap \operatorname{Sat}\left(b_{j}\right) \neq \varnothing\) then
```

if $\forall i \in\{1, \ldots, k\} . C \cap \operatorname{Sat}\left(b_{j}\right) \neq \varnothing$ then
return true
return true
else
else
choose an index $j \in\{1, \ldots, k\}$ with $C \cap \operatorname{Sat}\left(b_{j}\right)=\varnothing$;
choose an index $j \in\{1, \ldots, k\}$ with $C \cap \operatorname{Sat}\left(b_{j}\right)=\varnothing$;
if $C\left[\neg a_{j}\right]$ is acyclic (or empty) then
if $C\left[\neg a_{j}\right]$ is acyclic (or empty) then
return false
return false
else
else
compute the nontrivial SCCs of $C\left[\neg a_{j}\right]$;
compute the nontrivial SCCs of $C\left[\neg a_{j}\right]$;
for all nontrivial SCC $D$ of $C\left[\neg a_{j}\right]$ do
for all nontrivial SCC $D$ of $C\left[\neg a_{j}\right]$ do
if CheckFair $\left(D, k-1\right.$, sfair $_{1}, \ldots$, sfair $_{j-1}$, sfair $_{j+1}, \ldots$, sfair $\left._{k}\right)$ then
if CheckFair $\left(D, k-1\right.$, sfair $_{1}, \ldots$, sfair $_{j-1}$, sfair $_{j+1}, \ldots$, sfair $\left._{k}\right)$ then
return true
return true
fi
fi
od
od
fi
fi
fi
return false

```

\section*{Theorem 6.42. Time Complexity of Verifying \(\exists \square a\) under Fairness}

For transition system TS with \(N\) states and \(K\) transitions, and CTL strong fairness assumption fair with \(k\) conjuncts, the set \(\operatorname{Sat}_{\text {fair }}(\exists \square a)\) can be computed in \(\mathcal{O}((N+K) \cdot k)\).

The time complexity is thus linear in the size of the transition system and the number of imposed fairness constraints. The set \(T\) resulting from all SCCs \(C\) of \(G[a]\), for which CheckFair \((C, 1)\) returns the value true, can be computed (with appropriate implementation) in time \(\mathcal{O}(\operatorname{size}(G[a]) \cdot k)\). A reachability analysis (by, e.g., depth-first search) results in the set
\[
\operatorname{Sat}_{\text {fair }}(\exists \square a)=\left\{s \in S \mid \operatorname{Reach}_{G[a]}(s) \cap T \neq \varnothing\right\}
\]

Gathering these results yields that CTL model checking under strong fairness constraints and for CTL formulae in ENF can be done in time linear in the size of the transition system, the length of the formula, and the number of (strong) fairness constraints. This, in fact, also holds for arbitrary CTL formulae (with universal quantification which can be treated by techniques similar as the ones we presented for existential quantification) and weak fairness constraints, or any mixture of unconditional, strong, and weak fairness constraints. We thus obtain:

\section*{Theorem 6.43. Time Complexity of CTL Model Checking with Fairness}

For transition system \(T S\) with \(N\) states and \(K\) transitions, \(C T L\) formula \(\Phi\), and \(C T L\) fairness assumption fair with \(k\) conjuncts, the CTL model-checking problem \(T S \models_{\text {fair }} \Phi\) can be determined in time \(\mathcal{O}((N+K) \cdot|\Phi| \cdot k)\).

\subsection*{6.6 Counterexamples and Witnesses}

A major strength of model checking is the possibility of generating a counterexample in case a formula is refuted. Let us first explain what is meant by a counterexample. In the case of LTL, a counterexample for \(T S \nLeftarrow \varphi\) is a sufficiently long prefix of a path \(\pi\) that indicates why \(\pi\) refutes \(\varphi\). For instance, a counterexample for the LTL formula \(\diamond a\) is a finite prefix of just \(\neg a\)-states that ends with a single cycle traversal. Such counterexample suggests that there is a \(\square \neg a\)-path. Similarly, a counterexample for \(\bigcirc a\) consists of a path \(\pi\) for which \(\pi[1]\) violates \(a\).

For CTL the situation is somewhat more involved due to the existential path quantification. For CTL formulae of the form \(\forall \varphi\) a sufficiently long prefix of \(\pi\) with \(\pi \not \vDash \varphi\) provides-as in LTL-sufficient information about the source of the refutation. In the
case of a path formula of the form \(\exists \varphi\), it is unclear what a counterexample will be: if \(T S \not \vDash \exists \varphi\), then all paths violate \(\varphi\) and no path satisfies \(\varphi\). However, if one checks the formula \(\exists \varphi\) for a transition system \(T S\), then it is quite natural that for the answer "yes, \(T S \vDash \exists \varphi\) " one aims at an initial path where \(\varphi\) holds, while the answer "no" might be sufficient. \({ }^{6}\) Therefore, CTL model checking supports the system diagnosis by providing counterxamples or witnesses. Intuitively, counterexamples indicate the refutation of universally quantified path formulae, while witnesses indicate the satisfaction of existentially quantified path formulae. From a path-based view, the concepts are counterexamples and witnesses can be explained as follows:
- a sufficiently long prefix of a path \(\pi\) with \(\pi \not \vDash \varphi\) is a counterexample for the CTL path formula \(\forall \varphi\), and
- a sufficiently long prefix of a path \(\pi\) with \(\pi \models \varphi\) is a witness of the CTL path formula \(\exists \varphi\).

To exemplify the idea of generating a witness, consider the following well-known combinatorial problem.

\section*{Example 6.44. The Wolf-Goat-Cabbage Problem}

Consider the problem of bringing a ferryman (f), a goat (g), a cabbage (c) and a wolf (w) from one side of a river to the other side. The only available means to go from one riverside to another is a small boat that is capable of carrying at most two occupants. In order for the boat to be steered, one of the occupants needs to be the ferryman. Of course, the boat does not need to be fully occupied, and the ferryman can cross the river alone. For obvious reasons, neither the goat and the cabbage nor the goat and the wolf should be left unobserved by the ferryman. The question is whether there exists a series of boat movements such that all three items can be carried by the ferryman to the other side of the river.

This problem can be represented as a CTL model-checking problem in a rather natural way. The behavior of the goat, wolf, and cabbage is provided by a simple two-state transition system depicted in Figure 6.16. The state identifiers indicate the position of each of the items: 0 stands for the current (i.e., starting) riverside, and 1 for the other side

\footnotetext{
\({ }^{6}\) Since the standard CTL model-checking procedure calculates the set of states where the given (state) formula \(\Phi\) holds, also some information extracted from \(\operatorname{Sat}(\Phi)\) could be returned to the user. For instance, if \(T S \not \vDash \exists \varphi\), then the model checker might return the set of initial states \(s_{0}\) where \(s_{0} \not \vDash \exists \varphi\). This information could be understood as a counterexample and used for debugging purposes. This issue will not be addressed here. Instead we will discuss the concept of path-based counterexamples (also often called "error traces") and their duals, i.e., computations that provide a proof for the existence of computations with certain properties.
}


Figure 6.16: Transition systems for the wolf, goat, cabbage, and ferryman.
(i.e., the goal). The synchronization actions \(\alpha\) and \(\beta\) are used to describe the boat trips that are taking place together with the ferryman. The ferryman behaves very similarly, but has in addition the possibility to cross the river alone. This corresponds to the two \(\tau\)-labeled transitions. The resulting transition system representing the entire "system"
\[
T S=(\text { wolf }| | \mid \text { goat }| | \mid \text { cabbage }) \| \text { ferryman },
\]
has \(2^{4}=16\) states. The resulting transition system is depicted in Figure 6.17. Note that the transitions are all bidirectional as each boat movement can be reversed. The existence of a sequence of boat movements to bring the two animals and the cabbage to the other riverbank can be expressed as the CTL state formula \(\exists \varphi\) where
\[
\varphi=\left(\bigwedge_{i=0,1}\left(w_{i} \wedge g_{i} \rightarrow f_{i}\right) \wedge\left(c_{i} \wedge g_{i} \rightarrow f_{i}\right)\right) \cup\left(c_{1} \wedge f_{1} \wedge g_{1} \wedge w_{1}\right)
\]

Here, the left part of the until formula forbids the scenarios in which the wolf and the goat, or the cabbage and the goat are left unobserved. A witness of the CTL path formula \(\varphi\) is an initial finite path fragment which leads from
\[
\text { initial state }\left\langle c_{0}, f_{0}, g_{0}, w_{0}\right\rangle \text { to target state }\left\langle c_{1}, f_{1}, g_{1}, w_{1}\right\rangle
\]
and which does not pass through any of the (six) states in which the wolf and the goat or the goat and the cabbage are left alone on one riverbank. That is, e.g., the states \(\left\langle c_{0}, f_{0}, g_{1}, w_{1}\right\rangle\) and \(\left\langle c_{1}, f_{0}, g_{1}, w_{0}\right\rangle\) should be avoided. An example witness for \(\varphi\) is:
\[
\begin{array}{ll}
\left\langle c_{0}, f_{0}, g_{0}, w_{0}\right\rangle & \text { goat to riverbank } 1 \\
\left\langle c_{0}, f_{1}, g_{1}, w_{0}\right\rangle & \text { ferryman comes back to riverbank } 0 \\
\left\langle c_{0}, f_{0}, g_{1}, w_{0}\right\rangle & \text { cabbage to riverbank } 1 \\
\left\langle c_{1}, f_{1}, g_{1}, w_{0}\right\rangle & \text { goat back to riverbank } 0 \\
\left\langle c_{1}, f_{0}, g_{0}, w_{0}\right\rangle & \text { wolf to riverbank } 1 \\
\left\langle c_{1}, f_{1}, g_{0}, w_{1}\right\rangle & \text { ferryman comes back to riverbank } 0 \\
\left\langle c_{1}, f_{0}, g_{0}, w_{1}\right\rangle & \text { goat to riverbank } 1 \\
\left\langle c_{1}, f_{1}, g_{1}, w_{1}\right\rangle &
\end{array}
\]


Figure 6.17: Transition system of the wolf-goat-cabbage problem.

\subsection*{6.6.1 Counterexamples in CTL}

Now we explain how to generate counterexamples or witnesses for CTL (path) formulae. We consider here path formulae of the form \(\bigcirc \Phi, \Phi \cup \Psi\), and \(\square \Phi\). (Techniques for other operators, such as \(W\) or \(R\), can be derived. Alternatively, corresponding considerations can be made for them.)

In the sequel, let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a finite transition system without terminal states.

The Next Operator A counterexample of \(\varphi=\bigcirc \Phi\) is a pair of states \(\left(s, s^{\prime}\right)\) with \(s \in I\) and \(s^{\prime} \in \operatorname{Post}(s)\) such that \(s^{\prime} \not \models \Phi\). A witness of \(\varphi=\bigcirc \Phi\) is a pair of states \(\left(s, s^{\prime}\right)\) with \(s \in I\) and \(s^{\prime} \in \operatorname{Post}(s)\) with \(s^{\prime} \models \Phi\). Thus, counterexamples and witnesses for the next-step operator result from inspecting the immediate successors of the initial states of TS.

The Until Operator A witness of \(\varphi=\Phi U \Psi\) is an initial path fragment \(s_{0} s_{1} \ldots s_{n}\) for which
\[
s_{n} \models \Psi \quad \text { and } \quad s_{i} \models \Phi \text { for } 0 \leqslant i<n .
\]

Witnesses can be determined by a backward search starting in the set of \(\Psi\)-states.
A counterexample is an initial path fragment that indicates a path \(\pi\) for which either:
\[
\pi \models \square(\Phi \wedge \neg \Psi) \quad \text { or } \quad \pi \models(\Phi \wedge \neg \Psi) \cup(\neg \Phi \wedge \neg \Psi)
\]

For the first case, a counterexample is an initial path fragment of the form


For the second case, an initial path fragment of the form
\[
\underbrace{s_{0} s_{1} \ldots s_{n-1}}_{\text {satisfy } \Phi \wedge \neg \Psi} s_{n} \quad \text { with } s_{n} \models \neg \Phi \wedge \neg \Psi
\]
does suffice as counterexample. Counterexamples can be determined by an analysis of the digraph \(G=(S, E)\) where
\[
E=\left\{\left(s, s^{\prime}\right) \in S \times S \mid s^{\prime} \in \operatorname{Post}(s) \wedge s \models \Phi \wedge \neg \Psi\right\}
\]

Then the SCCs of \(G\) are determined. Each path in \(G\) that starts in an initial state \(s_{0} \in S\) and leads to a nontrivial SCC \(C\) in \(G\) provides a counterexample of the form
\[
s_{0} s_{1} \ldots s_{n} \underbrace{s_{1}^{\prime} \ldots s_{r}^{\prime}}_{\in C} \quad \text { with } \quad s_{n}=s_{r}^{\prime}
\]

Each path in \(G\) that leads from an initial state \(s_{0}\) to a trivial terminal SCC
\[
C=\left\{s^{\prime}\right\} \quad \text { with } \quad s^{\prime} \not \models \Psi
\]
provides a counterexample of the form \(s_{0} s_{1} \ldots s_{n}\) with \(s_{n} \models \neg \Phi \wedge \neg \Psi\).


Figure 6.18: Transition system of semaphore-based mutual exclusion algorithm.

The Always Operator A counterexample for the formula \(\varphi=\square \Phi\) is an initial path fragment \(s_{0} s_{1} \ldots s_{n}\) such that \(s_{i} \models \Phi\) for \(0 \leqslant i<n\) and \(s_{n} \not \vDash \Phi\). Counterexamples may be determined by a backward search starting in \(\neg \Phi\)-states.

A witness of \(\varphi=\)\(\Phi\) consists of an initial path fragment of the form
\[
\underbrace{s_{0} s_{1} \ldots s_{n} s_{1}^{\prime} \ldots s_{r}^{\prime}}_{\text {satisfy } \Phi} \quad \text { with } \quad s_{n}=s_{r}^{\prime}
\]

Witnesses can be determined by a simple cycle search in the digraph \(G=(S, E)\) where the set of edges \(E\) is obtained from the transitions emanating from \(\Phi\)-states, i.e., \(E=\) \(\left\{\left(s, s^{\prime}\right) \mid s^{\prime} \in \operatorname{Post}(s) \wedge s \models \Phi\right\}\).

\section*{Example 6.45. Counterexamples and Semaphore-Based Mutual Exclusion}

Recall the two-process mutual exclusion algorithm that exploits a binary semaphore \(y\) to resolve contention (see Eaxmple 3.9 on page 98). For convenience, the transition system \(T S_{S e m}\) of this algorithm is depicted in Figure 6.18. Consider the CTL formula over \(A P=\) \(\left\{c_{1}, c_{2}, n_{1}, n_{2}, w_{1}, w_{2}\right\}:\)
\[
\forall(\underbrace{\left(\left(n_{1} \wedge n_{2}\right) \vee w_{2}\right)}_{\Phi} \cup \underbrace{c_{2}}_{\Psi}) .
\]

It expresses that process \(P_{2}\) acquires access to the critical section once it starts waiting to enter it.

Note that the state labeling of \(T S_{S e m}\) can be directly obtained from the information in


Figure 6.19: Graph to detect counterexamples for \(\forall\left(\left(\left(n_{1} \wedge n_{2}\right) \vee w_{2}\right) \cup c_{2}\right)\).
the states. Evidently, the CTL formula is refuted by \(T S_{S e m}\). Counterexamples can be established as follows. First, the graph \(G\) is determined by only allowing edges that emanate from states satisfying \(\Phi \wedge \neg \Psi\). This entails that all edges emanating from \(s\) with \(s \vDash \neg\left(\left(n_{1} \wedge n_{2}\right) \vee w_{2}\right) \vee c_{2}\) are eliminated. This yields the graph depicted in Figure 6.19. The graph \(G\) contains a single nontrivial SCC \(C\) that is reachable from the initial state in \(T S_{\text {Sem }}\). The initial path fragment:
\[
\left\langle n_{1}, n_{2}, y=1\right\rangle \underbrace{\left\langle n_{1}, w_{2}, y=1\right\rangle\left\langle w_{1}, w_{2}, y=1\right\rangle\left\langle c_{1}, w_{2}, y=0\right\rangle}_{\in C}
\]
is a counterexample as it shows that there is a path \(\pi\) in \(T S_{\text {Sem }}\) satisfying \(\square\left(\left(\left(n_{1} \wedge n_{2}\right) \vee\right.\right.\) \(\left.w_{2}\right) \wedge \neg c_{2}\) ), i.e., a path for which \(c_{2}\) is never established.

Alternatively, any path from the initial state to the trivial terminal SCC satisfying \(\neg c_{2}\) is a counterexample as well. This only holds for the terminal SCC \(\left\{\left\langle w_{1}, n_{2}, y=1\right\rangle\right\}\).

\section*{Theorem 6.46. Time Complexity of Counterexample Generation}

Let TS be a transition system TS with \(N\) states and \(K\) transitions and \(\varphi\) a CTL- path formula. If TS \(\vDash \forall \varphi\), then a counterexample for \(\varphi\) in TS can be determined in time \(\mathcal{O}(N+K)\). The same holds for a witness for \(\varphi\), provided that \(T S \models \exists \varphi\).

\subsection*{6.6.2 Counterexamples and Witnesses in CTL with Fairness}

In the case CTL fairness assumptions are imposed, witnesses and counterexamples can be provided in a similar way as for CTL. Suppose fair is the fairness assumption of interest.

The Next Operator A witness for \(\pi \models_{\text {fair }} \bigcirc a\) originates from a witness for \(\pi \models\) \(\bigcirc\left(a \wedge a_{f a i r}\right)\). Note that the latter is obtained as for CTL, as no fairness is involved. A counterexample to \(\bigcirc a\) is a prefix of a fair path \(\pi=s_{0} s_{1} s_{2} \ldots\) in \(T S\) with \(\pi \not \vDash \bigcirc a\). As \(\pi\) is fair and \(\pi \not \vDash \bigcirc a\), it follows that \(s_{1} \vDash a_{\text {fair }}\) and \(s_{1} \not \vDash a\). So, \(s_{1} \not \vDash a_{\text {fair }} \rightarrow a\). A counterexample to \(\bigcirc a\) with respect to \(\models_{\text {fair }}\) thus results from a counterexample to the formula \(\bigcirc\left(a_{f a i r} \rightarrow a\right)\) for CTL without fairness.

The Until Operator A witness of \(a \cup a^{\prime}\) with respect to the fair semantics is a witness of \(a \cup\left(a^{\prime} \wedge a_{\text {fair }}\right)\) with respect to the standard CTL semantics. A counterexample for \(a \cup a^{\prime}\) with respect to the fair semantics is either a witness of \(\left(a \wedge a^{\prime}\right) \cup\left(\neg a \wedge \neg a^{\prime} \wedge a_{\text {fair }}\right)\) under the common semantics \(\models\) or a witness of \(\square\left(a \wedge \neg a^{\prime}\right)\) with respect to the fair semantics (explained below).

The Always Operator For a formula of the form \(\square a\), a counterexample with respect to the fair semantics is an initial path fragment \(s_{0} s_{1} \ldots s_{n}\) such that:
\[
s_{n} \models \neg a \wedge a_{\text {fair }} \quad \text { and } \quad s_{i} \models a \text { for } 0 \leqslant i<n .
\]

Consider the strong fairness assumption of the form:
\[
\text { sfair }=\bigwedge_{0<i \leqslant k}\left(\square \diamond a_{i} \rightarrow \square \diamond b_{i}\right) \text {. }
\]

A witness of \(\square a\) under sfair is an initial path fragment
\[
\underbrace{s_{0} s_{1} \ldots s_{n} s_{1}^{\prime} s_{2}^{\prime} \ldots s_{r}^{\prime}}_{\models a} \quad \text { with } s_{n}=s_{r}^{\prime}
\]
such that for all \(0<i \leqslant k\) it holds that
\[
\operatorname{Sat}\left(a_{i}\right) \cap\left\{s_{1}^{\prime}, \ldots, s_{r}^{\prime}\right\}=\varnothing \text { or } \operatorname{Sat}\left(b_{i}\right) \cap\left\{s_{1}^{\prime}, \ldots, s_{r}^{\prime}\right\} \neq \varnothing
\]

A witness can be computed by means of an analysis of the SCCs of a digraph that originates from the state graph of \(T S\) after some slight modifications. The costs are (multiplicatively) linear in the number of fairness conditions and in the size of the state graph.

\section*{Theorem 6.47. Time Complexity of Fair Counterexample Generation}

For transition system TS with \(N\) states and \(K\) transitions, CTL path formula \(\varphi\) and CTL fairness assumption fair with \(k\) conjuncts such that \(T S \not \models_{\text {fair }} \forall \varphi\), a counterexample for \(\varphi\) in TS can be determined in time \(\mathcal{O}((N+K) \cdot k)\). The same holds for a witness for \(\varphi\) if \(T S \models \exists \varphi\).

\section*{Example 6.48.}

Consider the transition system depicted in Figure 6.11 (page 350) and assume the formula of interest is:
\[
\exists \square(a \vee(b=c)) \quad \text { under fairness constraint } \quad \text { sfair }=\square \diamond(q \wedge r) \rightarrow \square \diamond \neg(q \vee r) .
\]

The strong CTL fairness constraint asserts that in case either state \(s_{0}\) or \(s_{2}\) is visited infinitely often, then also state \(s_{3}\) or \(s_{7}\) needs to be visited infinitely often. The path \(s_{1} s_{3} s_{0} s_{2} s_{0}\) is a witness of \(\exists \square(a \vee(b=c))\) in the absence of any fairness constraint. It is, however, no witness under sfair as it visits \(s_{0}\) (and \(s_{2}\) ) infinitely often, but not \(s_{3}\) or \(s_{7}\). On the other hand, the path \(s_{1} s_{3} s_{0} s_{2} s_{1}\) is a witness under sfair.

\subsection*{6.7 Symbolic CTL Model Checking}

The CTL model-checking procedure described so far relies on the assumption that the transition system has an explicit representation by the predecessor and successor lists per state. Such an enumerative representation is not adequate for very large transition systems. To attack the state explosion problem, the CTL model-checking procedure can be reformulated in a symbolic way where sets of states and sets of transitions are represented rather than single states and transitions. This set-based approach is very natural for CTL, since its semantics and model-checking algorithm rely on satisfaction sets for subformulae. There are several possibilities to realize the CTL model checking algorithm in a purely set-based stetting. The most prominent ones rely on a binary encoding of the states, which permits identifying subsets of the state space and the transition relation with switching functions. To obtain compact representations of switching functions, special data structures have been developed, such as ordered binary decision diagrams. Other forms for the representation of switching functions, such as conjunctive normal forms, could be used as well. They are widely used in the context of so-called SAT-based model checking where the model-checking problem is reduced to the satisfiability problem for propositional formulae (SAT). The SAT-based techniques will not be described in this book. Instead we will explain the main ideas of the approach with binary decision diagrams.

We first explain the general ideas behind symbolic approaches which operate on sets of states rather than single states and rely on a representation of transition systems by switching functions. In the sequel, let \(T S=(S, \rightarrow, I, A P, L)\) be a "large", but finite transition system. The set of actions is irrelevant here and has been skipped. That is, the transition relation \(\rightarrow\) is a subset of \(S \times S\). Let \(n \geqslant\lceil\log |S|\rceil\). (Since we suppose a large transition system, we can safely assume that \(|S| \geqslant 2\).) We choose an arbitrary (injective) encoding enc : S \(\rightarrow\{0,1\}^{n}\) of the states by bit vectors of length \(n\). Although enc might not be surjective, it is no restriction to suppose that enc \((S)=\{0,1\}^{n}\), since all elements \(\left(b_{1}, \ldots, b_{n}\right) \in\{0,1\}^{n} \backslash \operatorname{enc}(S)\) can be treated as the encoding of pseudo states that cannot be reached from any proper state \(s \in S\). The transitions of these pseudo states are arbitrary. The idea is now to identify the states \(s \in S=e n c^{-1}\left(\{0,1\}^{n}\right)\) with their encoding enc \((s) \in\{0,1\}^{n}\) and to represent any subset \(T\) of \(S\) by its characteristic function \(\chi_{T}:\{0,1\}^{n} \rightarrow\{0,1\}\), which evaluates to true exactly for the (encodings of the) states \(s \in T\). Similarly, the transition relation \(\rightarrow \subseteq S \times S\) can be represented by a Boolean function \(\Delta:\{0,1\}^{2 n} \rightarrow\{0,1\}\) that assigns 1 to exactly those pairs \(\left(s, s^{\prime}\right)\) of bit vectors of length \(n\) each where \(s \rightarrow s^{\prime}\).

On the basis of this encoding, the CTL model-checking procedure can be reformulated to an algorithm that operates on the representation of \(T S\) by binary decision diagrams for \(\Delta\) and the characteristic functions \(\chi_{S a t(a)}\) for the satisfaction sets of the atomic propositions \(a \in A P\). The remainder of this section is concerned with explanations on this approach. Section 6.7.1 summarizes our notations for switching functions and operations on them. The encoding of transition systems by switching functions and a corresponding reformulation of the CTL model-checking algorithm will be presented in Section 6.7.2. The main concepts of (ordered) binary decision diagrams are summarized in Section 6.7.3.

\subsection*{6.7.1 Switching Functions}

For technical reasons, it is more appropriate to consider switching functions as mappings from evaluations for certain Boolean variables to the values 0 or 1 rather than functions \(\{0,1\}^{n} \rightarrow\{0,1\}\). This permits simpler definitions of composition operators as we just have to identify the common variables, rather than refer to common arguments via their positions in bit tuples. Furthermore, the reference of the arguments of a switching function by means of variable names is also central for binary decision diagrams.

Let \(z_{1} \ldots, z_{m}\) be Boolean variables and \(\operatorname{Var}=\left\{z_{1}, \ldots, z_{m}\right\}\). Let \(\operatorname{Eval}\left(z_{1}, \ldots, z_{m}\right)\) denote the set of evaluations for \(z_{1}, \ldots, z_{m}\), i.e., functions \(\eta: \operatorname{Var} \rightarrow\{0,1\}\). Evaluations are written as \(\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\). We often use tuple-notations such as \(\bar{z}\) for the variable tuple \(\left(z_{1}, \ldots, z_{m}\right), \bar{b}\) for a bit tuple \(\left(b_{1}, \ldots, b_{m}\right) \in\{0,1\}^{m}\), and \([\bar{z}=\bar{b}]\) as a shorthand for the evaluation \(\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\).

\section*{Notation 6.49. Switching Function}

A switching function for \(\operatorname{Var}=\left\{z_{1}, \ldots, z_{m}\right\}\) is a function \(f: \operatorname{Eval}(\operatorname{Var}) \rightarrow\{0,1\}\). The special case \(m=0\) (i.e., Var \(=\varnothing\) ) is allowed. The switching functions for the empty variable set are just constants 0 or 1 .

To indicate the underlying set of variables of a switching function we often write \(f(\bar{z})\) or \(f\left(z_{1}, \ldots, z_{m}\right)\) rather than \(f\). When an enumeration of the variables in Var is clear from the context, say \(z_{1}, \ldots, z_{m}\), then we often simply write \(f\left(b_{1}, \ldots, b_{m}\right)\) or \(f(\bar{b})\) instead of \(f\left(\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\right)(\) or \(f([\bar{z}=\bar{b}]))\).

Disjunction, conjunction, negation and other Boolean connectives are defined for switching functions in the obvious way. For example, if \(f_{1}\) is a switching function for \(\left\{z_{1}, \ldots, z_{n}, \ldots z_{m}\right\}\) and \(f_{2}\) a switching function for \(\left\{z_{n}, \ldots z_{m}, \ldots, z_{k}\right\}\), where the \(z_{i}\) 's are supposed to be pairwise distinct and \(0 \leqslant n \leqslant m \leqslant k\), then \(f_{1} \vee f_{2}\) is a switching function for \(\left\{z_{1}, \ldots, z_{k}\right\}\) and the values of \(f_{1} \vee f_{2}\) are given by
\[
\begin{aligned}
& \left(f_{1} \vee f_{2}\right)\left(\left[z_{1}=b_{1}, \ldots, z_{k}=b_{k}\right]\right) \\
& =\max \left\{f_{1}\left(\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\right), f_{2}\left(\left[z_{n}=b_{n}, \ldots, z_{k}=b_{k}\right]\right)\right\} .
\end{aligned}
\]

We often simply write \(z_{i}\) for the projection function \(\operatorname{pr}_{z_{i}}: \operatorname{Eval}(\bar{z}) \rightarrow\{0,1\}, \operatorname{pr}_{z_{i}}([\bar{z}=\bar{b}])=\) \(b_{i}\) and 0 or 1 for the constant switching functions. With these notations, switching functions can be represented by Boolean connections of the variables \(z_{i}\) (viewed as projection functions) and constants. For instance, \(z_{1} \vee\left(z_{2} \wedge \neg z_{3}\right)\) stands for a switching function.

\section*{Notation 6.50. Cofactor and Essential Variable}

Let \(f: \operatorname{Eval}\left(z, y_{1}, \ldots, y_{m}\right) \rightarrow\{0,1\}\) be a switching function. The positive cofactor of \(f\) for variable \(z\) is the switching function \(\left.f\right|_{z=1}: \operatorname{Eval}\left(z, y_{1}, \ldots, y_{m}\right) \rightarrow\{0,1\}\) given by
\[
\left.f\right|_{z=1}\left(\mathfrak{c}, b_{1}, \ldots, b_{m}\right)=f\left(1, b_{1}, \ldots, b_{m}\right)
\]
where the bit tuple \(\left(\mathfrak{c}, b_{1}, \ldots, b_{m}\right) \in\{0,1\}^{m+1}\) is short for the evaluation \(\left[z=\mathfrak{c}, y_{1}=\right.\) \(\left.b_{1}, \ldots, y_{m}=b_{m}\right]\). Similarly, the negative cofactor of \(f\) for variable \(z\) is the switching function \(\left.f\right|_{z=0}: \operatorname{Eval}\left(z, y_{1}, \ldots, y_{m}\right) \rightarrow\{0,1\}\) given by \(\left.f\right|_{z=0}\left(\mathfrak{c}, b_{1}, \ldots, b_{m}\right)=f\left(0, b_{1}, \ldots, b_{m}\right)\). If \(f\) is a switching function for \(\left\{z_{1}, \ldots, z_{k}, y_{1}, \ldots, y_{m}\right\}\), then we write \(\left.f\right|_{z_{1}=b_{1}, \ldots, z_{k}=b_{k}}\) for the iterated cofactor, also simply called cofactor of \(f\), given by
\[
\left.f\right|_{z_{1}=b_{1}, \ldots, z_{k}=b_{k}}=\left.\left(\left.\ldots\left(\left.f\right|_{z_{1}=b_{1}}\right)\right|_{z_{2}=b_{2}} \ldots\right)\right|_{z_{k}=b_{k}}
\]

Variable \(z\) is called essential for \(f\) if \(\left.f\right|_{z=0} \neq\left. f\right|_{z=1}\).

The values of \(\left.f\right|_{z_{1}=b_{1}, \ldots, z_{k}=b_{k}}\) for \(f=f\left(z_{1}, \ldots, z_{k}, y_{1}, \ldots, y_{m}\right)\) are given by
\[
\left.f\right|_{z_{1}=b_{1}, \ldots, z_{k}=b_{k}}\left(\mathfrak{c}_{1}, \ldots, \mathfrak{c}_{k}, a_{1}, \ldots, a_{m}\right)=f\left(b_{1}, \ldots, b_{k}, a_{1}, \ldots, a_{m}\right)
\]
where \(\left(\mathfrak{c}_{1}, \ldots, \mathfrak{c}_{k}, a_{1}, \ldots, a_{m}\right)\) is identified with the evaluation \(\left[z_{1}=\mathfrak{c}_{1}, \ldots, z_{k}=\mathfrak{c}_{k}, y_{1}=\right.\) \(\left.a_{1}, \ldots, y_{m}=a_{m}\right]\). As a consequence, the definition of (iterated) cofactors does not depend on the order in which the cofactors for single variables are considered, i.e.:
\[
\left.f\right|_{z_{1}=b_{1}, \ldots, z_{k}=b_{k}}=\left.\left(\left.\ldots\left(\left.f\right|_{z_{i_{1}}=b_{i_{1}}}\right)\right|_{z_{i_{2}}=b_{i_{2}}} \ldots\right)\right|_{z_{i_{k}}=b_{i_{k}}}
\]
for each permutation \(\left(i_{1}, \ldots, i_{k}\right)\) of \((1, \ldots, k)\).
Obviously, variable \(z\) is not essential for the cofactors \(\left.f\right|_{z=0}\) and \(\left.f\right|_{z=1}\). Thus, at most the variables in \(\operatorname{Var} \backslash\left\{z_{1}, \ldots, z_{k}\right\}\) are essential for \(\left.f\right|_{z_{1}=b_{1}, \ldots, z_{k}=b_{k}}\), provided that \(f\) is a switching function for Var.

\section*{Example 6.51. Cofactors and Essential Variables}

Consider the switching function \(f\left(z_{1}, z_{2}, z_{3}\right)\) given by \(\left(z_{1} \vee \neg z_{2}\right) \wedge z_{3}\). Then, \(\left.f\right|_{z_{1}=1}=z_{3}\) and \(\left.f\right|_{z_{1}=0}=\neg z_{2} \wedge z_{3}\). In particular, variable \(z_{1}\) is essential for \(f\).

When we fix the variable set \(\left\{z_{1}, z_{2}, z_{3}\right\}\), then variables \(z_{2}\) and \(z_{3}\) are not essential for the projection function \(p r_{z_{1}}=z_{1}\). In fact, we have \(\left.z_{1}\right|_{z_{2}=0}=\left.z_{1}\right|_{z_{2}=1}=z_{1}\). On the other hand, \(z_{1}\) is essential for the projection function \(z_{1}\) as we have \(\left.z_{1}\right|_{z_{1}=1}=1 \neq 0=\left.z_{1}\right|_{z_{1}=0}\).

For another example, variables \(z_{1}\) and \(z_{2}\) are essential for \(f\left(z_{1}, z_{2}, z_{3}\right)=z_{1} \vee \neg z_{2} \vee\left(z_{1} \wedge z_{2} \wedge\right.\) \(\neg z_{3}\) ), while variable \(z_{3}\) is not, as \(\left.f\right|_{z_{3}=1}=z_{1} \vee \neg z_{2}\) agrees with \(\left.f\right|_{z_{3}=0}=z_{1} \vee \neg z_{2} \vee\left(z_{1} \wedge z_{2}\right)\).

The following lemma yields a decomposition of \(f\) into its cofactors. This simple observation relies on the fact that for any evaluation where \(z\) is assigned to 0 the value of \(f(z, \bar{y})\) agrees with the value of \(\left.f\right|_{z=0}\) under the same evaluation for \(\bar{y}\). And similarly, \(f([z=1, \bar{y}=\bar{b}])=\) \(\left.f\right|_{z=1}([\bar{y}=\bar{b}])\).

\section*{Lemma 6.52. Shannon Expansion}

If \(f\) is a switching function for Var, then for each variable \(z \in \operatorname{Var}\) :
\[
f=\left(\left.\neg z \wedge f\right|_{z=0}\right) \vee\left(\left.z \wedge f\right|_{z=1}\right)
\]

A simple consequence of the Shannon expansion is that \(z\) is not essential for \(f\) if and only if \(f=\left.f\right|_{z=0}=\left.f\right|_{z=1}\).


Figure 6.20: Binary decision tree for \(z_{1} \wedge\left(\neg z_{2} \vee z_{3}\right)\).

\section*{Remark 6.53. Binary Decision Trees}

The Shannon expansion is inherent in the representation of switching functions by binary decision trees. Given a switching function \(f\) for some variable set Var, one first fixes an arbitrary enumeration \(z_{1}, \ldots, z_{m}\) for the variables in Var and then represents \(f\) by a binary tree of height \(m\) such that the two outgoing edges of the inner nodes at level \(i\) stand for the cases \(z_{i}=0\) (depicted by a dashed line) and \(z_{i}=1\) (depicted by a solid line). Thus, the paths from the root to a leaf in that tree represent the evaluations and the corresponding value. The leaves stand for the function values 0 or 1 of \(f\). That is, given the evaluation \(s=\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\), then \(f(s)\) is the value of the terminal node that is reached by traversing the tree from the root using the branch \(z_{i}=b_{i}\) for the node at level \(i\). The subtree of node \(v\) of the binary decision tree for \(f\) and the variable ordering \(z_{1}, \ldots, z_{m}\) yields a representation of the iterated cofactor \(\left.f\right|_{z_{1}=b_{1}, \ldots, z_{i-1}=b_{i-1}}\) (viewed as a switching function for \(\left\{z_{i}, \ldots, z_{m}\right\}\) ) where \(z_{1}=b_{1}, \ldots, z_{i-1}=b_{i-1}\) is the sequence of decisions made along the path from the root to node \(v\).

An example of a binary decision tree for \(f\left(z_{1}, z_{2}, z_{3}\right)=z_{1} \wedge\left(\neg z_{2} \vee z_{3}\right)\) is given in Figure 6.20. We use dashed lines for the edges from an inner node for variable \(z\) representing the case \(z=0\) and solid edges for the case \(z=1\).

Further operators on switching functions that will be needed later are existential quantification over variables and renaming of variables.

\section*{Notation 6.54. Existential and Universal Quantification}

Let \(f\) be a switching function for Var and \(z \in \operatorname{Var}\). Then, \(\exists z . f\) is the switching function given by:
\[
\exists z . f=\left.\left.f\right|_{z=0} \vee f\right|_{z=1}
\]

If \(\bar{z}=\left(z_{1}, \ldots, z_{k}\right)\) and \(z_{i} \in \operatorname{Var}\) for \(1 \leqslant i \leqslant k\), then \(\exists \bar{z} . f\) is a short-form notation for \(\exists z_{1} \cdot \exists z_{2} \ldots . \exists z_{k} . f\). Similarly, universal quantification is defined by
\[
\forall z \cdot f=\left.\left.f\right|_{z=0} \wedge f\right|_{z=1}
\]
and \(\forall \bar{z} \cdot f=\forall z_{1} . \forall z_{2} \ldots \forall z_{k} . f\).

For example, if \(f\left(z, y_{1}, y_{2}\right)\) is given by \(\left(z \vee y_{1}\right) \wedge\left(\neg z \vee y_{2}\right)\), then
\[
\exists z . f=\left.\left.f\right|_{z=0} \vee f\right|_{z=1}=y_{1} \vee y_{2}
\]
and \(\forall z . f=\left.\left.f\right|_{z=0} \wedge f\right|_{z=1}=y_{1} \wedge y_{2}\).

The rename operator simply replaces certain variables with other ones. E.g., renaming variable \(z\) into \(y\) in \(f(z, x)=\neg z \vee x\) yields the switching function \(\neg y \vee x\). Formally:

\section*{Notation 6.55. Rename Operator}

Let \(\bar{z}=\left(z_{1}, \ldots, z_{m}\right), \bar{y}=\left(y_{1}, \ldots, y_{m}\right)\) be variable tuples of the same length and let \(\bar{x}=\) \(\left(x_{1}, \ldots, x_{k}\right)\) be a variable tuple such that none of the variables \(y_{i}\) or \(z_{i}\) is contained in \(\bar{x}\). For the evaluation \(s=[\bar{y}=\bar{b}] \in \operatorname{Eval}(\bar{y}, \bar{x}), s\{\bar{y} \leftarrow \bar{z}\}\) denotes the evaluation in \(\operatorname{Eval}(\bar{z}, \bar{x})\) that is obtained by composing the variable renaming function \(y_{i} \mapsto z_{i}\), for \(1 \leqslant i \leqslant m\) with the evaluation \(s\). That is, \(s\{\bar{y} \leftarrow \bar{z}\}\) agrees with \(s\) for the variables in \(\bar{x}\) and \(s\{\bar{y} \leftarrow \bar{z}\}\) assigns the same value \(b \in\{0,1\}\) to variable \(z_{i}\) as \(s\) to variable \(y_{i}\). Given a switching function \(f: \operatorname{Eval}(\bar{z}, \bar{x}) \rightarrow\{0,1\}\), then the switching function \(f\{\bar{z} \leftarrow \bar{y}\}: \operatorname{Eval}(\bar{y}, \bar{x}) \rightarrow\{0,1\}\) is given by
\[
f\{\bar{z} \leftarrow \bar{y}\}(s)=f(s\{\bar{y} \leftarrow \bar{z}\})
\]
i.e., \(f\{\bar{z} \leftarrow \bar{y}\}([\bar{y}=\bar{b}, \bar{x}=\overline{\mathfrak{c}}])=f([\bar{z}=\bar{b}, \bar{x}=\overline{\mathfrak{c}}])\). If it is clear from the context which variables have to be renamed, then we simply write \(f(\bar{y}, \bar{x})\) rather than \(f\{\bar{z} \leftarrow \bar{y}\}\).

\subsection*{6.7.2 Encoding Transition Systems by Switching Functions}

After this excursus on switching functions, we return to the question of a symbolic representation of a transition system \(T S=(S, \rightarrow, I, A P, L)\). As mentioned above (see page 382), the action set is irrelevant for our purposes and therefore omitted. For the encoding of the states \(s \in S\) we use \(n\) Boolean variables \(x_{1}, \ldots, x_{n}\) and identify any evaluation \(\left[x_{1}=b_{1}, \ldots, x_{n}=b_{n}\right] \in \operatorname{Eval}(\bar{x})\) with the unique state \(s \in S\) such that \(\operatorname{enc}(s)=\left(b_{1}, \ldots, b_{n}\right)\). In the sequel, we suppose \(S=\operatorname{Eval}(\bar{x})\). Given a subset \(B\) of \(S\), then the characteristic function \(\chi_{B}: S \rightarrow\{0,1\}\) of \(B\) assigns 1 to all states \(s \in B\)
and 0 to all states \(s \notin B\). As we assume \(S=\operatorname{Eval}(\bar{x})\), the characteristic function is the switching function given by
\[
\chi_{B}: \operatorname{Eval}(\bar{x}) \rightarrow\{0,1\}, \quad \chi_{B}(s)= \begin{cases}1 & \text { if } s \in B \\ 0 & \text { otherwise }\end{cases}
\]

In particular, for any atomic proposition \(a \in A P\), the satisfaction set \(S a t(a)=\{s \in S \mid\) \(s \models a\}\) can be represented by the switching function \(f_{a}=\chi_{\text {Sat }(a)}\) for \(\bar{x}\). This yields a symbolic representation of the labeling function by means of a family \(\left(f_{a}\right)_{a \in A P}\) of switching functions for \(\bar{x}\).

The symbolic representation of the transition relation \(\rightarrow \subseteq S \times S\) relies on the same idea: we identify \(\rightarrow\) with its characteristic function \(S \times S \rightarrow\{0,1\}\) where truth-value 1 is assigned to the state pair \((s, t)\) if and only if \(s \rightarrow t\). Formally, we deal with the variable tuple \(\bar{x}=\left(x_{1}, \ldots, x_{n}\right)\) to encode the starting state \(s\) of a transition and a copy \(\bar{x}^{\prime}=\left(x_{1}^{\prime}, \ldots, x_{n}^{\prime}\right)\) of \(\bar{x}\) to encode the target state. That is, for each of the variables \(x_{i}\) we introduce a new variable \(x_{i}^{\prime}\). The original (unprimed) variables \(x_{i}\) serve to encode the current state, while the primed variables \(x_{i}^{\prime}\) are used to encode the next state. Then, we identify the transition relation \(\rightarrow\) of \(T S\) with the switching function
\[
\Delta: \operatorname{Eval}\left(\bar{x}, \bar{x}^{\prime}\right) \rightarrow\{0,1\}, \quad \Delta\left(s, t\left\{\bar{x}^{\prime} \leftarrow \bar{x}\right\}\right)= \begin{cases}1 & \text { if } s \rightarrow t \\ 0 & \text { otherwise }\end{cases}
\]
where \(s\) and \(t\) are elements of the state space \(S=\operatorname{Eval}(\bar{x})\) and the second argument \(t\left\{\bar{x}^{\prime} \leftarrow \bar{x}\right\}\) in \(\Delta\left(s, t\left\{\bar{x} / \bar{x}^{\prime}\right\}\right)\) is the evaluation for \(\bar{x}^{\prime}\) that assigns the same value (1 or 0 ) to variable \(x_{i}^{\prime}\) as \(t\) assigns to variable \(x_{i}\) (cf. Notation 6.55).

\section*{Example 6.56. Symbolic Representation of Transition Relation}

Suppose that TS has two states \(s_{0}\) and \(s_{1}\) and the transitions \(s_{0} \rightarrow s_{0}, s_{0} \rightarrow s_{1}\) and \(s_{1} \rightarrow s_{0}\), then we need a single Boolean variable \(x_{1}=x\) for the encoding. Say we identify state \(s_{0}\) by 0 and state \(s_{1}\) by 1 . The transition relation \(\rightarrow\) is represented by the switching function \(\Delta: \operatorname{Eval}\left(x, x^{\prime}\right) \rightarrow\{0,1\}\),
\[
\Delta=\neg x \vee \neg x^{\prime}
\]

Let us check why. The satisfying assignments for \(\Delta\) are \(\left[x=0, x^{\prime}=0\right],\left[x=0, x^{\prime}=1\right]\) and \(\left[x=1, x^{\prime}=0\right]\). The first two evaluations (where \(x=0=0\) ) represent the two outgoing transitions from state \(s_{0}=0\), while \(\left[x=1, x^{\prime}=0\right]\) stands for the transition \(s_{1} \rightarrow s_{0}\).

\section*{Example 6.57. Symbolic Representation of a Ring}

Consider a transition system \(T S\) with states \(\left\{s_{0}, \ldots, s_{k-1}\right\}\) where \(k=2^{n}\) that are organized in a ring, i.e., \(T S\) has the transitions
\[
s_{i} \rightarrow s_{(i+1) \bmod k}
\]
for \(0 \leqslant i<k\). We use the encoding that identifies any state \(s_{i}\) with the binary encoding of its index \(i\). E.g., if \(k=16\), then \(n=4\) and state \(s_{1}\) is identified with 0001, state \(s_{10}\) with 1010 , and state \(s_{15}\) with 1111 . We use the Boolean variables \(x_{1}, \ldots, x_{n}\) where \(x_{n}\) represents the most significant bit (i.e., the evaluation \(\left[x_{n}=b_{n}, \ldots, x_{1}=b_{1}\right]\) stands for state \(\sum_{1 \leqslant i \leqslant n} b_{i} 2^{i-1}\) ). Then, \(\Delta\) is a function with \(2 n\) variables, namely \(x_{1}, x_{2}, \ldots, x_{n}\) and their copies \(x_{1}^{\prime}, x_{2}^{\prime}, \ldots x_{n}^{\prime}\). The values of the switching function \(\Delta\left(\bar{x}, \bar{x}^{\prime}\right)\) are given by
\[
\begin{gathered}
\bigwedge_{1 \leqslant i<n}\left(x_{1} \wedge \ldots \wedge x_{i} \wedge \neg x_{i+1} \rightarrow x^{\prime} \wedge \ldots \wedge x_{i}^{\prime} \wedge x_{i+1}^{\prime} \wedge \bigwedge_{j<i \leqslant n}\left(x_{j} \leftrightarrow x_{j}^{\prime}\right)\right) \\
\wedge\left(x_{1} \wedge \ldots \wedge x_{n} \rightarrow \neg x_{1}^{\prime} \wedge \ldots \wedge \neg x_{n}^{\prime}\right)
\end{gathered}
\]

The set \(B=\left\{s_{2 i} \mid 0 \leqslant i<2^{n-1}\right\}\) is given by the switching function \(\chi_{B}(\bar{x})=x_{1}\).

Given the switching function \(\Delta\) and a state \(s \in S=\operatorname{Eval}(\bar{x})\), then the successor set \(\operatorname{Post}(s)=\left\{s^{\prime} \in S \mid s \rightarrow s^{\prime}\right\}\) arises from \(\Delta\) by fixing evaluation \(s\) for \(\bar{x}\). More precisely, if \(s=\left[x_{1}=b_{1}, \ldots, x_{n}=b_{n}\right]\), then a switching function \(\chi_{\operatorname{Post}(s)}\) for \(\operatorname{Post}(s)\) is obtained from \(\Delta\) by building the cofactor for the variables \(x_{1}, \ldots, x_{n}\) and the values \(b_{1}, \ldots, b_{n}\) :
\[
\chi_{\operatorname{Post}(s)}=\left.\Delta\right|_{s}\left\{\bar{x} \leftarrow \bar{x}^{\prime}\right\}
\]
where \(\left.\Delta\right|_{s}\) stands for the iterated cofactor \(\left.\Delta\right|_{x_{1}=b_{1}, \ldots, x_{n}=b_{n}}\). As \(\left.\Delta\right|_{s}\) is a switching function for \(\left\{x_{1}^{\prime}, \ldots, x_{n}^{\prime}\right\}\), the renaming operator \(\left\{\bar{x} \leftarrow \bar{x}^{\prime}\right\}\) yields a representation of Post( \(s\) ) by the variables \(x_{1}, \ldots, x_{n}\).

\section*{Example 6.58.}

The successor set \(\operatorname{Post}\left(s_{0}\right)=\left\{s_{0}, s_{1}\right\}\) for the simple system in Example 6.56 is obtained symbolically by
\[
\left.\Delta\right|_{x=0}\left\{x \leftarrow x^{\prime}\right\}=\underbrace{\left.\left(\neg x \vee \neg x^{\prime}\right)\right|_{x=0}}_{=1}\left\{x \leftarrow x^{\prime}\right\}=1,
\]
which reflects the fact that after identifying state \(s_{0}\) with the evaluation \([x=0]\) and state \(s_{1}\) with \([x=1]\) the successor set of \(s_{0}\) is \(\operatorname{Eval}(x)=\left\{s_{0}, s_{1}\right\}\), and its characteristic function is the constant 1. For state \(s_{1}=[x=1]\), a symbolic representation of the successor set \(\operatorname{Post}\left(s_{1}\right)=\left\{s_{0}\right\}=\{[x=0]\}\) is obtained by
\[
\left.\Delta\right|_{x=1}\left\{x \leftarrow x^{\prime}\right\}=\underbrace{\left.\left(\neg x \vee \neg x^{\prime}\right)\right|_{x=1}}_{=\neg x^{\prime}}\left\{x \leftarrow x^{\prime}\right\}=\neg x
\]

\section*{Remark 6.59. Symbolic Composition Operators}

As we explained in Chapter 2, a crucial aspect for the feasibility of any algorithmic verification technique is the automatic construction of "large" transition systems to be analyzed by means of operators that compose several "small" transition systems \(T S_{1}, \ldots, T S_{m}\) representing the processes to be executed in parallel. Let us suppose that we have appropriate representations for the switching functions \(\Delta_{1}, \ldots, \Delta_{m}\) for transition systems \(T S_{1}, \ldots, T S_{m}\) at our disposal. If \(T S\) arises by \(T S_{1}, \ldots, T S_{m}\) through the synchronous product operator of \(T S\), then the transition relation of \(T S\) is given by
\[
\Delta\left(\bar{x}_{1}, \ldots, \bar{x}_{m}, \bar{x}_{1}^{\prime}, \ldots, \bar{x}_{m}^{\prime}\right)=\bigwedge_{1 \leqslant i \leqslant n} \Delta_{i}\left(\bar{x}_{i}, \bar{x}_{i}^{\prime}\right)
\]
where \(\bar{x}_{i}\) denotes the variable tuple that is used to encode the states in \(T S_{i}\). The justification for the above equation is that each transition \(\left\langle s_{1}, \ldots, s_{m}\right\rangle \rightarrow\left\langle s_{1}^{\prime}, \ldots, s_{m}^{\prime}\right\rangle\) of \(T S=T S_{1} \otimes \ldots \otimes T S_{m}\) is composed of individual transitions \(s_{i} \rightarrow s_{i}^{\prime}\) in \(T S_{i}\) for each of the transition systems \(T S_{i}\). For the other extreme, suppose that \(T S=T S_{1}\| \|\| \| S_{m}\) arises by the interleaving operator of the \(T S_{i}\) 's (without any synchronization or communication). Then,
\[
\Delta\left(\bar{x}_{1}, \ldots, \bar{x}_{m}, \bar{x}_{1}^{\prime}, \ldots, \bar{x}_{m}^{\prime}\right)=\bigvee_{1 \leqslant i \leqslant n}\left(\Delta_{i}\left(\bar{x}_{i}, \bar{x}_{i}^{\prime}\right) \wedge \bigwedge_{\substack{1 \leqslant j \leqslant m \\ i \neq j}} \bar{x}_{j}=\bar{x}_{j}^{\prime}\right)
\]
where for \(\bar{x}_{j}=\left(x_{1, j}, \ldots, x_{n_{j}, j}\right)\) and \(\bar{x}_{j}^{\prime}=\left(x_{1, j}^{\prime}, \ldots, x_{n_{j}, j}^{\prime}\right)\) notation \(\bar{x}_{j}=\bar{x}_{j}^{\prime}\) abbreviates \(\bigwedge_{1 \leqslant k \leqslant n_{j}}\left(x_{k, j} \leftrightarrow x_{k, j}^{\prime}\right)\). The justification for the above equation for \(\Delta\) is that each transition in TS has the form
\[
\left\langle s_{1}, \ldots, s_{i-1}, s_{i}, s_{i+1}, \ldots, s_{m}\right\rangle \rightarrow\left\langle s_{1}, \ldots, s_{i-1}, s_{i}^{\prime}, s_{i+1}, \ldots, s_{m}\right\rangle
\]
where exactly one transition system makes a move \(s_{i} \rightarrow s_{i}^{\prime}\), while the others do not change their local state. For the parallel operator \(\|_{H}\) which relies on an interleaving semantics for actions outside \(H\) and synchronization over the actions in \(H\), we need a combination of the above formulae for \(\Delta\). Here, in fact, we need a refined representation of the transition relation in \(T S_{i}\) with actions for the transitions by means of a switching function \(\Delta_{i}\left(\bar{x}_{i}, \bar{z}, \bar{x}_{i}^{\prime}\right)\) where the variable tuple \(\bar{z}\) serves to encode the action names. In case \(m=2\), the switching function \(\Delta\left(\bar{x}_{1}, \bar{x}_{2}, \bar{x}_{1}^{\prime}, \bar{x}_{2}^{\prime}\right)\) for the transition relation in \(T S=T S_{1} \|_{H} T S_{2}\) is given by
\[
\begin{aligned}
& \exists \bar{z} .\left(\quad\left(\quad \chi_{H}(\bar{z}) \wedge \Delta_{1}\left(\bar{x}_{1}, \bar{z}, \bar{x}_{1}^{\prime}\right) \wedge \Delta_{2}\left(\bar{x}_{1}, \bar{z}, \bar{x}_{2}^{\prime}\right)\right) \vee\right. \\
& \left(\neg \chi_{H}(\bar{z}) \wedge \Delta_{1}\left(\bar{x}_{1}, \bar{z}, \bar{x}_{1}^{\prime}\right) \wedge \bar{x}_{2}=\bar{x}_{2}^{\prime}\right) \vee \\
& \left(\neg \chi_{H}(\bar{z}) \wedge \Delta_{2}\left(\bar{x}_{2}, \bar{z}, \bar{x}_{2}^{\prime}\right) \wedge \bar{x}_{1}=\bar{x}_{1}^{\prime} \quad\right) \text { ). }
\end{aligned}
\]

In case \(T S\) is the transition system of a program graph or channel system then the Boolean variables \(x_{i}\) serve to encode the locations, the potential values of the variables, and the
channel contents. The effect of the actions has then to be rewritten in terms of these variables.

Given the switching function \(\Delta\left(\bar{x}, \bar{x}^{\prime}\right)\) and the characteristic function \(\chi_{B}(\bar{x})\) for some set \(B\) of states, we can now describe the backward BFS-based reachability analysis to compute all states in \(\operatorname{Pr}^{*}(B)=\{s \in S \mid s \vDash \exists \diamond B\}\) on the basis of switching functions. Initially, we start with the switching function \(f_{0}=\chi_{B}\) that characterizes the set \(T_{0}=B\). Then, we successively compute the characteristic functions \(f_{j+1}=\chi_{T_{j+1}}\) of
\[
T_{j+1}=T_{j} \cup\left\{s \in S \mid \exists s^{\prime} \in S \text { s.t. } s^{\prime} \in \operatorname{Post}(s) \wedge s^{\prime} \in T_{j}\right\}
\]

The set of states \(s\) where the condition \(\exists s^{\prime} \in S\) s.t. \(s^{\prime} \in \operatorname{Post}(s)\) and \(s^{\prime} \in T_{j}\) holds is given by the switching function:
\[
\exists \bar{x}^{\prime} \cdot(\underbrace{\Delta\left(\bar{x}, \bar{x}^{\prime}\right)}_{s^{\prime} \in \operatorname{Post}(s)} \wedge \underbrace{f_{j}\left(\bar{x}^{\prime}\right)}_{s^{\prime} \in T_{j}}) .
\]

Recall that \(f_{j}\left(\bar{x}^{\prime}\right)\) is just a short notation for \(f_{j}\left\{\bar{x}^{\prime} \leftarrow \bar{x}\right\}\), i.e., arises from \(f_{j}\) by renaming the unprimed variables \(x_{i}\) into their primed copies \(x_{i}^{\prime}\) for \(1 \leqslant i \leqslant n\). This BFS-based technique can easily be adapted to treat constrained reachability properties \(\exists(C \cup B)\) for subsets \(B, C\) of \(S\), as shown in Algorithm 20 on page 390 .
```

Algorithm 20 Symbolic computation of $\operatorname{Sat}(\exists(C \cup B))$
$f_{0}(\bar{x}):=\chi_{B}(\bar{x}) ;$
$j:=0$;
repeat
$f_{j+1}(\bar{x}):=f_{j+1}(\bar{x}) \vee\left(\chi_{C}(\bar{x}) \wedge \exists \bar{x}^{\prime} \cdot\left(\Delta\left(\bar{x}, \bar{x}^{\prime}\right) \wedge f_{j}\left(\bar{x}^{\prime}\right)\right)\right) ;$
$j:=j+1$
until $f_{j}(\bar{x})=f_{j-1}(\bar{x})$;
return $f_{j}(\bar{x})$.

```

If we are just interested in the one-step predecessors, i.e., properties of the form \(\exists \bigcirc B\), then no iteration is needed and the characteristic function of the set of states \(s \in S\) with \(\operatorname{Post}(s) \cap B \neq \varnothing\) is obtained by
\[
\exists \bar{x}^{\prime} \cdot(\underbrace{\Delta\left(\bar{x}, \bar{x}^{\prime}\right)}_{s^{\prime} \in \operatorname{Post}(s)} \wedge \underbrace{\chi_{B}\left(\bar{x}^{\prime}\right)}_{s^{\prime} \in B})
\]

In a similar way, the set \(\operatorname{Sat}(\exists \square B)\) of all states \(s\) that have an infinite path consisting of states in a given set \(B\) can be computed symbolically. Here, we mimic the computation of the largest set \(T \subseteq B\) with \(\operatorname{Post}(t) \cap T \neq \varnothing\) for all \(t \in T\) by the iteration \(T_{0}=B\) and
\[
T_{j+1}=T_{j} \cap\left\{s \in S \mid \exists s^{\prime} \in S \text { s.t. } s^{\prime} \in \operatorname{Post}(s) \wedge s^{\prime} \in T_{j}\right\}
\]
```

Algorithm 21 Symbolic computation of $\operatorname{Sat}(\exists \square B)$
$f_{0}(\bar{x}):=\chi_{B}(\bar{x}) ;$
$j:=0$;
repeat
$f_{j+1}(\bar{x}):=f_{j+1}(\bar{x}) \wedge \exists \bar{x}^{\prime} .\left(\Delta\left(\bar{x}, \bar{x}^{\prime}\right) \wedge f_{j}\left(\bar{x}^{\prime}\right)\right) ;$
$j:=j+1$
until $f_{j}(\bar{x})=f_{j-1}(\bar{x})$;
return $f_{j}(\bar{x})$.

```
in a symbolic way, as shown in Algorithm 21 on page 391.
These considerations show that the CTL model checking problem "Does CTL formula \(\Phi\) hold for TS?" can be solved symbolically by means of switching functions \(f_{\Psi}\) that represent the satisfaction sets \(\operatorname{Sat}(\Psi)\) of the subformulae \(\Psi\) of \(\Phi\). The satisfaction sets \(f_{a}\) for the atomic propositions \(a \in A P\) are supposed to be given. Union, intersection, and complementation of sets of states correspond to disjunction, conjunction, and negation on the level of switching functions. E.g., the satisfaction set for \(\Psi_{1} \wedge \neg \Psi_{2}\), is obtained by \(f_{\Psi_{1}} \wedge\) \(\neg f_{\Psi_{2}}\). The treatment of formulae \(\exists\left(\Psi_{1} \cup \Psi_{2}\right)\) and \(\exists \square \Psi\) relies on the techniques sketched in Algorithms 20 and 21. To treat full CTL, we can either transform any CTL formula into an equivalent CTL formula in existential normal form or use analogous symbolic algorithms for universally quantified formulae such as \(\forall(\Phi \cup \Psi)\).

The major challenge is to find an appropriate data structure for the switching functions. Besides yielding compact representations of the satisfaction sets and the transition relation, this data structure has to support the Boolean connectives (disjunction, conjunction, complementation) and the comparison of two switching functions. The latter is needed in the termination criteria for the repeat loops in Algorithms 20 and 21.

Before presenting the definition of binary decision diagrams that have proven to be very efficient in many applications, let us first discuss other potential data structures. Truth tables can be ruled as they always have the size \(2^{n}\) for switching functions with \(n\) variables. The same holds for binary decision trees since they always have \(2^{n+1}-1\) nodes. Conjunctive or disjunctive normal forms for the representation of switching functions by propositional logic formulae yield the problem that checking equivalence (i.e., equality for the represented switching functions) is expensive, namely coNP-complete. Furthermore, there are switching functions \(f_{m}\) with \(m\) essential variables where the length of any representation by a conjunctive normal form formula grows exponentially in \(m\). The same holds for disjunctive normal forms. However, the latter is no proper argument against conjunctive or disjunctive normal forms, because there is no data structure that ensures representations of polynomial size for all switching functions. The reason for this is that the
number of switching functions for \(m\) variables \(z_{1}, \ldots, z_{m}\) is double exponential in \(m\). Note that \(\left|\operatorname{Eval}\left(z_{1}, \ldots, z_{m}\right)\right|=2^{m}\), and hence, the number of functions \(\operatorname{Eval}\left(z_{1}, \ldots, z_{m}\right) \rightarrow\{0,1\}\) is \(2^{2^{m}}\). Suppose we are given a universal data structure for switching functions (i.e., a data structure that can represent any switching function) such that \(K_{m}\) is the number of switching functions for \(z_{1}, \ldots, z_{m}\) that can be represented by at most \(2^{m-1}\) bits. Then:
\[
K_{m} \leqslant \sum_{i=0}^{2^{m-1}} 2^{i}=2^{2^{m-1}+1}-1<2^{2^{m-1}+1}
\]

But then there are at least
\[
2^{2^{m}}-2^{2^{m-1}+1}=2^{2^{m-1}+1} \cdot\left(2^{2^{m}-2^{m-1}-1}-1\right)=2^{2^{m-1}+1} \cdot\left(2^{2^{m-1}-1}-1\right)
\]
switching functions for \(z_{1}, \ldots, z_{m}\) where the representation requires more than \(2^{m-1}\) bits. This calculation shows that we cannot expect a data structure which is efficient for all switching functions. Nevertheless there are data structures which yield compact representations for many switching functions that appear in practical applications. A data structure that has been proven to be very successful for model checking purposes, in particular in the area of hardware verification, is ordered binary decision diagrams (OBDDs). Besides yielding compact representation for many "realistic" transition systems, they enjoy the property that the Boolean connectives can be realized in time linear in the size of the input OBDDs and that (with appropriate implementation techniques) equivalence checking can even be performed in constant time.

In the remainder of this section, we explain those aspects of ordered binary decision diagrams that are relevant to understanding the main concepts of symbolic model checking with these data structures. Further theoretical aspects on binary decision diagrams, their variants and applications, can be found, e.g., in the textbooks [134, 180, 292, 300, 418]. For details on OBDD-based symbolic model checking we refer to [74, 92, 288, 374].

\subsection*{6.7.3 Ordered Binary Decision Diagrams}

Ordered binary decision diagrams (OBDDs for short), originally proposed by Bryant [70], yield a data structure for switching functions that relies on a compactification of binary decision trees. The rough idea is to skip redundant fragments of a binary decision tree. This means collapsing constant subtrees (i.e., subtrees where all terminal nodes have the same value) into a single node and identifying nodes with isomorphic subtrees. In this way, we obtain a directed acyclic graph of outdegree 2 where - as in binary decision trees - the inner nodes are labeled by variables and their outgoing edges stand for the possible evaluations of the corresponding variable. The terminal nodes are labeled by the function value.


Figure 6.21: Binary decision diagram for \(z_{1} \wedge\left(\neg z_{2} \vee z_{3}\right)\).

\section*{Example 6.60. From Binary Decision Tree to \(O B D D\)}

Before presenting the formal definition of binary decision diagrams, we explain the main ideas by means of the function \(f\left(z_{1}, z_{2}, z_{3}\right)=z_{1} \wedge\left(\neg z_{2} \vee z_{3}\right)\). A binary decision tree for \(f\) has been shown in Figure 6.20 on page 385. Since all terminal nodes in the right subtree of the root have value 0 (which reflects the fact that the cofactor \(\left.f\right|_{z_{1}=0}\) agrees with the constant function 0 ), the inner tests for variables \(z_{2}\) and \(z_{3}\) in that subtree are redundant and the whole subtree can be replaced by a terminal node with value 0 . Similarly, the subtree of the \(z_{3}\)-node representing the cofactor \(\left.f\right|_{z_{1}=1, z_{2}=0}=1\) can be replaced with a terminal node for the value 1. This leads to the graph shown on the left of Figure 6.21. Finally, we may identify all terminal nodes with the same value, which yields the graph shown on the right of Figure 6.21.

\section*{Example 6.61. From Binary Decision Tree to OBDD}

As another example, consider the switching function \(f=\left(z_{1} \wedge z_{3}\right) \vee\left(z_{2} \wedge z_{3}\right)\). The upper part of Figure 6.22 on page 394 shows a binary decision tree for \(f\). The subtree of the \(z_{3}\) node on the right is constant and can be replaced by a terminal node. The three subtrees of the \(z_{3}\)-nodes for the cofactors \(\left.f\right|_{z_{1}=0, z_{2}=1},\left.f\right|_{z_{1}=1, z_{2}=0}\), and \(\left.f\right|_{z_{1}=1, z_{2}=1}\) are isomorphic and can thus be collapsed. This yields the decision graph shown in the middle part of Figure 6.22. But now the \(z_{2}\)-node for the cofactor \(\left.f\right|_{z_{1}=1}\) becomes redundant, as regardless whether \(z_{2}=0\) or \(z_{2}=1\), the same node will be reached. This permits removing this \(z_{2}{ }^{-}\) node and redirecting its incoming edge. This yields the binary decision diagram depicted in the lower part of Figure 6.22.


Figure 6.22: Binary decision diagrams for \(f=\left(z_{1} \wedge z_{3}\right) \vee\left(z_{2} \wedge z_{3}\right)\).

\section*{Notation 6.62. Variable Ordering}

Let Var be a finite set of variables. A variable ordering for Var denotes any tuple \(\wp=\) \(\left(z_{1}, \ldots, z_{m}\right)\) such that \(\operatorname{Var}=\left\{z_{1}, \ldots, z_{m}\right\}\) and \(z_{i} \neq z_{j}\) for \(1 \leqslant i<j \leqslant m\). We write \(<_{\wp}\) for the induced total order on Var. I.e., for \(\wp=\left(z_{1}, \ldots, z_{m}\right)\) the binary relation \(<_{\wp}\) on Var is given by \(z_{i}<_{\wp} z_{j}\) if and only if \(i<j\). We write \(z_{i} \leqslant_{\wp} z_{j}\) iff either \(z_{i}<_{\wp} z_{j}\) or \(i=j\).

\section*{Definition 6.63. Ordered Binary Decision Diagram (OBDD)}

Let \(\wp\) be a variable ordering for Var. An \(\wp\)-ordered binary decision diagram ( \(\wp\)-OBDD for short) is a tuple
\[
\mathfrak{B}=\left(V, V_{I}, V_{T}, \text { succ }_{0}, \text { succ }_{1}, \text { var, val, } v_{0}\right)
\]
consisting of
- a finite set \(V\) of nodes, disjointly partitioned into \(V_{I}\) and \(V_{T}\) where the nodes in \(V_{I}\) are called inner nodes, while the nodes in \(V_{T}\) are called terminal nodes or drains;
- successor functions succ \(_{0}\), succ \(_{1}: V_{I} \rightarrow V\) that assign to each inner node \(v\) a 0 -successor \(\operatorname{succ}_{0}(v) \in V\) and a 1 -successor \(\operatorname{succ}_{1}(v) \in V\);
- a variable labeling function var : \(V_{I} \rightarrow\) Var that assigns to each inner node \(v\) a variable \(\operatorname{var}(v) \in \operatorname{Var} ;\)
- a value function val : \(V_{T} \rightarrow\{0,1\}\) that assigns to each drain a function value 0 or 1;
- a root (node) \(v_{0} \in V\).

Consistency of the variable labeling function with the variable ordering \(\wp\) is required in the following sense. If \(\wp=\left(z_{1}, \ldots, z_{m}\right)\), then for each inner node \(v\) : if \(\operatorname{var}(v)=z_{i}\) and \(w \in\left\{\operatorname{succ}_{0}(v), \operatorname{succ}_{1}(v)\right\} \cap V_{I}\), then \(\operatorname{var}(w)=z_{j}\) for some \(j>i\). Furthermore, it is required that each node \(v \in V \backslash\left\{v_{0}\right\}\) has at least one predecessor, i.e., \(v=\operatorname{succ}_{b}(w)\) for some \(w \in V\) and \(b \in\{0,1\}\).

The underlying digraph of a \(\wp-\mathrm{OBDD}\) is obtained by using \(V\) as node set and establishing an edge from node \(v\) to \(w\) if and only if \(w\) is an successor of \(v\), i.e., if \(w \in\) \(\left\{\operatorname{succ}_{0}(v), \operatorname{succ}_{1}(v)\right\}\). The last condition in Definition 6.63, which states that each node of an OBDD, except for the root, has a predecessor, is equivalent to the condition that all nodes of an OBDD are reachable from the root. The size of an OBDD \(\mathfrak{B}\), denoted \(\operatorname{size}(\mathfrak{B})\), is the number of nodes in \(\mathfrak{B}\).

An equivalent formulation for the order consistency of \(\mathfrak{B}\) (i.e., the condition stating that the variable of an inner node appears in \(\wp\) before the variables of its successors, provided they are inner nodes) is the requirement that for each path \(v_{0} v_{1} \ldots v_{n}\) in (the underlying graph of) \(\mathfrak{B}\) we have \(v_{i} \in V_{I}\) for \(1 \leqslant i<n\) and
\[
\operatorname{var}\left(v_{0}\right)<_{\wp} \operatorname{var}\left(v_{1}\right)<_{\wp} \ldots<_{\wp} \operatorname{var}\left(v_{n}\right)
\]
where for the drains we put \(\operatorname{var}(v)=\perp\) (undefined) and extend \(<_{\wp}\) by \(z<_{\wp} \perp\) for all variables \(z \in \operatorname{Var}\). That is, we regard \(<_{\wp}\) as a total order on \(\operatorname{Var} \cup\{\perp\}\) and consider the variable labeling function as a function var : V \(\rightarrow \operatorname{Var} \cup\{\perp\}\). This yields that the underlying graph of an OBDD is acyclic. In particular, \(v \neq \operatorname{succ}_{b}(v)\) for all inner nodes \(v\) and \(b \in\{0,1\}\).

Examples of OBDDs are the binary decision trees and graphs shown in Figures 6.20, 6.21, and 6.22. All these OBDDs rely on the variable ordering \(\wp=\left(z_{1}, z_{2}, z_{3}\right)\).

In the sequel, we often refer to an inner node \(v\) with \(\operatorname{var}(v)=z\) as a \(z\)-node. As the pictures for OBDDs suggest, we may think of the node set \(V\) of an OBDD to be partitioned into levels. Dealing with the ordering \(\wp=\left(z_{1}, \ldots, z_{m}\right)\), then the \(z_{i}\)-nodes constitute the nodes at level \(i\). The drains yields the bottom level, while the top level consists of the root node.

\section*{Definition 6.64. Semantics of OBDDs}

Let \(\mathfrak{B}\) be an \(\wp\)-OBDD as in Definition 6.63. The semantics \(\mathfrak{B}\) is the switching function \(f_{\mathfrak{B}}\) for Var where \(f_{\mathfrak{B}}\left(\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\right)\) is the value of the drain that will be reached when traversing the graph starting in the root \(v_{0}\) and branching according to the evaluation \(\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\). That is, if the current node \(v\) is a \(z_{i}\)-node, then the traversal continues with the \(b_{i}\)-successor of \(v\).

\section*{Definition 6.65. Sub-OBDD, Switching Function for the Nodes}

Let \(\mathfrak{B}\) be a \(\wp-O B D D\). If \(v\) is a node in \(\mathfrak{B}\), then the sub- \(O B D D\) induced by \(v\), denoted \(\mathfrak{B}_{v}\) arises from \(\mathfrak{B}\) by declaring \(v\) as the root node and removing all nodes that are not reachable from \(v\). The switching function for node, denoted \(f_{v}\), is the switching function for Var that is given by the sub-OBDD \(\mathfrak{B}_{v}\).

Clearly, at most the variables \(x\) with \(\operatorname{var}(v) \leqslant_{\wp} x\) can be essential for \(f_{v}\), since none of the variables \(z\) with \(z<_{\wp} \operatorname{var}(v)\) appear in the sub-OBDD. Hence, \(f_{v}\) could also be viewed as a switching function for the whole variable set Var or as a switching function for the variables \(x\) with \(\operatorname{var}(v) \leqslant_{\wp} x\), but this is irrelevant here. In particular, if \(v\) is a \(z\)-node, then the order condition \(z=\operatorname{var}(v)<_{\wp} \operatorname{var}\left(\operatorname{succ}_{b}(v)\right)\) yields that \(\left.f_{\text {Succ }_{b}(v)}\right|_{z=\mathfrak{c}}=f_{\text {Succ }_{b}(v)}\),
since variable \(z\) is not essential for \(f_{\text {Succ }_{b}(v)}\). But then:
\[
\begin{aligned}
\left.f_{v}\right|_{z=0} & =\left.\left(\neg z \wedge f_{\text {Succ }_{0}(v)}\right)\right|_{z=0} \vee \underbrace{\left.\left(z \wedge f_{\text {Succ }_{1}(v)}\right)\right|_{z=0}}_{=0} \\
& =\left.f_{\text {Succ }_{0}(v)}\right|_{z=0}=f_{\text {Succ }_{0}(v)}
\end{aligned}
\]
and, similarly, \(\left.f_{v}\right|_{z=1}=f_{\text {Succ }_{1}(v)}\). Thus, the Shannon expansion yields the following bottom-up characterization of the functions \(f_{v}\) :

\section*{Lemma 6.66. Bottom-up Characterization of the Functions \(f_{v}\)}

Let \(\mathfrak{B}\) be a \(\wp-O B D D\). The switching functions \(f_{v}\) for the nodes \(v \in V\) are given as follows:
- If \(v\) is a drain, then \(f_{v}\) is the constant switching function with value val(v).
- If \(v\) is a \(z\)-node, then \(f_{v}=\left(\neg z \wedge f_{\text {Succ }_{0}(v)}\right) \wedge\left(z \wedge f_{\text {Succ }_{1}(v)}\right)\).

Furthermore, \(f_{\mathfrak{B}}=f_{v_{0}}\) for the root \(v_{0}\) of \(\mathfrak{B}\).

This yields that \(f_{v}=\left.f_{\mathfrak{B}}\right|_{z_{1}=b_{1}, \ldots, z_{i}=b_{i}}\) where \(\left[z_{1}=b_{1}, \ldots, z_{i}=b_{i}\right]\) is an evaluation which leads from the root \(v_{0}\) of \(\mathfrak{B}\) to node \(v\). In fact, all concepts of OBDD-based approaches rely on the decomposition of switching functions into their cofactors. However, only the cofactors are relevant that arise from \(f\) by assigning values to the first \(i\) variables of \(\wp\) for some \(i\).

\section*{Notation 6.67. \(\wp\)-Consistent Cofactor}

Let \(f\) be a switching function for Var and \(\wp=\left(z_{1}, \ldots, z_{m}\right)\) a variable ordering for Var. A switching function \(f^{\prime}\) for Var is called a \(\wp\)-consistent cofactor of \(f\) if there exists some \(i \in\{0,1, \ldots, m\}\) such that \(f^{\prime}=\left.f\right|_{z_{1}=b_{1}, \ldots, z_{i}=b_{i}}\). (Including the case \(i=0\) means that we regard \(f\) as a cofactor of itself.)

For instance, if \(f=z_{1} \wedge\left(z_{2} \vee \neg z_{3}\right)\) and \(\wp=\left(z_{1}, z_{2}, z_{3}\right)\), then the \(\wp\)-consistent cofactors of \(f\) are the switching functions \(f,\left.f\right|_{z_{1}=1}=z_{2} \vee \neg z_{3},\left.f\right|_{z_{1}=1, z_{2}=0}=\neg z_{3}\) and the constants 0 and 1. The cofactors \(\left.f\right|_{z_{3}=0}=z_{1}\) and \(\left.f\right|_{z_{2}=0}=z_{1} \wedge \neg z_{3}\) are not \(\wp\)-consistent. Since it is possible that some cofactors of \(f\) arise by several evaluations, it is possible that cofactors \(\left.f\right|_{z_{i_{1}}=b_{1}, \ldots, z_{i_{k}}=b_{k}}\) are \(\wp\)-consistent for the variable ordering \(\wp=\left(z_{1}, \ldots, z_{m}\right)\), even if \(\left(z_{i_{1}}, \ldots, z_{i_{k}}\right)\) is not a permutation of \(\left(z_{1}, \ldots, z_{k}\right)\). For example, for \(f=z_{1} \wedge\left(z_{2} \vee \neg z_{3}\right)\) and \(\wp=\left(z_{1}, z_{2}, z_{3}\right)\) the cofactor \(\left.f\right|_{z_{2}=0, z_{3}=1}\) is \(\wp\)-consistent as it agrees with the cofactors \(f_{z_{1}=0}\) or \(\left.f\right|_{z_{1}=1, z_{2}=0, z_{3}=1}\). (They all agree with the constant function 0 .)

The observation made above can now be rephrased as follows:

\section*{Lemma 6.68. Nodes in OBDDs and \(\wp\)-Consistent Cofactors}

For each node \(v\) of an \(\wp-O B D D \mathfrak{B}\), the switching function \(f_{v}\) is a \(\wp\)-consistent cofactor of \(f_{\mathfrak{B}}\). Vice versa, for each \(\wp\)-consistent cofactor \(f^{\prime}\) of \(f_{\mathfrak{B}}\) there is at least one node \(v\) in \(\mathfrak{B}\) such that \(f_{v}=f^{\prime}\).

However, given a \(\wp\)-OBDD \(\mathfrak{B}\) and a \(\wp\)-consistent cofactor \(f^{\prime}\) of \(f_{\mathfrak{B}}\) there could be more than one node in \(\mathfrak{B}\) representing \(f^{\prime}\). This, for instance, applies to the binary decision tree shown in Figure 6.20 on page 385, viewed as \(\wp\)-OBDD for \(\wp=\left(z_{1}, z_{2}, z_{3}\right)\), where we have \(f_{\mathfrak{B}}=z_{1} \wedge\left(\neg z_{2} \vee z_{3}\right)\) and the \(\wp\)-consistent cofactors represented by the nodes in the left subtree of the root agree as we have
\[
\left.f\right|_{z_{1}=0}=\left.f\right|_{z_{1}=0, z_{2}=b}=\left.f\right|_{z_{1}=0, z_{2}=b, z_{3}=\mathfrak{c}}=0
\]
for all \(b, \mathfrak{c} \in\{0,1\}\). For the \(\wp\)-OBDD shown on the left of Figure 6.21 on page 393, all inner nodes represent different switching functions. However, the two drains with value 0 represent the same cofactors of \(f_{\mathfrak{B}}\). The same holds for the two drains with value 1 . However, in the \(\wp\)-OBDD shown on the right of Figure 6.21, every \(\wp\)-conistent cofactor is represented by a single node. In this sense, this \(\wp\)-OBDD is free of redundancies, and therefore called reduced:

\section*{Definition 6.69. Reduced OBDD}

Let \(\mathfrak{B}\) be a \(\wp\)-OBDD. \(\mathfrak{B}\) is called reduced if for every pair \((v, w)\) of nodes in \(\mathfrak{B}: v \neq w\) implies \(f_{v} \neq f_{w}\). Let \(\wp\)-ROBDD denote a reduced \(\wp\)-OBDD.

Thus, in reduced \(\wp\)-OBDDs any \(\wp\)-consistent cofactor is represented by exactly one node. This is the key property to prove that reduced OBDDs yield a universal and canonical data structure for switching functions. Universality means that any switching function can be represented by an OBDD. Canonicity means that any two \(\wp\)-OBDDs for the same function agree up to isomorphism, i.e., renaming of the nodes.

\section*{Theorem 6.70. Universality and Canonicity of ROBDDs}

Let Var be a finite set of Boolean variables and \(\wp\) a variable ordering for Var. Then:
(a) For each switching function \(f\) for Var there exists a \(\wp-R O B D D \mathfrak{B}\) with \(f_{\mathfrak{B}}=f\).
(b) Given two \(\wp-R O B D D s \mathfrak{B}\) and \(\mathfrak{C}\) with \(f_{\mathfrak{B}}=f_{\mathfrak{C}}\), then \(\mathfrak{B}\) and \(\mathfrak{C}\) are isomorphic, i.e., agree up to renaming of the nodes.

Proof: ad (a). Clearly, the constant functions 0 and 1 can be represented by a \(\wp\)-ROBDD consisting of a single drain. Given a nonconstant switching function \(f\) for Var and a variable ordering \(\wp\) for Var, we construct a reduced \(\wp\)-OBDD \(\mathfrak{B}\) for \(f\) as follows. Let \(V\) be the set of \(\wp\)-consistent cofactors of \(f\). The root of \(\mathfrak{B}\) is \(f\). The constant cofactors are the drains with the obvious values. For \(f^{\prime} \in V, f^{\prime} \notin\{0,1\}\), let
\[
\operatorname{var}\left(f^{\prime}\right)=\min \left\{z \in \operatorname{Var} \mid z \text { is essential for } f^{\prime}\right\}
\]
be the first essential variable where the minimum is taken according to the total order \(<_{\wp}\) induced by \(\wp\). (We use here the trivial fact that any nonconstant switching function has at least one essential variable.) The successor functions are given by
\[
\operatorname{succ}_{0}\left(f^{\prime}\right)=\left.f^{\prime}\right|_{z=0}, \quad \operatorname{succ}_{1}\left(f^{\prime}\right)=\left.f^{\prime}\right|_{z=1}
\]
where \(z=\operatorname{var}\left(f^{\prime}\right)\). The definition of \(\operatorname{var}(\cdot)\) yields that \(\mathfrak{B}\) is a \(\wp-O B D D\). By the Shannon expansion we get that the semantics of \(f^{\prime} \in V\) (i.e., the switching function of \(f^{\prime}\) as a node of \(\mathfrak{B}\) ) is \(f^{\prime}\). In particular, this yields that \(f_{\mathfrak{B}}=f\) (the function for the root \(f\) ) and the reducedness of \(\mathfrak{B}\) (as any two nodes represent different cofactors of \(f\) ).

To prove the statement in (b), it suffices to show that any reduced \(\wp\)-OBDD \(\mathfrak{C}\) with \(f_{\mathfrak{C}}=f\) is isomorphic to the \(\wp\)-ROBDD \(\mathfrak{B}\) constructed above. Let \(V^{\mathfrak{C}}\) be the node set of \(\mathfrak{C}\), \(v_{0}^{\mathfrak{C}}\) the root of \(\mathfrak{C}\), \(\operatorname{var}^{\mathfrak{C}}\) the variable labeling function, and \(\operatorname{succ}_{0}^{\mathfrak{C}}\), \(\operatorname{succ}_{1}^{\mathfrak{C}}\) the successor functions in \(\mathfrak{C}\). Let function \(\imath: V^{\mathfrak{C}} \rightarrow V\) be given by \(\imath(v)=f_{v}\). (Recall that the functions \(f_{v}\) are \(\wp\)-consistent cofactors of \(f_{\mathfrak{C}}=f\). This ensures that \(f_{v} \in V\).) Since \(\mathfrak{C}\) is reduced, \(\imath\) is a bijection. It remains to show that \(\imath\) preserves the variable labeling of inner nodes and maps the successors of an inner node \(v\) of \(\mathfrak{C}\) to the successors of \(f_{v}=\imath(v)\) in \(\mathfrak{B}\).

Let \(v\) be an inner node of \(\mathfrak{C}\), say a \(z\)-node, and let \(w_{0}\) and \(w_{1}\) be the 0 - and 1 -successors of \(v\) in \(\mathfrak{C}\). Then, the cofactor \(\left.f_{v}\right|_{z=0}\) agrees with \(f_{w_{0}}\), and similarly, \(f_{w_{1}}=\left.f_{v}\right|_{z=1}\). (This holds in any OBDD.) As \(\mathfrak{C}\) is reduced, \(f_{v}\) is nonconstant (since otherwise \(f_{v}=f_{w_{0}}=f_{w_{1}}\) ). Variable \(z\) must be the first essential variable of \(f_{v}\) according to \(<_{\wp}\), i.e., \(z=\operatorname{var}\left(f_{v}\right)\). Let us see why. Let \(y=\operatorname{var}\left(f_{v}\right)\). The assumption \(z<_{\wp} y\) yields that \(z\) is not essential for \(f_{v}\), and therefore \(f_{w_{0}}=\left.f_{v}\right|_{z=0}=f=\left.f_{v}\right|_{z=1}=f_{w_{1}}\). But then \(w_{0}\), \(w_{1}\) and \(v\) represent the same function. Since \(w_{0} \neq v\) and \(w_{1} \neq v\), this contradicts the assumption that \(\mathfrak{C}\) is reduced. The assumption \(y<_{\wp} z\) is also impossible since then no \(y\)-node would appear in the sub-OBDD \(\mathfrak{C}_{v}\), which is impossible as \(y=\operatorname{var}\left(f_{v}\right)\) is essential for \(f_{v}\) by definition.

But then \(\operatorname{var}(\imath(v))=z=\operatorname{var}^{\mathscr{C}}(v)\) and, for \(b \in\{0,1\}\) :
\[
\operatorname{succ}_{b}(\imath(v))=\left.f_{v}\right|_{z=b}=f_{\text {succ }_{b}^{\mathfrak{c}}(v)}=\imath\left(\operatorname{succ}_{b}^{\mathfrak{c}}(v)\right)
\]

Hence, \(\imath\) is an isomorphism.

Theorem 6.70 permits speaking about "the \(\wp\)-ROBDD" for a given switching function \(f\) for Var. The \(\wp\)-ROBDD-size of \(f\) denotes the size (i.e., number of nodes) in the \(\wp\)-ROBDD for \(f\).

\section*{Corollary 6.71. Minimality of Reduced \(O B D D s\)}

Let \(\mathfrak{B}\) be a \(\wp-O B D D\) for \(f\). Then, \(\mathfrak{B}\) is reduced if and only if size \((\mathfrak{B}) \leqslant \operatorname{size}(\mathfrak{C})\) for each \(\wp-O B D D \mathfrak{C}\) for \(f\).

Proof: This follows from the facts that (i) each \(\wp\)-consistent confactor of \(f\) is represented in any \(\wp-\) OBDD \(\mathfrak{C}\) for \(f\) by at least one node, and (ii) a \(\wp-O B D D \mathfrak{B}\) for \(f\) is reduced if and only if the nodes of \(\mathfrak{B}\) stand in one-to-one-correspondence to the \(\wp\)-consistent cofactors of \(f\).

Reduction rules. When the variable ordering \(\wp\) for Var is fixed, then reduced \(\wp\)-OBDDs provide unique representations of switching functions for Var. (Of course, uniqueness is up to isomorphism.) Although reducedness is a global condition for an OBDD, there are two simple local reduction rules (see Figure 6.23), which can successively be applied to transform a given nonreduced \(\wp-\mathrm{OBDD}\) into an equivalent \(\wp-\) ROBDD.

Elimination rule: If \(v\) is an inner node of \(\mathfrak{B}\) with \(\operatorname{succ}_{0}(v)=\operatorname{succ}_{1}(v)=w\), then remove \(v\) and redirect all incoming edges \(u \rightarrow v\) to \(w\).

Isomorphism rule: If \(v, w\) are nodes in \(\mathfrak{B}\) with \(v \neq w\) and either \(v, w\) are drains with \(\operatorname{val}(v)=\operatorname{val}(w)\) or \(v, w\) are inner nodes with
\[
\left\langle\operatorname{var}(v), \operatorname{succ}_{1}(v), \operatorname{succ}_{0}(v)\right\rangle=\left\langle\operatorname{var}(w), \operatorname{succ}_{1}(w), \operatorname{succ}_{0}(w)\right\rangle
\]
then remove node \(v\) and redirect all incoming edges \(u \rightarrow v\) to node \(w\).

Both rules delete node \(v\). The redirection of the incoming edges \(u \rightarrow v\) to node \(w\) means the replacement of the edges \(u \rightarrow v\) which \(u \rightarrow w\). Formally, this means that we deal with the modified successor functions given by
\[
\operatorname{succ}_{b}^{\prime}(u)= \begin{cases}\operatorname{succ}_{b}(u) & \text { if } \operatorname{succ}_{b}(u) \neq v \\ w & \text { if } \operatorname{succ}_{b}(u)=v\end{cases}
\]
for \(b=0,1\). The transformations described in Examples 6.60 and 6.61 rely on the application of the isomorphism and elimination rule.

becomes

becomes

becomes


Figure 6.23: Reduction rules for OBDDs.

Both reduction rules are sound in the sense that they do not affect the semantics, i.e., if \(\mathfrak{C}\) arises from a \(\wp-O B D D \mathfrak{B}\) by applying the elimination or isomorphism rule, then \(\mathfrak{C}\) is again a \(\wp-\mathrm{OBDD}\) and \(f_{\mathfrak{B}}=f_{\mathfrak{C}}\). This is due to the fact that both rules simply collapse two nodes \(v\) and \(w\) with \(f_{v}=f_{w}\). For the elimination rule applied to a \(z\)-node \(v\) with \(w=\operatorname{succ}_{0}(v)=\operatorname{succ}_{1}(v)\), we have
\[
f_{v}=\left(\neg z \wedge f_{\text {Succ }_{0}(v)}\right) \wedge\left(z \wedge f_{\text {Succ }_{1}(v)}\right)=\left(\neg z \wedge f_{w}\right) \wedge\left(z \wedge f_{w}\right)=f_{w}
\]

Similarly, if the isomorphism rule is applied to \(z\)-nodes \(v\) and \(w\) then
\[
f_{v}=\left(\neg z \wedge f_{\text {Succ }_{0}(v)}\right) \wedge\left(z \wedge f_{\text {Succ }_{1}(v)}\right)=\left(\neg z \wedge f_{\text {Succ }_{0}(w)}\right) \wedge\left(z \wedge f_{\text {Succ }_{1}(w)}\right)=f_{w}
\]

Since the application of the reduction rules decreases the number of nodes, the process to generate an equivalent \(\wp\)-OBDD by applying the reduction rules as long as possible always terminates. In fact, the resulting OBDD is reduced:

\section*{Theorem 6.72. Completeness of Reduction Rules}
\(\wp-O B D D \mathfrak{B}\) is reduced if and only if no reduction rule is applicable to \(\mathfrak{B}\).

Proof: \(\Rightarrow\) : The applicability of a reduction rule implies the existence of at least two nodes representing the same switching function. Thus, if \(\mathfrak{B}\) is reduced, then no reduction rule is applicable.
\(\Leftarrow\) : We prove the other direction by induction on the number of variables. More precisely, suppose that we are given a \(\wp\)-OBDD \(\mathfrak{B}\) for the variable ordering \(\wp=\left(z_{1}, \ldots, z_{m}\right)\) such that neither the elimination nor the isomorphism rule is applicable and show by induction on \(i\) that
\[
f_{v} \neq f_{w} \text { for all nodes } v, w \in V_{i} \text { where } v \neq w
\]

Here, \(V_{i}\) denotes the set of all nodes \(v \in V\) on level \(i\) or lower. Formally, \(V_{i}\) is the set of all nodes \(v\) in \(\mathfrak{B}\) such that \(z_{i} \leqslant_{\wp} \operatorname{var}(v)\). Recall that \(\operatorname{var}(v)=\perp\) (undefined) for each drain \(v\) and that \(z<_{\wp} \perp\) for all variables \(z\).

We start the induction with the bottom level \(i=m+1\). The statement \(f_{v} \neq f_{w}\) for all drains \(v, w\) where \(v \neq w\) is trivial since the nonapplicability of the isomorphism rule yields that there is at most one drain with value 0 and at most one drain with value 1 . In the induction step \(i+1 \Longrightarrow i(m \geqslant i \geqslant 0)\) we suppose that \(f_{v} \neq f_{w}\) for all nodes \(v, w \in V_{i+1}\) with \(v \neq w\) (induction hypothesis). Suppose there are two nodes \(v, w \in V_{i}\) with \(v \neq w\) and \(f_{v}=f_{w}\). At least one of the nodes \(v\) or \(w\) must be on level \(i\). Say \(v \in V_{i} \backslash V_{i+1}\). Then, \(\operatorname{var}(v)=z_{i}\).

Let us first suppose that \(w \in V_{i+1}\). Then, either \(w\) is a drain or a \(z_{j}\)-node for some \(j>i\). In either case, variable \(z_{i}\) is not essential for \(f_{v}=f_{w}\). As \(v\) is a \(z_{i}\)-node this yields that \(f_{v}\) agrees with the switching functions \(f_{v_{0}}\) and \(f_{v_{1}}\) of its successors \(v_{0}=\operatorname{succ}_{0}(v)\) and \(v_{1}=\operatorname{succ}_{1}(v)\). But then \(v_{0}, v_{1} \in V_{i+1}\) and \(f_{v_{0}}=f_{v_{1}}\). The induction hypothesis yields that \(v_{0}=v_{1}\). But then the elimination rule would be applicable. Contradiction.

Suppose now that \(w\) is a \(z_{i}\)-node too. Let \(v_{0}=\operatorname{succ}_{0}(v), v_{1}=\operatorname{succ}_{1}(v)\) and \(w_{0}=\operatorname{succ}_{0}(w)\), \(w_{1}=\operatorname{succ}_{1}(w)\). The assumption \(f_{v}=f_{w}\) yields that
\[
f_{v_{0}}=\left.f_{v}\right|_{z_{i}=0}=\left.f_{w}\right|_{z_{i}=0}=f_{w_{0}}
\]
and \(f_{v_{1}}=f_{w_{1}}\). As \(v_{0}, v_{1}, w_{0}, w_{1} \in V_{i+1}\) the induction hypothesis yields that \(v_{0}=w_{0}\) and \(v_{1}=w_{1}\). But then the ismorphism rule is applicable. Contradiction.

Theorem 6.70 suggests a reduction algorithm which takes as input a nonreduced \(\wp\)-OBDD \(\mathfrak{B}\) and constructs an equivalent \(\wp\)-OBDD by applying the reduction rules as long as possible. According to the inductive proof of the completeness of the reduction rules in Theorem 6.72, this technique is complete if the candidate nodes for the reduction rules are considered in a bottom-up manner. That is, initially we identify all drains with the same value. Then, for the levels \(m, m-1, \ldots, 1\) (in this order) we apply the elimination and isomorphism rule. At level \(i\), we first remove all nodes with identical successors (elimination rule) and then check the pairs of \(z_{i}\)-nodes where the isomorphism rule is applicable. To support the isomorphism rule a bucket technique can be used that groups together (1) all \(z_{i}\)-nodes with the same 0 -successor and (2) splits all buckets consisting of \(z_{i}\)-nodes with the same 0 -successor into buckets consisting of \(z_{i}\)-nodes with exactly the same successors. Then, application of the isomorphism rule simply means that the nodes in the buckets resulting from step (2) have to be collapsed into a single node. The time complexity of this algorithm is in \(\mathcal{O}(\operatorname{size}(\mathfrak{B}))\). In particular, given two \(\wp-\mathrm{OBDDs} \mathfrak{B}\) and \(\mathfrak{C}\), the equivalence problem "Does \(f_{\mathfrak{B}}=f_{\mathfrak{C}}\) hold?" can be solved by applying the reduction algorithm to both and then checking isomorphism for the reduced \(\wp\)-ROBDDs (see Exercise 6.12 on page 436). We will see later that with tricky implementation techniques, which integrate the steps of the reduction algorithm in the synthesis algorithms for ROBDDs and thus ensure that at any time the decision graph is reduced, the equivalence problem for ROBDDs can even be solved in constant time.

The Variable Ordering Problem The result stating the canonicity of reduced OBDDs crucially depends on the fact that the variable ordering \(\wp\) is assumed to be fixed. Varying the variable ordering can lead to totally different ROBDDs, possibly ranging from ROBDDs of linear size to ROBDDs of exponential size. The results established before yield that the size (i.e., number of nodes) in the \(\wp\)-ROBDD for a switching function \(f\) agrees


Figure 6.24: ROBDD for the function \(f_{3}=\left(z_{1} \wedge y_{1}\right) \vee\left(z_{2} \wedge y_{2}\right) \vee\left(z_{3} \wedge y_{3}\right)\) for the variable ordering \(\wp=\left(z_{1}, y_{1}, z_{2}, y_{2}, z_{3}, y_{3}\right)\).
with the number of \(\wp\)-consistent cofactors of \(f\). Thus, reasoning about the memory requirements of ROBDD-based approaches relies on counting the number of order-consistent cofactors.

\section*{Example 6.73. A Function with Small and Exponential-Size ROBDDs}

To illustrate how the size of ROBDDs can be determined by analyzing the cofactors we consider a simple switching function which has both ROBDDs of linear size and ROBDDs with exponentially many nodes. Let \(m \geqslant 1\) and
\[
f_{m}=\left(z_{1} \wedge y_{1}\right) \vee\left(z_{2} \wedge y_{2}\right) \vee \ldots \vee\left(z_{m} \wedge y_{m}\right) .
\]

For the variable ordering \(\wp=\left(z_{m}, y_{m}, z_{m-1}, y_{m-1}, \ldots, z_{1}, y_{1}\right)\), the \(\wp\)-ROBDD for \(f_{m}\) has \(2 m+2\) nodes, while \(\Omega\left(2^{m}\right)\) nodes are required for the ordering \(\wp^{\prime}=\left(z_{1}, z_{2}, \ldots, z_{m}\right.\), \(\left.y_{1}, \ldots, y_{m}\right)\). Figures 6.25 and 6.24 show the ROBDDs for the linear-size and exponentialsize variable orderings for \(m=3\). We first consider the ordering \(\wp\) which groups the variables \(z_{i}\) and \(y_{i}\) that appear in the same clause. In fact, the variable ordering \(\wp\) is optimal for \(f_{m}\) as the \(\wp\)-ROBDD for \(f_{m}\) contains one node for each variable. (This is the best we can hope to have as all \(2 n\) variables are essential for \(f_{m}\) and must appear in any ROBDD for


Figure 6.25: ROBDD for the function \(f_{3}=\left(z_{1} \wedge y_{1}\right) \vee\left(z_{2} \wedge y_{2}\right) \vee\left(z_{3} \wedge y_{3}\right)\) for the variable ordering \(\wp=\left(z_{1}, z_{2}, z_{3}, y_{1}, y_{2}, y_{3}\right)\).
\(f_{m}\).) Note that for \(1 \leqslant i \leqslant m\) :
\[
\begin{aligned}
& \left.f_{m}\right|_{z_{m}=a_{m}, z_{m}=b_{m}, \ldots, z_{i}=a_{i}, z_{i}=b_{i}}= \begin{cases}1 & \text { if } a_{j}=b_{j}=1 \\
f_{i-1} & \text { for some } j \in\{i, \ldots, m\} \\
\text { otherwise }\end{cases} \\
& \left.f_{m}\right|_{z_{m}=a_{m}, z_{m}=b_{m}, \ldots, z_{i+1}=a_{i+1}, z_{i+1}=b_{i+1}, z_{i}=a_{i} \in\left\{1, y_{i} \vee f_{i-1}\right\}}
\end{aligned}
\]
where \(f_{0}=0\). Hence, the \(\wp\)-ROBDD for \(f_{m}\) has exactly one \(z_{i}\)-node representing the function \(f_{i}\), exactly one \(y_{i}\)-node for the function \(y_{i} \vee f_{i-1}\) (for \(1 \leqslant i \leqslant m\) ), and two drains.

To see why the variable ordering \(\wp^{\prime}\) leads to a \(\wp^{\prime}\) - ROBDD of exponential size, we consider the \(\wp^{\prime}\)-consistent cofactors
\[
f_{\bar{b}}=\left.f_{m}\right|_{z_{1}=b_{1}, \ldots, z_{m}=b_{m}}=\bigvee_{i \in I_{\bar{b}}} y_{i}
\]
where \(\bar{b}=\left(b_{1}, \ldots, b_{n}\right)\) and \(I_{\bar{b}}=\left\{i \in\{1, \ldots, m\} \mid b_{i}=1\right.\). The set of essential variables of \(f_{\bar{b}}\) is \(\left\{y_{i} \mid i \in I_{\bar{b}}\right\}\). As \(\mathrm{f} \bar{b}, \overline{\mathfrak{c}} \in\{0,1\}^{m}, \bar{b} \neq \overline{\mathfrak{c}}\) the index sets \(I_{\bar{b}}\) and \(I_{\bar{c}}\) are different, the \(f_{\bar{b}}\) and \(f_{\bar{c}}\) have different essential variables. Therefore, \(f_{\bar{b}} \neq f_{\bar{c}}\) if \(\bar{b} \neq \overline{\mathfrak{c}}\). But then the number of \(\wp^{\prime}\)-consistent cofactors is at least \(2^{m}\). This yields that the \(\wp^{\prime}\)-ROBDD for \(f_{m}\) has at most \(2^{m}\) nodes.

Since for many switching functions the ROBDD sizes for different variable ordering can vary enormously, the efficiency of BDD-based computations crucially relies on the use of techniques that improve a given variable ordering. Although the problem to find an optimal variable ordering is known to be computationally hard (already the problem to decide whether a given variable ordering is optimal is NP-hard [56, 386]), there are several efficient heuristics for improving the current ordering. Most promiment is the so-called sifting algorithm [358] which relies on a local search for the best position for each variable, when the ordering of the other variables is supposed to be fixed. Explanations on such variable reordering algorithms and further discussions on the variable ordering problem are beyond the scope of this monograph. We refer the interested reader to the textbooks [292, 418] and conclude the comments on the influence of the variable ordering by the remark that there are also types of switching functions with ROBDDs of polynomial size under all variable orderings and types of switching functions where any variable ordering leads to a ROBDD of exponential size. An example of the latter is the middle bit of the multiplication function [71]. Examples of switching functions where each variable ordering leads to a ROBDD of at most quadratic size are symmetric functions. These are switching functions where the function values just depend on the number of variables that are assigned to 1 . Stated differently, \(f \in \operatorname{Eval}\left(z_{1}, \ldots, z_{m}\right)\) is symmetric if and only if
\[
f\left(\left[z_{1}=b_{1}, \ldots, z_{m}=b_{m}\right]\right)=f\left(\left[z_{1}=b_{i_{1}}, \ldots, z_{m}=b_{i_{m}}\right]\right)
\]
for each permutation \(\left(i_{1}, \ldots, i_{m}\right)\) of \((1, \ldots, m)\). Examples of symmetric functions for \(\operatorname{Var}=\left\{z_{1}, \ldots, z_{m}\right\}\) are \(z_{1} \vee z_{2} \vee \ldots \vee z_{m}, z_{1} \wedge z_{2} \wedge \ldots \wedge z_{m}\), the parity function \(z_{1} \oplus z_{2} \oplus \ldots \oplus z_{m}\) (which returns 1 iff the number of variables that are assigned to 1 is odd), and the majority function (which returns 1 iff the number of variables that are assigned to 1 is greater than the number of variables that are assigned to 0 ). The ROBDDs for symmetric functions have the same topological structure for all variable orderings. This follows from the fact that the \(\wp\)-ROBDD for a symmetric function can be transformed into the \(\wp^{\prime}\)-ROBDD by just modifying the variable labeling function.

\section*{Lemma 6.74. ROBDD-Size for Symmetric Functions}

If \(f\) is a symmetric function with \(m\) essential variables, then for each variable ordering \(\wp\) the \(\wp-R O B D D\) has size \(\mathcal{O}\left(m^{2}\right)\).

Proof: Given a symmetric function \(f\) for \(m\) variables and a variable ordering \(\wp\), say \(\wp=\left(z_{1}, \ldots, z_{m}\right)\), then the \(\wp\)-consistent cofactors \(\left.f\right|_{z_{1}=b_{1}, \ldots, z_{i}=b_{i}}\) and \(\left.f\right|_{z_{1}=\mathfrak{c}_{1}, \ldots, z_{i}=\mathfrak{c}_{i}}\) agree for all bit tuples \(\left(b_{1}, \ldots, b_{i}\right)\) and \(\left(\mathfrak{c}_{1}, \ldots, \mathfrak{c}_{i}\right)\) that contain the same number of 1 's. Thus, there are at most \(i+1\) different \(\wp\)-consistent cofactors of \(f\) that arise by assigning values to the first \(i\) variables. Hence, the total number of \(\wp\)-consistent cofactors is bounded above by \(\sum_{i=0}^{m}(i+1)=\mathcal{O}\left(m^{2}\right)\).

ROBDDs vs. CNF /DNF Both the parity and the majority function provide examples for switching functions with small ROBDD representations, while any representation of them by conjunctive or disjunctive normal forms (CNF, DNF) requires formulae of exponential length. Vice versa, there are also switching functions with short conjunctive or disjunctive normal forms, while the ROBDD under any variable ordering has exponential length (see, e.g., [418]). In fact, ROBDDs yield a totally different picture for complexity theoretic considerations than CNF or DNF. For example, given a CNF representation for a switching function \(f\), the task to generate a CNF for \(\neg f\) is expensive, as \(f\) might be expressible by a CNF of polynomial length, while any CNF for \(\neg f\) has at least exponentially many clauses. For ROBDDs, however, negation is trivial as we may simply swap the values of the drains. In particular, for any variable ordering \(\wp\), the \(\wp\)-ROBDDs for \(f\) and \(\neg f\) have the same size. For another example, regard the satisfiability problem, which is known to be NP-complete for CNF, but again trivial for ROBDDs, since \(f \neq 0\) if and only if the \(\wp\)-ROBDD for \(f\) does not contain a 0 -drain. Similarly, the question whether two CNFs are equivalent is computationally hard (coNP-complete), but can be solved for \(\wp\)-ROBDDs \(\mathfrak{B}\) and \(\mathfrak{C}\) by checking isomorphism. The latter can be performed by a simultaneous traversal of the \(\wp\)-OBDDs in time linear in the sizes of \(\mathfrak{B}\) and \(\mathfrak{C}\). See Exercise 6.12, page 436. Note that these results do not contradict the complexity theoretic lower bounds, since "linear time" for a ROBDD-based algorithm means linear in the size of the input ROBDDs, which could be exponentially larger than equivalent input formulae (e.g., CNF).

\subsection*{6.7.4 Implementation of ROBDD-Based Algorithms}

The efficiency of ROBDD-based algorithms to manipulate switching functions crucially relies on appropriate implementation techniques. In fact, with tricky implementation techniques, equivalence checking for ROBDDs can even be realized in constant time. In the sequel, we will explain the main ideas of such techniques, which provide the basis for most BDD packages and serve as a platform for an efficient realization of synthesis algorithms on ROBDDs. The purpose of synthesis algorithms is to construct a \(\wp\)-ROBDD for a function \(f_{1}\) op \(f_{2}\) (where op is a Boolean connective such as disjunction, conjunction, implication, etc.), when \(\wp\)-ROBDDs for \(f_{1}\) and \(f_{2}\) are given. Recall that the symbolic realization of the CTL model-checking procedure relies on such synthesis operations.

The idea, originally proposed in [301], is to use a single reduced decision graph with one global variable ordering \(\wp\) to represent several switching functions, rather than using separate \(\wp\)-ROBDDs for each of the switching functions. All computations on these decision graphs are interleaved with the reduction rules to guarantee redundance-freedom at any time. Thus, the comparison of two represented functions simply requires checking equality of the nodes for them, rather than analyzing their sub-OBDDs.


Figure 6.26: Example of a shared OBDD.

We start with the formal definition of a shared \(\wp\)-OBDD which is the same as a \(\wp-\operatorname{ROBDD}\), the only difference being that we can have more than more root node.

\section*{Definition 6.75. Shared OBDD}

Let Var be a finite set of Boolean variables and \(\wp\) a variable ordering for Var. A shared \(\wp-\mathrm{OBDD}\left(\wp-\mathrm{SOBDD}\right.\) for short) is a tuple \(\overline{\mathfrak{B}}=\left(V, V_{I}, V_{T}\right.\), \(\operatorname{succ}_{0}\), succ \(_{1}\), var, val, \(\left.\bar{v}_{0}\right)\) where \(V, V_{I}, V_{T}\), succ \(_{0}\), succ \(_{1}\), var, and val are as in \(\wp\)-OBDDs (see Definition 6.63 on page 395). The last component is a tuple \(\bar{v}_{0}=\left(v_{0}^{1}, \ldots, v_{0}^{k}\right)\) of roots. The requirements are as in \(\wp\)-ROBDDs, i.e., for all nodes \(v, w \in V\), (1) \(\operatorname{var}(v)<_{\wp} \operatorname{var}\left(\operatorname{succ}_{b}(v)\right)\) if \(v \in V_{I}\) and \(b \in\{0,1\}\) and (2) \(v \neq w\) implies \(f_{v} \neq f_{w}\), where the switching function \(f_{v}\) for the nodes \(v \in V\) is defined as for OBDDs.

Figure 6.26 shows a shared OBDD with four root nodes that represent the functions \(z_{1} \wedge \neg z_{2}, \neg z_{2}, z_{1} \oplus z_{2}\) and \(\neg z_{1} \vee z_{2}\).

If \(v\) is a node in a \(\wp-\operatorname{SOBDD} \overline{\mathfrak{B}}\), then the sub-OBDD \(\overline{\mathfrak{B}}_{v}\) is the \(\wp-\operatorname{ROBDD}\) that results from \(\overline{\mathfrak{B}}\) by removing all nodes that are not reachable from \(v\) and declaring \(v\) to be the root node. In fact, \(\overline{\mathfrak{B}}_{v}\) is the \(\wp\)-ROBDD for \(f_{v}\), and thus, the size \(N_{v}\) of \(\overline{\mathfrak{B}}_{v}\) is with the \(\wp\) - \(\operatorname{ROBDD}\) size of \(f_{v}\). Thus, an alternative view of an SOBDD is the combination of several ROBDDs for the same variable ordering \(\wp\) by sharing nodes for common \(\wp\)-consistent cofactors. In particular, an SOBDD has exactly two drains for the constant functions 0 and 1 (where we ignore the pathological case of an SOBDD with a single root node representing a constant function). Thus, if \(f_{1}, \ldots, f_{k}\) are the functions for the root nodes \(v_{0}^{1}, \ldots, v_{0}^{k}\) of \(\overline{\mathfrak{B}}\), then the size (i.e., total number of nodes) of \(\overline{\mathfrak{B}}\) is often smaller than but at most \(N_{f_{1}}+\ldots+N_{f_{k}}\) where \(N_{f}\) denotes the \(\wp\)-ROBDD size of \(f\).

For the symbolic representation of a transition system by means of switching functions \(\Delta\left(\bar{x}, \bar{x}^{\prime}\right)\) for the transition relation and switching functions \(f_{a}(\bar{x}), a \in A P\), for the satisfaction sets of the atomic propositions (see page 386), one might use a shared OBDD with
root nodes for \(\Delta\) and the \(f_{a}\) 's. As we mentioned before, the chosen variable ordering \(\wp\) can be crucial for the size of the SOBDD representing a transition system. Experimental studies have shown that typically good variable orderings are obtained when grouping together the unprimed variables \(x_{i}\) and their copies \(x_{i}^{\prime}\). Later we will give some formal arguments why such interleaved variable orderings, like \(\wp=\left(x_{1}, x_{1}^{\prime}, \ldots, x_{n}, x_{n}^{\prime}\right)\), are advantageous.

To perform the CTL model-checking procedure in a symbolic way, the shared OBDD \(\overline{\mathfrak{B}}\) with root nodes for \(\Delta\) and the \(f_{a}\) 's has to be extended by new root nodes representing the characteristic functions of the satisfaction sets \(\operatorname{Sat}(\Psi)\) for the state subformulae \(\Psi\) of the CTL formula \(\Phi\) to be checked. For instance, if \(\Phi=a \wedge \neg b\) with atomic propositions \(a, b\), then we first have to insert a root node for the characteristic function \(f_{\neg b}=\neg f_{b}\) for \(\operatorname{Sat}(\neg b)\) and then a root node for the switching function \(f_{a} \wedge f_{\neg b}\). The treatment of formulae of the form, e.g., \(\exists \diamond \Psi\) or \(\exists \square \Psi\) by means of Algorithms 20 or 21 , requires creating additional root nodes for the functions \(f_{i}\) representing the current approximations of satisfaction sets. Of course, adding a new root node for some switching function \(f\) also means that we have to add nodes for all order-consistent cofactors of \(f\) that are not yet represented by a node in \(\overline{\mathfrak{B}}\).

To support such dynamic changes of the set of switching functions to be represented, the realization of shared OBDDs typically relies on the use of two tables: the unique table, which contains the relevant information about the nodes and serves to keep the diagram reduced during the execution of synthesis algorithms, and a computed table, which is needed for efficiency reasons. Let us first explain the ideas behind the use of the unique table. The implementation of synthesis algorithms and the use of the computed table will be explained later.

The Unique Table The entries of the unique table are triples of the form
\[
\operatorname{info}(v)=\left\langle\operatorname{var}(v), \operatorname{succ}_{0}(v), \operatorname{succ}_{0}(v)\right\rangle
\]
for each inner node \(v\). Note that these info-triples contain the relevant information which is necessary for the applicability of the isomorphism rule. Accessing the unique table is supported by a find_or_add-operation which takes as argument a triple \(\left\langle z, v_{1}, v_{0}\right\rangle\) consisting of a variable \(z\) and two nodes \(v_{1}\) and \(v_{0}\) with \(v_{1} \neq v_{0}\). The task of the find_or_addoperation is to check whether there exists a node \(v\) in the shared OBDD \(\overline{\mathfrak{B}}\) such that \(\operatorname{info}(v)=\left\langle z, v_{1}, v_{0}\right\rangle\). If so, then it returns node \(v\), otherwise it creates a new \(z\)-node \(v\) with 1 -successor \(v_{1}\) and 0 -successor \(v_{0}\), and makes a corresponding entry in the unique table. Thus, the find_or_add operation can be viewed as the SOBDD realization of the isomorphism rule. In most BDD packages, the unique table is organized using appropriate hashing techniques. We skip such details here and assume constant expected time to access the into-triple for any node, and to perform the find_or_add-operation.

Boolean Operators Let us now consider how synthesis algorithms can be realized on SOBDDs, using the unique table. An elegant, but also very efficient way is to support a ternary operator, called ITE for "if-then-else", that covers all Boolean connectives. The ITE operator takes as arguments three switching functions \(g, f_{1}, f_{2}\) and composes them according to "if \(g\) then \(f_{1}\) else \(f_{2}\) ". Formally:
\[
\operatorname{ITE}\left(g, f_{1}, f_{2}\right)=\left(g \wedge f_{1}\right) \vee\left(\neg g \wedge f_{2}\right)
\]

For the special case where \(g\) is constant we have \(\operatorname{ITE}\left(0, f_{1}, f_{2}\right)=f_{2}\) and \(\operatorname{ITE}\left(1, f_{1}\right.\), \(\left.f_{2}\right)=f_{1}\). The ITE operator fits very well with the representation of the SOBDD nodes in the unique table by their info-triples as we have:
\[
f_{v}=\operatorname{ITE}\left(z, f_{\text {Succ }_{1}(v)}, f_{\text {Succ }_{0}(v)}\right)
\]

The negation operator is obtained by \(\neg f=\operatorname{ITE}(f, 0,1)\). Also all other Boolean connectives can be expressed by the ITE-operator. For example:
\[
\begin{aligned}
f_{1} \vee f_{2} & =\operatorname{ITE}\left(f_{1}, 1, f_{2}\right) \\
f_{1} \wedge f_{2} & =\operatorname{ITE}\left(f_{1}, f_{2}, 0\right) \\
f_{1} \oplus f_{2} & =\operatorname{ITE}\left(f_{1}, \neg f_{2}, f_{2}\right)=\operatorname{ITE}\left(f_{1}, \operatorname{ITE}\left(f_{2}, 0,1\right), f_{2}\right) \\
f_{1} \rightarrow f_{2} & =\operatorname{ITE}\left(f_{1}, f_{2}, 1\right)
\end{aligned}
\]

The realization of the ITE operator on an \(\operatorname{SOBDD} \overline{\mathfrak{B}}\) requires a procedure that takes as input three nodes \(u, v_{1}, v_{2}\) of \(\overline{\mathfrak{B}}\) and returns a possibly new node \(w\) such that \(f_{w}=\) \(\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\), by reusing existing nodes whenever possible and adding new nodes to \(\overline{\mathfrak{B}}\) if necessary. For this, the sub-OBDDs for the input nodes \(u, v_{1}\), and \(v_{2}\) are traversed simultaneously in a top-down fashion, while the synthesis of the sub-ROBDD for \(w\) (and generation of new nodes) is performed in a bottom-up manner. This method relies on the following observation.

\section*{Lemma 6.76. Cofactors of \(\operatorname{ITE}(\cdot)\)}

If \(g, f_{1}, f_{2}\) are switching functions for Var, \(z \in \operatorname{Var}\) and \(b \in\{0,1\}\), then
\[
\left.\operatorname{ITE}\left(g, f_{1}, f_{2}\right)\right|_{z=b}=\operatorname{ITE}\left(\left.g\right|_{z=b},\left.f_{1}\right|_{z=b},\left.f_{2}\right|_{z=b}\right)
\]

Proof: For simplicity, let us assume that \(g, f_{1}, f_{2}\) are switching functions for the same variable set \(\operatorname{Var}=\left\{z, y_{1}, \ldots, y_{m}\right\}\). This, in fact, is no proper restriction as we may simply take the union of the variable sets of \(g, f_{1}\) and \(f_{2}\) and regard all three functions as switching functions for the resulting variable sets. Let \((a, \bar{c})\) be a short-form notation for
the evaluation \([z=a, \bar{y}=\overline{\boldsymbol{c}}] \in \operatorname{Eval}(\operatorname{Var})\). Then, we have
\[
\begin{aligned}
& \left.\operatorname{ITE}\left(g, f_{1}, f_{2}\right)\right|_{z=b}(a, \overline{\mathfrak{c}}) \\
= & \operatorname{ITE}\left(g, f_{1}, f_{2}\right)(b, \overline{\mathfrak{c}}) \\
= & \left(g(b, \overline{\mathfrak{c}}) \wedge f_{1}(b, \overline{\mathfrak{c}})\right) \vee\left(\neg g(b, \overline{\mathfrak{c}}) \wedge f_{2}(b, \overline{\mathfrak{c}})\right) \\
= & \left(\left.\left.g\right|_{z=b}(a, \overline{\mathfrak{c}}) \wedge f_{1}\right|_{z=b}(a, \overline{\mathfrak{c}})\right) \vee\left(\left.\left.\neg g\right|_{z=b}(a, \overline{\mathfrak{c}}) \wedge f_{2}\right|_{z=b}(a, \overline{\mathfrak{c}})\right) \\
= & \operatorname{ITE}\left(\left.g\right|_{z=b},\left.f_{1}\right|_{z=b},\left.f_{2}\right|_{z=b}\right)(a, \overline{\mathfrak{c}}) .
\end{aligned}
\]

Thus, a node in a \(\wp-\operatorname{SOBDD}\) for representing \(\operatorname{ITE}\left(g, f_{1}, f_{2}\right)\) is a node \(w\) such that \(\operatorname{info}(w)=\) \(\left\langle z, w_{1}, w_{0}\right\rangle\) where
- \(z\) is the minimal essential variable of \(\operatorname{ITE}\left(g, f_{1}, f_{2}\right)\) according to \(<_{\wp}\),
- \(w_{1}, w_{0}\) are SOBDD nodes with:
\[
f_{w_{1}}=\operatorname{ITE}\left(\left.g\right|_{z=1},\left.f_{1}\right|_{z=1},\left.f_{2}\right|_{z=1}\right) \quad \text { and } \quad f_{w_{0}}=\operatorname{ITE}\left(\left.g\right|_{z=0},\left.f_{1}\right|_{z=0},\left.f_{2}\right|_{z=0}\right)
\]

This observation suggests a recursive algorithm which determines \(z\) and then recursively computes the nodes for ITE applied to the cofactors of \(g=f_{u}, f_{1}=f_{v_{1}}, f_{2}=f_{v_{2}}\) for variable \(z\). Since the explicit computation of \(z\) can be hard, we use the decomposition into cofactors for the minimal variable \(z\) that is essential for \(f_{u}, f_{v_{1}}\) or \(f_{v_{2}}\) :
\[
z=\min \left\{\operatorname{var}(u), \operatorname{var}\left(v_{1}\right), \operatorname{var}\left(v_{2}\right)\right\}
\]
where the minimum is taken according to the total order \(<_{\wp}\) on \(\operatorname{Var} \cup\{\perp\}\). (Recall that we put \(\operatorname{var}(v)=\perp\) for any drain \(v\) and that \(x<_{\wp} \perp\) for all \(x \in \operatorname{Var}\).) If \(z^{\prime}\) is the first essential variable of \(\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\), then \(z \leqslant_{\wp} z^{\prime}\), since no variable \(y<_{\wp} z\) appears in the sub-OBDDs of nodes \(u, v_{1}, v_{2}\), and hence, no such variable \(y\) can be essential for \(\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\). The case \(z<_{\wp} z^{\prime}\) is possible, if accidentally the cofactors \(\left.\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\right|_{z=0}\) and \(\left.\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\right|_{z=1}\) agree. In this case, however, we are in the situation of the elimination rule and the ITE algorithm returns the node representing \(\left.\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\right|_{z=0}\). Otherwise, i.e., if the nodes \(w_{0}\) and \(w_{1}\) that have been recursively determined for \(\left.\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\right|_{z=0}\) and \(\left.\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\right|_{z=1}\), respectively, are different, then \(z^{\prime}=z\) and a node for representing \(\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\) is obtained by the find_or_addoperation applied to the info-triple \(\left\langle z, w_{1}, w_{0}\right\rangle\).

The question remains how to obtain the cofactors \(\left.f_{u}\right|_{z=b},\left.f_{v_{1}}\right|_{z=b}\), and \(\left.f_{v_{2}}\right|_{z=b}\). Nodes that represent these functions are obtained easily since (by choice of variable \(z\) ) nodes \(u, v_{1}\), and \(v_{2}\) are on the \(z\)-level or below, i.e., \(z \leqslant_{\wp} \operatorname{var}(v)\) for \(v \in\left\{u, v_{1}, v_{2}\right\}\). If \(\operatorname{var}(v)=z\), then \(\left.f_{v}\right|_{z=b}\) is represented by the \(b\)-successor of \(v\). If \(z<_{\wp} \operatorname{var}(v)\), then \(z\) is not essential for \(f_{v}\) and we have \(\left.f_{v}\right|_{z=b}=f_{v}\). Thus, if we define
\[
\left.v\right|_{z=b}= \begin{cases}\operatorname{succ}_{b}(v) & \text { if } \operatorname{var}(v)=z \\ u & \text { if } z<_{\wp} \operatorname{var}(v)\end{cases}
\]
then \(\left.v\right|_{z=b}\) is the node in \(\overline{\mathfrak{B}}\) representing \(\left.f_{v}\right|_{z=b}\). Hence, a node representing the function \(\left.\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\right|_{z=b}\) is obtained by a recursive call of the ITE algorithm with the arguments \(\left.u\right|_{z=b},\left.v_{1}\right|_{z=b}\) and \(\left.v_{2}\right|_{z=b}\) (Lemma 6.76). Note that these are already existing nodes in the \(\operatorname{SOBDD} \overline{\mathfrak{B}}\).

The steps to realize the ITE-operator on a shared OBDD by means of a DFS-based traversal of the sub-OBDDs of \(u, v_{1}\), and \(v_{2}\) to determine the relevant cofactors (where recursive calls of the ITE-algorithm are required) have been summarized in Algorithm 22 on page 412 .
```

Algorithm $22 \operatorname{ITE}\left(u, v_{1}, v_{2}\right)$ (first version)
if $u$ is terminal then
if $\operatorname{val}(u)=1$ then
$w:=v_{1} \quad\left(* \operatorname{ITE}\left(1, f_{v_{1}}, f_{v_{2}}\right)=f_{v_{1}}{ }^{*}\right)$
else
$w:=v_{2} \quad\left(* \operatorname{ITE}\left(0, f_{v_{1}}, f_{v_{2}}\right)=f_{v_{2}}{ }^{*}\right)$
fi
else
$z:=\min \left\{\operatorname{var}(u), \operatorname{var}\left(v_{1}\right), \operatorname{var}\left(v_{2}\right)\right\} ;$
$w_{1}:=\operatorname{ITE}\left(\left.u\right|_{z=1},\left.v_{1}\right|_{z=1},\left.v_{2}\right|_{z=1}\right) ;$
$w_{0}:=\operatorname{ITE}\left(\left.u\right|_{z=0},\left.v_{1}\right|_{z=0},\left.v_{2}\right|_{z=0}\right)$;
if $w_{0}=w_{1}$ then
$w:=w_{1} ; \quad \quad(*$ elimination rule *)
else
$w:=$ find_or_add $\left(z, w_{1}, w_{0}\right) ; \quad(*$ isomorphism rule (?) *)
fi
fi
return $w$

```

Before discussing the complexity of the ITE algorithms, we will first study how the size of the SOBDD can change through performing the ITE algorithm. The size of the sub-OBDD for the (possibly new) node \(w\) representing \(\operatorname{ITE}\left(u, v_{1}, v_{2}\right)\) is bounded by \(N_{u} \cdot N_{v_{1}} \cdot N_{v_{2}}\) where \(N_{v}\) denotes the number of the nodes in the sub-OBDD \(\overline{\mathfrak{B}}_{v}\) for node \(v\). This follows from
the fact that each node \(w^{\prime}\) in the generated sub-OBDD for \(\operatorname{ITE}\left(u, v_{1}, v_{2}\right)\) corresponds to one or more triples \(\left(u^{\prime}, v_{1}^{\prime}, v_{2}^{\prime}\right)\), where \(u^{\prime}\) is a node in \(\overline{\mathfrak{B}}_{u}\) and \(v_{i}^{\prime}\) a node in \(\overline{\mathfrak{B}}_{v_{i}}\) Formally:

\section*{Lemma 6.77. ROBDD Size of \(\operatorname{ITE}\left(g, f_{1}, f_{2}\right)\)}

The size of the \(\wp-R O B D D\) for \(\operatorname{ITE}\left(g, f_{1}, f_{2}\right)\) is bounded by \(N_{g} \cdot N_{f_{1}} \cdot N_{f_{2}}\) where \(N_{f}\) denotes the size of the \(\wp-R O B D D\) for \(f\).

Proof: Let \(\wp=\left(z_{1}, \ldots, z_{m}\right)\) and let \(\mathfrak{B}_{f}\) denote the \(\wp\)-ROBDD for \(f\) where the nodes are the \(\wp\)-consistent cofactors of \(f\) (see the proof of part (a) of Theorem 6.70). We write \(V_{f}\) for the node set of \(\mathfrak{B}_{f}\), i.e.,
\[
V_{f}=\left\{\left.f\right|_{z_{1}=b_{1}, \ldots, z_{i}=b_{i}} \mid 0 \leqslant i \leqslant m, b_{1}, \ldots, b_{i} \in\{0,1\}\right\} .
\]

Note that several of the cofactors \(\left.f\right|_{z_{1}=b_{1}, \ldots, z_{i}=b_{i}}\) might agree, and hence, they stand for the same element (node) of \(V_{f}\). By Lemma 6.76, the node set \(V_{\text {ITE }\left(g, f_{1}, f_{2}\right)}\) of the \(\wp\)-ROBDD \(\mathfrak{B}_{\operatorname{ITE}\left(g, f_{1}, f_{2}\right)}\) for \(\operatorname{ITE}\left(g, f_{1}, f_{2}\right)\) agrees with the set of switching functions
\[
\operatorname{ITE}\left(\left.g\right|_{z_{1}=b_{1}, \ldots, z_{i}=b_{i}},\left.f_{1}\right|_{z_{1}=b_{1}, \ldots, z_{i}=b_{i}}, f_{2} z_{1}=b_{1}, \ldots, z_{i}=b_{i}\right)
\]
where \(0 \leqslant i \leqslant m, b_{1}, \ldots, b_{i} \in\{0,1\}\). Thus, the function
that maps any triple \(\left(g^{\prime}, f_{1}^{\prime}, f_{2}^{\prime}\right)\) where \(g^{\prime}\) is a node in \(\mathfrak{B}_{g}\) (i.e., a \(\wp\)-consistent cofactor of \(g\) ) and \(f_{i}^{\prime}\) a node in \(\mathfrak{B}_{f_{i}}\) (i.e., a \(\wp\)-consistent cofactor of \(f_{i}\) ) to the node \(\operatorname{ITE}\left(g^{\prime}, f_{1}^{\prime}, f_{2}^{\prime}\right)\) of \(\mathfrak{B}_{\operatorname{ITE}\left(g, f_{1}, f_{2}\right)}\) yields a surjective mapping from \(V_{g} \times V_{f_{1}} \times V_{f_{2}}\) to some superset of \(V_{\text {ITE }}\left(g, f_{1}, f_{2}\right)\). Hence:
\[
N_{\operatorname{ITE}\left(g, f_{1}, f_{2}\right)}=\left|V_{\operatorname{ITE}\left(g, f_{1}, f_{2}\right)}\right| \leqslant\left|V_{g} \times V_{f_{1}} \times V_{f_{2}}\right|=N_{g} \cdot N_{f_{1}} \cdot N_{f_{2}}
\]

Observe that only the triples \(\left(g^{\prime}, f_{1}^{\prime}, f_{2}^{\prime}\right) \in V_{g} \times V_{f_{1}} \times V_{f_{2}}\) where \(g^{\prime}, f_{1}^{\prime}\), \(f_{2}^{\prime}\) arise from \(g, f_{1}, f_{2}\) by the same evaluation \(\left[z_{1}=b_{1}, \ldots, z_{i}=b_{i}\right]\) are mapped via \(\imath\) to nodes of \(\mathfrak{B}_{\text {ITE }\left(g, f_{1}, f_{2}\right)}\). Furthermore, \(\operatorname{ITE}\left(g^{\prime}, f_{1}^{\prime}, f_{2}^{\prime}\right)=\operatorname{ITE}\left(g^{\prime \prime}, f_{1}^{\prime \prime}, f_{2}^{\prime \prime}\right)\) is possible for \(\left(g^{\prime}, f_{1}^{\prime}, f_{2}^{\prime}\right) \neq\left(g^{\prime \prime}, f_{1}^{\prime \prime}, f_{2}^{\prime \prime}\right)\). Therefore, \(N_{\text {ITE }\left(g, f_{1}, f_{2}\right)}\) can be much smaller than \(N_{g} \cdot N_{f_{1}} \cdot N_{f_{2}}\).

As a consequence of Lemma 6.77, the size of the \(\wp-\operatorname{ROBDD}\) for \(f_{1} \vee f_{2}\) is bounded above by the product of the sizes of the \(\wp\)-ROBDDs for \(f_{1}\) and \(f_{2}\). Recall that \(f_{1} \vee f_{2}=\operatorname{ITE}\left(f_{1}, 1, f_{2}\right)\) and hence
\[
N_{f_{1} \vee f_{2}} \leqslant N_{f_{1}} \cdot N_{1} \cdot N_{f_{2}}=N_{f_{1}} \cdot N_{f_{2}}
\]

The same holds for conjunction and even for any other binary Boolean connective. This also applies to operators like \(\oplus\) (xor, parity) where \(f \oplus g=\operatorname{ITE}(f, \neg g, g)\), i.e., negation
is needed to express \(f \oplus g\) by ITE, \(f\) and \(g\). Lemma 6.77 yields the bound \(N_{f} \cdot N_{g}^{2}\) for the \(\wp\)-ROBDD size for \(f \oplus g\). However, since the \(\wp\)-ROBDDs for \(g\) and \(\neg g\) are isomorphic up to exchanging the values of the drains, the recursive calls in the ITE-algorithms have the form \(\operatorname{ITE}(u, v, w)\) where \(f_{v}=\neg f_{w}\). Hence, the number of nodes in the \(\wp\)-ROBDD for \(f \oplus g\) is bounded by the number of triples \(\left(f^{\prime}, \neg g^{\prime}, g^{\prime}\right)\) where \(f^{\prime}\) is a \(\wp\)-consistent cofactor of \(f\) and \(g^{\prime}\) a \(\wp\)-consistent cofactor of \(g\). This yields the upper bound \(N_{f} \cdot N_{g}\) for the size of the \(\wp\)-ROBDD of \(f \oplus g\).

Computed Table The problem with Algorithm 22 is that its worst-case running time is exponential. The reason for this is that, if there are several paths that lead from \(\left(u, v_{1}, v_{2}\right)\) to \(\left(u^{\prime}, v_{1}^{\prime}, v_{2}^{\prime}\right)\), then \(\operatorname{ITE}\left(u^{\prime}, v_{1}^{\prime}, v_{2}^{\prime}\right)\) is invoked several times and the whole subOBDDs of these nodes \(u^{\prime}, v_{1}^{\prime}, v_{2}^{\prime}\) are traversed in each of these recursive invocations. To avoid such redundant recursive invocations one uses a computed table that stores the tuples \(\left(u, v_{1}, v_{2}\right)\), where \(\operatorname{ITE}\left(u, v_{1}, v_{2}\right)\) has already been executed, together with the result, i.e., the SOBDDD-node \(w\) with \(f_{w}=\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\). Thus, we may refine the ITE-algorithm as shown in Algorithm 23 on page 414.
```

Algorithm $23 \operatorname{ITE}\left(u, v_{1}, v_{2}\right)$
if there is an entry for $\left(u, v_{1}, v_{2}, w\right)$ in the computed table then
return node $w$
else
(* no entry for $\operatorname{ITE}\left(u, v_{1}, v_{2}\right)$ in the computed table *)
if $u$ is terminal then
if $\operatorname{val}(u)=1$ then $w:=v_{1}$ else $w:=v_{2} \mathrm{fi}$
else
$z:=\min \left\{\operatorname{var}(u), \operatorname{var}\left(v_{1}\right), \operatorname{var}\left(v_{2}\right)\right\} ;$
$w_{1}:=\operatorname{ITE}\left(\left.u\right|_{z=1},\left.v_{1}\right|_{z=1},\left.v_{2}\right|_{z=1}\right) ;$
$w_{0}:=\operatorname{ITE}\left(\left.u\right|_{z=0},\left.v_{1}\right|_{z=0},\left.v_{2}\right|_{z=0}\right)$;
if $w_{0}=w_{1}$ then $w:=w_{1}$ else $w:=\operatorname{find}_{1}$ r_add $_{-}\left(z, w_{1}, w_{0}\right) \mathbf{f i}$;
insert ( $u, v_{1}, v_{2}, w$ ) in the computed table;
return node $w$
fi
fi

```

The number of recursive calls in Algorithm 23 for the input-nodes \(u, v_{1}, v_{2}\) agrees with the \(\wp\)-ROBDD size of \(\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\), which is bounded by \(N_{u} \cdot N_{v_{1}} \cdot N_{v_{2}}\), where \(N_{v}=N_{f_{v}}\) denotes the number of nodes in the sub-OBDD of node \(v\). The cost per recursive call is constant when the time to access the computed table and to perform the find_or_addoperation is assumed to be constant. This is an adequate assumption if suitable hashing techniques are used to organize both tables. However, in practice the running time of the ITE algorithm is often much better than this upper bound. First, only in extreme cases
is the size of the \(\wp\) - \(\operatorname{ROBDD}\) for \(\operatorname{ITE}\left(f_{u}, f_{v_{1}}, f_{v_{2}}\right)\) roughly \(N_{u} \cdot N_{v_{1}} \cdot N_{v_{2}}\). Second, the use of the ITE operator yields the advantage that all synthesis algorithms that are expressible via ITE rely on the same computed table. This increases the hit rate and makes it possible that the computation aborts due to an entry in the computed table that has been made during the synthesis of another function. Moreover, there are several tricks to intensify this phenomenon. One simple trick is to make use of equivalence rules, such as
\[
f_{1} \vee f_{2}=\operatorname{ITE}\left(f_{1}, 1, f_{2}\right)=\operatorname{ITE}\left(f_{2}, 1, f_{1}\right)=\operatorname{ITE}\left(f_{1}, f_{1}, f_{2}\right)=\ldots,
\]
and to transform the arguments of ITE into so-called standard triples. Furthermore, the ITE-algorithm can be refined by terminating the traversal in certain special cases. E.g., we have \(\operatorname{ITE}(g, f, f)=f\) and \(\operatorname{ITE}(g, 1,0)=g\), which allows aborting the computation if either the last two arguments agree or they equal the pair consisting of the 1 - and 0 -drains.

\section*{Remark 6.78. The Negation Operator}

We noticed above that the negation operator can be realized as an instance of the ITE operator as we have \(\neg f=\operatorname{ITE}(f, 0,1)\). However, applying the ITE algorithm seems to be unnecessarily complicated since the \(\wp\)-ROBDDs for \(f\) and \(\neg f\) just differ in the values of the drains. In fact, swapping the values of the drains is an adequate technique to realize negation on ROBDDs, but it is not for shared OBDDs (since changing the values of the drains also affects the functions of all other root nodes). However, there is a simple trick to perform negation in SOBDDs in constant time. It relies on the use of complement bits for the edges. This permits the representation of \(f\) and \(\neg f\) by a single node. Negation then just means swapping the value of the complement bit of the incoming edge. Besides leading to smaller SOBDD sizes, the use of complement bits also tightens the effect of standard triples, since now more equivalence rules can be used to identify the input triples for ITE or for early termination. E.g., we have
\[
\operatorname{ITE}\left(f_{1}, 1, f_{2}\right)=f_{1} \vee f_{2}=\neg\left(\neg f_{1} \wedge \neg f_{2}\right)=\neg \operatorname{ITE}\left(\neg f_{1}, \neg f_{2}, 0\right)
\]
and \(\operatorname{ITE}(g, 0,1)=\neg g\). However, to ensure canonicity some extra requirements are needed. For instance, the constant function 0 could be represented by the 0 -drain (with unnegated complement bit) or the 1-drain with negated complement bit. To guarantee the uniqueness, one could require that only the 1-drain be used and that complement bits be used only for the 0 -edges (i.e., edges from the inner nodes to their 0 -successors) and the pointers to the root nodes. For more information on such advanced implementation techniques, further theoretical considerations on OBDDs and variants thereof, we refer to textbooks on BDDs, such as \([134,292,300,418]\).

Other Operators on OBDDs Although all Boolean connectives can be expressed by the ITE operator, some more operators are required to perform, e.g., the CTL model
checking procedure with an SOBDD representation of the transition system. Recall that in the symbolic computation of \(\operatorname{Sat}(\exists \diamond B)\) and \(\operatorname{Sat}(\exists \square B)\) (see Algorithms 20 and 21 on page 391) we use iterations of the form
\[
f_{j+1}(\bar{x}):=f_{j}(\bar{x}) \text { op } \exists \bar{x}^{\prime} \cdot\left(\Delta\left(\bar{x}, \bar{x}^{\prime}\right) \wedge f_{j}\left(\bar{x}^{\prime}\right)\right)
\]
where \(o p \in\{\vee, \wedge\}\) and \(f_{j}=\chi_{T}\) is the characteristic function of some set \(T\). Thus, \(f_{j+1}\) is the characteristic function of \(T \cap \operatorname{Pre}(T)\) (if \(o p=\wedge\) ) and \(T \cup \operatorname{Pre}(T)\) (if \(o p=\vee\) ). (For the treatment of constrained reachability properties like \(\exists(C \cup B)\), we have an additional conjunction with \(\chi_{C}\), but this is just a technical detail.) Besides disjunction and conjunction, these iterations use existential quantification and renaming. The major difficulty is the preimage computation, i.e., the computation of the symbolic representation of \(\operatorname{Pre}(T)\) by means of the expression \(\exists \bar{x}^{\prime} \cdot\left(\Delta\left(\bar{x}, \bar{x}^{\prime}\right) \wedge f_{j}\left(\bar{x}^{\prime}\right)\right)\), which is often called a relational product.

Let us start with the rename operator which is inherent in \(f_{j}\left(\bar{x}^{\prime}\right)\) since \(f_{j}\) is a switching function for the variables in \(\bar{x}\) and \(f_{j}\left(\bar{x}^{\prime}\right)=f_{j}\left\{\bar{x}^{\prime} \leftarrow \bar{x}\right\}\) means the function that results from \(f_{j}\) when renaming the unprimed variables \(x_{i}\) into their primed copies \(x_{i}^{\prime}\). At first glance, the renaming operator appears to be trivial as we simply may modify the variable labeling function by replacing \(x_{i}\) with \(x_{i}^{\prime}\). This operation certainly transforms a given ROBDD for \(f(\bar{x})\) into a ROBDD for \(f\left(\bar{x}^{\prime}\right)\). However, if we are given an arbitrary variable ordering \(\wp\) where the relative order of the unprimed variables can be different from the relative order of the primed variables (i.e., \(x_{i}<_{\wp} x_{j}\) while \(x_{j}^{\prime}<_{\wp} x_{i}^{\prime}\) ) then this renaming operator is no longer adequate, since the resulting ROBDD for \(f\left(\bar{x}^{\prime}\right)\) would rely on another ordering than \(\wp\). Furthermore, for an implementation with shared OBDDs, modifying existing nodes is not appropriate since then the functions for all root nodes might be affected. In fact, it is not possible to design a general rename operator which runs in time polynomial in the size of the input ROBDD. To see why, consider the function
\[
f=\left(z_{1} \wedge y_{1}\right) \vee\left(z_{2} \wedge y_{2}\right) \vee \ldots \vee\left(z_{m} \wedge y_{m}\right)
\]
of Example 6.73 (page 404). Suppose \(\operatorname{Var}=\left\{z_{i}, y_{i}, z_{i}^{\prime}, y_{i}^{\prime} \mid 1 \leqslant i \leqslant m\right\}\) and \(\wp=\) \(\left(z_{m}, y_{m}, \ldots, z_{1}, y_{1}, z_{1}^{\prime}, \ldots, z_{m}^{\prime}, y_{1}^{\prime}, \ldots, y_{m}^{\prime}\right)\) and that we are given the \(\wp\) - \(\operatorname{ROBDD}^{\mathfrak{B}_{f}}\) for \(f\) in the form of a root node \(v\) of a \(\wp-S O B D D\). The goal is now to rename \(z_{i}\) into \(z_{i}^{\prime}\) and \(y_{i}\) into \(y_{i}^{\prime}\) in \(f\), i.e., to compute the \(\wp\)-ROBDD representation of
\[
f\left\{z_{i}^{\prime} \leftarrow z_{i}, y_{i}^{\prime} \leftarrow y_{i} \mid 1 \leqslant i \leqslant m\right\}=\left(z_{1}^{\prime} \wedge y_{1}^{\prime}\right) \vee \ldots \vee\left(z_{m}^{\prime} \wedge y_{m}^{\prime}\right)
\]

By the results of Example 6.73: while the \(\wp\)-ROBDD size of \(f\) is \(2 m+2\), the \(\wp\)-ROBDD size of \(f\{\ldots\}\) is \(\Omega\left(2^{m}\right)\). This observation shows that there is no linear-time algorithm that realizes the rename operator for arbitrary variable orderings. However, if we suppose that \(x_{i}\) and \(x_{i}^{\prime}\) are neighbors in the ordering \(\wp\), e.g., \(x_{i}<_{\wp} x_{i}^{\prime}\) and there is no variable \(z\) with \(x_{s}<_{\wp}\) \(z<_{\wp} x_{i}^{\prime}\), then renaming \(x_{i}\) into \(x_{i}^{\prime}\) for a function \(f(\bar{x})\) is simple. As for the ITE operator, we can work with a DFS-based traversal of the sub-OBDD for the node representing \(f(\bar{x})\) and
generate the ROBDD for \(f\left(\bar{x}^{\prime}\right)\) in a bottom-up manner; see Algorithm 24. The algorithm takes as input a node \(v\) of a \(\pi\)-SOBDD and tuples \(\bar{x}=\left(x_{1}, \ldots, x_{n}\right), \bar{x}^{\prime}=\left(x_{1}^{\prime}, \ldots, x_{n}^{\prime}\right)\), of pairwise distinct variables such that \(x_{1}^{\prime}, \ldots, x_{n}^{\prime}\) are not essential for \(f_{v}\) and \(x_{i}\) und \(x_{i}^{\prime}\) are neighbors in \(\pi\). The output is a node \(w\) such that \(f_{w}=f_{v}\left\{\bar{x} \leftarrow \bar{x}^{\prime}\right\}\). To avoid multiple invocations of the algorithms with the same input node \(v\), we use a computed table that stores all nodes \(v\) where \(\operatorname{Rename}\left(v, \bar{x} \leftarrow \bar{x}^{\prime}\right)\) has already been executed together with the output node \(w\), i.e., the node \(w\) with \(f_{w}=f_{v}\left\{\bar{x} \leftarrow \bar{x}^{\prime}\right\}\).
```

Algorithm $24 \operatorname{Rename}\left(v, \bar{x} \leftarrow \bar{x}^{\prime}\right)$
if there is an entry $(v, w)$ in the computed table then
return $w$
else
if $v$ ist terminal then
$w:=v$
else
$w_{0}:=$ Rename $\left(\operatorname{succ}_{0}(v), \bar{x} \leftarrow \bar{x}^{\prime}\right) ;$
$w_{1}:=\operatorname{Rename}\left(\operatorname{succ}_{1}(v), \bar{x} \leftarrow \bar{x}^{\prime}\right) ;$
if $\operatorname{var}(v)=z_{j}$ for some $j \in\{1, \ldots, n\}$ then
$z:=z_{j}^{\prime} \quad \quad$ (* replace $z_{j}$ with $\left.z_{j}^{\prime}{ }^{*}\right)$
else
$z:=\operatorname{var}(v)$
fi
$w:=$ find_or_add $\left(z, w_{1}, w_{0}\right)$;
fi
insert $(u, w)$ in the computed table;
return $w$
fi

```

\section*{Remark 6.79. Interleaved Variable Orderings for Transition Systems}

We noticed before (page 409) that interleaved variable orderings, such as \(\left(x_{1}, x_{1}^{\prime}, \ldots, x_{n}, x_{n}^{\prime}\right)\), are favorable for the representation of transition systems. The rename operator yields one reason, as interleaved variable orderings permit use of the renaming that is inherent in the OBDD algorithms for the preimage computation by the above algorithm. Another formal argument for interleaved variable orderings is that they are beneficial for the construction of the ROBDD representation for the transition relation of a composite transition system. In Remark 6.59 (page 389) we saw that if \(T S\) arises from the synchronous product of transition systems \(T S_{1}, \ldots, T S_{m}\), then the switching function \(\Delta\left(\bar{x}_{1}, \ldots, \bar{x}_{n}, \bar{x}_{1}^{\prime}, \ldots, \bar{x}_{n}^{\prime}\right)\) for \(T S\) 's transition relation is obtained by the conjunction of the switching functions \(\Delta_{i}\left(\bar{x}_{i}, \bar{x}_{i}^{\prime}\right)\) for the transition relations in \(T S_{i}, i=1, \ldots, m\). Since the \(\Delta_{i}\) 's do not have common variables the \(\wp\)-ROBDD size of \(\Delta\) is bounded by
\[
N_{\Delta} \leqslant N_{\Delta_{1}}+\ldots+N_{\Delta_{m}}
\]
whenever \(\wp\) is an interleaved variable ordering where all variables in \(\bar{x}_{i}\) and \(\bar{x}_{i}^{\prime}\) are grouped together. Thus, there is no exponential blowup for the ROBDD sizes! Although Lemma 6.77 yields the upper bound \(N_{\Delta_{1}} \cdot \ldots \cdot N_{\Delta_{m}}\) for any variable ordering, for such interleaved variable orderings \(\wp\) at most linear growth of the \(\wp\)-ROBDD sizes is guaranteed. This is due to the fact that the \(\wp\)-ROBDD for \(\Delta\) arises by linking the \(\wp\)-ROBDDs for \(\Delta_{1}, \ldots, \Delta_{m}\). E.g., if we suppose that all variables in \(\bar{x}_{i}, \bar{x}_{i}^{\prime}\) appear after the variables in \(\bar{x}_{1}, \bar{x}_{1}^{\prime}, \ldots, \bar{x}_{i-1}, \bar{x}_{i-1}^{\prime}\) in \(\wp\), then we may simply redirect any edge to the 1 -drain in the \(\wp\)-ROBDD for \(\Delta_{i-1}\) to the root of the \(\wp\)-ROBDD for \(\Delta_{i}\) (for \(1 \leqslant i<m\) ). This yields the \(\wp\)-ROBDD for \(\Delta\).

A slightly more involved argument applies to the interleaving \(T S=T S_{1}\| \| \ldots \| \mid S_{m}\) where \(\Delta\) arises by the disjunction of the \(\Delta_{i}\) 's together with the side conditions \(\bar{x}_{j}=\bar{x}_{j}^{\prime}\) for \(i \neq j\). With an interleaved variable ordering where the variables for \(T S_{i}\) appear before the variables of \(T S_{i+1}, \ldots, T S_{m}\), we can guarantee that the \(\wp\)-ROBDD size \(N_{\Delta}\) is bounded by \(\mathcal{O}\left(\left(N_{\Delta_{1}}+\ldots+N_{\Delta_{m}}\right) \cdot n^{2}\right)\) where the \(n\) is the total number of variables in \(T S\). The additional factor \(\mathcal{O}\left(n^{2}\right)\) stands for the representation of the conditions \(\bar{x}_{i}=\bar{x}_{i}^{\prime}\).

Existential quantification is reducible to ITE and the cofactor operator, as we have \(\exists x . f=\) \(\left.\left.f\right|_{x=0} \vee f\right|_{x=1}\). As we mentioned in the explanations for the ITE operator, the cofactor operator \(\left.f \mapsto f\right|_{x=b}\) is trivial if \(x \leqslant_{\wp} z\) for the first essential variable \(z\) of \(f\) in the given variable ordering \(\wp\), since then \(\left.f\right|_{z=b}\) is represented by the \(b\)-successor of the node representing \(f\), if \(x=z\), and \(\left.f\right|_{x=b}=f\), if \(x<_{\wp} z\) or \(f\) is constant. If a representation for \(\left.f\right|_{x=b}\) is required where \(z<_{\wp} x\), then we may use the fact that \(\left.\left(\left.f\right|_{x=b}\right)\right|_{z=\mathfrak{c}}=\left.\left(\left.f\right|_{z=\mathfrak{c}}\right)\right|_{x=b}\) and apply the cofactor-operator recursively to the successors of the node representing \(f\). This leads to Algorithm 25 on 419. Here, again, we use a computed table that organizes all pairs \((v, w)\) where the cofactor \(\left.f_{v}\right|_{x=b}\) is known to be represented by node \(w\).

The time complexity of the renaming algorithm, as well as the algorithm for obtaining cofactors is bounded by \(\mathcal{O}\left(\operatorname{size}\left(\overline{\mathfrak{B}}_{v}\right)\right)\) as both rely on a DFS-based traversal of the subOBDD \(\overline{\mathfrak{B}}_{v}\), assuming constant time for accessing the entries in the unique and computed table. The \(\wp\)-ROBDD size of \(f\) agrees with the \(\wp\)-ROBDD size of \(f\left\{\bar{x} \leftarrow \bar{x}^{\prime}\right\}\) under the assumptions we made for \(\bar{x}, \bar{x}^{\prime}\) and \(\wp\). The \(\wp\)-ROBDD size of \(\left.f\right|_{x=b}\) is at most the \(\wp-\) ROBDD size of \(f\). This follows from the fact that given the \(\wp\)-ROBDD \(\mathfrak{B}_{f}\) for \(f\), an \(\wp\)-ROBDD for \(\left.f\right|_{x=\bar{b}}\) is obtained by redirecting any edge \(w \rightarrow u\) that leads to an \(x\)-node \(u\) to the \(b\)-successor of \(u\) and then removing all \(x\)-nodes. \({ }^{7}\) In summary, the preimage computation via the relational product
\[
\exists \bar{x} \cdot\left(\Delta \wedge f\left\{\bar{x}^{\prime} \leftarrow \bar{x}\right\}\right)
\]
- which is required for, e.g., the symbolic computation of \(\operatorname{Sat}(\exists \square B)\) - could be performed by first applying the rename operator to \(f\), then the conjunction operator (as an instance of

\footnotetext{
\({ }^{7}\) Although this yields a correct operator for constructing the \(\wp\)-ROBDD for \(\left.f\right|_{x=b}\) from the \(\wp\)-ROBDD for \(f\), the redirection of edges is not adequate for an implementation with shared OBDDs.
}
```

Algorithm $25 \operatorname{Cof}(v, x, b)$
if there is an entry $(v, w)$ in the computed table then
return $w$
else
if $v$ ist terminal then
$w:=v$
else
if $x \leqslant_{反} \operatorname{var}(v)$ then
$w:=\left.v\right|_{x=b}$
else
$z:=\operatorname{var}(v)$;
$w_{1}:=\operatorname{Cof}\left(\operatorname{succ}_{1}(v), x, b\right) ; \quad w_{0}:=\operatorname{Cof}\left(\operatorname{succ}_{0}(v), x, b\right) ;$
if $w_{0}=w_{1}$ then $w:=w_{1}$ else $w:={\text { find_or_add }\left(z, w_{1}, w_{0}\right)} \mathbf{f i}$
fi
fi
insert $(u, w)$ in the computed table;
return node $w$
fi

```

ITE operator) to the nodes representing \(\Delta\) and \(f\left(\bar{x}^{\prime}\right)\), and finally computing the existential quantifications by cofactors and disjunctions. This naive approach is very time-consuming as it relies on several top-down traversals in the shared OBDD. It also yields the problem that the ROBDD representation for \(\Delta \wedge f\left\{\bar{x}^{\prime} \leftarrow \bar{x}\right\}\) could be very large.

A more elegant approach for a BDD-based preimage computation by means of the relational product \(\exists \bar{x}^{\prime} .\left(\Delta \wedge f\left\{\bar{x}^{\prime} \leftarrow \bar{x}\right\}\right)\) is to realize the existential quantifications, renaming and conjunction simultaneously by a single DFS traversal of the sub-OBDDs of the nodes representing \(\Delta\) and \(f\), with intermediate calls of the ITE operator to realize the disjunctions that are inherent in the existential quantification. Algorithm 26 on page 420 summarizes the main steps of this approach.

The input of Algorithm 26 are two nodes \(u\) and \(v\) of a \(\wp\)-SOBDD such that \(f_{u}=\Delta\left(\bar{x}, \bar{x}^{\prime}\right)\) and \(f_{v}=f\left(\bar{x}^{\prime}\right)\). The assumptions about the variable tuples \(\bar{x}, \bar{x}^{\prime}\) are as above, i.e., \(\bar{x}=\left(x_{1}, \ldots, x_{n}\right)\) and \(\bar{x}^{\prime}=\left(x_{1}^{\prime}, \ldots, x_{n}^{\prime}\right)\) are tuples consisting of pairwise distinct variables such that the unprimed variables \(x_{i}\) and their primed copies \(x_{i}^{\prime}\) are neighbours in \(\wp\). For simplicity, let us suppose that \(\wp\) interleaves the unprimed and primed variables, say
\[
x_{1}<_{\wp} x_{1}^{\prime}<_{\wp} x_{2}<_{\wp} x_{2}^{\prime}<_{\wp} \ldots<_{\wp} x_{n}<_{\wp} x_{n}^{\prime}
\]

The output of Algorithm 26 is a (possibly new) node \(w\) with \(f_{w}=\exists \bar{x}^{\prime} \cdot\left(\Delta\left(\bar{x}, \bar{x}^{\prime}\right) \wedge f\left(\bar{x}^{\prime}\right)\right)=\) \(\exists \bar{x}^{\prime} \cdot\left(f_{u} \wedge f_{v}\right)\). The termination condition of Algorithm 26 is given by the cases: (i) there exists an entry in the computed table, or (ii) \(\Delta=0\) or \(f=0\) in which case \(\exists \bar{x}^{\prime} .(\Delta \wedge f)=0\),
```

Algorithm 26 Relational product $\operatorname{RelProd}(u, v)$
if there exists an entry $(u, v, w)$ in the computed table then return $w \mathbf{f}$;
if $u$ or $v$ is the 0 -drain then return the 0 -drain fi;
if $u$ and $v$ are the 1-drain then return the 1-drain fi;
$y:=\min \{\operatorname{var}(u), \operatorname{var}(v)\}$, say $y \in\left\{x_{i}, x_{i}^{\prime}\right\}$
if $y=x_{i}$ then
$w_{1,0}:=\operatorname{RelProd}\left(\left.u\right|_{x_{i}=1, x_{i}^{\prime}=0},\left.v\right|_{x_{i}=0}\right)$;
$w_{1,1}:=\operatorname{RelProd}\left(\left.u\right|_{x_{i}=1, x_{i}^{\prime}=1},\left.v\right|_{x_{i}=1}\right)$;
$w_{1}:=\operatorname{ITE}\left(w_{1,0}, 1, w_{1,1}\right)$;
$w_{0,0}:=\operatorname{RelProd}\left(\left.u\right|_{x_{i}=0, x_{i}^{\prime}=0},\left.v\right|_{x_{i}=0}, \bar{x}, \bar{x}^{\prime}\right)$;
$w_{0,1}:=\operatorname{RelProd}\left(\left.u\right|_{x_{i}=0, x_{i}^{\prime}=1},\left.v\right|_{x_{i}=1}, \bar{x}, \bar{x}^{\prime}\right)$;
$w_{0}:=\operatorname{ITE}\left(w_{0,0}, 1, w_{0,1}\right)$;
if $w_{1}=w_{0}$ then
$w:=w_{1} \quad$ (* elimination rule *)
else
$w:=$ find_or_add $\left(x_{i}, w_{1}, w_{0}\right)$
fi
else
$w_{0}:=\operatorname{RelProd}\left(\left.u\right|_{x_{i}^{\prime}=0}, v\right) ; \quad w_{1}:=\operatorname{RelProd}\left(\left.u\right|_{x_{i}^{\prime}=1}, v\right) ;$
$w:=\operatorname{ITE}\left(w_{0}, 1, w_{1}\right)$
fi
insert $(u, v, w)$ in the computed table;
return $w$

```
or (iii) \(\Delta=f=1\) in which case \(\exists \bar{x}^{\prime} .(\Delta \wedge f)=1\). In the remaining cases, the traversal of the sub-OBDDs of \(u\) and \(v\) relies on the expansion rule:
\[
\begin{aligned}
& \left.\exists x_{1}^{\prime} \exists x_{2}^{\prime} \ldots \exists x_{n}^{\prime} \cdot\left(\Delta \wedge f\left\{x_{1}^{\prime} \leftarrow x_{1}, x_{2}^{\prime} \leftarrow x_{2}, \ldots, x_{n}^{\prime} \leftarrow x_{n}\right\}\right)\right)\left.\right|_{x_{1}=b} \\
& =\quad \exists x_{2}^{\prime} \ldots \exists x_{n}^{\prime} \cdot\left(\left.\left.\Delta\right|_{x_{1}=b, x_{1}^{\prime}=0} \wedge f\right|_{x_{1}=0}\left\{x_{2}^{\prime} \leftarrow x_{2}, \ldots, x_{n}^{\prime} \leftarrow x_{n}\right\}\right) \vee \\
& \quad \exists x_{2}^{\prime} \ldots \exists x_{n}^{\prime} \cdot\left(\left.\left.\Delta\right|_{x_{1}=b, x_{1}^{\prime}=1} \wedge f\right|_{x_{1}=1}\left\{x_{2}^{\prime} \leftarrow x_{2}, \ldots, x_{n}^{\prime} \leftarrow x_{n}\right\}\right)
\end{aligned}
\]

In the literature, several techniques have been proposed that serve to improve the image or preimage computation. These range from techniques that rely on partitionings of the variables that attempt to perform the existential quantification as soon as possible (as they decrease the number of essential variables and often lead to smaller ROBDD sizes). Other techniques rely on so-called input- or output splitting (which use alternative expansion rules), and special ROBDD operators that attempt to replace, e.g., \(\Delta\) with other switching functions \(\tilde{\Delta}\) such that \(\Delta \wedge f=\tilde{\Delta} \wedge f\) and such that the \(\wp\)-ROBDD for \(\bar{\Delta}\) is smaller than the \(\wp\)-ROBDD for \(\Delta\). In the case of an iterated preimage computation by \(T_{0}=B\) and \(T_{j+1}=T_{j} \cup \operatorname{Pre}\left(T_{j}\right)\) for the symbolic computation of \(\operatorname{Pre}(B)\), one might also switch from \(T_{j}\) to any set \(\tilde{T}\) such that \(T_{j} \backslash T_{j-1} \subseteq \tilde{T} \subseteq T_{j}\) and compute \(T_{j} \cup \operatorname{Pre}(\tilde{T})\). For such advanced techniques, we refer the interested reader to \([92,292,374]\) and the literature mentioned there.

Summary We now have all ingredients for a symbolic realization of the standard CTL model-checking approach which recursively computes the satisfaction sets of the subformulae by means of shared OBDDs. Initially, one has to construct the ROBDD representation of the transition system to be analyzed. This can be done in a compositional way by means of synthesis operators (disjunction, conjunction, etc.) as mentioned in Remark 6.59. Furthermore, we assume that ROBDD representations of the satisfaction sets for the atomic propositions are given. This assumption is justified since often the atomic propositions can serve as variables for the encoding of the states. (And in this case their satisfaction set is just given by a projection function.) The CTL model-checking procedure can then be performed by means of the ITE algorithm (to treat the propositional logic fragment of CTL) and the symbolic BFS-based algorithms sketched in Algorithms 20 and 21. Both rely on an iterative preimage computation. Techniques to do this efficiently have been discussed above. The termination condition requires checking the equality of two switching functions. This, in fact, is trivial for shared OBDDs since it simply amounts to the comparison of the corresponding nodes and can be performed in constant time.

\subsection*{6.8 CTL*}

We saw in Theorem 6.21 on page 337 that CTL and LTL have incomparable expressiveness. An extension of CTL, proposed by Emerson and Halpern, called CTL*, combines the features of both logics, and thus is more expressive than either of them.

\subsection*{6.8.1 Logic, Expressiveness, and Equivalence}

CTL* is an extension of CTL as it allows path quantifiers \(\exists\) and \(\forall\) to be arbitrarily nested with linear temporal operators such as \(\bigcirc\) and U. In contrast, in CTL each linear temporal operator must be immediately preceded by a path quantifier. As in CTL, the syntax of CTL* distinguishes between state and path formulae. The syntax of CTL* state formulae is roughly as in CTL, while the CTL* path formulae are defined as LTL formulae, the only difference being that arbitrary CTL* state formulae can be used as atoms. For example, \(\forall \bigcirc \bigcirc a\) is a legal CTL* formula, but does not belong to CTL. The same applies to the CTL* formulae \(\exists \square \diamond a\) and \(\forall \square \diamond a\). (However, \(\forall \square \diamond a\) is equivalent to the CTL formula \(\forall \square \forall \diamond\) a.)

\section*{Definition 6.80. Syntax of CTL*}

CTL* state formulae over the set AP of atomic propositions, briefly called CTL* formulae, are formed according to the following grammar:
\[
\Phi::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\quad \neg \Phi| \exists \varphi
\]
where \(a \in A P\) and \(\varphi\) is a path formula. The syntax of CTL* path formulae is given by the following grammar:
\[
\varphi::=\Phi\left|\varphi_{1} \wedge \varphi_{2}\right| \neg \varphi|\bigcirc \varphi| \varphi_{1} \cup \varphi_{2}
\]
where \(\Phi\) is a state formula, and \(\varphi, \varphi_{1}\), and \(\varphi_{2}\) are path formulae.

As for LTL or CTL, we use derived propositional logic operators like \(\vee, \rightarrow, \ldots\) and let
\[
\diamond \varphi=\operatorname{true} \mathrm{U} \varphi \quad \text { and } \quad \square \varphi=\neg \diamond \neg \varphi .
\]

The universal path quantifier \(\forall\) can be defined in CTL* by existential quantification and negation:
\[
\forall \varphi=\neg \exists \neg \varphi .
\]
(Note that this is not the case for CTL.)

For example, the following formulae are syntactically correct CTL* formulae:
\[
\forall \square(\bigcirc \diamond a \wedge \neg(b \cup \square c))
\]
and
\[
\forall \bigcirc \square \neg a \wedge \exists \diamond \square(a \vee \forall(b \cup a)) .
\]

Note that these formulae are not CTL formulae.

\section*{Definition 6.81. Satisfaction Relation for CTL*}

Let \(a \in A P\) be an atomic proposition, \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system without terminal states, state \(s \in S, \Phi, \Psi\) be CTL* state formulae, and \(\varphi, \varphi_{1}\) and \(\varphi_{2}\) be CTL* path formulae. The satisfaction relation \(\vDash\) is defined for state formulae by
\[
\begin{array}{llll}
s \models a & \text { iff } & & a \in L(s), \\
s \models \neg \Phi & & \text { iff } & \\
\text { not } s \models \Phi, \\
s \models \Phi \wedge \Psi & & \text { iff } & \\
(s \models \Phi) \text { and }(s \models \Psi), \\
s \models \exists \varphi & & \text { iff } & \\
\pi \models \varphi \text { for some } \pi \in \operatorname{Paths}(s) .
\end{array}
\]

For path \(\pi\), the satisfaction relation \(\models\) for path formulae is defined by:
\[
\begin{array}{llll}
\pi & =\Phi & \text { iff } & \\
s_{0} \models \Phi, \\
\pi & =\varphi_{1} \wedge \varphi_{2} & \text { iff } & \pi \models \varphi_{1} \text { and } \pi \models \varphi_{2}, \\
\pi & =\neg \varphi & \text { iff } & \pi \not \models \varphi, \\
\pi & =\bigcirc \varphi & \text { iff } & \pi[1 . .] \models \varphi, \\
\pi & =\varphi_{1} \cup \varphi_{2} & \text { iff } & \exists j \geqslant 0 .\left(\pi[j . .] \models \varphi_{2} \wedge\left(\forall 0 \leqslant k<j . \pi[k . .] \models \varphi_{1}\right)\right)
\end{array}
\]
where for path \(\pi=s_{0} s_{1} s_{2} \ldots\) and integer \(i \geqslant 0, \pi[i .\).\(] denotes the suffix of \pi\) from index \(i\) on.

\section*{Definition 6.82. CTL* Semantics for Transition Systems}

For CTL*-state formula \(\Phi\), the satisfaction \(\operatorname{set} \operatorname{Sat}(\Phi)\) is defined by
\[
\operatorname{Sat}(\Phi)=\{s \in S \mid s \models \Phi\} .
\]

The transition system \(T S\) satisfies CTL* formula \(\Phi\) if and only if \(\Phi\) holds in all initial states of TS:
\[
T S \models \Phi \quad \text { if and only if } \forall s_{0} \in I . s_{0} \models \Phi .
\]

Thus, \(T S \vDash \Phi\) if and only if all initial states of \(T S\) satisfy the formula \(\Phi\).
LTL formulae are CTL* path formulae in which the elementary state formulae \(\Phi\) are restricted to atomic propositions. Apparently, the interpretation of LTL over the paths of a transition system (see Definition 5.7, page 237) corresponds to the semantics of LTL obtained as a sublogic of CTL*. The following theorem demonstrates that the corresponding claim also holds for states. Thereby, every LTL formula \(\varphi\) is identified with the CTL* formula \(\forall \varphi\), and the semantics of LTL over states is taken as a reference. Recall that according to the LTL semantics in paths, \(s \models \varphi\) if and only if \(\pi \models \varphi\) for all \(\pi \in \operatorname{Paths}(s)\).

\section*{Theorem 6.83. Embedding of LTL in CTL*}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system without terminal states. For each LTL formula \(\varphi\) over AP and for each \(s \in S\) :
\[
\underbrace{s \models \varphi}_{\text {LTL semantics }} \text { if and only if } \underbrace{s \models \forall \varphi}_{C T L^{*} \text { semantics }} .
\]

In particular, \(T S \models \varphi\) (with respect to the LTL semantics) if and only if \(T S \models \forall \varphi\) (with respect to the CTL* semantics)

As a result, it is justified to understand LTL (with interpretation over the states of a transition system) as a sublogic of CTL*. Theorem 6.21 (page 337) stated that the expressivenesses of LTL and CTL are incomparable. Since LTL is a sublogic of CTL*, it now follows that CTL* subsumes LTL and CTL, i.e., there exist CTL* formulae which can be expressed neither in LTL nor in CTL.

Theorem 6.84. CTL \(^{*}\) is More Expressive Than LTL and CTL
For the \(C T L^{*}\) formula over \(A P=\{a, b\}\),
\[
\Phi=(\forall \diamond \square a) \vee(\forall \square \exists \diamond b),
\]
there does not exist any equivalent LTL or CTL formula.

Proof: This follows directly from the fact that \(\forall \square \exists \diamond b\) is a CTL formula that cannot be expressed in LTL, whereas \(\diamond \square a\) is an LTL formula that cannot be expressed in CTL. Both these facts follow from Theorem 6.21 (page 337).

The relationship between LTL, CTL, and CTL* is depicted in Figure 6.27.


Figure 6.27: Relationship between LTL CTL and CTL*.

One of the consequences of this fact, is that-as in LTL-fairness assumptions can be expressed syntactically in CTL*. For instance, for fairness assumption fair, formulae of the form
\[
\forall(\text { fair } \rightarrow \varphi) \quad \text { or } \quad \exists(\text { fair } \wedge \varphi)
\]
are legal CTL* formulae. As for CTL or LTL, semantic equivalence \(\equiv\) can be defined for CTL* formulae. Apart from the equivalence laws for CTL and the laws resulting from the equivalence laws for LTL, there exists a series of other important laws that are specific to CTL*. This includes, among others, the laws listed in Figure 6.28.

Example instances of the duality laws for the path quantifiers are
\[
\neg \forall \square \diamond a \equiv \exists \diamond \square \neg a \quad \text { and } \quad \neg \exists \square \diamond a \equiv \forall \diamond \square \neg a
\]

As usual, universal quantification does not distribute over disjunction, and the same applies to existential quantification and conjunction:
\[
\forall(\varphi \vee \psi) \not \equiv \forall \varphi \vee \forall \psi \quad \text { and } \quad \exists(\varphi \wedge \psi) \not \equiv \exists \varphi \wedge \exists \psi
\]

Path quantifier elimination should also be considered with care. For instance,
\[
\forall \diamond \square \varphi \not \equiv \forall \diamond \forall \square \varphi \quad \text { and } \quad \exists \square \diamond \Phi \not \equiv \exists \square \exists \diamond \varphi
\]

Finally, we remark that for \(\mathrm{CTL}^{*}\)-state formula \(\Phi\) we have that
\[
\exists \Phi \equiv \Phi \quad \text { and } \quad \forall \Phi \equiv \Phi
\]

This is illustrated by means of an example. Consider the CTL* formula \(\exists \forall \diamond a\). This formula holds in state \(s\) whenever there exists an infinite path fragment \(\pi=s_{0} s_{1} s_{2} \ldots \in\) \(\operatorname{Paths}(s)\), such that \(\pi \models \forall \diamond a\). Since \(\pi \models \forall \diamond a\) holds if and only if \(s=s_{0} \models \forall \diamond a\), the formula \(\exists \forall \diamond a\) is equivalent to \(\forall \diamond a\).
duality laws for path quantifiers
\[
\begin{aligned}
\neg \forall \varphi & \equiv \exists \neg \varphi \\
\neg \exists \varphi & \equiv \forall \neg \varphi
\end{aligned}
\]
distributive laws
\[
\begin{aligned}
\forall\left(\varphi_{1} \wedge \varphi_{2}\right) & \equiv \forall \varphi_{1} \wedge \forall \varphi_{2} \\
\exists\left(\varphi_{1} \vee \varphi_{2}\right) & \equiv \exists \varphi_{1} \vee \exists \varphi_{2}
\end{aligned}
\]
quantifier absorption laws
\[
\begin{aligned}
& \forall \square \diamond \varphi \equiv \forall \square \forall \diamond \varphi \\
& \exists \diamond \square \varphi \equiv \exists \diamond \exists \square \varphi
\end{aligned}
\]

Figure 6.28: Some equivalence laws for CTL*.

Remark 6.85. Extending CTL with Boolean Connectors for Path Formulae (CTL \({ }^{+}\)) Consider the following fragment, called CTL \({ }^{+}\), of CTL* , which extends CTL by allowing Boolean operators in path formulae. \(\mathrm{CTL}^{+}\)state formulae over the set \(A P\) of atomic proposition are formed according to the following grammar:
\[
\Phi::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\neg \Phi| \exists \varphi \mid \forall \varphi
\]
where \(a \in A P\) and \(\varphi\) is a path formula. \(\mathrm{CTL}^{+}\)path formulae are formed according to the following grammar:
\[
\varphi::=\varphi_{1} \wedge \varphi_{2}|\neg \varphi| \bigcirc \Phi \mid \Phi_{1} \cup \Phi_{2}
\]
where \(\Phi, \Phi_{1}\), and \(\Phi_{2}\) are state formulae, and \(\varphi_{1}, \varphi_{2}\) are path formulae. Surprisingly, CTL \({ }^{+}\) is as expressive as CTL, i.e., for any CTL \(^{+}\)state formula \(\Phi^{+}\)there exists an equivalent CTL formula \(\Phi\). For example:
\[
\underbrace{\exists(a \mathrm{~W} b)}_{\text {CTL formula }} \equiv \underbrace{\exists((a \cup b) \vee \square a)}_{\text {CTL }^{+} \text {formula }}
\]
or
\[
\underbrace{\exists(\diamond a \wedge \diamond b)}_{\text {CTL }^{+} \text {formula }} \equiv \underbrace{\exists \diamond(a \wedge \exists \diamond b) \wedge \exists \diamond(b \wedge \exists \diamond a)}_{\text {CTL formula }}
\]

We do not provide here a full proof for transforming CTL \({ }^{+}\)formulae into equivalent CTL formulae. The transformation relies on equivalence laws such as
\[
\begin{aligned}
\exists(\neg \bigcirc \Phi) \equiv & \exists \bigcirc \neg \Phi \\
\exists\left(\neg\left(\Phi_{1} \cup \Phi_{2}\right)\right) \equiv & \exists\left(\left(\Phi_{1} \wedge \neg \Phi_{2}\right) \cup\left(\neg \Phi_{1} \wedge \neg \Phi_{2}\right)\right) \vee \exists \square \neg \Phi_{2} \\
\exists\left(\bigcirc \Phi_{1} \wedge \bigcirc \Phi_{2}\right) \equiv & \exists \bigcirc\left(\Phi_{1} \wedge \Phi_{2}\right) \\
\exists\left(\bigcirc \Phi \wedge\left(\Phi_{1} \cup \Phi_{2}\right)\right) \equiv & \left(\Phi_{2} \wedge \exists \bigcirc \Phi\right) \vee\left(\Phi_{1} \wedge \exists \bigcirc\left(\Phi \wedge \exists\left(\Phi_{1} \cup \Phi_{2}\right)\right)\right) \\
\exists\left(\left(\Phi_{1} \cup \Phi_{2}\right) \wedge\left(\Psi_{1} \cup \Psi_{2}\right)\right) \equiv & \exists\left(\left(\Phi_{1} \wedge \Psi_{1}\right) \cup\left(\Phi_{2} \wedge \exists\left(\Psi_{1} \cup \Psi_{2}\right)\right)\right) \vee \\
& \exists\left(\left(\Phi_{1} \wedge \Psi_{1}\right) \cup\left(\Psi_{2} \wedge \exists\left(\Phi_{1} \cup \Phi_{2}\right)\right)\right)
\end{aligned}
\]

Thus, CTL can be expanded by means of a Boolean operator for path formulae without changing the expressiveness. However, \(\mathrm{CTL}^{+}\)formulae can be much shorter than the shortest equivalent CTL formulae.

\subsection*{6.8.2 CTL* Model Checking}

This section treats a model-checking algorithm for CTL*. The CTL* model-checking problem is to establish whether \(T S \models \Phi\) holds for a given finite transition system \(T S\) (without terminal states) and the CTL* state formula \(\Phi\). As we will see, an appropriate combination of the model-checking algorithms for LTL and CTL suffices.

As for CTL, the model-checking algorithm for CTL* is based on a bottom-up traversal of the syntax tree of the formula \(\Phi\) to be checked. Due to the bottom-up nature of the algorithm, the satisfaction set \(\operatorname{Sat}(\Psi)\) for any state sub formulae \(\Psi\) of \(\Phi\) has been computed before, and can be used to determine \(\operatorname{Sat}(\Phi)\). This holds in particular for the maximal proper state subformulae of \(\Phi\).

\section*{Definition 6.86. Maximal Proper State Subformula}

State formula \(\Psi\) is a maximal proper state subformula of \(\Phi\) whenever \(\Psi\) is a subformula of \(\Phi\) that differs from \(\Phi\) and that is not contained in any other proper state subformula of \(\Phi\).

The basic concept is to replace all maximal proper state subformulae of \(\Phi\) by fresh atomic propositions \(a_{1}, \ldots, a_{k}\), say. These propositions do not occur in \(\Phi\) and are such that \(a_{i} \in L(s)\) if and only if \(s \in \operatorname{Sat}\left(\Psi_{i}\right)\), the \(i\) th maximal state subformula of \(\Phi\). For state
subformulae whose "top-level" operator is a Boolean operator (such as negation or conjunction), the treatment is obvious. Let us consider the more interesting case of \(\Psi=\exists \varphi\). By replacing all maximal state subformulae in \(\varphi\), an LTL formula results! Since
\[
s \models \exists \varphi \text { iff } \underbrace{s \not \models \forall \neg \varphi}_{\text {CTL }^{*} \text { semantics }} \text { iff } \underbrace{s \not \models \neg \varphi}_{\text {LTL semantics }},
\]
it suffices to compute the satisfaction set
\[
\operatorname{Sat}_{L T L}(\neg \varphi)=\left\{s \in S|s|_{L T L} \neg \varphi\right\}
\]
by means of an LTL model checker. (Here, the notations \(\operatorname{Sat}_{L T L}(\cdot)\) and \(\models_{L T L}\) are used to emphasize that the basis is the LTL satisfaction relation.) The satisfaction set for \(\Phi=\exists \varphi\) is now obtained by complementation:
\[
\operatorname{Sat}_{C T L^{*}}(\exists \varphi)=S \backslash \operatorname{Sat}_{L T L}(\neg \varphi)
\]

For CTL* formulae where the outermost operator is an universal quantification we simply may deal with
\[
\operatorname{Sat}_{C T L^{*}}(\forall \varphi)=\operatorname{Sat}_{L T L}(\varphi)
\]
where, as before, it is assumed that \(\varphi\) is an LTL formula resulting from the replacement of the maximal state subformula with fresh atomic propositions.

The main steps of the CTL* model-checking procedure are presented in Algorithm 27 on page 429 .

\section*{Example 6.87. Abstract Example of CTL* Model Checking}

The CTL* model-checking approach is illustrated by considering the CTL* formula:
\[
\exists \varphi \quad \text { where } \quad \varphi=\bigcirc(\forall \square \exists \diamond a) \wedge \diamond \square \exists(\bigcirc a \wedge \square b) .
\]

The maximal proper state subformulae of \(\varphi\) are
\[
\Phi_{1}=\forall \square \exists \diamond a \quad \text { and } \quad \Phi_{2}=\exists(\bigcirc a \wedge \square b)
\]

Thus:
\[
\varphi=\bigcirc \underbrace{(\forall \square \exists \diamond a)}_{\Phi_{1}} \wedge \diamond \square \underbrace{\exists(\bigcirc a \wedge \square b)}_{\Phi_{2}}=\bigcirc \Phi_{1} \wedge \diamond \square \Phi_{2} .
\]

According to the model-checking algorithm for CTL*, the satisfaction sets \(\operatorname{Sat}\left(\Phi_{i}\right)\) are computed recursively. Subsequently, \(\Phi_{1}\) and \(\Phi_{2}\) are replaced with the atomic propositions \(a_{1}\) and \(a_{2}\), say. This yields the following LTL formula over the set of propositions \(A P^{\prime}=\) \(\left\{a_{1}, a_{2}\right\}:\)
\[
\varphi^{\prime}=\bigcirc a_{1} \wedge \diamond \square a_{2}
\]
```

Algorithm 27 CTL* model checking algorithm (basic idea)
Input: finite transition system $T S$ with initial states $I$, and CTL* formula $\Phi$
Output: $I \subseteq \operatorname{Sat}(\Phi)$
for all $i \leqslant|\Phi|$ do
for all $\Psi \in \operatorname{Sub}(\Phi)$ with $|\Psi|=i$ do
switch $(\Psi)$ :
true : $\operatorname{Sat}(\Psi):=S$;
$a \quad: \quad \operatorname{Sat}(\Psi):=\{s \in S \mid a \in L(s)\} ;$
$a_{1} \wedge a_{2} \quad: \quad \operatorname{Sat}(\Psi):=\operatorname{Sat}\left(a_{1}\right) \cap \operatorname{Sat}\left(a_{2}\right) ;$
$\neg a \quad: \quad \operatorname{Sat}(\Psi):=S \backslash \operatorname{Sat}(a) ;$
$\exists \varphi \quad: \quad$ determine $\operatorname{Sat}_{L T L}(\neg \varphi)$ by means of an LTL model-checker;
: $\operatorname{Sat}(\Psi):=S \backslash \operatorname{Sat}_{L T L}(\neg \varphi)$
end switch
$A P:=A P \cup\left\{a_{\Psi}\right\} ; \quad$ (* introduce fresh atomic proposition *)
replace $\Psi$ with $a_{\Psi}$
forall $s \in \operatorname{Sat}(\Psi)$ do $L(s):=L(s) \cup\left\{a_{\Psi}\right\} ;$ od
od
od
return $I \subseteq \operatorname{Sat}(\Phi)$

```

The labeling function \(L^{\prime}: S \rightarrow 2^{A P^{\prime}}\) is given by:
\[
a_{i} \in L^{\prime}(s) \quad \text { if and only if } s \in \operatorname{Sat}\left(\Phi_{i}\right) \text { for } i \in\{1,2\} .
\]

Applying the LTL model-checking algorithm to the formula \(\neg \varphi^{\prime}\) yields the set of states satisfying (with respect to the LTL semantics) \(\neg \varphi^{\prime}\), i.e., \(\operatorname{Sat}_{L T L}\left(\neg \varphi^{\prime}\right)\). The complement of \(\operatorname{Sat}_{L T L}\left(\neg \varphi^{\prime}\right)\) provides the set \(\operatorname{Sat}_{C T L^{*}}(\varphi)\).

Evidently, the time complexity of the CTL* model-checking algorithm is dominated by the LTL model-checking phases. The additional effort that is necessary for CTL* model checking is polynomial in the size of the transition system and the length of the formula. Hence, a time complexity is obtained which is exponential in the length of the formula and linear in the size of the transition system.

\section*{Theorem 6.88. Time Complexity of CTL* Model Checking}

For transition system \(T S\) with \(N\) states and \(K\) transitions, and \(C T L^{*}\) formula \(\Phi\), the CTL* model-checking problem \(T S=\Phi\) can be determined in time \(\mathcal{O}\left((N+K) \cdot 2^{|\Phi|}\right)\).

Note that CTL* model checking can be solved by any LTL model-checking algorithm. These considerations show that there is a polynomial reduction of the CTL* model-
\begin{tabular}{|c|c|c|c|}
\hline & CTL & LTL & CTL* \(^{*}\) \\
\hline \hline model checking & PTIME & \(\operatorname{PSPACE-complete~}\) & PSPACE-complete \\
without fairness & \(\operatorname{size}(T S) \cdot|\Phi|\) & \(\operatorname{size}(T S) \cdot \exp (|\Phi|)\) & \begin{tabular}{c}
\(\operatorname{size}(T S) \cdot \exp (|\Phi|)\)
\end{tabular} \\
with fairness & \(\operatorname{size}(T S) \cdot|\Phi| \cdot \mid\) fair \(\mid\) & \(\operatorname{size}(T S) \cdot \exp (|\Phi|) \cdot \mid\) fair \(\mid\) & \(\operatorname{size(TS)\cdot \operatorname {exp}(|\Phi |)\cdot |\text {fair}|}\) \\
\hline \begin{tabular}{c} 
for fixed specifications \\
(model complexity)
\end{tabular} & \(\operatorname{size(TS)}\) & \(\operatorname{size(TS)}\) & \(\operatorname{size(TS)}\) \\
\hline satisfiability check \\
best known technique & \(\operatorname{EXPTIME}\) \\
\(\exp (|\Phi|)\) & \begin{tabular}{c} 
PSPACE-complete \\
\(\exp (|\Phi|)\)
\end{tabular} & \begin{tabular}{c} 
2EXPTIME \\
\(\exp (\exp (|\Phi|))\)
\end{tabular} \\
\hline
\end{tabular}

Figure 6.29: Complexity of the model-checking algorithms and satisfiability checking.
checking problem to the LTL model-checking problem. As a result, the theoretical complexity results for LTL also apply to CTL*. Table 6.29 summarizes the complexity results for model checking CTL, CTL*, and LTL.

\section*{Theorem 6.89. Theoretical Complexity of CTL* Model Checking}

The CTL* model-checking problem is PSPACE-complete.

\subsection*{6.9 Summary}
- Computation Tree Logic (CTL) is a logic for formalizing properties over computation trees, i.e., the branching structure of the states.
- The expressivenesses of LTL and CTL are incomparable.
- Although fairness constraints cannot be encoded in CTL formulae, fairness assumptions can be incorporated in CTL by adapting the CTL semantics such that quantification is over fair paths, rather than over all paths.
- The CTL model-checking problem can be solved by a recursive descent procedure over the parse tree of the state formula to be checked. The set of states satisfying \(\exists(\Phi \cup \Psi)\) can be determined using a smallest fixed-point procedure; for \(\exists \square \Phi\) this is a largest fixed-point procedure.
- The time complexity of the CTL model-checking algorithm is linear in the size of the transition system and the length of the formula. In case fairness constraints are considered, an additional multiplicative factor that is proportional to the number of fairness constraints needs to be taken into account.
- Counterexamples and witnesses for CTL path formulae can be determined using a standard graph analysis.
- The CTL model-checking procedure can be realized symbolically by means of ordered binary decision diagrams. These provide a universal and canonical data structure for switching functions.
- Extended Computation Tree Logic (CTL*) is more expressive than either CTL and LTL.
- The CTL* model-checking problem can be solved by an appropriate combination of the recursive descent procedure (as for CTL) and the LTL model-checking algorithm.
- The CTL* model-checking problem is PSPACE-complete.

\subsection*{6.10 Bibliographic Notes}

Branching temporal logics. Various types of branching temporal logic have been proposed in the literature. We mention a few important ones in increasing expressive power: Hennessy-Milner logic (HML [197]), Unified System of Branching-Time Logic [42], Computation Tree Logic (CTL [86]), Extended Computation Tree Logic (CTL* [86]), and modal \(\mu\)-Calculus [243]. The modal \(\mu\)-calculus is the most expressive among these languages, and HML is the least expressive. CTL has been extended with fairness by Emerson and Halpern [140] and by Emerson and Lei [143].

Not treated in this textbook is the task of proving satisfiability of formulae by means of algorithms or deductive techniques. The satisfiability problems for CTL, CTL* , and other temporal logics have been addressed by many researchers and used, e.g., in the context of synthesis problems where the goal is to design a system model from a given temporal specification. Emerson [138] has shown that checking CTL satisfiability is in the complexity class EXPTIME. This means that the time complexity of checking CTL satisfiability is exponential in the length of the formula. For CTL* this problem is double exponential [141] in the length of the formula. A complete axiomatization of CTL has been given by Ben-Ari, Manna and Pnueli [42] and Emerson and Halpern [139].

Branching vs. linear temporal logics. The discussion of the relative merits of linear- vs. branching-time logics goes back to the early eighties. Pnueli [338] established that linear
and branching temporal logics are based on two distinct notions of time. Various papers [85, 140, 259] show that the expressivenesses of LTL and CTL are incomparable. A somewhat more practical view on comparing the usefulness of LTL vs. CTL was recently given by Vardi [410]. The logic CTL* that encompasses LTL and CTL was defined by Clarke and Emerson [86].

CTL model checking. The first algorithms for CTL model checking were presented by Clarke and Emerson [86] in 1981 and (for a logic similar to CTL) by Queille and Sifakis [347] in 1982. The algorithm by Clarke and Emerson was polynomial in both the size of the transition system and the length of the formula, and could handle fairness. Clarke, Emerson, and Sistla [87] presented an efficiency improvement using the detection of strongly connected components and backward breadth-first search, yielding an algorithm that is linear in both the size of the system and the length of the formula. CTL model checking based on a forward search has been proposed by Iwashita, Nakata, and Hirose [224]. Emerson and Lei [143] showed that CTL* can be checked with essentially the same complexity as LTL, using a combination of the algorithms for LTL and CTL. The same authors consider in [142] CTL model checking under a broad class of fairness assumptions. Practical aspects of CTL* model checking have been reported by Bhat, Cleaveland, and Grumberg [50], and more recently by Visser and Barringer [414]. Algorithms for generating counterexamples and witnesses originate from the works by Clarke et al. [91] and Hojati, Brayton, and Kurshan [204]. More recent developments are the use of satisfiability solvers for propositional logic or quantified Boolean formulae to find counterexamples up to certain length, as proposed by Clarke et al. [84], and the use of tree-like counterexamples as opposed to linear ones by Clarke [93].

CTL model checkers. Clarke and Emerson [86] reported the first (fair) CTL model checker, called EMC. About the same time, Queille and Sifakis [347] announced CESAR, a model checker for a branching logic very similar to CTL. EMC was improved in [87] and constituted the basis for SMV (Symbolic Model Verifier), an efficient CTL model checker by McMillan based on a symbolic OBDD-based representation of the state space [288]. The concept of ordered binary decision diagrams has been proposed by Bryant [70]. The milestone for symbolic model checking with OBDDs is the paper by Burch et al.[74]. Further references for OBDD-based approaches have been given in Section 6.7. Recent variants of SMV are NuSMV [83] developed by the teams of Cimatti et al., and SMV by McMillan and colleagues at Cadence Berkeley Laboratories that is focused on compositionality. Both tools are freely available. Another symbolic CTL model checker is VIS [62].

\subsection*{6.11 Exercises}

Exercise 6.1. Consider the following transition system over \(A P=\{b, g, r, y\}\) :


The following atomic propositions are used: \(r\) (red), \(y\) (yellow), \(g\) (green), and \(b\) (black). The model is intended to describe a traffic light that is able to blink yellow. You are requested to indicate for each of the following CTL formulae the set of states for which these formulae hold:
(a) \(\forall \Delta y\)
(g) \(\exists \square \neg g\)
(b) \(\forall \square y\)
(h) \(\forall(b U \neg b)\)
(c) \(\forall \square \forall \Delta y\)
(i) \(\exists(b \cup \neg b)\)
(d) \(\forall \diamond g\)
(j) \(\quad \forall(\neg b \cup \exists \diamond b)\)
(e) \(\exists \diamond g\)
(k) \(\forall(g \cup \forall(y \cup r))\)
(f) \(\exists \square g\)
(1) \(\forall(\neg b \cup b)\)

Exercise 6.2. Consider the following CTL formulae and the transition system TS outlined on the right:
\[
\begin{aligned}
& \Phi_{1}=\forall(a \cup b) \vee \exists \bigcirc(\forall \square b) \\
& \Phi_{2}=\forall \square \forall(a \cup b) \\
& \Phi_{3}=(a \wedge b) \rightarrow \exists \square \exists \bigcirc \forall(b \mathrm{~W} a) \\
& \Phi_{4}=\left(\forall \square \exists \diamond \Phi_{3}\right)
\end{aligned}
\]


Determine the satisfaction sets \(\operatorname{Sat}\left(\Phi_{i}\right)\) and decide whether \(T S \models \Phi_{i}(1 \leqslant i \leqslant 4)\).

Exercise 6.3. Which of the following assertions are correct? Provide a proof or a counterexample.
(a) If \(s \models \exists \square a\), then \(s \models \forall \square a\).
(b) If \(s \models \forall \square a\), then \(s \models \exists \square a\).
(c) If \(s \models \forall \diamond a \vee \forall \diamond b\), then \(s \models \forall \diamond(a \vee b)\).
(d) If \(s \models \forall \diamond(a \vee b)\), then \(s \models \forall \diamond a \vee \forall \diamond b\).
(e) If \(s \models \forall(a \cup b)\), then \(s \models \neg(\exists(\neg b \cup(\neg a \wedge \neg b)) \vee \exists \square \neg b)\).

Exercise 6.4. Let \(\Phi\) and \(\Psi\) be arbitrary CTL formulae. Which of the following equivalences for CTL formulae are correct?
(a) \(\forall \bigcirc \forall \diamond \Phi \equiv \forall \diamond \forall \bigcirc \Phi\)
(b) \(\exists \bigcirc \exists \diamond \Phi \equiv \exists \diamond \exists \bigcirc \Phi\)
(c) \(\forall \bigcirc \forall \square \Phi \equiv \forall \square \forall \bigcirc \Phi\)
(d) \(\exists \bigcirc \exists \square \Phi \equiv \exists \square \exists \bigcirc \Phi\)
(e) \(\exists \diamond \exists \square \Phi \equiv \exists \square \exists \diamond \Phi\)
(f) \(\forall \square(\Phi \Rightarrow(\neg \Psi \wedge \exists \bigcirc \Phi)) \equiv(\Phi \Rightarrow \neg \forall \diamond \Psi)\)
(g) \(\forall \square(\Phi \Rightarrow \Psi) \equiv(\exists \bigcirc \Phi \Rightarrow \exists \bigcirc \Psi)\)
(h) \(\neg \forall(\Phi \cup \Psi) \equiv \exists(\Phi \cup \neg \Psi)\)
(i) \(\exists((\Phi \wedge \Psi) \cup(\neg \Phi \wedge \Psi)) \equiv \exists(\Phi \cup(\neg \Phi \wedge \Psi))\)
(j) \(\forall(\Phi W \Psi) \equiv \neg \exists(\neg \Phi W \neg \Psi)\)
(k) \(\exists(\Phi \cup \Psi) \equiv \exists(\Phi \cup \Psi) \wedge \exists \diamond \Psi\)
(l) \(\exists(\Psi \mathrm{W} \neg \Psi) \vee \forall(\Psi \cup\) false \() \equiv \exists \bigcirc \Phi \vee \forall \bigcirc \neg \Phi\)
(m) \(\forall \square \Phi \wedge(\neg \Phi \vee \exists \bigcirc \exists \diamond \neg \Phi) \equiv \exists X \neg \Phi \wedge \forall \bigcirc \Phi\)
(n) \(\forall \square \forall \diamond \Phi \equiv \Phi \wedge(\forall \bigcirc \forall \square \forall \diamond \Phi) \vee \forall \bigcirc(\forall \diamond \Phi \wedge \forall \square \forall \diamond \Phi)\)
(o) \(\forall \square \Phi \equiv \Phi \vee \forall \bigcirc \forall \square \Phi\)

Exercise 6.5. Consider an elevator system that services \(N>0\) floors numbered 0 through \(N-1\). There is an elevator door at each floor with a call button and an indicator light that signals whether or not the elevator has been called. In the elevator cabin there are \(N\) send buttons (one per floor) and \(N\) indicator lights that inform to which floor(s) is going to be sent. For simplicity consider \(N=4\). Present a set of atomic propositions-try to minimize the number of propositions-that are needed to describe the following properties of the elevator system as CTL formulae and give the corresponding CTL formulae:
(a) The doors are "safe", i.e., a floor door is never open if the cabin is not present at the given floor.
(b) The indicator lights correctly reflect the current requests. That is, each time a button is pressed, there is a corresponding request that needs to be memorized until fulfillment (if ever).
(c) The elevator only services the requested floors and does not move when there is no request.
(d) All requests are eventually satisfied.

Exercise 6.6. Consider the single pulser circuit, a hardware circuit that is part of a set of benchmark circuits for hardware verification. The single pulser has the following informal specification: "For every pulse at the input inp there appears exactly one pulse of length 1 at output outp, independent of the length of the input pulse". Thus, the single pulser circuit is required to generate an output pulse between two rising edges of the input signal. The following questions require the formulation of the circuit in terms of CTL. Suppose we have the proposition rise_edge at our disposal which is true if the input was low (0) at time instant \(n-1\) and high (1) at time instant \(n\) (for natural \(n>0\) ). It is assumed that input sequences of the circuit are well behaved, i.e., more that the rising edge appears in the input sequence.

Questions: specify the following requirements of the circuit in CTL:
(a) A rising edge at the inputs leads to an output pulse.
(b) There is at most one output pulse for each rising edge.
(c) There is at most one rising edge for each output pulse.
(This exercise is taken from [246].)

Exercise 6.7. Transform the following CTL formulae into ENF and PNF. Show all intermediate steps.
\[
\begin{aligned}
& \Phi_{1}=\forall((\neg a) \mathrm{W}(b \rightarrow \forall \bigcirc c)) \\
& \Phi_{2}=\forall \bigcirc(\exists((\neg a) \cup(b \wedge \neg c)) \vee \exists \square \forall \bigcirc a)
\end{aligned}
\]

ExERCISE 6.8. Provide two finite transition systems \(T S_{1}\) and \(T S_{2}\) (without terminal states, and over the same set of atomic propositions) and a CTL formula \(\Phi\) such that \(\operatorname{Traces}\left(T S_{1}\right)=\) \(\operatorname{Traces}\left(T S_{2}\right)\) and \(T S_{1} \models \Phi\), but \(T S_{2} \not \models \Phi\).

Exercise 6.9. Consider the CTL formula
\[
\Phi=\forall \square(a \rightarrow \forall \diamond(b \wedge \neg a))
\]
and the following CTL fairness assumption:
\[
\text { fair }=\forall \diamond \forall \bigcirc(a \wedge \neg b) \rightarrow \forall \diamond \forall \bigcirc(b \wedge \neg a) \wedge \diamond \square \exists \diamond b \rightarrow \square \diamond b \text {. }
\]

Prove that \(T S \models_{\text {fair }} \Phi\) where transition system \(T S\) is depicted below.


Exercise 6.10. Let \(\wp=\left(z_{1}, z_{2}, z_{3}, z_{4}, z_{5}, z_{6}\right)\). Depict the \(\wp\)-ROBDD for the majority function
\[
\operatorname{MAJ}\left(\left[z_{1}=b_{1}, z_{2}=b_{2}, \ldots, z_{6}=b_{6}\right]\right)= \begin{cases}1 & \text { if } b_{1}+b_{2}+\ldots+b_{6} \geqslant 4 \\ 0 & \text { otherwise }\end{cases}
\]

Exercise 6.11. Consider the function
\[
F\left(x_{0}, \ldots, x_{n-1}, a_{0}, \ldots, a_{k-1}\right)=x_{m}
\]
where \(n=2^{k}\), and \(m=\sum_{j=0}^{k-1} a_{j} 2^{j}\). Let \(k=3\). Questions:
(a) Depict the \(\wp\)-ROBDD for \(\wp=\left(a_{0}, \ldots, a_{k-1}, x_{0}, \ldots, x_{n-1}\right)\).
(b) Depict the \(\wp\)-ROBDD for \(\wp=\left(a_{0}, x_{0}, \ldots, a_{k-1}, x_{k-1}, x_{k}, \ldots, x_{n-1}\right)\).

Exercise 6.12. Let \(\mathfrak{B}\) and \(\mathfrak{C}\) be two \(\wp\)-ROBDDs. Design an algorithm that checks whether \(f_{\mathfrak{B}}=f_{\mathfrak{C}}\) and runs in time linear in the sizes of \(\mathfrak{B}\) and \(\mathfrak{C}\).
(Hint: It is assumed that \(\mathfrak{B}\) and \(\mathfrak{C}\) are given as separate graphs (and not by nodes of a shared OBDD).)

Exercise 6.13. Let \(T S\) be a finite transition system (over AP) without terminal states, and \(\Phi\) and \(\Psi\) be CTL state formulae (over \(A P\) ). Prove or disprove
\[
T S \models \exists(\Phi \cup \Psi) \quad \text { if and only if } \quad T S^{\prime} \models \exists \diamond \Psi
\]
where \(T S^{\prime}\) is obtained from \(T S\) by eliminating all outgoing transitions from states \(s\) such that \(s \models \Psi \vee \neg \Phi\).

Exercise 6.14. Check for each of the following formula pairs ( \(\Phi_{i}, \varphi_{i}\) ) whether the CTL formula \(\Phi_{i}\) is equivalent to the LTL formula \(\varphi_{i}\). Prove the equivalence or provide a counterexample that illustrates why \(\Phi_{i} \not \equiv \varphi_{i}\).
(a) \(\Phi_{1}=\forall \square \forall \bigcirc a\). and \(\varphi_{1}=\square \bigcirc a\)
(b) \(\Phi_{2}=\forall \diamond \forall \bigcirc a\) and \(\varphi_{2}=\diamond \bigcirc a\).
(c) \(\Phi_{3}=\forall \diamond(a \wedge \exists \bigcirc a)\) and \(\varphi_{3}=\diamond(a \wedge \bigcirc a)\).
(d) \(\Phi_{4}=\forall \diamond a \quad \vee \quad \forall \diamond b\) and \(\varphi_{4}=\diamond(a \vee b)\).
(e) \(\Phi_{5}=\forall \square(a \rightarrow \forall \Delta b)\) and \(\varphi_{5}=\square(a \rightarrow \diamond b)\).
(f) \(\Phi_{6}=\forall(b \cup(a \wedge \forall \square b))\) and \(\varphi_{6}=\diamond a \wedge \square b\).

\section*{Exercise 6.15.}
(a) Prove, using Theorem 6.18, that there does not exist an equivalent LTL formula for the CTL formula \(\Phi_{1}=\forall \diamond(a \wedge \exists \bigcirc a)\).
(b) Now prove directly (i.e. without Theorem 6.18), that there does not exist an equivalent LTL formula for the CTL formula \(\Phi_{2}=\forall \diamond \exists \bigcirc \forall \diamond \neg a\). (Hint: Argument by contraposition.)

Exercise 6.16.

Consider the following CTL formulae
\[
\Phi_{1}=\exists \Delta \forall \square c \quad \text { and } \quad \Phi_{2}=\forall(a \cup \forall \diamond c)
\]
and the transition system \(T S\) outlined on the right. Decide whether \(T S \models \Phi_{i}\) for \(i=1,2\) using the CTL model-checking algorithm. Sketch its main steps.


Exercise 6.17. Provide an algorithm in pseudo code to compute \(\operatorname{Sat}(\forall(\Phi \cup \Psi))\) in a direct manner, i.e., without transforming the formula into ENF.

Exercise 6.18.
(a) Prove that \(\operatorname{Sat}(\exists(\Phi \mathrm{W} \Psi))\) is the largest set \(T\) such that
\[
T \subseteq \operatorname{Sat}(\Psi) \cup\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \cap T \neq \varnothing\} .
\]
(b) Prove that \(\operatorname{Sat}(\forall(\Phi \mathrm{W} \Psi))\) is the largest set \(T\) such that
\[
T \subseteq \operatorname{Sat}(\Psi) \cup\{s \in \operatorname{Sat}(\Phi) \mid \operatorname{Post}(s) \subseteq T\}
\]

Use the above characterizations to provide efficient algorithms for computing the sets \(\operatorname{Sat}(\exists(\Phi \mathrm{W} \Psi))\) and \(\operatorname{Sat}(\forall(\Phi W \Psi))\) in a direct manner.

Exercise 6.19. Consider the fragment ECTL of CTL which consists of formulae built according to the following grammar:
\[
\begin{aligned}
\Phi & ::=a|\neg a| \Phi \wedge \Phi \mid \exists \varphi \\
\varphi & ::=\bigcirc \Phi|\square \Phi| \Phi \cup \Phi
\end{aligned}
\]

For two transition systems \(T S_{1}=\left(S_{1}, A c t, \rightarrow_{1}, I_{1}, A P, L_{1}\right)\) and \(T S_{2}=\left(S_{2}, A c t, \rightarrow_{2}, I_{2}, A P, L_{2}\right)\), let \(T S_{1} \subseteq T S_{2}\) iff \(S_{1} \subseteq S_{2}, \rightarrow_{1} \subseteq \rightarrow_{2}, I_{1}=I_{2}\) and \(L_{1}(s)=L_{2}(s)\) for all \(s \in S_{1}\).
(a) Prove that for all ECTL formulae \(\Phi\) and all transition systems \(T S_{1}, T S_{2}\) with \(T S_{1} \subseteq T S_{2}\), it holds:
\[
T S_{1} \models \Phi \quad \Longrightarrow \quad T S_{2} \models \Phi .
\]
(b) Give a CTL formula which is not equivalent to any other ECTL formula. Justify your answer.

Exercise 6.20. In CTL, the release operator is defined by
\[
\exists(\Phi \mathrm{R} \Psi)=\neg \forall((\neg \Phi) \cup(\neg \Psi)) \quad \text { and } \quad \forall(\Phi \mathrm{R} \Psi)=\neg \exists((\neg \Phi) \cup(\neg \Psi)) .
\]
(a) Provide expansion laws for \(\exists(\Phi \mathrm{R} \Psi)\) and \(\forall(\Phi \mathrm{R} \Psi)\).
(b) Give a pseudo code algorithm for computing \(\operatorname{Sat}(\exists(\Phi \mathrm{R} \Psi))\) and do the same for computing \(\operatorname{Sat}(\forall(\Phi \mathrm{R} \Psi))\).

Exercise 6.21. Consider the CTL formula \(\Phi\) and the strong fairness assumption sfair:
\[
\begin{array}{ll}
\Phi & =\forall \square \forall \diamond a \\
\text { sfair } & =\square \diamond \underbrace{(b \wedge \neg a)}_{\Phi_{1}} \rightarrow \square \diamond \underbrace{\exists(b \cup(a \wedge \neg b))}_{\Psi_{1}}
\end{array}
\]
and transition system \(T S\) over \(A P=\{a, b\}\) which is given by


Questions:
(a) Determine \(\operatorname{Sat}\left(\Phi_{1}\right)\) and \(\operatorname{Sat}\left(\Psi_{1}\right)\) (without fairness).
(b) Determine \(\operatorname{Sat}_{\text {sfair }}(\exists \square\) true).
(c) Determine \(\operatorname{Sat}_{\text {sfair }}(\Phi)\).

Exercise 6.22. Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system without terminal states, \(a, b \in A P\) and \(s \in S\). Furthermore, let fair be a CTL fairness assumption and \(a_{f a i r} \in A P\) an atomic proposition with \(a_{\text {fair }} \in L(s)\) iff \(s \models_{\text {fair }} \exists \square\) true.

Which of the following assertions are correct? Give a proof or counterexample.
(a) \(s \models_{\text {fair }} \forall(a \cup b) \quad\) iff \(\quad s \models \forall\left(a \cup\left(b \wedge a_{\text {fair }}\right)\right)\).
(b) \(s \models_{\text {fair }} \exists(a \mathrm{~W} b) \quad\) iff \(s \models \exists\left(a \mathrm{~W}\left(b \wedge a_{\text {fair }}\right)\right)\).
(c) \(s \models_{\text {fair }} \forall(a \mathrm{~W} b) \quad\) iff \(\quad s \models \forall\left(a \mathrm{~W}\left(a_{\text {fair }} \rightarrow b\right)\right)\).

Exercise 6.23. Consider the following transition system \(T S\) over \(A P=\left\{a_{1}, \ldots, a_{6}\right\}\).


Let \(\Phi=\exists \bigcirc\left(a_{1} \rightarrow \exists\left(a_{1} \cup a_{2}\right)\right)\) and sfair \(=\) sfair \(_{1} \wedge\) sfair \(_{2} \wedge\) sfair \(_{3}\) a strong CTL fairness assumption where
\[
\begin{aligned}
& \text { sfair }_{1}=\square \diamond \forall \diamond\left(a_{1} \vee a_{3}\right) \longrightarrow \square \diamond a_{4} \\
& \text { sfair }_{2}=\square \diamond\left(a_{3} \wedge \neg a_{4}\right) \longrightarrow \square \diamond a_{5} \\
& \text { sfair }_{3}=\square \diamond\left(a_{2} \wedge a_{5}\right) \longrightarrow \square \diamond a_{6}
\end{aligned}
\]

Sketch the main steps for computing the satisfaction sets \(\operatorname{Sat}_{\text {sfair }}(\exists \square\) true \()\) and \(\operatorname{Sat}_{\text {sfair }}(\Phi)\).

Exercise 6.24. Consider the CTL* formula over \(A P=\{a, b\}:\)
\[
\Phi=\forall \diamond \square \exists \bigcirc(a \cup \exists \square b)
\]
and the transition system TS outlined below:


Apply the CTL* model-checking algorithm to compute \(\operatorname{Sat}(\Phi)\) and decide whether \(T S \vDash \Phi\). (Hint: You may infer the satisfaction sets for LTL formulae directly.)

EXERCISE 6.25. The model-checking algorithm presented in Section 6.5 treats CTL with strong fairness assumptions. Explain which modifications are necessary to deal with weak CTL fairness assumptions:
\[
\text { wfair }=\bigwedge_{1 \leqslant i \leqslant k}\left(\diamond \square a_{i} \longrightarrow \square \diamond b_{i}\right)
\]

You may assume that \(a_{i}, b_{i} \in\{\) true \(\} \cup A P\).

Exercise 6.26. Which of the following equivalences for CTL* are correct? Provide a proof or a counterexample.
(a) \(\forall \bigcirc \forall \square \Phi \equiv \forall \bigcirc \square \Phi\)
(b) \(\exists \bigcirc \exists \square \Phi \equiv \exists \bigcirc \square \Phi\)
(c) \(\forall(\varphi \wedge \psi) \equiv \forall \varphi \wedge \forall \psi\)
(d) \(\exists(\varphi \wedge \psi) \equiv \exists \varphi \wedge \exists \psi\)
(e) \(\neg \forall(\varphi \rightarrow \psi) \equiv \exists(\varphi \wedge \neg \psi)\)
(f) \(\exists \square \exists \bigcirc \bigcirc \neg \neg \bigcirc \neg \Phi \equiv \exists \square(\neg \bigcirc \neg \Phi)\)
(g) \(\forall(\diamond \Psi \wedge \square \Phi) \equiv \forall \diamond(\Psi \wedge \forall \square \Phi) \wedge \forall \square(\Phi \wedge \forall \diamond \Psi)\)
(h) \(\exists(\diamond \Psi \wedge \square \Phi) \equiv \exists \diamond(\Psi \wedge \exists \square \Phi)\)

Here, \(\Phi, \Psi\) are arbitrary CTL* state formulae and \(\psi, \varphi\) are CTL* path formulae.

\section*{Exercise 6.27.}

Consider the transition system TS and the CTL* formula
\[
\Phi=\exists(\bigcirc(a \wedge \neg b) \wedge \bigcirc \forall(b \cup \square a)) .
\]

Apply the CTL* model-checking algorithm to check whether \(T S \models \Phi\) and sketch its main steps as well as its output. (Hint: You may compute the LTL satisfaction sets directly.)


Exercise 6.28. Consider the transition system TS depicted below and the CTL* formula
\[
\Phi=\exists(\square \diamond b \wedge \square \exists \diamond \bigcirc a) \wedge \forall \square \diamond \bigcirc c
\]


Sketch the main steps that the CTL* model checking algorithm performs for checking whether \(T S \models \Phi\).

Exercise 6.29. Provide an example of a CTL* formula which is not a \(\mathrm{CTL}^{+}\)formula, but there exists an equivalent CTL formula.

ExERCISE 6.30. Provide equivalent CTL formulae for the CTL \({ }^{+}\)formulae \(\forall(\diamond a \wedge \square b)\) and \(\forall(\bigcirc a \wedge\) \(\neg(a \cup(\square b)))\).

\section*{Practical Exercises}

The remaining exercises are modeling and veriifcation exercises using the CTL model checker NuSMV [83].

Exercise 6.31. The following program is a mutual exclusion protocol for two processes due to Pnueli (taken from [118]). There is a single shared variable \(s\) which is either 0 or 1 , and initially equals 1. Besides, each process has a local Boolean variable \(y\) that initially equals 0 . The program text for process \(P_{i}(i=0,1)\) is as follows:
```

10: loop forever do
begin
11: Noncritical section
12:}(\mp@subsup{y}{i}{},s):=(1,i)
13: wait until ((y-i = 0) \vee (s\not=i));
14: Critical section
15: }\mp@subsup{y}{i}{}:=
end.

```

Here, the statement \(\left(y_{i}, s\right):=(1, i)\) is a multiple assignment in which variable \(y_{i}:=1\) and \(s:=i\) is a single, atomic step.

The intuition behind this protocol is as follows. The variables \(y_{0}\) and \(y_{1}\) are used by each process to signal the other process of active interest in entering the critical section. On leaving the noncritical section, process \(P_{i}\) sets its own local variable \(y_{i}\) to 1 . In a similar way this variable is reset to 0 once the critical section is left. The global variable \(s\) is used to resolve a tie situation between the processes. It serves as a logbook in which each process that sets its \(y\) variable to 1 signs at the same time. The test at line 13 says that \(P_{0}\) may enter its critical section if either \(y_{1}\) equals 0 implying that its competitor is not interested in entering its critical section - or if \(s\) differs from 0 - implying that the competitor process \(P_{1}\) performed its assignment to \(y_{1}\) after \(p_{0}\) assigned 1 to \(y_{0}\).

Questions concerning this mutual exclusion protocol:
(a) Model this protocol in NuSMV; formulate the property of mutual exclusion in CTL and check this property.
(b) Check whether Pnueli's protocol ensures absence of unbounded overtaking, i.e., when a process wants to enter its critical section, it eventually will be able to do so. Provide a counterexample (and an explanation thereof) in case this property is violated.
(c) Add the fairness constraint FAIRNESS running to the process specification in your NuSMV model of the mutual exclusion program, and check again the property of absence of unbounded overtaking. Compare the obtained results with the results obtained in the previous question without using the fairness constraint.
(d) Express in CTL that each process will occupy its critical section infinitely often. Check the property (use again the FAIRNESS running).
(e) A practical problem with this mutual exclusion protocol is that it is too demanding in the sense that it enforces performing the assignments to \(y_{i}\) and \(s\) (in line l2) in a single step. Most existing hardware systems cannot perform such assignments in one step. Therefore, it is requested to investigate whether any of the four possible realizations of this protocol - in which the aforementioned assignments are not atomic anymore - is a correct mutual exclusion protocol.
(I) Report for each possible implementation your results, including possible counterexamples and their explanation.
(II) Compare your results with the results of your PROMELA experiments with this exercise in the previous exercise series.

EXERCISE 6.32. In this exercise you are confronted with a nonstandard example for model checking. The purpose of this exercise is to present the model checker as a solver for combinatorial problems rather than a tool for correctness analysis. These problems involve a search (involving backtracking) of optimal or cost-minimizing strategies such as schedulers or puzzle solutions. The exercise is concerned with Loyd's puzzle that consists of an \(N \cdot K\) grid in which there are \(N \cdot K-1\) numbered tiles and a single blank space. The goal of the puzzle is to achieve a predetermined order on the tiles. The initial and final configuration of the tiles for \(N=3\) and \(K=3\) is as follows:

initial configuration

final configuration

Note that there are approximately \(4 \cdot(N \cdot K)\) ! possible moves in this puzzle. For \(N=3\) and \(K=3\) this already amounts to about \(1.4510^{6}\) possible configurations.

Questions concerning Loyd's puzzle:
(a) Complete the (partial) model of Loyd's puzzle in NuSMV that is given below. In this model, there is an array \(h\) that keeps track of the horizontal positions of the tiles and an array v that records the vertical positions of the tiles such that the position of tile \(i\) is given by the pair h[i], v[i]. Tile 0 represents the blank tile. Positionh[i] = 1 and \(v[i]=1\) is the lowest left corner of the puzzle.
```

MODULE main
DEFINE N := 3; K := 3;
VAR move: {u, d, l, r}; -- the possible tile-moves
h: array 0..8 of 1..3; -- the horizontal positions of all tiles
v: array 0..8 of 1..3; -- .... and their vertical positions
ASSIGN -- the initial horizontal and vertical positions of all tiles
init(h[0]) := 1; init(v[0]) := 3;
init(h[1]) := 2; init(v[1]) := 3;
init(h[2]) := 3; init(v[2]) := 3;
init(h[3]) := 1; init(v[3]) := 2;
init(h[4]) := 2; init(v[4]) := 2;
init(h[5]) := 3; init(v[5]) := 2;
init(h[6]) := 1; init(v[6]) := 1;
init(h[7]) := 2; init(v[7]) := 1;
init(h[8]) := 3; init(v[8]) := 1;
ASSIGN
-- determine the next positions of the blank tile
next(h[0]) := -- horizontal position of the blank tile
case
-- one position right
-- one position left
1 : h[0]; -- keep the same horizontal position
esac;
next(v[0]) := -- vertical position of the blank tile
case
-- one position down
-- one position up
1 : v[0]; -- keep the same vertical position
esac;
-- determine the next positions of all non-blank tiles
next(h[1]) := -- horizontal position of tile 1
case
esac;

```
```

next(v[1]) := -- vertical position of tile 1
case
esac;
-- and similar for all remaining tiles

```

A possible way to proceed is as follows:
(i) First, consider the possible moves of the blank tile (i.e., the blank space). Notice that the blank space cannot be moved to the left in all positions. The same applies to moves upward, downward and to the right.
(ii) Then try to find the possible moves of tile [1]. The code for tiles [2] through [8] are obtained by simply copying the code for tile [1] while replacing all references to [1] with references of the appropriate tile number.
(iii) Test the possible moves by running a simulation.
(b) Define an atomic proposition goal that describes the desired goal configuration of the puzzle. Add this definition to your NuSMV specification by incorporating the following line(s) in your NuSMV model:

DEFINE goal :=
where the dotted lines contain your description of the goal configuration.
(c) Find a solution to the puzzle by imposing the appropriate CTL formula to the NuSMV specification, and running the model checker on this formula.

This exercise has been taken from [95].

Exercise 6.33. Consider the mutual exclusion algorithm by the Dutch mathematician Dekker. There are two processes \(P_{1}\) and \(P_{2}\), two Boolean-valued variables \(b_{1}\) and \(b_{2}\) whose initial values are false, and a variable \(k\) which may take the values 1 and 2 and whose initial value is arbitrary. The \(i\) th process \((i=1,2)\) may be described as follows, where \(j\) is the index of the other process:
```

while true do
begin }\mp@subsup{b}{i}{}:=\mathrm{ true;
while }\mp@subsup{b}{j}{}\mathrm{ do
if k=j then begin
b
while }k=j\mathrm{ do skip;
b
end;
\langle critical section \rangle;
k:= j;
b}:==\mathrm{ false
end

```

Questions:
(a) Model Dekker's algorithm in NuSMV.
(b) Verify whether this algorithm satisfies the following properties:
(c) Mutual exclusion: two processes cannot be in their critical section at the same time.
(d) Absence of individual starvation: if a process wants to enter its critical section, it is eventually able to do so.
(Hint: use the FAIRNESS running statement in your NuSMV specification for proving the latter property in order to prohibit unfair executions that might trivially violate these requirements.)

EXERCISE 6.34. In the original mutual exclusion protocol by Dijkstra in 1965 (another Dutch mathematician), it is assumed that there are \(n \geqslant 2\) processes, and global variables \(b, c\) : array \([1 \ldots n]\) of Boolean and an integer \(k\). Initially all elements of \(b\) and of \(c\) have the value true and the value of \(k\) belongs to \(1,2, \ldots, n\). The \(i\) th process may be represented as follows:
```

var $j$ : integer;
while true do
begin $b[i]:=$ false;
$L_{i}:$ if $k \neq i$ then begin $c[i]:=$ true;
if $b[k]$ then $k:=i$;
goto $L_{i}$
end;
else begin $c[i]:=$ false;
for $j:=1$ to $n$ do
if $(j \neq i \wedge \neg(c[j]))$ then goto $L_{i}$
end
〈 critical section 〉;
$c[i]:=$ true;
$b[i]:=$ true
end

```

Questions:
(a) Model this algorithm in NuSMV.
(b) Check the mutual exclusion property (at most one process can be in its critical section at any point in time) in two different ways: by means of a CTL formula using SPEC and by using invariants. Try to check this property for \(n=2\) through \(n=5\) by increasing the number of processes gradually and compare the sizes of the state spaces and the runtime needed for the two ways of verifying the mutual exclusion property.
(c) Check the absence of individual starvation property: if a process wants to enter its critical section, it is eventually able to do so.

Exercise 6.35. In order to find a fair solution for \(N\) processes, Peterson proposed in 1981 the following protocol. Let \(Q[1 \ldots N]\) ( \(Q\) for queue) and \(T[1 \ldots N]\) ( \(T\) for turn), be two shared arrays which are initially 0 and 1 , respectively. The variables \(i\) and \(j\) are local to the process with \(i\) containing the process number. The code of process \(i\) is as follows:
```

while true do
for $j:=1$ to $N-1$ do
begin
$Q[i]:=j ;$
$T[j]:=i$;
wait until $(T[j] \neq i \vee($ forall $k \neq i . Q[k]<j))$
end;
$\langle$ critical section $\rangle$;
$Q[i]:=0$
end

```

Questions:
(a) Model Peterson's algorithm in NuSMV.
(b) Verify whether this algorithm satisfies the following properties:
(i) Mutual exclusion.
(ii) Absence of individual starvation.

\section*{Chapter 7}

\section*{Equivalences and Abstraction}

Transition systems can model a piece of software or hardware at various abstraction levels. The lower the abstraction level, the more implementation details are present. At high abstraction levels, such details are deliberately left unspecified. Binary relations between states (henceforth implementation relations) are useful to relate or to compare transition systems, possibly at different abstraction levels. When two models are related, one model is said to be refined by the other, or, reversely, the second is said to be an abstraction of the first. If the implementation relation is an equivalence, then it identifies all transition systems that cannot be distinguished; such models fulfill the same observable properties at the relevant abstraction level.

Implementation relations are predominantly used for comparing two models of the same system. Given a transition system \(T S\) that acts as an abstract system specification, and a more detailed system model \(T S^{\prime}\), implementation relations allow checking whether \(T S^{\prime}\) is a correct implementation (or: refinement) of TS. Alternatively, for system analysis purposes, implementation relations provide ample means to abstract from certain system details, preferably details that are irrelevant for the analysis of the property, \(\varphi\) say, at hand. In this way, a transition system \(T S^{\prime}\) comprising very many, maybe even infinitely many, states can be abstracted by a smaller model TS. Provided the abstraction preserves the property to be checked, the analysis of the (hopefully small) abstract model TS suffices to decide the satisfaction of the properties in \(T S^{\prime}\). Formally, from \(T S \models \varphi\) we may safely conclude \(T S^{\prime}=\varphi\).

This chapter will introduce several implementation relations, ranging from very strict ones ("strong" relations) that require transition systems to mutually mimic each transition, to more liberal ones ("weak" relations) in which this is only required for certain transitions.

In fact, in this monograph we have already encountered implementation relations that were aimed at comparing transition systems by considering their linear-time behavior, i.e., (finite or infinite) traces. Examples of such relations are trace inclusion and trace equivalence. Linear-time properties or LTL formulae are preserved by trace relations based on infinite traces; for trace-equivalent \(T S\) and \(T S^{\prime}\) and linear-time property \(P\), we have \(T S^{\prime} \vDash P\) whenever \(T S \vDash P\). A similar result is obtained when replacing \(P\) by an LTL formula.

Besides the introduction of a weak variant of trace equivalence, the goal of this chapter is primarily to study relations that respect the branching-time behavior. Classical representatives of such relations are bisimulation equivalences and simulation preorder relations. Whereas bisimulation relates states that mutually mimic all individual transitions, simulation requires that one state can mimic all stepwise behavior of the other, but not the reverse. Weak variants of these relations only require this for certain ("observable") transitions, and not for other ("silent") transitions. This chapter will formally define strong and weak variants of (bi)simulation, and will treat their relationship to trace-based relations. The preservation of CTL and CTL* formulae is shown; for bisimulation the truth value of all such formulae is preserved, while for simulation this applies to a (large) fragment thereof. These results provide us with the tools to simplify establishing \(T S^{\prime} \models \varphi\), for CTL (or CTL*) formula \(\varphi\) by checking whether \(T S \models \varphi\).

This provides the theoretical underpinning of exploiting bisimulation and simulation relations for the purpose of abstraction. The remaining issue is how to obtain the more abstract TS from the (larger, and more concrete) transition system \(T S^{\prime}\). This chapter will treat polynomial-time algorithms for several notions of (bi)simulation. These algorithms allow checking whether two given transition systems are (bi)similar, and can be used to generate an abstract transition system from a more concrete one in an automated manner.

As in the previous chapters, we will follow the state-based approach. This means that we consider branching-time relations that refer to the state labels, i.e., the atomic propositions that hold in the states. Action labels of the transitions are not considered. All concepts (definitions, theorems, algorithms) treated in this chapter can, however, easily be reformulated for the approach that is focused on action labels rather than state labels. This connection is discussed in Section 7.1.2.

Throughout this chapter, the transition systems under consideration may have terminal states, and hence, may exhibit both finite and infinite paths and traces. When considering temporal logics, though, transition systems without terminal states (thus only having infinite behaviors) are assumed.

\subsection*{7.1 Bisimulation}

Bisimulation equivalence aims to identify transition systems with the same branching structure, and which thus can simulate each other in a stepwise manner. Roughly speaking, a transition system \(T S^{\prime}\) can simulate transition system \(T S\) if every step of \(T S\) can be matched by one (or more) steps in \(T S^{\prime}\). Bisimulation equivalence denotes the possibility of mutual, stepwise simulation. We first introduce bisimulation equivalence as a binary relation between transition systems (over the same set of atomic propositions); later on, bisimulation is also treated as a relation between states of a single transition system. Bisimulation is defined coinductively, i.e., as the largest relation satisfying certain properties.

\section*{Definition 7.1. Bisimulation Equivalence}

Let \(T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right), i=1,2\), be transition systems over AP. A bisimulation for \(\left(T S_{1}, T S_{2}\right)\) is a binary relation \(\mathcal{R} \subseteq S_{1} \times S_{2}\) such that
(A) \(\forall s_{1} \in I_{1}\left(\exists s_{2} \in I_{2} .\left(s_{1}, s_{2}\right) \in \mathcal{R}\right)\) and \(\forall s_{2} \in I_{2}\left(\exists s_{1} \in I_{1} .\left(s_{1}, s_{2}\right) \in \mathcal{R}\right)\)
(B) for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) it holds:
(1) \(L_{1}\left(s_{1}\right)=L_{2}\left(s_{2}\right)\)
(2) if \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) then there exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\)
(3) if \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) then there exsist \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
\(T S_{1}\) and \(T S_{2}\) are bisimulation-equivalent (bisimilar, for short), denoted \(T S_{1} \sim T S_{2}\), if there exists a bisimulation \(\mathcal{R}\) for \(\left(T S_{1}, T S_{2}\right)\).

Condition (A) asserts that every initial state of \(T S_{1}\) is related to an initial state of \(T S_{2}\), and vice versa. According to condition (B.1), the states \(s_{1}\) and \(s_{2}\) are equally labeled. This can be considered as ensuring the "local" equivalence of \(s_{1}\) and \(s_{2}\). Condition (B.2) states that every outgoing transition of \(s_{1}\) must be matched by an outgoing transition of \(s_{2}\); the reverse is stated by (B.3). Figure 7.1 summarises the latter two conditions.

\section*{Example 7.2. Two Beverage Vending Machines}

Let \(A P=\{\) pay, beer, soda \(\}\). Consider the transition systems depicted in Figure 7.2. These model a beverage vending machine, but differ in the number of possibilities for supplying beer. The fact that the right-hand transition system \(\left(T S_{2}\right)\) has an additional option to deliver beer is not observable by a user. This suggests an equivalence between


Figure 7.1: Conditions (B.2) and (B.3) of bisimulation equivalence.


Figure 7.2: Two bisimilar beverage vending machines
the transition systems. Indeed, the equivalence of \(T S_{1}\) and \(T S_{2}\) follows from the fact that the relation
\[
\mathcal{R}=\left\{\left(s_{0}, t_{0}\right),\left(s_{1}, t_{1}\right),\left(s_{2}, t_{2}\right),\left(s_{2}, t_{3}\right),\left(s_{3}, t_{4}\right)\right\}
\]
is a bisimulation for \(\left(T S_{1}, T S_{2}\right)\). It can easily be verified that \(\mathcal{R}\) indeed satisfies all requirements of Definition 7.1.

Now consider an alternative model \(\left(T S_{3}\right)\) of the vending machine where the user selects the drink on inserting a coin, see Figure 7.3, depicting \(T S_{1}\) (left) and \(T S_{3}\) (right). AP is as before. It follows that \(T S_{1}\) and \(T S_{3}\) are not bisimilar, since the state \(s_{1}\) in \(T S_{1}\) cannot be mimicked by a state in \(T S_{3}\). This can be seen as follows. Due to the labeling condition (B.1), the only candidates for mimicking state \(s_{1}\) are the states \(u_{1}\) and \(u_{2}\) in \(T S_{3}\). However, neither of these states can mimic all transitions of \(s_{1}\) in \(T S_{1}\) : either the possibility for soda or for beer is missing. Thus, \(T S_{1} \nsim T S_{3}\) for \(A P=\{\) pay, beer, soda \(\}\).


Figure 7.3: Nonbisimulation-equivalent beverage vending machines.

As a final example, reconsider \(T S_{1}\) and \(T S_{3}\) for \(A P=\{\) pay, drink \(\}\). The labelings of the transition systems are obvious: \(L\left(s_{0}\right)=L\left(u_{0}\right)=\{\) pay \(\}, L\left(s_{1}\right)=L\left(u_{1}\right)=L\left(u_{2}\right)=\varnothing\), and all remaining states are labeled with \(\{d r i n k\}\). It can now be established that the relation
\[
\left\{\left(s_{0}, u_{0}\right),\left(s_{1}, u_{1}\right),\left(s_{1}, u_{2}\right),\left(s_{2}, u_{3}\right),\left(s_{2}, u_{4}\right),\left(s_{3}, u_{3}\right),\left(s_{3}, u_{4}\right)\right\}
\]
is a bisimulation for \(\left(T S_{1}, T S_{3}\right)\). Henceforth, \(T S_{1} \sim T S_{3}\) for \(A P=\{\) pay, drink \(\}\).

\section*{Remark 7.3. The Relevant Set of Atomic Propositions}

The fixed set AP plays a crucial role in comparing transition systems using bisimulation. Intuitively, AP stands for the set of all "relevant" atomic propositions. All other atomic propositions are understood as negligible and are ignored in the comparison. In case \(T S\) is a refinement of \(T S^{\prime}\), e.g., \(T S\) is obtained from \(T S^{\prime}\) by incorporating some implementation details, then the set \(A P\) of atomic propositions of \(T S\) generally is a proper superset of the set \(A P^{\prime}\) of propositions of \(T S^{\prime}\). To compare \(T S\) and \(T S^{\prime}\), the set of common atomic propositions, \(A P^{\prime}\), is a reasonable choice. In this way, it is possible to check whether the branching structure of \(T S\) agrees with that of \(T S^{\prime}\) when considering all observable information in \(A P^{\prime}\). If we are only interested in checking the equivalence of \(T S\) and \(T S^{\prime}\) with respect to the satisfaction of a temporal logic formula \(\Phi\), it suffices to consider \(A P\) as the atomic propositions that occur in \(\Phi\).

\section*{Lemma 7.4. Reflexivity, Transitivity, and Symmetry of \(\sim\)}

For a fixed set AP of atomic propositions, the relation \(\sim\) is an equivalence relation.

Proof: Let \(A P\) be a set of atomic propositions. We show reflexivity, symmetry, and transitivity of \(\sim\) :
- Reflexivity: For transition system \(T S\) with state space \(S\), the identity relation \(\mathcal{R}=\) \(\{(s, s) \mid s \in S\}\) is a bisimulation for (TS, TS).
- Symmetry: Sssume that \(\mathcal{R}\) is a bisimulation for \(\left(T S_{1}, T S_{2}\right)\). Consider
\[
\mathcal{R}^{-1}=\left\{\left(s_{2}, s_{1}\right) \mid\left(s_{1}, s_{2}\right) \in \mathcal{R}\right\}
\]
that is obtained by swapping the states in any pair in \(\mathcal{R}\). Clearly, relation \(\mathcal{R}^{-1}\) satisfies conditions (A) and (B.1). By symmetry of (B.2) and (B.3), we immediately conclude that \(\mathcal{R}^{-1}\) is a bisimulation for \(\left(T S_{2}, T S_{1}\right)\).
- Transitivity: Let \(\mathcal{R}_{1,2}\) and \(\mathcal{R}_{2,3}\) be bisimulations for \(\left(T S_{1}, T S_{2}\right)\) and \(\left(T S_{2}, T S_{3}\right)\), respectively. The relation \(\mathcal{R}=\mathcal{R}_{1,2} \circ \mathcal{R}_{2,3}\), given by
\[
\mathcal{R}=\left\{\left(s_{1}, s_{3}\right) \mid \exists s_{2} \in S_{2} .\left(s_{1}, s_{2}\right) \in \mathcal{R}_{1,2} \wedge\left(s_{2}, s_{3}\right) \in \mathcal{R}_{2,3}\right\},
\]
is a bisimulation for \(\left(T S_{1}, T S_{3}\right)\) where \(S_{2}\) denotes the set of states in \(T S_{2}\). This can be seen by checking all conditions for a bisimulation.
(A) Consider the initial state \(s_{1}\) of \(T S_{1}\). Since \(\mathcal{R}_{1,2}\) is a bisimulation, there is an initial state \(s_{2}\) of \(T S_{2}\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}_{1,2}\). As \(\mathcal{R}_{2,3}\) is a bisimulation, there is an initial state \(s_{3}\) of \(T S_{3}\) with \(\left(s_{2}, s_{3}\right) \in \mathcal{R}_{2,3}\). Thus, \(\left(s_{1}, s_{3}\right) \in \mathcal{R}\). In the same way we can check that for any initial state \(s_{3}\) of \(T S_{3}\), there is an initial state \(s_{1}\) of \(T S_{1}\) with \(\left(s_{1}, s_{3}\right) \in \mathcal{R}\).
(B.1) By definition of \(\mathcal{R}\), there is a state \(s_{2}\) in \(T S_{2}\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}_{1,2}\) and \(\left(s_{2}, s_{3}\right) \in\) \(\mathcal{R}_{2,3}\). Then, \(L_{1}\left(s_{1}\right)=L_{2}\left(s_{2}\right)=L_{3}\left(s_{3}\right)\).
(B.2) Assume \(\left(s_{1}, s_{3}\right) \in \mathcal{R}\). As \(\left(s_{1}, s_{2}\right) \in \mathcal{R}_{1,2}\), it follows that if \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) then \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}_{1,2}\) for some \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\). Since \(\left(s_{2}, s_{3}\right) \in \mathcal{R}_{2,3}\), we have \(\left(s_{2}^{\prime}, s_{3}^{\prime}\right) \in\) \(\mathcal{R}_{2,3}\) for some \(s_{3}^{\prime} \in \operatorname{Post}\left(s_{3}\right)\). Hence, \(\left(s_{1}^{\prime}, s_{3}^{\prime}\right) \in \mathcal{R}\).
(B.3) Similar to the proof for (B.2).

Bisimulation is defined in terms of the direct successors of states. By using an inductive argument over states, we obtain a relation between (finite or infinite) paths.

\section*{Lemma 7.5. Bisimulation on Paths}

Let \(T S_{1}\) and \(T S_{2}\) be transition systems over AP, \(\mathcal{R}\) a bisimulation for \(\left(T S_{1}, T S_{2}\right)\), and \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\). Then for each (finite or infinite) path \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots \in \operatorname{Paths}\left(s_{1}\right)\) there exists a path \(\pi_{2}=s_{0,2} s_{1,2} s_{2,2} \ldots \in \operatorname{Paths}\left(s_{2}\right)\) of the same length such that \(\left(s_{j, 1}, s_{j, 2}\right) \in \mathcal{R}\) for all \(j\).


Figure 7.4: Construction of statewise bisimilar paths.

Proof: Let \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots \in \operatorname{Paths}\left(s_{1}\right)\) be a maximal path fragment in \(T S_{1}\) starting in \(s_{1}=s_{0,1}\) and assume \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\). We successively define a "corresponding" maximal path fragment in \(T S_{2}\) starting in \(s_{2}=s_{0,2}\), where the transitions \(s_{i, 1} \rightarrow_{1} s_{i+1,1}\) are matched by transitions \(s_{i, 2} \rightarrow_{2} s_{i+1,2}\) such that \(\left(s_{i+1,1}, s_{i+1,2}\right) \in \mathcal{R}\). This is done by induction on \(i\), see Figure 7.4 on page 455 . For each case we distinguish between \(s_{i}\) being a terminal state or not.
- Base case: \(i=0\). In case \(s_{1}\) is a terminal state, it follows directly from \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) (by condition (B.3)) that \(s_{2}\) is a terminal state too. Thus \(s_{2}=s_{0,2}\) is a maximal path fragment in \(T S_{2}\). If \(s_{1}\) is not a terminal state, it follows from \(\left(s_{0,1}, s_{0,2}\right)=\left(s_{1}, s_{2}\right) \in \mathcal{R}\) that the transition \(s_{1}=s_{0,1} \rightarrow_{1} s_{1,1}\) can be matched by a transition \(s_{2} \rightarrow_{2} s_{1,2}\) such that \(\left(s_{1,1}, s_{1,2}\right) \in \mathcal{R}\). This yields the path fragment \(s_{2} s_{1,2}\) in \(T S_{2}\).
- Induction step: Assume \(i \geqslant 0\) and that the path fragment \(s_{2} s_{1,2} s_{2,2} \ldots s_{i, 2}\) is already constructed with \(\left(s_{j, 1}, s_{j, 2}\right) \in \mathcal{R}\) for \(j=1, \ldots, i\).

If \(\pi_{1}\) has length \(i\), then \(\pi_{1}\) is maximal and \(s_{i, 1}\) is a terminal state. By condition (B.3), \(s_{i, 2}\) is terminal too. Thus, \(\pi_{2}=s_{2} s_{1,2} s_{2,2} \ldots s_{i, 2}\) is a maximal path fragment in \(T S_{2}\) which is statewise related to \(\pi_{1}=s_{1} s_{1,1} s_{2,1} \ldots s_{i, 1}\).

Now assume that \(s_{i, 1}\) is not terminal. We consider the step \(s_{i, 1} \rightarrow_{1} s_{i+1,1}\) in \(\pi_{1}\). Since \(\left(s_{i, 1}, s_{i, 2}\right) \in \mathcal{R}\), there exists a transition \(s_{i, 2} \rightarrow_{2} s_{i+1,2}\) with \(\left(s_{i+1,1}, s_{i+1,2}\right) \in \mathcal{R}\). This yields a path fragment \(s_{2} s_{1,2} \ldots s_{i, 2} s_{i+1,2}\) which is statewise related to the prefix \(s_{1} s_{1,1} \ldots s_{i, 1} s_{i+1,1}\) of \(\pi_{1}\).

By symmetry, for each path \(\pi_{2} \in \operatorname{Paths}\left(s_{2}\right)\) there exists a path \(\pi_{1} \in \operatorname{Paths}\left(s_{1}\right)\) of the same length which is statewise related to \(\pi_{2}\). As one can construct statewise bisimilar paths, bisimilar transition systems are trace-equivalent. It is mostly easier to prove that two transition systems are bisimilar rather than prove their trace equivalence. The intuitive
reason for this discrepancy is that proving bisimulation equivalence just requires "local" reasoning about state behavior instead of considering entire paths. The following result is thus of importance in checking trace equivalence as well.

\section*{Theorem 7.6. Bisimulation and Trace Equivalence}
\(T S_{1} \sim T S_{2}\) implies \(\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)\).

Proof: Let \(\mathcal{R}\) be a bisimulation for \(\left(T S_{1}, T S_{2}\right)\). By Lemma 7.5, any path \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots\) in \(T S_{1}\) can be lifted to a path \(\pi_{2}=s_{0,2} s_{1,2} s_{2,2} \ldots\) in \(T S_{2}\) such that \(\left(s_{i, 1}, s_{i, 2}\right) \in \mathcal{R}\) for all indices \(i\). According to condition (B.1), \(L_{1}\left(s_{i, 1}\right)=L_{2}\left(s_{i, 2}\right)\) for all \(i\). Thus, \(\operatorname{trace}\left(\pi_{1}\right)=\operatorname{trace}\left(\pi_{2}\right)\). This shows \(\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right)\). By symmetry, one obtains that \(T S_{1}\) and \(T S_{2}\) are trace equivalent.

As trace-equivalent transition systems fulfill the same linear-time properties, it thus now follows that bisimilar transition systems fulfill the same linear-time properties.

\subsection*{7.1.1 Bisimulation Quotient}

So far, bisimulation has been defined as a relation between transition systems. This enables comparing different transition systems. An alternative perspective is to consider bisimulation as a relation between states within a single transition system. By considering the quotient transition system under such a relation, smaller models are obtained. This minimization recipe can be exploited for efficient model checking. In the following, we define bisimulation as a relation on states, relate it to the notion of bisimulation between transition systems, and define the quotient transition system under such relation.

\section*{Definition 7.7. Bisimulation Equivalence as Relation on States}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system. A bisimulation for \(T S\) is a binary relation \(\mathcal{R}\) on \(S\) such that for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) :
1. \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
2. If \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\), then there exists an \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
3. If \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\), then there exists an \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).

States \(s_{1}\) and \(s_{2}\) are bisimulation-equivalent (or bisimilar), denoted \(s_{1} \sim_{T S} s_{2}\), if there exists a bisimulation \(\mathcal{R}\) for \(T S\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).

Thus, a bisimulation (on states) for \(T S\) is a bisimulation (on transition systems) for the pair (TS, TS ), except that condition (A) is not required. This condition could be ensured by adding the pairs \((s, s)\) to \(\mathcal{R}\) for any state \(s\). Moreover, for all states \(s_{1}\) and \(s_{2}\) in \(T S\) it holds that
\[
\underbrace{s_{1} \sim_{\text {TS }} s_{2}}_{\text {as states of } T S \text { (Def. } 7.7 \text { ) }} \text { iff } \underbrace{T S_{s_{1}} \sim T S_{s_{2}}}_{\text {in the sense of Def. } 7.1}
\]
where \(T S_{s_{i}}\) denotes the transition system obtained from \(T S\) by declaring \(s_{i}\) as the unique initial state. Vice versa, the definition of bisimulation between transition systems (Definition 7.1) arises from Definition 7.7 as follows. Take transition systems \(T S_{1}\) and \(T S_{2}\) over \(A P\), and combine them in a single transition system \(T S_{1} \oplus T S_{2}\), which basically results from the disjoint union of state spaces (see below). We then subsequently "compare" the initial states of \(T S_{1}\) and \(T S_{2}\) as states of the composite transition system \(T S_{1} \oplus T S_{2}\) to ensure condition (A).

The formal definition of \(T S_{1} \oplus T S_{2}\) is as follows. For \(T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right)\), \(i=1,2\) :
\[
T S_{1} \oplus T S_{2}=\left(S_{1} \uplus S_{2}, A c t_{1} \cup A c t_{2}, \rightarrow_{1} \cup \rightarrow_{2}, I_{1} \cup I_{2}, A P, L\right)
\]
where \(\uplus\) stands for disjoint union and where \(L(s)=L_{i}(s)\) if \(s \in S_{i}\). Then \(T S_{1} \sim T S_{2}\) if and only if, for every initial state \(s_{1}\) of \(T S_{1}\), there exists a bisimilar initial state \(s_{2}\) of \(T S_{2}\), and vice versa. That is, \(s_{1} \sim_{T S_{1} \oplus T S_{2}} s_{2}\). Stated in terms of equivalence classes, \(T S_{1} \sim T S_{2}\) if and only if
\[
\forall C \in\left(S_{1} \uplus S_{2}\right) / \sim_{T S_{1} \oplus T S_{2}} . I_{1} \cap C \neq \varnothing \quad \text { iff } \quad I_{2} \cap C \neq \varnothing .
\]

Here, \(\left(S_{1} \uplus S_{2}\right) / \sim_{T S_{1} \oplus T S_{2}}\) denotes the quotient space with respect to \(\sim_{T S_{1} \oplus T S_{2}}\), i.e., the set of all bisimulation equivalence classes in \(S_{1} \uplus S_{2}\). The latter observation is based on the fact that \(\sim_{T S_{1} \oplus T S_{2}}\) is an equivalence relation, see the first part of the following lemma.

\section*{Lemma 7.8. Coarsest Bisimulation}

For transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) it holds that:
1. \(\sim_{T S}\) is an equivalence relation on \(S\).
2. \(\sim_{\text {TS }}\) is a bisimulation for TS.
3. \(\sim_{T S}\) is the coarsest bisimulation for \(T S\).

Proof: The first claim follows directly from the characterization of \(\sim_{T S}\) in terms of \(\sim\), and Lemma 7.4 on page 453. The last claim states that each bisimulation \(\mathcal{R}\) for \(T S\) is
finer than \(\sim_{T S}\); this immediately follows from the definition of \(\sim_{T S}\). It remains to prove that \(\sim_{T S}\) is a bisimulation for \(T S\). We show that \(\sim_{T S}\) satisfies conditions (1) and (2) of Definition 7.7. Condition (3) follows by symmetry. Let \(s_{1} \sim_{T S} s_{2}\). Then, there exists a bisimulation \(\mathcal{R}\) that contains \(\left(s_{1}, s_{2}\right)\). From condition (1), it follows that \(L\left(s_{1}\right)=L\left(s_{2}\right)\). Condition (2) yields that for any transition \(s_{1} \rightarrow s_{1}^{\prime}\) there is a transition \(s_{2} \rightarrow s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\). Hence, \(s_{1}^{\prime} \sim_{T S} s_{2}^{\prime}\).

Stated differently, the relation \(\sim_{T S}\) is the coarsest equivalence on the state space of \(T S\) such that equivalent states are equally labeled and can simulate each other as shown in Figure 7.5.


Figure 7.5: Condition (2) for bisimulation equivalence \(\sim_{T S}\) on states.

\section*{Remark 7.9. Union of Bisimulations}

For finite index set \(I\) and \(\left(\mathcal{R}_{i}\right)_{i \in I}\) a family of bisimulation relations for \(T S, \bigcup_{i \in I} \mathcal{R}_{i}\) is a bisimulation for \(T S\) too (see Exercise 7.2). Since \(\sim_{T S}\) is the coarsest bisimulation for TS, \(\sim_{T S}\) coincides with the union of all bisimulation relations for \(T S\).

As stated before, bisimilar transition systems satisfy the same linear-time properties. Such properties - and, as we will see later, all temporal formulae that can be expressed in CTL* - can thus be checked on the quotient system instead of on the original (and possibly much larger) transition system. Before providing the definition of quotient transition systems with respect to \(\sim_{T S}\), let us first fix some notations.

\section*{Notation 7.10. Equivalence Classes, Quotient Space}

Let \(S\) be a set and \(\mathcal{R}\) an equivalence on \(S\). For \(s \in S,[s]_{\mathcal{R}}\) denotes the equivalence class of state \(s\) under \(\mathcal{R}\), i.e., \([s]_{\mathcal{R}}=\left\{s^{\prime} \in S \mid\left(s, s^{\prime}\right) \in \mathcal{R}\right\}\). Note that for \(s^{\prime} \in[s]_{\mathcal{R}}\) we have \(\left[s^{\prime}\right]_{\mathcal{R}}=[s]_{\mathcal{R}}\). The set \([s]_{\mathcal{R}}\) is often referred to as the \(\mathcal{R}\)-equivalence class of \(s\). The quotient space of \(S\) under \(\mathcal{R}\), denoted by \(S / \mathcal{R}=\left\{[s]_{\mathcal{R}} \mid s \in S\right\}\), is the set consisting of all \(\mathcal{R}\)-equivalence classes.

\section*{Definition 7.11. Bisimulation Quotient}

For transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) and bisimulation \(\sim_{T S}\), the quotient transition system \(T S / \sim_{T S}\) is defined by
\[
T S / \sim_{T S}=\left(S / \sim_{T S},\{\tau\}, \rightarrow^{\prime}, I^{\prime}, A P, L^{\prime}\right)
\]
where:
\[
\text { - } I^{\prime}=\left\{[s]_{\sim} \mid s \in I\right\}
\]
- \(\rightarrow^{\prime}\) is defined by
\[
\frac{s \xrightarrow{\alpha} s^{\prime}}{[s]_{\sim} \xrightarrow{\tau}\left[s^{\prime}\right]_{\sim}},
\]
- \(L^{\prime}\left([s]_{\sim}\right)=L(s)\).

In the sequel, \(T S / \sim_{T S}\) is referred to as the bisimulation quotient of \(T S\). For the sake of simplicity, we write \(T S / \sim\) rather than \(T S / \sim_{T S}\).

The state space of \(T S / \sim\) is the quotient of \(S\) under \(\sim\). The initial states in \(T S / \sim\) are the \(\sim\)-equivalence classes of the initial states in \(T S\). Each transition \(s \rightarrow s^{\prime}\) in \(T S\) induces a transition \([s]_{\sim} \rightarrow^{\prime}\left[s^{\prime}\right]_{\sim}\). As the action label is irrelevant, these labels are omitted from now on; this is reflected in the definition by replacing any action \(\alpha \in\) Act by an arbitrary action, \(\tau\) say. The state-labeling function \(L^{\prime}\) is well-defined, since all states in \([s]_{\sim}\) are equally labeled (see the definition of bisimulation). According to the definition of \(\rightarrow^{\prime}\) it follows for any \(B, C \in S / \sim\) :
\[
B \rightarrow^{\prime} C \quad \text { if and only if } \quad \exists s \in B . \exists s^{\prime} \in C . s \rightarrow s^{\prime}
\]

By condition (2) of Definition 7.7, this is equivalent to
\[
B \rightarrow^{\prime} C \quad \text { if and only if } \forall s \in B . \exists s^{\prime} \in C . s \rightarrow s^{\prime}
\]

The following two examples show that enormous state-space reductions may be obtained by considering the bisimulation quotient. In some cases, it even allows obtaining a finite quotient transition system for infinite transition systems.

\section*{Example 7.12. Many Printers}

Consider a system consisting of \(n\) printers, each represented as extremely simplified by two states, ready and print. The initial state is ready, and once started, each printer alternates between being ready and printing. The entire system is given by
\[
T S_{n}=\underbrace{\text { Printer }\| \| \ldots \| \text { Printer }}_{n \text { times }}
\]

Assume the states of \(T S_{n}\) are labeled with atomic propositions from the set \(A P=\) \(\{0,1, \ldots, n\}\). Intuitively, \(L(s)\) denotes the number of printers available in state \(s\), i.e., which are in the local state ready. The number of states of \(T S_{n}\) is exponential in \(n\) (it is \(2^{n}\) ); for \(n=3, T S_{n}\) is depicted in Figure 7.6, where \(r\) denotes ready, and \(p\) denotes print. The quotient transition system \(T S_{n} / \sim\), however, only contains \(n+1\) states. For \(n=3\), \(T S_{n} / \sim\) is depicted in Figure 7.7.


Figure 7.6: Transition system \(T S_{3}\) for three independent printers


Figure 7.7: Bisimulation quotient \(T S_{3} \nsim\).

\section*{Example 7.13. The Bakery Algorithm}

We consider a mutual exclusion algorithm, originally proposed by Lamport, which is known as the Bakery algorithm. Although the principle of the Bakery algorithm allows guaranteeing mutual exclusion for an arbitrary number of processes, we consider the simpler setting with two processes. Let \(P_{1}\) and \(P_{2}\) be two processes, and \(x_{1}\) and \(x_{2}\) be two shared variables that both initially equal zero, see the following program text:
```

Process 1:
while true {
n
w1: wait until ( }\mp@subsup{x}{2}{}=0|\mp@subsup{|}{1}{}<\mp@subsup{x}{2}{})
c
x := 0;
}

```
Process 2:
```

while true {
x := x
wait until ( }\mp@subsup{x}{1}{}=0|\mp@subsup{x}{2}{}<\mp@subsup{x}{1}{})
...critical section ...}
x2 := 0;
.....
}

```

These variables are used to resolve a conflict if both processes want to enter the critical section. (One might consider the value of a variable as a ticket, i.e., a number that one typically gets upon entering a bakery. The holder of the lowest number is the next customer to be served.) If process \(P_{i}\) is waiting, and \(x_{i}<x_{j}\) or \(x_{j}=0\), then it may enter the critical section. We have \(x_{i}>0\) whenever process \(P_{i}\) is either waiting to acquire access to the critical section, or is in the critical section. On requesting access to the critical section, process \(P_{i}\) sets \(x_{i}\) to \(x_{j}+1\), where \(i \neq j\). Intuitively, process \(P_{i}\) gives priority to its opponent, process \(P_{j}\).

As the value of \(x_{1}\) and \(x_{2}\) may grow unboundedly, the underlying transition system of the parallel composition of \(P_{1}\) and \(P_{2}\) is infinite; a fragment of the transition system is depicted in Figure 7.8. An example of a path fragment that visits infinitely many different states is:
\begin{tabular}{|l|l|c|c|l|}
\hline process \(P_{1}\) & process \(P_{2}\) & \(x_{1}\) & \(x_{2}\) & effect \\
\hline \hline noncrit \(_{1}\) & noncrit \(_{2}\) & 0 & 0 & \(P_{1}\) requests access to critical section \\
wait \(_{1}\) & noncrit \(_{2}\) & 1 & 0 & \(P_{2}\) requests access to critical section \\
wait \(_{1}\) & wait \(_{2}\) & 1 & 2 & \(P_{1}\) enters the critical section \\
crit \(_{1}\) & wait \(_{2}\) & 1 & 2 & \(P_{1}\) leaves the critical section \\
noncrit \(_{1}\) & wait \(_{2}\) & 0 & 2 & \(P_{1}\) requests access to critical section \\
wait \(_{1}\) & wait \(_{2}\) & 3 & 2 & \(P_{2}\) enters the critical section \\
wait \(_{1}\) & crit \(_{2}\) & 3 & 2 & \(P_{2}\) leaves the critical section \\
wait \(_{1}\) & noncrit \(_{2}\) & 3 & 0 & \(P_{2}\) requests access to critical section \\
wait \(_{1}\) & wait \(_{2}\) & 3 & 4 & \(P_{2}\) enters the critical section \\
\(\ldots\) & \(\ldots\) & \(\ldots\) &.. & \(\ldots\) \\
\hline
\end{tabular}

Although algorithmic analysis, such as LTL model checking, is impossible due to the infinity of the transition system, it is not difficult to check that the Bakery algorithm suffers neither from deadlock nor starvation:
- A deadlock only occurs whenever none of the processes can enter the critical section, i.e., if \(x_{1}=x_{2}>0\). It is easy to see, however, that apart from the initial situation, we always have \(x_{1} \neq x_{2}\).
- Starvation only occurs if a process that wants to enter the critical section is never able to do so. Such situation, however, can never occur: in case both processes want to enter the critical section, it is impossible that a process acquires access to the critical section twice in a row.

Alternatively, the correctness of the Bakery algorithm can also be established algorithmically. This is done by considering an abstraction of the infinite transition system such that, instead of the concrete values of \(x_{1}\) and \(x_{2}\), it is only recorded whether
\[
x_{1}>x_{2}>0 \quad \text { or } \quad x_{2}>x_{1}>0 \quad \text { or } \quad x_{1}=0 \quad \text { or } \quad x_{2}=0
\]

Note that this information is sufficient to determine which process may acquire access to the critical section. By means of this data abstraction, we obtain a finite transition system \(T S_{B a k}^{a b s t r a c t}\), e.g., the infinite set of states for which \(x_{1}\) and \(x_{2}\) exceed 0 is mapped onto the single abstract state \(\left\langle\right.\) wait \(_{1}\), wait \(\left.{ }_{2}, x_{1}>x_{2}>0\right\rangle\). When considering the atomic propositions
\[
\left\{\text { noncrit }_{i}, \text { wait }_{i}, \text { crit }_{i} \mid i=1,2\right\} \cup\left\{x_{1}>x_{2}>0, x_{2}>x_{1}>0, x_{1}=0, x_{2}=0\right\}
\]
the finite transition system \(T S_{\text {Bak }}^{\text {abstract }}\) (with the obvious state labeling) is trace-equivalent to the original infinite transition system \(T S_{B a k}\). Due to the fact that trace-equivalent transition systems satisfy the same LT properties, it follows that each LT property that is
shown to hold for the finite (abstract) transition system also holds for the original infinite transition system! The following LT properties, expressed as LTL formulae, indeed hold for \(T S_{\text {Bak }}^{\text {abstract }}\) :
\[
\square\left(\neg \text { crit }_{1} \vee \neg c \text { rit }_{2}\right) \quad \text { and } \quad\left(\square \diamond \text { wait }_{1} \Rightarrow \square \diamond \text { crit }_{1}\right) \wedge\left(\square \diamond \text { wait }_{2} \Rightarrow \square \diamond \text { crit }_{2}\right) .
\]


Figure 7.8: Fragment of the infinite transition system of the Bakery algorithm.

Transition systems \(T S_{\text {Bak }}\) and \(T S_{\text {Bak }}^{\text {abstract }}\) are bisimilar. This can be easily proven by indicating a bisimulation relation. The data abstraction described above is formalized by means of the function \(f: S \rightarrow S^{\prime}\) where \(S\) and \(S^{\prime}\) denote the set of reachable states of \(T S_{B a k}\) and \(T S_{B a k}^{a b s t r a c t}\), respectively. The function \(f\) associates to any reachable state \(s\) of \(T S_{\text {Bak }}\), an (abstract) state \(f(s)\) of \(T S_{\text {Bak }}^{\text {abstract }}\). Let \(s=\left\langle\ell_{1}, \ell_{2}, x_{1}=b_{1}, x_{2}=b_{2}\right\rangle\) be a state of \(T S_{\text {Bak }}\) with \(\ell_{i} \in\left\{\right.\) noncrit \(_{i}\), wait \(t_{i}\), crit \(\left.{ }_{i}\right\}\) and \(b_{i} \in \mathbb{N}\) for \(i=1,2\). Then, we define
\[
f(s)= \begin{cases}\left\langle\ell_{1}, \ell_{2}, x_{1}=0, x_{2}=0\right\rangle & \text { if } b_{1}=b_{2}=0 \\ \left\langle\ell_{1}, \ell_{2}, x_{1}=0, x_{2}>0\right\rangle & \text { if } b_{1}=0 \text { and } b_{2}>0 \\ \left\langle\ell_{1}, \ell_{2}, x_{1}>0, x_{2}=0\right\rangle & \text { if } b_{1}>0 \text { and } b_{2}=0 \\ \left\langle\ell_{1}, \ell_{2}, x_{1}>x_{2}>0\right\rangle & \text { if } b_{1}>b_{2}>0 \\ \left\langle\ell_{1}, \ell_{2}, x_{2}>x_{1}>0\right\rangle & \text { if } b_{2}>b_{1}>0\end{cases}
\]


Figure 7.9: Bisimulation quotient transition system of the Bakery algorithm

It follows by straightforward reasoning that \(\mathcal{R}=\{(s, f(s)) \mid s \in S\}\) is a bisimulation for \(\left(T S_{\text {Bak }}, T S_{B a k}^{a b s t r a c t}\right)\) for any subset of \(A P=\left\{\right.\) noncrit \(_{i}\), wait \({ }_{i}\), crit \(\left._{i} \mid i=1,2\right\}\). The transition system \(T S_{B a k}^{a b s t r a c t}\) in Figure 7.9 above is the bisimulation quotient system
\[
T S_{B a k}^{a b s t r a c t}=T S_{B a k} / \sim \text { for } A P=\left\{\text { crit }_{1}, \text { crit }_{2}\right\} .
\]

Whereas the original transition system is infinite, its bisimulation quotient is finite.

Theorem 7.14. Bisimulation Equivalence of TS and TS/~
For any transition system TS it holds that TS \(\sim T S / \sim\).

Proof: Follows immediately from the fact that \(\left\{\left(s, s^{\prime}\right) \mid s^{\prime} \in[s]_{\sim}, s \in S\right\}\) is a bisimulation for \((T S, T S / \mathcal{R})\) in the sense of Definition 7.1 (page 451).

In general, the quotient transition system \(T S / \mathcal{R}\) with respect to a bisimulation \(\mathcal{R}\) contains more states than \(T S / \sim\) since \(\sim\) is the coarsest bisimulation relation. It is often simple to manually indicate a (meaningful) bisimulation, while the computation of the quotient space \(S / \sim\) requires an algorithmic analysis of the complete transition system (see Section 7.3 on page 476 and further). Therefore, it may be advantageous to generate the quotient system \(T S / \mathcal{R}\) instead of \(T S / \sim\).

\subsection*{7.1.2 Action-Based Bisimulation}

As the prime interest of this monograph is model checking, the notions of bisimulation are all focused on the state labelings; labels of transitions are simply ignored. In other contexts, most notably process algebras, analogous notions of bisimulation are considered that do the reverse - these notions ignore state labelings and are focused on transitions labels, i.e. actions. The aim of this subsection is to relate these notions.

\section*{Definition 7.15. Action-Based Bisimulation Equivalence}

Let \(T S_{i}=\left(S_{i}, A c t, \rightarrow_{i}, I_{i}, A P_{i}, L_{i}\right), i=1,2\), be transition systems over the set Act of actions. An action-based bisimulation for \(\left(T S_{1}, T S_{2}\right)\) is a binary relation \(\mathcal{R} \subseteq S_{1} \times S_{2}\) such that
(A) \(\forall s_{1} \in I_{1} \exists s_{2} \in I_{2} .\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(\forall s_{2} \in I_{2} \exists s_{1} \in I_{1} .\left(s_{1}, s_{2}\right) \in \mathcal{R}\)
(B) for any \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) it holds that
(2') if \(s_{1} \xrightarrow{\alpha} 1 s_{1}^{\prime}\), then \(s_{2} \xrightarrow{\alpha} 2 s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\) for some \(s_{2}^{\prime} \in S_{2}\)
\(\left(3^{\prime}\right)\) if \(s_{2} \xrightarrow{\alpha} 2 s_{2}^{\prime}\), then \(s_{1} \xrightarrow{\alpha} 1 s_{1}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\), for some \(s_{1}^{\prime} \in S_{1}^{\prime}\).
\(T S_{1}\) and \(T S_{2}\) are action-based bisimulation equivalent (or action-based bisimilar), denoted \(T S_{1} \sim^{\text {Act }} T S_{2}\), if there exists an action-based bisimulation \(\mathcal{R}\) for \(\left(T S_{1}, T S_{2}\right)\).

Action-based bisimulation differs from the variant for state labels (see Definition 7.1) in the following way: the state-labeling condition (B.1) is reformulated by means of the transition labels, and thus is encoded by conditions (B.2') and (B.3'). All results and concepts presented for \(\sim\) can be adapted for \(\sim^{A c t}\) in a straightforward manner. For instance, \(\sim^{A c t}\) is an equivalence and can be adapted to an equivalence \(\sim_{T S}^{A c t}\) for the states of a single transition system \(T S\) in a similar way as before. The action-based bisimulation quotient system \(T S / \sim^{A c t}\) is defined as \(T S / \sim\) except that we deal with an empty set of atomic propositions and lift any transition \(s \xrightarrow{\alpha} s^{\prime}\) to an equally labeled transition \(B \xrightarrow{\alpha} B^{\prime}\)
where \(B\) and \(B^{\prime}\) denote the action-based bisimulation equivalence classes of states \(s\) and \(s^{\prime}\), respectively.

In the context of process calculi, an important aspect of bisimulation is whether it enjoys a substitutivity property with respect to syntactic operators in the process calculus, such as parallel composition. The following result states that action-based bisimulation is a congruence for the parallel composition \(\|_{H}\) with synchronization over handshaking actions , see Definition 2.26 on page 48 .

\section*{Lemma 7.16. Congruence w.r.t. Handshaking}

For transition systems \(T S_{1}, T S_{1}^{\prime}\) over Act \(1_{1}, T S_{2}, T S_{2}^{\prime}\) over Act \({ }_{2}\), and \(H \subseteq A c t_{1} \cap A c t_{2}\) it holds that
\[
T S_{1} \sim^{\text {Act }} T S_{1}^{\prime} \quad \text { and } \quad T S_{2} \sim^{\text {Act }} T S_{2}^{\prime} \quad \text { implies } T S_{1}\left\|_{H} T S_{2} \sim^{\text {Act }} T S_{1}^{\prime}\right\|_{H} T S_{2}^{\prime}
\]

Proof: Let \(T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right)\) and \(T S_{i}^{\prime}=\left(S_{i}^{\prime}, A c t_{i}, \rightarrow_{i}^{\prime}, I_{i}^{\prime}, A P, L_{i}^{\prime}\right)\) and let \(\mathcal{R}_{i} \subseteq S_{i} \times S_{i}^{\prime}\) be an action-based bisimulation for \(\left(T S_{i}, T S_{i}^{\prime}\right), i=1,2\). Then, the relation:
\[
\mathcal{R}=\left\{\left(\left\langle s_{1}, s_{2}\right\rangle,\left\langle s_{1}^{\prime}, s_{2}^{\prime}\right\rangle\right) \mid\left(s_{1}, s_{1}^{\prime}\right) \in \mathcal{R}_{1} \wedge\left(s_{2}, s_{2}^{\prime}\right) \in \mathcal{R}_{2}\right\}
\]
is an action-based bisimulation for \(\left(T S_{1}\left\|_{H} T S_{2}, T S_{1}^{\prime}\right\|_{H} T S_{2}^{\prime}\right)\). This can be seen as follows. The fulfillment of condition (A) is obvious. To check condition (B.2'), assume that (1) there is a transition \(\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha}\left\langle t_{1}, t_{2}\right\rangle\) in \(T S_{1} \|_{H} T S_{2}\), and (2) \(\left(\left\langle s_{1}, s_{2}\right\rangle,\left\langle t_{1}, t_{2}\right\rangle\right)\) in \(\mathcal{R}\). Distinguish two cases.
1. \(\alpha \in\) Act \(\backslash H\). Then \(\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha}\left\langle t_{1}, t_{2}\right\rangle\) arises by an individual move of either \(T S_{1}\) or \(T S_{2}\). By symmetry, we may assume that \(s_{1} \xrightarrow{\alpha}{ }_{1} t_{1}\) and \(s_{2}=t_{2}\). Since \(\left(s_{1}, s_{1}^{\prime}\right)\) belongs to bisimulation \(\mathcal{R}_{1}\), there exists a transition \(s_{1}^{\prime} \xrightarrow{\alpha}{ }_{1}^{\prime} t_{1}^{\prime}\) in \(T S_{1}^{\prime}\) with \(\left(t_{1}, t_{1}^{\prime}\right) \in\) \(\mathcal{R}_{1}\). Thus, \(\left\langle s_{1}^{\prime}, s_{2}^{\prime}\right\rangle \xrightarrow{\alpha}\left\langle t_{1}^{\prime}, s_{2}^{\prime}\right\rangle\) is a transition in \(T S_{1}^{\prime} \|_{H} T S_{2}^{\prime}\) and \(\left(\left\langle t_{1}, s_{2}\right\rangle,\left\langle t_{1}^{\prime}, s_{2}^{\prime}\right\rangle\right) \in \mathcal{R}\).
2. \(\alpha \in H\). Then \(\left\langle s_{1}, s_{2}\right\rangle \xrightarrow{\alpha}\left\langle t_{1}, t_{2}\right\rangle\) arises by synchronizing transitions in \(T S_{1}\) and \(T S_{2}\). That is, \(s_{1} \xrightarrow{\alpha}{ }_{1} t_{1}\) is a transition in \(T S_{1}\) and \(s_{2} \xrightarrow{\alpha}{ }_{2} t_{2}\) a transition in \(T S_{2}\). Since \(\left(s_{i}, s_{i}^{\prime}\right) \in \mathcal{R}_{i}\), there exists a transition \(s_{i}^{\prime} \xrightarrow{\alpha}{ }_{i}^{\prime} t_{i}^{\prime}\) in \(T S_{i}^{\prime}(\) for \(i=1,2)\) with \(\left(t_{i}, t_{i}^{\prime}\right) \in \mathcal{R}_{i}\). Thus, \(\left\langle s_{1}^{\prime}, s_{2}^{\prime}\right\rangle \xrightarrow{\alpha}\left\langle t_{1}^{\prime}, t_{2}^{\prime}\right\rangle\) is a transition in \(T S_{1}^{\prime} \|_{H} T S_{2}^{\prime}\) and \(\left(\left\langle t_{1}, t_{2}\right\rangle,\left\langle t_{1}^{\prime}, t_{2}^{\prime}\right\rangle\right) \in \mathcal{R}\).

The fulfillment of condition (B.3') follows by a symmetric argument.

Let us now consider the relation between state-based and action-based bisimulation in more detail. We first discuss how an action-based bisimulation can be obtained from
a state-based bisimulation, and the reverse. This is done for transition system \(T S=\) \((S, A c t, \rightarrow, I, A P, L)\).

Consider the bisimulation \(\sim_{T S}\) on \(S\). The intention is to define a transition system
\[
T S_{a c t}=\left(S_{a c t}, A c t, \rightarrow{ }_{a c t}, I_{a c t}, A P_{a c t}, L_{a c t}\right)
\]
such that \(\sim_{T S}\) and the action-based bisimulation \(\sim_{T S}^{A c t}\) coincide. As our interest is in actionbased bisimulation on \(T S_{a c t}, A P_{a c t}\) and \(L_{a c t}\) are irrelevant, and can be taken as \(A P\) and \(L\), respectively. Let \(S_{a c t}=S \cup\{t\}\) where \(t\) is a new state (i.e., \(t \notin S\) ). \(T S_{a c t}\) has the same initial states as \(T S\), i.e., \(I_{a c t}=I\), and is equipped with the action set Act \(=2^{A P} \cup\{\tau\}\). The transition relation \(\rightarrow_{a c t}\) is given by the rules:
\[
\frac{s \longrightarrow s^{\prime}}{s \xrightarrow{L(s)} \text { act } s^{\prime}} \quad \text { and } \quad \frac{s \text { is a terminal state in } T S}{s \xrightarrow{\tau}{ }_{a c t} t}
\]

Thus, the new state \(t\) serves to characterize the terminal states in \(T S\). For bisimulation \(\mathcal{R}\) for \(T S, \mathcal{R}_{\text {act }}=\mathcal{R} \cup\{(t, t)\}\) is an action-based bisimulation for \(T S_{\text {act }}\). Vice versa, for action-based bisimulation \(\mathcal{R}_{\text {act }}\) for \(T S_{a c t}, \mathcal{R}_{\text {act }} \cap(S \times S)\) is a bisimulation for \(T S\). Thus, for all states \(s_{1}, s_{2} \in S\) :
\[
s_{1} \sim_{T S} s_{2} \quad \text { if and only if } s_{1} \sim_{T S} A_{\text {act }} s_{2}
\]

Let us now consider the reverse direction. Consider the action-based bisimulation \(\sim_{T S}^{A c t}\) on \(S\). The intention is to define a transition system
\[
T S_{\text {state }}=\left(S_{\text {state }}, A c t_{\text {state }}, \rightarrow_{\text {state }}, I_{\text {state }}, A P_{\text {state }}, L_{\text {state }}\right)
\]
such that \(\sim_{T S}^{A c t}\) and the bisimulation \(\sim_{T S_{s t a t e}}\) coincide. As our interest is in a state-based bisimulation, the action-set Act \(_{\text {state }}\) is not of importance. Let \(S_{\text {state }}=S \cup(S \times A c t)\) where it is assumed that \(S \cap(S \times A c t)=\varnothing\). (Such construction is also used to compare action-based vs. state-based fairness on page 263) Take \(I_{\text {state }}=I\). The actions of \(T S\) serve as atomic propositions in \(T S_{\text {state }}\), i.e., \(A P_{\text {state }}=A c t\). The labeling function of \(L_{\text {state }}\) is defined by \(L(s)=\varnothing\) and \(L(\langle s, \alpha\rangle)=\{\alpha\}\). The transition relation \(\rightarrow_{\text {state }}\) is defined by the rules:
\[
\frac{s \xrightarrow{\alpha} s^{\prime}}{s \longrightarrow \text { state }\left\langle s^{\prime}, \alpha\right\rangle} \quad \text { and } \quad \frac{s \xrightarrow{\alpha} s^{\prime}, \beta \in A c t}{\langle s, \beta\rangle \longrightarrow \text { state }\left\langle s^{\prime}, \alpha\right\rangle}
\]

That is, state \(\langle s, \alpha\rangle\) in \(T S_{\text {state }}\) serves to simulate state \(s\) in \(T S\). In fact, the second component \(\alpha\) indicates the action via which \(s\) is entered. It now follows that for all states \(s_{1}, s_{2} \in S\) (see Exercise 7.5):
\[
s_{1} \sim_{T S}^{A c t} s_{2} \quad \text { if and only if } s_{1} \sim_{T S_{s t a t e}} s_{2}
\]

\subsection*{7.2 Bisimulation and CTL* Equivalence}

This section considers the equivalence relations induced by the temporal logics CTL and CTL* and discusses their connection to bisimulation equivalence. As in the chapters on temporal logic, this section is restricted to transition systems that do not have any terminal states. These transition systems thus only have infinite paths.

States in a transition system are equivalent with respect to a logic whenever these states cannot be distinguished by the truth value of any of such formulae. Stated differently, whenever there is a formula in the logic that holds in one state, but not in the other, these states are not equivalent.

\section*{Definition 7.17. CTL* Equivalence}

Let \(T S, T S_{1}\), and \(T S_{2}\) be transition systems over \(A P\) without terminal states.
1. States \(s_{1}, s_{2}\) in \(T S\) are \(C T L^{*}\)-equivalent, denoted \(s_{1} \equiv_{C T L^{*}} s_{2}\), if
\[
s_{1} \models \Phi \quad \text { iff } \quad s_{2} \models \Phi \quad \text { for all CTL* }{ }^{*} \text { state formulae over } A P .
\]
2. \(T S_{1}\) and \(T S_{2}\) are \(C T L^{*}\)-equivalent, denoted \(T S_{1} \equiv_{C T L^{*}} T S_{2}\), if
\[
T S_{1} \models \Phi \quad \text { iff } \quad T S_{2} \models \Phi \quad \text { for all CTL* state formulae over } A P .
\]

States \(s_{1}\) and \(s_{2}\) are CTL* equivalent if there does not exist a CTL* state formula that holds in \(s_{1}\) and not in \(s_{2}\), or, vice versa, holds in \(s_{2}\), but not in \(s_{1}\). This definition can easily be adapted to any subset of CTL*, e.g., \(s_{1}\) and \(s_{2}\) are CTL equivalent, denoted \(s_{1} \equiv{ }_{C T L} s_{2}\), if for all CTL formulae \(\Phi\) over \(A P,\left\{s_{1}, s_{2}\right\} \subseteq \operatorname{Sat}(\Phi)\) or \(\left\{s_{1}, s_{2}\right\} \cap \operatorname{Sat}(\Phi)=\varnothing\). Similarly, \(s_{1}\) and \(s_{2}\) are LTL-equivalent, denoted \(s_{1} \equiv_{L T L} s_{2}\), if \(s_{1}\) and \(s_{2}\) satisfy the same LTL formula over \(A P\).

The fact that trace-equivalent transition systems satisfy the same LTL formulae (see Theorem 3.15 on page 104) can now be stated as follows:

Theorem 7.18. Trace Equivalence is Finer Than LTL Equivalence \(\equiv_{\text {trace }} \subseteq \equiv_{L T L}\).

\section*{Remark 7.19. Distinguishing Nonequivalent States by Formulae}

Let \(T S\) be a transition system without terminal states and \(s_{1}, s_{2}\) states in \(T S\). If \(s_{1} \not \equiv 三_{C T L}\) \(s_{2}\), then there exists a CTL state formula \(\Phi\) with \(s_{1} \models \Phi\) and \(s_{2} \not \models \Phi\). This follows from the definition of CTL equivalence which yields the existence of a formula \(\Phi\) with (i) \(s_{1} \models \Phi\) and \(s_{2} \not \vDash \Phi\) or (ii) \(s_{1} \not \models \Phi\) and \(s_{2} \models \Phi\). In case (ii), one may switch from \(\Phi\) to \(\neg \Phi\) and obtain \(s_{1} \models \neg \Phi\) and \(s_{2} \not \vDash \neg \Phi\).

A corresponding result holds for CTL*, but not for LTL. This can be seen as follows. Assume that \(s_{1} \not 三_{L T L} s_{2}\) and \(\operatorname{Traces}\left(s_{1}\right)\) is a proper subset of \(\operatorname{Traces}\left(s_{2}\right)\). Then, all LTL formulae that hold for \(s_{2}\) also hold for \(s_{1}\). However, since there are traces in \(\operatorname{Traces}\left(s_{2}\right)\) that are not in \(\operatorname{Traces}\left(s_{1}\right)\), there exists an LTL formula \(\varphi\) (e.g., characterizing such trace) with \(s_{2} \models \varphi\), but \(s_{1} \not \models \varphi\).

The main result of this section is stated in Theorem 7.20 (see below). It asserts that CTL equivalence, CTL* equivalence, and bisimulation equivalence coincide. This theorem has various important consequences. First, and for all, it relates the notion of bisimulation equivalence to logical equivalences. As a result, bisimulation equivalence preserves all formulae that can be formulated in CTL* or CTL. In principle this result allows performing model checking on the bisimulation quotient transition system-assuming that we can obtain this in an algorithmic manner-while preserving both affirmative and negative outcomes of the model checking. If a CTL* formula holds for the quotient, it also holds for the original transition system. Moreover, if the formula is refuted by the quotient, the original transition system refutes it too. On the other hand, it indicates that a single CTL* formula that holds for one but not for the other state suffices to show the nonbisimilarity of the states.

Secondly, the fact that CTL and CTL* equivalence coincide is perhaps surprising as CTL* is strictly more expressive than CTL as CTL* subsumes LTL, but the expressivenesses of CTL and LTL are incomparable, see Theorem 6.21 on page 337. So, although the expressiveness of CTL* is strictly larger than that of CTL, their logical equivalence is the same. In particular, to show CTL* equivalence it suffices to show CTL equivalence!

\section*{Theorem 7.20. \(C T L^{*} / C T L\) and Bisimulation Equivalence}

For finite transition system TS without terminal states:
\[
\sim_{T S}=\equiv_{C T L}=\equiv_{C T L^{*}}
\]

Proof: The proof of this central result is divided into three steps. First, it is shown that CTL equivalence is finer than bisimulation equivalence, i.e., \(\equiv_{C T L} \subseteq \sim_{T S}\); see Lemma
7.21 below. Subsequently, it is proven that bisimulation equivalence is finer than CTL* equivalence, i.e., \(\sim_{T S} \subseteq \equiv_{C T L^{*}}\); see Lemma 7.26 on page 473 . The obvious fact that CTL* equivalence is finer than CTL equivalence (since CTL* subsumes CTL) completes the proof. Summarizing, we have:
\[
\underbrace{\sim_{T S} \subseteq \equiv_{C T L^{*}}}_{\text {Lemma } 7.26} \quad \text { and } \quad \equiv_{C T L^{*}} \subseteq \equiv_{C T L} \quad \text { and } \quad \underbrace{\equiv_{C T L} \subseteq \sim_{T S}}_{\text {Lemma } 7.21}
\]

\section*{Lemma 7.21. CTL Equivalence is Finer Than Bisimulation}

For finite transition system TS without terminal states, and \(s_{1}, s_{2}\) states in TS:
\[
s_{1} \equiv{ }_{C T L} s_{2} \quad \text { implies } \quad s_{1} \sim_{T S} s_{2} .
\]

Proof: It suffices to show that the relation:
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid s_{1} \equiv_{C T L} s_{2}\right\}
\]
is a bisimulation for \(T S\). This is proven by checking the conditions of being a bisimulation (see Definition 7.7 on page 456). As \(\mathcal{R}\) is, by definition, an equivalence relation, it suffices to consider the first two conditions. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\), i.e., \(s_{1} \equiv_{C T L} s_{2}\).
1. Consider the following CTL state formula \(\Phi\) over \(A P\) :
\[
\Phi=\bigwedge_{a \in L\left(s_{1}\right)} a \wedge \bigwedge_{a \in A P \backslash L\left(s_{1}\right)} \neg a .
\]

Clearly, \(s_{1} \models \Phi\). Since \(s_{1} \equiv{ }_{C T L} s_{2}\), it follows that \(s_{2} \models \Phi\). As \(\Phi\) characterizes the labeling of state \(s_{1}\), we immediately get \(L\left(s_{1}\right)=L\left(s_{2}\right)\). Thus, condition (1) of Definition 7.7 holds.
2. For any equivalence class \(C \in S / \mathcal{R}\), let CTL formula \(\Phi_{C}\) be such that
\[
\begin{equation*}
\operatorname{Sat}\left(\Phi_{C}\right)=C \tag{*}
\end{equation*}
\]
\(\Phi_{C}\) is obtained as follows. For every pair \((C, D)\) of equivalence classes \(C, D \in S / \mathcal{R}\) with \(C \neq D\), let CTL formula \(\Phi_{C, D}\) be such that \(\operatorname{Sat}\left(\Phi_{C, D}\right) \supseteq C\) and \(\operatorname{Sat}\left(\Phi_{C, D}\right) \cap\) \(D=\varnothing\). Since \(T S\) is finite, there are only finitely many equivalence classes under \(\mathcal{R}\). Hence, \(\Phi_{C}\) can be defined as the conjunction of all formulae \(\Phi_{C, D}\) :
\[
\Phi_{C}=\bigwedge_{\substack{D \in S / \mathcal{R} \\ D \neq C}} \Phi_{C, D}
\]

Clearly, condition (*) is satisfied.
Let \(B \in S / \mathcal{R}\) and \(s_{1}, s_{2} \in B\), i.e., \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(B\) is the equivalence class of \(s_{1}, s_{2}\) with respect to \(\mathcal{R}\). Further, let \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) and \(C\) be the equivalence class of \(s_{1}^{\prime}\) with respect to \(\mathcal{R}\), i.e., \(C=\left[s_{1}^{\prime}\right]_{\mathcal{R}}\). Now we show the existence of a transition \(s_{2} \rightarrow s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
Due to \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right) \cap C\) and \((*)\), we have \(s_{1} \vDash \exists \bigcirc \Phi_{C}\). Since \(s_{1} \equiv_{C T L} s_{2}\), we get \(s_{2} \models \exists \bigcirc \Phi_{C}\). Thus, there is a state \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(s_{2}^{\prime} \models \Phi_{C}\). But then equation \((*)\) yields \(s_{2}^{\prime} \in C\). Since \(C=\left[s_{1}^{\prime}\right]_{\mathcal{R}}\), we obtain \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).

\section*{Remark 7.22. Master Formulae}

The proof of Lemma 7.21 relies on establishing so-called master formulae \(\Phi_{C}\) for equivalence class \(C\) such that \(\operatorname{Sat}\left(\Phi_{C}\right)=C\). To indicate how to obtain such master formulae, consider the bisimulation quotient of the Bakery algorithm (see Figure 7.9 on page 464), and let \(A P=\left\{c r i t_{1}, c r i t_{2}\right\}\). Instead of considering the conjunction of formulae \(\Phi_{C, D}\), master formulae can be obtained by inspecting the transition system. The equivalence class \(C=\left\langle\right.\) wait \(_{1}\), wait \(\left.{ }_{2}, x_{1}>x_{2}>0\right\rangle\) is, given the atomic propositions in \(A P\), uniquely characterized by the fact that none of the processes is currently in the critical section and the second process acquires access to the critical section immediately. A master formula \(\Phi_{C}\) for \(C\) thus is
\[
\Phi_{C}=\neg \text { crit }_{1} \wedge \neg \text { crit }_{2} \wedge \forall \bigcirc \text { crit }_{2}
\]

A master formula for the equivalence class \(D=\left\langle\right.\) wait \(_{1}\), crit \(\left._{2},--\right\rangle\) is
\[
\Phi_{D}=c r i t_{2} \wedge \forall \bigcirc \neg c r i t_{2} .
\]

It is easy to check that none of the other equivalence classes under bisimulation satisfies \(\Phi_{D}\).

\section*{Remark 7.23. Logical Characterization by a Sublogic of CTL}

We stress that the proof of Lemma 7.21 only exploits the propositional fragment of CTL, i.e., atomic propositions, conjunction, and negation, and the modal operator \(\exists \bigcirc\). It does not, however, rely on the presence of the until operator. Thus, bisimulation equivalence (and CTL* equivalence) agrees with the equivalence induced by a simple logic that only contains atomic propositions, negation, conjunction and \(\exists \bigcirc\). In particular, any two
nonbisimilar states can be distinguished by a CTL formula that does contain neither the until operator U nor one of the derived operators \(\diamond, \square, \mathrm{W}\) or R .

\section*{Remark 7.24. Infinite Transition Systems}

Lemma 7.21 is restricted to finite transition systems. This restriction is reflected in the proof as follows. The master formula for equivalence class \(C\), formula \(\Phi_{C}\) is defined as the conjunction of CTL formulae for the form \(\Phi_{C, D}\) with \(C \neq D\). As CTL only allows for finite conjunctions, there should only be finitely many equivalence classes \(D \neq C\). This is guaranteed, of course, if the transition system is finite. An extension of CTL with infinite conjunctions would be needed to obtain a similar result for infinite transition systems.

An example of an infinite transition system \(T S\) where CTL equivalence is not finer than bisimulation equivalence is as follows. Assume the set \(A P\) of atomic propositions is infinite. The states in \(T S\) are \(s_{1}, s_{2}\) and the states \(t_{A}\) where \(A\) is a subset of \(A P\). States \(s_{1}, s_{2}\) are labelled with \(\varnothing\), while the label of state \(t_{A}\) is \(A\). State \(s_{1}\) has a transition to all states \(t_{A}\) where \(A \neq A P\), while state \(s_{2}\) has transitions to all states \(t_{A}\), including \(A=A P\). Thus, \(\operatorname{Post}\left(s_{1}\right)=\left\{t_{A} \mid A \subseteq A P, A \neq A P\right\}\) and \(\operatorname{Post}\left(s_{2}\right)=\left\{t_{A} \mid A \subseteq A P\right\}\). States \(t_{A}\) have a single outgoing transition which leads to \(s_{1}\). Clearly, \(s_{1}\) and \(s_{2}\) are not bisimulationequivalent as \(s_{2}\) has a transition to state \(t_{A P}\) but there is no successor of \(s_{1}\) with the label \(L\left(t_{A P}\right)=A P\). On the other hand, \(s_{1}\) and \(s_{2}\) are CTL equivalent with respect to \(A P\). This follows from the fact that (1) the set of atomic propositions that appear in a given CTL formula over \(A P\) is finite and (2) \(s_{1}, s_{2}\) are bisimulation-equivalent when shrinking the set \(A P^{\prime}\) of atomic propositions to any finite subset of \(A P\). Note that then \(\mathcal{R}=\left\{\left(s_{1}, s_{2}\right)\right\} \cup\left\{\left(t_{A}, t_{B}\right): A \cap A P^{\prime}=B \cap A P^{\prime}\right\}\) is a bisimulation. But then Lemma 7.26 yields that \(s_{1}, s_{2}\) satisfy the same CTL formulae over \(A P^{\prime}\).

However, the result stated in Lemma 7.21 can also be established for (possibly infinite) transition systems that are finitely branching. This is shown in the following lemma.

Recall that a transition system is finitely branching if (i) the set of initial states is finite and (ii) for each state \(s\) the set of successors (i.e., the set \(\operatorname{Post}(s)\) ) is finite.

\section*{Lemma 7.25. CTL Equivalence is Finer Than Bisimulation (Revisited)}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a finitely branching transition system without terminal states, and \(s_{1}, s_{2}\) states in TS. Then \(s_{1} \equiv C T L s_{2}\) implies \(s_{1} \sim_{T S} s_{2}\).

Proof: As in the proof of Lemma 7.25 we show that
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid s_{1} \equiv_{C T L} s_{2}\right\}
\]
is a bisimulation for \(T S\). Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\). These states are equally labeled, since for any atomic proposition \(a\) we have \(s_{1} \models a\) iff \(s_{2} \models a\), i.e., \(a \in L\left(s_{1}\right)\) if and only if \(a \in L\left(s_{2}\right)\). Condition (2) is proven by contraposition. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\). Assume there is no \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\). Since \(T S\) is finitely branching, the set \(\operatorname{Post}\left(s_{2}\right)\) is finite, say it agrees with \(\left\{t_{1}, \ldots, t_{k}\right\}\). The assumption \(\left(s_{1}^{\prime}, t_{j}\right) \notin \mathcal{R}\) and the definition of \(\mathcal{R}\) yields the existence of a CTL formula \(\Psi_{j}\) such that
\[
s_{1}^{\prime} \models \Psi_{j} \quad \text { and } \quad t_{j} \not \models \Psi_{j} \quad \text { for } 0<j \leqslant k
\]

Now consider the formula
\[
\Phi=\exists \bigcirc\left(\Psi_{1} \wedge \ldots \wedge \Psi_{k}\right)
\]

Clearly, we have \(s_{1}^{\prime} \models \Psi_{1} \wedge \ldots \wedge \Psi_{k}\). Hence, \(s_{1} \models \Phi\). On the other hand, \(t_{j} \not \vDash \Psi_{1} \wedge \ldots \wedge \Psi_{k}\) for \(0<j \leqslant k\). This yields \(s_{2} \not \vDash \Phi\). But then \(\left(s_{1}, s_{2}\right) \notin \mathcal{R}\). Contradiction.

The following lemma asserts that bisimilar states are CTL* equivalent, and is not restricted to finite transition systems. Let \(\pi_{1} \sim_{T S} \pi_{2}\) denote that the path fragments \(\pi_{1}\) and \(\pi_{2}\) are statewise bisimilar, i.e., for \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots\) and \(\pi_{2}=s_{0,2} s_{1,2} s_{2,2} \ldots\), we have \(s_{j, 1} \sim_{T S} s_{j, 2}\) for all \(j \geqslant 0\).

\section*{Lemma 7.26. Bisimulation is Finer Than CTL* Equivalence}

Let TS be a transition system TS (over AP) without terminal states, \(s_{1}, s_{2}\) be states in \(T S\), and \(\pi_{1}, \pi_{2}\) infinite path fragments in TS. Then:
(a) If \(s_{1} \sim_{T S} s_{2}\), then for any CTL* formula \(\Phi: s_{1} \models \Phi\) if and only if \(s_{2} \models \Phi\).
(b) If \(\pi_{1} \sim_{\text {TS }} \pi_{2}\), then for any CTL* path formula \(\varphi: \pi_{1} \models \varphi\) if and only if \(\pi_{2} \models \varphi\).

Proof: We simultaneously prove (a) and (b) by means of induction over the structure of the formula.

Induction basis. Let \(s_{1} \sim_{T S} s_{2}\). For \(\Phi=\) true, statement (a) obviously holds. Since \(L\left(s_{1}\right)=L\left(s_{2}\right)\), it follows that for \(\Phi=a \in A P\) :
\[
s_{1} \models a \text { iff } a \in L\left(s_{1}\right) \text { iff } a \in L\left(s_{2}\right) \text { iff } s_{2} \models a .
\]

\section*{Induction step}
(a) Assume \(\Phi_{1}, \Phi_{2}, \Psi\) are CTL* \(^{*}\) state formulae for which proposition (a) holds, and \(\varphi\) to be a CTL* path formula for which proposition (b) holds. Let \(s_{1} \sim_{T S} s_{2}\). The proof is by structural induction on \(\Phi\) :

Case 1: \(\Phi=\Phi_{1} \wedge \Phi_{2}\). The induction hypothesis for \(\Phi_{1}\) and \(\Phi_{2}\) provides
\[
\begin{array}{lll}
s_{1} \models \Phi_{1} \wedge \Phi_{2} & \text { iff } & s_{1} \models \Phi_{1} \text { and } s_{1} \models \Phi_{2} \\
& \text { iff } & s_{2} \models \Phi_{1} \text { and } s_{2} \models \Phi_{2} \\
& \text { iff } & s_{2} \models \Phi_{1} \wedge \Phi_{2} .
\end{array}
\]

Case 2: \(\Phi=\neg \Psi\). By applying the induction hypothesis for \(\Psi\), one obtains
\[
\begin{array}{lll}
s_{1} \models \neg \Psi & \text { iff } & s_{1} \not \vDash \Psi \\
& \text { iff } & s_{2} \not \models \Psi \\
& \text { iff } & s_{2} \neq \neg \Psi .
\end{array}
\]

Case 3: \(\Phi=\exists \varphi\). For symmetry reasons, it suffices to show
\[
s_{1} \models \exists \varphi \quad \Longrightarrow \quad s_{2} \models \exists \varphi .
\]

Assume \(s_{1} \models \exists \varphi\). Then there exists \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots \in \operatorname{Paths}\left(s_{1}\right)\) starting in \(s_{1}=s_{0,1}\), with \(\pi_{1} \models \varphi\). According to the path-lifting lemma (Lemma 7.5 on page 454) there exists a path \(\pi_{2}=s_{0,2} s_{1,2} s_{2,2} \ldots \in \operatorname{Paths}\left(s_{2}\right)\) starting in \(s_{2}=s_{0,2}\), such that \(\pi_{1} \sim_{\text {TS }} \pi_{2}\). From the induction hypothesis (on part (b)), it follows that \(\pi_{2} \models \varphi\). Thus, \(s_{2} \models \exists \varphi\).
(b) Assume statement (a) holds for CTL* state formula \(\Phi\), and (b) holds for CTL* path formulae \(\varphi_{1}, \varphi_{2}\), and \(\psi\). Let \(\pi_{1} \sim_{T S} \pi_{2}\). Recall that \(\pi_{i}[j \ldots]\) denotes the suffix \(s_{j, i} s_{j+1, i} s_{j+2, i} \ldots\) of \(\pi_{i}\). The proof proceeds by structural induction on \(\varphi\) :

Case 1: \(\varphi=\Phi\). It follows from the CTL* semantics and the induction hypothesis on \(\Phi\) that
\[
\pi_{1} \models \varphi \text { iff } s_{0,1} \models \Phi \text { iff } s_{0,2} \models \Phi \text { iff } \pi_{2} \models \varphi .
\]

Case 2: \(\varphi=\varphi_{1} \wedge \varphi_{2}\). By applying the induction hypothesis to \(\varphi_{1}\) and \(\varphi_{2}\) we obtain
\[
\begin{array}{lll}
\pi_{1} \models \varphi_{1} \wedge \varphi_{2} & \text { iff } & \pi_{1} \models \varphi_{1} \text { and } \pi_{1} \models \varphi_{2} \\
& \text { iff } & \pi_{2} \models \varphi_{1} \text { and } \pi_{2} \models \varphi_{2} \\
& \text { iff } & \pi_{2} \models \varphi_{1} \wedge \varphi_{2} .
\end{array}
\]

Case 3: \(\varphi=\neg \psi\). We apply the induction hypothesis to \(\psi\) and obtain
\[
\begin{array}{lll}
\pi_{1} \models \neg \psi & \text { iff } & \pi_{1} \not \vDash \psi \\
& \text { iff } & \pi_{2} \not \vDash \psi \\
& \text { iff } & \pi_{2} \not \models \neg \psi .
\end{array}
\]

Case 4: \(\varphi=\bigcirc \psi\). The induction hypothesis for \(\psi\) and the path fragments \(\pi_{l}[1 \ldots]\), \(l=1,2\), provides
\[
\begin{array}{lll}
\pi_{1} \models \bigcirc \psi & \text { iff } & \pi_{1}[1 \ldots] \models \psi \\
& \text { iff } & \pi_{2}[1 \ldots] \models \psi \\
& \text { iff } & \pi_{2} \models \bigcirc \psi .
\end{array}
\]

Case 5: \(\varphi=\varphi_{1} \cup \varphi_{2}\). The induction hypothesis for \(\varphi_{1}\) and \(\varphi_{2}\) and the path fragments \(\pi_{l}[i \ldots], l=1,2, i=0,1,2, \ldots\), provides
\[
\pi_{1} \models \varphi_{1} \cup \varphi_{2} \quad \text { iff } \quad \text { there exists an index } j \in \mathbb{N} \text { with }
\]
\[
\pi_{1}[j \ldots] \models \varphi_{2} \text { and }
\]
\[
\pi_{1}[i \ldots] \vDash \varphi_{1}, i=0,1, \ldots, j-1
\]
iff there exists an index \(j \in \mathbb{N}\) with
\(\pi_{2}[j \ldots] \models \varphi_{2}\) and
\(\pi_{2}[i \ldots] \vDash \varphi_{1}, i=0,1, \ldots, j-1\)
iff \(\quad \pi_{2} \models \varphi_{1} \cup \varphi_{2}\).

\section*{Corollary 7.27. CTL*/CTL-vs. Bisimulation Equivalence}

For finite transition systems \(T S_{1}, T S_{2}\) (over AP) without terminal states, the following three statements are equivalent:
(a) \(T S_{1} \sim T S_{2}\)
(b) \(T S_{1} \equiv_{C T L} T S_{2}\), i.e., \(T S_{1}\) and \(T S_{2}\) satisfy the same CTL formulae.
(c) \(T S_{1} \equiv_{C T L} T S_{2}\), i.e., \(T S_{1}\) and \(T S_{2}\) satisfy the same CTL* formulae.

Thus, bisimilar transition systems are equivalent with respect to all formulae that can be expressed in CTL*. The fact that CTL equivalence is finer than bisimulation equivalence proves that bisimulation equivalence is the coarsest equivalence which preserves all CTL formulae. Stated differently, a relation which is strictly coarser than \(\sim\) may yield a smaller quotient transition system, but cannot guarantee the preservation of all CTL formulae. The fact that CTL equivalence is finer than \(\sim\) is useful to prove that two finite transition systems are not bisimilar - it suffices to indicate a single CTL formula which holds for one transition system, but not for the other.

Example 7.28. Distinguishing Nonbisimilar Systems by CTL Formulae
Consider the beverage vending machines \(T S_{1}\) and \(T S_{2}\) in Figure 7.3 on page 453. For \(A P=\{\) pay, beer, soda \(\}\), we have \(T S_{1} \not \chi_{T S} T S_{2}\). In fact, they can be distinguished by the CTL formula
\[
\Phi=\exists \bigcirc(\exists \bigcirc \text { beer } \wedge \exists \bigcirc \text { soda })
\]
as \(T S_{1} \models \Phi\), but \(T S_{2} \not \vDash \Phi\).

\subsection*{7.3 Bisimulation-Quotienting Algorithms}

This section presents algorithms for obtaining the bisimulation quotient for a finite transition system TS. Such an algorithm serves two purposes. First, it can be used to verify the bisimilarity of two finite transition systems \(T S_{1}\) and \(T S_{2}\) by considering the quotient of \(T S_{1} \oplus T S_{2}\) (see page 457). Since bisimulation is finer than trace equivalence, algorithms that verify whether \(T S_{1} \sim T S_{2}\) can also be used as a sound, though incomplete, proof technique for proving the trace equivalence of \(T S_{1}\) and \(T S_{2}\). Secondly, such algorithms can be used to obtain the abstract (and thus smaller) transition system \(T S / \sim\) in a fully automated manner. As \(T S \sim T S / \sim\) (see Theorem 7.14 on page 464), it follows from Corollary 7.27 on page 475 that any verification result for \(T S / \sim\) - either being negative or positive - carries over to TS. This applies to any formula expressed in either LTL, CTL, or CTL*.

This section treats two algorithms for computing the bisimulation quotient \(T S / \sim\). Both algorithms rely on a partition refinement technique. Roughly speaking, the finite state space \(S\) is partitioned in blocks, i.e., pairwise disjoint sets of states. Starting from a straightforward initial partition where, e.g., all equally labeled states form a partition, the algorithm successively refines these partitions such that ultimately partitions only contain bisimilar states. This strategy strongly resembles minimization algorithms for deterministic finite automata (DFA).

In the sequel, let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a finite transition system with \(S \neq \varnothing\).

\section*{Definition 7.29. Partition, Block, Superblock}

A partition for \(S\) is a set \(\Pi=\left\{B_{1}, \ldots, B_{k}\right\}\) such that \(B_{i} \neq \varnothing(\) for \(0<i \leqslant k), B_{i} \cap B_{j}=\varnothing\) (for \(0<i, j \leqslant k\) and \(i \neq j\) ), and \(S=\bigcup_{0<i \leqslant k} B_{i}\).
\(B_{i} \in \Pi\) is called a block. \(C \subseteq S\) is a superblock of \(\Pi\) if \(C=B_{i_{1}} \cup \ldots \cup B_{i_{\ell}}\) for some \(B_{i_{1}}, \ldots, B_{i_{\ell}} \in \Pi\).

Let \([s]_{\Pi}\) denote the unique block of partition \(\Pi\) containing \(s\). For partitions \(\Pi_{1}\) and \(\Pi_{2}\) of \(S, \Pi_{1}\) is called finer than \(\Pi_{2}\), or \(\Pi_{2}\) is called coarser than \(\Pi_{1}\), if:
\[
\forall B_{1} \in \Pi_{1} \exists B_{2} \in \Pi_{2} . B_{1} \subseteq B_{2}
\]

In this case, every block of \(\Pi_{2}\) can be written as a disjoint union of blocks in \(\Pi_{1}\). \(\Pi_{1}\) is strictly finer than \(\Pi_{2}\) (and \(\Pi_{2}\) is strictly coarser than \(\Pi_{1}\) ) if \(\Pi_{1}\) is finer than \(\Pi_{2}\) and \(\Pi_{1} \neq \Pi_{2}\).

\section*{Remark 7.30. Partitions and Equivalences}

There is a close connection between equivalence relations and partitions. For equivalence relation \(\mathcal{R}\) on \(S\), the quotient space \(S / \mathcal{R}\) is a partition for \(S\). Vice versa, partition \(\Pi\) for \(S\) induces the equivalence relation:
\[
\begin{aligned}
\mathcal{R}_{\Pi} & =\left\{\left(s_{1}, s_{2}\right) \mid \exists B \in \Pi . s_{1} \in B \wedge s_{2} \in B\right\} \\
& =\left\{\left(s_{1}, s_{2}\right) \mid\left[s_{1}\right]_{\Pi}=\left[s_{2}\right]_{\Pi}\right\}
\end{aligned}
\]
such that \(S / \mathcal{R}_{\Pi}\) corresponds to \(\Pi\). For equivalence relation \(\mathcal{R}\), the equivalence relation \(\mathcal{R}_{\Pi}\) induced by \(\Pi=S / \mathcal{R}\) corresponds to \(\mathcal{R}\).

The partition refinement algorithm works with a representation of equivalences by means of the induced partition (i.e., the quotient space). In the initial partition \(\Pi_{0}=\Pi_{A P}\), each group of equally labeled states forms a block. Subsequently, the current partition \(\Pi_{i}\) is successively replaced with a finer partition \(\Pi_{i+1}\), (for details, see Section 7.3.2). This iterative refinement procedure is halted as soon as \(\Pi_{i}\) cannot be further refined, i.e., when \(\Pi_{i}=\Pi_{i+1}\). Such situation is guaranteed to occur as \(S\) is finite. The resulting partition \(\Pi_{i}\) is the bisimulation quotient space. The main steps are outlined in Algorithm 28.
```

Algorithm 28 Partition refinement algorithm (basic idea)
Input: finite transition system $T S$ with state space $S$
Output: bisimulation quotient space $S / \sim$

```
\(\Pi_{0}:=\Pi_{A P} ;\)
\(i:=0 ;\)
repeat \(\quad\left(*\right.\) see Section \(\left.7.3 .1{ }^{*}\right)\)
\(\quad \Pi_{i+1}:=\operatorname{Refine}\left(\Pi_{i}\right) ;\)
\(\quad i:=i+1 ;\)
until \(\Pi_{i}=\Pi_{i-1} \quad\left(*\right.\) no further refinement possible \(\left.{ }^{*}\right)\)
return \(\Pi_{i}\)

The termination of the partition refinement procedure is clear, as the partition \(\Pi_{i+1}\) is strictly finer than \(\Pi_{i}\). Thus, for the induced equivalence relations \(\mathcal{R}_{\Pi_{i}}\) we have
\[
S \times S \supseteq \mathcal{R}_{\Pi_{0}} \supsetneqq \mathcal{R}_{\Pi_{1}} \supsetneqq \mathcal{R}_{\Pi_{2}} \supsetneqq \ldots \supsetneqq \mathcal{R}_{\Pi_{i}}=\sim_{T S} .
\]

Due to the fact that \(S\) is finite, a partition \(\Pi_{i}\) with \(\Pi_{i}=\Pi_{i-1}\) is reached after at most \(|S|\) iterations. After \(|S|\) proper refinements, any block in \(\Pi_{i}\) is a singleton, and a further refinement is thus impossible.

\subsection*{7.3.1 Determining the Initial Partition}

Since bisimilar states are equally labeled, it is rather natural to use this in determining the initial partition \(\Pi_{0}=\Pi_{A P}\).

\section*{Definition 7.31. The AP Partition}

The \(A P\) partition of \(T S\), denoted \(\Pi_{A P}\), is the quotient space \(S / \mathcal{R}_{A P}\) induced by \(\mathcal{R}_{A P}=\) \(\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid L\left(s_{1}\right)=L\left(s_{2}\right)\right\}\).

The initial partition \(\Pi_{0}=\Pi_{A P}\) can be computed as follows. The basic idea is to generate a decision tree for \(a \in A P\). For \(A P=\left\{a_{1}, \ldots, a_{k}\right\}\), the height of the decision tree for \(A P\) is \(k\). The vertices at depth \(i<k\) represent the decision "is \(a_{i+1} \in L(s)\) ?". The left branch of vertex \(v\) at depth \(i<k\) represents the case " \(a_{i+1} \notin L(s)\) ", while the right branch represents " \(a_{i+1} \in L(s)\) ". The full decision tree for, e.g., \(k=2, a_{1}=a\), and \(a_{2}=b\) is fo the form:


Leaf \(v\) represents a set of states. More precisely, states \((v)\) contains the states \(s \in S\) for which \(L(s)\) is represented by the path from the root to \(v\). The decision tree for \(A P\) is constructed successively by considering all states in \(S\) separately. The initial decision tree consists only of the root \(v_{0}\). On considering state \(s\), the tree is traversed in a top-down manner, and new vertices are inserted when necessary, i.e., when \(s\) is the first encountered state with labeling \(L(s)\). Once the tree traversal for state \(s\) reaches leaf \(w\), \(\operatorname{states}(w)\) is extended with \(s\). The essential steps are outlined in Algorithm 29.

\section*{Example 7.32. Determining the Initial Partition}

Consider the transition system \(T S\) over \(A P=\{a, b\}\) in Figure 7.10. We assume that \(L\left(s_{0}\right)=L\left(s_{2}\right)=L\left(s_{5}\right)=L\left(s_{6}\right)=\{a\}, L\left(s_{1}\right)=\{a, b\}\), and \(L\left(s_{3}\right)=L\left(s_{4}\right)=\varnothing\). The first three steps for generating the initial partition and the resulting decision tree are indicated in Figure 7.11. The resulting partition \(\Pi_{A P}\) consists of three blocks. Block \(B_{1}=\left\{s_{0}, s_{2}, s_{5}, s_{6}\right\}\) represents all states labeled with \(\{a\}\). The two other blocks stand for the states labeled with \(\{a, b\}\) or \(\varnothing\).
```

Algorithm 29 Computing the initial partition $\Pi_{A P}$
Input: $T S=(S, A c t, \rightarrow, I, A P, L)$ over $A P=\left\{a_{1}, \ldots, a_{k}\right\}$ and finite $S$
Output: partition $\Pi_{A P}$

```



Figure 7.10: Transition system \(T S\) over \(A P=\{a, b\}\).
We conclude this section by considering the time complexity of computing the initial partition. For each state \(s\) in \(S\), the decision tree has to be traversed from root to leaf. This takes \(\Theta(|A P|)\) time. The overall time complexity is now given as follows.

\section*{Lemma 7.33. Time Complexity of Computing the Initial Partition}

The initial partition \(\Pi_{A P}\) can be computed in time \(\Theta(|S| \cdot|A P|)\).

\subsection*{7.3.2 Refining Partitions}

Since any partition \(\Pi_{i}(i \geqslant 0)\) is finer than the initial partition \(\Pi_{A P}\), each block in \(\Pi_{i}\) is guaranteed to contain only equally labeled states. The initial partition, however, does not consider the one-step successors of states. This is taken care of in the successive refinement steps.

\section*{Lemma 7.34. Coarsest Partition}

The bisimulation quotient space \(S / \sim\) is the coarsest partition \(\Pi\) for \(S\) such that
(i) \(\Pi\) is finer than \(\Pi_{A P}\).


Figure 7.11: Generating the Partition \(\Pi_{A P}\).
(ii) for all \(B, C \in \Pi: B \cap \operatorname{Pre}(C)=\varnothing\) or \(B \subseteq \operatorname{Pre}(C)\).

Moreover, if partition \(\Pi\) fulfills (ii), then \(B \cap \operatorname{Pre}(C)=\varnothing\) or \(B \subseteq \operatorname{Pre}(C)\) for all \(B \in \Pi\) and all superblocks \(C\) of \(\Pi\).

Recall that \(\operatorname{Pre}(C)=\{s \in S \mid \operatorname{Post}(s) \cap C \neq \varnothing\}\) denotes the set of states in \(S\), which have at least one successor in \(C\). Thus, for every block \(B\), we have that \(B \cap \operatorname{Pre}(C)=\varnothing\) if and only if no state of \(B\) has an immediate successor in \(C\), and \(B \subseteq \operatorname{Pre}(C)\) if and only if all states in \(B\) have an immediate successor in \(C\).

Proof: Let \(\Pi\) be a partition of \(S\) and \(\mathcal{R}_{\Pi}\) the equivalence relation on \(S\) induced by \(\Pi\). The proof is carried out in two steps. We first show that \(\mathcal{R}_{\Pi}\) is a bisimulation if and only if the conditions (i) and (ii) are satisfied. Then, we show that \(S / \sim\) is the coarsest partition satisfying (i) and (ii).
\(\Leftarrow\) : Assume that \(\Pi\) satisfies (i) and (ii). We prove that \(\mathcal{R}_{\Pi}\) induced by \(\Pi\) is a bisimulation. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}_{\Pi}\) and \(B=\left[s_{1}\right]_{\Pi}=\left[s_{2}\right]_{\Pi}\).
1. Since \(\Pi\) is finer than \(\Pi_{A P}\) (condition (i)), there exists a block \(B^{\prime}\) of \(\Pi_{A P}\) containing \(B\). Thus, \(s_{1}, s_{2} \in B \subseteq B^{\prime} \in \Pi_{A P}\), and, therefore, \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
2. Let \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) and \(C=\left[s_{1}^{\prime}\right]_{\Pi}\). Then, \(s_{1} \in B \cap \operatorname{Pre}(C)\). By condition (ii), we obtain \(B \subseteq \operatorname{Pre}(C)\). Hence, \(s_{2} \in \operatorname{Pre}(C)\). So, there exists a state \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right) \cap C\). Since \(s_{2}^{\prime} \in C=\left[s_{1}^{\prime}\right]_{\Pi}\), it follows that \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}_{\Pi}\).
\(\Rightarrow\) : Assume \(\mathcal{R}_{\Pi}\) is a bisimulation. The proof obligation is to show that the conditions (i) and (ii) are satisfied.
(i) By contraposition. Assume that \(\Pi\) is not finer than \(\Pi_{A P}\). Then, there exist a block \(B \in \Pi\) and states \(s_{1}, s_{2} \in B\) with \(\left[s_{1}\right]_{\Pi_{A P}} \neq\left[s_{2}\right]_{\Pi_{A P}}\). Then, \(L\left(s_{1}\right) \neq L\left(s_{2}\right)\). Hence, \(\mathcal{R}_{\Pi}\) is not a bisimulation. Contradiction.
(ii) Let \(B, C\) be blocks of \(\Pi\). We assume that \(B \cap \operatorname{Pre}(C) \neq \varnothing\) and show that \(B \subseteq \operatorname{Pre}(C)\). As \(B \cap \operatorname{Pre}(C) \neq \varnothing\), there exists a state \(s_{1} \in B\) with \(\operatorname{Post}\left(s_{1}\right) \cap C \neq \varnothing\). Let \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right) \cap C\) and \(s_{2}\) be an arbitrary state of \(B\). We demonstrate that \(s_{2} \in \operatorname{Pre}(C)\). Since \(s_{1}, s_{2} \in B\), we get that \(\left(s_{1}, s_{2}\right) \in \mathcal{R}_{\Pi}\). Due to \(s_{1} \rightarrow s_{1}^{\prime}\), there exists a transition \(s_{2} \rightarrow s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}_{\Pi}\). But then \(s_{1}^{\prime} \in C\) yields \(s_{2}^{\prime} \in C\). Hence, \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right) \cap C\). Thus, \(s_{2} \in \operatorname{Pre}(C)\).

Assume \(\Pi\) satisfies (ii). We show that \(B \cap \operatorname{Pre}(C)=\varnothing\) or \(B \subseteq \operatorname{Pre}(C)\) for any \(B \in \Pi\) and superblocks \(C\) of \(\Pi\). The proof is by contraposition. Let \(B \in \Pi\) and \(C\) a superblock, i.e., \(C\) is of the form \(C=C_{1} \cup \ldots \cup C_{\ell}\) for blocks \(C_{1}, \ldots, C_{\ell}\) of \(\Pi\). Assume that \(B \cap \operatorname{Pre}(C) \neq \varnothing\) and \(B \nsubseteq \operatorname{Pre}(C)\). Then, there exists an index \(i \in\{1, \ldots, \ell\}\) with \(B \cap \operatorname{Pre}\left(C_{i}\right) \neq \varnothing\). Further, it is clear that \(B \nsubseteq \operatorname{Pre}\left(C_{i}\right)\), since otherwise \(B \subseteq \operatorname{Pre}\left(C_{i}\right) \subseteq \operatorname{Pre}(C)\). Thus, condition (ii) is not satisfied for block \(C_{i} \in \Pi\). Contradiction.

It remains to show that the bisimulation partition \(\Pi=S / \sim\) is the coarsest partition of \(S\) satisfying the conditions (i) and (ii). This immediately follows from the fact that \(\sim_{\text {TS }}\) is the coarsest bisimulation (Lemma 7.8 on page 457).

Let us now consider how partitions are successively refined. Every refinement aims at satisfying condition (ii). To that end, a superblock \(C\) of \(\Pi\) is considered and every block \(B\) of the current partition \(\Pi_{i}\) is decomposed ("splitted") into the subblocks \(B \cap \operatorname{Pre}(C)\) and \(B \backslash \operatorname{Pre}(C)\), provided that these subblocks are nonempty. If one of the subblocks is empty, then \(B\) satisfies condition (ii) and no decomposition takes place.

\section*{Definition 7.35. The Refinement Operator}

Let \(\Pi\) be a partition for \(S\) and \(C\) be a superblock of \(\Pi\). Then:
\[
\operatorname{Refine}(\Pi, C)=\bigcup_{B \in \Pi} \operatorname{Refine}(B, C)
\]
where \(\operatorname{Refine}(B, C)=\{B \cap \operatorname{Pre}(C), B \backslash \operatorname{Pre}(C)\} \backslash\{\varnothing\}\).


Figure 7.12: Refinement operator.
The basic scheme of the refinement operator is depicted in Figure 7.12. Refine \((B, C)=\) \(\{B\}\), i.e., \(B\) is not decomposed with respect to \(C\), if all states in \(B\) have a direct \(C\) successor or if no state in \(B\) has a direct \(C\)-successor. In case some states in \(B\) have a direct \(C\)-successor, while others have not, \(B\) is refined accordingly.

The following result shows that successive refinements, starting with partition \(\Pi_{A P}\), yield a series of partitions \(\Pi_{0}=\Pi_{A P}, \Pi_{1}, \Pi_{2}, \Pi_{3}, \ldots\), which become increasingly finer and which all are coarser than \(S / \sim\). This property is essential for establishing the correctness of the iterative approach for as outlined in Algorithm 28 (see page 477).

\section*{Lemma 7.36. Correctness of the Refinement Operator}

Let \(\Pi\) be a partition of \(S\), which is finer than \(\Pi_{A P}\) and coarser than \(S / \sim\). Further, let \(C\) be a superblock for \(\Pi\). Then:
(a) Refine \((\Pi, C)\) is finer than \(\Pi\).
(b) Refine \((\Pi, C)\) is coarser than \(S / \sim\).

\section*{Proof:}
(a) This follows directly from the definition of Refine, since every block \(B \in \Pi\) is either contained in Refine \((\Pi, C)\) or is decomposed into \(B \cap \operatorname{Pre}(C)\) and \(B \backslash \operatorname{Pre}(C)\).
(b) Let \(B \in S / \sim\). We show that \(B\) is contained in a block of \(\operatorname{Refine}(\Pi, C)\). Since \(\Pi\) is coarser than \(S / \sim\), there exists a block \(B^{\prime} \in \Pi\) with \(B \subseteq B^{\prime} . B^{\prime}\) is of the form \(B^{\prime}=B \cup D\) where \(D\) is a (possibly empty) superblock of \(S / \sim\). If \(B^{\prime} \in \operatorname{Refine}(\Pi, C)\), then \(B \subseteq B^{\prime} \in \operatorname{Refine}(\Pi, C)\). Otherwise, i.e., if \(B^{\prime} \notin \operatorname{Refine}(\Pi, C)\), then \(B^{\prime}\) is decomposed into the subblocks \(B^{\prime} \cap \operatorname{Pre}(C)\) and \(B^{\prime} \backslash \operatorname{Pre}(C)\). We now show that \(B\) is contained in one of these two new subblocks. Condition (ii) of Lemma 7.34 implies that either \(B \cap \operatorname{Pre}(C)=\varnothing\) (thus, \(B \backslash \operatorname{Pre}(C)=B\) ) or \(B \backslash \operatorname{Pre}(C)=\varnothing\) (thus, \(B \cap \operatorname{Pre}(C)=B)\). Since \(B^{\prime}=B \cup D, B\) is either contained in block
- \(B^{\prime} \backslash \operatorname{Pre}(C)=(B \backslash \operatorname{Pre}(C)) \cup(D \backslash \operatorname{Pre}(C))\)
- or in \(B^{\prime} \cap \operatorname{Pre}(C)=(B \cap \operatorname{Pre}(C)) \cup(D \cap \operatorname{Pre}(C))\).

\section*{Definition 7.37. Splitter, Stability}

Let \(\Pi\) be a partition for \(S\) and \(C\) a superblock of \(\Pi\).
1. \(C\) is a splitter for \(\Pi\) if there exists a block \(B \in \Pi\) with \(B \cap \operatorname{Pre}(C) \neq \varnothing\) and \(B \backslash \operatorname{Pre}(C) \neq \varnothing\).
2. Block \(B\) is stable with respect to \(C\) if \(B \cap \operatorname{Pre}(C)=\varnothing\) or \(B \backslash \operatorname{Pre}(C)=\varnothing\).
3. \(\Pi\) is stable with respect to \(C\) if each block \(B \in \Pi\) is stable wrt. \(C\).
\(C\) is thus a splitter for \(\Pi\) if and only if \(\Pi \neq \operatorname{Refine}(\Pi, C)\), that is, if and only if \(\Pi\) is not stable for \(C . B\) is stable with respect to \(C\) whenever \(\{B\}=\operatorname{Refine}(B, C)\). Note that \(S / \sim\) is the coarsest stable partition that is finer than \(\Pi_{A P}\).
```

Algorithm 30 Algorithm for computing the bisimulation quotient space
Input: finite transition system $T S$ over $A P$ with state space $S$
Output: bisimulation quotient space $S / \sim$
$\Pi:=\Pi_{A P} ;$
while there exists a splitter for $\Pi$ do
choose a splitter $C$ for $\Pi$;
$\Pi:=\operatorname{Refine}(\Pi, C) ; \quad \quad\left(* \operatorname{Refine}(\Pi, C)\right.$ is strictly finer than $\left.\Pi{ }^{*}\right)$
od
return $\Pi$

```

The refine operator and the concept of a splitter can be effectively exploited in computing \(S / \sim\), see Algorithm 30. Its correctness follows from Lemmas 7.34 and 7.36, Termination follows from the following result.

\section*{Lemma 7.38. Termination Criterion of Algorithm 30}

Let \(\Pi\) be a partition for \(S\), which is finer than \(\Pi_{A P}\) and coarser than \(S / \sim\). Then, \(\Pi\) is strictly coarser than \(S / \sim\) if and only if there exists a splitter for \(\Pi\).

Proof: Follows immediately from Lemma 7.34 on page 480.

\section*{Example 7.39. Partition Refinement Algorithm}

Figure 7.13 illustrates the principle of the partition refinement algorithm on a small example. The initial partition \(\Pi_{A P}\) identifies all equally labeled states:
\[
\Pi_{0}=\Pi_{A P}=\left\{\left\{s_{1}, s_{2}, s_{3}\right\},\left\{t_{1}, t_{2}, t_{3}\right\},\left\{u_{1}, u_{2}\right\},\left\{v_{1}, v_{2}\right\}\right\}
\]

In the first step, consider \(C_{1}=\left\{v_{1}, v_{2}\right\}\). This leaves the blocks \(\left\{s_{1}, s_{2}, s_{3}\right\},\left\{u_{1}, u_{2}\right\}\) and \(\left\{v_{1}, v_{2}\right\}\) unaffected, but splits block of the \(t\)-states into \(\left\{t_{1}\right\}=\left\{t_{1}, t_{2}, t_{3}\right\} \backslash \operatorname{Pre}\left(C_{1}\right)\) and
\(\left\{t_{2}, t_{3}\right\}=\left\{t_{1}, t_{2}, t_{3}\right\} \cap \operatorname{Pre}\left(C_{1}\right)\). Thus, we obtain the partition
\[
\Pi_{1}=\left\{\left\{s_{1}, s_{2}, s_{3}\right\},\left\{t_{1}\right\},\left\{t_{2}, t_{3}\right\},\left\{u_{1}, u_{2}\right\},\left\{v_{1}, v_{2}\right\}\right\} .
\]

In the second refinement step, consider \(C_{2}=\left\{t_{1}\right\}\). This splitter divides the \(s\)-block into


Figure 7.13: Execution of the partition refinement algorithm for a small example. \(\left\{s_{1}, s_{2}\right\}=\left\{s_{1}, s_{2}, s_{3}\right\} \cap \operatorname{Pre}\left(C_{2}\right)\) and \(\left\{s_{3}\right\}=\left\{s_{1}, s_{2}, s_{3}\right\} \backslash \operatorname{Pre}\left(C_{2}\right)\), yielding
\[
\Pi_{2}=\left\{\left\{s_{1}, s_{2}\right\},\left\{s_{3}\right\},\left\{t_{1}\right\},\left\{t_{2}, t_{3}\right\},\left\{u_{1}, u_{2}\right\},\left\{v_{1}, v_{2}\right\}\right\} .
\]

In the third refinement step, consider \(C_{3}=\left\{t_{2}, t_{3}\right\}\). This splitter distinguishes states \(s_{1}\) and \(s_{2}-s_{1}\) has no transition to \(C_{3}\), while \(s_{2}\) does-and yields the partition
\[
\Pi_{3}=\left\{\left\{s_{1}\right\},\left\{s_{2}\right\},\left\{s_{3}\right\},\left\{t_{1}\right\},\left\{t_{2}, t_{3}\right\},\left\{u_{1}, u_{2}\right\},\left\{v_{1}, v_{2}\right\}\right\} .
\]

No further refinements are possible, and \(\Pi_{3}\) thus agrees with the bisimulation quotient of the original transition system.

A possible implementation of the operator \(\operatorname{Refine}(\Pi, C)\) is based on a list representation of the sets of predecessors \(\operatorname{Pre}\left(s^{\prime}\right), s^{\prime} \in S\). For any state \(s^{\prime} \in C\), the list of predecessors \(\operatorname{Pre}\left(s^{\prime}\right)\) is traversed, and \(s \in \operatorname{Pre}\left(s^{\prime}\right)\) is "moved" from the (data structure representing the) current block \(B=[s]_{\Pi}\) to the block
\[
[s]_{\operatorname{Refine}(\Pi, C)}=B \cap \operatorname{Pre}(C)
\]

The states remaining in \(B\) form the subblock \(B \backslash \operatorname{Pre}(C)\). In case all states \(s \in B\) are moved from \(B\) to \(B \cap \operatorname{Pre}(C)\), the set of states in the data structure representing block \(B\) is empty and we have \(B \cap \operatorname{Pre}(C)=B\) and \(B \backslash \operatorname{Pre}(C)=\varnothing\). In case no state is moved from \(B\) to \(B \cap \operatorname{Pre}(C)\), we have, \(B=B \backslash \operatorname{Pre}(C)\) and \(B \cap \operatorname{Pre}(C)=\varnothing\).

By means of the above-described technique, every state \(s^{\prime} \in C\) causes the costs \(\mathcal{O}\left(\left|\operatorname{Pre}\left(s^{\prime}\right)\right|+\right.\) 1), provided appropriate data structures are used to represent \(T S\) (or its state graph), and the partition \(\Pi\). (The summand 1 in the complexity bound reflects the case \(\operatorname{Pre}\left(s^{\prime}\right)=\varnothing\); although these summands can be omitted when considering the complexity per state, they might be relevant when considering the complexity over all states.) Examples of such data structures are, e.g., an adjacency list representation for \(\operatorname{Pre}(\cdot)\), and a bit-vector representation of the satisfaction sets \(\operatorname{Sat}(a)=\{s \in S \mid a \in L(s)\}\) to represent the labeling function. Furthermore, list representations of partition \(\Pi\) and of the blocks \(B \in \Pi\) can be used. Using the fact that
\[
\mathcal{O}\left(\sum_{s^{\prime} \in C}\left(\left|\operatorname{Pre}\left(s^{\prime}\right)\right|+1\right)\right)=\mathcal{O}(|\operatorname{Pre}(C)|+|C|)
\]
we obtain for the time complexity of the refinement operator:

\section*{Lemma 7.40. Time Complexity of the Refinement Operator}

Refine \((\Pi, C)\) can be computed in time \(\mathcal{O}(|\operatorname{Pre}(C)|+|C|)\).

\subsection*{7.3.3 A First Partition Refinement Algorithm}

So far we presented the partition refinement algorithm without specifying any search strategy for the splitters. Such search strategy prescribes how to determine a splitter \(C\) for a given partition \(\Pi_{i+1}\). A simple strategy that will be pursued in this section, is to use the blocks of the previous partition \(\Pi_{i}\) (where \(\Pi_{-1}=\{S\}\) ) as splitter candidates for \(\Pi_{i+1}\), see Algorithm 31. In every outermost iteration, the refine operator causes for a state \(s \in S\) (as part of \(\Pi_{\text {old }}\) ), the cost \(\mathcal{O}(|\operatorname{Pre}(s)|+1\) ), see page 486 . The outermost iteration is traversed maximally \(|S|\) times; this occurs when in every iteration exactly one state is separated, i.e., constitutes a separate (singleton) block. This yields for the total costs of the iteration:
\[
\mathcal{O}\left(|S| \cdot \sum_{s^{\prime} \in S}\left(\left|\operatorname{Pre}\left(s^{\prime}\right)\right|+1\right)\right)=\mathcal{O}(|S| \cdot(M+|S|))
\]
where
\[
M=\sum_{s^{\prime} \in S}\left|\operatorname{Pre}\left(s^{\prime}\right)\right|
\]
```

Algorithm 31 A first partition refinement algorithm
Input: finite transition system $T S$ with the state space $S$
Output: bisimulation quotient space $S / \sim$
$\Pi:=\Pi_{A P} ;$
$\Pi_{\text {old }}:=\{S\} ; \quad\left(* \Pi_{\text {old }}\right.$ is the previous partition $\left.*\right)$
(* loop invariant: $\Pi$ is coarser than $S / \sim$ and finer than $\Pi_{A P}$ and $\left.\Pi_{o l d}{ }^{*}\right)$
repeat
$\Pi_{\text {old }}:=\Pi$;
for all $C \in \Pi_{\text {old }}$ do
$\Pi:=\operatorname{Refine}(\Pi, C)$;
od
until $\Pi=\Pi_{\text {old }}$
return $\Pi$

```
denotes the number of edges in the state graph \(G(T S)\). Assuming \(M \geqslant|S|\), this can be simplified to \(\mathcal{O}(|S| \cdot M)\). To obtain the total time complexity of Algorithm 31, it remains to consider the cost of computing \(\Pi_{A P}\), which is \(\Theta(|S| \cdot|A P|)\), as stated in Lemma 7.33 on page 480 . This yields:

\section*{Theorem 7.41. Time Complexity of Algorithm 31}

The bisimulation quotient space of TS can be computed by Algorithm 31 with time complexity \(\mathcal{O}(|S| \cdot(|A P|+M))\), under the assumption that \(M \geqslant|S|\), where \(M\) is the number of edges in the state graph \(G(T S)\).

Computing the bisimulation quotient is thus linear in the number of states of TS. In the next section, a strategy is presented that improves this such that the time complexity is logarithmic in \(|S|\).

\subsection*{7.3.4 An Efficiency Improvement}

The crucial observation that allows for an improvement of the time complexity is that it is not necessary to refine with respect to all blocks \(C \in \Pi_{\text {old }}\) as in Algorithm 31. Instead, it suffices to only consider roughly half of them, viz. the smaller subblocks of a previous refinement. More precisely, if a block \(C^{\prime}\) of the current partition \(\Pi\) is decomposed into subblocks \(C_{1}=C^{\prime} \cap \operatorname{Pre}(D)\) and \(C_{2}=C^{\prime} \backslash \operatorname{Pre}(D)\), only the smaller subblock is used as a splitter candidate in the following iteration. Now, let \(C \in\left\{C_{1}, C_{2}\right\}\) such that
\[
|C| \leqslant\left|C^{\prime}\right| / 2, \text { thus }|C| \leqslant\left|C^{\prime} \backslash C\right|
\]

The decomposition of the blocks \(B \in \operatorname{Refine}(\Pi, D)\) with respect to \(C\) and \(C^{\prime} \backslash C\) is now slightly modified by combining the refinement steps with respect to \(C\) and \(C^{\prime} \backslash C\). To make such "simultaneous" refinement possible, the algorithm exploits a ternary (instead of the previous binary) refinement operator:
\[
\operatorname{Refine}\left(\Pi, C, C^{\prime} \backslash C\right)=\operatorname{Refine}\left(\operatorname{Refine}(\Pi, C), C^{\prime} \backslash C\right)
\]
where it is assumed that \(|C| \leqslant\left|C^{\prime} \backslash C\right|\). This modification of the refinement operator is necessary to ensure that the decomposed blocks are stable with respect to \(C\) and \(C^{\prime} \backslash C\). As before, the ternary refinement operator decomposes each block into subblocks:
\[
\operatorname{Refine}\left(\Pi, C, C^{\prime} \backslash C\right)=\bigcup_{B \in \Pi} \operatorname{Refine}\left(B, C, C^{\prime} \backslash C\right)
\]

The effect of the refinement operator \(\operatorname{Refine}\left(B, C, C^{\prime} \backslash C\right)\) is schematically depicted for \(B \subseteq \operatorname{Pre}\left(C^{\prime}\right)\) in Figure 7.14. Thus, every block \(B \in \Pi\) with \(B \subseteq \operatorname{Pre}\left(C^{\prime}\right)\) is decomposed

block \(B\)
Figure 7.14: Ternary refinement operator.
into maximally three subblocks: states in \(B\) that only have direct successors in \(C\), those that only have such direct successors in \(C^{\prime}\), and the rest, i.e., states that have direct successors in both \(C\) and \(C^{\prime}\). Formally:
\[
\operatorname{Refine}\left(B, C, C^{\prime} \backslash C\right)=\left\{B_{1}, B_{2}, B_{3}\right\} \backslash\{\varnothing\}
\]
where
\[
\begin{aligned}
& B_{1}=B \cap \operatorname{Pre}(C) \cap \operatorname{Pre}\left(C^{\prime} \backslash C\right), \\
& B_{2}=(B \cap \operatorname{Pre}(C)) \backslash \operatorname{Pre}\left(C^{\prime} \backslash C\right), \\
& B_{3}=\left(B \cap \operatorname{Pre}\left(C^{\prime} \backslash C\right)\right) \backslash \operatorname{Pre}(C) .
\end{aligned}
\]

Note that the blocks \(B_{1}, B_{2}, B_{3}\) are stable with respect to \(C\) and \(C^{\prime} \backslash C\).
If \(B \cap \operatorname{Pre}\left(C^{\prime}\right)=\varnothing\), then \(B\) is stable with respect to \(C\) and \(C^{\prime} \backslash C\), in which case \(\operatorname{Refine}\left(B, C, C^{\prime} \backslash C\right)=\{B\}\). This suggests as loop invariant of the algorithm:
each block \(B \in \Pi\) is stable with respect to all blocks in \(\Pi_{o l d}\)

From this invariant it follows that only the two cases \(B \cap \operatorname{Pre}\left(C^{\prime}\right)=\varnothing\) and \(B \subseteq \operatorname{Pre}\left(C^{\prime}\right)\) are possible.

Algorithm 32 outlines the main steps of the improved partition refinement algorithm. To establish the aforementioned loop invariant, the initial partition is \(\operatorname{Refine}\left(\Pi_{A P}, S\right)\), i.e., each block only contains equally-labeled states that are all either terminal or not. This is based on the observation that \(\operatorname{Pre}(S)=\{s \in S \mid s\) is nonterminal \(\}\). Instead of first computing \(\Pi_{A P}\) and then applying the refinement operator, Refine \(\left(\Pi_{A P}, S\right)\) can be obtained by executing Algorithm 29 with \(A P\) extended with a special symbol which identifies nonterminal states. The initial partition can thus (as before) be determined in time \(\Theta(|S| \cdot|A P|)\).
```

Algorithm 32 An improved partition refinement algorithm
Input: finite transition system $T S$ with state space $S$
Output: bisimulation quotient space $S / \sim$

```
\(\Pi_{\text {old }}:=\{S\} ;\)
\(\Pi:=\operatorname{Refine}\left(\Pi_{A P}, S\right) ;\)
\(\quad\left(*\right.\) loop invariant: \(\Pi^{\Pi}\) is coarser than \(S / \sim\) and finer than \(\Pi_{A P}\) and \(\left.\Pi_{\text {old }}, *\right)\)
\(\quad\left(*\right.\) and \(\Pi\) is stable with respect to any block in \(\left.\Pi_{\text {old }}{ }^{*}\right)\)
repeat
    \(\Pi_{\text {old }}:=\Pi\);
    choose block \(C^{\prime} \in \Pi_{\text {old }} \backslash \Pi\) and block \(C \in \Pi\) with \(C \subseteq C^{\prime}\) and \(|C| \leqslant \frac{\left|C^{\prime}\right|}{2}\);
    \(\Pi:=\operatorname{Refine}\left(\Pi, C, C^{\prime} \backslash C\right)\);
until \(\Pi=\Pi_{\text {old }}\)
return \(\Pi\)

Example 7.42. Abstract Example of Algorithm 32
Consider the quotienting of the transition system in Figure 7.15. Since all black states are terminal, and all white states are not, we have \(\Pi_{A P}=\operatorname{Refine}\left(\Pi_{A P}, S\right)\).

In the first iteration, one may split with respect to the white or the black states. As \(\left|\left\{u_{1}, \ldots, u_{8}, w_{1}, w_{2}, w_{3}\right\}\right|>\frac{|S|}{2}\), the set of white states is not a suitable splitter. In fact, the only splitter candidate is block \(C=\left\{v_{1}, v_{2}\right\}\).
\[
\operatorname{Refine}(\Pi_{A P}, \underbrace{\left\{v_{1}, v_{2}\right\}}_{C}, \underbrace{\left\{u_{1}, u_{2}, \ldots, u_{8}, w_{1}, w_{2}, w_{3}\right\}}_{C^{\prime} \backslash C})
\]
\[
\mathrm{AP}=\{\bigcirc, \bigcirc\}
\]


Figure 7.15: Example of a transition system.
(with \(C^{\prime}=S\) ), decomposes block \(B=\left\{u_{1}, u_{2}, \ldots, u_{6}, u_{7}, u_{8}, w_{1}, w_{2}, w_{3}\right\}\) into
\[
\begin{aligned}
& B_{1}=B \cap \operatorname{Pre}(C) \cap \operatorname{Pre}\left(C^{\prime} \backslash C\right)=\left\{u_{7}\right\}, \\
& B_{2}=(B \cap \operatorname{Pre}(C)) \backslash \operatorname{Pre}\left(C^{\prime} \backslash C\right)=\left\{u_{1}, u_{2}, \ldots, u_{6}\right\} \cup\left\{u_{8}\right\}, \\
& B_{3}=\left(B \cap \operatorname{Pre}\left(C^{\prime} \backslash C\right)\right) \backslash \operatorname{Pre}(C)=\left\{w_{1}, w_{2}, w_{3}\right\} .
\end{aligned}
\]

This yields \(\Pi_{\text {old }}=\left\{C, C^{\prime} \backslash C\right\}\) and \(\Pi=\left\{C, B_{1}, B_{2}, B_{3}\right\}\).
In the second iteration, the blocks \(B_{1}=\left\{u_{7}\right\}\) and \(B_{3}=\left\{w_{1}, w_{2}, w_{3}\right\}\) are potential splitters. \(C\) is not considered as \(C \notin \Pi_{\text {old }} \backslash \Pi\). Block \(B_{2}\) is not considered as splitter, since it is too large with respect to its superblock in \(\Pi_{\text {old }}:\left|B_{2}\right|=7>11 / 2=\frac{\left|C^{\prime} \backslash C\right|}{2}\). Suppose \(B_{1}(\) called \(D)\) is selected as splitter; its superblock in \(\Pi_{\text {old }}\) equals \(D^{\prime}=C^{\prime} \backslash C \in \Pi_{\text {old }}\). Figure 7.16 illustrates the partition obtained by
\[
\operatorname{Refine}(\Pi, \underbrace{\left\{u_{7}\right\}}_{=D}, \underbrace{\left\{u_{1}, \ldots, u_{6}, u_{8}, w_{1}, w_{2}, w_{3}\right\}}_{=D^{\prime} \backslash D}) .
\]

The blocks \(C=\left\{v_{1}, v_{2}\right\}\) and \(B_{2}\) remain unchanged, since they do not have any direct successor in \(D^{\prime} \backslash D\). Block \(B_{3}\) is refined as follows:
\[
\operatorname{Refine}\left(\left\{w_{1}, w_{2}, w_{3}\right\},\left\{u_{7}\right\},\left\{u_{1}, \ldots, u_{6}, u_{8}, w_{1}, w_{2}, w_{3}\right\}\right)=\left\{\left\{w_{1}\right\},\left\{w_{2}\right\},\left\{w_{3}\right\}\right\}
\]
since \(w_{1}\) has only direct successors in \(D^{\prime} \backslash D, w_{2}\) can only move to \(D\), and \(w_{3}\) can move


Figure 7.16: Example for Algorithm 32.
to \(D\) as well as to \(D^{\prime} \backslash D\). Thus, at the end of the second iteration:
\[
\begin{aligned}
\Pi & =\left\{\left\{v_{1}, v_{2}\right\},\left\{u_{7}\right\},\left\{u_{1}, \ldots, u_{6}, u_{8}\right\},\left\{w_{1}\right\},\left\{w_{2}\right\},\left\{w_{3}\right\}\right\}, \\
\Pi_{\text {old }} & =\{\underbrace{\left\{v_{1}, v_{2}\right\}}_{=C}, \underbrace{\left\{u_{7}\right\}}_{=D}, \underbrace{\left\{u_{1}, \ldots, u_{6}, u_{8}, w_{1}, w_{2}, w_{3}\right\}}_{=D^{\prime} \backslash D}\} .
\end{aligned}
\]

The next refinement does not impose any further splittings, and \(\Pi=S / \sim\).

Refine \(\left(\Pi, C, C^{\prime} \backslash C\right)\) can be realized with time complexity \(\mathcal{O}(|\operatorname{Pre}(C)|+|C|)\) provided every state \(s^{\prime} \in \operatorname{Pre}(C)\) causes the costs \(\mathcal{O}\left(\left|\operatorname{Pre}\left(s^{\prime}\right)\right|+1\right)\). This can be established as follows. For every pair \(\left(s, C^{\prime}\right)\) with \(s \in \operatorname{Pre}\left(C^{\prime}\right), C^{\prime} \in \Pi_{\text {old }}\), a counter \(\delta\left(s, C^{\prime}\right)\) is introduced that keeps track of the number of direct successors of \(s\) in \(C^{\prime}\) :
\[
\delta\left(s, C^{\prime}\right)=\left|\operatorname{Post}(s) \cap C^{\prime}\right|
\]

During the execution of Refine \(\left(\Pi, C, C^{\prime} \backslash C\right)\), the values \(\delta(s, C)\) and \(\delta\left(s, C^{\prime} \backslash C\right)\) are computed for any \(s \in \operatorname{Pre}(C)\) as follows:
```

for all $s^{\prime} \in C$ do
for all $s \in \operatorname{Pre}\left(s^{\prime}\right)$ do
$\delta(s, C):=\delta(s, C)+1 ;$
od
od
for all $s \in \operatorname{Pre}(C)$ do
$\delta\left(s, C^{\prime} \backslash C\right):=\delta\left(s, C^{\prime}\right)-\delta(s, C) ;$
od

```
where it is assumed that initially \(\delta(s, C)=0\).

Let \(B \in \Pi\) be a block with \(B \subseteq \operatorname{Pre}\left(C^{\prime}\right)\), which should be decomposed into \(B_{1}, B_{2}, B_{3}\) by means of \(\operatorname{Refine}\left(B, C, C^{\prime} \backslash C\right)\). Then \(B_{1}, B_{2}\), and \(B_{3}\) are obtained as follows:
\[
\begin{aligned}
& B_{1}=B \cap \operatorname{Pre}(C) \cap \operatorname{Pre}\left(C^{\prime} \backslash C\right)=\left\{s \in B \mid \delta(s, C)>0, \delta\left(s, C^{\prime} \backslash C\right)>0\right\} \\
& B_{2}=B \cap \operatorname{Pre}(C) \backslash \operatorname{Pre}\left(C^{\prime} \backslash C\right)=\left\{s \in B \mid \delta(s, C)>0, \delta\left(s, C^{\prime} \backslash C\right)=0\right\} \\
& B_{3}=B \cap \operatorname{Pre}\left(C^{\prime} \backslash C\right) \backslash \operatorname{Pre}(C)=\left\{s \in B \mid \delta(s, C)=0, \delta\left(s, C^{\prime} \backslash C\right)>0\right\}
\end{aligned}
\]

The decomposition of block \(B \in \Pi\) is realized by "moving" the states \(s \in \operatorname{Pre}(C)\) from block \(B\) to block \(B_{1}\) or \(B_{2}\). The states remaining in \(B\) represent block \(B_{3}\).

The initial values \(\delta(s, C)\) and \(\delta\left(s, C^{\prime} \backslash C\right)\) need only be determined for the states in Pre \((C)\). The initial values of counters for \(s \in \operatorname{Pre}\left(C^{\prime}\right) \backslash \operatorname{Pre}(C)\) are derived by \(\delta\left(s, C^{\prime} \backslash C\right)=\delta\left(s, C^{\prime}\right)\), and \(\delta(s, C)=0\). For these states, the variable \(\delta(s, C)\) is not needed. Variable \(\delta\left(s, C^{\prime}\right)\) for state \(s \in \operatorname{Pre}\left(C^{\prime}\right) \backslash \operatorname{Pre}(C)\) and block \(C^{\prime}\) can be identified with variable \(\delta\left(s, C^{\prime} \backslash C\right)\) for the new block \(C^{\prime} \backslash C\). These considerations lead to the following lemma:

\section*{Lemma 7.43. Time Complexity of the Ternary Refinement Operator}

The time complexity of Refine \(\left(\Pi, C, C^{\prime} \backslash C\right)\) is in \(\mathcal{O}(|\operatorname{Pre}(C)|+|C|)\).

As asserted by the following theorem, the time complexity of the improved quotienting algorithm (see Algorithm 32) is logarithmic in the number of states.

\section*{Theorem 7.44. Time Complexity of Algorithm 32}

The bisimulation quotient of a finite transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) can be computed with Algorithm 32 in \(\mathcal{O}(|S| \cdot|A P|+M \cdot \log |S|)\).

Proof: The time complexity of Algorithm 32 is bounded above by
\[
\mathcal{O}\left(|S| \cdot|A P|+\sum_{s \in S} K(s) \cdot(|\operatorname{Pre}(s)|+1)\right)
\]
where \(K\left(s^{\prime}\right)\) denotes the number of blocks \(C\) containing \(s^{\prime}\) for which Refine \((\Pi, C \ldots)\) is invoked.

The first summand represents the asymptotic time needed to compute the initial partition Refine \(\left(\Pi_{A P}, S\right)\). This partition can be computed in the same way as \(\Pi_{A P}\) by extending the set of atomic propositions with a new label that indicates the terminal states. Recall from Lemma 7.33 (page 480) that \(\Pi_{A P}\) can be computed in \(\Theta(|S| \cdot|A P|)\). The fact that an additional atomic proposition is needed is not relevant.

The second summand stands for the cost of the refinement steps; recall that state \(s \in C\) induces the costs \(\mathcal{O}(|\operatorname{Pre}(s)|+1)\) on executing Refine \(\left(\Pi, C, C^{\prime} \backslash C\right)\), see Lemma 7.43 (page 492). This summand can be bounded from above as follows. We first observe that
\[
K(s) \leqslant \log |S|+1 \quad \text { for any state } s
\]

This is proven as follows. Let \(s \in S\) and \(C_{i}\) be the \(i\) th block with \(s \in C_{i}\), for which Refine \(\left(\Pi, C_{i}, \ldots\right)\) is executed. Then:
\[
\left|C_{i+1}\right| \leqslant \frac{\left|C_{i}\right|}{2} \quad \text { and } \quad\left|C_{1}\right| \leqslant|S|
\]

Let \(K(s)=k\). Then:
\[
1 \leqslant\left|C_{k}\right| \leqslant \frac{\left|C_{k-1}\right|}{2} \leqslant \frac{\left|C_{k-2}\right|}{4} \leqslant \ldots \leqslant \frac{\left|C_{k-i}\right|}{2^{i}} \leqslant \ldots \leqslant \frac{\left|C_{1}\right|}{2^{k-1}} \leqslant \frac{|S|}{2^{k-1}}
\]

From this, we conclude \(2^{k-1} \leqslant|S|\), or equivalently \(k-1 \leqslant \log |S|\). This yields \(K(s)=\) \(k \leqslant \log |S|+1\).

Let \(M=\sum_{s \in S}|\operatorname{Pre}(s)|\) and assume that a representation of the sets \(\operatorname{Pre}(\cdot)\) can be obtained in time \(\mathcal{O}(M)\). Thus:
\[
\begin{aligned}
\sum_{s^{\prime} \in S} K\left(s^{\prime}\right) \cdot\left(\left|\operatorname{Pre}\left(s^{\prime}\right)\right|+1\right) & \leqslant(\log |S|+1) \sum_{s^{\prime} \in S}\left(\left|\operatorname{Pre}\left(s^{\prime}\right)\right|+1\right) \\
& =(\log |S|+1) \cdot(M+|S|) \\
& \leqslant 2 \cdot(\log |S|+1) \cdot M=\mathcal{O}(M \cdot \log |S|)
\end{aligned}
\]

\subsection*{7.3.5 Equivalence Checking of Transition Systems}

The partition refinement algorithms can be used to verify the bisimilarity of the finite transition systems \(T S_{1}\) and \(T S_{2}\). To that end, the bisimulation quotient space of the composite transition system \(T S=T S_{1} \oplus T S_{2}\) (see page 457) is computed. Subsequently, it is checked whether
\[
C \cap I_{1}=\varnothing \quad \text { if and only if } \quad C \cap I_{2}=\varnothing
\]
for each bisimulation equivalence class \(C\) of the composite system \(T S\). Here, \(I_{i}\) denotes the set of initial states of \(T S_{i}, i=1,2\).

\section*{Corollary 7.45. Complexity of Checking Bisimulation Equivalence}

Checking whether \(T S_{1} \sim T S_{2}\) for finite transition systems \(T S_{1}\) and \(T S_{2}\) over AP with the state spaces \(S_{1}\) and \(S_{2}\), respectively, can be performed in time
\[
\mathcal{O}\left(\left(\left|S_{1}\right|+\left|S_{2}\right|\right) \cdot|A P|+\left(M_{1}+M_{2}\right) \cdot \log \left(\left|S_{1}\right|+\left|S_{2}\right|\right)\right) .
\]

Here, \(M_{i}\) denotes the number of edges in \(G\left(T S_{i}\right)\) and is assumed to be at least \(\left|S_{i}\right|(i=1,2)\).

Checking whether two transition systems are trace-equivalent is much harder. In the following theorem, the problem which asks whether two finite transition systems are trace equivalent is shown to be complete for the complexity class PSPACE. This means that the trace equivalence problem belongs to PSPACE, i.e., is solvable by a polynomial spacebounded algorithm, and is PSPACE-hard.

\section*{Theorem 7.46. Lower Bound on Checking Trace Equivalence}

Let \(T S_{1}\) and \(T S_{2}\) be finite transition systems over AP. Then:
(a) The problem whether \(\operatorname{Traces}_{f i n}\left(T S_{1}\right)=\operatorname{Traces}_{f i n}\left(T S_{2}\right)\) is PSPACE-complete.
(b) The problem whether \(\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)\) is PSPACE-complete.

Proof: We first prove membership of PSPACE. We prove that problems (a) and (b) are in PSPACE by providing a polynomial reduction from these problems to the equivalence problem for nondeterministic finite automata (NFA). As the latter problem is in PSPACE, the existence of this reduction implies that problems (a) and (b) are in PSPACE. The equivalence problem for NFA \(\mathcal{A}_{1}\) and \(\mathcal{A}_{2}\) is whether \(\mathcal{L}\left(\mathcal{A}_{1}\right)=\mathcal{L}\left(\mathcal{A}_{2}\right)\).

The main proof idea is to define NFA \(\mathcal{A}_{T S}\) for transition system \(T S\) such that \(\mathcal{L}\left(\mathcal{A}_{T S}\right)\) can be obtained from \(\operatorname{Traces}(T S)\), and vice versa. For finite transition system \(T S=(S, A c t, \rightarrow\), \(I, A P, L)\), let NFA \(\mathcal{A}_{T S}=\left(Q, \Sigma, \delta, Q_{0}, F\right)\) be defined as follows:
- \(Q=S \cup\{t\}\) with \(t \notin S\) (state \(t\) identifies terminal states in \(T S\) )
- \(\Sigma=2^{A P} \cup\{\tau\}\) with \(\tau \notin 2^{A P}\)
- The transition relation \(\delta\) in \(\mathcal{A}_{T S}\) is given by
\[
\delta(s, A)= \begin{cases}\operatorname{Post}(s) & \text { if } s \in S, \operatorname{Post}(s) \neq \varnothing \text { and } L(s)=A \\ \{t\} & \text { if }(s \in S \wedge \operatorname{Post}(s)=\varnothing \wedge L(s)=A) \\ & \vee(s=t \wedge A=\tau) \\ \varnothing & \text { otherwise }\end{cases}
\]
- \(Q_{0}=I\)
- \(F=Q\)

It now follows that \(\operatorname{Traces}(T S)\) is the set consisting of all finite, nonempty words \(A_{0} A_{1} \ldots A_{n}\) where \(A_{0} A_{1} \ldots A_{n} \tau \in \mathcal{L}\left(\mathcal{A}_{T S}\right)\) and all infinite words \(A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}\) where all prefixes \(A_{0} \ldots A_{n}\) belong to \(\mathcal{L}\left(\mathcal{A}_{T S}\right)\). Hence, \(\operatorname{Traces}(T S)\) can be derived from \(\mathcal{L}\left(\mathcal{A}_{T S}\right)\), and vice versa.

For finite transition systems \(T S_{1}\) and \(T S_{2}\) we have
\[
\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right) \quad \text { if and only if } \quad \mathcal{L}\left(\mathcal{A}_{T S_{1}}\right)=\mathcal{L}\left(\mathcal{A}_{T S_{2}}\right)
\]

As the NFA \(\mathcal{A}_{T S}\) are obtained in polynomial time, this yields a polynomial reduction from problem (b) to the equivalence problem for NFA.

A polynomial reduction from (a) to the equivalence problem for NFA is obtained by considering a similar transformation \(T S \mapsto \mathcal{A}_{T S}^{\prime}\) where \(\mathcal{A}_{T S}^{\prime}\) agrees with \(\mathcal{A}_{T S}\), except that the self-loop at state \(t\) is omitted. State \(t\) is thus the only terminal in \(\mathcal{A}_{T S}^{\prime}\). Then, \(\operatorname{Traces}_{f i n}(T S)\) agrees with the set of all nonempty words accepted by \(\mathcal{A}_{T S}^{\prime}\), i.e., \(\operatorname{Traces}_{f i n}(T S)=\mathcal{L}\left(\mathcal{A}_{T S}^{\prime}\right) \backslash\) \(\{\varepsilon\}\). Thus:
\[
\operatorname{Traces}_{f i n}\left(T S_{1}\right)=\operatorname{Traces}_{f i n}\left(T S_{2}\right) \quad \text { if and only if } \quad \mathcal{L}\left(\mathcal{A}_{T S_{1}}^{\prime}\right)=\mathcal{L}\left(\mathcal{A}_{T S_{2}}^{\prime}\right)
\]

It remains to show the PSPACE-hardness. This is shown by a polynomial reduction from the language-equivalence problem for NFA, a problem that is PSPACE-complete. PSPACE-hardness of this problem follows from the fact that the (universality) problem whether \(\mathcal{L}(E)=\Sigma^{*}\) for regular expression \(E\) over \(\Sigma\) is PSPACE-hard; see e.g. [383]. Since there are polynomial transformations from regular expressions to NFA and there is a singlestate NFA for \(\Sigma^{*}\), the universality problem is polynomial reducible to the equivalence problem for NFA.

The main idea is to map an NFA \(\mathcal{A}=\left(Q, \Sigma, \delta, Q_{0}, F\right)\) to a finite transition system \(T S_{\mathcal{A}}=\) \((S,\{\tau\}, \rightarrow, I, A P, L)\) such that \(\operatorname{Traces}\left(T S_{\mathcal{A}}\right)\) encodes \(\mathcal{L}(\mathcal{A})\). As a preprocessing step we delete all states \(q\) in \(\mathcal{A}\) with \(\delta^{*}(q, w) \cap F=\varnothing\) for any \(w \in \Sigma^{*}\), as from these states it is impossible to reach a final state. \(T S_{\mathcal{A}}\) is defined as
- \(S=Q_{0} \cup(Q \times \Sigma) \cup\{\) accept \(\}\) with \(Q_{0} \cap(Q \times \Sigma)=\varnothing\) and accept \(\notin Q \cup(Q \times \Sigma)\)
- \(\rightarrow\) is defined by the following rules:
\[
\frac{q_{0} \in Q_{0} \wedge p \in \delta\left(q_{0}, A\right)}{q_{0} \xrightarrow{\tau}\langle p, A\rangle} \quad \frac{q \in Q \wedge B \in \Sigma \wedge p \in \delta(q, A)}{\langle q, B\rangle \xrightarrow{\tau}\langle p, A\rangle}
\]
and
\[
\frac{q \in F \wedge B \in \Sigma}{\langle q, B\rangle \xrightarrow{\tau} \text { accept }} \quad \xrightarrow[{\text { accept } \xrightarrow{\tau} \text { accept }}]{ }
\]
- \(I=Q_{0}\)
- \(A P=\Sigma \cup\{\) accept \(\}\) with accept \(\notin \Sigma\)
- \(L\left(q_{0}\right)=\varnothing\) for any \(q_{0} \in Q_{0}, L(\langle q, A\rangle)=\{A\}\), and \(L\) (accept \()=\{\) accept \(\}\)

It is not difficult to establish that \(\operatorname{Traces}_{f i n}\left(T S_{\mathcal{A}}\right)\) consists of all prefixes of words of the form \(\varnothing\left\{A_{1}\right\} \ldots\left\{A_{n}\right\}\{\text { accept }\}^{m}\) with \(m \geqslant 0\) and \(A_{1} \ldots A_{n} \in \mathcal{L}(\mathcal{A}) . \mathcal{L}(\mathcal{A})\) can be derived from \(\operatorname{Traces}\left(T S_{\mathcal{A}}\right)\) as follows:
\[
A_{1} \ldots A_{n} \in \mathcal{L}(\mathcal{A}) \quad \text { iff } \quad \varnothing\left\{A_{1}\right\} \ldots\left\{A_{n}\right\}\{\operatorname{accept}\} \in \operatorname{Traces}_{f i n}\left(T S_{\mathcal{A}}\right)
\]

Thus, for NFA \(\mathcal{A}_{1}\) and \(\mathcal{A}_{2}\) over the alphabet \(\Sigma\) :
\[
\begin{array}{lll}
\mathcal{L}\left(\mathcal{A}_{1}\right)=\mathcal{L}\left(\mathcal{A}_{2}\right) & \text { iff } & \operatorname{Traces}_{f i n}\left(T S_{\mathcal{A}_{1}}\right)=\operatorname{Traces}_{f i n}\left(T S_{\mathcal{A}_{2}}\right) \\
& \text { iff } & \operatorname{Traces}\left(T S_{\mathcal{A}_{1}}\right)=\operatorname{Traces}\left(T S_{\mathcal{A}_{2}}\right)
\end{array}
\]

The equivalence of finite trace inclusion and trace inclusion follows from Theorem 3.30 on page 117 as the transition systems \(T S_{\mathcal{A}_{i}}\) are finite and do not have terminal states. Since \(T S_{\mathcal{A}}\) is obtained in time \(\mathcal{O}(|\mathcal{A}|)\) from \(\mathcal{A}\), the above yields a polynomial reduction from the equivalence problem for NFA to problems (a) and (b).

\subsection*{7.4 Simulation Relations}

Bisimulation relations are equivalences requiring two bisimilar states to exhibit identical stepwise behavior. On the contrary, simulation relations are preorders on the state space requiring that whenever \(s \preceq s^{\prime}\left(s^{\prime}\right.\) simulates \(\left.s\right)\), state \(s^{\prime}\) can mimic all stepwise behavior
of \(s\); the reverse, i.e., \(s^{\prime} \preceq s\) is not guaranteed, so state \(s^{\prime}\) may perform transitions that cannot be matched by state \(s\). Thus, if \(s^{\prime}\) simulates \(s\) then every successor of \(s\) has a corresponding, i.e., related successor of \(s^{\prime}\), but the reverse does not necessarily hold. Simulation can be lifted to entire transition systems by comparing (according to \(\preceq\) ) their initial states. Simulation relations are often used for verification purposes to show that one system correctly implements another, more abstract system. One of the interesting aspects of simulation relations is that they allow a verification by "local" reasoning. The transitivity of \(\preceq\) allows a stepwise verification in which the correctness is established via several intermediate systems. Simulation relations are therefore used as a basis for abstraction techniques where the rough idea is to replace the model to be verified by a smaller abstract model and to verify the latter instead of the original one.

The simulation order \(\preceq\) induces an equivalence which is coarser than bisimulation equivalence, and hence yields a better abstraction (i.e., a smaller quotient space), while still preserving a wide range of logical formulae in LTL and CTL. As bisimulation equivalence is the coarsest equivalence that preserves CTL and CTL* , the simulation order \(\preceq\) preserves a fragment of these logics. The use of simulation thus relies on the preservation of certain classes of formulae, not of all formulae. For instance, if \(s \preceq s^{\prime}\), then for any safety property \(\forall \varphi\) it follows that \(s^{\prime} \models \forall \varphi\) implies \(s \models \forall \varphi\), since any path starting in \(s\) is mimicked by a similar path that starts in \(s^{\prime}\). The reverse, \(s^{\prime} \not \vDash \forall \varphi\), cannot be used to deduce that \(\forall \varphi\) does not hold in the simulated state \(s\); the paths starting in \(s^{\prime}\) that violate \(\varphi\) might be behaviors that \(s\) cannot perform at all.

As for bisimulation relations, the formal definition of the simulation order relies on a coinductive approach which defines the simulation order as the greatest relation satisfying certain conditions:

\section*{Definition 7.47. Simulation Order}

Let \(T S_{i}=\left(S_{i}\right.\), Act \(\left._{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right), i=1,2\), be transition systems over AP. A simulation for \(\left(T S_{1}, T S_{2}\right)\) is a binary relation \(\mathcal{R} \subseteq S_{1} \times S_{2}\) such that
(A) \(\forall s_{1} \in I_{1} .\left(\exists s_{2} \in I_{2} .\left(s_{1}, s_{2}\right) \in \mathcal{R}\right)\)
(B) for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) it holds that:
(1) \(L_{1}\left(s_{1}\right)=L_{2}\left(s_{2}\right)\)
(2) if \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\), then there exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
\(T S_{1}\) is simulated by \(T S_{2}\) (or, equivalently, \(T S_{2}\) simulates \(T S_{1}\) ), denoted \(T S_{1} \preceq T S_{2}\), if there exists a simulation \(\mathcal{R}\) for \(\left(T S_{1}, T S_{2}\right)\).

Condition (A) requires that all initial states in \(T S_{1}\) are related to an initial state of \(T S_{2}\). As the reverse is not required, there might be initial states of \(T S_{2}\) that are not matched by an initial state of \(T S_{1}\). Conditions (B.1) and (B.2) are as for bisimulations; note that the symmetric counterpart of (B.2) -as for bisimulation-is not required.

\section*{Example 7.48. Two Beverage Vending Machines}

Consider the transition systems in Figure 7.17. Let \(A P=\{\) pay, beer, soda \(\}\) with the obvious labeling functions. The relation
\[
\mathcal{R}=\left\{\left(s_{0}, t_{0}\right),\left(s_{1}, t_{1}\right),\left(s_{2}, t_{1}\right),\left(s_{3}, t_{2}\right),\left(s_{4}, t_{3}\right)\right\}
\]
is a simulation for \(\left(T S_{1}, T S_{2}\right)\). Since \(\mathcal{R}\) contains the pair of initial states \(\left(s_{0}, t_{0}\right), T S_{1} \preceq\) \(T S_{2}\). The reverse does not hold, i.e., \(T S_{2} \npreceq T S_{1}\), since there is no state in \(T S_{1}\) that can mimic state \(t_{1}\). This is due to the fact that the options "beer" and "soda" are possible in state \(t_{1}\), but in no state in \(T S_{1}\).


Figure 7.17: The vending machine on the right simulates the one on the left.
Consider now \(A P=\{\) pay, drink \(\}\), and assume states \(s_{3}, s_{4}, t_{2}\) and \(t_{3}\) equals \(\{d r i n k\}\). Relation \(\mathcal{R}\) is again a simulation for \(\left(T S_{1}, T S_{2}\right)\), and thus \(T S_{1} \preceq T S_{2}\). Its inverse,
\[
\mathcal{R}^{-1}=\left\{\left(t_{0}, s_{0}\right),\left(t_{1}, s_{1}\right),\left(t_{1}, s_{2}\right),\left(t_{2}, s_{3}\right),\left(t_{3}, s_{4}\right)\right\}
\]
is a simulation for \(\left(T S_{2}, T S_{1}\right)\). We thus also obtain \(T S_{2} \preceq T S_{1}\).

The simulation relation \(\preceq\) is a preorder, i.e., it is reflexive and transitive. As the conditions of simulation relations are not symmetric, the simulation preorder is not an equivalence.

\section*{Lemma 7.49. Reflexivity and Transitivity of \(\preceq\)}

For a fixed set AP of atomic propositions, the relation \(\preceq\) is reflexive and transitive.

Proof: Similar to the proof of reflexivity and transitivity of \(\sim\); see Lemma 7.4 (page 453).
\(T S_{1} \preceq T S_{2}\) holds if \(T S_{1}\) originates from \(T S_{2}\) by deleting transitions from \(T S_{2}\), e.g., in case of a nondeterministic choice in \(T S_{1}\), only one alternative is retained. In such cases, \(T S_{1}\) can be understood as a refinement of \(T S_{2}\), since \(T S_{1}\) resolves some nondeterminism in \(T S_{2} . T S_{1} \preceq T S_{2}\) also holds when \(T S_{2}\) arises from \(T S_{1}\) by means of an abstraction. In fact, abstraction is a fundamental concept that permits the analysis of large or even infinite transition systems. Abstraction is identified by a set of abstract states \(\widehat{S}\); an abstraction function \(f\), which associates to each (concrete) state \(s\) of the transition system \(T S\) the abstract state \(f(s)\) which represents it; and a set AP of atomic propositions which label the concrete and abstract states. Abstractions differ in the choice of the set \(\widehat{S}\) of abstract states, the abstraction function \(f\), and the relevant propositions AP. The concept of abstraction functions has, in fact, already been used in several examples in this monograph. For instance, an abstraction function was used for the Bakery algorithm in order to prove that the infinite transition system \(T S_{\text {Bak }}\) has a finite bisimulation quotient, see Example 7.13 (page 461). Due to the special structure of the transition system, here a bisimulation-equivalent transition system results. Typically, however, an abstract transition system results that simulates TS.

We briefly outline the essential ideas of abstractions that are obtained by aggregating disjoint sets of concrete states into single abstract states. Abstraction functions map concrete states onto abstract ones, such that abstract states are associated with equally labeled concrete states only.

\section*{Definition 7.50. Abstraction Function}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system, and \(\widehat{S}\) be a set of (abstract) states. \(f: S \rightarrow \widehat{S}\) is an abstraction function if for any \(s, s^{\prime} \in S: f(s)=f\left(s^{\prime}\right)\) implies \(L(s)=L\left(s^{\prime}\right)\).

The abstract transition system \(T S_{f}\) originates from \(T S\) by identifying all states that are represented by the same abstract state under abstraction function \(f\). An abstract state is initial whenever it represents an initial concrete state. Similarly, there is a transition from abstract state \(f(s)\) to state \(f\left(s^{\prime}\right)\) if there is a transition from \(s\) to \(s^{\prime}\).

\section*{Definition 7.51. Abstract Transition System}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a (concrete) transition system, \(\widehat{S}\) a set of (abstract) states, and \(f: S \rightarrow \widehat{S}\) an abstraction function. The abstract transition system \(T S_{f}=\) \(\left(\widehat{S}, A c t, \rightarrow_{f}, I_{f}, A P, L_{f}\right)\) induced by \(f\) on \(T S\) is defined by:
- \(\rightarrow_{f}\) is defined by the rule: \(\frac{s \xrightarrow{\alpha} s^{\prime}}{f(s) \xrightarrow{\alpha} f\left(s^{\prime}\right)}\)
- \(I_{f}=\{f(s) \mid s \in I\}\)
- \(L_{f}(f(s))=L(s)\) for all states \(s \in S\)

\section*{Lemma 7.52.}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a (concrete) transition system, \(\widehat{S}\) a set of (abstract) states, and \(f: S \rightarrow \widehat{S}\) an abstraction function. Then \(T S \preceq T S_{f}\).

Proof: Follows from the fact that \(\mathcal{R}=\{(s, f(s)) \mid s \in S\}\) is a simulation for \(\left(T S, T S_{f}\right)\).

Example 7.53. An Automatic Door Opener
Consider an automatic door opener as modeled by the (concrete) transition system in Figure 7.18; for simplicity, the action labels are omitted from the transitions. Let \(A P=\) \{alarm, open \(\}\). The door opener requires a three-digit code \(d_{1} d_{2} d_{3}\) as input with \(d_{i} \in\) \(\{0, \ldots, 9\}\). It allows an erroneous digit to be entered, but this may happen at most twice. The variable error keeps track of the number of wrong digits that have been provided, and is initially zero. In case error exceeds two, the door opener issues an alarm signal. On a successful input of the door code, the door is opened. Once locked again, it returns to its initial state. Location \(\ell_{i}\) (for \(i=0,1,2\) ) indicates that the first \(i\) digits of the code have been correctly entered; the second component of a state indicates the value of the variable error (if applicable).

Consider the following two abstractions. In the first abstraction, the concrete states are aggregated as indicated by the dashed ovals in Figure 7.18. In fact, this is a data abstraction in which the domain of the variable error is restricted to \(\{\leqslant 1,2\}\), i.e., the values 0 and 1 are not distinguished in the abstract transition system. The corresponding abstraction
function \(f\) is defined by
\[
f(\langle\ell, \text { error }=k\rangle)= \begin{cases}\langle\ell, \text { error } \leqslant 1\rangle & \text { if } k \in\{0,1\} \\ \langle\ell, \text { error }=2\rangle & \text { if } k=2\end{cases}
\]

For all other concrete states, \(f\) is the identity function. It immediately follows that \(f\) is indeed an abstraction function, as only equally labeled states are mapped onto the same abstract state. The abstract transition system \(T S_{f}\) is depicted in Figure 7.19; by construction we have \(T S \preceq T S_{f}\).


Figure 7.18: Transition system of the door opener


Figure 7.19: Abstract transition system of the door opener (under abstraction function \(f\) )

Now consider an abstraction in which we completely abstract from the value of the variable error. The concrete states that are aggregated are indicated by the dashed ovals in Figure
7.20. The corresponding abstraction function \(g\) is defined by \(g(\langle\ell\), error \(=k\rangle)=\ell\), for any location \(\ell \in\left\{\ell_{0}, \ell_{1}, \ell_{2}\right\} ; g\) is the identity function otherwise. This yields the transition system \(T S_{g}\) in Figure 7.21. As \(g\) is indeed an abstraction function, it follows that \(T S \preceq\) \(T S_{g}\).


Figure 7.20: Alternative aggregation of states for the door opener.


Figure 7.21: Abstract transition system of the door opener (under abstraction function g).

The abstract transition systems in the previous example have been obtained by choosing for the variable error in the system an abstract domain that is smaller than its original domain \(\{0,1,2\}\). Let us illustrate what such a type of abstraction (also known as data abstraction) means for program graphs. The essential idea of abstraction in program graphs is to abstract from concrete values of certain program variables or locations, i.e., program counters. Rather than keeping track of the concrete and precise value of variables, only abstract values are considered. For instance, a possible abstraction of an integer \(x\) could be a three-valued variable which is -1 whenever \(x<0,0\) whenever \(x=0\), and 1 if \(x>1\). Such abstraction is useful when only the sign of \(x\) is of importance.

\section*{Example 7.54. Data Abstraction applied to a Simple Sequential Program}

Consider the following program fragment on the nonnegative integer variables \(x\) and \(y\) :
```

\ell
\ell}\quadx:=x-1
\ell2 }\quady:=y+1
od;
\ell if even(y) then return " 1" else return "0" fi;
\ell

```

Let \(P G\) be the program graph of this program, and \(T S\) its underlying transition system \(T S(P G)\). Each state of \(T S\) is of the form \(s=\langle\ell, x=n, y=m\rangle\), where \(\ell\) is a program location and \(m\) and \(n\) are natural numbers. By means of abstraction one of the variables may be completely ignored or the variable domains are restricted. We exemplify the second kind of abstraction and restrict \(x\) and \(y\) to the domains
\[
\operatorname{dom}_{\text {abstract }}(x)=\{\text { gzero, zero }\} \quad \text { and } \quad \operatorname{dom}_{\text {abstract }}(y)=\{\text { even, odd }\} .
\]

Stated in words, we only keep track of whether \(x>0\) or \(x=0\), and whether \(y\) is even or odd. The precise values of \(x\) and \(y\) are not explicitly administered.

The abstraction function \(f\) which maps a concrete state \(\langle\ell, x=n, y=m\rangle\) (with \(m, n \in \mathbb{N}\) ) onto an abstract state \(\langle\ell, x=V, y=W\rangle\) (with \(V \in\{\) gzero, zero \(\}\) and \(W \in\{\) even, odd \(\}\) ) is defined as follows:
\[
f(\langle\ell, x=v, y=w\rangle)= \begin{cases}\langle\ell, x=\text { gzero, } y=\text { even }\rangle & \text { if } x>0 \wedge y \text { is even } \\ \langle\ell, x=\text { gzero, } y=\text { odd }\rangle & \text { if } x>0 \wedge y \text { is odd } \\ \langle\ell, x=\text { zero, } y=\text { even }\rangle & \text { if } x=0 \wedge y \text { is even } \\ \langle\ell, x=\text { zero, } y=o d d\rangle & \text { if } x=0 \wedge y \text { is odd }\end{cases}
\]

To obtain an abstract transition system \(T S_{f}\) with \(T S \preceq T S_{f}\), the operations in \(T S\) (e.g., incrementing \(y\) ) have to be replaced with corresponding abstract operations that yield values from the abstract domains. Accordingly, the statement \(y:=y+1\) is replaced with the abstract operation:
\[
y \mapsto \begin{cases}\text { even } & \text { if } y=\text { odd } \\ \text { odd } & \text { if } y=\text { even }\end{cases}
\]

The outcome of the statement \(x:=x-1\) depends on the value of \(x\). In the abstract setting, the precise value of \(x\) is, however, not known. Thus, \(x:=x-1\) corresponds to the abstract nondeterministic statement:
\[
x:=\text { gzero or } \quad x:=\text { zero }
\]
(If the guard of the while loop would be \(x>1\), then there would be no reason for nondeterminism in the abstract program.)

The abstract transition system \(T S_{f}\) results from the following piece of program:
```

$\ell_{0} \quad$ while ( $x=$ gzero ) do
$\ell_{1} \quad x:=$ gzero or $x:=$ zero;
$\ell_{2} \quad$ if $y=$ even then $y:=o d d$ else $y:=$ even fi;
od;
$\ell_{3} \quad$ if $y=e v e n$ then return " 1 " else return " 0 " fi;
$\ell_{4} \quad \vdots$

```

Note that the abstract program originates from syntactic transformations which can be completely automated. The previous considerations show that
\[
\mathcal{R}=\{(s, f(s)) \mid s \in \operatorname{Loc} \times \operatorname{Eval}(x \in \mathbb{N}, y \in \mathbb{N})\}
\]
is a simulation for \(\left(T S, T S_{f}\right)\), provided that the initial values of the abstract variables are chosen in accordance with the initial values of the concrete variables, and the set of atomic propositions is (a subset of)
\[
A P=\left\{\ell_{0}, \ell_{1}, \ell_{2}, \ell_{3}, \ell_{4}, x>0, x=0, \operatorname{even}(y), \operatorname{odd}(y)\right\}
\]
used with the obvious labeling function.

If \(\mathcal{R}\) is a simulation and \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) then each path fragment \(\pi_{1}\) of \(s_{1}\) can be lifted to a path fragment \(\pi_{2}\) of \(s_{2}\) such that \(\pi_{1}\) and \(\pi_{2}\) are statewise related via \(\mathcal{R}\), see Figure 7.22.

\section*{Lemma 7.55. Simulation on Path Fragments}

Let \(T S_{1}\) and \(T S_{2}\) be transition systems over \(A P, \mathcal{R}\) a simulation for \(\left(T S_{1}, T S_{2}\right)\), and \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) Then for each (finite or infinite) path fragment \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots\) starting in \(s_{0,1}=s_{1}\) there exists a path fragment \(\pi_{2}=s_{0,2} s_{1,2} s_{2,2} \ldots\) from state \(s_{0,2}=s_{2}\) of the same length such that \(\left(s_{j, 1}, s_{j, 2}\right) \in \mathcal{R}\) for all \(j\).

Proof: The proof is similar to Lemma 7.5 (page 454).

Hence, whenever \(\mathcal{R}\) is a simulation that contains \(\left(s_{1}, s_{2}\right)\) and \(\pi_{1}\) an infinite path from state \(s_{1}\) then there exists an infinite path from \(s_{2}\) such that \(\pi_{1}\) and \(\pi_{2}\) are statewise \(\mathcal{R}\)-related. The labeling condition ensures that the traces of \(\pi_{1}\) and \(\pi_{2}\) agree. This yields that all infinite traces of \(s_{1}\) are at the same time infinite traces of \(s_{2}\). If, however, \(\pi_{1}\) is a finite


Figure 7.22: Path fragment lifting for simulations.
path from \(s_{1}\) then \(\pi_{2}\) is a finite path fragment from \(s_{2}\), but possibly not maximal (i.e., the last state of \(\pi_{2}\) might be nonterminal). Hence, state \(s_{1}\) can have finite traces, which are trace fragments, but not traces of \(s_{2}\).

\subsection*{7.4.1 Simulation Equivalence}

The simulation relation \(\preceq\) is transitive and reflexive, but not symmetric. An example where \(T S_{1} \preceq T S_{2}\), but \(T S_{2} \npreceq T S_{1}\), was provided in Example 7.48 (page 498). However, as any preorder, \(\preceq\) induces an equivalence relation, the so-called kernel of \(\preceq\), which is defined by \(\simeq=\preceq \cap \preceq^{-1}\). It consists of all pairs \(\left(T S_{1}, T S_{2}\right)\) of transition systems that can mutually simulate each other. The relation \(\simeq\) is called simulation equivalence.

\section*{Definition 7.56. Simulation Equivalence}

Transition systems \(T S_{1}\) and \(T S_{2}\) (over AP) are simulation-equivalent, denoted \(T S_{1} \simeq T S_{2}\), if \(T S_{1} \preceq T S_{2}\) and \(T S_{2} \preceq T S_{1}\).

\section*{Example 7.5\%. Simulation Equivalent Transition Systems}

The transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) in Figure 7.23 are simulation-equivalent. This can be seen as follows. Since \(T S_{1}\) is a subgraph of \(T S_{2}\) (up to isomorphism, i.e., state identities), it is clear that \(T S_{1} \preceq T S_{2}\). Consider now the reverse direction. The transition \(t_{1} \rightarrow t_{2}\) in \(T S_{2}\) is simulated by \(s_{1} \rightarrow s_{2}\). Formally,
\[
\mathcal{R}=\left\{\left(t_{1}, s_{1}\right),\left(t_{2}, s_{2}\right),\left(t_{3}, s_{2}\right),\left(t_{4}, s_{3}\right)\right\}
\]
is a simulation for \(\left(T S_{2}, T S_{1}\right)\). From this, we derive \(T S_{2} \preceq T S_{1}\) and obtain \(T S_{1} \simeq T S_{2}\).


Figure 7.23: Simulation-equivalent transition systems.

The beverage vending machines described in Example 7.48 (page 498) are not simulationequivalent if (as before) the set of propositions \(A P=\{\) pay, beer, soda \(\}\) is used. However, \(T S_{1} \simeq T S_{2}\) holds for \(A P=\{\) pay \(\}\) or for \(A P=\{\) pay, drink \(\}\).

The simulation preorder \(\preceq\) and its induced equivalence may also be adopted to compare states of a single transition system.

\section*{Definition 7.58. Simulation Order as a Relation on States}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system. A simulation for \(T S\) is a binary relation \(\mathcal{R} \subseteq S \times S\) such that for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) :
1. \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
2. If \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\), then there exists an \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).

State \(s_{1}\) is simulated by \(s_{2}\) (or \(s_{2}\) simulates \(s_{1}\) ), denoted \(s_{1} \preceq_{T S} s_{2}\), if there exists a simulation \(\mathcal{R}\) for \(T S\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\). States \(s_{1}\) and \(s_{2}\) of \(T S\) are simulation-equivalent, denoted \(s_{1} \simeq_{T S} s_{2}\), if \(s_{1} \preceq_{T S} s_{2}\) and \(s_{2} \preceq_{T S} s_{1}\).

For state \(s\), let \(\operatorname{Sim}_{T S}(s)\) denote the simulator set of \(s\), i.e., the set of states that simulate \(s\). Formally:
\[
\operatorname{Sim}_{T S}(s)=\left\{s^{\prime} \in S \mid s \preceq_{T S} s^{\prime}\right\}
\]

For convenience, the branching-time relations introduced so far are summarized in Figure 7.24.

As stated on page 457 for bisimulations, the relation \(\preceq_{T S}\) on \(S \times S\) results from \(\preceq\) via
\[
s_{1} \preceq_{T S} s_{2} \quad \text { if and only if } T S_{s_{1}} \preceq T S_{s_{2}}
\]
simulation order
\(s_{1} \preceq_{T S} s_{2}: \Leftrightarrow\) there exists a simulation \(\mathcal{R}\) for \(T S\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\)
simulation equivalence
\(s_{1} \simeq_{T S} s_{2}: \Leftrightarrow \quad s_{1} \preceq_{T S} s_{2}\) and \(s_{2} \preceq_{T S} s_{1}\)
bisimulation equivalence
\(s_{1} \sim_{\text {TS }} s_{2}: \Leftrightarrow\) there exists a bisimulation \(\mathcal{R}\) for \(T S\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\)

Figure 7.24: Summary of the relations \(\preceq_{T S}, \sim_{T S}\), and \(\simeq_{T S}\).
where \(T S_{s}\) arises from \(T S\) by declaring \(s\) as the unique initial state. Vice versa, \(T S_{1} \preceq T S_{2}\) if each initial state \(s_{1}\) of \(T S_{1}\) is simulated by an initial state of \(T S_{2}\) in the transition system \(T S_{1} \oplus T S_{2}\). Analogous observations hold for the simulation equivalence. These observations allow us to say that a state \(s_{1}\) in a transition system \(T S_{1}\) is simulated by a state \(s_{2}\) of another transition system \(T S_{2}\), provided that the pair \(\left(s_{1}, s_{2}\right)\) is contained in some simulation for \(\left(T S_{1}, T S_{2}\right)\).

In analogy to Lemma \(7.8, \preceq_{T S}\) is the coarsest simulation. Moreover, the simulation order on \(T S\) is a preorder (i.e., transitive and reflexive) on \(T S\) 's state space and is the union of all simulations on \(T S\).

\section*{Lemma 7.59. \(\preceq_{\text {TS }}\) is a Preorder and the Coarsest Simulation}

For transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) it holds that:
1. \(\preceq_{T S}\) is a preorder on \(S\).
2. \(\preceq_{T S}\) is a simulation on \(T S\).
3. \(\preceq_{\text {TS }}\) is the coarsest simulation for TS.

Proof: The first claim follows from Lemma 7.49 (page 453). To prove the second statement we show \(\preceq_{T S}\) fulfills conditions (1) and (2) of simulations on \(T S\) (Definition 7.58). Let \(s_{1} \preceq_{\text {TS }} s_{2}\) for \(s_{1}, s_{2}\) states in \(T S\). Then, there exists a simulation \(\mathcal{R}\) that contains ( \(s_{1}, s_{2}\) ). Since conditions (1) and (2) of Definition 7.58 hold for all pairs in \(\mathcal{R}, L\left(s_{1}\right)=L\left(s_{2}\right)\) and for any transition \(s_{1} \rightarrow s_{1}^{\prime}\), there exists a transition \(s_{2} \rightarrow s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\). Hence, \(s_{1}^{\prime} \preceq_{T S} s_{2}^{\prime}\). Thus, \(\preceq_{T S}\) is a simulation on TS.

The third claim follows immediately from the second statement ( \(\preceq_{T S}\) is a simulation) and the fact that \(s_{1} \preceq_{T S} s_{2}\) if there exists some simulation containing ( \(s_{1}, s_{2}\) ). Hence, each simulation \(\mathcal{R}\) is contained in \(\preceq_{T S}\).

Let us now define the quotient transition system under simulation equivalence for transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\). For simplicity, we omit the subscript \(T S\) and simply write \(\preceq\) rather than \(\preceq_{T S}\) and \(\simeq\) rather than \(\simeq_{T S}\).

\section*{Definition 7.60. Simulation Quotient System}

For transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\), the simulation quotient transition system \(T S / \simeq\) is defined as follows:
\[
T S / \simeq=\left(S / \simeq,\{\tau\}, \rightarrow_{\simeq}, I_{\simeq}, A P, L_{\simeq}\right)
\]
where \(I_{\simeq}=\left\{[s]_{\simeq} \mid s \in I\right\}, L_{\simeq}\left([s]_{\simeq}\right)=L(s)\), and
\[
\rightarrow_{\simeq} \text { is defined by: } \frac{s \xrightarrow{\alpha} s^{\prime}}{[s]_{\simeq}^{\tau} \simeq\left[s^{\prime}\right]_{\simeq}}
\]

Note that the transition relation in Definition 7.60 is similar to that of the bisimulation quotient of \(T S\), but considers equivalence classes under \(\simeq\) rather than \(\sim\). From the definition of \(\rightarrow \simeq\) it follows that
\[
B \rightarrow \simeq C \quad \text { if and only if } \quad \exists s \in B . \exists s^{\prime} \in C . s \rightarrow s^{\prime}
\]

In contrast to the bisimulation quotient, this is not equivalent to
\[
B \rightarrow \simeq C \quad \text { if and only if } \quad \forall s \in B . \exists s^{\prime} \in C . s \rightarrow s^{\prime}
\]

Stated in words, \(B \rightarrow \simeq C\) does not imply that \(\operatorname{Post}(s) \cap C \neq \varnothing\) for any \(s \in B\). Thus, in general, we have \(T S \nsim T S / \simeq\). However, simulations can be established for \(T S\) and \(T S / \simeq\) in both directions. That is to say, \(T S\) and \(T S / \simeq\) are simulation equivalent:

Theorem 7.61. Simulation equivalence of TS and TS/ \(\simeq\)
For any transition system \(T S\) it holds that \(T S \simeq T S / \simeq\).

Proof: \(T S \preceq T S / \simeq\) follows directly from the fact that the relation \(\mathcal{R}=\left\{\left(s,[s]_{\simeq}\right) \mid s \in S\right\}\) is a simulation for \((T S, T S / \simeq)\). We show that \(T S / \simeq \preceq T S\) by establishing a simulation
\(\mathcal{R}^{\prime}\) for \((T S / \simeq, T S)\). First, observe that \(\mathcal{R}^{\prime}=\mathcal{R}^{-1}=\left\{\left([s]_{\simeq}, s\right) \mid s \in S\right\}\) is not adequate, as it is possible that \(s^{\prime} \simeq s, s^{\prime} \rightarrow t^{\prime}\), but \(\left[t^{\prime}\right] \simeq \cap \operatorname{Post}(s)=\varnothing\). Instead,
\[
\mathcal{R}^{\prime}=\left\{\left([s]_{\simeq}, t\right) \mid s \preceq t\right\}
\]
is a simulation for \((T S / \simeq, T S)\). This is proven as follows. Each initial state \(\left[s_{0}\right] \simeq\) in \(T S / \simeq\) is simulated by an initial state in \(T S\), since \(\left[s_{0}\right] \simeq \preceq s_{0}\) and \(s_{0} \in I\). In addition, \([s]_{\simeq}\) and \(t\) are equally labeled as \(s \preceq t\) and all states in the simulator set \(\operatorname{Sim}_{T S}(s)=\left\{s^{\prime} \in S \mid s \preceq_{\text {TS }} s^{\prime}\right\}\) of \(s\) are equally labeled. It remains to check whether each transition of \([s] \simeq\) is mimicked by a transition of \(t\) for \(s \preceq t\). Let \(B \in S / \simeq,(B, t) \in \mathcal{R}^{\prime}\), and \(B \rightarrow \simeq C\) be a transition in \(T S / \simeq\). By definition of \(\rightarrow_{\simeq}\), there exists a state \(s \in B\) and a transition \(s \rightarrow s^{\prime}\) in \(T S\) such that \(C=\left[s^{\prime}\right] \simeq\). As \((B, t) \in \mathcal{R}^{\prime}, t\) simulates some state \(u \in B\), i.e., \(u \preceq t\) for some \(u \in B\). Since \(s, u \in B\), and all states in \(B\) are simulation-equivalent, we get \(u \preceq t\) and \(s \preceq u\) (and \(u \preceq s\) ). By transitivity of \(\preceq\), we have \(s \preceq t\). There is thus a transition \(t \rightarrow t^{\prime}\) in \(T S\) with \(s^{\prime} \preceq t^{\prime}\). By definition of \(\mathcal{R}^{\prime},\left(C, t^{\prime}\right)=\left(\left[s^{\prime}\right] \simeq, t^{\prime}\right) \in \mathcal{R}^{\prime}\).

\section*{Remark 7.62. Alternative Definition of the Simulation Quotient System}

Let us consider the following alternative definition of the simulation quotient system. Let \(T S\) be a transition system as before and
\[
T S / \simeq^{\prime}=\left(S / \simeq,\{\tau\}, \rightarrow_{\simeq}^{\prime}, I_{\simeq}, A P, L_{\simeq}\right)
\]
where \(I_{\simeq}\) and \(L_{\simeq}\) are as in Definition 7.60 and where the transitions in \(T S / \simeq^{\prime}\) arise by the rule
\[
\frac{B, B^{\prime} \in S / \simeq \wedge \forall s \in B \exists s^{\prime} \in B^{\prime} . s \rightarrow s^{\prime}}{[s]_{\simeq} \rightarrow_{\simeq}^{\prime}\left[s^{\prime}\right] \simeq}
\]
(where the action labels are omitted since they are not of importance here). Clearly, \(T S / \simeq^{\prime}\) is simulated by \(T S\) as \(\mathcal{R}=\left\{\left([s]_{\simeq}, s\right) \mid s \in S\right\}\) is a simulation for \(\left(T S / \simeq^{\prime}, T S\right)\). However, we cannot guarantee that the reverse holds, i.e., that \(T S\) is simulated by \(T S / \simeq^{\prime}\).

Let us illustrate this by means of an example. Assume that \(T S\) has two initial states \(s_{1}\) and \(s_{2}\) with \(\operatorname{Post}\left(s_{1}\right)=\{2 i+1 \mid i \geqslant 0\}\) and \(\operatorname{Post}\left(s_{2}\right)=\{2 i \mid i \geqslant 0\}\). Moreover, assume that
\[
1 \preceq 2 \preceq 3 \preceq \ldots
\]
while state \(n+1 \npreceq n\) for all \(n \in \mathbb{N}\). For instance, the state space of \(T S\) could be \(\left\{s_{1}, s_{2}\right\} \cup\) \(\mathbb{N} \cup\left\{t_{n} \mid n \in \mathbb{N}\right\}\) with \(s_{j} \rightarrow 2 i+j\) for all \(i \geqslant 0, j=1,2, n \rightarrow t_{n}\) and \(t_{n+1} \rightarrow t_{n}\) for all \(n \geqslant 0\), while state \(t_{0}\) is terminal. Moreover, we deal with \(A P=\{a, b\}\) and \(L\left(s_{1}\right)=L\left(s_{2}\right)=\{a\}\), \(L(n)=\{b\}\) and \(L\left(t_{n}\right)=\varnothing\) for all \(n \geqslant 0\). Then \(t_{i} \preceq_{T S} t_{j}\) if and only if \(i<j\) which yields
\[
i \preceq_{T S} j \text { if and only if } i<j
\]
and \(s_{1} \simeq_{T S} s_{2}\). In the quotient system \(T S / \simeq^{\prime}\), states \(s_{1}\) and \(s_{2}\) are collapsed into their simulation equivalence class \(B=\left\{s_{1}, s_{2}\right\}\). Since the states in \(\mathbb{N}\) are pairwise not simulationequivalent and since \(s_{1}\) and \(s_{2}\) do not have any common direct successor, \(B\) is a terminal state in \(T S / \simeq^{\prime}\). In particular, the reachable fragment of \(T S / \simeq^{\prime}\) just consists of the initial state \(B\) and does not have any transition. Thus, \(T S / \simeq^{\prime}\) and \(T S\) are not simulation equivalent.

In the above example, \(T S\) is infinite. If, however, we are given a finite transition system, then \(T S / \simeq^{\prime}\) simulates \(T S\), which yields the simulation equivalence of \(T S\) and \(T S / \simeq^{\prime}\). To see why it suffices to establish a simulation for \(\left(T S, T S / \simeq^{\prime}\right)\). Let
\[
\mathcal{R}^{\prime}=\{(s,[t] \simeq) \mid s \preceq t\}
\]
and show that \(\mathcal{R}^{\prime}\) is a simulation for \(\left(T S, T S / \simeq^{\prime}\right)\). Conditions (A) and (B.1) are obvious. Let us check condition (B.2). We use the following claim:

Claim. If \(B \in S / \simeq, t \in B\) and \(t \rightarrow t^{\prime}\), then there exists a transition \(t \rightarrow t_{\text {max }}^{\prime}\) such that \(t^{\prime} \preceq t_{\text {max }}^{\prime}\) and \(\left.B \rightarrow{ }^{\prime} \simeq t_{\text {max }}^{\prime}\right]_{\simeq}\).

Proof of the claim. Let \(t_{\text {max }}^{\prime}\) be such that \(t_{\text {max }}^{\prime}\) is "maximal" in the sense that for any transition \(t \rightarrow v\) where \(t_{\text {max }}^{\prime} \preceq v\) we have \(t_{\max }^{\prime} \simeq v\). (Such "maximal" elements exist since \(T S\) is finite.) For each state \(u \in B\), we have \(t \simeq u\). Since \(t \preceq u\), there exists a transition \(u \rightarrow u^{\prime}\) with \(t_{\text {max }}^{\prime} \preceq u^{\prime}\). And vice versa, since \(u \preceq t\), there exists a transition \(t \rightarrow v\) with \(u^{\prime} \preceq v\). But then \(t_{\max }^{\prime} \preceq v\) by the transitivity of \(\preceq\). The maximality of \(t_{\text {max }}^{\prime}\) yields \(t_{\text {max }}^{\prime} \simeq v\). Since \(t_{\text {max }}^{\prime} \preceq u^{\prime} \preceq v\) we get \(t_{\text {max }}^{\prime} \simeq u^{\prime}\). We conclude that all states \(u \in B\) have an outgoing transition leading to a state in \(\left[t_{\text {max }}^{\prime}\right] \simeq\). This yields \(B \rightarrow{ }_{\sim}^{\prime}\left[t_{\text {max }}^{\prime}\right] \simeq\).

We now use the above claim to show (B.2) for \(\mathcal{R}^{\prime}\) when \(T S\) is finite. Let us now suppose that \((s, B) \in \mathcal{R}^{\prime}\) and that \(s \rightarrow s^{\prime}\) is a transition in \(T S\). We take an arbitrary representative \(t \in B\). By definition of \(\mathcal{R}^{\prime}\), state \(t\) simulates \(s\). Thus, there exists a transition \(t \rightarrow t^{\prime}\) with \(s^{\prime} \preceq t^{\prime}\). The above claim yields the existence of a transition \(t \rightarrow t_{\text {max }}^{\prime}\) with \(t^{\prime} \preceq t_{\text {max }}^{\prime}\) and \(B \rightarrow{ }^{\prime} C\) where \(C=\left[t_{\text {max }}^{\prime}\right] \simeq\). But then \(s^{\prime} \preceq t_{\text {max }}^{\prime}\) and hence, \(\left(s^{\prime}, C\right) \in \mathcal{R}^{\prime}\).

\subsection*{7.4.2 Bisimulation, Simulation, and Trace Equivalence}

In this monograph, various equivalences and preorder relations have been defined on transition systems. This section compares bisimulation, simulation, and (infinite and finite) trace equivalence as well as the simulation preorder and trace inclusion. In the sequel of this section, let \(T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right), i=1,2\), be transition systems over \(A P\). We first observe that bisimulation equivalence implies simulation equivalence. This simply
follows by the symmetry of the conditions required for bisimulations. However, there are simulation equivalent transition systems which are not bisimulation equivalent. Note that bisimulation equivalence requires that whenever \(s_{1} \sim s_{2}\), then every transition \(s_{1} \rightarrow s_{1}^{\prime}\) can be mimicked by a transition \(s_{2} \rightarrow s_{2}^{\prime}\) such that \(s_{1}^{\prime} \sim s_{2}^{\prime}\). Simulation equivalence, however, requires that whenever \(s_{1} \simeq s_{2}\), then \(s_{1} \rightarrow s_{1}^{\prime}\) can be mimicked by \(s_{2} \rightarrow s_{2}^{\prime}\) such that \(s_{1}^{\prime} \preceq s_{2}^{\prime}\) (but not necessarily \(s_{1}^{\prime} \simeq s_{2}^{\prime}!\) ). The fact that \(T S_{1} \simeq T S_{2}\) does not always imply \(T S_{1} \sim T S_{2}\) is illustrated by the following example.

\section*{Example 7.63. Similar but not Bisimilar Transition Systems}

Consider the transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) in Figure 7.25. \(T S_{1} \nsim T S_{2}\), as there is no bisimilar state in \(T S_{2}\) that mimics state \(s_{2}\); the only candidate would be \(t_{2}\), but \(s_{2}\) cannot mimic \(t_{2} \rightarrow t_{4}\). \(T S_{1}\) and \(T S_{2}\) are, however, simulation-equivalent. As \(T S_{2}\) is a subgraph (up to isomorphism) of \(T S_{1}\), we obtain \(T S_{2} \preceq T S_{1}\). In addition, \(T S_{1} \preceq T S_{2}\) as
\[
\mathcal{R}=\left\{\left(s_{1}, t_{1}\right),\left(s_{2}, t_{2}\right),\left(s_{3}, t_{2}\right),\left(s_{4}, t_{3}\right),\left(s_{5}, t_{4}\right)\right\}
\]
is a simulation relation for \(\left(T S_{1}, T S_{2}\right)\).


Figure 7.25: Simulation-, but not bisimulation-equivalent transition systems.

Theorem 7.64. Bisimulation is Strictly Finer than Simulation Equivalence
\(T S_{1} \sim T S_{2}\) implies \(T S_{1} \simeq T S_{2}\), but \(T S_{1} \nsim T S_{2}\) and \(T S_{1} \simeq T S_{2}\) is possible.

Proof: Suppose \(T S_{1} \sim T S_{2}\). Then, there exists a bisimulation \(\mathcal{R}\) for \(\left(T S_{1}, T S_{2}\right)\). It follows directly that \(\mathcal{R}\) is a simulation for \(\left(T S_{1}, T S_{2}\right)\), and that \(\mathcal{R}^{-1}\) is a simulation for \(\left(T S_{2}, T S_{1}\right)\). Hence, \(T S_{1} \simeq T S_{2}\). Example 7.63 provides transition systems \(T S_{1}\) and \(T S_{2}\) such that \(T S_{1} \simeq T S_{2}\), but \(T S_{1} \nsucc T S_{2}\).

Recall the notion of AP-determinism (see Definition 2.5 on page 24 ).

Definition 7.65. AP-Deterministic Transition System
Transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) is AP-deterministic if
1. for \(A \subseteq A P:|I \cap\{s \mid L(s)=A\}| \leqslant 1\), and
2. for \(s \in S\) : if \(s \xrightarrow{\alpha} s^{\prime}\) and \(s \xrightarrow{\alpha} s^{\prime \prime}\) and \(L\left(s^{\prime}\right)=L\left(s^{\prime \prime}\right)\), then \(s^{\prime}=s^{\prime \prime}\).

The difference between \(\sim\) and \(\simeq\) crucially relies on \(A P\)-determinism, e.g., transition system \(T S_{1}\) in Figure 7.25 (left) is not \(A P\)-deterministic as its initial state has two distinct \(\varnothing\) successors.

Theorem 7.66. AP-Determinism Implies \(\sim\) and \(\simeq\) Coincide If \(T S_{1}\) and \(T S_{2}\) are AP-deterministic, then \(T S_{1} \sim T S_{2}\) if and only if \(T S_{1} \simeq T S_{2}\).

Proof: see Exercise 7.3.

This completes the comparison between bisimulation and simulation equivalence. To enable the comparison with trace equivalence, we first consider the relation between \(\preceq\) and (finite and infinite) trace inclusion.

\section*{Theorem 7.67. Simulation Order vs. Finite Trace Inclusion}
\(T S_{1} \preceq T S_{2}\) implies \(\operatorname{Traces}_{f i n}\left(T S_{1}\right) \subseteq \operatorname{Traces}_{f i n}\left(T S_{2}\right)\).

Proof: Assume \(T S_{1} \preceq T S_{2}\). Let \(\widehat{\pi_{1}}\) be \(s_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots s_{n, 1} \in \operatorname{Paths}_{f i n}\left(s_{1}\right)\) where \(s_{1} \in I_{1}\), i.e., \(s_{1}\) is an initial state in \(T S_{1}\). Since \(T S_{1} \preceq T S_{2}\), there exists \(s_{2} \in I_{2}\) such that \(s_{1} \preceq s_{2}\). As \(\preceq\) can be lifted to finite path fragments (see Lemma 7.55 on page 504 ), it follows that there exists \(\widehat{\pi_{2}} \in \operatorname{Paths}_{f i n}\left(s_{2}\right)\), say, of the form \(s_{2}=s_{0,2} s_{1,2} s_{2,2} \ldots s_{2, n}\) such that \(\widehat{\pi_{1}} \preceq \widehat{\pi_{2}}\), i.e., \(s_{j, 1} \preceq s_{j, 2}\) for \(0 \leqslant j \leqslant n\). But then \(L\left(s_{j, 1}\right)=L\left(s_{j, 2}\right)\) for \(0 \leqslant j \leqslant n\), that is, \(\operatorname{trace}\left(\pi_{1}\right)=\operatorname{trace}\left(\pi_{2}\right)\).

\section*{Corollary 7.68. Simulation Preserves Safety Properties}

Let \(P_{\text {safe }}\) be a safety \(L T\) property and \(T S_{1}\) and \(T S_{2}\) transition systems (all over AP) without terminal states. Then:
\[
T S_{1} \preceq T S_{2} \quad \text { and } \quad T S_{2} \models P_{\text {safe }} \quad \text { implies } \quad T S_{1} \models P_{\text {safe }} .
\]

Proof: Follows directly from Theorem 7.67 and the fact that \(\operatorname{Traces}_{f i n}\left(T S_{2}\right) \supseteq \operatorname{Traces}_{f i n}\left(T S_{1}\right)\) implies that if \(T S_{2} \models P_{\text {safe }}\), then \(T S_{1} \models P_{\text {safe }}\) (see Corollary 3.15 on page 104).

Theorem 7.67 relates simulation preorder and finite trace inclusion, and cannot be obtained for trace inclusion. That is, in general, \(T S_{1} \preceq T S_{2}\) does not imply that \(\operatorname{Traces}\left(T S_{1}\right) \subseteq\) \(\operatorname{Traces}\left(T S_{2}\right)\). This is illustrated by the following example.

\section*{Example 7.69. Simulation Preorder Does Not Imply Trace Inclusion}

Consider the transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) depicted in Figure 7.26. We have \(T S_{1} \preceq T S_{2}\), but \(\operatorname{Traces}\left(T S_{1}\right) \nsubseteq \operatorname{Traces}\left(T S_{2}\right)\) since \(\{a\} \varnothing \in \operatorname{Traces}\left(T S_{1}\right)\) but \(\{a\} \varnothing \notin\) \(\operatorname{Traces}\left(T S_{1}\right)\). This is due to the fact that \(s_{2} \preceq t_{2}\), but whereas \(s_{2}\) is a terminal state, \(t_{2}\) is


Figure 7.26: \(T S_{1} \preceq T S_{2}\), but \(\operatorname{Traces}\left(T S_{1}\right) \nsubseteq \operatorname{Traces}\left(T S_{2}\right)\).
not. (Note that terminal states are simulated by any equally labeled state, be it terminal or not). As a result, a trace in \(T S_{1}\) may end in \(s_{2}\), but a trace in \(T S_{2}\) cannot end in \(t_{2}\).

The example indicates that terminal states are an essential factor of not preserving trace inclusion. Indeed, if terminal states are absent in \(T S_{1}\) and \(T S_{1} \preceq T S_{2}\) then all paths in \(T S_{1}\) are infinite and the path fragment lifting for simulations yields that each path of \(T S_{1}\) can be lifted to an infinite path in \(T S_{2}\). This yields:

\section*{Theorem 7.70. Simulation Order and Trace Inclusion}

If \(T S_{1} \preceq T S_{2}\) and \(T S_{1}\) does not have terminal states then \(\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right)\).

Thus, for transition systems without terminal states, the result stated in Corollary 7.68 holds for all LT properties and not just the safety properties.

For the class of AP-deterministic transition systems one obtains a similar result. Here, however, we have to deal with simulation equivalence and trace equivalence (rather than the simulation preorder and trace inclusion).

\section*{Theorem 7.71. Simulation in AP-Deterministic Systems}

If \(T S_{1} \simeq T S_{2}\) and \(T S_{1}\) and \(T S_{2}\) are AP-deterministic then \(\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)\).

Proof: See Exercise 7.3.

\section*{Corollary 7.72. Simulation and (Finite) Trace Equivalence}

For transition systems \(T S_{1}\) and \(T S_{2}\) over AP:
(a) If \(T S_{1} \simeq T S_{2}\), then \(\operatorname{Traces}_{f i n}\left(T S_{1}\right)=\operatorname{Traces}_{f i n}\left(T S_{2}\right)\).
(b) If \(T S_{1}\) and \(T S_{2}\) do not have terminal states and \(T S_{1} \simeq T S_{2}\), then \(\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)\).
(c) If \(T S_{1}\) and \(T S_{2}\) are AP-deterministic, then \(T S_{1} \simeq T S_{2}\) if and only if \(\operatorname{Traces}\left(T S_{1}\right)=\) Traces \(\left(T S_{2}\right)\).

Proof: Claim (a) follows from Theorem 7.67, claim (b) from Theorem 7.70. Claim (c) follows from (and only if) Theorem 7.71 and (if) Exercise 7.3.

Simulation equivalent transition systems without terminal states satisfy the same LT properties, and hence, the same LTL formulae. Claim (c) together with Theorem 7.66 yield that for \(A P\)-deterministic transition systems, \(\sim, \simeq\), and trace equivalence coincide. (Finite) Trace equivalence does not imply simulation equivalence. This is illustrated by the two beverage vending machines described in Example 7.48 on page 498.

\section*{Remark 7.73. Bisimulation and Trace Equivalence}

Since \(\sim\) is finer than \(\simeq\), it follows from Theorem 7.67 (page 512) that \(\sim\) is finer than finite trace equivalence:
\[
T S_{1} \sim T S_{2} \text { implies } \operatorname{Traces}_{f i n}\left(T S_{1}\right)=\operatorname{Traces}_{f i n}\left(T S_{2}\right)
\]

However, bisimulation equivalent transition systems are even trace eqivalent. That is, \(T S_{1} \sim T S_{2}\) implies \(\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)\). This applies to any transition system


Figure 7.27: Relation between equivalences and preorders on transition systems.
\(T S_{1}\), i.e., the absence of terminal states (as in Theorem 7.70 on page 513) is not required. This stems from the fact that a terminal state cannot be bisimilar to a nonterminal state, while terminal states are simulated by any state with the same label.

Figure 7.27 summarizes the relationship between bisimulation, simulation, and (finite and infinite) trace equivalence as well as simulation order and trace inclusion in the form of a Hasse diagram. The vertices of the diagram are relations, and an edge from \(\mathcal{R}\) to \(\mathcal{R}^{\prime}\) means that \(\mathcal{R}\) is strictly finer than \(\mathcal{R}^{\prime}\), that is to say, \(\mathcal{R}\) is more distinctive than \(\mathcal{R}^{\prime}\). Recall that for \(A P\)-deterministic transition systems, \(\sim, \simeq\) and trace equivalence coincide.

\subsection*{7.5 Simulation and \(\forall\) CTL \(^{*}\) Equivalence}

The aim of this section is to provide a logical characterization of the simulation order \(\preceq\). This amounts establishing an analogous result to the theorem stating that bisimulation equivalence coincides with CTL* (and CTL) equivalence for finite transition systems without terminal states. That is to say, given two transition systems \(T S_{1}\) and \(T S_{2}\), the aim is to characterize the relation \(\preceq\) in terms of the satisfaction relations on \(T S_{1}\) and \(T S_{2}\) for some temporal logic. In fact, from the results of the previous section, it can be inferred that for transition systems without terminal states, \(T S_{1} \preceq T S_{2}\) implies trace inclusion of \(T S_{1}\) and \(T S_{2}\), and thus whenever \(T S_{2} \models \varphi\) for a LTL formula \(\varphi\), it follows \(T S_{1} \models \varphi\). In this section, it is shown that this not only applies to any LTL formula, but for a fragment
of CTL* that includes LTL.

Since \(\preceq\) is not symmetric, it cannot be expected that \(\preceq\) can be characterized by a sublogic of CTL* which allows negation of arbitrary state formulae. This can be seen as follows. Let \(\mathbf{L}\) be a fragment of CTL* which is closed under negation, i.e., \(\Phi \in \mathbf{L}\) implies \(\neg \Phi \in \mathbf{L}\), such that for any transition system \(T S\) without terminal states and states \(s_{1}, s_{2}\) of \(T S\) :
\[
s_{1} \preceq_{T S} s_{2} \quad \text { iff } \quad \text { for all state formulae } \Phi \text { of } \mathbf{L}: s_{2} \models \Phi \Longrightarrow s_{1} \models \Phi \text {. }
\]

Let \(s_{1} \preceq_{T S} s_{2}\). Note that \(s_{2} \models \neg \Phi\) implies \(s_{1} \models \neg \Phi\), i.e., \(s_{1} \not \models \neg \Phi\) implies \(s_{2} \not \models \neg \Phi\). Then, for any state formula \(\Phi\) of \(\mathbf{L}\) :
\[
s_{1} \models \Phi \quad \Longrightarrow s_{1} \not \models \neg \Phi \quad \Longrightarrow \quad s_{2} \not \models \neg \Phi \quad \Longrightarrow \quad s_{2} \vDash \Phi
\]

Hence, \(s_{2} \preceq_{T S} s_{1}\). Thus, a logic that characterizes \(\preceq_{T S}\) cannot be closed under negation, as that would require \(\preceq_{T S}\) to be symmetric which, however, is not the case.

The proof of the path-lifting lemma (Lemma 7.55 on page 504) informally explains why for every LTL formula \(\varphi\) such that \(s_{2} \models \varphi\) we have \(s_{1} \models \varphi\). This observation can be generalized for CTL* formulae of the form \(\forall \varphi\), provided that the state formulae in \(\varphi\) contain neither negation nor existential quantification. These considerations motivate the definition of the universal fragment of CTL*, denoted \(\forall \mathrm{CTL}^{*}\).

\section*{Definition 7.74. Universal Fragment of CTL*}

The universal fragment of CTL*, denoted \(\forall \mathrm{CTL}^{*}\), consists of the state formulae \(\Phi\) and path formulae \(\varphi\) given, for \(a \in A P\), by
\[
\begin{aligned}
& \Phi::=\text { true } \mid \text { false }|a| \neg a\left|\Phi_{1} \wedge \Phi_{2}\right| \Phi_{1} \vee \Phi_{2} \mid \forall \varphi \\
& \varphi::=\Phi|\bigcirc \varphi| \varphi_{1} \wedge \varphi_{2}\left|\varphi_{1} \vee \varphi_{2}\right| \varphi_{1} \cup \varphi_{2} \mid \varphi_{1} \mathrm{R} \varphi_{2}
\end{aligned}
\]

State formulae in \(\forall C T L^{*}\) are required to be in positive normal form, (negations may only occur adjacent to atomic propositions) and do not contain existential path quantifiers. Due to the positive normal form (PNF, for short), the release operator R is considered as a basic operator in the logic. The universal fragment of CTL, denoted \(\forall\) CTL, is obtained from the definition of \(\forall \mathrm{CTL}^{*}\) by restricting the path formulae to
\[
\varphi::=\bigcirc \Phi\left|\Phi_{1} \cup \Phi_{2}\right| \Phi_{1} R \Phi_{2}
\]

The eventually and always operators are obtained by the \(\forall \mathrm{CTL}^{*}\) path formulae \(\diamond \varphi=\) true \(\mathrm{U} \varphi\) and \(\square \varphi=\) false \(\mathrm{R} \varphi\). Similarly, in \(\forall \mathrm{CTL}\) these modalities are obtained by \(\forall \diamond \Phi=\) \(\forall(\) true \(U \Phi)\) and \(\forall \square \Phi=\forall(\) false \(R \Phi)\).
\(\forall\) CTL* \(^{*}\) covers all safety properties that can be expressed in CTL*. The following result shows that for any LTL formula there exists an equivalent \(\forall\) CTL* formula.

\section*{Lemma 7.75. \(\forall\) CTL \(^{*}\) Includes LTL}

For every LTL formula \(\varphi\) there exists an equivalent \(\forall C T L^{*}\) formula.

Proof: It follows from Theorem 5.24 (page 257) that for any LTL formula \(\varphi\) there exists an equivalent LTL formula in PNF. From the definition of \(\forall\) CTL* and positive normal form LTL, it follows directly that any LTL formula in PNF can be regarded as a \(\forall \mathrm{CTL}^{*}\) formula.

Vice versa, the universal fragment of CTL* is more expressive than LTL, as, e.g., \(\forall \Delta \forall \square a\) is a \(\forall \mathrm{CTL}^{*}\) formula that has no equivalent LTL formula. The following theorem indicates a temporal logic characterization of the simulation order by \(\forall \mathrm{CTL}^{*}\) and \(\forall \mathrm{CTL}\).

\section*{Theorem 7.76. Simulation Order and \(\forall C T L^{*} / \forall C T L\)}

Let TS be a finite transition system without terminal states and \(s_{1}, s_{2}\) be states of TS. The following statements are equivalent:
(a) \(s_{1} \preceq_{T S} s_{2}\).
(b) For all \(\forall C T L^{*}\) formulae \(\Phi\) : \(s_{2} \models \Phi\) implies \(s_{1} \models \Phi\).
(c) For all \(\forall C T L\) formulae \(\Phi\) : \(s_{2} \models \Phi\) implies \(s_{1} \models \Phi\).

Proof: The proof technique is analogous to Theorem 7.20 (page 469).
\((\mathrm{a}) \Longrightarrow(\mathrm{b})\) follows from the following claims that hold for arbitrary (i.e., not necessarily finite) transition system \(T S\) with state space \(S\) :
(i) if \(s_{1} \preceq_{T S} s_{2}\), then for all \(\forall\) CTL \(^{*}\) state formulae \(\Phi\) : \(s_{2} \models \Phi\) implies \(s_{1} \models \Phi\).
(ii) if \(\pi_{1} \preceq_{T S} \pi_{2}\), then for all \(\forall\) CTL \(^{*}\) path formulae \(\varphi: \pi_{2} \models \varphi\) implies \(\pi_{1} \models \varphi\).

The proof of these claims is by structural induction, similar to Lemma 7.26 (page 473), and is omitted here. See also Exercise 7.14. For the treatment of formulas \(\forall \varphi\) in the step induction for (i), the assumption that the given transition system does not have terminal states is crucial, since it permits path lifting rather than just path fragment lifting.
\((\mathrm{b}) \Longrightarrow(\mathrm{c})\) is obvious since \(\forall \mathrm{CTL}\) is a sublogic of \(\forall \mathrm{CTL}^{*}\).
\((\mathrm{c}) \Longrightarrow(\mathrm{a})\). Let \(S\) be the state space of the finite transition system \(T S\). It suffices to show that
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid \forall \Phi \in \forall \text { CTL }^{*} . s_{2} \models \Phi \Rightarrow s_{1} \models \Phi\right\}
\]
is a simulation for \(T S\). This is proven by checking the conditions of a simulation relation. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).
1. If \(A P\) is finite, then we can argue by means of the formula
\[
\Phi=\bigwedge_{a \in L\left(s_{2}\right)} a \wedge \bigwedge_{a \in A P \backslash L\left(s_{2}\right)} \neg a .
\]

Note that \(\Phi\) is a propositional formula in PNF, and hence a \(\forall C T L\) formula. Since \(s_{2} \models \Phi\) and \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\), we obtain \(s_{1} \models \Phi\). As \(\Phi\) uniquely characterizes the labeling of \(s_{1}\), we have \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
If \(A P\) is infinite we need a slightly different argument. Let \(a \in A P\). If \(a \in L\left(s_{2}\right)\), then \(s_{2} \models a\). Since the atomic proposition \(a\) is a \(\forall\) CTL formula, we get \(s_{1} \models a\) (by definition of \(\mathcal{R}\) ), and therefore \(a \in L\left(s_{1}\right)\). Similarly, if \(a \notin L\left(s_{2}\right)\), then \(s_{2} \models \neg a\). Again, as \(\neg a\) is a \(\forall\) CTL formula and by definition of \(\mathcal{R}\), we get \(s_{1} \models \neg a\), and hence \(a \notin L\left(s_{1}\right)\). This yields \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
2. The idea for establishing the condition (2) of simulation relations is to define a \(\forall \mathrm{CTL}\) master formula \(\Phi_{u}\) for the downward closure of state \(u\) with respect to \(\mathcal{R}\), i.e.,
\[
\operatorname{Sat}\left(\Phi_{u}\right)=u \downarrow=\{t \in S \mid(t, u) \in \mathcal{R}\}
\]

The definition of \(\Phi_{u}\) is as follows. If \(u, t\) are states in \(T S\) with \(t \notin u \downarrow\) then, according to the definition of \(u \downarrow\) and \(\mathcal{R}\), there exists a \(\forall\) CTL formula \(\Phi_{u, t}\) such that \(u \models \Phi_{u, t}\) and \(t \not \vDash \Phi_{u, t}\). Let \(\Phi_{u}\) be the conjunction of all formulae \(\Phi_{u, t}\) where \(t \notin u \downarrow\) :
\[
\Phi_{u}=\bigwedge_{\substack{t \in S \\(t, u) \notin \mathcal{R}}} \Phi_{u, t}
\]

As TS is finite, \(\Phi_{u}\) arises by a finite conjunction, and thus, it is a \(\forall\) CTL formula. In fact, \(\Phi_{u}\) is a master formula for \(u \downarrow\) :
\[
\operatorname{Sat}\left(\Phi_{u}\right)=u \downarrow
\]

Let us check this. If \(v \notin u \downarrow\) then \((v, u) \notin \mathcal{R}\) and \(\Phi_{u}\) has the form \(\ldots \wedge \Phi_{u, v} \wedge \ldots\).. Hence, \(v \not \vDash \Phi_{u}\) (as \(v \not \vDash \Phi_{u, v}\) ), and therefore \(v \notin \operatorname{Sat}\left(\Phi_{u}\right)\). Vice versa, if \(v \in u \downarrow\) then \((v, u) \in \mathcal{R}\). As \(u \models \Phi_{u}\) we get \(v \models \Phi_{u}\) (by definition of \(\mathcal{R}\) ). Hence, \(v \in \operatorname{Sat}\left(\Phi_{u}\right)\).
It remains to prove that whenever \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1} \rightarrow s_{1}^{\prime}\), we have that \(s_{2} \rightarrow s_{2}^{\prime}\) such that \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\). Since \(T S\) is finite, the set of direct successors of \(s_{2}\) is finite, say \(\operatorname{Post}\left(s_{2}\right)=\left\{u_{1}, \ldots, u_{k}\right\}\). Let \(\Phi_{i}=\Phi_{u_{i}}\) be the master formula for \(u_{i} \downarrow\), for \(0<i \leqslant k\). From \(u_{i} \in u_{i} \downarrow\), it follows \(u_{i} \models \Phi_{i}\). Hence:
\[
s_{2} \models \forall \bigcirc \bigvee_{0<i \leqslant k} \Phi_{i}
\]

Since \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(\forall \bigcirc \underset{0<i \leqslant k}{\bigvee} \Phi_{i}\) is a \(\forall\) CTL formula, we have:
\[
s_{1} \models \forall \bigcirc \bigvee_{1 \leqslant i \leqslant k} \Phi_{i}
\]

Since \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right), s_{1}^{\prime} \in \operatorname{Sat}\left(\Phi_{i}\right)\) for some \(i \in\{1, \ldots, k\}\). Hence, \(s_{1}^{\prime} \in u_{i} \downarrow\), i.e., \(\left(s_{1}^{\prime}, u_{i}\right) \in \mathcal{R}\). Taking \(s_{2}^{\prime}=u_{i}\), we obtain \(s_{2} \rightarrow s_{2}^{\prime}\) and \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).

Theorem 7.76 can be reformulated for the simulation order between transition systems. This yields
\[
T S_{1} \preceq T S_{2} \quad \text { if and only if } \quad T S_{2} \models \Phi \Rightarrow T S_{1} \models \Phi
\]
for any \(\forall \mathrm{CTL}^{*}\) (or \(\forall \mathrm{CTL}\) ) formula \(\Phi\). As the proof of Theorem 7.76 does not exploit the until operator, this result applies to the fragment of CTL* that consists of literals (i.e., atomic propositions and their negations), conjunction, disjunction, and the modality \(\forall \bigcirc\).

\section*{Example 7.77. Distinguishing Nonsimilar Transition Systems}

Consider the transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) in Figure 7.28. \(T S_{1}\) is simulated by \(T S_{2}\), but due to the self-loop in the initial state, \(T S_{2} \npreceq T S_{1}\). This can also be established by providing a \(\forall\) CTL formula \(\Phi\) such that, e.g., \(T S_{1} \models \Phi\) whereas \(T S_{2} \not \vDash \Phi\). Such a \(\forall\) CTL formula is
\[
\Phi=\forall \bigcirc(\forall \bigcirc \neg a \vee \forall \bigcirc a)
\]

An alternative logical characterization of \(\preceq\) is obtained by replacing the universal quantifier \(\forall\) by the existential quantifier \(\exists\) using the CTL* duality rule: \(\forall \varphi \equiv \neg \exists \neg \varphi\). Theorem 7.76 yields:
\[
s_{1} \preceq_{T S} s_{2} \text { iff for all formulae } \exists \varphi: s_{1} \models \exists \varphi \text { implies } s_{2} \models \exists \varphi .
\]


Figure 7.28: \(T S_{1} \preceq T S_{2}\), but \(T S_{2} \npreceq T S_{1}\).

Here, \(\varphi\) is an arbitrary CTL* path formula that does not contain universal path quantifiers, and is in PNF. Such formulae are \(\exists\) CTL* formulae.

\section*{Definition 7.78. Existential Fragment of CTL*}

The existential fragment of CTL*, denoted \(\exists \mathrm{CTL}^{*}\), consists of the state formulae \(\Phi\) and path formulae \(\varphi\) given, for \(a \in A P\), by
\[
\begin{aligned}
& \Phi::=\text { true } \mid \text { false }|a| \neg a\left|\Phi_{1} \wedge \Phi_{2}\right| \Phi_{1} \vee \Phi_{2} \mid \exists \varphi \\
& \varphi::=\Phi|\bigcirc \varphi| \varphi_{1} \wedge \varphi_{2}\left|\varphi_{1} \vee \varphi_{2}\right| \varphi_{1} \cup \varphi_{2} \mid \varphi_{1} \mathrm{R} \varphi_{2}
\end{aligned}
\]

The existential fragment of CTL, denoted \(\exists \mathrm{CTL}\), is obtained by restricting the path formulae in the above definition to
\[
\varphi::=\bigcirc \Phi\left|\Phi_{1} \cup \Phi_{2}\right| \Phi_{1} \mathrm{R} \Phi_{2}
\]

The modalities eventually and always can be derived as for \(\forall \mathrm{CTL}^{*}\) and \(\forall \mathrm{CTL}\).

\section*{Theorem 7.79. Simulation Order and \(\exists C T L^{*} / \exists C T L\)}

Let TS be a finite transition system without terminal states, and \(s_{1}, s_{2}\) states of TS. Then, the following statements are equivalent:
(a) \(s_{1} \preceq_{T S} s_{2}\).
(b) For all \(\exists C T L^{*}\) formulae \(\Phi\) : \(s_{1} \models \Phi\) implies \(s_{2} \models \Phi\).
(c) For all \(\exists C T L\) formulae \(\Phi\) : \(s_{1} \models \Phi\) implies \(s_{2} \models \Phi\).

Proof: Directly from Theorem 7.76, as statements (b) and (c) here are the duals to statements (b) and (c) in Theorem 7.76.

\section*{Corollary 7.80. Characterizations of Simulation Equivalence}

Let TS be a finite transition system without terminal states and \(s_{1}, s_{2}\) be states in TS. Then the following five statements are equivalent:
(a) \(s_{1} \simeq_{T S} s_{2}\).
(b) \(s_{1}\) and \(s_{2}\) satisfy the same \(\forall C T L^{*}\) formulae.
(c) \(s_{1}\) and \(s_{2}\) satisfy the same \(\forall C T L\) formulae.
(d) \(s_{1}\) and \(s_{2}\) satisfy the same \(\exists C T L^{*}\) formulae.
(e) \(s_{1}\) and \(s_{2}\) satisfy the same \(\exists C T L\) formulae.

Proof: From Theorems 7.76 and 7.79.

In fact, this result can be strengthened as the listed equivalences coincide with the logical equivalence on the sublogic of CTL that just contains atomic propositions and their negations, conjunction, disjunction, and the next-step operator \(\bigcirc\), with either a universal or existential path quantifier.

\subsection*{7.6 Simulation-Quotienting Algorithms}

This section presents algorithms for obtaining the simulation quotient \(T S / \simeq\) for finite transition system \(T S\). Such algorithms serve two purposes. First, they can be used to verify whether \(T S_{1} \simeq T S_{2}\) by considering the simulation quotient of \(T S_{1} \oplus T S_{2}\) (see page 457). Secondly, such algorithms can be used to obtain the abstract (and thus smaller) transition system \(T S / \simeq\) in a fully automated manner. As \(T S \simeq T S / \simeq\), it follows that any verification result for \(T S / \simeq\), either being negative or positive, carries over to \(T S\). This applies to any formula expressed in either the universal or existential fragment of CTL or CTL*. Since simulation equivalence is coarser than bisimulation equivalence, the quotient \(T S / \simeq\) is at most as large as-but maybe significantly smaller than- \(T S / \sim\). The bisimulation quotient of a given infinite state transition system might be infinite whereas its simulation quotient is finite.

The goal of this section is to present an algorithm which takes as input a finite transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\), possibly with terminal states, and computes the simulation order \(\preceq_{T S}\). Clearly, this algorithm yields at the same time an automatic approach to check whether one finite transition system simulates another one and a sound, but incomplete technique for proving finite trace inclusion. The basic scheme for computing the simulation order is sketched in Algorithm 33.
```

Algorithm 33 Computation of the simulation order (basic idea)

```

Input: finite transition system \(T S\) over \(A P\) with state space \(S\)
Output: simulation order \(\preceq_{T S}\)
```

$\mathcal{R}:=\left\{\left(s_{1}, s_{2}\right) \mid L\left(s_{1}\right)=L\left(s_{2}\right)\right\} ;$

```
while \(\mathcal{R}\) is not a simulation do
    choose \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) such that \(s_{1} \rightarrow s_{1}^{\prime}\), but there is no \(s_{2}^{\prime}\) such that \(s_{2} \rightarrow s_{2}^{\prime}\) and
    \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\);
    \(\mathcal{R}:=\mathcal{R} \backslash\left\{\left(s_{1}, s_{2}\right)\right\}\)
od
return \(\mathcal{R}\)

The number of iterations is bounded above by \(|S|^{2}\), since
\[
S \times S \supseteq \mathcal{R}_{0} \supsetneqq \mathcal{R}_{1} \supsetneqq \mathcal{R}_{2} \supsetneqq \ldots \supsetneqq \mathcal{R}_{n}=\preceq
\]
where \(\mathcal{R}_{i}\) denotes the relation \(\mathcal{R}\) at the start of the \((i+1)\) st iteration.
In the following, we discuss some details that permit an efficient implementation of this algorithm. Instead of explicitly representing \(\mathcal{R}\), we use
\[
\operatorname{Sim}_{\mathcal{R}}\left(s_{1}\right)=\left\{s_{2} \in S \mid\left(s_{1}, s_{2}\right) \in \mathcal{R}\right\}
\]

Note that \(\operatorname{Sim}_{\mathcal{R}}\left(s_{1}\right)\) is a superset of \(\operatorname{Sim}_{T S}\left(s_{1}\right)\). This yields Algorithm 34 .
A straightforward implementation of Algorithm 34 yields a time complexity \(\mathcal{O}\left(M \cdot|S|^{3}\right)\), where \(M\), the number of edges in the state graph \(G(T S)\), is assumed to be at least \(|S|\). Let us briefly sketch how this time complexity is established. In every iteration, it is checked for each transition \(s_{1} \rightarrow s_{1}^{\prime}\), whether there exists a state \(s_{2} \in \operatorname{Sim}\left(s_{1}\right)\) that cannot simulate this transition. Let counter \(\delta\left(s_{1}^{\prime}, s_{2}\right)\) denote the number of successors of \(s_{2}\) that belong to the current simulator set of \(s_{1}^{\prime}\), i.e.,
\[
\delta\left(s_{1}^{\prime}, s_{2}\right)=\left|\operatorname{Post}\left(s_{2}\right) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right)\right| .
\]

The initialization of these counters can be done in \(\mathcal{O}(M \cdot|S|)\) by setting \(\delta\left(s_{1}^{\prime}, s_{2}\right)=\left|\operatorname{Post}\left(s_{2}\right)\right|\) for each \(s_{2} \in \operatorname{Sim}\left(s_{1}\right)\) and \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\). Checking \(\operatorname{Post}\left(s_{2}\right) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right)=\varnothing\) reduces to check whether \(\delta\left(s_{1}^{\prime}, s_{2}\right)=0\). Given a matrix representation of \(\delta\), this takes \(\mathcal{O}(1)\). Consider now the operations during an iteration. On removing state \(s_{2}\) from \(\operatorname{Sim}\left(s_{1}\right)\), we set
```

Algorithm 34 Computation of the simulation order (first refinement)
Input: finite transition system $T S$ over $A P$ with state space $S$
Output: simulation order $\preceq_{\text {TS }}$

```
    for all \(s_{1} \in S\) do
        \(\operatorname{Sim}\left(s_{1}\right):=\left\{s_{2} \in S \mid L\left(s_{1}\right)=L\left(s_{2}\right)\right\} ;\)
        (* initialization *)
    od
    while \(\exists\left(s_{1}, s_{2}\right) \in S \times \operatorname{Sim}\left(s_{1}\right) . \exists s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\operatorname{Post}\left(s_{2}\right) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right)=\varnothing\) do
        choose such a pair of states \(\left(s_{1}, s_{2}\right)\);
        ( \(\left.{ }^{*} s_{1} \not \nwarrow_{\text {TS }} s_{2}{ }^{*}\right)\)
        \(\operatorname{Sim}\left(s_{1}\right):=\operatorname{Sim}\left(s_{1}\right) \backslash\left\{s_{2}\right\} ;\)
    od
                                    \(\left({ }^{*} \operatorname{Sim}(s)=\operatorname{Sim}_{\text {TS }}(s)\right.\) for any \(\left.s^{*}\right)\)
return \(\left\{\left(s_{1}, s_{2}\right) \mid s_{2} \in \operatorname{Sim}\left(s_{1}\right)\right\}\)
\[
\delta\left(s_{1}, v_{2}\right):=\delta\left(s_{1}, v_{2}\right)-1 \text { for all } v_{2} \in \operatorname{Pre}\left(s_{2}\right)
\]
(The number of direct successors of \(v_{2}\) that can simulate \(s_{1}\) is reduced by one, viz. states \(s_{2}\).) Assuming a list presentation of the sets \(\operatorname{Pre}(\cdot)\) and bit vector representations of the simulator sets \(\operatorname{Sim}(\cdot)\), the total time complexity of the body of the while-loop (when ranging over all iterations) is in \(\mathcal{O}(M \cdot|S|)\). Note that each state \(s_{2}\) is removed from \(\operatorname{Sim}\left(s_{1}\right)\) at most once. Hence, the total number of steps performed in the body of the while-loop is bounded by
\[
\mathcal{O}(\sum_{s_{1} \in S} \underbrace{\sum_{s_{2} \in S}\left|\operatorname{Pre}\left(s_{2}\right)\right|}_{=M})=\mathcal{O}(M \cdot|S|) .
\]

In order to check whether \(\mathcal{R}\) is a simulation, we may inspect all transitions \(s_{1} \rightarrow s_{1}^{\prime}\) and all states \(s_{2}\) and check whether the counter \(\delta\left(s_{1}^{\prime}, s_{2}\right)\) is 0 . If so and \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\), then \(\mathcal{R}\) is not a simulation. Otherwise we have found a pair \(\left(s_{1}, s_{2}\right)\) to be removed from \(\mathcal{R}\) in the body of the while-loop. As \(|S|^{2}\) is an upper bound for the number of iterations of the while-loop and the costs required to check the condition of the while-loop are in \(\mathcal{O}(M \cdot|S|)\), the total time complexity of Algorithm 34 is \(\mathcal{O}\left(M \cdot|S|^{3}\right)\).

With a simple trick, the time complexity can be reduced to \(\mathcal{O}\left(M \cdot|S|^{2}\right)\). The idea is to organize all pairs \(\left(s_{1}^{\prime}, s_{2}\right)\) where \(\delta\left(s_{1}^{\prime}, s_{2}\right)=0\) in a list and to pick one such pair \(\left(s_{1}^{\prime}, s_{2}\right)\) per iteration, rather than seeking for a pair \(\left(s_{1}, s_{2}\right)\) that violates the simulation condition. In each iteration, we run through the predecessor list of \(s_{1}^{\prime}\) and check for each state \(s_{1} \in \operatorname{Pre}\left(s_{1}^{\prime}\right)\) whether \(s_{2} \in \operatorname{Sim}\left(s_{1}\right)\). If so, then we remove \(s_{2}\) from \(\operatorname{Sim}\left(s_{1}\right)\) and decrement the counters \(\delta\left(s_{1}, v_{2}\right)\) for all states \(v_{2} \in \operatorname{Pre}\left(s_{2}\right)\). In case \(\delta\left(s_{1}, v_{2}\right)\) becomes 0 , the pair \(\left(s_{1}, v_{2}\right)\) is inserted in the list organizing all pairs where \(\delta(\cdot)\) is 0 .

An Efficiency Improvement We now discuss a further refinement of Algorithm 34 that leads to the worst-case time complexity \(\mathcal{O}(M \cdot|S|)\). The crux of this more efficient realization is the following observation. Suppose \(s_{1} \rightarrow s_{1}^{\prime}\) and \(s_{2} \rightarrow s_{2}^{\prime}\), and we are about to remove \(s_{2}^{\prime}\) from \(\operatorname{Sim}\left(s_{1}^{\prime}\right)\). If state \(s_{2}^{\prime}\) is the only direct successor of \(s_{2}\) that belongs to the current simulator set of \(s_{1}^{\prime}\), i.e,
\[
s_{1} \in \operatorname{Pre}\left(s_{1}^{\prime}\right), s_{2} \in \operatorname{Sim}\left(s_{1}\right) \text { and } \operatorname{Sim}\left(s_{1}^{\prime}\right) \cap \operatorname{Post}\left(s_{2}\right)=\left\{s_{2}^{\prime}\right\},
\]
then there does not exist a transition \(s_{2} \rightarrow s_{2}^{\prime}\) which can simulate \(s_{1} \rightarrow s_{1}^{\prime}\). Hence, \(s_{2}\)-in fact, any direct predecessor of \(s_{2}^{\prime}\) for which this holds - can be safely removed from \(\operatorname{Sim}\left(s_{1}\right)\).

This observation can be generalized to sets of states. Let \(\operatorname{Sim}_{\text {old }}\left(s_{1}\right) \supseteq \operatorname{Sim}\left(s_{1}\right)\) denote the simulator set that preceded the last removal of states from \(\operatorname{Sim}\left(s_{1}\right)\); initially \(\operatorname{Sim}_{\text {old }}\left(s_{1}\right)=\) \(S\). On the removal of \(s_{2}^{\prime}\) from \(\operatorname{Sim}\left(s_{1}^{\prime}\right)\), we consider all direct predecessors of \(s_{1}^{\prime}\), and remove all states in
\[
\operatorname{Remove}\left(s_{1}^{\prime}\right)=\operatorname{Pre}\left(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right)\right) \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)
\]
from \(\operatorname{Sim}\left(s_{1}\right)\). This is justified as all the states in this set do not have a successor simulating \(s_{1}^{\prime}\), and thus these states cannot simulate any of the predecessors of \(s_{1}^{\prime}\). (Note that \(\operatorname{Post}\left(s_{2}\right) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right) \neq \varnothing\) if and only if \(s_{2} \in \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\).) This yields Algorithm 35, which also takes into account that a terminal state cannot be in the simulator set of a nonterminal state. Thus, when \(s_{1}^{\prime}\) is considered the first time, then Remove \(\left(s_{1}^{\prime}\right)\) consists of \(\operatorname{Pre}(S) \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\) and all terminal states (that is, we define \(\left.\operatorname{Remove}\left(s_{1}^{\prime}\right)=S \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\right)\).

The next observation is that there is no need to explicitly represent the sets \(\operatorname{Sim} \operatorname{Sold}(\cdot)\). The sets Remove (•) are dynamically adapted on modifying \(\operatorname{Sim}\left(s_{1}\right)\). The termination condition of the iteration can be replaced by \(\operatorname{Remove}\left(s_{1}^{\prime}\right)=\varnothing\) for all \(s_{1}^{\prime} \in S\). Intuitively speaking, this means that there are no states that need to be removed from the sets of simulators \(\operatorname{Sim}\left(s_{1}\right)\) for \(s_{1} \in \operatorname{Pre}\left(s_{1}^{\prime}\right)\). Let \(s_{1}^{\prime}\) be a state such that \(\operatorname{Remove}\left(s_{1}^{\prime}\right) \neq \varnothing\). Now consider all pairs \(\left(s_{1}, s_{2}\right) \in S \times S\) such that
\[
s_{2} \in \operatorname{Remove}\left(s_{1}^{\prime}\right)=\operatorname{Pre}\left(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right)\right) \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)
\]
and \(s_{1} \in \operatorname{Pre}\left(s_{1}^{\prime}\right)\). Then, \(s_{1} \rightarrow s_{1}^{\prime}\), but there is no transition in \(s_{2} \rightarrow s_{2}^{\prime} \in \operatorname{Sim}\left(s_{1}^{\prime}\right)\). This yields \(s_{1} \preceq_{\text {TS }} s_{2}\). Therefore, \(s_{2}\) can be removed from \(\operatorname{Sim}\left(s_{1}\right)\). Accordingly, the set Remove \(\left(s_{1}\right)\) is extended with any state \(s\) such that
\[
s \in \operatorname{Pre}\left(s_{2}\right) \text { and } \operatorname{Post}(s) \cap \operatorname{Sim}\left(s_{1}\right)=\varnothing
\]

For these states \(s\), we have: if \(u \rightarrow s_{1}\), then there is no matching transition \(s \rightarrow t\) with \(t \in \operatorname{Sim}\left(s_{1}\right)\). Thus, \(u Ł_{\text {TS }} s\). Hence, if in a later iteration of the while-loop, state \(s_{1}\) is
```

Algorithm 35 Computation of the simulation order (second refinement)
Input: finite transition system $T S$ over $A P$ with state space $S$
Output: simulation order $\preceq_{T S}$

```
```

for all $s_{1} \in S$ do

```
for all \(s_{1} \in S\) do
    \(\operatorname{Sim}_{\text {old }}\left(s_{1}\right):=\) undefined;
    \(\operatorname{Sim}_{\text {old }}\left(s_{1}\right):=\) undefined;
    \(\operatorname{Sim}\left(s_{1}\right):=\left\{s_{2} \in S \mid L\left(s_{1}\right)=L\left(s_{2}\right)\right\}\)
    \(\operatorname{Sim}\left(s_{1}\right):=\left\{s_{2} \in S \mid L\left(s_{1}\right)=L\left(s_{2}\right)\right\}\)
od
od
while \(\exists s \in S\) with \(\operatorname{Sim}_{\text {old }}(s) \neq \operatorname{Sim}(s)\) do
while \(\exists s \in S\) with \(\operatorname{Sim}_{\text {old }}(s) \neq \operatorname{Sim}(s)\) do
    choose \(s_{1}^{\prime}\) such that \(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right) \neq \operatorname{Sim}\left(s_{1}^{\prime}\right)\)
    choose \(s_{1}^{\prime}\) such that \(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right) \neq \operatorname{Sim}\left(s_{1}^{\prime}\right)\)
    if \(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right)=\) undefined then
    if \(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right)=\) undefined then
        Remove \(\left(s_{1}^{\prime}\right):=S \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\)
        Remove \(\left(s_{1}^{\prime}\right):=S \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\)
    else
    else
        \(\operatorname{Remove}\left(s_{1}^{\prime}\right):=\operatorname{Pre}\left(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right)\right) \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\)
        \(\operatorname{Remove}\left(s_{1}^{\prime}\right):=\operatorname{Pre}\left(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right)\right) \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\)
    fi
    fi
    for all \(s_{1} \in \operatorname{Pre}\left(s_{1}^{\prime}\right)\) do
    for all \(s_{1} \in \operatorname{Pre}\left(s_{1}^{\prime}\right)\) do
        \(\operatorname{Sim}\left(s_{1}\right):=\operatorname{Sim}\left(s_{1}\right) \backslash \operatorname{Remove}\left(s_{1}^{\prime}\right)\)
        \(\operatorname{Sim}\left(s_{1}\right):=\operatorname{Sim}\left(s_{1}\right) \backslash \operatorname{Remove}\left(s_{1}^{\prime}\right)\)
    od
    od
    \(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right):=\operatorname{Sim}\left(s_{1}^{\prime}\right)\)
    \(\operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right):=\operatorname{Sim}\left(s_{1}^{\prime}\right)\)
od
od
return \(\left\{\left(s_{1}, s_{2}\right) \mid s_{2} \in \operatorname{Sim}\left(s_{1}\right)\right\}\)
```

return $\left\{\left(s_{1}, s_{2}\right) \mid s_{2} \in \operatorname{Sim}\left(s_{1}\right)\right\}$

```
chosen, then the simulator sets of the predecessors \(u \in \operatorname{Pre}\left(s_{1}\right)\) are regarded and state \(s\) is to be removed from \(\operatorname{Sim}(u)\). This yields Algorithm 36 .
```

Algorithm 36 Computation of the simulation order
Input: finite transition system $T S$ with state space $S$
Output: simulation order $\preceq_{\text {TS }}$

```
```

for all $s_{1} \in S$ do
$\operatorname{Sim}\left(s_{1}\right):=\left\{s_{2} \in S \mid L\left(s_{1}\right)=L\left(s_{2}\right)\right\} ;$
Remove $\left(s_{1}\right):=S \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}\right)\right)$
od
(* loop invariant: Remove $\left.\left(s_{1}^{\prime}\right) \subseteq S \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)^{*}\right)$
while $\left(\exists s_{1}^{\prime} \in S\right.$ with $\left.\operatorname{Remove}\left(s_{1}^{\prime}\right) \neq \varnothing\right)$ do
choose $s_{1}^{\prime}$ such that Remove $\left(s_{1}^{\prime}\right) \neq \varnothing$;
for all $s_{2} \in \operatorname{Remove}\left(s_{1}^{\prime}\right)$ do
for all $s_{1} \in \operatorname{Pre}\left(s_{1}^{\prime}\right)$ do
if $s_{2} \in \operatorname{Sim}\left(s_{1}\right)$ then
$\operatorname{Sim}\left(s_{1}\right):=\operatorname{Sim}\left(s_{1}\right) \backslash\left\{s_{2}\right\} ; \quad\left({ }^{*} s_{2} \in \operatorname{Sim}_{\text {old }}\left(s_{1}\right) \backslash \operatorname{Sim}\left(s_{1}\right)^{*}\right)$
for all $s \in \operatorname{Pre}\left(s_{2}\right)$ with $\operatorname{Post}(s) \cap \operatorname{Sim}\left(s_{1}\right)=\varnothing$ do
$\left({ }^{*} s \in \operatorname{Pre}\left(\operatorname{Sim}_{\text {old }}\left(s_{1}\right)\right) \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}\right)\right){ }^{*}\right)$
Remove $\left(s_{1}\right):=\operatorname{Remove}\left(s_{1}\right) \cup\{s\}$
od
fi
od
od
$\operatorname{Remove}\left(s_{1}^{\prime}\right):=\varnothing ; \quad\left(* \operatorname{Sim}_{\text {old }}\left(s_{1}^{\prime}\right):=\operatorname{Sim}\left(s_{1}^{\prime}\right) *\right)$
od
return $\left\{\left(s_{1}, s_{2}\right) \mid s_{2} \in \operatorname{Sim}\left(s_{1}\right)\right\}$

```

Theorem 7.81. Partial Correctness of Algorithm 36
On termination, Algorithm 36 returns \(\preceq_{\text {TS }}\).

Proof: First, observe that the outermost loop (i.e., the while-loop) maintains the following loop invariant. For all states \(s_{1} \in S\) :
(a) \(\operatorname{Remove}\left(s_{1}\right) \subseteq S \backslash \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}\right)\right)\).
(b) \(\left\{s_{2} \in S \mid s_{1} \preceq_{\text {TS }} s_{2}\right\} \subseteq \operatorname{Sim}\left(s_{1}\right) \subseteq\left\{s_{2} \in S \mid L\left(s_{1}\right)=L\left(s_{2}\right)\right\}\).
(c) For all \(s_{2} \in \operatorname{Sim}\left(s_{1}\right)\), one of the following two statements holds:
- \(\exists s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\operatorname{Post}\left(s_{2}\right) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right)=\varnothing\) and \(s_{2} \in \operatorname{Remove}\left(s_{1}^{\prime}\right)\),
- \(\operatorname{Post}\left(s_{2}\right) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right) \neq \varnothing\) for all \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\).

From (c), it follows that whenever \(\operatorname{Remove}\left(s_{1}^{\prime}\right)=\varnothing\) for all \(s_{1}^{\prime} \in S\), then:
\[
\forall s_{1} \in S . \forall s_{2} \in \operatorname{Sim}\left(s_{1}\right) . \forall s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right) . \operatorname{Post}\left(s_{2}\right) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right) \neq \varnothing
\]

Therefore, on termination, the relation \(\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \mid s_{2} \in \operatorname{Sim}\left(s_{1}\right)\right\}\) is a simulation for TS. Assertion (b) implies that \(\mathcal{R}\) agrees with \(\preceq_{T S}\), as \(\preceq_{T S}\) is the coarsest simulation for TS.

\section*{Lemma 7.82. Termination of Algorithm 36}

For each pair \(\left(s_{2}, s_{1}^{\prime}\right) \in S \times S\), state \(s_{2}\) is inserted in (and removed from) the set Remove ( \(s_{1}^{\prime}\) ) at most once.

In particular, Algorithm 36 requires at most \(\mathcal{O}\left(|S|^{2}\right)\) iterations.
Proof: Assume \(s_{2} \in \operatorname{Remove}\left(s_{1}^{\prime}\right)\) and let \(s_{1}^{\prime}\) be the state that is selected in the outermost iteration. Then \(s_{2} \notin \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\) (by part (a) of the loop invariant established in the proof of Theorem 7.81). Since the simulator sets are decreasing, \(s_{2} \notin \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\) in all further iterations: the only possibility to insert \(s=s_{2}\) in \(\operatorname{Remove}\left(s_{1}^{\prime}\right)\) is when \(s \in \operatorname{Pre}\left(\bar{s}_{2}\right)\) for some state \(\bar{s}_{2} \in \operatorname{Sim}\left(s_{1}^{\prime}\right)\) with \(\operatorname{Post}(s) \cap \operatorname{Sim}\left(s_{1}^{\prime}\right)=\left\{\bar{s}_{2}\right\}\). But then \(s_{2}=s \in \operatorname{Pre}\left(\operatorname{Sim}\left(s_{1}^{\prime}\right)\right)\). Thus, \(s_{2}\) will never be added to Remove \(\left(s_{1}^{\prime}\right)\) once it has been deleted from Remove \(\left(s_{1}^{\prime}\right)\).

This lemma is the starting point for deriving the time complexity of Algorithm 36. Let \(M\) be the number of edges in the state graph of \(T S\) and assume \(M \geqslant|S|\). Assume that list representations for the sets of predecessors \(\operatorname{Pre}(\cdot)\) are available. As for the initial algorithm to compute \(\preceq_{T S}\), we deal with counters
\[
\delta\left(s_{1}, s\right)=\left|\operatorname{Post}(s) \cap \operatorname{Sim}\left(s_{1}\right)\right| .
\]

Using Algorithm 29 (page 479), the initial simulator sets \(\operatorname{Sim}\left(s_{1}\right)=\left\{s_{2} \in S \mid L\left(s_{1}\right)=\right.\) \(\left.L\left(s_{2}\right)\right\}\) can be obtained in time \(\mathcal{O}(|S| \cdot|A P|)\). We then compute the sets Remove \(\left(s_{1}\right)\) and the counters \(\delta\left(s_{1}, s\right)\) in time \(\mathcal{O}(M \cdot|S|)\). In each iteration the counters \(\delta\left(s_{1}, s\right)\) need to be updated. This is done as follows. On removing \(s_{2}\) from \(\operatorname{Sim}\left(s_{1}\right)\), the list \(\operatorname{Pre}\left(s_{2}\right)\) is traversed and we set
\[
\delta\left(s_{1}, s\right):=\delta\left(s_{1}, s\right)-1 \quad \text { for every } s \in \operatorname{Pre}\left(s_{1}\right)
\]

The check \(\operatorname{Post}(s) \cap \operatorname{Sim}\left(s_{1}\right)=\varnothing\) boils down to \(\delta\left(s_{1}, s\right)=0\) and can be performed in constant time. The total costs of the outermost iteration can be estimated as follows. For
each state \(s_{2}\) and each edge \(s \rightarrow s_{2}\), the condition " \(s_{2} \in \operatorname{Remove}\left(s_{1}^{\prime}\right) \wedge s_{2} \in \operatorname{Sim}\left(s_{1}\right)\) " is satisfied at most once. Therefore, the code fragment
```

if }\mp@subsup{s}{2}{}\in\operatorname{Sim}(\mp@subsup{s}{1}{})\mathrm{ then
Sim}(\mp@subsup{s}{1}{}):=\operatorname{Sim}(\mp@subsup{s}{1}{})\{\mp@subsup{s}{2}{}}
for all }s\in\operatorname{Pre}(\mp@subsup{s}{2}{})\mathrm{ do
if Post(s)\cap\operatorname{Sim}(\mp@subsup{s}{1}{})=\varnothing\mathrm{ then}
Remove(s)}:=\mathrm{ Remove (s1) }\cup{s}
fi
od
fi

```
causes the cost:
\[
\mathcal{O}(\underbrace{\sum_{s_{2} \in S}\left|\operatorname{Pre}\left(s_{2}\right)\right|}_{M} \cdot \underbrace{\sum_{s_{1} \in S} 1}_{|S|})=\mathcal{O}(M \cdot|S|)
\]
where we add up the costs over all iterations for which \(s_{2} \in \operatorname{Sim}\left(s_{1}\right)\). The condition \(s_{2} \in \operatorname{Remove}\left(s_{1}^{\prime}\right)\) of the outermost for-loop is fulfilled at most once (see Lemma 7.82). Thus, the total costs for all inner loops
```

for all $s_{2} \in \operatorname{Remove}\left(s_{1}^{\prime}\right)$ do
for all $s_{1} \in \operatorname{Pre}\left(s_{2}^{\prime}\right)$ do
od
od

```
is bounded by \(\mathcal{O}(M)\). Under the assumption that \(M \geqslant|S|\), we thus obtain:

\section*{Theorem 7.83. Complexity of Algorithm 36}

The simulation order \(\preceq\) of finite transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) can be computed with Algorithm 36 in time \(\mathcal{O}(M \cdot|S|+|S| \cdot|A P|)\).

The simulation equivalence \(\simeq_{T S}\) and the simulation quotient system \(T S / \simeq\) can be computed with the same complexity.

\subsection*{7.7 Stutter Linear-Time Relations}

The equivalence and preorder relations considered so far require that each outgoing transition of state \(s\) is mimicked by an outgoing transition of the related state \(s^{\prime}\). We now consider variants of trace equivalence and bisimulation that relax this requirement. Rather than considering all transitions as "visible", it is allowed to mimic an outgoing transition \(s\) by a sequence of transitions starting from \(s^{\prime}\). Such sequences of transitions need to be "invisible", that is to say, all state changes (except the last state change) in such sequences should not change the truth value of the atomic propositions that hold in \(s^{\prime}\). Such state changes are called stutter steps. Equivalences and preorders that abstract from these stutter steps, also referred to as internal or nonobservable steps, are called weak. In contrast, trace equivalence, \(\sim, \preceq\), and \(\simeq\) are strong implementation relations as these relations consider stutter steps as any other transition. Weak implementation relations are important for system synthesis as well as system analysis. To compare transition systems that model a given system at different abstraction levels, it is often too demanding to require a statewise equivalence. Instead, an action in a transition system at a high level of abstraction can be modeled by a sequence of actions in the more concrete transition system. Establishing, e.g., trace equivalence is then simply impossible. This is illustrated by the following example.

\section*{Example 7.84. Abstraction of Internal Moves}

Consider the abstract program fragment \(x:=y\) !, and let \(T S_{a b s}\) be its underlying transition system. Let \(T S_{\text {conc }}\) model the concrete program fragment
```

i:= y;z:= 1;
while }i>1\mathrm{ do
z:=z*i;i:= i-1;
od
x := z;

```
that computes the factorial of \(y\) iteratively. Clearly, \(T S_{a b s}\) and \(T S_{\text {conc }}\) are not traceequivalent. They, however, are related after abstracting from the iteration (and initial assignments) in the concrete program under the assumption that the iteration (plus initial assignments) does not affect the truth value of atomic propositions. This is guaranteed when restricting to atomic propositions that only refer to the values of \(x\) and \(y\), and, e.g., not to the individual program locations.

Secondly, by abstracting from internal steps, quotient transition systems are obtained that may be significantly smaller than the quotient under the corresponding strong implementation relation. This is due to the fact that quotienting with respect to a weak relation
allows for abstracting from sequences of transitions such that all states on such paths can be aggregated. Interestingly, though, still a rather rich set of properties is preserved under such abstractions.

This section is concerned with a weak version of trace inclusion and trace equivalence. Section 7.8 (from page 536 onward) deals with weak versions of bisimulation.

\subsection*{7.7.1 Stutter Trace Equivalence}

Internal steps are transitions that do not affect the state labels of successive states. Thus, intuitively, an internal step operates on program or control variables that are either not visible from the outside or viewed to be irrelevant at a certain abstraction level. Such transitions are called stutter steps.

\section*{Definition 7.85. Stutter Step}

Transition \(s \rightarrow s^{\prime}\) in transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) is a stutter step if \(L(s)=L\left(s^{\prime}\right)\).

The notion of stuttering is lifted to paths as follows. Two paths are called stutterequivalent if their traces only differ in their stutter steps, i.e., if there is a sequence \(A_{0} A_{1} A_{2} \ldots\) of sets of atomic propositions \(A_{i} \subseteq A P\) such that the traces of both paths have the form \(A_{0}^{+} A_{1}^{+} A_{2}^{+} \ldots\)

\section*{Definition 7.86. Stutter-Equivalence of Paths}

Let \(T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right)\) be transition systems without terminal states and \(\pi_{i} \in \operatorname{Paths}\left(T S_{i}\right), i=1,2 . \pi_{1}\) and \(\pi_{2}\) are stutter-equivalent, denoted \(\pi_{1} \triangleq \pi_{2}\), if there exists an infinite sequence \(A_{0} A_{1} A_{2} \ldots\) with \(A_{i} \subseteq A P\) and natural numbers \(n_{0}, n_{1}, n_{2}, \ldots\), \(m_{0}, m_{1}, m_{2}, \ldots \geqslant 1\) such that
\[
\begin{aligned}
& \operatorname{trace}\left(\pi_{1}\right)=\underbrace{A_{0} \ldots A_{0}}_{n_{0} \text {-times }} \underbrace{A_{1} \ldots A_{1}}_{n_{1} \text {-times }} \underbrace{A_{2} \ldots A_{2}}_{m_{0} \text {-times }} \ldots \\
& \operatorname{trace}\left(\pi_{2}\right)=\underbrace{A_{0} \ldots A_{0}}_{m_{1} \text {-times }} \underbrace{A_{1} \ldots A_{1}}_{m_{2} \text {-times }} \underbrace{A_{2} \ldots A_{2}}_{2} \ldots
\end{aligned}
\]

Finite path fragments \(\widehat{\pi}_{1}\) in \(T S_{1}\) and \(\widehat{\pi}_{2}\) in \(T S_{2}\) are stutter equivalent, denoted \(\widehat{\pi}_{1} \triangleq \widehat{\pi}_{2}\), if there exists a finite sequence \(A_{0} \ldots A_{n} \in\left(2^{A P}\right)^{+}\)such that trace \(\left(\widehat{\pi}_{1}\right)\) and trace \(\left(\widehat{\pi}_{2}\right)\) are contained in the language given by the regular expression \(A_{0}^{+} A_{1}^{+} \ldots A_{n}^{+}\).

Note that it is not required that \(A_{0}, A_{1}, A_{2}, \ldots\) are distinct. The notion of stutter equivalence can be applied to (finite or infinite) path fragments of transition system \(T S\) by setting \(T S_{1}=T S_{2}=T S\).

\section*{Example 7.87. Stutter-Equivalent Paths}

Consider the transition system \(T S_{S e m}\) depicted in Figure 7.29 and let \(A P=\left\{\right.\) crit \(_{1}\), crit \(\left._{2}\right\}\).


Figure 7.29: Transition system of semaphore-based mutual exclusion algorithm.
All infinite paths in \(T S_{S e m}\) that agree on the first process acquiring access to the critical section, and in which the two processes acquire access in a strictly alternating fashion, are stutter-equivalent. For the case that process \(P_{1}\) is the first to enter the critical section, all such paths have a trace of the form
\[
\varnothing \ldots \varnothing \underbrace{\left\{\operatorname{crit}_{1}\right\} \ldots\left\{\text { crit }_{1}\right\}}_{P_{1} \text { in cs }} \varnothing \ldots \varnothing \underbrace{\left\{\operatorname{crit}_{2}\right\} \ldots\left\{\text { crit }_{2}\right\}}_{P_{2} \text { in cs }} \varnothing \ldots \varnothing \underbrace{\left\{\operatorname{crit}_{1}\right\} \ldots\left\{\text { crit }_{1}\right\}}_{P_{1} \text { in cs }} \ldots
\]

Consider, for instance, the following two infinite paths in \(T S_{S e m}\) :
\[
\begin{aligned}
\pi_{1}= & \left\langle n_{1}, n_{2}\right\rangle \rightarrow\left\langle w_{1}, n_{2}\right\rangle \rightarrow\left\langle w_{1}, w_{2}\right\rangle \rightarrow\left\langle c_{1}, w_{2}\right\rangle \rightarrow\left\langle n_{1}, w_{2}\right\rangle \rightarrow \\
& \left\langle n_{1}, c_{2}\right\rangle \rightarrow\left\langle n_{1}, n_{2}\right\rangle \rightarrow\left\langle w_{1}, n_{2}\right\rangle \rightarrow\left\langle w_{1}, w_{2}\right\rangle \rightarrow\left\langle c_{1}, w_{2}\right\rangle \rightarrow \ldots \\
\pi_{2}= & \left\langle n_{1}, n_{2}\right\rangle \rightarrow\left\langle w_{1}, n_{2}\right\rangle \rightarrow\left\langle c_{1}, n_{2}\right\rangle \rightarrow\left\langle c_{1}, w_{2}\right\rangle \rightarrow\left\langle n_{1}, w_{2}\right\rangle \rightarrow \\
& \left\langle w_{1}, w_{2}\right\rangle \rightarrow\left\langle w_{1}, c_{2}\right\rangle \rightarrow\left\langle w_{1}, n_{2}\right\rangle \rightarrow\left\langle c_{1}, n_{2}\right\rangle \rightarrow \ldots
\end{aligned}
\]
where for the sake of simplicity, we omitted the value of variable \(y\) in each state. Hence,
\(\pi_{1} \triangleq \pi_{2}\), since for \(A P=\left\{\right.\) crit \(_{1}\), crit \(\left.t_{2}\right\}\)
\[
\begin{aligned}
\operatorname{trace}\left(\pi_{1}\right) & =\varnothing^{3}\left\{\text { crit }_{1}\right\} \varnothing\left\{\text { crit }_{2}\right\} \varnothing^{3}\left\{\text { crit }_{1}\right\} \ldots \text { and } \\
\operatorname{trace}\left(\pi_{2}\right) & =\varnothing^{2}\left(\left\{\text { crit }_{1}\right\}\right)^{2} \varnothing^{2}\left\{\text { crit }_{2}\right\} \varnothing\left\{\text { crit }_{1}\right\} \ldots
\end{aligned}
\]

Figure 7.30 indicates trace \(\left(\pi_{1}\right)\) and \(\operatorname{trace}\left(\pi_{2}\right)\) and their stutter equivalence.


Figure 7.30: Stutter-equivalent paths in \(T S_{S e m}\).

\section*{Notation 7.88. Stutter Equivalence for Executions and Traces}

The notion of stutter equivalence can be adapted to execution fragments and words over \(2^{A P}\) in the obvious way. Two execution fragments \(\rho_{1}\) and \(\rho_{2}\) of \(T S_{1}\) and \(T S_{2}\) are stutterequivalent, denoted \(\rho_{1} \triangleq \rho_{2}\), if the induced path fragments are stutter-equivalent. Accordingly, traces \(\sigma_{1}\) and \(\sigma_{2}\) over \(2^{A P}\) are stutter-equivalent, denoted \(\sigma_{1} \triangleq \sigma_{2}\), if they are both of the form \(A_{0}^{+} A_{1}^{+} A_{2}^{+} \ldots\) for \(A_{0}, A_{1}, A_{2}, \ldots \subseteq A P\).

Transition systems \(T S_{1}\) and \(T S_{2}\) are stutter-equivalent whenever each trace of \(T S_{1}\) can be mimicked by a stutter equivalent trace in \(T S_{2}\), and vice versa.

\section*{Definition 7.89. Stutter Equivalence of Transition Systems}

Transition systems \(T S_{i}\) over \(A P, i=1,2\), are stutter trace equivalent, denoted \(T S_{1} \triangleq T S_{2}\), if \(T S_{1} \unlhd T S_{2}\) and \(T S_{2} \unlhd T S_{1}\), where \(\unlhd\) is defined by:
\[
T S_{1} \unlhd T S_{2} \quad \text { iff } \quad \forall \sigma_{1} \in \operatorname{Traces}\left(T S_{1}\right) \quad\left(\exists \sigma_{2} \in \operatorname{Traces}\left(T S_{2}\right) . \quad \sigma_{1} \triangleq \sigma_{2}\right)
\]

Evidently, it follows that \(\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}(T S) 2\) implies \(T S_{1} \unlhd T S_{2}\).

\section*{Example 7.90. Stutter Trace Inclusion}

Consider the transition systems \(T S_{1}\) (left), \(T S_{2}\) (middle) and \(T S_{3}\) (right) in Figure 7.31. It follows:
\[
\begin{aligned}
\operatorname{Traces}\left(T S_{1}\right) & =\left\{(a \varnothing)^{\omega},(a \varnothing)^{*} a^{\omega}\right\} \\
\operatorname{Traces}\left(T S_{2}\right) & =\left\{\left(a^{+} \varnothing\right)^{\omega},\left(a^{+} \varnothing\right)^{*} a^{\omega}\right\} \\
\operatorname{Traces}\left(T S_{3}\right) & =\left\{a^{\omega}, a^{+}(\varnothing a)^{\omega}\right\} .
\end{aligned}
\]

We have \(T S_{1} \unlhd T S_{2}\) since \(\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right)\), and \(T S_{2} \unlhd T S_{1}\) as trace \(\left(a^{+} \varnothing\right)^{\omega}\) is stutter-equivalent to \((a \varnothing)^{\omega}\) and trace \(\left(a^{+} \varnothing\right)^{*} a^{\omega}\) is stutter-equivalent to \((a \varnothing)^{*} a^{\omega}\). Accordingly,
\[
T S_{1} \triangleq T S_{2}
\]

Note that \(\operatorname{Traces}\left(T S_{2}\right) \nsubseteq \operatorname{Traces}\left(T S_{1}\right)\), as e.g., \(a^{2} \varnothing \ldots \in \operatorname{Traces}\left(T S_{2}\right)\) but is not a trace of \(T S_{1}\). It follows that
\[
T S_{1} \notin T S_{3} \text { and } T S_{2} \npreceq T S_{3},
\]
as \(T S_{1}\) and \(T S_{2}\) have traces of the form \(\left(a^{+} \varnothing\right)^{+} a^{\omega}\) whereas there is no trace in \(T S_{3}\) that contains \(\varnothing\) and ends with \(a^{\omega}\). Since \(\operatorname{Traces}\left(T S_{3}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right)\), it follows \(T S_{3} \unlhd T S_{2}\). \(T S_{3} \unlhd T S_{1}\) since \(a^{\omega}\) is a trace of both transition systems, and \(a^{+}(\varnothing a)^{\omega}\) of \(T S_{3}\) is stutterequivalent to \((a \varnothing)^{\omega}\).


Figure 7.31: Three transition systems.

\subsection*{7.7.2 Stutter Trace and LTL \(_{\backslash \bigcirc}\) Equivalence}

In analogy to the result stating that trace equivalence is finer than LTL equivalence, we now consider the fragment of LTL that is preserved by stutter trace equivalence. To that end, we first determine the properties that are preserved by stutter-equivalent paths and words over \(2^{A P}\), respectively. Let us discuss why a restriction to logics without the next operator is necessary. Consider, e.g., the traces \(\sigma_{1}=A B B B \ldots\) and \(\sigma_{2}=A A A B B B B \ldots\) with \(A, B \subseteq A P\) and \(A \neq B\). Clearly,
\[
\sigma_{1} \triangleq \sigma_{2} \quad \text { but } \quad \sigma_{1} \models \bigcirc b \quad \text { and } \quad \sigma_{2} \not \models \bigcirc b \text { for } b \in B \backslash A
\]

Stated in words: stutter equivalence does not preserve the truth value of formulae with the next operator. In fact, it turns out that this is the only modal operator that is not preserved.

\section*{Notation 7.91. LTL without Next Step}
\(\mathrm{LTL}_{\backslash \bigcirc}\) denotes the class of LTL formulae without the next step operator \(\bigcirc\).

Theorem 7.92. Stutter Equivalence and LTL \({ }_{\backslash \bigcirc}\) Equivalence
For \(\sigma_{1}, \sigma_{2} \in\left(2^{A P}\right)^{\omega}\) :
\[
\sigma_{1} \triangleq \sigma_{2} \Rightarrow \quad\left(\sigma_{1} \models \varphi \text { if and only if } \sigma_{2} \models \varphi\right)
\]
for any \(L T L_{\backslash \bigcirc}\) formula \(\varphi\) over \(A P\).

Proof: The proof is by structural induction over the formula \(\varphi\). Let \(A_{0} A_{1} A_{2} \ldots\) be an infinite word over \(2^{A P}\) and
\[
\sigma_{1}=A_{0}^{n_{0}} A_{1}^{n_{1}} A_{2}^{n_{2}} \ldots \text { and } \sigma_{2}=A_{0}^{m_{0}} A_{1}^{m_{1}} A_{2}^{m_{2}} \ldots
\]
where \(n_{0}, n_{1}, n_{2}, \ldots\) and \(m_{0}, m_{1}, m_{2}, \ldots\) are positive natural numbers. Hence: \(\sigma_{1} \triangleq \sigma_{2}\).

Basis: For \(\varphi=\) true the proposition is clear. For \(\varphi=a \in A P\), we have
\[
\sigma_{1} \models a \quad \text { iff } \quad a \in A_{0} \quad \text { iff } \quad \sigma_{2} \models a .
\]

Induction step: For \(\varphi=\varphi_{1} \wedge \varphi_{2}\) or \(\varphi=\neg \varphi^{\prime}\), the claim follows immediately from the induction hypothesis applied to \(\varphi_{1}\) and \(\varphi_{2}\) or \(\varphi^{\prime}\), respectively. The remaining case is \(\varphi=\varphi_{1} \mathrm{U} \varphi_{2}\). Assume \(\sigma_{1} \models \varphi\). From the semantics of \(\mathrm{LTL}_{\backslash \bigcirc}\), it follows that there exists a natural number \(j\) such that
\[
\sigma_{1}[j . .] \models \varphi_{2} \quad \text { and } \quad \sigma_{1}[i . .] \models \varphi_{1} \text { for all } 0 \leqslant i<j .
\]

Recall that for \(\sigma=B_{0} B_{1} B_{2} \ldots\) and \(h \geqslant 0\), the suffix \(B_{h} B_{h+1} \ldots\) of \(\sigma\) is denoted by \(\sigma[h .\).\(] .\)
Let \(r \geqslant 0\) be such that
\[
n_{0}+\ldots+n_{r-1}<j \leqslant n_{0}+\ldots+n_{r-1}+n_{r}
\]

Stated in words, \(r\) is the index of the \(A\)-block in \(\sigma_{1}\) that contains \(\sigma_{1}[j]\). Then, \(\sigma_{1}[j .\).\(] is\) obtained from \(\sigma_{1}\) by eliminating the prefix \(A_{0}^{n_{0}} \ldots A_{r-1}^{n_{r-1}} A_{r}^{n}\) where \(n=n_{0}+\ldots+n_{r-1}+\) \(n_{r}-j\). Note that \(0 \leqslant n<n_{r}\). Thus, \(\sigma_{1}[j .\).\(] is of the form A_{r}^{+} A_{r+1}^{+} A_{r+2}^{+} \ldots\). Since \(\sigma_{1} \triangleq \sigma_{2}\), it follows that for:
\[
k=m_{0}+\ldots+m_{r-1}+1,
\]
\(\sigma_{2}[k .\).\(] is of the form A_{r}^{+} A_{r+1}^{+} A_{r+2}^{+} \ldots\). More precisely, we have
1. \(\sigma_{1}[j ..] \triangleq \sigma_{2}[k .\).\(] since both words are of the form A_{r}^{+} A_{r+1}^{+} A_{r+2}^{+} \ldots\), and
2. for all \(0 \leqslant h<k\), there is an index \(0 \leqslant i<j\) such that \(\sigma_{1}[i ..] \triangleq \sigma_{2}[h .\).\(] .\)

As \(\sigma_{1}[j ..] \models \varphi_{2}\) and \(\sigma_{1}[i ..] \models \varphi_{1}\), for all \(i<j\), applying the induction hypothesis yields \(\sigma_{2}[k ..] \models \varphi_{2}\) and \(\sigma_{2}[h ..] \models \varphi_{1}\), for \(h<k\). Hence, \(\sigma_{2} \models \varphi_{1} \cup \varphi_{2}\).

By symmetry, we get the equivalence of \(\sigma_{1}\) and \(\sigma_{2}\) for \(\mathrm{LTL}_{\backslash \bigcirc}\).

\section*{Corollary 7.93. Stutter Trace Relations and LTL}

For transition systems \(T S_{1}, T S_{2}\) (over AP) without terminal states:
(a) \(T S_{1} \triangleq T S_{2}\) implies \(T S_{1} \equiv_{L T L \_{\bigcirc}} T S_{2}\).
(b) if \(T S_{1} \unlhd T S_{2}\), then for any \(L T L_{\backslash \bigcirc}\) formula \(\varphi\) : \(T S_{2} \models \varphi\) implies \(T S_{1} \models \varphi\).

A slightly more general preservation result can be established for stutter trace equivalence and inclusion.

\section*{Definition 7.94. Stutter-Insensitive LT property}

LT property \(P\) is stutter-insensitive if \([\sigma]_{\triangleq} \subseteq P\), for any \(\sigma \in P\).

Stated in words, \(P\) is stutter-insensitive if it is closed under stutter equivalence, i.e., for any \(\sigma \in P\) all stutter-equivalent words are also contained in \(P\). It follows immediately that stutter trace equivalent transition systems satisfy the same stutter-insensitive LT properties. Moreover, for any stutter-insensitive LT property \(P\) :
\[
T S_{1} \unlhd T S_{2} \text { and } T S_{2} \models P \quad \text { implies } \quad T S_{1} \models P .
\]

This is, in fact, a more general statement than Corollary 7.93 since for any LTL \(_{\backslash \bigcirc}\) formula \(\varphi\) the induced LT property \(\operatorname{Words}(\varphi)\) is stutter-insensitive. On the other hand, there exist stutter-insensitive LT properties that cannot be expressed in LTL \(\backslash \bigcirc\). For instance, consider the set \(P\) of words over \(\{a, b\}\) that contain an odd number of occurrences of the subword \(a b\). The LT property that is stutter-insensitive and contains \(P\), cannot be expressed in \(\operatorname{LTL}_{\backslash \bigcirc}\). However, if \(\varphi\) is an LTL formula such that \(\operatorname{Words}(\varphi)\) is stutter insensitive, then \(\varphi\) is equivalent to some \(\mathrm{LTL}_{\backslash \bigcirc}\) formula \(\psi\); see Exercise 7.20.

\subsection*{7.8 Stutter Bisimulation}

This section is concerned with stutter bisimulation. Stutter bisimulation is defined in a coinductive manner, as bisimulation. Whereas bisimulation requires for equivalent states \(s_{1}\) and \(s_{2}\) that each transition \(s_{1} \rightarrow t_{1}\) (with \(s_{1}\) inequivalent to \(t_{1}\) ) is matched by some transition \(s_{2} \rightarrow t_{2}\), stutter bisimulation allows \(s_{1} \rightarrow t_{1}\) to be matched by a path fragment \(s_{2} u_{1} u_{2} \ldots u_{n} t_{2}\) (for \(n \geqslant 0\) ) such that \(t_{1}\) and \(t_{2}\) are equivalent, and \(u_{i}\) is equivalent to \(s_{2}\). That is, single transitions may be matched by (suitable) path fragments.

The following definition considers the notion of a stutter bisimulation for a single transition system. Later on, this notion will be adapted for pairs of transition systems.

\section*{Definition 7.95. Stutter Bisimulation}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system. A stutter bisimulation for \(T S\) is a binary relation \(\mathcal{R}\) on \(S\) such that for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) :
1. \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
2. If \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\left(s_{1}^{\prime}, s_{2}\right) \notin \mathcal{R}\), then there exists a finite path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) with \(n \geqslant 0\) and \(\left(s_{1}, u_{i}\right) \in \mathcal{R}, i=1, \ldots, n\) and \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
3. If \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}, s_{2}^{\prime}\right) \notin \mathcal{R}\), then there exists a finite path fragment \(s_{1} v_{1} \ldots v_{n} s_{1}^{\prime}\) with \(n \geqslant 0\) and \(\left(v_{i}, s_{2}\right) \in \mathcal{R}, i=1, \ldots, n\) and \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
\(s_{1}, s_{2}\) are stutter bisimulation equivalent (stutter-bisimilar, for short), denoted \(s_{1} \approx_{T S} s_{2}\), if there exists a stutter bisimulation \(\mathcal{R}\) for \(T S\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).

Condition (1) is standard, and requires equivalent states to be equally labeled. According to condition (2), every outgoing transition \(s_{1} \rightarrow t_{1}\) (where \(s_{1}\) is not equivalent to \(t_{1}\) ) must be matched by a path fragment that leads from \(s_{2}\) to \(t_{2}\) such that \(t_{1}\) and \(t_{2}\) are equivalent, and all intermediate states in the path fragment are equivalent to \(s_{2}\). Roughly speaking, if \(s_{1}\) changes its equivalence class and moves to \(t_{1}\), this must be mimicked by \(s_{2}\), but only after some transitions that are internal to the equivalence class of \(s_{2}\). Condition (3) is the symmetric counterpart of condition (2).

\section*{Lemma 7.96. Coarsest Stutter Bisimulation}

For transition system \(T S\) with state space \(S\) :
1. \(\approx_{T S}\) is an equivalence relation on \(S\).
2. \(\approx_{T S}\) is a stutter bisimulation for TS.
3. \(\approx_{T S}\) is the coarsest stutter bisimulation for \(T S\) and coincides with the union of all stutter bisimulations for TS.

Proof: Similar to the case for bisimulation; see Exercise 7.26.
(Note that \(\approx_{T S}\) is an equivalence, but this does not hold for any stutter bisimulation relation.) Since \(\approx_{T S}\) is a stutter bisimulation, condition (2) of Definition 7.95 can be rephrased by means of \(\approx_{T S}\), as illustrated in Figure 7.32.

\section*{Example 7.97. Stutter Bisimulation in Semaphore-Based Mutual Exclusion}

Consider the semaphore-based solution to the mutual exclusion problem; see the transition system \(T S_{S e m}\) in Figure 5.5 (page 239). Let \(A P=\left\{\right.\) crit \(_{1}\), crit \(\left._{2}\right\}\). Relation \(\mathcal{R}\) that induces the following partitioning of the state space is a stutter bisimulation:
\[
\left\{\left\{\left\langle n_{1}, n_{2}\right\rangle,\left\langle n_{1}, w_{2}\right\rangle,\left\langle w_{1}, n_{2}\right\rangle,\left\langle w_{1}, w_{2}\right\rangle\right\},\left\{\left\langle c_{1}, n_{2}\right\rangle,\left\langle c_{1}, w_{2}\right\rangle\right\},\left\{\left\langle c_{2}, n_{1}\right\rangle,\left\langle w_{1}, c_{2}\right\rangle\right\}\right\} .
\]

This can be checked by verifying the conditions in Definition 7.95. For instance, \(\left\langle n_{1}, n_{2}\right\rangle \approx_{T S}\) \(\left\langle w_{1}, w_{2}\right\rangle\), since \(\left\langle w_{1}, w_{2}\right\rangle \rightarrow\left\langle w_{1}, c_{2}\right\rangle\) can be mimicked by the path fragment \(\left\langle n_{1}, n_{2}\right\rangle \rightarrow\) \(\left\langle n_{1}, w_{2}\right\rangle \rightarrow\left\langle n_{1}, c_{2}\right\rangle\), and \(\left\langle w_{1}, w_{2}\right\rangle \rightarrow\left\langle c_{1}, w_{2}\right\rangle\) is matched by path fragment \(\left\langle n_{1}, n_{2}\right\rangle \rightarrow\) \(\left\langle w_{1}, n_{2}\right\rangle \rightarrow\left\langle c_{1}, n_{2}\right\rangle\). As \(\left\langle n_{1}, n_{2}\right\rangle\) does not have any direct successor that is inequivalent to


Figure 7.32: Condition (2) for stutter bisimulation equivalence.
it, no requirements are imposed on \(\left\langle w_{1}, w_{2}\right\rangle\). In a similar way, the stutter-bisimilarity of other pairs of states can be checked.

\section*{Example 7.98. Stutter Bisimulation in Peterson's Mutual Exclusion Algorithm}

Consider the transition system \(T S_{P e t}\) for Peterson's mutual exclusion algorithm (see Figure 7.33 on page 539) and let \(A P=\left\{c r i t_{1}, c_{1} t_{2}\right\}\). The initial states
\[
s_{0,1}=\left\langle n_{1}, n_{2}, x=1\right\rangle \quad \text { and } \quad s_{0,2}=\left\langle n_{1}, n_{2}, x=2\right\rangle
\]
are stutter-bisimilar. All other states represent separate stutter bisimulation equivalence classes. Essentially, this is due to the fact that
\[
\underbrace{\left\langle c_{1}, n_{2}, x=2\right\rangle}_{s_{1}} \not \overbrace{T S} \underbrace{\left\langle c_{1}, w_{2}, x=1\right\rangle}_{s_{2}}
\]

This follows from the fact that \(s_{1}\) can move to one of the initial states, whereas \(s_{2}\) cannot mimic this by a series of stutter steps. In state \(s_{1}\), it is possible that process 1 reenters the critical section next, whereas in state \(s_{2}\), process 2 is always granted access to the critical section next. For the other states, a similar reasoning applies.

Stutter bisimulation for pairs of transition systems is defined as follows.

\section*{Definition 7.99. Stutter-Bisimilar Transition Systems}

Let \(T S_{i}=\left(S_{i}, A c t_{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right), i=1,2\), be transition systems over AP. \(T S_{1}\) and


Figure 7.33: Transition system for Peterson's mutual exclusion algorithm.
\(T S_{2}\) are stutter bisimulation equivalent (stutter-bisimilar, for short), denoted \(T S_{1} \approx T S_{2}\), if there exists a stutter bisimulation \(\mathcal{R}\) on \(\left(S_{1} \times S_{2}\right) \cup\left(S_{1} \times S_{2}\right)\) such that
\[
\forall s_{1} \in I_{1} \cdot\left(\exists s_{2} \in I_{2} \cdot\left(s_{1}, s_{2}\right) \in \mathcal{R}\right) \text { and } \forall s_{2} \in I_{2} \cdot\left(\exists s_{1} \in I_{1} \cdot\left(s_{1}, s_{2}\right) \in \mathcal{R}\right) .
\]

The connection between Definitions 7.95 and 7.99 is as for ordinary bisimulation. We have \(T S_{1} \approx T S_{2}\) if and only if for every initial state of \(T S_{1}\) there exists a stutter-bisimilar initial state of \(T S_{2}\), and vice versa, where stutter bisimulation equivalence of states \(s_{1}\) in \(T S_{1}\) and \(s_{2}\) in \(T S_{2}\) thereby refers to the stutter bisimulation equivalence \(\approx_{T S_{1} \oplus T S_{2}}\). Recall that the operation \(\oplus\) on transition systems amounts to the union of transition systems, see page 457). Vice versa, stutter bisimulation equivalence \(\approx_{T S}\) of a single transition system can be obtained from Definition 7.99 by the observation that \(s_{1} \approx_{T S} s_{2}\) if and only if \(T S_{s_{1}} \approx T S_{s_{2}}\), where \(T S_{s}\) is obtained from \(T S\) by declaring \(s\) as the unique initial state.

Figure 7.34 surveys the stutter implementation relations in this monograph. Stutter bisimulation equivalence with divergence is a variant of stutter bisimulation and will be introduced in Section 7.8.1 (page 543 and further).

\section*{Example 7.100. Door Opener}

Consider an automatic door opener as modeled by the (concrete) transition system in Figure 7.35; for simplicity, the action labels are omitted from the transitions. Let \(A P=\) \(\{\) alarm, open \(\}\). The door opener requires a three-digit code \(d_{1} d_{2} d_{3}\) as input with \(d_{i} \in\)
stutter trace inclusion:
\[
T S_{1} \unlhd T S_{2} \quad \text { iff } \quad \forall \sigma_{1} \in \operatorname{Traces}\left(T S_{1}\right) \exists \sigma_{2} \in \operatorname{Traces}\left(T S_{2}\right) . \quad \sigma_{1} \triangleq \sigma_{2}
\]
stutter trace equivalence:
\(T S_{1} \triangleq T S_{2} \quad\) iff \(\quad T S_{1} \unlhd T S_{2}\) and \(\quad T S_{2} \unlhd T S_{1}\)
stutter bisimulation equivalence:
\(T S_{1} \approx T S_{2}\) iff there exists a stutter bisimulation for \(\left(T S_{1}, T S_{2}\right)\)
stutter bisimulation equivalence with divergence:
\(T S_{1} \approx{ }^{d i v} T S_{2}\) iff there exists a divergence-sensitive stutter bisimulation for \(\left(T S_{1}, T S_{2}\right)\)

Figure 7.34: Stutter implementation relations.
\(\{0, \ldots, 9\}\). It allows an erroneous digit to be entered, but this may happen at most twice. Location \(\ell_{i}\) (for \(i=0,1,2\) ) indicates that the first \(i\) digits of the code have been correctly entered.


Figure 7.35: Transition system of the door opener
The transition system in Figure 7.36 is stutter-bisimilar to the transition system in Figure 7.35 .

The quotient transition system under stutter bisimulation \(\approx_{T S}\) is defined as the quotient under bisimulation. As a stutter bisimulation is not guaranteed to be transitive, the


Figure 7.36: Stutter-bisimilar transition system for the door opener.
quotient under the equivalence \(\approx_{T S}\) is considered. Note that, of course, states are now equivalence classes under \(\approx_{T S}\) If \(s \rightarrow s^{\prime}\) and \(s \not \approx s^{\prime}\), there is a transition from \([s]_{\approx}\) to \(\left[s^{\prime}\right] \approx\). As a result, \(T S / \approx\) contains no self-loops.

\section*{Definition 7.101. Stutter Bisimulation Quotient System}

For transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\), the stutter bisimulation quotient transition system \(T S / \approx\) is defined as follows:
\[
T S / \approx=\left(S / \approx_{T S},\{\tau\}, \rightarrow \approx, I_{\approx}, A P, L_{\approx}\right)
\]
where
- \(I_{\approx}=\left\{[s]_{\approx} \mid s \in I\right\}\),
\(\rightarrow \approx\) is defined by \(\frac{s \xrightarrow{\alpha} s^{\prime} \wedge s \not \approx s^{\prime}}{[s]_{\approx} \xrightarrow{\tau} \approx\left[s^{\prime}\right]_{\approx}}\),
- \(L_{\approx}\left([s]_{\approx}\right)=L(s)\).

Theorem 7.102. Stutter Bisimulation Equivalence of TS and TS/ \(\approx\)
For any transition system \(T S\), we have \(T S \approx T S / \approx\).

Proof: Follows from the fact that \(\mathcal{R}=\left\{\left(s, s^{\prime}\right) \mid s^{\prime} \in[s]_{\approx}, s \in S\right\}\) is a stutter bisimulation for \((T S, T S / \approx)\).

\section*{Example 7.103. Semaphore-Based Mutual Exclusion}

Consider again the transition system \(T S_{S e m}\) for the semaphore-based solution to the mutual exclusion problem (see Figure 5.5 on page 239). Let \(A P=\left\{\right.\) crit \(_{1}, c^{\prime}\) crit \(\left._{2}\right\}\). Recall that \(\mathcal{R}\) inducing the following partitioning of the state space is a stutter bisimulation:
\[
\left\{\left\{\left\langle n_{1}, n_{2}\right\rangle,\left\langle n_{1}, w_{2}\right\rangle,\left\langle w_{1}, n_{2}\right\rangle,\left\langle w_{1}, w_{2}\right\rangle\right\},\left\{\left\langle c_{1}, n_{2}\right\rangle,\left\langle c_{1}, w_{2}\right\rangle\right\},\left\{\left\langle c_{2}, n_{1}\right\rangle,\left\langle w_{1}, c_{2}\right\rangle\right\}\right\} .
\]

In fact, this is the coarsest stutter bisimulation, i.e., \(\mathcal{R}\) equals \(\approx_{T S}\). It is not difficult to see that the quotient under bisimulation, i.e., \(T S_{S e m} / \sim\) does not yield any state space reduction. The quotient transition system under \(\approx\), i.e., \(T S_{S e m} / \approx\), is depicted in Figure 7.37 and does yield a substantial reduction of the state space.


Figure 7.37: The stutter bisimulation quotient system of \(T S_{\text {Sem }}\).

Let us now discuss the relationship between stutter trace equivalence and stutter bisimulation. Obviously, for transition systems that do not have any stutter steps, \(\sim\) and \(\approx\) coincide. The same applies to trace equivalence and stutter trace equivalence ( \(\triangleq\) ). Therefore, in general, stutter trace equivalent transition systems are not stutter-bisimilar. As an example, consider trace equivalent transition systems without stutter steps, which are not bisimilar. As we have seen, \(T S_{1} \sim T S_{2}\) implies that \(T S_{1}\) and \(T S_{2}\) are trace equivalent, see Theorem 7.6 (page 456). Based on this fact, one is tempted to assume that stutter bisimulation is finer than stutter trace equivalence. This is, however, not true, as stutter bisimulation does not impose any restrictions on paths that just consist of stutter steps. Only paths that "switch" equivalence class have to be matched, but pure stutter paths not - condition (2) of Definition 7.95 (page 536) only requires that, if \(s_{1}\) and \(s_{2}\) are stutterbisimilar, then every transition \(s_{1} \rightarrow s_{1}^{\prime}\) with \(s_{1} \not \approx s_{1}^{\prime}\) can be simulated by a path fragment \(s_{2} \rightarrow \ldots \rightarrow s_{2}^{\prime}\). For stutter steps \(s_{1} \rightarrow s_{1}^{\prime}\), within the stutter bisimulation equivalence class of \(s_{1}\) (i.e., \(s_{1} \approx s_{1}^{\prime}\) ), no conditions are imposed. That is, it is possible that
\[
s_{1} \approx s_{2} \quad \text { while } \quad \exists \pi_{1} \in \operatorname{Paths}\left(s_{1}\right) . \forall \pi_{2} \in \operatorname{Paths}\left(s_{2}\right) . \pi_{1} \neq \pi_{2}
\]

Theorem 7.104. Stutter Trace vs. Stutter Bisimulation Equivalence \(\triangleq\) and \(\approx\) are incomparable.

\(T S_{1}\)

\(T S_{2}\)

\(T S_{3}\)

\(T S_{4}\)

Figure 7.38: Stutter trace vs. stutter bisimulation equivalence.
Proof: see Figure 7.38. For the two leftmost transition systems \(T S_{1}\) and \(T S_{2}\), we have
\[
T S_{1} \triangleq T S_{2}, \text { but } T S_{1} \not \approx T S_{2} .
\]
(In fact, we have \(\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)\) and \(T S_{1} \nsim T S_{2}\).) For the two rightmost transition systems, we have \(T S_{3} \approx T S_{4}\), while \(T S_{3} \neq T S_{4}\). The fact that the transition systems are not stutter trace equivalent follows from the fact that \(T S_{3}\) exhibits the trace \(\varnothing^{\omega}\), whereas \(T S_{4}\) cannot generate this trace. Note that the trace \(\varnothing^{\omega}\) just consists of stutter steps.

As a consequence, stutter bisimulation does not preserve the truth value of all \(\mathrm{LTL}_{\backslash \bigcirc}\) formulae. This is illustrated by the following example.

Example 7.105. Stutter Bisimulation Equivalence vs. \(L T L_{\backslash \bigcirc}\) Equivalence
Consider the transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) in Figure 7.39. Then \(T S_{1} \approx T S_{2}\) and \(T S_{2} \vDash \diamond a\). However, \(T S_{1} \not \models \diamond a\) since the path \(s_{0}^{\omega}\) violates \(\diamond a\). This is due to the fact that the path \(s_{0}^{\omega}\) just consists of stutter steps, and by definition of \(\approx\) is not required to have a stutter-equivalent path starting in \(t_{0}\).

\subsection*{7.8.1 Divergence-Sensitive Stutter Bisimulation}

The fact that stutter trace equivalence ( \(\triangleq\) ) and stutter bisimulation \((\approx)\) are incomparable is caused by stutter paths, i.e., paths that only consists of stutter steps. The existence of


Figure 7.39: Stutter bisimulation, but not \(\mathrm{LTL}_{\backslash \bigcirc}\) equivalent systems.
these paths is also the reason that \(\approx\) does not imply LTL \(_{\backslash \bigcirc}\) equivalence. Stutter paths stay forever in an equivalence class without performing any visible step. This behavior is called divergent.

The topic of this section is to adapt stutter bisimulation such that states may only be related if they both exhibit divergent paths or none of them has a divergent path. This yields a variant of stutter bisimulation that will be shown to coincide with CTL \(_{\backslash \bigcirc}^{*}\) equivalence, i.e., equivalence with respect to CTL* formulae that do not contain the next-step operator. (Note that CTL \(_{\backslash \bigcirc}^{*}\) includes \(\mathrm{LTL}_{\backslash \bigcirc}\).) This divergence-sensitive variant of stutter bisimulation is the coarsest equivalence that preserves all CTL \(_{{ }^{*}}^{*}\) formulae. Besides, in contrast to \(\approx\), this variant is strictly finer that stutter trace equivalence.

\section*{Definition 7.106. \(\mathcal{R}\)-Divergent State, Divergence Sensitivity}

Let \(T S\) be a transition system and \(\mathcal{R}\) an equivalence relation on \(S\).
- \(s \in S\) is \(\mathcal{R}\)-divergent if there exists an infinite path fragment \(\pi=s s_{1} s_{2} \ldots \in\) \(\operatorname{Paths}(s)\) such that \(\left(s, s_{j}\right) \in \mathcal{R}\) for all \(j>0\).
- \(\mathcal{R}\) is divergence-sensitive if for any \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) : if \(s_{1}\) is \(\mathcal{R}\)-divergent, then \(s_{2}\) is \(\mathcal{R}\)-divergent.

Stated in words, a state is \(\mathcal{R}\)-divergent if there is an infinite path starting in \(s\) that only visits states in \([s]_{\mathcal{R}}\). As \(\mathcal{R}\) is an equivalence, for \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) we have \(s_{1}\) is \(\mathcal{R}\)-divergent if and only if \(s_{2}\) is \(\mathcal{R}\)-divergent. Equivalence \(\mathcal{R}\) is thus divergence-sensitive if in any \(\mathcal{R}\) equivalence class, either all states are \(\mathcal{R}\)-divergent or none is \(\mathcal{R}\)-divergent.

\section*{Example 7.107. Divergence Sensitivity}

For the transition system in Figure 7.40 , we have \(s_{0} \approx_{T S} s_{1} \approx_{T S} s_{2}\). States \(s_{0}\) and \(s_{1}\) are
\(\approx_{T S}\)-divergent as in both states it is possible to infinitely often alternate between \(s_{0}\) and \(s_{1}\), i.e., stay forever in \(\left[s_{0}\right] \approx\). As state \(s_{2}\) is not \(\approx_{T S}\)-divergent, it follows that \(\approx_{T S}\) is not divergence-sensitive. Equivalence \(\mathcal{R}=\left\{\left(s_{0}, s_{1}\right),\left(s_{1}, s_{0}\right)\right\} \cup I d\) is divergence-sensitive.


Figure 7.40: \(s_{0}\) and \(s_{1}\) are \(\approx_{T S}\)-divergent, but \(s_{2}\) is not.

\section*{Example 7.108. Alternating Bit Protocol}

Consider the alternating bit protocol as introduced in Chapter 2; see Example 2.32 on page 57 . Recall that messages from the sender to the receiver are equipped with a bit that toggles between transmissions of new messages and stays constant on retransmissions. Accordingly, the sender can be either in a " 0 -mode" to transmit a message with bit 0 , or in the "1-mode". Similarly, the receiver can be either expecting a message with bit 0 or with bit 1. Let
\[
A P=\left\{s \_m o d e=0, s_{-} m o d e=1, r_{-} \text {mode }=0, r_{-} \text {mode }=1\right\}
\]
where s_mode indicates the mode of the sender, and r_mode that of the receiver. The state space of the underlying \(T S_{A B P}\) of the alternating bit protocol consists of hundreds of states for a channel capacity of one (for both channels). The stutter bisimulation quotient just consists of four states; see Figure 7.41 where the state labeling \((i, j)\) is a shorthand for s_mode \(=i\) and r_mode \(=j, i, j \in\{0,1\}\). By definition, no state in \(T S_{A B P} / \approx\)


Figure 7.41: Stutter bisimulation quotient of the alternating bit protocol.
is \(\approx\)-divergent. However, \(T S_{A B P}\) contains (various) states that are \(\approx\)-divergent, as in these states it is possible to continuously lose a message. For instance, in states in the equivalence class \(\{(0,0)\}\) it is possible that the transmitted message with bit 0 as sent by the sender is lost infinitely often. Thus:
\[
T S_{A B P} \not \vDash \forall \square \diamond(\text { s_mode }=0) \wedge \quad \forall \square \diamond(\text { s_mode }=1),
\]
but
\[
T S_{A B P} / \approx \models \forall \square \diamond(\text { s_mode }=0) \wedge \forall \square \diamond(\text { s_mode }=1) .
\]

This confirms our earlier observation that \(\approx\) does not guarantee the preservation of \(\mathrm{LTL}_{\backslash \bigcirc}\) formulae.

\section*{Definition 7.109. Stutter Bisimulation with Divergence}

States \(s_{1}, s_{2}\) in transition system \(T S\) are divergent stutter bisimilar, denoted \(s_{1} \approx_{T S}^{d i v} s_{2}\), if there exists a divergence-sensitive stutter bisimulation \(\mathcal{R}\) on \(T S\) such that \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).

It is not difficult to show that \(\approx_{T S}^{d i v}\) is an equivalence on \(S\), and the coarsest divergencesensitive stutter bisimulation for \(T S\) which is obtained by the union of all divergencesensitive stutter bisimulations for \(T S\).

\section*{Example 7.110. Stutter Bisimulation with Divergence}

Consider the transition system \(T S\) in Figure 7.40 (page 545). Its equivalence classes under \(\approx_{T S}^{d i v}\) are \(\left\{s_{0}, s_{1}\right\},\left\{s_{2}\right\}\) and \(\left\{s_{3}\right\}\). State \(s_{2}\) is \(\approx_{T S}^{d i v}\)-divergent whereas \(s_{0}\) and \(s_{1}\) are not. Thus, \(s_{2} \not \not_{T S}^{d i v} s_{0}\) and \(s_{2} \not \nsim T S_{d i v}^{d} s_{1}\).

As a second example, consider the transition system in Figure 7.42 where the state labeling is indicated by the grey scale. \(s_{1} \not \overbrace{T S}^{d i v} s_{2}\) as \(u_{3}\) and \(u_{1}, u_{2}\) are not equivalent under \(\approx_{T S}^{d i v}\) (since only \(u_{3}\) is divergent) and only \(s_{2}\) (but not \(s_{1}\) ) has a transition to \(u_{3}\).


Figure 7.42: \(s_{1} \not \overbrace{T S}^{d i v} s_{2}\).
\(T S_{1} \approx^{\text {div }} T S_{2}\) if and only if each initial state of \(T S_{1}\) is divergence stutter bisimilar (according to \(\approx_{T S_{1} \oplus T S_{2}}^{d i v}\) ) to an initial state in \(T S_{2}\), and vice versa. The quotient transition
system with respect to stutter bisimulation \(\approx_{T S}^{d i v}\) is defined as usual with equivalence classes under \(\approx_{T S}^{d i v}\) as states. In addition to the usual transitions, every equivalence class \(C\) consisting of divergent states is equipped with a self-loop. This self-loop indicates the divergence. (Recall that \(T S / \mathcal{R}\) for stutter bisimulation \(\mathcal{R}\) does not contain any self-loop.) That is, the transition relation of \(T S / \approx^{d i v}\) is defined by
\[
\frac{s \stackrel{\alpha}{\longrightarrow} s^{\prime} \wedge s \not \not^{d i v} s^{\prime}}{[s]_{d i v} \xrightarrow{\tau}{ }_{d i v}\left[s^{\prime}\right]_{d i v}} \quad \text { and } \quad \frac{s \text { is } \approx^{\text {div -divergent }}}{[s]_{\text {div }} \xrightarrow{\tau}{ }_{\text {div }}[s]_{\text {div }}}
\]
where \([s]_{\text {div }}\) denotes the equivalence class of \(s\) under \(\approx{ }^{d i v}\).


transition system \(T S / \approx\)

transition system \(T S / \approx^{d i v}\)

Figure 7.43: Transition system \(T S, T S / \approx\), and \(T S / \approx^{d i v}\).

\section*{Example 7.111. Quotient Transition System under \(\approx\) and \(\approx\) div}

Consider the transition system \(T S\) in Figure 7.43 (upper part). Its quotient under stutter bisimulation, \(T S / \approx\), is depicted in the left lower part of the figure. States \(s_{0}, s_{1}\), and \(s_{2}\) are all stutter-bisimilar. Note that, by definition, no self-loops occur in \(T S / \approx\); only the transitions that change equivalence class are of importance. \(T S / \approx^{d i v}\) is depicted in the right lowerpart of Figure 7.43. State \(s_{2}\) is not equivalent to \(s_{0}\) and \(s_{1}\) as it is not divergent, whereas \(s_{0}\) and \(s_{1}\) are divergent.

Theorem 7.112. TS and TS/ \(\approx^{\text {div }}\) are Divergence Stutter Bisimilar
For any transition system TS, we have TS \(\approx^{d i v} T S / \approx^{d i v}\).

Proof: Follows from the fact that \(\mathcal{R}=\left\{\left(s,[s]_{d i v}\right) \mid s \in S\right\}\) is a divergence-sensitive stutter bisimulation for \(\left(T S, T S / \approx^{d i v}\right)\).

Since divergence-sensitive stutter bisimulations are special cases of arbitrary stutter bisimulations, stutter bisimulation with divergence \(\approx^{d i v}\) is strictly finer than stutter bisimulation \(\approx\). In the example of Figure 7.40 on page \(545, s_{0} \approx_{T S} s_{2}\) but \(s_{0} \not \not_{T S}^{d i v} s_{2}\).

Lemma 7.113. \(\approx^{\text {div }}\) is Strictly Finer Than \(\approx\)
For transition systems \(T S_{1}\) and \(T S_{2}\) :

while the reverse does not hold in general.

Remark 7.114. Terminal and Purely Divergent States
State \(s\) is purely divergent if all paths starting in \(s\) are infinite and completely consist of stutter steps. Since \(\approx\) imposes no restrictions on transitions inside equivalence classes, any purely divergent state is stutter-bisimilar to any equally labeled state which is terminal. However, \(\approx{ }^{d i v}\) distinguishes terminal states and purely divergent states. Thus,
\[
s_{t} \approx_{T S} s_{p d}, \quad \text { while } s_{t} \not \nsim T S_{d i v}^{d i v} s_{p d}
\]
where \(s_{t}\) is a terminal state and \(s_{p d}\) a purely divergent state and \(L\left(s_{t}\right)=L\left(s_{p d}\right)\).
Both \(\approx_{T S}\) and \(\approx_{T S}^{d i v}\) identify any terminal state \(s_{t}\) with any state \(s\) with \(L(s)=L\left(s_{t}\right)\) and for which \(\operatorname{Paths}(s)\) consists of finite stutter paths only. In other words, all reachable states from \(s\) have the same labeling as \(s\) and there is no infinite path starting in \(s\). In fact, there are no other states that are divergence stutter bisimilar to \(s_{t}\). That is, if \(s_{t}\) is terminal and \(s_{t} \approx_{T S}^{d i v} s\), then \(L(s)=L\left(s_{t}\right)\) and each path of \(s\) is finite and only consists of stutter steps.

\section*{Remark 7.115. Divergence-Sensitive Transition Systems}

Transition system \(T S\) is called divergence-sensitive if \(\approx_{T S}\) on \(T S\) is divergence-sensitive. Clearly, for divergence-sensitive transition systems, \(\approx_{T S}\) corresponds to \(\approx_{T S}^{d i v}\). For instance, the transition system shown in Figure 7.40 on page 545 is not divergence-sensitive.

Recall that \(\triangleq\) (stutter trace equivalence) and \(\approx\) (stutter bisimulation) are incomparable, see Theorem 7.104 (page 543). This is due to the fact that \(\approx\) ignores divergent paths, whereas \(\triangleq\) does not. In the sequel, it will be shown that \(\approx^{d i v}\) is strictly finer than \(\triangleq\). The proof technique is rather similar to proving that bisimulation is finer than trace equivalence and is based on a lifting of \(\approx^{d i v}\) to paths.

\section*{Definition 7.116. Divergence Stutter-Bisimilar Paths}

Let TS be a transition system.
1. For infinite path fragments \(\pi_{i}=s_{0, i} s_{1, i} s_{2, i} \ldots, i=1,2\), in \(T S\) :
\[
\pi_{1} \approx_{T S}^{d i v} \pi_{2}
\]
if and only if there exists an infinite sequence of indices \(0=j_{0}<j_{1}<j_{2}<\ldots\) and \(0=k_{0}<k_{1}<k_{2}<\ldots\) with:
\[
s_{j, 1} \approx_{T S}^{d i v} s_{k, 2} \text { for all } j_{r-1} \leqslant j<j_{r} \text { and } k_{r-1} \leqslant k<k_{r} \text { with } r=1,2, \ldots
\]
2. For finite paths fragments \(\widehat{\pi}_{1}=s_{0, i} s_{1, i} \ldots s_{K_{i}, i}, i=1,2\), in \(T S\) :
\[
\widehat{\pi}_{1} \approx_{T S}^{d i v} \widehat{\pi}_{2}
\]
if and only if there exists a finite sequence of indices \(0=j_{0}<j_{1}<\ldots<j_{\ell}=K_{1}+1\) and \(0=k_{0}<k_{1}<\ldots<k_{\ell}=K_{2}+1\) with \(s_{j, 1} \approx_{T S}^{d i v} s_{k, 2}\) for all \(j_{r-1} \leqslant j<j_{r}\) and \(k_{r-1} \leqslant k<k_{r}\) with \(r=1,2, \ldots, \ell\).

Paths \(\pi_{1}\) and \(\pi_{2}\) are stutter-bisimilar with divergence if both paths can be divided into segments \(s_{j_{r}, 1} s_{j_{r}+1,1} s_{j_{r}+2,1} \ldots s_{j_{r+1}, 1}\) and \(s_{k_{r}, 2} s_{k_{r}+1,2} s_{k_{r}+2,2} \ldots s_{k_{r+1}, 2}\), respectively, that are statewise bisimilar (under \(\approx^{d i v}\) ).

The following lemma follows directly from the definition of \(\approx^{d i v}\) on paths and \(\triangleq\) on paths; see Definition 7.86 on page 530 .

\section*{Lemma 7.117. Equivalences \(\approx{ }^{\text {div }}\) and \(\triangleq\) for Paths}

For all infinite paths \(\pi_{1}\) and \(\pi_{2}\), we have \(\pi_{1} \approx_{T S}^{\text {div }} \pi_{2}\) implies \(\pi_{1} \triangleq{ }_{T S} \pi_{2}\).

Divergence stutter bisimilar states have divergence stutter bisimilar paths:

\section*{Lemma 7.118. Path Lifting for Divergence Stutter Bisimilar States}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system, \(s_{1}, s_{2} \in S\). Then:
\[
s_{1} \approx_{T S}^{d i v} s_{2} \quad \text { implies } \quad \forall \pi_{1} \in \operatorname{Paths}\left(s_{1}\right) \cdot\left(\exists \pi_{2} \in \operatorname{Paths}\left(s_{2}\right) \cdot \pi_{1} \approx_{T S}^{d i v} \pi_{2}\right)
\]

Proof: Let \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots \in \operatorname{Paths}\left(s_{1}\right)\). The proof technique is to successively define a statewise stutter-bisimilar path \(\pi_{2}\) starting in \(s_{2}\) by "lifting" the transitions \(s_{i, 1} \rightarrow s_{i+1,1}\) with \(s_{i, 1} \not \boldsymbol{\psi}_{T S}^{d i v} s_{i+1,1}\) to finite path fragments \(s_{i, 2} u_{i, 1} \ldots u_{i, n_{i}} s_{i+1,2}\) such that \(s_{i+1,1} \approx_{T S}^{d i v}\) \(s_{i+1,2}\) and \(s_{i, 2} \approx_{T S}^{d i v} u_{i, 1} \approx_{T S}^{d i v} \ldots \approx_{T S}^{d i v} u_{i, n_{i}}\).

The proof is by induction on \(i\). The base case \(i=0\) is straightforward and omitted. Assume \(i \geqslant 0\) and that the path fragment:
\[
s_{2}=s_{0,2} u_{0,1}, \ldots, u_{0, n_{0}} s_{1,2} u_{1,1} \ldots u_{1, n_{1}} s_{2,2} \ldots s_{i, 2}(*)
\]
is already constructed. In particular, \(s_{i, 1} \approx_{T S}^{d i v} s_{i, 2}\). If \(s_{i, 1}\) is a terminal state, then there exists a finite path fragment \(s_{i, 2} v_{1} \ldots v_{m}\) consisting of stutter steps such that \(v_{m}\) is a terminal (see Remark 7.114 on page 548). Hence, the path \(\pi_{2}\) resulting by concatenating the above path fragment \((*)\) from \(s_{2}\) to \(s_{i, 2}\) and the path fragment \(s_{i, 2} v_{1} \ldots v_{m}\) fulfills the desired conditions. In the sequel, we assume that \(s_{i, 1}\) is not a terminal state, and, hence, \(\pi_{1}\) does not end in state \(s_{i, 1}\). Distinguish two cases:
1. \(s_{i, 1} \not \ddot{z}_{T S}^{d i v} s_{i+1,1}\). Since \(s_{i, 1} \approx_{T S}^{d i v} s_{i, 2}\) and \(s_{i, 1} \rightarrow s_{i+1,1}\), there exists a finite path fragment \(s_{i, 2} u_{i, 1} \ldots u_{i, n_{i}} s_{i+1,2}\) such that:
\[
s_{i+1,1} \approx_{T S}^{d i v} s_{i+1,2} \quad \text { and } \quad s_{i, 2} \approx_{T S}^{d i v} u_{i, 1} \approx_{T S}^{d i v} \ldots \approx_{T S}^{d i v} u_{i, n_{i}}
\]

Concatenating the path \((*)\) with the path fragment \(s_{i, 2} u_{i, 1} \ldots u_{i, n_{i}} s_{i+1,2}\) yields a path fragment that fulfills the desired conditions.
2. \(s_{i, 1} \approx_{T S}^{d i v} s_{i+1,1}\). Distinguish between \(s_{i, 1}\) is divergent and not divergent:
(a) \(s_{i, 1}\) is not divergent, i.e, there exists an index \(j>i+1\) with \(s_{i, 1} \not \nsim T S_{d i v}^{d i} s_{j, 1}\). Without loss of generality, assume that \(j\) is minimal, i.e., \(s_{i, 1}, s_{i+1,1}, \ldots, s_{j-1,1}\) are pairwise equivalent under \(\approx^{d i v}\). In particular,
\[
s_{i, 2} \approx_{T S}^{d i v} s_{j-1,1} \quad \text { and } \quad s_{j-1,1} \quad \not \overbrace{T S}^{d i v} \quad s_{j, 1}
\]

As \(s_{i, 2} \approx_{T S}^{d i v} s_{j-1,1}\) and \(s_{j-1,1} \rightarrow s_{j, 1}\), there exists a finite path fragment \(s_{i, 2} u_{i, 1} \ldots u_{i, n_{i}} s_{i+1,2}\) such that
\[
s_{j, 1} \approx_{T S}^{d i v} \quad s_{i+1,2} \quad \text { and } \quad s_{i, 2} \approx_{T S}^{d i v} u_{i, 1} \approx_{T S}^{d i v} \ldots \approx_{T S}^{d i v} u_{i, n_{i}}
\]

Concatenation of the path fragment \((*)\) with \(s_{i, 2} u_{i, 1} \ldots u_{i, n_{i}} s_{i+1,2}\) yields a path that fulfills the desired conditions.


Figure 7.44: Relation between weak equivalences and preorders on transition systems.
(b) \(s_{i, 1}\) is divergent, i.e., \(s_{i, 1} \approx_{T S}^{d i v} s_{j, 1}\) for all \(j \geqslant i\). As \(s_{i, 1} \approx_{T S}^{d i v} s_{i, 2}\), and \(s_{i, 1}\) is divergent, \(s_{i, 2}\) is divergent. That is, there is a path \(s_{i, 2} s_{i+1,2} s_{i+2,2} \ldots\) with \(s_{i, 2} \approx_{T S}^{d i v} s_{i+1,2} \approx_{T S}^{d i v} s_{i+2,2} \approx_{T S}^{d i v} \ldots\). Concatenation of the path fragment \((*)\) with \(s_{i, 2} s_{i+1,2} s_{i+2,2} \ldots\) yields a path that fulfills the desired conditions.

The thus resulting path fragment \(\pi_{2}\) is divergence stutter bisimilar to \(\pi_{1}\).

The fact that \(\approx^{d i v}\) can be lifted from states to paths enables to establish that \(\approx^{d i v}\) on transition systems is strictly finer than \(\triangleq\), i.e., stutter trace-equivalence.

\section*{Theorem 7.119. Stutter Trace vs. Divergence Stutter Bisimulation}

Let \(T S_{1}\) and \(T S_{2}\) be transition systems over AP. Then:
\[
\underbrace{\text { with divergence }}_{\text {stutter bisimulation equivalence }} ⿺ 辶 \mathrm{TS}_{1} \approx \operatorname{div} T S_{2} \quad \text { implies } \underbrace{T S_{1} \triangleq}_{\text {stutter trace equivalence }}
\]
whereas the reverse implication does not hold in general.

Proof: Assume \(T S_{1} \approx{ }^{d i v} T S_{2}\). Consider the transition system \(T S_{1} \oplus T S_{2}\), and let \(\left(s_{1}, s_{2}\right)\) be a pair of initial states of \(T S_{1}\) and \(T S_{2}\) such that \(s_{1} \approx_{T S}^{d i v} s_{2}\). From Lemma 7.118, it follows that for every \(\pi_{1} \in \operatorname{Paths}\left(s_{1}\right)\), there exists \(\pi_{2} \in \operatorname{Paths}\left(s_{2}\right)\) such that \(\pi_{1} \approx_{T S}^{d i v} \pi_{2}\). By Lemma 7.117, \(\pi_{1} \triangleq \pi_{2}\). Thus, \(T S_{1} \triangleq T S_{2}\). The reverse direction does not apply in general. Let \(T S_{1}, T S_{2}\) be trace equivalent but not bisimilar, and neither contains any stutter steps. The absence of stutter steps yields \(T S_{1} \triangleq T S_{2}\), but \(T S_{1} \not \approx^{d i v} T S_{2}\).

The relationship between the various stutter equivalences and stutter trace preorder relations are summarized in Figure 7.44. An arrow from relation \(\mathcal{R}\) to \(\mathcal{R}^{\prime}\) denotes that \(\mathcal{R}\) is strictly finer than \(\mathcal{R}^{\prime}\). Note that for \(A P\)-deterministic transition systems, trace equivalence coincides with bisimulation, and stutter trace equivalence coincides with divergencesensitive stutter bisimulation. The latter fact is stated without proof, and is left as an exercise for the reader.

\subsection*{7.8.2 Normed Bisimulation}

This section introduces the notion of normed bisimulation. These bisimulations are defined using so-called norm functions. Normed bisimulation turns out to be strictly finer than \(\approx^{d i v}\). As norm functions allow establishing a bisimulation by just local reasoning about the states - only considering their direct successors - means that it is often simpler to establish a normed bisimulation than a divergence-sensitive stutter bisimulation. We introduce a norm function which yields a sufficient (but not necessary) requirement for \(\approx^{d i v}\). This will be used in Chapter 8.

\section*{Definition 7.120. Normed (Bi)Simulation}

Let \(T S_{1}=\left(S_{1}, A c t_{1}, \rightarrow_{1}, I_{1}, A P, L_{1}\right)\) and \(T S_{2}=\left(S_{2}, A c t_{2}, \rightarrow_{2}, I_{2}, A P, L_{2}\right)\) be transition systems over \(A P\).

A normed simulation for \(\left(T S_{1}, T S_{2}\right)\) is a triple ( \(\left.\mathcal{R}, \nu_{1}, \nu_{2}\right)\) consisting of a binary relation \(\mathcal{R} \subseteq S_{1} \times S_{2}\) such that:
\[
\forall s_{1} \in I_{1} \cdot \exists s_{2} \in I_{2} \cdot\left(s_{1}, s_{2}\right) \in \mathcal{R}
\]
and functions \(\nu_{1}, \nu_{2}: S_{1} \times S_{2} \rightarrow \mathbb{N}\) such that for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) :
(NI) \(L_{1}\left(s_{1}\right)=L_{2}\left(s_{2}\right)\).
(NII) For all \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\), at least one of the following three conditions holds:
(N1) There exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
(N2) \(\left(s_{1}^{\prime}, s_{2}\right) \in \mathcal{R}\) and \(\nu_{1}\left(s_{1}^{\prime}, s_{2}\right)<\nu_{1}\left(s_{1}, s_{2}\right)\).
(N3) There exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}, s_{2}^{\prime}\right) \in \mathcal{R}\) and \(\nu_{2}\left(s_{1}, s_{2}^{\prime}\right)<\nu_{2}\left(s_{1}, s_{2}\right)\).

A normed bisimulation for \(\left(T S_{1}, T S_{2}\right)\) is a normed simulation \(\left(\mathcal{R}, \nu_{1}, \nu_{2}\right)\) for \(\left(T S_{1}, T S_{2}\right)\) such that \(\left(\mathcal{R}^{-1}, \nu_{2}^{-}, \nu_{1}^{-}\right)\)is a normed simulation for \(\left(T S_{2}, T S_{1}\right)\). Here, \(\nu_{i}^{-}\)denotes the function \(S_{2} \times S_{1} \rightarrow \mathbb{N}\) that results from \(\nu_{i}\) by swapping the arguments, i.e., \(\nu_{i}^{-}(u, v)=\) \(\nu_{i}(v, u)\) for all \(u \in S_{2}\) and \(v \in S_{1}\).
\(T S_{1}\) and \(T S_{2}\) are normed-bisimilar, denoted \(T S_{1} \approx^{n} T S_{2}\), if there exists a normed bisimulation for \(\left(T S_{1}, T S_{2}\right)\).

For transition system \(T S\), a normed bisimulation for \(T S\) is a normed bisimulation for \((T S, T S)\). States \(s_{1}\) and \(s_{2}\) of \(T S\) are called normed-bisimilar, denoted \(s_{1} \approx_{T S}^{n} s_{2}\), if there exists a normed bisimulation \(\left(\mathcal{R}, \nu_{1}, \nu_{2}\right)\) for \(T S\) such that \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).
\(\nu_{1}\) and \(\nu_{2}\) are norm functions. The values \(\nu_{i}\left(s_{1}, s_{2}\right)\) for \(\left(s_{1}, s_{2}\right) \notin \mathcal{R}\) are irrelevant. Hence, \(\nu_{i}\) could have been defined as function \(\mathcal{R} \rightarrow \mathbb{N}\). The intuitive meaning of \(\nu_{1}\left(s_{1}, s_{2}\right)\) for \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) is as follows. \(\nu_{1}\left(s_{1}, s_{2}\right)\) serves as a count down for the number "allowed" stutter steps from \(s_{1}\) that cannot be mimicked by transitions of \(s_{2}\). Similarly, \(\nu_{2}\left(s_{1}, s_{2}\right)\) can be regarded as a counter for the number of stutter steps that \(s_{2}\) may perform to reach a state where the visible (nonstutter) steps of \(s_{1}\) can be simulated.

\section*{Example 7.121. Normed Bisimulation Equivalence}

Consider the transition system TS in Figure 7.40 on page 545. The coarsest equivalence \(\mathcal{R}\) which identifies \(s_{0}\) and \(s_{1}\) together with \(\nu_{1}\left(s_{0}, s_{1}\right)=\nu_{2}\left(s_{1}, s_{0}\right)=1\) and \(\nu_{2}\left(s_{1}, s_{2}\right)=1\), \(\nu_{1}(s, s)=\nu_{2}(s, s)=0\) for all \(s \in\left\{s_{0}, s_{1}, s_{2}, s_{3}\right\}\) (and arbitrary values for \(\nu_{1}\) and \(\nu_{2}\) for the remaining cases) is a normed bisimulation. This is checked as follows. It suffices to only consider \(\left(s_{0}, s_{1}\right),\left(s_{1}, s_{0}\right) \in \mathcal{R}\) and their outgoing transitions.
- For \(s_{0} \rightarrow s_{1}\), case (N1) applies as \(s_{0} \rightarrow s_{1}\) and \(\left(s_{0}, s_{1}\right) \in \mathcal{R}\).
- For \(s_{1} \rightarrow s_{0}\), case (N1) applies as \(s_{1} \rightarrow s_{0}\) and \(\left(s_{0}, s_{1}\right) \in \mathcal{R}\).
- For \(s_{0} \rightarrow s_{2}\), case (N3) applies as \(s_{1} \rightarrow s_{0},\left(s_{0}, s_{0}\right) \in \mathcal{R}\) and \(0=\nu_{2}\left(s_{0}, s_{0}\right)<\) \(\nu_{2}\left(s_{0}, s_{1}\right)=1\).
- For \(s_{1} \rightarrow s_{3}\), case (N3) applies as \(s_{0} \rightarrow s_{1},\left(s_{1}, s_{1}\right) \in \mathcal{R}\) and \(0=\nu_{2}\left(s_{1}, s_{1}\right)<\) \(\nu_{2}\left(s_{1}, s_{0}\right)\).

Thus, \(s_{0} \approx_{T S}^{n} s_{1}\).
We have \(s_{1} \not \ddot{\chi}_{T S}^{n} s_{2}\). This can be seen as follows. Assume there is normed bisimulation \(\left(\mathcal{R}, \nu_{1}, \nu_{2}\right)\) for \(T S\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\). Neither case (N1) nor (N3) applies to \(s_{1} \rightarrow s_{0}\) since \(L\left(s_{3}\right) \neq L\left(s_{0}\right)\), and thus \(s_{3} \not \nsim T S_{n}^{n} s_{0}\). In order for case (N2) to apply to \(s_{1} \rightarrow s_{0}\) and \(s_{2} \rightarrow s_{3}\), we should have
\[
s_{0} \approx_{T S}^{n} s_{2} \quad \text { and } \nu_{1}\left(s_{0}, s_{2}\right)<\nu_{1}\left(s_{1}, s_{2}\right)
\]

But for \(s_{0} \approx_{T S}^{n} s_{2}\), neither case ( N 1 ) nor ( N 3 ) applies to \(s_{0} \rightarrow s_{1}\), as \(s_{2}\) can only move to a state that is not labeled with \(\{a\}\). Thus, case (N2) applies to \(s_{0} \rightarrow s_{1}\). This yields
\[
\nu_{1}\left(s_{1}, s_{2}\right)<\nu_{1}\left(s_{0}, s_{2}\right)
\]
and contradicts the condition \(\nu_{1}\left(s_{0}, s_{2}\right)<\nu_{1}\left(s_{1}, s_{2}\right)\).

It is not difficult to prove that \(\approx_{T S}^{n}\) is an equivalence. Moreover, \(\approx_{T S}^{n}\) can be equipped with norm functions \(\nu_{1}, \nu_{2}\) such that \(\left(\approx_{T S}^{n}, \nu_{1}, \nu_{2}\right)\) is a normed bisimulation; see Exercise 7.24.

Lemma 7.122. \(\approx_{T S}^{n}\) is Finer Than \(\approx_{T S}^{d i v}\)
If \(s_{1}, s_{2}\) are states in a transition system \(T S\), then \(s_{1} \approx_{T S}^{n} s_{2}\) implies \(s_{1} \approx_{T S}^{d i v} s_{2}\).

Proof: Let \(\left(\mathcal{R}, \nu_{1}, \nu_{2}\right)\) be a normed bisimulation. We show that the symmetric, reflexive, and transitive closure of \(\mathcal{R}\) - the coarsest equivalence that contains \(\mathcal{R}\)-is a divergencesensitive stutter bisimulation. This amounts showing that \(\mathcal{R}\) fulfills the conditions of a stutter bisimulation and that \(\mathcal{R}\) is divergence-sensitive.
1. As normed bisimulation requires states to be equally labeled, it suffices to first prove condition (2) of a stutter bisimulation (see Definition 7.95 on page 536).
Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\left(s_{1}^{\prime}, s_{2}\right) \notin \mathcal{R}\). It is to be proven that there is a path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\) and \(\left(s_{2}, u_{j}\right) \in \mathcal{R}\) for \(j \geqslant 0\). By (NII), one of the conditions (N1), (N2) or (N3) holds. The assumption \(\left(s_{1}^{\prime}, s_{2}\right) \notin \mathcal{R}\) rules out case (N2). If (N1) holds then the claim directly follows. Assume case (N3) applies. Then there exists \(u_{1} \in \operatorname{Post}\left(s_{2}\right)\) with
\[
\left(s_{1}, u_{1}\right) \in \mathcal{R} \text { and } \nu_{2}\left(s_{1}, u_{1}\right)<\nu_{2}\left(s_{1}, s_{2}\right)
\]

For state \(u_{1}\) the same argument as for \(s_{2}\) applies: only the cases ( N 1 ) and (N3) could apply. If (N1) holds, then there exists \(s_{2}^{\prime} \in \operatorname{Post}\left(u_{1}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\) and it follows that \(s_{2} u_{1} s_{2}^{\prime}\) is a path that satisfies the necessary requirements. Assume case (N3) applies. Then there exists \(u_{2} \in \operatorname{Post}\left(u_{1}\right)\) with
\[
\left(s_{1}, u_{2}\right) \in \mathcal{R} \text { and } \nu_{2}\left(s_{1}, u_{2}\right)<\nu_{2}\left(s_{1}, u_{1}\right)
\]

The reasoning of \(u_{1}\) can now be applied to \(u_{2}\), and if applicable \(u_{3}, u_{4}\), and so on: if (N1) applies, the claim directly follows; otherwise (N3) applies. Assume that in this way we obtain a path fragment \(s_{2} u_{1} u_{2} \ldots u_{n}\) with \(\left(s_{1}, u_{i}\right) \in \mathcal{R}\), for \(0<i \leqslant n\) and
\[
0 \leqslant \nu_{2}\left(s_{1}, u_{n}\right)<\nu_{2}\left(s_{1}, u_{n-1}\right)<\ldots<\nu_{2}\left(s_{1}, u_{1}\right)<\nu_{2}\left(s_{1}, s_{2}\right)
\]

Since the values of \(\nu_{2}\) are natural numbers, case (N3) can only apply finitely many times. Assume (N3) does not apply to \(\left(s_{1}, u_{n}\right)\). Then (N1) holds for \(\left(s_{1}, u_{n}\right)\) and transition \(s_{1} \rightarrow s_{1}^{\prime}\) for some \(n \leqslant \nu_{2}\left(s_{1}, s_{2}\right)\). Hence, there exists a transition \(u_{n} \rightarrow s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\). This yields a path fragment \(s_{2} u_{1} u_{2} \ldots u_{n} s_{2}^{\prime}\) with \(\left(s_{1}, u_{i}\right) \in \mathcal{R}\), and as \(\mathcal{R}\) is an equivalence with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\), we have
\[
\left(s_{2}, u_{i}\right) \in \mathcal{R} \text { for } 0<i \leqslant n \text { and }\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R},
\]
as required in condition (2) of stutter bisimulations.
2. The divergence sensitivity of \(\mathcal{R}\) is shown as follows. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and assume \(s_{1}\) is \(\mathcal{R}\)-divergent. We have to prove that \(s_{2}\) is \(\mathcal{R}\)-divergent. This is done by showing the existence of a transition \(s_{2} \rightarrow v_{1}\) such that \(\left(s_{2}, v_{1}\right) \in \mathcal{R}\). We then may repeat this argument for \(\left(s_{1}, v_{1}\right)\). This yields the existence of a transition \(v_{1} \rightarrow v_{2}\) with \(\left(s_{1}, v_{2}\right) \in \mathcal{R}\). By repeated application, we obtain an infinite path \(s_{2} v_{1} v_{2} v_{3} \ldots\) with \(\left(s_{1}, v_{i}\right) \in \mathcal{R}\) for all \(i>0\). Thus \(s_{2}\) is \(\mathcal{R}\)-divergent.
Let us now consider the proof. Since \(s_{1}\) is \(\mathcal{R}\)-divergent, there is an infinite path \(u_{0} u_{1} u_{2} u_{3} \ldots\) starting in \(s_{1}=u_{0}\) for which \(\left(s_{1}, u_{0}\right) \in \mathcal{R}\) for all \(i \geqslant 0\). If case (N2) applies to \(\left(u_{i}, s_{2}\right) \in \mathcal{R}\) and transitions \(u_{i} \rightarrow u_{i+1}, i=0,1, \ldots, k\), then
\[
\nu_{1}\left(s_{1}, s_{2}\right)=\nu_{1}\left(u_{0}, s_{2}\right)>\nu_{1}\left(u_{1}, s_{2}\right)>\nu_{1}\left(u_{2}, s_{2}\right)>\ldots>\nu_{1}\left(u_{k}, s_{2}\right) \geqslant 0
\]

Let \(k_{1} \geqslant 0\) be the smallest index such that case (N2) does not apply to \(\left(u_{k_{1}}, s_{2}\right) \in\) \(\mathcal{R}\) and transition \(u_{k_{1}} \rightarrow u_{k_{1}+1}\). If case (N1) holds, then there exists a transition \(s_{2} \rightarrow v_{1}\) where \(\left(u_{k_{1}+1}, v_{1}\right) \in \mathcal{R}\) and hence \(\left(s_{2}, v_{1}\right) \in \mathcal{R}\). If case ( N 3 ) holds for \(\left(u_{k_{1}}, s_{2}\right) \in \mathcal{R}\) and \(u_{k_{1}} \rightarrow u_{k_{1}+1}\), then there exists \(v_{1} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(u_{k_{1}}, v_{1}\right) \in \mathcal{R}\) and \(\nu_{2}\left(u_{k_{1}}, v_{1}\right)<\nu_{2}\left(u_{k_{1}}, s_{2}\right)\). Thus, \(s_{2} \rightarrow v_{1}\) and \(\left(s_{2}, v_{1}\right) \in \mathcal{R}\).

Lemma 7.122 can be rewritten for pairs of transition systems as follows:

Corollary 7.123. Normed and Stutter Bisimulation with Divergence
For all transition systems \(T S_{1}\) and \(T S_{2}\) over \(A P\), we have:
\[
T S_{1} \approx^{n} T S_{2} \text { implies } T S_{1} \approx^{d i v} T S_{2}
\]

The reverse of Corollary 7.123 does not hold; see Exercise 7.25.

To obtain a sufficient and necessary local norm function criterion for stutter bisimulation equivalence with divergence, a more complex notion of norm functions is needed which also depends on the transitions to be simulated. We concentrate here on the case of finite transition systems where the natural numbers can serve as a range for the norm functions. (To treat transition systems with infinitely many states, one has to deal with well-founded sets of arbitrary cardinality.)

\section*{Definition 7.124. Step-Dependent Normed Bisimulation}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system. A step-dependent normed bisimulation for \(T S\) is a pair \((\mathcal{R}, \nu)\) where \(\mathcal{R}\) is an equivalence on \(S\) and \(\nu: S \times S \times S \rightarrow \mathbb{N}\) a function such that for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) :
(SI) \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
(SII) For all \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\), at least one of the following three conditions holds:
(S1) There exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
(S2) \(\left(s_{1}^{\prime}, s_{2}\right) \in \mathcal{R}\) and \(\nu\left(s_{1}^{\prime}, s_{1}^{\prime}, s_{2}\right)<\nu\left(s_{1}, s_{1}, s_{2}\right)\).
(S3) There exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}, s_{2}^{\prime}\right) \in \mathcal{R}\) and \(\nu\left(s_{1}^{\prime}, s_{1}, s_{2}^{\prime}\right)<\nu\left(s_{1}^{\prime}, s_{1}, s_{2}\right)\).

States \(s_{1}\) and \(s_{2}\) are step-dependent normed bisimilar, denoted \(s_{1} \approx_{T S}^{s} s_{2}\), if there exists a step-dependent normed bisimulation \((\mathcal{R}, \nu)\) for \(T S\) such that \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).

For \(\nu\left(s_{1}^{\prime}, s_{1}, s_{2}\right)\), the last two arguments denote a pair \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) while the \(s_{1}^{\prime}\) stands either for a direct successor of \(s_{1}\) or \(s_{1}\) itself. If \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right), \nu\left(s_{1}^{\prime}, s_{1}, s_{2}\right)\) denotes an upper bound for the number of stutter steps \(s_{2} u_{1} \ldots u_{n}\) inside \(\left[s_{2}\right]_{\mathcal{R}}\) that \(s_{2}\) may perform before taking a matching transition \(u_{n} \rightarrow s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\). In case \(s_{1}^{\prime}=s_{1}, \nu\left(s_{1}, s_{1}, s_{2}\right)\) is an upper bound for the stutter steps \(s_{1} u_{1} \ldots u_{n}\) inside \(\left[s_{1}\right]_{\mathcal{R}}\) that cannot be matched by \(s_{2}\).

\section*{Example 7.125. Step-Dependent Bisimulation Equivalence}

For the transition system \(T S\) shown in Figure 7.45, we have \(s_{1} \approx_{T S}^{s} s_{2}\). This is justified by establishing a step-dependent normed bisimulation for \(T S\) that contains \(\left(s_{1}, s_{2}\right)\).


Figure 7.45: \(s_{1} \approx_{T S}^{s} s_{2}\).

Let \(\mathcal{R}\) be the equivalence that identifies states \(s_{1}, u_{1}, u_{2}, u_{3}, s_{2}\), states \(t_{1}, t_{2}\), and states
\(w_{1}, w_{2}\). Let the step-dependent norm function \(\nu: S^{3} \rightarrow \mathbb{N}\) be defined as follows:
\[
\begin{aligned}
& \nu\left(s_{1}, s_{1}, s_{2}\right)=2 \quad \nu\left(u_{1}, u_{1}, s_{2}\right)=1 \\
& \nu\left(w_{1}, s_{1}, s_{2}\right)=2 \quad \nu\left(w_{1}, s_{1}, u_{2}\right)=1 \\
& \nu\left(w_{1}, u_{1}, s_{2}\right)=2 \quad \nu\left(w_{1}, u_{1}, u_{2}\right)=1 \\
& \nu\left(t_{1}, u_{1}, s_{2}\right)=2 \quad \nu\left(t_{1}, u_{1}, u_{2}\right)=1 \\
& \nu\left(s_{2}, s_{2}, s_{1}\right)=2 \quad \nu\left(u_{2}, u_{2}, s_{1}\right)=1 \\
& \nu\left(u_{3}, u_{3}, s_{1}\right)=1 \quad \nu\left(t_{2}, u_{3}, s_{1}\right)=1
\end{aligned}
\]
and \(\nu(\cdot)=0\) in all other cases. Then, \((\mathcal{R}, \nu)\) is a step-dependent normed bisimulation for \(T S\). Let us check the required conditions. Condition (SI) is obvious since \(\mathcal{R}\) identifies only equally labeled states. Consider the pair \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and check whether (SII) holds for the outgoing transitions of \(s_{1}\).
- For the transition \(s_{1} \rightarrow w_{1}\), we consider the transition \(s_{2} \rightarrow u_{2}\) and get \(\left(s_{1}, u_{2}\right) \in \mathcal{R}\) and \(\nu\left(w_{1}, s_{1}, u_{2}\right)=1<2=\nu\left(w_{1}, s_{1}, s_{2}\right)\) Hence, condition (S3) is fulfilled.
- For the transition \(s_{1} \rightarrow u_{1}\), case (S2) applies as we have \(\left(u_{1}, s_{2}\right) \in \mathcal{R}\) and \(\nu\left(u_{1}, u_{1}, s_{2}\right)=\) \(1<2=\nu\left(s_{1}, s_{1}, s_{2}\right)\).

Let us now consider the pair \(\left(u_{1}, s_{2}\right)\). For the transition \(u_{1} \rightarrow v\) where \(v \in\left\{w_{1}, t_{1}\right\}\) we consider the stutter step \(s_{2} \rightarrow u_{2}\) and get \(\nu\left(x, u_{1}, u_{2}\right)=1<2=\nu\left(x, u_{1}, s_{2}\right)\), which yields condition (S3). Similarly, for the pair \(\left(u_{1}, u_{2}\right) \in \mathcal{R}\) and the transitions \(u_{1} \rightarrow x\) case (S3) applies as we may take the transition \(u_{2} \rightarrow u_{3}\) and get
\[
\nu\left(x, u_{1}, u_{3}\right)=0<1=\nu\left(x, u_{1}, u_{2}\right) .
\]

For ( \(u_{1}, u_{3}\) ) and transition \(u_{1} \rightarrow x\) condition (S1) is fulfilled as we may take the corresponding transition from \(u_{3}\).

Let us now prove condition (SII) for the pair \(\left(s_{2}, s_{1}\right) \in \mathcal{R}\). We have to regard the transition \(s_{2} \rightarrow u_{2}\). In fact, case (S2) applies since \(s_{2}\) and \(u_{2}\) are \(\mathcal{R}\) equivalent and
\[
\nu\left(s_{2}, s_{2}, s_{1}\right)=2>1=\nu\left(u_{2}, u_{2}, s_{1}\right)
\]

Similarly, for the pair \(\left(u_{2}, s_{1}\right) \in \mathcal{R}\) we have to consider the transition \(u_{2} \rightarrow u_{3}\) where again condition (S2) is fulfilled as we have
\[
\nu\left(u_{2}, u_{2}, s_{1}\right)=1>0=\nu\left(u_{3}, u_{3}, s_{1}\right)
\]

For \(\left(u_{3}, s_{1}\right) \in \mathcal{R}\) and transition \(u_{3} \rightarrow w_{2}\) condition (S1) holds as we may move from \(s_{1}\) to \(w_{1}\). For the transition \(u_{3} \rightarrow t_{2}\) case (S3) applies for the stutter step \(s_{1} \rightarrow u_{1}\) as we have \(\nu\left(t_{2}, u_{3}, s_{1}\right)=1>0=\nu\left(t_{2}, u_{3}, u_{1}\right)\).

The following theorem shows that step-dependent norm functions yield a sound and complete criterion for divergence stutter bisimilarity.

\section*{Theorem 7.126. Alternative Characterization of \(\approx_{T S}^{\text {div }}\)}

Let TS be a finite transition system and \(s_{1}\) and \(s_{2}\) states in TS. Then:
\[
s_{1} \approx_{T S}^{s} s_{2} \quad \text { if and only if } s_{1} \approx_{T S}^{d i v} s_{2} .
\]

Proof: We show that \(\approx_{T S}^{S}\) is finer than \(\approx_{T S}^{d i v}\), and vice versa.
Claim 1. \(\approx_{T S}^{s}\) is finer than \(\approx_{T S}^{d i v}\).
Proof of Claim 1. Similar arguments as in the proof of Lemma 7.122 can be used. In fact, for this direction the cardinality of \(T S\) is irrelevant. We take a step-dependent normed bisimulation \((\mathcal{R}, \nu)\) and show that \(\mathcal{R}\) is a divergence-sensitive stutter bisimulation. Let us first check conditions (1) and (2) of stutter bisimulations. (Condition (3) is obtained from (2) by the symmetry of \(\mathcal{R}\).) (1) is immediate from the definition of normed bisimulations (see condition (SI)). Let us check condition (2). Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\left(s_{1}^{\prime}, s_{2}\right) \notin \mathcal{R}\). (SII) yields that one of the conditions (S1), (S2), or (S3) holds. (S2) is impossible as \(\left(s_{1}^{\prime}, s_{2}\right) \notin \mathcal{R}\). If (S1) holds, then (2) is obvious. If (S3) applies, then we pick a state \(u_{1} \in \operatorname{Post}\left(s_{2}\right)\) such that \(u_{1}\) is \(\mathcal{R}\)-equivalent to \(s_{1}\) and \(s_{2}\) and
\[
\nu\left(s_{1}^{\prime}, s_{1}, u_{1}\right)<\nu\left(s_{1}^{\prime}, s_{1}, s_{2}\right)
\]

Since \(\left(s_{1}, u_{1}\right) \in \mathcal{R}\) we may apply the same argument: either (S1) applies, i.e., there exists \(s_{2}^{\prime} \in \operatorname{Post}\left(u_{1}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\), or (S3) applies, which yields a state \(u_{2} \in \operatorname{Post}\left(u_{1}\right)\) such that \(u_{2}\) is \(\mathcal{R}\)-equivalent to \(s_{1}, s_{2}, u_{1}\) and
\[
\nu\left(s_{1}^{\prime}, s_{1}, u_{2}\right)<\nu\left(s_{1}^{\prime}, s_{1}, u_{1}\right)
\]

Repeating this argument yields a path fragment \(s_{2} u_{1} u_{2} \ldots u_{n}\) such that \(\left(s_{1}, u_{i}\right) \in \mathcal{R}\), \(1 \leqslant i \leqslant n\) and
\[
0 \leqslant \nu\left(s_{1}^{\prime}, s_{1}, u_{n}\right)<\nu\left(s_{1}^{\prime}, s_{1}, u_{2}\right)<\ldots<\nu\left(s_{1}^{\prime}, s_{1}, u_{1}\right)<\nu\left(s_{1}^{\prime}, s_{1}, s_{2}\right)
\]
and such that (S3) does not apply to \(\left(s_{1}, u_{n}\right) \in \mathcal{R}\) and the transition \(s_{1} \rightarrow s_{1}^{\prime}\). Since \(s_{1}^{\prime}\) is not \(\mathcal{R}\)-equivalent to \(s_{1}\), (S1) applies which yields a transition \(u_{n} \rightarrow s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\). Hence, we have a path fragment \(s_{2} u_{1} u_{2} \ldots u_{n} s_{2}^{\prime}\) with \(\left(s_{1}, u_{i}\right) \in \mathcal{R}\) and \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\), as required in condition (2).

It remains to show the divergence sensitivity of \(\mathcal{R}\). Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and let \(s_{1}\) be \(\mathcal{R}\) divergent. We have to prove the \(\mathcal{R}\)-divergency of \(s_{2}\). For this, we show the existence
of a transition \(s_{2} \rightarrow v_{1}\) such that \(s_{2}\) and \(v_{1}\) are \(\mathcal{R}\)-equivalent. We then may repeat this argument for \(\left(s_{1}, v_{1}\right)\) which yields the existence of a transition \(v_{1} \rightarrow v_{2}\) with \(\left(s_{1}, v_{2}\right) \in \mathcal{R}\) and the existence of a transition \(v_{2} \rightarrow v_{3}\) with \(\left(s_{1}, v_{3}\right) \in \mathcal{R}\), and so on. In this way we obtain an infinite path \(s_{2} v_{1} v_{2} v_{3} \ldots\) with \(\left(s_{1}, v_{i}\right) \in \mathcal{R}\) for all \(i \geqslant 0\), and thus, the \(\mathcal{R}\)-divergency of \(s_{2}\).

Since \(s_{1}\) is \(\mathcal{R}\)-divergent there is an infinite path \(u_{0} u_{1} u_{2} u_{3} \ldots\) starting in \(s_{1}=u_{0}\) where \(s_{1}\) and all states \(u_{i}\) are \(\mathcal{R}\)-equivalent. If case (S2) applies to the pair \(\left(u_{i}, s_{2}\right) \in \mathcal{R}\) and transitions \(u_{i} \rightarrow u_{i+1}, i=0,1, \ldots, k\), then
\[
\nu\left(s_{1}, s_{1}, s_{2}\right)>\nu\left(u_{1}, u_{1}, s_{2}\right)>\nu\left(u_{2}, u_{2}, s_{2}\right)>\ldots>\nu\left(u_{k}, u_{k}, s_{2}\right) \geqslant 0
\]

Let \(k_{1} \geqslant 0\) be the smallest index such that case ( S 2 ) does not apply to the pair \(\left(u_{k_{1}}, s_{2}\right) \in \mathcal{R}\) and transition \(u_{k_{1}} \rightarrow u_{k_{1}+1}\). If case (S1) holds, then there is a transition \(s_{2} \rightarrow v_{1}\) where \(\left(u_{k_{1}+1}, v_{1}\right) \in \mathcal{R}\). Since all \(u\)-states are \(\mathcal{R}\)-equivalent to \(s_{1}\) and \(s_{2}\), this yields \(\left(s_{2}, v_{1}\right) \in \mathcal{R}\), and we are done. If case (S3) holds for \(\left(u_{k_{1}}, s_{2}\right) \in \mathcal{R}\) and transition \(u_{k_{1}} \rightarrow u_{k_{1}+1}\), then there exists \(v_{1} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(u_{k_{1}}, v_{1}\right) \in \mathcal{R}\) and \(\nu\left(u_{k_{1}+1}, u_{k_{1}}, v_{1}\right)<\nu_{2}\left(u_{k_{1}+1}, u_{k_{1}}, s_{2}\right)\). Again, we have \(s_{2} \rightarrow v_{1}\) and the \(\mathcal{R}\)-equivalence of \(s_{2}\) and \(v_{1}\).

Claim 2. \(\approx_{T S}^{d i v}\) is finer than \(\approx_{T S}^{S}\), provided that \(T S\) is finite.
Proof of Claim 2. We establish a ternary norm function \(\nu\) that satisfies the conditions in Definition 7.124 for \(\mathcal{R}=\approx_{T S}^{d i v}\).
- For \(s_{1} \approx_{T S}^{d i v} s_{2}\) and \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(s_{1} \not \ddot{\chi}_{T S}^{d i v} s_{1}^{\prime}\) we define \(\nu\left(s_{1}^{\prime}, s_{1}, s_{2}\right)\) as the length of a shortest path fragment \(s_{2} u_{1} u_{2} \ldots u_{n}\) such that
\(-u_{i} \approx_{T S}^{\text {div }} s_{2}, i=1, \ldots, n\)
\(-\left[s_{1}^{\prime}\right]_{\text {div }} \cap \operatorname{Post}\left(u_{n}\right) \neq \varnothing\), where \([s]_{\text {div }}\) denotes the equivalence class of \(s\) under \(\approx_{T S}^{d i v}\).
- If \(s_{1} \approx_{T S}^{d i v} s_{2}\) and \(s_{1}, s_{2}\) are \(\approx_{T S}^{d i v}\)-divergent then we put \(\nu\left(s_{1}, s_{1}, s_{2}\right)=0\).
- If \(s_{1} \approx_{T S}^{d i v} s_{2}\) and \(s_{1}, s_{2}\) are not \(\approx_{T S}^{d i v}\)-divergent then we define \(\nu\left(s_{1}, s_{1}, s_{2}\right)\) as the length of a longest path fragment \(s_{1} v_{1} v_{2} \ldots v_{n}\) such that
\[
v_{i} \approx_{T S}^{d i v} s_{1} \text { for } i=1, \ldots, n \quad \text { and } \quad\left[s_{1}\right]_{d i v} \cap \operatorname{Post}\left(v_{n}\right)=\varnothing .
\]

Note that such a longest finite path fragment exists since \(T S\) is finite and \(s_{1}\) is not divergent.

We now show that \(\left(\approx_{T S}^{d i v}, \nu\right)\) is a step-dependent normed bisimulation for TS. Obviously, \(\approx_{T S}^{d i v}\) is an equivalence on \(S\) where (SI) holds. Let us check condition (SII) for states \(s_{1}, s_{2}\) with \(s_{1} \approx_{T S}^{d i v} s_{2}\) and transition \(s_{1} \rightarrow s_{1}^{\prime}\).
- If \(s_{1} \not \overbrace{T S}^{d i v} s_{1}^{\prime}\), then either \(\left[s_{1}^{\prime}\right]_{d i v} \cap \operatorname{Post}\left(s_{2}\right) \neq \varnothing\) (in which case condition (S1) holds) or there is a path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) with \(n \geqslant 1\) and
\[
u_{i} \approx_{T S}^{d i v} s_{2}, i=1, \ldots, n \text { and } s_{1}^{\prime} \approx_{T S}^{d i v} s_{2}^{\prime} .
\]

We may assume that \(n\) is minimal. Then, \(s_{2} \rightarrow u_{1}\) is a transition with \(\left(u_{1}, s_{2}\right) \in \mathcal{R}\) and \(n-1=\nu\left(s_{1}^{\prime}, s_{1}, u_{1}\right)<\nu\left(s_{1}^{\prime}, s_{1}, s_{2}\right)=n\). Hence, condition (S3) holds.
- Let us now consider a stutter step \(s_{1} \rightarrow s_{1}^{\prime}\) inside the equivalence class of \(s_{1}\), i.e., we suppose \(s_{1} \approx_{T S}^{d i v} s_{1}^{\prime}\).
- If \(s_{1}\) is divergent, then so is \(s_{2}\) and there is a step \(s_{2} \rightarrow s_{2}^{\prime}\) where \(s_{2}^{\prime}\) is \(\approx_{T S}^{d i v}\) equivalent to \(s_{1}\) and \(s_{2}\). Thus, condition (S1) holds.
- Let us now suppose that \(s_{1}\) and \(s_{2}\) are not divergent. For \(s \approx_{T S}^{d i v} s_{1}\) let \(\ell(s)\) be the length of a longest path fragment \(s v_{1} v_{2} \ldots v_{n}\) such that \(v_{i} \approx_{T S}^{d i v} s\), \(i=1, \ldots, n\) and \([s]_{\text {div }} \cap \operatorname{Post}\left(v_{n}\right)=\varnothing\). Clearly, we have
\[
\ell(s)=\max _{\left.s^{\prime} \in \text { Post }(s) \cap[s]\right]_{\text {div }}} \ell\left(s^{\prime}\right)+1
\]
\(\nu\left(s, s, s_{2}\right)=\ell(s)\) (by definition of \(\left.\nu\right)\). Hence,
\[
\nu\left(s_{1}, s_{1}, s_{2}\right)=\ell\left(s_{1}\right) \geqslant \ell\left(s_{1}^{\prime}\right)+1>\nu\left(s_{1}^{\prime}, s_{1}^{\prime}, s_{2}\right)
\]

This shows that condition (S2) holds.

\subsection*{7.8.3 Stutter Bisimulation and \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) Equivalence}

Stutter bisimulation \((\approx)\) does not preserve LTL \(_{\backslash \bigcirc}\) formulae since it does not impose any restrictions on divergent paths. That is, if \(T S_{1} \approx T S_{2}\), and \(T S_{1}\) contains a divergent path violating some \(\mathrm{LTL}_{\backslash \bigcirc}\) formula \(\varphi\), whereas \(T S_{2}\) does not exhibit such behaviour, then \(T S_{1} \not \vDash \varphi\) whereas \(T S_{2} \vDash \varphi\). To avoid the complications induced by stutter paths, we have considered \(\approx{ }^{d i v}\) that requires two equivalent states to either both have a divergent path, or both have no divergent paths. The central result in this section is that \(\approx^{d i v}\) coincides with the logical equivalences for the next-free fragments of CTL* and CTL. This entails that any formula in these logics can be checked on a transition system that is stutter bisimulation-equivalent with divergence, in particular, the quotient under \(\approx^{d i v}\). Moreover, to show \(T S_{1} \not \boldsymbol{z}^{\text {div }} T S_{2}\) it suffices to provide a CTL* formula \(\Phi\) that does not contain \(\bigcirc\), such that \(T S_{1} \models \Phi\) and \(T S_{2} \not \models \Phi\).

\section*{Notation 7.127. CTL* and CTL without Next Step}

Let \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) denote the class of all CTL* formulae that do not contain any occurrence of the next step operator \(\bigcirc\). Similarly, \(\mathrm{CTL}_{\backslash \bigcirc}\) denotes the sublogic of CTL which does not permit the next-step operator \(\bigcirc\).

Theorem 7.128. \(\quad C T L_{\backslash \bigcirc}^{*} / C T L_{\backslash \bigcirc}\) Equivalence and \(\approx^{\text {div }}\)
For finite transition system TS without terminal states, and states \(s_{1}, s_{2}\) in TS:
\[
s_{1} \approx_{T S}^{d i v} s_{2} \quad \text { iff } \quad s_{1} \equiv{ }_{C T L_{\backslash ○}^{*}} \quad s_{2} \quad \text { iff } \quad s_{1} \equiv{ }_{C T L_{\backslash ○}} s_{2} .
\]

Proof: The proof strategy is analogous to the proof of Theorem 7.20 (page 469).
1. Lemma 7.129 (see below) asserts that \(\approx_{T S}^{d i v}\) is finer than \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) equivalence for arbitrary (possibly infinite) transition systems.
2. \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) equivalence is finer than \(\mathrm{CTL}_{\backslash \bigcirc}\) equivalence, since \(\mathrm{CTL}_{\backslash \bigcirc}\) is a sublogic of CTL \({ }^{*}\)
3. Lemma 7.130 (see below) asserts that \(\mathrm{CTL}_{\backslash \bigcirc}\) equivalence is finer than \(\approx_{T S}^{d i v}\).

Note that the proof of (3) for CTL equivalence and bisimulation ( \(\sim\) ) exploits the next-step operator (and not the until operator). As this operator is absent in \(\mathrm{CTL}_{\backslash \bigcirc}\), an alternative logical characterization has to be pursued.

Lemma 7.129. Divergence Stutter Bisimilarity Implies CTL \({ }_{\backslash \bigcirc}^{*}\) Equivalence
Let TS be a transition system without terminal states, \(s_{1}, s_{2}\) states in TS, and \(\pi_{1}, \pi_{2} \in\) Paths(TS). Then:
1. if \(s_{1} \approx_{T S}^{\text {div }} s_{2}\), then for all \(C T L_{\backslash \bigcirc}^{*}\) state formulae \(\Phi\) : \(s_{1} \models \Phi\) iff \(s_{2} \models \Phi\).
2. if \(\pi_{1} \approx_{T S}^{\text {div }} \pi_{2}\), then for all CTL \({ }_{\backslash \bigcirc}^{*}\) path formulae \(\varphi: \pi_{1} \models \varphi\) iff \(\pi_{2} \models \varphi\).

Proof: The proof is provided by structural induction over the CTL \({ }_{\backslash \bigcirc}^{*}\) formulae and is similar to the proof of Lemma 7.26 (page 473). We omit the details and only consider the case \(\Phi=\exists \varphi\), where \(\varphi\) is a \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) path formula satisfying claim 2. Assume \(s_{1} \approx_{T S}^{d i v} s_{2}\)
and \(s_{1} \exists \varphi\). Let \(\pi_{1} \in \operatorname{Paths}\left(s_{1}\right)\) such that \(\pi_{1} \models \varphi\). According to Lemma 7.118 (page 549), there exists \(\pi_{2} \in \operatorname{Paths}\left(s_{2}\right)\) with \(\pi_{1} \approx_{T S}^{d i v} \pi_{2}\). Applying the induction hypothesis to \(\varphi\) and paths \(\pi_{1}, \pi_{2}\) yields \(\pi_{2} \models \varphi\). Hence, \(s_{2} \models \exists \varphi\).

\section*{Lemma 7.130. \(\quad \boldsymbol{C T L} L_{\backslash \bigcirc}\) Equivalence is Finer Than \(\approx_{T S}^{\text {div }}\)}

For finite transition system TS without terminal states, and \(s_{1}, s_{2}\) states in TS:
\(s_{1}\) and \(s_{2}\) are \(C T L_{\backslash \bigcirc}\) equivalent implies \(s_{1} \approx_{T S}^{d i v} s_{2}\).

Proof: Let \(S\) be the state space of \(T S\). It suffices to show that
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid s_{1} \equiv_{C T L_{\bigcirc \bigcirc}} s_{2}\right\}
\]
is a divergence-sensitive stutter bisimulation for \(T S\). We show that (1) \(\mathcal{R}\) is a stutter bisimulation for \(T S\) and (2) \(\mathcal{R}\) is divergence-sensitive.

Claim 1. \(\mathcal{R}\) is a stutter bisimulation for \(T S\).
Proof of Claim 1. This is proven by checking the conditions of stutter bisimulation. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).
1. Consider the propositional logic formula
\[
\Phi=\bigwedge_{a \in L\left(s_{1}\right)} a \wedge \bigwedge_{a \in A P \backslash L\left(s_{1}\right)} \neg a
\]

Clearly, \(\Phi\) is a \(\mathrm{CTL}_{\backslash \bigcirc}\) formula with \(s_{1} \models \Phi\). Hence, by definition of \(\mathcal{R}, s_{2} \models \Phi\). By construction of \(\Phi\), we have \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
2. For each equivalence class \(C \in S / \mathcal{R}\), let \(\mathrm{CTL}_{\backslash \bigcirc}\) formula \(\Phi_{C}\) be defined by
\[
\Phi_{C}=\bigwedge_{\substack{D \in S / \mathcal{R} \\ D \neq C}} \Phi_{C, D}
\]
where \(\Phi_{C, D}\) is a CTL \({ }_{\backslash \bigcirc}\) formula such that \(\operatorname{Sat}\left(\Phi_{C, D}\right) \supseteq C\) and \(\operatorname{Sat}\left(\Phi_{C, D}\right) \cap D=\varnothing\), for each pair \((C, D)\) of equivalence classes \(C, D\) under \(\mathcal{R}\) with \(C \neq D\). This yields \(\operatorname{Sat}\left(\Phi_{C}\right)=C\).
It remains to prove that for all \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\left(s_{1}, s_{1}^{\prime}\right) \notin \mathcal{R}\), there exists a finite path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\) and \(\left(s_{2}, u_{i}\right) \in \mathcal{R}, i=1, \ldots, n\). Let
\(B \in S / \mathcal{R}\) and \(s_{1}, s_{2} \in B\), i.e., \(B=\left[s_{1}\right]_{\mathcal{R}}=\left[s_{2}\right]_{\mathcal{R}}\). Further, let \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) such that \(\left(s_{1}, s_{1}^{\prime}\right) \notin \mathcal{R}\), say \(C=\left[s_{1}^{\prime}\right]_{\mathcal{R}}\) with \(B \neq C\). Then:
\[
s_{1} \models \exists\left(\Phi_{B} \cup \Phi_{C}\right) \quad \text { as } s_{1} \in B \text { and } s_{1}^{\prime} \in C
\]

Since \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(\Phi_{B} \cup \Phi_{C}\) is a \(\mathrm{CTL}_{\backslash \bigcirc}\) formula, we get
\[
s_{2} \models \exists\left(\Phi_{B} \cup \Phi_{C}\right)
\]

Thus, there exists a finite path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) with \(s_{2}^{\prime} \models \Phi_{C}\) and \(s_{2} \models \Phi_{B}\), \(u_{1} \models \Phi_{B}, \ldots, u_{n} \models \Phi_{B}\). Since \(\operatorname{Sat}\left(\Phi_{C}\right)=C\) and \(\operatorname{Sat}\left(\Phi_{B}\right)=B\), we obtain \(s_{2}^{\prime} \in C\) and \(u_{1}, \ldots, u_{n} \in B\). As \(C=\left[s_{1}^{\prime}\right]_{\mathcal{R}}\) and \(B=\left[s_{1}\right]_{\mathcal{R}}=\left[s_{2}\right]_{\mathcal{R}}\), we conclude that \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\) and \(\left(s_{2}, u_{i}\right) \in \mathcal{R}\), for \(i=1, \ldots, n\).

Claim 2. \(\mathcal{R}\) is divergence-sensitive.

Proof of Claim 2. Assume \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1}\) is \(\mathcal{R}\)-divergent. We have to show that \(s_{2}\) is \(\mathcal{R}\)-divergent. As \(s_{1}\) is \(\mathcal{R}\)-divergent there exists an \(\mathcal{R}\)-divergent path fragment:
\[
\pi_{1}=s_{1} s_{1,1} s_{2,1} s_{3,1} \ldots \in \operatorname{Paths}\left(s_{1}\right)
\]

Let \(C=\left[s_{1}\right]_{\mathcal{R}}\) and \(\Phi_{C}\) be a \(C T L_{\backslash \bigcirc}\) formula such that \(\operatorname{Sat}\left(\Phi_{C}\right)=C\). Then \(s_{j, 1} \in C\), i.e., \(s_{j, 1} \models \Phi_{C}\) for all \(j>0\). Hence, \(s_{1} \models \exists \square \Phi_{C}\). As \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(C=\left[s_{1}\right]_{\mathcal{R}}\), we get \(s_{2} \models \exists \square \Phi_{C}\). Thus, there exists an infinite path fragment:
\[
\pi_{2}=s_{2} s_{1,2} s_{2,2} s_{3,2} \ldots \in \operatorname{Paths}\left(s_{2}\right)
\]
with \(s_{j, 2} \models \Phi_{C}\) for \(j>0\). As \(\operatorname{Sat}\left(\Phi_{C}\right)=C\), we get \(s_{j, 2} \in C\) for \(j>0\). Since \(C=\left[s_{2}\right]_{\mathcal{R}}, \pi_{2}\) is an \(\mathcal{R}\)-divergent path. Hence, \(s_{2}\) is \(\mathcal{R}\)-divergent.

A few remarks are in order. \(\approx_{T S}^{d i v}\) is the coarsest equivalence which preserves all \(\mathrm{CTL}_{\backslash \bigcirc}\) formulae. Theorem 7.128 can be reformulated for pairs of transition systems in the usual way. Thus, for all finite transition systems \(T S_{1}, T S_{2}\) without terminal states:
\[
\begin{array}{lll}
T S_{1} \approx \operatorname{div} T S_{2} & \text { iff } & T S_{1} \text { and } T S_{2} \text { satisfy the same } \mathrm{CTL}_{\backslash \bigcirc}^{*} \text { formulae } \\
& \text { iff } & T S_{1} \text { and } T S_{2} \text { satisfy the same } \mathrm{CTL}_{\backslash \bigcirc} \text { formulae. }
\end{array}
\]

We finally remark that in the proof of the logical characterization of \(\approx_{T S}^{d i v}\), the until operator is explicitly used. Thus, whereas bisimulation and simulation can be characterized by a CTL (or CTL*) fragment that does not contain the until operator, this is not the case for \(\approx_{T S}^{d i v}\). This is not surprising, as \(\approx_{T S}^{d i v}\) does not preserve the next-step operator and some modal operator is needed to characterize finite stutter path fragments.


Figure 7.46: \(T S_{A B P} / \approx^{d i v}\).
In particular, since \(T S \approx^{d i v} T S / \approx^{d i v}\), it suffices to check \(\mathrm{CTL}_{\wedge_{○}}^{*}\) formula \(\Phi\) on \(T S / \approx d i v\) to decide whether \(T S \models \Phi\). This is exemplified by the alternating bit protocol.

\section*{Example 7.131. Alternating Bit Protocol}

Consider again the alternating bit protocol; see Example 2.32 on page 57. Let
\[
A P=\left\{s \_m o d e=0, s_{-} \text {mode }=1, \text { r_mode }=0, \text { r_mode }=1\right\}
\]
where s_mode indicates the mode of the sender, and r_mode that of the receiver. In Example 7.108 (page 545), it was concluded that \(T S_{A B P}\), the transition system underlying the ABP, and its stutter bisimulation quotient \(T S_{A B P} / \approx\), do not satisfy the same \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) formulae. This is due to the fact that the transition system
\[
T S=\left(T S_{A B P} \oplus T S_{A B P}\right) / \approx
\]
is not divergence-sensitive, i.e., there are stutter-bisimilar states in \(T S\) such that one of them is \(\approx\)-divergent, while the other is not. In fact, in order to prove a \(\mathrm{CTL}_{\backslash \bigcirc}^{*}, \mathrm{CTL}_{\backslash \bigcirc}\), or \(\operatorname{LTL}_{\backslash \bigcirc}\) formula of the ABP, we must consider \(T S_{A B P} / \approx^{d i v}\) instead of \(T S_{A B P} / \approx\).

To obtain the quotient under \(\approx^{\text {div }}\), each state in \(T S_{A B P} / \approx\) for which s_mode \(\neq\) r_mode needs to be decomposed in two states. Accordingly, the transition system \(T S_{A B P} / \approx^{d i v}\) is obtained as depicted in Figure 7.46. Two of those states are labeled with
\[
A_{1}=\left\{s \_m o d e=0, r_{-} \text {mode }=1\right\}, \quad A_{2}=\left\{s \_m o d e=1, \text { r_mode }=0\right\} .
\]

One of these two states represents the divergent \(A_{i}\)-states in \(T S_{A B P}\), while the other represents the nondivergent \(A_{i}\)-states \(\left\langle\operatorname{ch} k_{-} a c k(i), \ldots, x=i, \ldots\right\rangle\) in \(T S_{A B P}\). The CTL \({ }^{*}\) ○ safety property
\[
\Phi=\forall \square\left(\left(s \_m o d e=0\right) \longrightarrow\left(s \_m o d e=0\right) \mathrm{W}(\text { r_mode }=1)\right)
\]
expresses that the sender does not leave the 0 -mode (in which it only transmits message with bit 0) before the receiver changes to the 1-mode (in which it expects to receive messages with bit 1, i.e., before the receiver has acknowledged the correct receipt of the message with bit 0 . It is easy to establish that \(T S_{A B P} / \approx^{d i v} \models \Phi\). Theorem 7.128 yields \(T S_{A B P}=\Phi\).

\section*{Example 7.132. Producer-Consumer System}

Consider a concurrent program that involves a producer and consumer process that shares a buffer of data items of capacity \(m>0\). The producer repeatedly produces \(n\) items and inserts these data items one by one in the buffer. Local variable in indicates the next buffer cell that needs to be written. The producer can only insert an element in cell buffer at index in when this cell is empty. An empty buffer cell is indicated by \(\perp\). The consumer process successively attempts to get items from the buffer.
```

Producer
in : $=0$;
while true \{
produce $d_{1}, \ldots, d_{n}$;
for $i=1$ to $n\{$
wait until (buffer[in] $=\perp$ ) $\{$
buffer[in] := $d_{i}$;
in $:=(i n+1) \bmod m ;\}$
\}
\}

```

Consumer
```

out := 0;
while true {
for }j=1\mathrm{ to }n
wait until (buffer[out]}\not=\perp)
e}\mp@subsup{e}{j}{}:= buffer[out]
buffer[out]:= \perp;
out:= (out + 1) mod m; }
}
consume e}\mp@subsup{e}{1}{},···,\mp@subsup{e}{n}{
}

```

Let \(T S(m, n)\) denote the transition system underlying this producer-consumer program. The size of \(T S(n, m)\) grows exponentially in the number \(n\) of data items generated per cycle and in the buffer capacity \(m\).

Suppose we are interested in the situation in which the producer and the consumer are simultaneously in their produce and consume phase, respectively. Let \(A P=\{\) prod_and_cons \(\}\). Observe that the content of the buffer can be ignored. This also applies to the variables in and out. In order to keep track of the number of free buffer cells, we introduce the bounded integer variable free with domain \(\{0,1, \ldots, m\}\). On completion of the production phase, variable \(i\) indicates the number of data items the producer has already stored in the buffer. Accordingly, \(j\) stands for the number of data items the consumer has withdrawn from the buffer. This yields the following abstract programs:
```

Producer
while true \{
produce;
for $i=1$ to $n\{$
wait until (free >0) \{
free := free - 1;
\}
\}

```

Consumer
```

while true {
for }j=1\mathrm{ to }n
wait until (free < m) {
free := free + 1;
}
consume
}

```

Let \(T S_{\text {abstract }}(m, n)\) denote the transition system that is obtained for this abstract program. \(T S_{\text {abstract }}(2,2)\) is indicated in Figure 7.47, where it is assumed that executing the body of the for loop is executed atomically. Each state is of the form \(\left\langle\ell_{i}, \ell_{i}^{\prime}, v_{f}, v_{i}, v_{j}\right\rangle\) where \(\ell_{i}\) and \(\ell_{i}^{\prime}\) indicate the program locations of the producer and consumer, respectively; \(v_{f}\) is the value of free, and \(v_{i}, v_{j}\) the value of integers \(i\) and \(j\). In particular, we have:
\[
\begin{aligned}
\ell_{0} & : \text { produce } \\
\ell_{1} & :\langle\text { if }(\text { free }>0) \text { then } i:=1 ; \text { free }--\mathbf{f i}\rangle \\
\ell_{2} & :\langle\text { if }(\text { free }>0) \text { then } i:=0 ; \text { free }-\mathbf{f i}\rangle
\end{aligned}
\]

The labels \(\ell_{i}^{\prime}\) are defined for the consumer process in a similar way. Note that the states with control locations \(\ell_{0}\) and \(\ell_{2}^{\prime}\) indicate the global states in which the producer and consumer are in their produce and consume phase, respectively. Thus, all states of the form \(\langle 0,2,-,-,-\rangle\) are labeled with the proposition prod_and_cons; all other states are labeled with \(\varnothing\). The size of \(T S_{\text {abstract }}(n, m)\) is polynomial in \(n\) and \(m\). It follows that
\[
T S(m, n) \approx^{d i v} T S_{a b s t r a c t}(m, n)
\]

Observe that
\[
T S_{a b s t r a c t}(m, n) \not \models \forall \square \forall \diamond \text { prod_and_cons . }
\]

As this is a \(\mathrm{CTL}_{\backslash \bigcirc}\) formula, it follows that \(T S(m, n)\) refutes it. A counterexample in \(T S_{\text {abstract }}(m, n)\) is, e.g.,
\[
\begin{aligned}
& \left\langle\ell_{0}, \ell_{0}^{\prime}, 2,0,0\right\rangle \rightarrow\left\langle\ell_{1}, \ell_{0}^{\prime}, 2,0,0\right\rangle \rightarrow\left\langle\ell_{2}, \ell_{0}^{\prime}, 1,1,0\right\rangle \rightarrow\left\langle\ell_{0}, \ell_{0}^{\prime}, 0,0,0\right\rangle \rightarrow \\
& \left\langle\ell_{0}, \ell_{1}^{\prime}, 1,0,1\right\rangle \rightarrow\left\langle\ell_{1}, \ell_{1}^{\prime}, 1,0,1\right\rangle \rightarrow\left\langle\ell_{0}, \ell_{2}^{\prime}, 2,0,0\right\rangle \rightarrow\left\langle\ell_{0}, \ell_{0}^{\prime}, 2,0,0\right\rangle \rightarrow \ldots
\end{aligned}
\]

The fact that \(T S_{a b s t r a c t}(n, m) \not \vDash \forall \square \forall \vee\) prod_and_cons can easily be shown by considering its quotient under \(\approx^{d i v}\); see the self-loop in the initial states in Figure 7.48.


Figure 7.47: Transition system \(\operatorname{TS}_{\text {abstract }}(2,2)\).

\subsection*{7.8.4 Stutter Bisimulation Quotienting}

This section is concerned with algorithms to compute the quotient transition system under the stutter bisimulations \(\approx\) and \(\approx{ }^{d i v}\). Such algorithms serve the following two purposes. First, they can be used to check whether two transition systems are stutter bisimulation equivalent. Secondly, the algorithm for \(\approx^{d i v}\) can be used to minimize a transition system with respect to divergence-sensitive stutter bisimulation. This can be used as a preprocessing phase prior to model checking a CTL \(_{\backslash \bigcirc}^{*}\) formula. We first outline how to compute the equivalence classes under \(\approx\) (i.e., without divergence) and then explain how this algorithm can be adapted to \(\approx^{d i v}\).

In the remainder of this section, let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a finite transition system, possibly with terminal states. The quotienting algorithms for \(\approx\) and \(\approx^{d i v}\) are based on partition refinement, as for bisimulation. Recall that partition refinement algorithms are based on successively refining partitions until a stable partition is reached. The refinement is based on splitting a block in the current partition with respect to a superblock, i.e., a disjoint union of blocks. The idea of refinement for \(\approx\) is to split a given block \(B\) according to another block \(C\) into two subblocks: a block of states in \(B\) that can reach \(C\) via a path fragment that only visits states in \(B\), and a block that contains all other states in \(B\).


Figure 7.48: Quotient transition system \(T S_{\text {abstract }}(2,2) / \approx{ }^{d i v}\).

\section*{Notation 7.133. Constrained Predecessors}

For partition \(\Pi\) of \(S\) and blocks \(B, C\) in \(\Pi\) with \(s \in B\), we have \(s \in \operatorname{Pr} e_{\Pi}^{*}(C)\) whenever there exists a finite path fragment:
\[
s=s_{1} s_{2} \ldots s_{n-1} s_{n} \in \operatorname{Paths}(s) \text { with } s_{n} \in C \text { and } s_{i} \in B \text { for all } 0<i<n .
\]

Stated in words, \(\operatorname{Pre}_{\Pi}^{*}(C)\) contains any state \(s\) that can reach a state in \(C\) via a (possibly single-state) path that solely consists of states that are in the same block of \(\Pi\) as \(s\).

\section*{Definition 7.134. \(\quad \Pi\)-Splitter, \(C\)-Stability}

Let \(\Pi\) be a partition of \(S\) and let \(C, B \in \Pi\).
1. \(C\) is a \(\Pi\)-splitter for \(B\) if and only if
\[
B \neq C \quad \text { and } \quad B \cap \operatorname{Pre}(C) \neq \varnothing \quad \text { and } \quad B \backslash \operatorname{Pr} e_{\Pi}^{*}(C) \neq \varnothing .
\]
2. \(\Pi\) is \(C\)-stable if there is no block \(B \in \Pi\) such that \(C\) is a \(\Pi\)-splitter for \(B\).
3. \(\Pi\) is stable if \(\Pi\) is \(C\)-stable for all blocks \(C \in \Pi\).

Let us consider the requirements of \(C\) being a \(\Pi\)-splitter of \(B\). As a block cannot be split by itself, the requirement that \(B \neq C\) is natural. Furthermore, \(B\) can only be splitted according to \(C\) when \(C\) can be reached in a single transition from at least one state in \(B\). Finally, the third conjunct asserts that \(B\) should contain states that can only reach \(C\) via some state that is neither in \(B\) nor in \(C\).
```

Algorithm 37 Computing the stutter bisimulation quotient
Input: finite transition system $T S$ with state space $S$
Output: stutter bisimulation quotient space $S / \approx$
$\Pi:=\Pi_{A P} ;$
(* see Algorithm 29, page 479 *)
while ( $\exists B, C \in \Pi . C$ is a $\Pi$-splitter for $B$ ) do
choose such $B, C \in \Pi$;
$\Pi:=(\Pi \backslash\{B\}) \cup$ Refine $_{\approx}(B, C)$
od
return $\Pi$

```

\section*{Lemma 7.135. Coarsest Partition}

The stutter bisimulation quotient space \(S / \approx\) is the coarsest partition \(\Pi\) of \(S\) such that
(i) \(\Pi\) is finer than \(\Pi_{A P}\).
(ii) for all \(B, C \in \Pi: B \cap \operatorname{Pre}(C)=\varnothing\) or \(B \subseteq \operatorname{Pre}_{\Pi}^{*}(C)\).

Proof: Similar to the proof of Lemma 7.34 on page 480.

That is, \(S / \approx\) is the coarsest partition such that (i) states are equally labeled, and (ii) either there is no transition between \(B\) and \(C\), or in case such transition exists, there are some states in \(B\) that cannot reach \(C\) by only visiting states in \(B\). For such states, the only possibility to reach \(C\) is via some other block \(D \neq B, C\). This lemma suggests the following refinement operator:

\section*{Definition 7.136. The Refinement Operator}

Let \(\Pi\) be a partition for \(S\) and \(C\) be a block of \(\Pi\). Then:
\[
\text { Refine }_{\approx}(\Pi, C)=\bigcup_{B \in \Pi} \operatorname{Refine}_{\approx}(B, C)
\]
where Refine \(\approx(B, C)=\left\{B \cap \operatorname{Pr} e_{\Pi}^{*}(C), B \backslash \operatorname{Pre}_{\Pi}^{*}(C)\right\} \backslash\{\varnothing\}\).

The essential steps of the partition refinement technique for quotienting with respect to \(\approx\) are summarized in Algorithm 37. Note that unlike the partition refinement algorithms for ordinary bisimulation \(\sim\), Algorithm 37 refines per iteration just one block \(B\). As \(C\) is a \(\Pi\)-splitter for \(B\), each iteration yields a proper refinement. Therefore, the number of
iterations of the while-loop is bounded by the number of states. In the remainder of this section, some implementation details will be provided to enable an efficient realization of the refinement operator.

Removal of Stutter Cycles As a first step, the transition system TS is simplified by removing all its stutter cycles, i.e., cycles that completely consist of stutter steps. Formally:

\section*{Definition 7.137. Stutter Cycle}

Cycle \(s_{0} s_{1} s_{2} \ldots s_{n}\) (with \(s_{0}=s_{n}\) ) for \(n>0\) in transition system \(T S\) is a stutter cycle when \(s_{i} s_{i+1}\) is a stutter step for \(0 \leqslant i<n\).

All states on a stutter cycle belong to the same equivalence class under \(\approx\), and thus belong to the same block of \(S / \approx\). In fact, this even holds for \(\approx^{d i v}\) :

\section*{Lemma 7.138. Stutter Cycle}

For stutter cycle \(s_{0} s_{1} s_{2} \ldots s_{n}\) in transition system TS:
\[
s_{0} \approx_{T S}^{d i v} s_{1} \approx_{T S}^{d i v} \ldots \approx_{T S}^{d i v} s_{n}
\]

Proof: Let \(\mathcal{R}\) be the smallest equivalence relation on \(S\) which identifies the states \(s_{0}=\) \(s_{n}, s_{1}, \ldots, s_{n-1}\) and no other states. That is, \(\mathcal{R}\) is the equivalence relation that induces the partition:
\[
S / \mathcal{R}=\left\{\left\{s_{0}, \ldots, s_{n-1}\right\}\right\} \cup\left\{\{s\} \mid s \in S \backslash\left\{s_{0}, \ldots, s_{n-1}\right\}\right\}
\]

All states in \(\left\{s_{0}, \ldots, s_{n-1}\right\}\) are stutter-bisimilar as they all lie on the same stutter cycle, i.e., any visible step by \(s_{i} \rightarrow s^{\prime}\) can be matched by a path fragment starting in \(s_{j}(i \neq j)\) by first traversing the cycle to \(s_{i}\) and then taking the transition \(s_{i} \rightarrow s^{\prime}\). Clearly, these states are all divergent. All singleton partitions are stutter-bisimilar and nondivergent. Hence, \(\mathcal{R}\) is a divergence-sensitive stutter bisimulation.

Consequently, all states on a stutter cycle are divergent. Note that for finite transition systems it holds that any divergent path must contain a cycle. Since all states on such cycle are pairwise \(\approx^{d i v}\) equivalent-and thus equally labeled-they form a stutter cycle.

\section*{Corollary 7.139. Complete Characterization of Divergent States}

For finite transition system TS and state s in TS:
\[
\begin{aligned}
s \text { is } \approx \text { div -divergent } i f f & \text { a stutter cycle is reachable from } s \\
& \text { via a path fragment in }[s]_{\approx d i v .} .
\end{aligned}
\]

To simplify the refinement operator, we first remove stutter cycles from transition system \(T S\). This is done in the following way. In the state graph \(G(T S)\), determine the strongly connected components (SCCs) that only contain stutter steps. This can be carried out using a (standard) depth-first search algorithm. Then, any stutter SCC is collapsed into a single state. This yields a new transition system \(T S^{\prime}\) where the states are the stutter SCCs in \(G(T S)\), and \(C \rightarrow{ }^{\prime} C^{\prime}\) with \(C \neq C^{\prime}\) whenever \(s \rightarrow s^{\prime}\) is a transition in \(T S\) with \(s \in C\) and \(s^{\prime} \in C^{\prime}\). By construction, \(T S^{\prime}\) does not contain any stutter cycles. The time complexity of this step is \(\mathcal{O}(M+|S|)\). Note that
\[
s_{1} \approx_{T S} s_{2} \text { if and only if } C_{1} \approx_{T S^{\prime}} C_{2}
\]
where \(C_{i}\) denotes the stutter SCC which contains state \(s_{i}\).

Efficient Realization of Refining a Partition From now on, it is assumed that transition system TS has no stutter cycles. We now focus our attention on an efficient realization of the refinement operator and the search for a splitter (see the loop condition in Algorithm 37). Assume that list representations for the sets \(\operatorname{Pre}(s)\) and \(\operatorname{Post}(s)\) are available for any state \(s\). Let \(\Pi\) be a partition of \(S\) that is coarser than \(S / \approx\) and finer than \(\Pi_{A P}\). The goal is to check the instability of \(\Pi\) for a block \(C\) by a linear search in the lists \(\operatorname{Pre}(s)\) for \(s \in C\) rather than to consider the set \(\operatorname{Pr} e_{\Pi}^{*}(C)\). To that end, we aim at a characterization of a splitter in terms of direct predecessors of \(C\). An important notion to enable such characterization is that of an exit state in a block \(B\), say, i.e., a state whose only outgoing transitions lead to states that are outside \(B\).

\section*{Definition 7.140. Exit States}

Let \(T S\) be a finite transition system without stutter cycles and \(B\) be a block of a partition \(\Pi\) for \(S\), the state space of \(T S\). The exit states of \(B\) and partition \(\Pi\) are defined by
\[
\operatorname{Bottom}(B)=\{s \in B \mid \operatorname{Post}(s) \cap B=\varnothing\} \text { and } \operatorname{Bottom}(\Pi)=\bigcup_{B \in \Pi} \operatorname{Bottom}(B)
\]

Since \(T S\) does not have stutter cycles and is finite, any block \(B\) of partition \(\Pi\) that is finer than \(\Pi_{A P}\) has at least one exit state. This can be seen by contraposition. Assume block \(B \in \Pi\) does not contain an exit state. Then there exists an infinite path \(s_{0} s_{1} s_{2} \ldots\) consisting of states in \(B\). Since \(T S\) is finite, there exists a pair of indices \(i, j\) such that
\(0 \leqslant i<j\) with \(s_{i}=s_{j}\). As \(\Pi\) is finer than \(\Pi_{A P}\), all states in \(B\) are equally labeled. Hence, the path fragment \(s_{i} s_{i+1} \ldots s_{j}\) is a stutter cycle. This contradicts \(T S\) having no stutter cycles. Thus, \(\operatorname{Bottom}(B) \neq \varnothing\) for any block \(B \in \Pi\) where \(\Pi\) refines \(\Pi_{A P}\).

The following lemma is the key for checking whether a block \(C \in \Pi\) is a \(\Pi\)-splitter by a linear search in the lists \(\operatorname{Pre}(s)\) for \(s \in C\).

\section*{Lemma 7.141. Local Splitter Criterion}

Let TS be a finite transition system without stutter cycles, \(\Pi\) a partition of TS's state space \(S\) and \(C, B \in \Pi\). Then:
\[
C \text { is a } \Pi \text {-splitter for } B \quad \operatorname{iff}(B \neq C \wedge B \cap \operatorname{Pre}(C) \neq \varnothing \wedge \operatorname{Bottom}(B) \backslash \operatorname{Pre}(C) \neq \varnothing) .
\]

Proof: \(\Leftarrow\) : Assume \(B \neq C, B \cap \operatorname{Pre}(C) \neq \varnothing\) and \(\operatorname{Bottom}(B) \backslash \operatorname{Pre}(C) \neq \varnothing\). Then, there exist states \(t \in \operatorname{Bottom}(B) \backslash \operatorname{Pre}(C)\) and \(s \in B \cap \operatorname{Pre}(C)\). Since \(t \in \operatorname{Exit}(B)\) and \(t \notin \operatorname{Pre}(C)\), all direct successors of \(t\) are outside \(B \cup C\), and thus \(t \in B \backslash \operatorname{Pr} \mathrm{E}_{\Pi}^{*}(C)\). Thus, by Definition 7.134, \(C\) is a \(\Pi\)-splitter for \(B\).
\(\Rightarrow\) : Let \(C\) be a \(\Pi\)-splitter for \(B\). By Definition 7.134 of a \(\Pi\)-splitter, it follows that \(B \neq C\) and \(s \in B \cap \operatorname{Pre}(C)\) for some \(s\). It remains to show that \(B \backslash \operatorname{Pre} e_{\Pi}^{*}(C) \neq \varnothing\) implies \(\operatorname{Bottom}(B) \backslash \operatorname{Pre}(C) \neq \varnothing\). This is done by contraposition. Assume \(\operatorname{Bottom}(B) \backslash \operatorname{Pre}(C)=\) \(\varnothing\), i.e., \(\operatorname{Bottom}(B) \subseteq \operatorname{Pre}(C)\). Let \(u \in B\). Since there are no stutter cycles in \(T S\) and \(T S\) is finite, there exists a path fragment \(u \ldots t\) consisting of \(B\)-states only with \(t \in \operatorname{Bottom}(B)\). But, since \(\operatorname{Bottom}(B) \subseteq \operatorname{Pre}(C)\), we have \(t \in \operatorname{Pre}(C)\), and \(u \in \operatorname{Pre} e_{\Pi}^{*}(C)\). Thus, \(u \in B \cap \operatorname{Pr} e_{\Pi}^{*}(C)\). As this applies to any \(u \in B\), it follows \(B \subseteq \operatorname{Pr} e_{\Pi}^{*}(C)\), which contradicts that \(C\) is a \(\Pi\)-splitter of \(B\).

This result allows checking whether block \(C\) of partition \(\Pi\) is a \(\Pi\)-splitter for some \(B \in \Pi\), by investigating the direct predecessors of \(C\). For any state \(s \in C\), the list \(\operatorname{Pre}(s)\) is traversed and all states \(s^{\prime} \in \operatorname{Pre}(s)\) and blocks \(\left[s^{\prime}\right]_{\Pi}\) (not all states in the block, just the block itself) are marked. Then, for each block \(B \in \Pi, B \neq C\), we have
\[
\begin{array}{lll}
C \text { is a } \Pi \text {-splitter for } B & \text { iff } & \begin{array}{l}
B \text { is marked and } \operatorname{Bottom}(B) \\
\text { contains an unmarked state }
\end{array}
\end{array}
\]

In this way, the \(C\)-stability of \(\Pi\) can be checked in time \(\mathcal{O}(|\operatorname{Pre}(C)|)\) when using appropriate data structures (such as lists) for \(\Pi\) and \(\operatorname{Bottom}(B)\) for \(B \in \Pi\). For instance, a list representation of \(\Pi\) with pointers from \(\Pi\) 's blocks to the states in the blocks and vice versa, enables generating a list consisting of (pointers to) the marked blocks in time \(\mathcal{O}(|\operatorname{Pre}(C)|)\).

To check whether there is a marked block \(B\) where \(\operatorname{Bottom}(B)\) contains an unmarked state, a linear search through the list of marked blocks suffices. If \(B\) is the current (marked) block, then we traverse the list representing \(\operatorname{Bottom}(B)\) until an unmarked state (i.e., a state \(t \notin \operatorname{Pre}(C)\) ) has been found (or the end of the list has been reached). The time complexity of this strategy is \(\mathcal{O}(|\operatorname{Pre}(C)|)\), since the number of marked blocks is at most \(|\operatorname{Pre}(C)|\) and the total number of elements to be considered when traversing the lists for the exit states is bounded by \(|\operatorname{Pre}(C)|+1\). (The " +1 " covers the case where \(t \notin \operatorname{Pre}(C)\) has been reached.) Ranging over all blocks \(C \in \Pi\), the stability of \(\Pi\) can be checked in time \(\mathcal{O}(M)\) since \(M=\sum_{C \in \Pi}|\operatorname{Pre}(C)|\), where (as before) \(M\) denotes the number of edges in the state graph of \(T S\).

Moreover, if \(C\) is a \(\Pi\)-splitter (i.e., \(\Pi\) is not \(C\)-stable), then the above technique returns a block \(B \in \Pi\) such that \(C\) is a \(\Pi\)-splitter for \(B\). It remains to explain how the refinement of \(B\) into the two subblocks:
\[
B_{1}=B \cap \operatorname{Pre}_{\Pi}^{*}(C) \text { and } B_{2}=B \backslash \operatorname{Pre}_{\Pi}^{*}(C)
\]
can be realized in time \(\mathcal{O}(M)\). For this, we proceed as follows. We start with two new (variables for the) blocks \(B_{1}\) and \(B_{2}\), which are initially empty. On traversing the list of all exit states of \(B\), any marked state \(s \in \operatorname{Bottom}(B)\) is added to \(B_{1}\) and any unmarked state \(s \in \operatorname{Bottom}(B)\) is added to \(B_{2}\). (Recall that exactly the states in \(\operatorname{Pre}(C)\) are marked.) We then apply a standard graph algorithm to determine a reversed topological order \(s_{1}, \ldots, s_{k}\) of the nonexit states in \(B\), i.e., \(B \backslash \operatorname{Bottom}(B)=\left\{s_{1}, \ldots, s_{k}\right\}\) and \(i>j\) whenever there is a transition \(s_{i} \rightarrow s_{j}\). Note that such a topological order exists since \(T S\) has no stutter cycles. It can be computed in \(\mathcal{O}(M+|S|)\). The nonexit states of \(B\) are traversed in reversed topological order. On encountering state \(s=s_{i}\) for some \(0<i \leqslant k\), we perform the following. If \(s\) is unmarked, i.e., \(s \in B \backslash \operatorname{Pre}(C)\), the list \(\operatorname{Post}(s)\) is traversed. If \(t \in \operatorname{Post}(s) \cap B_{1}\) is encountered for some \(t\), then \(s\) is added to \(B_{1}\); otherwise, \(s\) is added to \(B_{2}\). This is justified by the observation that
- either \(s\) is marked in which case \(s \in B \cap \operatorname{Pre}(C)\),
- or \(s\) is unmarked and \(s \rightarrow t\) for some \(t \in B_{1}\), in which case either \(t \in \operatorname{Bottom}(B) \cap\) \(\operatorname{Pre}(C)\) or \(t \in\left\{s_{1}, \ldots, s_{i-1}\right\}\) and there exists a path fragment \(t \ldots v\) consisting of states in \(B_{1}\) with \(v \in \operatorname{Bottom}(B) \cap \operatorname{Pre}(C)\).

If all states \(s \in\left\{s_{1}, \ldots, s_{k}\right\}=B \backslash \operatorname{Bottom}(B)\) have been considered, then \(B_{1}\) consists of all states in \(B \cap \operatorname{Pr}_{\Pi}^{*}(C)\), while \(B_{2}\) contains all states \(B \backslash \operatorname{Pr} e_{\Pi}^{*}(C)\).

The above yields that there is an algorithm with time complexity \(\mathcal{O}(M)\) that checks
whether for a given partition \(\Pi\) there is a pair \((B, C)\) of blocks in \(\Pi\) such that \(C\) is a \(\Pi\)-splitter for \(C\), and returns such a pair \((B, C)\), if there is some. Moreover:

\section*{Lemma 7.142. Time Complexity of the Refinement Operator}

Refine \(_{\approx}(B, C)\) can be computed in time \(\mathcal{O}(M)\).

Hence, each iteration in Algorithm 37 causes the \(\operatorname{cost} \mathcal{O}(M)\). As each iteration yields a proper refinement of the current partition \(\Pi\), the number of iterations is at most \(|S|\). We conclude:

\section*{Theorem 7.143. Time Complexity of Algorithm 37}

The stutter bisimulation quotient of TS can be computed with Algorithm 37 in time \(\mathcal{O}(|S|\). \((|A P|+M)\) ), under the assumption that \(M \geqslant|S|\), where \(M\) is the number of edges in the state graph \(G(T S)\).

Proof: The costs of computing the initial partition are \(\mathcal{O}(|S| \cdot|A P|)\). At most \(|S|\) iterations take place. Each refinement in an iteration is in \(\mathcal{O}(M)\). Checking the loop condition can also be done in \(\mathcal{O}(M)\).

Although irrelevant for the asymptotic time complexity, the following observation can be helpful in practice to increase the efficiency of the splitter search: it provides a criterion to skip checking the \(\Pi\)-splitter-condition for certain blocks \(C\).

\section*{Lemma 7.144. Stability Criterion}

Let TS be a finite transition system without stutter cycles and let \(\Pi\) and \(\Pi^{\prime}\) partitions such that \(\Pi^{\prime}\) is finer than \(\Pi\) and \(\operatorname{Bottom}(\Pi)=\operatorname{Bottom}\left(\Pi^{\prime}\right)\). Then: there is no common block \(C\) of \(\Pi\) and \(\Pi^{\prime}\) such that \(\Pi\) is stable for \(C\), while \(\Pi^{\prime}\) is not.

Proof: By contraposition. Assume \(C \in \Pi \cap \Pi^{\prime}\) and \(C\) is a \(\Pi^{\prime}\)-splitter for some block \(B^{\prime} \in \Pi^{\prime}\), and \(\Pi\) is stable for \(C\). By Lemma 7.141, \(C \neq B^{\prime}\), there exists \(s \in B^{\prime} \cap \operatorname{Pre}(C)\), and there exists \(t \in \operatorname{Bottom}\left(B^{\prime}\right) \backslash \operatorname{Pre}(C)\). Since \(\Pi\) refines \(\Pi^{\prime}\), there exists a block \(B \in \Pi\) which subsumes \(B^{\prime} . C \neq B\), since otherwise \(C=B^{\prime}\) as \(C \in \Pi \cap \Pi^{\prime}\), and \(s \in B \cap \operatorname{Pre}(C)\). From \(\operatorname{Bottom}(\Pi)=\operatorname{Bottom}\left(\Pi^{\prime}\right)\), we derive that \(t \in \operatorname{Bottom}\left(B^{\prime}\right) \subseteq \operatorname{Bottom}(B)\). By Lemma 7.141, \(C\) is a \(\Pi\)-splitter for \(B\). This, however, contradicts the assumption that \(\Pi\) is stable for \(C\).

Quotienting for Divergence-Sensitive Stutter Bisimulation The computation of the equivalence classes under \(\approx^{d i v}\) is essentially based on the following steps. Let \(T S=\) \((S, A c t, \rightarrow, I, A P, L)\) be a finite transition system. As an initial step, \(T S\) is transformed into a divergence-sensitive transition system \(\overline{T S}\) such that the equivalence classes under \(\approx\) in \(\overline{T S}\) coincide with the equivalence classes under \(\approx^{d i v}\) in \(T S\). (Recall that \(T S\) is divergencesensitive if \(\approx_{T S}\) is divergence-sensitive, and hence coincides with \(\approx_{T S}^{d i v}\).) Applying Algorithm 37 to \(\overline{T S}\) then provides the quotient space \(S / \approx^{d i v}\) of \(T S\).

The divergence-sensitive transition system \(\overline{T S}\) is obtained as follows. \(T S\) is extended with a new state \(s_{d i v}\) that is not stutter-bisimilar to any other state in \(T S\). This is established by equipping \(s_{d i v}\) with a unique label, div say. The goal is to define the transitions in \(\overline{T S}\) such that the \(\approx_{T S}^{d i v}\)-divergent states of \(T S\) have a path fragment of the form \(s v_{1} \ldots v_{m} s_{d i v}\) in \(\overline{T S}\) such that
\[
s \approx_{T S}^{d i v} v_{1} \approx_{T S}^{d i v} \ldots \approx_{T S}^{d i v} v_{m}
\]

That is to say, the new state \(s_{d i v}\) can only be reached from a \(\approx_{T S}^{d i v}\)-divergent state. Since \(T S\) is finite, every \(\approx_{T S}^{d i v}\)-divergent path contains a cycle containing only states that are equivalent under \(\approx_{T S}^{d i v}\). By Lemma 7.138 on page 570 , these are exactly the stutter cycles! So, \(\overline{T S}\) is obtained from \(T S\) by inserting transitions from \(s\) to \(s_{d i v}\) for any state \(s\) that lies on a stutter cycle.

\section*{Definition 7.145. Divergence-Sensitive Expansion \(\overline{T S}\)}

The divergence-sensitive expansion of finite transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\) is:
\[
\overline{T S}=\left(S \cup\left\{s_{d i v}\right\}, A c t \cup\{\tau\}, \rightarrow, I, A P \cup\{\operatorname{div}\}, \bar{L}\right)
\]
where \(s_{d i v} \notin S, \rightarrow\) extends the transition relation of TS by the transitions \(s_{d i v} \xrightarrow{\tau} s_{d i v}\) and \(s \xrightarrow{\tau} s_{\text {div }}\) for every state \(s \in S\) on a stutter cycle in \(T S\), and \(\bar{L}(s)=L(s)\) if \(s \in S\) and \(\bar{L}\left(s_{d i v}\right)=\{\operatorname{div}\}\).

\section*{Example 7.146. Divergence-Sensitive Expansion}

Consider the transition system \(T S\) over \(A P=\{a\}\) in Figure 7.49 (upper part). Its divergence-sensitive expansion is depicted below it. It is not difficult to establish that
\[
S / \approx_{T S}^{d i v}=\left\{\left\{s_{0}, s_{1}\right\},\left\{s_{2}\right\},\left\{s_{3}\right\}\right\}=S / \approx_{\overline{T S}}
\]

Hence, the \(\approx\) equivalence classes of \(\overline{T S}\) with respect to \(\overline{A P}=\{a\), div \(\}\) correspond to the equivalence classes of \(T S\) under \(\approx^{d i v}\) for \(A P=\{a\}\). Note that we ignored the stutter bisimulation equivalence class \(\left\{s_{d i v}\right\}\) in \(\overline{T S}\).

The following theorem shows that \(\approx^{d i v}\) in \(T S\) coincides with \(\approx\) in \(\overline{T S}\).


Figure 7.49: Transition system \(T S\) (upper) and its divergence-sensitive expansion \(\overline{T S}\) (lower).

\section*{Theorem 7.147. Connection between TS and \(\overline{T S}\)}

For finite transition system TS:
1. \(\overline{T S}\) is divergence-sensitive, and
2. for all states \(s_{1}, s_{2} \in S: s_{1} \approx_{T S}^{d i v} s_{2}\) if and only if \(s_{1} \approx_{\overline{T S}} s_{2}\).

Proof: We first show that
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid s_{1} \approx_{T S}^{d i v} s_{2}\right\} \cup\left\{\left(s_{d i v}, s_{d i v}\right)\right\}
\]
is a stutter bisimulation for \(\overline{T S}\). Clearly, all states in \(\mathcal{R}\) are equally labeled. It remains to check that for \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) we have that \(s_{2}\) simulates transitions of \(s_{1}\). This is evident for \(\left(s_{d i v}, s_{d i v}\right) \in \mathcal{R}\); consider the remaining cases.

Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1}^{\prime} \in \operatorname{Post}_{\overline{T S}}\left(s_{1}\right)\) such that \(\left(s_{1}, s_{1}^{\prime}\right) \notin \mathcal{R}\). Distinguish two cases:
- If \(s_{1}^{\prime} \in S\) (i.e., \(s_{1}^{\prime} \neq s_{d i v}\) ), then \(s_{1}^{\prime} \in \operatorname{Post}_{T S}\left(s_{1}\right)\) and \(s_{1} \not \nsim T S_{d i v}^{d} s_{1}^{\prime}\). Since \(s_{1} \approx_{T S}^{d i v} s_{2}\), there exists a path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) in \(T S\) with
\[
s_{2} \approx_{T S}^{d i v} u_{1} \approx_{T S}^{d i v} \ldots \approx_{T S}^{d i v} u_{n} \quad \text { and } \quad s_{1}^{\prime} \approx_{T S}^{d i v} s_{2}^{\prime} .
\]

Thus, \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) is a path fragment satisfying the property \(\left(s_{2}, u_{j}\right) \in \mathcal{R}\), for \(0<j \leqslant n\), and \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
- If \(s_{1}^{\prime}=s_{d i v}\), then \(s_{1}\) is on a stutter cycle in \(T S\), and thus \(s_{1}\) is divergent. As \(s_{1} \approx_{T S}^{d i v} s_{2}\), state \(s_{2}\) is divergent. Since \(T S\) is finite, there exists a path fragment \(s_{2} u_{1} \ldots u_{n}\) such that
\[
s_{2} \approx_{T S}^{d i v} u_{1} \approx_{T S}^{d i v} \ldots \approx_{T S}^{d i v} u_{n}
\]
where \(u_{n}\) is on a stutter cycle. Therefore, \(u_{n} \rightarrow s_{d i v}\) is a transition in \(\overline{T S}\) and \(s_{2} u_{1} \ldots u_{n} s_{d i v}\) is a path fragment in \(\overline{T S}\) that matches \(s_{1} \rightarrow s_{1}^{\prime}\).

As a next step, we show that \(\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid s_{1} \approx_{\overline{T S}} s_{2}\right\}\) is a divergence-sensitive stutter bisimulation for \(T S\). Obviously, \(\mathcal{R}\) only contains equally labeled states; condition (2) of being a stutter bisimulation follows from the following reasoning:
- Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1}^{\prime} \in \operatorname{Post}_{T S}\left(s_{1}\right)\) such that \(\left(s_{1}, s_{1}^{\prime}\right) \notin \mathcal{R}\). Then, \(s_{1} \approx_{\overline{T S}} s_{2}\), \(s_{1}^{\prime} \in \operatorname{Post}_{\overline{T S}}\left(s_{1}\right)\), and \(s_{1} \not \overbrace{\overline{T S}} s_{1}^{\prime}\). Thus, there exists a path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) in \(\overline{T S}\) with
\[
s_{2} \approx_{\overline{T S}} u_{1} \approx_{\overline{T S}} \ldots \approx_{\overline{T S}} u_{n} \quad \text { and } \quad s_{1}^{\prime} \approx_{\overline{T S}} s_{2}^{\prime}
\]

As \(s_{1}^{\prime} \neq s_{d i v}\), we have \(s_{2}^{\prime} \neq s_{\text {div }}\). Therefore, \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) is a path fragment in \(T S\) that matches \(s_{1} \rightarrow s_{1}^{\prime}\).

It remains to check the divergence sensitivity of \(\mathcal{R}\). Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) such that \(s_{1}\) is \(\mathcal{R}\)-divergent. Then, there exists a path fragment \(s_{1} v_{1} \ldots v_{m} s_{1}^{\prime} u_{1} \ldots u_{n} s_{1}^{\prime}\) such that:
\[
s_{1} \approx_{\overline{T S}} v_{1} \approx_{\overline{T S}} \ldots \approx_{\overline{T S}} v_{m} \approx_{\overline{T S}} s_{1}^{\prime} \approx_{\overline{T S}} u_{1} \approx_{\overline{T S}} \ldots \approx_{\overline{T S}} u_{n}
\]

In particular, \(s_{1}^{\prime}\) is on a stutter cycle. Therefore, \(s_{1}^{\prime} \rightarrow s_{d i v}\) is a transition in \(\overline{T S}\). As \(\left(s_{1}, s_{2}\right) \in \mathcal{R}, s_{1} \approx_{\overline{T S}} s_{2}\). Together with \(s_{1} \approx_{\overline{T S}} s_{1}^{\prime}\), we get \(s_{1}^{\prime} \approx_{\overline{T S}} s_{2}\). Moreover, \(s_{d i v} \in \operatorname{Post}_{\overline{T S}}\left(s_{1}^{\prime}\right)\), and \(s_{1}^{\prime} \not \chi_{\overline{T S}} s_{d i v}\). Hence, there exists a path fragment \(s_{2} w_{1} \ldots w_{k} s_{2}^{\prime} s_{d i v}\) such that
\[
s_{2} \approx_{\overline{T S}} w_{1} \approx_{\overline{T S}} \ldots \approx_{\overline{T S}} w_{k} \approx_{\overline{T S}} s_{2}^{\prime} \quad \text { and } \quad s_{2}^{\prime} \rightarrow s_{d i v}
\]

Note, since only state \(s_{d i v}\) is labeled with \(\operatorname{div}\), if \(s \approx_{\overline{T S}} s_{d i v}\) then \(s=s_{d i v}\).
Since \(s_{2}^{\prime} \rightarrow s_{d i v}\) is a transition in \(\overline{T S}, s_{2}^{\prime}\) belongs to a stutter cycle \(s_{2}^{\prime} t_{1} \ldots t_{\ell} s_{2}^{\prime}\). By Lemma 7.138 (page 570), \(s_{2}^{\prime} \approx_{T S} t_{1} \approx_{\overline{T S}} \ldots \approx_{\overline{T S}} t_{\ell}\). As \(s_{2} \approx_{T S} s_{2}^{\prime}\), it follows that \(s_{2}\) is \(\mathcal{R}\)-divergent.

Finally, we consider the divergence sensitivity of \(\overline{T S}\). By definition, \(\overline{T S}\) is divergencesensitive if \(\approx_{\overline{T S}}\) is divergence-sensitive. This follows from the fact that \(\approx_{\overline{T S}}\) coincides with \(\approx_{T S}^{d i v}\).

Theorem 7.147 shows that the quotient space \(S / \approx\) of transition system \(T S\) can be computed by applying Algorithm 37 to the divergence-sensitive expansion \(\overline{T S}\). In order to
construct \(\overline{T S}\), all states on a stutter cycle must be determined, as these states need to be equipped with a transition to \(s_{d i v}\). This can be done by generating the nontrivial SCCs in the digraph that is obtained from \(G(T S)\) by only considering stutter steps. Let \(G_{\text {stutter }}(T S)=(V, E)\) with \(V=S\), the state space of \(T S\) and \(E=\{(s, t) \in S \times S \mid L(s)=\) \(L(t)\}\).

Summarizing, the required steps to compute the quotient transition system \(T S / \approx^{d i v}\) are:
1. Construct the expansion \(\overline{T S}\) by determining the SCCs in \(G_{\text {stutter }}(T S)\), and inserting transitions \(s \rightarrow s_{d i v}\) and \(s_{d i v} \rightarrow s_{d i v}\) for any state \(s\) in a nontrivial SCC of \(G_{s t u t t e r}\).
2. Apply Algorithm 37 to \(\overline{T S}\) to obtain the quotient space
\[
S / \approx_{T S}^{d i v}=S / \approx_{\overline{T S}}
\]

The transition system \(\overline{T S} / \approx\) is constructed as follows. Any \(C \in S / \approx{ }^{d i v}\) that contains an initial state of \(T S\) is identified as initial state. The labeling of \(C \in S / \approx^{d i v}\) equals the labeling of any \(s \in C\). All transitions \(s \rightarrow s^{\prime}\) with \(s \not \chi_{T S}^{d i v} s^{\prime}\) are lifted to transitions of the corresponding state classes.
3. \(T S / \approx{ }^{d i v}\) is obtained from \(\overline{T S} / \approx\) as follows. The transitions \(s \rightarrow s_{d i v}\) in \(\overline{T S}\) are replaced by self-loops \([s]_{d i v} \rightarrow[s]_{d i v}\), and the state \(s_{d i v}\) is deleted.

\section*{Example 7.148.}

Consider again the transition system \(T S\) over \(A P=\{a\}\) in Figure 7.49 (upper part) and its divergence-sensitive expansion \(\overline{T S}\) (see lower part of the same figure). The quotient transition system \(\overline{T S} / \approx\) is depicted in Figure 7.50 (upper part). The corresponding transition system \(T S / \approx\) is obtained by turning all transitions to \([s]_{d i v}\) into self-loops at the source state, and deleting \([s]_{\text {div }}\). This yields the transition system depicted in the lower part of Figure 7.50.

The costs of the described technique are dominated by the costs of Algorithm 37 and the computation of the SCCs in digraph \(G_{\text {stutter }}\). Theorem 7.143 on page 574 yields:

Theorem 7.149. Time Complexity of Constructing TS/ \(\approx^{\text {div }}\)
The quotient space of TS under \(\approx^{\text {div }}\) can be computed in time \(\mathcal{O}((|S|+M)+|S| \cdot(|A P|+\) \(M)\) ), under the assumption that \(M \geqslant|S|\), where \(M\) denotes the number of edges in the state graph \(G(T S)\).


Figure 7.50: Example of computing the quotient system under \(\approx^{d i v}\).

\subsection*{7.9 Summary}
- A bisimulation equivalence only relates a pair of states if both states are equally labeled and can mutually mimic their outgoing transitions. In a simulation preorder it suffices that one state can mimic the other, but not necessarily the reverse.
- Bisimulation is strictly finer than simulation equivalence and trace equivalence.
- In general, simulation and trace equivalence are incomparable. For transition systems without terminal states, simulation equivalence is strictly finer than trace equivalence.
- For \(A P\)-deterministic transition systems, bisimulation, simulation, and trace equivalence coincide.
- Bisimulation equivalence and the simulation order can be computed for finite transition systems in polynomial time. The problem of checking trace equivalence or trace inclusion is PSPACE-complete.
- For finite transition systems without terminal states, bisimulation equivalence corresponds to CTL* and CTL equivalence. (This result is considerable, since CTL* includes the two incomparable logics CTL and LTL.)
- A logical characterization of the simulation preorder can be provided by the universal or the existential fragments of CTL* and CTL.
- Stutter trace equivalence is a variant of trace equivalence that abstracts from stutter steps, i.e, transitions between equally labeled states. It coincides with LTL \(_{\backslash \bigcirc}\) equivalence.
- Stutter bisimulation is a variant of bisimulation where transitions can be mimicked by path fragments (rather than by single transitions). It is incomparable to stutter trace equivalence.
- A state is divergent if a path consisting only of stutter steps emanates from it. Divergence-sensitive stutter bisimulation is a stutter bisimulation that distinguishes divergent from nondivergent states.
- Divergence-sensitive stutter bisimulation is strictly finer than stutter trace equivalence and coincides with \(\mathrm{CTL}_{{ }^{*}-\text { - and }} \mathrm{CTL}_{\backslash \bigcirc}\) equivalence.
- Divergence-sensitive bisimulation can be computed for finite transition systems in polynomial time.

Figure 7.51 on page 580 summarizes the results concerning the treated notions of bisimulation. For the sake of simplicity, the costs of the initialization phase of the quotienting algorithms have been omitted.
\begin{tabular}{|c|c|c|c|c|}
\hline & bisimulation equivalence & simulation order & stutter equivalence with divergence & trace equivalence \\
\hline preservation of temporal-logical properties & \[
\begin{gathered}
\text { CTL* } \\
\text { CTL }
\end{gathered}
\] & \[
\begin{gathered}
\forall \mathrm{CTL}^{*} / \exists \mathrm{CTL} * \\
\forall \mathrm{CTL} / \exists \mathrm{CTL}
\end{gathered}
\] & \[
\begin{aligned}
& \text { CTL }_{\stackrel{*}{*}}^{*} \\
& \text { CTL }_{\backslash \bigcirc}
\end{aligned}
\] & \begin{tabular}{l}
LTL \\
(LT properties)
\end{tabular} \\
\hline checking equivalence & PTIME & PTIME & PTIME & PSPACEcomplete \\
\hline graph minimization & \[
\begin{gathered}
\text { PTIME } \\
\mathcal{O}(M \cdot \log |S|)
\end{gathered}
\] & \[
\begin{aligned}
& \text { PTIME } \\
& \mathcal{O}(M \cdot|S|)
\end{aligned}
\] & \[
\begin{gathered}
\text { PTIME } \\
\mathcal{O}(M \cdot|S|)
\end{gathered}
\] & - \\
\hline
\end{tabular}

Figure 7.51: Overview of equivalences on transition systems.

\subsection*{7.10 Bibliographic Notes}

Bisimulation and simulation. The notions of bisimulation and simulation in this chapter are state-based versions of the action-based bisimulation and simulation. Action-based
bisimulation has been brought up independently by Milner [296] and Park [322]. Actionbased simulation originates from Milner [295]. Milner [296] also introduced observational bisimulation, a weak bisimulation that abstracts away from internal steps (i.e., \(\tau\)-labeled transitions). (Exercise 7.22 considers a state-based variant of observational equivalence.) Stutter bisimulation is the state-based variant of branching bisimulation introduced by van Glabbeek and Weijland [406]. As noticed by Groote and Vaandrager [176], divergencesensitive stutter bisimulation coincides with stutter equivalence introduced by Browne, Clarke and Grumberg [67]. The alternative characterizations of stutter bisimulation by norm functions for finite and infinite transition systems is due to Namjoshi [311]. (Theorem 7.126 treats only finite transition systems.) Griffioen and Vaandrager [175] apply a (slightly) different notion of norm functions to define simulation relations. A comprehensive comparison of various trace-based and (bi)simulation relations has been provided by van Glabbeek [404, 405]. The concept of bisimulation and simulation relations has been refined and studied in various directions (such as axiomatization, refinement, coinduction, domain theoretic approaches); see, e.g., [298, 1, 4, 279, 397, 101]. The use of simulation relations for abstraction has been studied in, e.g., [90, 277, 133, 109, 105].

Logical characterizations. The first logical characterization of bisimulation equivalence has been provided in the action-based setting by Hennessy and Milner [197] for a modal logic, called Hennessy-Milner logic, with an action-labeled next step operator. Browne, Clarke and Grumberg [67] showed that state-based bisimulation coincides with CTL- and CTL* equivalence; see Theorem 7.20, and the analogous result for divergence-sensitive stutter bisimulation and \(\mathrm{CTL}_{\backslash \bigcirc} / \mathrm{CTL}_{\backslash \bigcirc}^{*}\); see Theorem 7.128. Kucera and Schnoebelen [247] recently presented a refinement of the latter result. De Nicola and Vaandrager [314] studied the connection between temporal logics and variants of stutter bisimulation in the action- and state-based setting. The logical characterizations of the simulation order by means of the existential and universal fragment of CTL and CTL* is due to Clarke, Grumberg and Long [90]. The observation that any LTL formula which induces a stutterinsensitive LT property can be expressed in LTL \({ }_{\backslash \bigcirc}\) (see Exercise 7.20) is by Peled and Wilke [329]; see also Etessami [147]. Etessami [146] defined a variant LTL that is as expressive as stutter-invariant \(\omega\)-regular languages.

Quotienting algorithms. The first partition refinement algorithm for bisimulation is due to Kanellakis and Smolka [231]. The refined version with the ternary refinement operator (see Algorithm 32) originates from Paige and Tarjan [318]. Kanellakis and Smolka [231] also showed that the trace equivalence problem is PSPACE-complete; see Theorem 7.46. The initial algorithm for calculating the simulation order has been suggested by Cleaveland, Parrow, and Steffen [97]. Henzinger, Henzinger, and Kopke [198] developed a more efficient algorithm; see Algorithm 36. Other algorithms for computing the simulation order have been suggested by, among others, Tan and Cleaveland [385] and Bustan and Grumberg [76]. A survey of bisimulation- and simulation-quotienting algorithms has been
given by Cleaveland and Sokolsky [98]. The quotienting algorithms for stutter bisimulation quotient (see Algorithm 37) go back to Groote and Vaandrager [176]. Moller and Smolka [302] discuss the computational complexity of checking the bisimilarity of several classes of processes. Thanks to extensive studies by Fisler and Vardi [152, 153, 154], it is known that bisimulation minimization for LTL model checking and invariant verification leads to drastic state space reductions (up to logarithmic savings) but at a time penalty: the time to minimize and model-check the resulting quotient significantly exceeds the time to verify the original transition system. In this monograph, we considered quotienting algorithms for finite transition systems. Semialgorithms for bisimulation quotienting of infinite transition systems (provided the bisimulation quotient is finite) have been presented by Bouajjani, Fernandez, and Halbwachs [60] and Lee and Yannakakis [266]. Henzinger, Henzinger, and Kopke [198] introduced a semialgorithmic approach for the simulation order to treat infinite state spaces. A classification and algorithms for infinite systems with finite quotients for other equivalences have been given by Henzinger, Majumdar, and Raskin [199].

For further reading we refer to the Handbook of Process Algebra [46] (and the literature mentioned therein) which treats many aspects of (bi)simulations and other implementation relations.

\subsection*{7.11 Exercises}

Exercise 7.1. Consider the four transition systems in Figure 7.52. Determine for each pair \(\left(T S_{i}, T S_{j}\right)\) for \(0<i \neq j \leqslant 4\) of these transition systems whether they are bisimilar. Justify your answer by either providing the bisimulation relation for \(\left(T S_{i}, T S_{j}\right)\) or a CTL formula \(\Phi\) such that \(T S_{i} \models \Phi\) and \(T S_{j} \not \models \Phi\).

Exercise 7.2. Let \(T S\) be a transition system.
(a) Show that the union of bisimulation relations for \(T S\) is a bisimulation for \(T S\).
(b) Show that the union of simulation relations for \(T S\) is a simulation for \(T S\).

Exercise 7.3. The goal of this exercise is to discuss the role of AP-determinism (see Definition 7.65) for the connection between (bi)simulation and trace relations. In particular, this exrecise completes the proofs for Theorem 7.64 on page 511 and Corollary 7.72 on page 514.

Suppose that \(T S_{1}\) and \(T S_{2}\) are transition systems over AP that are both AP-deterministic.


transition system \(T S_{3}\)

transition system \(T S_{4}\)

Figure 7.52: Four transition systems for Exercises 7.1, 7.11, 7.17, 7.21 and 7.22.
(a) Prove the equivalence of the following statements (1), (2), (3) and (4):
(1) \(T S_{1} \sim T S_{2}\)
(2) \(T S_{1} \simeq T S_{2}\)
(3) \(\operatorname{Traces}_{\text {fin }}\left(T S_{1}\right)=\operatorname{Traces}_{\text {fin }\left(T S_{2}\right)}\)
(4) \(\operatorname{Traces}\left(T S_{1}\right)=\operatorname{Traces}\left(T S_{2}\right)\)
(b) Prove or disprove: If \(\operatorname{Traces}_{f i n}\left(T S_{1}\right) \subseteq \operatorname{Traces}_{f i n}\left(T S_{2}\right)\) then \(T S_{1} \preceq T S_{2}\).
(c) Prove or disprove: If \(T S_{1} \preceq T S_{2}\) then \(\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right)\).
(d) Prove or disprove: If \(\operatorname{Traces}_{f i n}\left(T S_{1}\right) \subseteq \operatorname{Traces}_{f i n}\left(T S_{2}\right)\) then \(\operatorname{Traces}\left(T S_{1}\right) \subseteq \operatorname{Traces}\left(T S_{2}\right)\).

EXERCISE 7.4. Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a transition system. The relations \(\sim_{n} \subseteq S \times S\) are inductively defined by:
(a) \(s_{1} \sim_{0} s_{2}\) iff \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
(b) \(s_{1} \sim_{n+1} s_{2}\) iff
(i) \(L\left(s_{1}\right)=L\left(s_{2}\right)\),
(ii) for all \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) there exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(s_{1}^{\prime} \sim_{n} s_{2}^{\prime}\),
(iii) for all \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) there exists \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(s_{1}^{\prime} \sim_{n} s_{2}^{\prime}\).

Show that for finite \(T S\) it holds that \(\sim_{T S}=\bigcap_{n \geqslant 0} \sim_{n}\), i.e.,
\[
s_{1} \sim_{T S} s_{2} \text { if and only if } s_{1} \sim_{n} s_{2} \text { for all } n \geqslant 0 .
\]

Does this also hold for infinite transition systems?

\section*{Exercise 7.5.}
(a) Consider transition system \(T S\) shown in Figure 7.53. Apply the transformation \(T S \mapsto T S_{\text {act }}\) (see page 467) and check that \(\sim_{T S}\) coincides with action-based bisimulation \(\sim_{T S_{\text {act }}}^{A c t}\) on \(T S_{\text {act }}\).


Figure 7.53: Transition system TS for Exercise 7.5.
(b) Prove that for all states \(s_{1}, s_{2}\) in \(T S\),
\[
s_{1} \sim_{T S}^{A c t} s_{2} \text { if and only if } s_{1} \sim_{T S_{s t a t e}} s_{2}
\]
where \(T S_{\text {state }}\) is obtained from \(T S\) by the transformation defined on page 467 .

\section*{Exercise 7.6.}
(a) Give a transition system \(T S\) without terminal states and containing states \(s_{1}, s_{2}\) of \(T S\) such that \(s_{1} \not 三_{L T L} s_{2}\) and there is no LTL formula \(\varphi\) with \(s_{2} \models \varphi\) and \(s_{1} \not \vDash \varphi\). (See the remark on page 469.)
(b) Let \(T S_{1}\) and \(T S_{2}\) be transition systems over \(A P\) without terminal states such that \(T S_{1} \not \equiv_{C L T}\) \(T S_{2}\). Prove or disprove the claim: there exists a CTL formula \(\Phi\) over \(A P\) such that \(T S_{1} \models \Phi\) and \(T S_{2} \not \vDash \Phi\).


Figure 7.54: Transition system \(T S\) for Exercises 7.7, 7.8 and 7.23.

Exercise 7.7. Consider the transition system \(T S\) over \(A P=\{a, b\}\) shown in Figure 7.54.
(a) Determine the bisimulation equivalence \(\sim_{T S}\) and depict the bisimulation quotient system \(T S / \sim\).
(b) Provide CTL master formulae \(\Phi_{C}\) for each bisimulation equivalence class \(C\).

EXERCISE 7.8. Consider again the transition system \(T S\) with state space \(S=\left\{s_{1}, s_{2}, \ldots, s_{13}\right\}\) shown in Figure 7.54. Apply Algorithms 31 and 32 to compute the bisimulation quotient \(S / \sim_{T S}\).

ExERCISE 7.9. Let TS be a finite transition system over \(A P\) without terminal states. Provide an algorithm that computes for each bisimulation equivalence class \(C\) a CTL master formula (i.e., a CTL formula \(\Phi_{C}\) with \(\left.\operatorname{Sat}\left(\Phi_{C}\right)=C\right)\).
(Hint: There is a simple algorithm which relies on an extension of the partition refinement technique).

ExERCISE 7.10. Consider the transition systems \(T S_{1}\) and \(T S_{2}\) shown in Figure 7.55 . Apply Algorithms 31 (page 487) and 32 (page 489) to check whether \(T S_{1} \sim T S_{2}\).


Figure 7.55: Are \(T S_{1}\) (left) and \(T S_{2}\) (right) bisimilar (Exercise 7.10)?

Exercise 7.11. Consider the four transition systems \(T S_{1}, T S_{2}, T S_{3}, T S_{4}\) in Figure 7.52 on page 583. Check whether \(T S_{i} \preceq T S_{j}\) and \(T S_{i} \simeq T S_{j}\) for all indices \(i, j\). Justify your answer by either establishing a simulation for \(\left(T S_{i}, T S_{j}\right)\) or by providing a \(\forall C T L\) formula \(\Phi\) with \(T S_{j} \models \Phi\) and \(T S_{i} \not \models \Phi\) (if \(T S_{i} \npreceq T S_{j}\) ).

Exercise 7.12. Consider the transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) in Figure 7.56.
(a) Show that \(T S_{2} \preceq T S_{1}\) by providing a simulation relation for \(\left(T S_{2}, T S_{1}\right)\).
(b) Show that \(T S_{1} \npreceq T S_{2}\) by providing a \(\forall\) CTL formula \(\Phi_{\forall}\) and a \(\exists\) CTL formula \(\Phi_{\exists}\) such that \(T S_{1} \not \models \Phi_{\forall}\), but \(T S_{2} \models \Phi_{\forall}\) and \(T S_{1} \models \Phi_{\exists}\), but \(T S_{2} \not \models \Phi_{\exists}\).

\(T S_{2}:\)


Figure 7.56: \(T S_{2} \preceq T S_{1}\), but \(T S_{1} \npreceq T S_{2}\) (Exercise 7.12).

Exercise 7.13. Consider the transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) in Figure 7.57.
(a) Check whether \(T S_{1} \preceq T S_{2}, T S_{2} \preceq T S_{1}\) or \(T S_{1} \simeq T S_{2}\).


Figure 7.57: Transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) for Exercise 7.13.
(b) Consider the composite transition system \(T S=T S_{1} \oplus T S_{2}\). Depict the simulation quotient system \(T S / \preceq\) and verify that \(T S / \preceq\) and \(T S\) are simulation equivalent, but not bisimilar.

Exercise 7.14. Complete the proof of Theorem 7.76 by showing the correctness of the following two statements (a) and (b):
(a) If \(s_{1}\) and \(s_{2}\) are states in \(T S\) with \(s_{1} \preceq_{T S} s_{2}\) then for all \(\forall\) CTL \(^{*}\) state formulae \(\Phi: s_{2} \models \Phi\) implies \(s_{1} \models \Phi\).
(b) If \(\pi_{1}=s_{0,1} s_{1,1} s_{2, i} \ldots\) and \(\pi_{1}=s_{0,2} s_{1,2} s_{2,2} \ldots\) infinite path fragments in \(T S\) such that \(s_{j, 1} \preceq_{\text {TS }} s_{j, 2}\) for all \(j \geqslant 0\) then for all \(\forall\) CTL \(^{*}\) path formulae \(\varphi: \pi_{2} \models \varphi\) implies \(\pi_{1} \models \varphi\).

Exercise 7.15. Indicate for each of the following claims whether it is a correct statement for finite transition systems:
(a) \(\forall \mathrm{CTL}\) equivalence is finer than LTL equivalence.
(b) LTL equivalence is finer than \(\forall \mathrm{CTL}\) equivalence.
(c) \(\exists \mathrm{CTL}\) equivalence is finer than \(\forall \mathrm{CTL}\) equivalence.
(d) \(\exists \mathrm{CTL}\) equivalence is finer than LTL equivalence.
(e) \(\exists \mathrm{CTL}^{*}\) equivalence is finer than CTL equivalence.

Justify your answers.


Figure 7.58: Transition system \(T S\) over \(A P=\{a, b\}\) for Exercise 7.16.

Exercise 7.16. Consider the transition system \(T S\) as depicted in Figure 7.58. The initial states are irrelevant and have been omitted. Apply Algorithm 36 (page 526) to compute the simulation order \(\preceq_{\text {TS }}\).

Exercise 7.17. Which of the transition systems in Figure 7.52 on page 583 are stutter trace equivalent?

EXERCISE 7.18. Consider the transition systems depicted below, where the gray filling of a state indicates the propositions that hold in that state.

\(\{a\}\) \(\square\)
\(\{b\}\)

For each \(i, j \in\{1 \ldots 3\} \times\{1 \ldots 3\}, i \neq j\), determine whether \(T S_{i} \cong T S_{j}, T S_{i} \sqsubseteq T S_{j}\) or \(T S_{i} \nsubseteq T S_{j}\). Justify your answer.

Exercise 7.19. Consider the transition systems depicted below, where the gray filling of a state indicates the propositions that are hold in that state.


For each \(i, j \in\{1 \ldots 3\} \times\{1 \ldots 3\}, i \neq j\), determine whether \(T S_{i} \approx T S_{j}\) or \(T S_{i} \not \approx T S_{j}\). Justify your answer.

Exercise 7.20. Let \(\varphi\) be an LTL formula such that \(\operatorname{Words}(\varphi)\) is stutter-insensitive. Show that \(\varphi\) is equivalent to some \(\operatorname{LTL}_{\backslash \bigcirc}\) formula \(\psi\).

Exercise 7.21. Which of the transition systems in Figure 7.52 on page 583 are stutter bisimulation equivalent?

Exercise 7.22. Observational equivalence \(\approx_{o b s}\) is a slight variant of stutter bisimulation equivalence where state \(s_{2}\) is allowed to perform a path fragment
\[
\underbrace{s_{2} u_{1} \ldots u_{m}}_{\text {stutter steps }} \underbrace{v_{1} \ldots v_{k} s_{2}^{\prime}}_{\text {stutter steps }}
\]
with arbitrary stutter steps at the beginning and at the end and \(s_{1}^{\prime} \approx_{o b s} s_{2}^{\prime}\) to simulate a transition \(s_{1} \rightarrow s_{1}^{\prime}\) of an observational equivalent state \(s_{1}\). I.e., it is not required that \(s_{2}\) and states \(u_{i}\) are observationally equivalent, or that \(s_{2}^{\prime}\) and \(v_{i}\) are observationally equivalent. For the special case where \(s_{1} \rightarrow s_{1}^{\prime}\) is a stutter step the path fragment of length 0 (consisting of state \(s_{2}=s_{2}^{\prime}\) ) can be used to simulate \(s_{1} \rightarrow s_{1}^{\prime}\).

The formal definition of observational equivalence is as follows. Let \(T S_{1}\) and \(T S_{2}\) be two transition systems over \(A P\) with state spaces \(S_{1}\) and \(S_{2}\), respectively. A binary relation \(\mathcal{R} \subseteq S_{1} \times S_{2}\) is an observational bisimulation for \(\left(T S_{1}, T S_{2}\right)\) iff the following conditions are satisfied:
(A) Every initial state of \(T S_{1}\) is related to an initial state of \(T S_{2}\), and vice versa. That is,
\[
\forall s_{1} \in I_{1} \exists s_{2} \in I_{2} .\left(s_{1}, s_{2}\right) \in \mathcal{R} \quad \text { and } \forall s_{2} \in I_{2} \exists s_{1} \in I_{1} .\left(s_{1}, s_{2}\right) \in \mathcal{R}
\]
(B) For all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) the following conditions (1), (2) and (3) hold:
(1) If \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) then \(L_{1}\left(s_{1}\right)=L_{2}\left(s_{2}\right)\)
(2) If \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) then there exists a path fragment \(u_{0} u_{1} \ldots u_{n}\) such that \(n \geqslant 0\) and \(u_{0}=s_{2},\left(s_{1}^{\prime}, u_{n}\right) \in \mathcal{R}\) and, for some \(m \leqslant n, L_{2}\left(u_{0}\right)=L_{2}\left(u_{1}\right)=\ldots=L_{2}\left(u_{m}\right)\) and \(L_{2}\left(u_{m+1}\right)=L_{2}\left(u_{m+2}\right)=\ldots=L_{2}\left(u_{n}\right)\).
(3) If \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) then there exists a path fragment \(u_{0} u_{1} \ldots u_{n}\) such that \(n \geqslant 0\) and \(u_{0}=s_{2},\left(u_{n}, s_{2}^{\prime}\right) \in \mathcal{R}\) and, for some \(m \leqslant n, L_{1}\left(u_{0}\right)=L_{1}\left(u_{1}\right)=\ldots=L_{1}\left(u_{m}\right)\) and \(L_{1}\left(u_{m+1}\right)=L_{1}\left(u_{m+2}\right)=\ldots=L_{1}\left(u_{n}\right)\).
\(T S_{1}\) and \(T S_{2}\) are called observational-equivalent, denoted \(T S_{1} \approx_{o b s} T S_{2}\), if there exists an observational bisimulation for \(\left(T S_{1}, T S_{2}\right)\).
(a) Which of the transition systems in Figure 7.52 on page 583 are observational-equivalent?
(b) Show that \(T S_{1} \approx T S_{2}\) implies \(T S_{1} \approx_{o b s} T S_{2}\).
(c) Consider the two transition system \(T S_{1}\) (left) and \(T S_{2}\) (right) shown in Figure 7.59 where the colors stand for the state labels, e.g., we may deal with \(L_{1}\left(s_{i}\right)=L_{2}\left(t_{j}\right)=\{a\}\) for \(i \in\) \(\{1,2,3,4\}\) and \(j \in\{1,2,4\}\) and \(L_{1}\left(s_{6}\right)=L_{2}\left(t_{6}\right)=\varnothing\) and \(L_{1}\left(s_{7}\right)=L_{1}\left(s_{5}\right)=L_{2}\left(t_{7}\right)=\{b\}\).

\(T S_{1}\)


Figure 7.59: Transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right) for Exercise 7.22.
Show that \(T S_{1} \not \approx T S_{2}\) and \(T S_{1} \approx_{o b s} T S_{2}\).

Exercise 7.23. Consider the transition system TS shown in Figure 7.54 on page 585.
(a) Which states of \(T S\) are stutter-bisimilar (according to \(\approx_{T S}\) )? Depict the stutter bisimulation quotient system \(T S / \approx\).
(b) Which states of \(T S\) are divergence stutter bisimilar? Justify the equivalence of states (according to \(\approx_{T S}^{d i v}\) ) by providing a step-dependent norm function. Depict the quotient system \(T S / \approx^{d i v}\) under stutter bisimulation equivalence with divergence.
(c) Depict the divergence-sensitive expansion \(\overline{T S}\) and apply Algorithm 37 to compute the divergence stutter bisimulation equivalence classes.
(d) Provide \(\mathrm{CTL}_{\backslash \bigcirc}\) master formulae for the divergence stutter bisimulation equivalence classes.

Exercise 7.24. Let \(T S\) be a transition system with state space \(S\). Define functions \(\nu_{1}^{*}, \nu_{2}^{*}\) : \(S \times S \rightarrow \mathbb{N}\) such that \(\left(\approx_{T S}^{n}, \nu_{1}^{*}, \nu_{2}^{*}\right)\) is a normed bisimulation for \(T S\).

Exercise 7.25. Provide an example for a transition system \(T S\) where \(\approx_{T S}^{n}\) (normed bisimulation equivalence) is strictly finer than \(\approx_{T S}^{d i v}\).

Exercise 7.26. Provide the proof for Lemma 7.96 (page 537).

Exercise 7.27. Let CTL \(_{\backslash u}\) be the sublogic of CTL that does not permit the until operator. Similarly, CTL* \({ }_{\cup}\) means CTL without U. Which of the following statements are correct for finite transition systems?
(a) \(\mathrm{CTL}_{\backslash \cup}\) equivalence is finer than \(\mathrm{CTL}_{\backslash \bigcirc}\) equivalence.
(b) \(\mathrm{CTL}_{\backslash \cup}\) equivalence is finer than divergence-sensitive stutter trace equivalence.
(c) \(\mathrm{CTL}_{\backslash \bigcirc}\) equivalence is finer than \(\mathrm{LTL}_{\backslash \bigcirc}\) equivalence.
(d) Divergence-sensitive stutter bisimulation equivalence is finer than \(\mathrm{CTL}_{\backslash \cup}\) equivalence.
(e) Stutter trace equivalence is finer than \(\mathrm{CTL}_{\backslash \cup}\) equivalence.
(f) For AP-deterministic transition systems, stutter trace equivalence is finer than trace-equivalence.
(g) For \(A P\)-deterministic transition systems, trace equivalence is finer than \(\mathrm{CTL}_{\backslash_{\cup}}^{*}\) equivalence.

Exercise 7.28. Check the correctness of the following statement. If \(T S_{1}\) and \(T S_{2}\) are stutterbisimilar transition systems over \(A P\) that are divergence-sensitive then \(T S_{1} \approx d i v T S_{2}\). Provide either a proof or a counterexample.

Exercise 7.29. Let \(T S_{i}=\left(S_{i}, A c t i, \rightarrow i, I_{i}, A P, L_{i}\right), i=1,2\), be two transition systems. A stutter simulation for \(\left(T S_{1}, T S_{2}\right)\) is a relation \(\mathcal{R} \subseteq S_{1} \times S_{2}\) such that
(A) Each initial state of \(T S_{1}\) is \(\mathcal{R}\)-related to an initial state of \(T S_{2}\), that is, \(\forall s_{1} \in I_{1} \exists s_{2} \in\) \(I_{2} .\left(s_{1}, s_{2}\right) \in \mathcal{R}\).
(B) For all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) the following conditions (1) and (2) hold:
(1) \(L_{1}\left(s_{1}\right)=L_{2}\left(s_{2}\right)\).
(2) If \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\) with \(\left(s_{1}, s_{1}^{\prime}\right) \notin \mathcal{R}\), then there exists a finite path fragment \(s_{2} u_{1} \ldots u_{n} s_{2}^{\prime}\) with \(n \geqslant 0\) and \(\left(s_{1}, u_{i}\right) \in \mathcal{R}, i=1, \ldots, n\) and \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\).
\(T S_{1}\) is said to be stutter-simulated by \(T S_{2}\), denoted \(T S_{1} \preceq_{s t} T S_{2}\), iff there exists a stutter simulation for \(\left(T S_{1}, T S_{2}\right)\).
(a) Provide an example for transition systems \(T S_{1}, T S_{2}\) such that \(T S_{1}\) is not simulated by \(T S_{2}\), but \(T S_{1} \preceq_{s t} T S_{2}\).
(b) Provide an example for transition systems \(T S_{1}, T S_{2}\) such that \(T S_{1} \unlhd T S_{2}\) and \(T S_{1} \not \bigwedge_{s t} T S_{2}\).
(c) Provide an example for transition systems \(T S_{1}, T S_{2}\) such that \(\left.T S_{1}\right\urcorner \unlhd T S_{2}\) and \(T S_{1} \preceq_{s t} T S_{2}\).
(d) Provide an example for transition systems \(T S_{1}, T S_{2}\) such that \(T S_{1} \not \approx T S_{2}\), while \(T S_{1} \preceq_{s t} T S_{2}\) and \(T S_{2} \preceq_{s t} T S_{1}\).

A stutter simulation \(\mathcal{R}\) for \(\left(T S_{1}, T S_{2}\right)\) is called divergence-sensitive if for all pairs \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and each infinite path fragment \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots\) in \(T S_{1}\) with \(s_{0,1}=s_{1}\) and \(\left(s_{i, 1}, s_{2}\right) \in \mathcal{R}\) for all \(i \geqslant 0\) there exists a transition \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{j, 1}, s_{2}^{\prime}\right) \in \mathcal{R}\) for some \(j \geqslant 1\). We write
\[
T S_{1} \preceq_{s t}^{d i v} T S_{2}
\]
iff there exists a divergence-sensitive stutter simulation \(\mathcal{R}\) for \(\left(T S_{1}, T S_{2}\right)\).
(e) Provide an example for transition systems \(T S_{1}, T S_{2}\) such that \(T S_{1} \not \overbrace{}^{d i v} T S_{2}\), while \(T S_{1} \preceq_{s t}^{d i v} T S_{2}\) and \(T S_{2} \preceq_{s t}^{d i v} T S_{1}\).
(f) Show that \(T S_{1} \approx^{d i v} T S_{2}\) if and only if there exists a divergence-sensitive stutter simulation \(\mathcal{R}\) for \(\left(T S_{1}, T S_{2}\right)\) such that \(\mathcal{R}^{-1}=\left\{\left(s_{2}, s_{1}\right) \mid\left(s_{1}, s_{2}\right) \in \mathcal{R}\right\}\) is a divergence-sensitive stutter simulation for \(\left(T S_{2}, T S_{1}\right)\).
(g) Show that \(T S_{1} \approx d i v T S_{2}\) implies divergence stutter simulation equivalence of \(T S_{1}\) and \(T S_{2}\), i.e., \(T S_{1} \preceq_{s t}^{d i v} T S_{2}\) and \(T S_{2} \preceq_{s t}^{d i v} T S_{1}\).

Show that the universal fragment of \(\mathrm{CTL}_{\widehat{\circ}}^{*}\) yields a logical characterization of the stutter simulation order with divergence. For this, provide proofs for the following statements (h) and (i) where \(T S_{1}\) and \(T S_{2}\) are supposed to be finite transition systems without terminal states.
(h) If \(T S_{1} \preceq_{s t}^{d i v} T S_{2}\) and \(\Phi\) is a \(\forall \mathrm{CLL}_{\uparrow \bigcirc}^{*}\) formula with \(T S_{2} \models \Phi\), then \(T S_{1} \models \Phi\).
(i) Assume that for all \(\forall C T L_{\backslash}^{*}\) formulae \(\Phi\) we have \(T S_{2} \models \neg \Phi\) or \(T S_{1} \vDash \Phi\). Show that \(T S_{1} \preceq_{s t}^{d i v} T S_{2}\).

Exercise 7.30. Consider the following transition system TS:


Questions:
(a) Give the divergence-sensitive expansion \(\overline{T S}\).
(b) Determine the divergence stutter bisimulation quotient \((\overline{T S}) / \approx\). Give for each iteration the partition of the state space.
(c) Give \(T S / \approx^{d i v}\).
(d) Provide CTL \(\backslash \bigcirc\) master formulae for each divergence stutter bisimulation equivalence class.

\section*{Chapter 8}

\section*{Partial Order Reduction}

Consider the parallel composition of a number of processes \(\mathcal{P}_{1}\) through \(\mathcal{P}_{n}\). The size of the state space of \(\mathcal{P}_{1}\left\|\mathcal{P}_{2}\right\| \ldots \| \mathcal{P}_{n}\), where \(\|\) denotes some parallel composition operator, is exponential in the number \(n\) of processes. To check the validity of a linear-time property of this system requires an inspection of all states in the underlying transition system. In the simple setting where there are no synchronizations between the individual processesneither through shared variables nor via communication channels or the like - there are \(n\) ! different orderings of the interleaved execution of \(n\) local actions. The effect of concurrent actions, however, is often independent of their ordering. Consider, e.g., the assignments


Figure 8.1: Interleaving diamond for \(\alpha \|| | \beta\).
\(x:=x+1\) and \(y:=y-3\) in the concurrent system \(\mathcal{P}_{1}| | \mathcal{P}_{2}\), where \(x\) is a local variable of \(\mathcal{P}_{1}\), say, and \(y\) of \(\mathcal{P}_{2}\), and \(\|\|\) denotes the interleaving operator. It is evident that regardless of the ordering of these assignments, the result will be the same. This is illustrated in Figure
8.1, where actions \(\alpha\) and \(\beta\) denote the assignments of \(\mathcal{P}_{1}\) and \(\mathcal{P}_{2}\), respectively. Instead of analyzing the 2 ! orderings of \(x:=x+1\) and \(y:=y-3\) it may suffice to check just a single ordering. This is correct as long as the intermediate states reached after the execution of either \(\alpha\) or \(\beta\) (see states \(t\) and \(u\) in Figure 8.1), are irrelevant for the properties to be proved. Extending the simple example with a third process \(\mathcal{P}_{3}\) that, e.g., resets its variable \(z\) to 0 , yields following an analogous reasoning that it suffices to consider just one of the 3 ! possible orderings. This approach can be generalized for action sequences \(\alpha_{1} \alpha_{2} \ldots \alpha_{n}\) and \(\beta_{1} \beta_{2} \ldots \beta_{m}\) that are executed independently by processes \(\mathcal{P}_{1}\) and \(\mathcal{P}_{2}\). The transition system of \(\mathcal{P}_{1} \| \mid \mathcal{P}_{2}\) represents all interleavings of these action sequences, whereas a single path fragment respecting the ordering in these sequences suffices, provided the intermediate states are irrelevant.

Put in a nutshell, the aim of partial order reduction, the technique that is treated in this chapter, is to reduce the number of possible orderings that need to be analyzed for checking formulae stated in a temporal logic such as LTL or CTL*. This main concept is to reduce the state space of the transition system that needs to be analyzed. Thus, the idea is to replace the full transition system for \(\mathcal{P}_{1} \mid\left\|\mathcal{P}_{2}\right\|\|\ldots\| \mathcal{P}_{n}\) by a small fragment. Figure 8.2 illustrates this for two processes that execute the action sequences \(\alpha_{1} \alpha_{2}\) and \(\beta_{1} \beta_{2}\), respectively. The transition system on the left contains all possible interleavings, while the reduced transition system on the right just consists a single path that might serve as a representative for all possible interleavings. On increasing the number of concurrent processes, this effect becomes even more drastic - the size of the full transition system grows exponentially in the number of processes, whereas the reduced system consists of a single path that grows linear in \(n\).


Figure 8.2: Parallel execution \(\left(\alpha_{1} ; \alpha_{2}\right)\left|\left|\mid\left(\beta_{1} ; \beta_{2}\right)\right.\right.\).

To avoid peak memory requirements, such a reduced transition system is obtained without
the necessity of ever generating the entire transition system that incorporates all possible orderings. This is typically done by a static analysis of the high-level description of the concurrent system, e.g., the channel systems as introduced in Chapter 2.

Clearly, the reduction technique crucially relies on the assumption that all processes are fully autonomous, i.e., no synchronizations are involved via, e.g., shared variables or communication channels. Moreover, it is assumed that the property of interest does not depend on the intermediate states. To treat realistic systems where the processes may communicate and thus depend on one another, the partial order reduction approach attempts to identify path fragments of the full transition system which only differ in the order of the concurrently executed activities. Since such path fragments represent the same behavior, it seems obvious to restrict the analysis of state space to one (or a few) representatives of every possible interleaving.

Central for the partial order reduction approach is the notion of independent actions which will be introduced in Section 8.1. Section 8.2 treats the partial order reduction approach for model-checking LT properties that are specified by LTL \(_{\backslash \bigcirc}\) formulae. Section 8.3 deals with partial order reduction for \(\mathrm{CTL}_{\backslash \bigcirc}\) and \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\). The intuition behind the omission of the next step operator is that we have to abstract away from certain intermediate states, as we sketched for the reduction in Figure 8.2.

Throughout this chapter \(T S=(S, A c t, \rightarrow, I, A P, L)\) is a finite transition system without terminal states. As partial order reduction is most effective for concurrent systems that are "loosely" coupled, it is implicitly assumed that TS models an asynchronous concurrent system where processes interact, e.g., either by shared variables or channel communication. In a synchronous setting where concurrent processes evolve in a lockstep fashion, each global transition involves all processes and thus cannot be considered as independent. This assumption is not relevant for the theoretical considerations in this chapter, but is important when considering static analysis of concurrent programs to enable the detection of independent actions syntactically.

It is assumed that \(T S\) is action-deterministic. This entails that for any state \(s \in S\) and any action \(\alpha \in A c t, s\) has at most one outgoing transition with action label \(\alpha\). Formally, \(s \xrightarrow{\alpha} s^{\prime}\) and \(s \xrightarrow{\alpha} s^{\prime \prime}\) implies \(s^{\prime}=s^{\prime \prime}\). (The reader should not confuse this notion with \(A P\) determinism.) The assumption that \(T S\) is action-deterministic is not a severe restriction, since actions can always be renamed such that an action-deterministic transition system results. When, e.g., TS results from the parallel composition of several action-deterministic processes, indices may be used for the actions to indicate which process performs the action.

\subsection*{8.1 Independence of Actions}

Let us first introduce some notations that are used throughout the remainder of this chapter.

\section*{Notation 8.1. The Set of Actions Act \((s)\)}

For state \(s\) in transition system \(T S\) with action set Act, let \(\operatorname{Act}(s)=\{\alpha \in \operatorname{Act} \mid\) \(\left.\exists s^{\prime} . s \xrightarrow{\alpha} s^{\prime}\right\}\).

Thus, \(\operatorname{Act}(s)\) denotes the set of actions that are enabled in state \(s\). Since an actiondeterministic transition system is assumed, for any \(\alpha \in \operatorname{Act}(s)\) there is a unique \(\alpha\)-successor of \(s\), denoted by \(\alpha(s)\).

\section*{Notation 8.2. The States \(\alpha(s)\)}

For action-deterministic transition system \(T S, s\) a state in \(T S\) and \(\alpha \in \operatorname{Act}(s)\), let \(\alpha(s)\) denote the unique \(\alpha\)-successor of \(s\), i.e., \(s \xrightarrow{\alpha} \alpha(s)\). For action sequence \(\alpha_{1} \ldots \alpha_{n}\) with \(\alpha_{1} \in \operatorname{Act}(s)\) and \(\alpha_{i+1} \in \operatorname{Act}\left(s_{i}\right)\) where \(s_{i}=\alpha_{i}\left(s_{i-1}\right)\) for \(1<i \leqslant n,\left(\alpha_{1} \ldots \alpha_{n}\right)(s)\) denotes \(s_{n}\), the state that is reached from \(s\) by performing \(\alpha_{1} \ldots \alpha_{n}\).

In Figure 8.1, e.g., \(\operatorname{Act}(s)=\{\alpha, \beta\}, t=\beta(s)\) and \(u=\alpha(s)\). Moreover, \(v=(\beta \alpha)(s)\) and \(v=(\alpha \beta)(s)\).

The notion of independence of actions plays a central role in partial order reduction. As we will describe later, the transition system \(T S\) is reduced by omitting the redundancies in \(T S\) that are caused by the different orderings of independent actions. Intuitively, the pair of actions \(\alpha\) and \(\beta\) with \(\alpha \neq \beta\) is independent when these actions access disjoint variables. In this case, the effect of executing \(\alpha \beta\) or \(\beta \alpha\) is the same. Actions \(\alpha\) and \(\beta\) are also independent when they represent actions of different processes such that one of these actions only operates on local variables. Actions within a process may also be independent, viz. when their order of execution is irrelevant.

The characteristic feature of the independence of actions \(\alpha\) and \(\beta\) is their commutativity which asserts that the effect of the execution orders \(\alpha \beta\) and \(\beta \alpha\) is the same. This means that if \(\alpha\) and \(\beta\) are enabled in state \(s\),
- the execution of action \(\alpha\) cannot disable \(\beta\), and vice versa, and
- the action sequences \(\alpha \beta\) and \(\beta \alpha\) executed in \(s\) yield the same state.

These properties are characteristic for the interleaving diamond (see Figure 8.1 on page 595).

\section*{Definition 8.3. Independence of Actions}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be an action-deterministic transition system with \(\alpha, \beta \in\) Act, \(\alpha \neq \beta\).
1. \(\alpha\) and \(\beta\) are independent (in \(T S\) ) if for any \(s \in S\) with \(\alpha, \beta \in \operatorname{Act}(s)\) :
\[
\beta \in \operatorname{Act}(\alpha(s)) \quad \text { and } \quad \alpha \in \operatorname{Act}(\beta(s)) \quad \text { and } \quad \alpha(\beta(s))=\beta(\alpha(s)) .
\]
2. \(\alpha\) and \(\beta\) are dependent (in \(T S\) ) if \(\alpha\) and \(\beta\) are not independent in TS.

As in most cases the transition system \(T S\) is fixed, the part (in \(T S\) ) is often omitted. The notions of dependence and independence can be lifted to relations between actions and sets of actions as follows. For \(A \subseteq\) Act and \(\beta \in A c t \backslash A, \beta\) is independent of \(A\) (in TS) if for any \(\alpha \in A, \beta\) is independent of \(\alpha\) (in TS). \(\beta\) is dependent on \(A\) in TS if \(\beta \in\) Act \(\backslash A\) and \(\beta\) and \(\alpha\) are dependent in \(T S\) for some \(\alpha \in A\).

\section*{Example 8.4. Independence of Actions (Parallel Operator \(\|_{H}\) )}

Let \(T S_{1}, T S_{2}\) be two action-deterministic transition systems with action sets \(A c t_{1}\) and \(A c t_{2}\), respectively, and \(H=A c t_{1} \cap A c t_{2}\). The actions \(\alpha \in A c t_{1} \backslash H\) and \(\beta \in A c t_{2} \backslash H\) are independent in \(T S_{1} \|_{H} T S_{2}\). (The handshaking operator \(\|_{H}\) has been defined on page 48.) When \(H=\varnothing\), all actions of \(T S_{1}\) are independent of the actions of \(T S_{2}\).

\section*{Example 8.5. Independence of Actions (Program Graphs)}

Consider the program graphs \(P G_{1}\) and \(P G_{2}\) and assume they do not communicate over channels. The actions \(\alpha\) and \(\beta\), by means of which at least one of the two processes accesses local variables only, are independent in \(T S\left(P G_{1} \| \mid P G_{2}\right)\), the transition system underlying the parallel composition of \(P G_{1}\) and \(P G_{2}\). More precisely, we impose the following condition. Let \(\alpha\) be an action only appearing in \(P G_{1}\) such that for variable valuation \(\eta\) :
- \(\operatorname{Effect}(\alpha, \eta)(x)=\eta(x)\) for all variables \(x\) accessed by \(P G_{2}\), and
- for any edge \(\ell \stackrel{g: \alpha}{\hookrightarrow} \ell^{\prime}\) in \(P G_{1}\), guard \(g\) does not refer to the variables that appear in \(P G_{2}\).

Under these conditions, \(\alpha\) is independent of every action \(\beta\) in \(P G_{2}\).


Figure 8.3: The semaphore-based mutual exclusion algorithm.

Let us illustrate this by means of the semaphore-based mutual exclusion program, see the transition system in Figure 8.3. The actions \(\alpha=\) request \(_{1}\) and \(\beta=\) request \(_{2}\) only appear in the following edges of the program graph:
\[
\text { noncrit }_{i} \stackrel{\text { request }_{i}}{ } \text { wait }_{i} \text {. }
\]

In fact, \(\alpha\) and \(\beta\) satisfy the above-mentioned conditions, and therefore are independent. This reflects the fact that the request operations-as well as all activities in the noncritical section - can be concurrently executed. On the other hand, the actions enter \({ }_{i}\) that are executed on entering the critical section,
\[
\text { wait }_{i} \xrightarrow{y>0: \text { enter }_{i}} \text { crit }_{i},
\]
access the shared variable \(y\) (the semaphore). Note that

Therefore, the actions enter \({ }_{1}\) and enter 2 are dependent, since e.g., for state \(s=\left\langle w_{1}, w_{2}, y=\right.\) 1), we have
\[
\text { enter }_{1}, \text { enter }_{2} \in \operatorname{Act}(s) \quad \text { and } \quad \text { enter }_{2} \notin \operatorname{Act}\left(\operatorname{enter}_{1}(s)\right)=\operatorname{Act}\left(\left\langle c_{1}, w_{2}, y=0\right\rangle\right) .
\]

The pairs of actions (request \({ }_{1}\), enter \({ }_{2}\) ), (enter \({ }_{1}\), request \({ }_{2}\) ), (rel, request \({ }_{1}\) ) and (rel, request \({ }_{2}\) ) are independent. For instance, the only state in which both actions rel and request \({ }_{1}\) are enabled is \(s^{\prime}=\left\langle n_{1}, c_{2}, y=0\right\rangle\). We have
\[
\text { request }_{1} \in \operatorname{Act}\left(\operatorname{rel}\left(s^{\prime}\right)\right) \quad \text { rel } \in \operatorname{Act}\left(\text { request }_{1}\left(s^{\prime}\right)\right)
\]
and
\[
\operatorname{rel}\left(\text { request }_{1}\left(s^{\prime}\right)\right)=\operatorname{request}_{1}\left(\operatorname{rel}\left(s^{\prime}\right)\right)=\left\langle w_{1}, n_{2}, y=1\right\rangle .
\]

A similar reasoning applies to the other pairs of independent actions.

The following lemma is a central result for the partial order reduction approach. It relies on the successive exchange of independent actions \(\beta_{i}\) and \(\alpha\) in the action sequence \(\beta_{1} \ldots \beta_{n} \alpha\).

\section*{Lemma 8.6. Permuting Independent Actions}

Let TS be an action-deterministic transition system, s a state in TS, and let
\[
s=s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} s_{2} \xrightarrow{\beta_{3}} \ldots \xrightarrow{\beta_{n}} s_{n}
\]
be an execution fragment from \(s\) with the action sequence \(\beta_{1} \ldots \beta_{n}\). Then, for any \(\alpha \in\) \(\operatorname{Act}(s)\) which is independent of \(\left\{\beta_{1}, \ldots, \beta_{n}\right\}\) we have \(\alpha \in \operatorname{Act}\left(s_{i}\right)\), and
\[
s=s_{0} \xrightarrow{\alpha} t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n-1}} t_{n-1} \xrightarrow{\beta_{n}} t_{n}
\]
is an execution fragment in TS with the action sequence \(\alpha \beta_{1} \ldots \beta_{n}\) such that \(t_{i}=\alpha\left(s_{i}\right)\) for \(0 \leqslant i \leqslant n\).

Proof: Let \(T S\) be action-deterministic transition system. By induction on \(i \geqslant 1\) it is shown that
- \(\alpha\) and \(\beta_{i+1}\) are enabled in state \(s_{i}=\left(\beta_{1} \ldots \beta_{i}\right)(s)\),
- \(\beta_{i}\) is enabled in state \(t_{i-1}=\left(\alpha \beta_{1} \ldots \beta_{i-1}\right)(s)\), and
- \(\alpha\left(s_{i}\right)=\beta_{i}\left(t_{i-1}\right)\).

Base: \((i=1)\). Let \(\alpha, \beta_{1} \in \operatorname{Act}(s)\) and \(s_{1}=\beta_{1}(s)\) and \(t_{0}=\alpha(s)\). Since \(\alpha\) and \(\beta_{1}\) are independent, \(\alpha \in \operatorname{Act}\left(s_{1}\right)\) and \(\beta_{1} \in \operatorname{Act}\left(t_{0}\right)\), and \(\alpha\left(\beta_{1}(s)\right)=\beta_{1}(\alpha(s))\), i.e., \(\alpha\left(s_{1}\right)=\beta_{1}\left(t_{0}\right)\).

Induction step: \((i-1 \Longrightarrow i\) for \(1<i \leqslant n)\). Assume \(\alpha\) and \(\beta_{i}\) are enabled in state \(s_{i-1}=\) \(\left(\beta_{1} \ldots \beta_{i-1}\right)(s), \beta_{i-1}\) is enabled in state \(t_{i-2}=\left(\alpha \beta_{1} \ldots \beta_{i-2}\right)(s)\), and the \(\alpha\)-successor of \(s_{i-1}\) and the \(\beta_{i-1}\)-successor of \(t_{i-2}\) agree, i.e.:
\[
t_{i-1}=\alpha\left(s_{i-1}\right)=\beta_{i-1}\left(t_{i-2}\right)
\]


Figure 8.4: Permuting \(\alpha\) with the independent actions \(\beta_{1}\) through \(\beta_{n}\).

The independence of \(\alpha\) and \(\beta_{i}\) yields that \(\alpha\) is enabled in state \(\beta_{i}\left(s_{i-1}\right)=s_{i}, \beta_{i}\) is enabled in state \(\alpha\left(s_{i-1}\right)=\beta_{i-1}\left(t_{i-2}\right)=t_{i-1}\), and \(\alpha\left(s_{i}\right)=\beta_{i}\left(t_{i-1}\right)=t_{i}\).

Lemma 8.6 is illustrated in Figure 8.4.
Consider the infinite execution fragment \(\rho\) starting in \(s\) with the action sequence \(\beta_{1} \beta_{2} \beta_{3} \ldots\) and an action \(\alpha \in \operatorname{Act}(s)\) that is independent on all \(\beta_{j}\) 's (in particular, \(\alpha \neq \beta_{j}\) for all \(j>0)\). Then Lemma 8.6 applied to the finite prefixes of \(\rho\) yields the existence of finite execution fragments from \(s\) with the action sequences \(\alpha \beta_{1} \beta_{2} \ldots \beta_{n}\) for all \(n>0\). Since \(T S\) is action-deterministic, we obtain an infinite execution fragment which first executes action \(\alpha\) and then the infinite action sequence \(\beta_{1} \beta_{2} \beta_{3} \ldots\). More precisely, we have:

\section*{Lemma 8.7. Adding an Independent Action}

Let TS be an action-deterministic transition system, s a state in TS, and let
\[
s=s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} s_{2} \xrightarrow{\beta_{3}} \ldots
\]
be an infinite execution fragment from \(s\) with the action sequence \(\beta_{1} \beta_{2} \ldots\). Then, for \(\alpha \in \operatorname{Act}(s)\) which is independent of \(\left\{\beta_{1}, \beta_{2}, \ldots\right\}\) we have \(\alpha \in \operatorname{Act}\left(s_{i}\right)\) for all \(i \geqslant 0\) and
\[
s=s_{0} \xrightarrow{\alpha} t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} t_{2} \xrightarrow{\beta_{3}} \ldots
\]
is an infinite execution fragment in TS with the action sequence \(\alpha \beta_{1} \beta_{2} \ldots\) and \(t_{i}=\alpha\left(s_{i}\right)\) for \(i \geqslant 0\).

If no further assumptions are made, the traces induced by the execution fragments
\[
\begin{array}{llllllll}
s_{0} \xrightarrow{\beta_{1}} & s_{1} & \xrightarrow{\beta_{2}} & \ldots & \xrightarrow{\beta_{n}} & s_{n} & \xrightarrow{\alpha} & t, \\
\text { and } \\
s_{0} \xrightarrow{\alpha} & t_{0} & \xrightarrow{\beta_{1}} & \ldots & \xrightarrow{\beta_{n-1}} & t_{n-1} & \xrightarrow{\beta_{n}} & t
\end{array}
\]
will be distinct and not related via any form of (stutter) trace equivalence or (stutter) trace inclusion. However, if the action \(\alpha\) which is moved from the "right" to the "left" (by successive swapping the order \(\beta_{i} \alpha\) into \(\alpha \beta_{i}\) ) does not affect the state labeling, then the execution fragments are stutter-equivalent. Recall that execution fragments \(\varrho, \varrho^{\prime}\) are stutter-equivalent, denoted \(\varrho \triangleq \varrho^{\prime}\), if their traces only differ in the number of repetitions of state labels; see Section 7.7.1. These actions are called stutter actions or invisible actions.

\section*{Definition 8.8. Stutter Action}

The action \(\alpha \in\) Act is a stutter action if for each transition \(s \xrightarrow{\alpha} s^{\prime}\) in transition system \(T S\) we have \(L(s)=L\left(s^{\prime}\right)\).

Since action-deterministic transition systems are considered, \(\alpha\) is a stutter action in TS if and only if \(L(s)=L(\alpha(s))\) for all states \(s\) in \(T S\) with \(\alpha \in \operatorname{Act}(s)\). For example, the actions \(\beta\) and \(\gamma\) in the transition system in Figure 8.5 are stutter actions, while \(\alpha\) is not a stutter action.


Figure 8.5: Actions \(\beta\) and \(\gamma\) are stutter actions.

Remark 8.9. Stutter Steps vs. Stutter Actions
Let us briefly explain the difference between stutter steps and stutter actions. Recall that
a stutter step is a transition \(s \rightarrow t\) such that \(L(s)=L(t)\). The fact whether or not an \(\alpha\)-labeled transition is a stutter step depends on the state in which \(\alpha\) is executed. For example, let \(\alpha\) be the action corresponding to the assignment \(x:=2 * x\) for some integer variable \(x\). In all states in which \(x\) evaluates to \(0, \alpha\) does not affect the value of \(x\). If the atomic propositions only consider the values of \(x\) (and other program variables), but not the program locations, then the transitions
\[
\langle\ldots, x=0\rangle \xrightarrow{x:=2 * x}\langle\ldots, x=0\rangle
\]
are stutter steps, whereas the transitions
\[
\langle\ldots, x=v\rangle \xrightarrow{x:=2 * x}\langle\ldots, x=2 v\rangle
\]
for \(v \neq 0\) are not. Action \(\alpha\) is a stutter action whenever all transitions \(s \xrightarrow{\alpha} s^{\prime}\) are stutter steps.

\section*{Lemma 8.10. Permuting Independent Stutter Actions}

Let TS be an action-deterministic transition system, s a state in TS, and \(\varrho\) and \(\varrho^{\prime}\) be finite execution fragments starting in \(s\) with action sequences \(\beta_{1} \ldots \beta_{n} \alpha\) and \(\alpha \beta_{1} \ldots \beta_{n}\), respectively, such that \(\alpha\) is a stutter action which is independent of \(\left\{\beta_{1}, \ldots, \beta_{n}\right\}\). Then \(\varrho \triangleq \varrho^{\prime}\).

Proof: Let
\[
\begin{array}{lllllllllll}
\varrho=s_{0} & \xrightarrow{\beta_{1}} & s_{1} & \xrightarrow{\beta_{2}} & s_{2} & \xrightarrow{\beta_{3}} & \ldots & \xrightarrow{\beta_{n}} & s_{n} & \xrightarrow{\alpha} & v \\
\varrho^{\prime}=s_{0} & \xrightarrow{\alpha} & t_{0} & \xrightarrow{\beta_{1}} & t_{1} & \xrightarrow{\beta_{2}} & \ldots & \xrightarrow{\beta_{n-1}} & t_{n-1} & \xrightarrow{\beta_{n}} & t_{n},
\end{array}
\]
where \(s=s_{0}\). By Lemma 8.6, \(\varrho\) and \(\varrho^{\prime}\) end in the same state, i.e., \(t_{n}=v\), and \(\alpha\left(s_{i}\right)=t_{i}\) for \(0 \leqslant i \leqslant n\). Since \(\alpha\) is a stutter action:
\[
L\left(s_{i}\right)=L\left(\alpha\left(s_{i}\right)\right)=L\left(t_{i}\right) \quad \text { for } \quad 0 \leqslant i \leqslant n
\]

Let \(A_{i}=L\left(s_{i}\right), 0 \leqslant i \leqslant n\). Then:
\[
\begin{aligned}
& \operatorname{trace}(\varrho)=L\left(s_{0}\right) L\left(s_{1}\right) \ldots L\left(s_{n}\right) L\left(t_{n}\right)=A_{0} A_{1} \ldots A_{n} A_{n} \text { and } \\
& \operatorname{trace}\left(\varrho^{\prime}\right)=L\left(s_{0}\right) L\left(t_{0}\right) L\left(t_{1}\right) \ldots L\left(t_{n}\right)=A_{0} A_{0} A_{1} \ldots A_{n} \text {. }
\end{aligned}
\]

Thus, both traces have the form \(A_{0}^{+} A_{1} \ldots A_{n-1} A_{n}^{+}\). Hence, \(\varrho \triangleq \varrho^{\prime}\).

The following lemma describes a transformation of an infinite execution fragment with the action sequence \(\beta_{1} \beta_{2} \beta_{3} \ldots\) into a stutter-equivalent execution fragment with the action sequence \(\alpha \beta_{1} \beta_{2} \beta_{3} \ldots\) where \(\alpha\) is a stutter action that is independent of any \(\beta_{i}\).

\section*{Lemma 8.11. Adding an Independent Stutter Action}

Let TS be an action-deterministic transition system, \(s\) a state in TS, and \(\rho\) and \(\rho^{\prime}\) be infinite execution fragments starting in \(s\) with the action sequences \(\beta_{1} \beta_{2} \beta_{3} \ldots\) and \(\alpha \beta_{1} \beta_{2} \beta_{3} \ldots\). respectively, such that \(\alpha\) is a stutter action which is independent of \(\left\{\beta_{1}, \beta_{2}, \beta_{3}, \ldots\right\}\). Then \(\rho \triangleq \rho^{\prime}\).

Proof: Let \(\rho=s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} s_{2} \xrightarrow{\beta_{3}} \ldots\) and \(\rho^{\prime}=s_{0} \xrightarrow{\alpha} t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} t_{2} \xrightarrow{\beta_{3}} \ldots\) where \(s=s_{0}\). Then, \(s_{i}=\alpha\left(t_{i}\right)\) for all \(i \geqslant 0\). Since \(\alpha\) is a stutter action we have \(L\left(s_{i}\right)=L\left(t_{i}\right)\) for all \(i \geqslant 0\). With \(A_{i}=L\left(s_{i}\right)\) we get
\[
\begin{aligned}
\operatorname{trace}(\rho) & =L\left(s_{0}\right) L\left(s_{1}\right) L\left(s_{2}\right) \ldots \\
\operatorname{trace}\left(\rho^{\prime}\right) & =L\left(s_{0}\right) L\left(t_{0}\right) L\left(t_{1}\right) L\left(t_{2}\right) \ldots
\end{aligned}=A_{0} A_{1} A_{2} \ldots .
\]

Thus, both traces have the form \(A_{0}^{+} A_{1} A_{2} \ldots\) which yields \(\rho \triangleq \rho^{\prime}\).

Lemmas 8.10 and 8.11 yield the basis of the partial order reduction approach. During partial order reduction, any stutter equivalence class of executions in the full system TS is represented by at least one execution in the reduced system \(\hat{T S}\). (One might say that partial order reduction amounts to model checking using representative executions.) The representatives in \(\hat{T S}\) of \(T S\) 's stutter equivalence classes arise by permuting independent actions and adding independent stutter actions.

\subsection*{8.2 The Linear-Time Ample Set Approach}

We consider partial order reduction for LTL using so-called ample sets. The basic idea is the following. Consider a high-level specification of an asynchronous system. Using traditional state space generation, for each encountered states all direct successors are explored. That is, for each action \(\alpha \in \operatorname{Act}(s)\), the successor state \(\alpha(s)\) is determined, and when encountered for the first time, generated. With partial order reduction using ample sets, the set ample \((s) \subseteq \operatorname{Act}(s)\) will be explored instead of the entire set \(\operatorname{Act}(s)\). That is, all direct successors in \(\operatorname{Act}(s) \backslash \operatorname{ample}(s)\) are not explored, and possibly not generated at all. By choosing appropriate action sets ample \((\cdot)\), this approach yields a-hopefully small-fragment of the full transition system \(T S=(S, A c t, \rightarrow, I, A P, L)\). As \(T S\) will never be generated, the peak memory requirements are determined by the size of the fragment \(\hat{T S}\) rather than by TS. The reduced transition system \(\hat{T S}\) results from the transition relation \(\Longrightarrow\) which is defined by
\[
\frac{s \xrightarrow{\alpha} s^{\prime} \wedge \alpha \in \operatorname{ample}(s)}{s \underset{\alpha}{\Longrightarrow} s^{\prime}}
\]

More precisely, \(\hat{T} S=(\hat{S}, A c t, \Longrightarrow, I, A P, \widehat{L})\) where the state space \(\hat{S}\) consists of those states that are reachable (under \(\Longrightarrow\) ) from some initial state \(s_{0} \in I\) and \(\widehat{L}(s)=L(s)\) for any \(s \in \hat{S}\). Thus, \(\hat{S}\) might be a proper subset of the original state space \(S\). The following correctness criterion needs to be established:
(1) \(\hat{T S}\) and \(T S\) are equivalent with respect to the formulae to be checked.

It turns out that stutter trace equivalence is an appropriate notion of equivalence for partial order reduction when checking LT properties; for branching-time properties, divergencesensitive stutter bisimulation is convenient. Besides this formal soundness criterion, the following more informal requirements are of importance:
(2) \(\hat{T S}\) should be considerably smaller (and therefore more efficient to analyze) than \(T S\).
(3) The effort to generate \(\hat{T S}\) should be relatively low compared to the verification of TS.

To fulfill the last constraint, we aim at an algorithm to generate \(\hat{T} S\) with a time complexity that is linear in the size of \(\hat{T S}\). Typically, \(\hat{T S}\) is obtained by a static analysis of a highlevel description of \(T S\), e.g., as a channel system or sequential program graph. If \(\hat{T S}\) is significantly smaller than \(T S\), then such an approach can be viewed as efficient.

The following subsections treat the ample-set method for verifying LT properties specified as LTL formulae. It is assumed that the original transition system \(T S=(S, A c t, \rightarrow\) , \(I, A P, L)\) is finite, action-deterministic, and does not have terminal states. The aim is to replace the verification whether \(T S \vDash \varphi\), for \(\varphi\) an LTL formula over AP, by checking whether \(\hat{T S} \models \varphi\). \(\hat{T S}\) is obtained from \(T S\) by choosing appropriate ample sets. We first deal with LT properties; in Section 8.3, this approach will be extended to the verification of branching-time properties stated in CTL*.

\subsection*{8.2.1 Ample Set Constraints}

To ensure that \(T S\) and \(\hat{T S}\) are equivalent, the ample sets have to fulfill a number of conditions. As we will see, these constraints ensure that \(T S \triangleq \hat{T S}\), i.e., \(T S\) and \(\hat{T S}\) are stutter trace equivalent (see Section 7.7.1, see page 530 and further). Since stutter trace equivalence preserves all stutter-insensitive LT properties, the reduction of \(T S\) to \(\hat{T S}\) is sound for such LT properties. The fragment of LTL that does not contain the next step
operator, \(\mathrm{LTL}_{\backslash \bigcirc}\), is an appropriate logical formalism to specify stutter-insensitive LT properties, and thus
\[
\hat{T S} \models \varphi \quad \text { if and only if } \quad T S \models \varphi
\]
for any \(\operatorname{LTL}_{\backslash \bigcirc}\)-formulae \(\varphi\), see Corollary 7.93 (page 535). (As the next step operator is not very useful for asynchronous systems, the kind of systems that partial order reduction is aimed at, the absence of the \(\bigcirc\) operator is not a severe restriction.)

Given a technique to generate the ample sets, such that \(T S \triangleq \hat{T S}\), standard LTL modelchecking techniques may be applied to check \(\hat{T S} \vDash \varphi\) (and thus \(T S \models \varphi\) ). There are two main approaches to the computation of the ample sets, i.e., for determining \(\hat{T S}\) : static vs. dynamic partial order reduction. In dynamic (or on-the-fly) partial order reduction, the reduced transition system \(\hat{T S}\) is generated during the LTL model checking of \(\hat{T S}\). The main advantage of this approach is that \(\hat{T S} \not \vDash \varphi\) might be established without the need for generating the entire transition system \(\hat{T S}\) - only the relevant part of the product \(\hat{T S} \otimes \mathcal{A}_{\neg \varphi}\), where \(\mathcal{A}_{\neg \varphi}\) is the Büchi automaton for \(\neg \varphi\), is needed to show the refutation of \(\varphi\). This approach is treated in Section 8.2.3. In the static approach, a symbolic representation (e.g., in terms of program graphs) of \(\hat{T S}\) is generated prior to the verification. The reduction of the transition system may be viewed as a preprocessing phase of model checking. The static approach is treated in Section 8.2.4.

In the remainder of this section, conditions for the ample sets are established that ensure \(T S \triangleq \hat{T S}\). Subsequently, techniques will be discussed to algorithmically determine appropriate ample sets.

Since every execution in \(\hat{T} S\) is an execution in TS, we establish sufficient conditions to assign to each execution \(\rho_{0}\) in TS a stutter-equivalent execution \(\hat{\rho}\) in \(\hat{T S}\). In order to avoid any confusion, we stress that this assignment of executions in TS to executions in \(\hat{T S}\) is not part of the algorithm for generating \(\hat{T S}\), but is only needed to prove the stutter trace equivalence of \(T S\) and \(\hat{T S}\). The basic idea for the transformation
\[
\text { "execution } \rho_{0} \text { in } T S \mapsto \hat{\rho}_{0} \text { in } \hat{T} S \text { with } \rho_{0} \triangleq \hat{\rho}_{0} "
\]
is the following. Let \(\rho_{0}\) be an infinite execution in \(T S\) that is not an execution in \(\hat{T S}\). A stutter-equivalent execution \(\hat{\rho}_{0}\) in \(\hat{T S}\) will be obtained from \(\rho_{0}\) by successively permuting the order of independent actions and possibly adding stutter steps, according to the transformations described in cases 1 and 2 (see below). The correctness of these transformations is ensured by Lemmas 8.10 and 8.11. These transformations allow the replacement of \(\rho_{0}\) through a possibly infinite series of transformations:
\[
\rho_{0} \mapsto \rho_{1} \mapsto \rho_{2} \mapsto \ldots \mapsto \hat{\rho}_{0} \quad \text { where } \rho_{i} \triangleq \rho_{0} \text { for all } i \geqslant 0
\]
such that at least the first \(i\) steps in \(\rho_{i}\) are transitions (i.e., according to \(\Rightarrow\) ) in \(\hat{T} S\) and
agree with the first \(i\) transitions of \(\rho_{j}\) for all \(j>i\). That is, \(\rho_{i}[. . i]=\rho_{j}[. . i]\) for all \(j>i\). In this way, an execution \(\hat{\rho}_{0}\) in \(\hat{T S}\) is obtained which, for any \(i \geqslant 0\), agrees with \(\rho_{i}\) for the first \(i\) transitions. It can be regarded as the "limit" of the sequence \(\rho_{0}, \rho_{1}, \rho_{2}\), and so on.

Let us consider the transformation of \(\rho_{i}\) to \(\rho_{i+1}\) in more detail. For simplicity consider the case \(i=0\). Let \(m\) be the minimal index in \(\rho_{0}\) such that \(s=\rho_{0}[m] \xrightarrow{\alpha} \rho_{0}[m+1]\) for \(\alpha \notin \operatorname{ample}(s)\). Execution \(\rho_{0}\) thus consists of a finite prefix \(\varrho_{0}\) containing transitions in \(\hat{T S}\) that ends in state \(s\), and the infinite execution fragment \(\rho\) starting in \(s\) with action sequence \(\beta_{1} \beta_{2} \ldots\), say, such that \(\beta_{1} \notin \operatorname{ample}(s)\) :
\[
\rho_{0}=\underbrace{\stackrel{\gamma_{1}}{\Longrightarrow} \ldots \stackrel{\gamma_{m}}{\Longrightarrow} s}_{\text {prefix } \varrho_{0}} \underbrace{s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} s_{2} \ldots}_{\text {suffix } \rho \text { with } \beta_{1} \notin \operatorname{ample}(s)} \text { for } m \geqslant 0 \text {. }
\]

Execution \(\rho_{1}\) then starts with the prefix \(\varrho_{0}\) and continues with the infinite execution fragment \(\rho^{\prime}\) that is obtained from \(\rho\) according to one of the transformations described below.

Case 1: There exists \(n>0\) such that \(\alpha=\beta_{n+1} \in \operatorname{ample}(s)\) and \(\beta_{1}, \ldots, \beta_{n} \notin \operatorname{ample}(s)\), i.e., some action of the suffix \(\rho=s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots\) of \(\rho_{0}\) belongs to ample(s). The conditions imposed on the ample sets will ensure that \(\alpha\) is a stutter action that is independent of \(\left\{\beta_{1}, \ldots, \beta_{n}\right\}\). The execution fragment \(\rho^{\prime}\) (i.e., the suffix of \(\rho_{1}\) ) results from \(\rho\) by replacing the action sequence \(\beta_{1} \ldots \beta_{n} \alpha\) with \(\alpha \beta_{1} \ldots \beta_{n}\). That is, the action \(\alpha\) is shifted to occur prior to any \(\beta_{i}(0<i \leqslant n)\). This transformation is described by Lemma 8.6 (page 601). Pictorially this amounts to
\[
\begin{aligned}
& \rho_{0}=u \xrightarrow{\gamma_{1}} \ldots \xlongequal{\gamma_{m}} s \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\alpha} t \xrightarrow{\beta_{n+2}} s_{n+2} \xrightarrow{\beta_{n+3}} \ldots \\
& \rho_{1}=u \xrightarrow{\gamma_{1}} \ldots \xrightarrow{\gamma_{m}} s \xrightarrow{\alpha} t_{0} \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{n}} t \xrightarrow{\beta_{n+2}} s_{n+2} \xrightarrow{\beta_{n+3}} \ldots \\
& \underbrace{+} \underbrace{\underbrace{\underbrace{2}}} \\
& \text { common prefix } \varrho_{0} \\
& \text { stutter-equivalent } \\
& \text { common suffix } \\
& \text { execution fragments }
\end{aligned}
\]

Since \(\alpha\) is a stutter action, \(\rho \triangleq \rho^{\prime}\) (see Lemma 8.10 on page 604), and thus \(\rho_{0} \triangleq \rho_{1}\).
Case 2: For all \(i>0, \beta_{i} \notin \operatorname{ample}(s)\), i.e., none of the actions occurring in the suffix \(\rho\) belongs to ample(s). The conditions on the ample sets will ensure the independence of \(\beta_{i}\) and \(\operatorname{ample}(s)\) for any \(i\), and that any \(\alpha \in \operatorname{ample}(s)\) is a stutter action. The execution fragment \(\rho^{\prime}\) then results by replacing \(\rho\) by the execution fragment that starts in \(s\) and successively executes the actions \(\alpha \beta_{1} \beta_{2} \beta_{3} \ldots\), for some \(\alpha \in \operatorname{ample}(s)\).

Schematically:
\[
\begin{aligned}
\rho_{0}= & u \stackrel{\gamma_{1}}{\Longrightarrow} \ldots \stackrel{\gamma_{m}}{\Longrightarrow} s \stackrel{\beta_{1}}{\Longrightarrow} s_{1} \xrightarrow{\beta_{2}} s_{2} \xrightarrow{\beta_{3}} \ldots \\
\rho_{1}= & \underbrace{u \stackrel{\gamma_{1}}{\Longrightarrow} \ldots \stackrel{\gamma_{m}}{\Longrightarrow}}_{\text {common prefix } \varrho_{0}} s \stackrel{\alpha}{\Longrightarrow} t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} t_{2} \xrightarrow{\beta_{3}} \ldots
\end{aligned}
\]

Since \(\alpha\) is a stutter action, Lemma 8.11 yields \(\rho \triangleq \rho^{\prime}\) and thus \(\rho_{0} \triangleq \rho_{1}\).

In both cases, an execution fragment \(\rho_{1}\) is obtained that starts in the same state as \(\rho_{0}\) such that the first transition that is not a transition in \(\hat{T S}\) occurs at some position \(\geqslant 2\). This recipe is applied iteratively to \(\rho_{j} \mapsto \rho_{j+1}\) for \(j \geqslant 1\). This yields a stutter-equivalent execution fragment \(\rho_{j+1}\) such that the first transition that is not a transition in \(\hat{T S}\) (i.e., not an ample transition) occurs at some position \(>j\) and \(\rho_{j}[. . j+1]=\rho_{j+1}[. . j+1]\). Continuing in this way, we finally obtain an execution fragment in \(\hat{T S}\).

To ensure that the above-described transformations (i.e., cases 1 and 2 ) are applicable and yield an execution \(\hat{\rho}_{0}\) in \(\hat{T S}\) such that \(\rho_{0} \triangleq \hat{\rho}_{0}\) with \(\rho_{0}\) in TS, four conditions are imposed on ample sets referred to as (A1) through (A4). Conditions (A1) through (A3) are imposed on each state \(s\) in \(\hat{S}\); no restrictions are imposed on states that are unreachable via \(\Longrightarrow\). Condition (A4) is imposed on all cycles in \(\hat{T} S\).

Let us consider the constraints in more detail.

> (A1) Nonemptiness condition
> \(\varnothing \neq \operatorname{ample}(s) \subseteq A c t(s)\)

The first condition asserts that if a state has at least one direct successor in \(T S\), then it has least one direct successor in \(\hat{T S}\). As TS does not have terminal states, condition (A1) ensures that \(\hat{T S}\) does not have any terminal states.

\section*{(A2) Dependency condition}

Let \(s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\alpha} t\) be a finite execution fragment in TS. If \(\alpha\) depends on \(\operatorname{ample}(s)\), then \(\beta_{i} \in \operatorname{ample}(s)\) for some \(0<i \leqslant n\).

The key condition for the correctness is the dependency condition (A2). It asserts that in every (!) finite execution fragment of \(T S\), an action depending on ample(s) cannot occur before some action from \(\operatorname{ample}(s)\) is occurring first. Note that this condition is imposed
on every (finite) execution of the original transition system TS. We will see later that this condition ensures that for any state \(s\) which is not fully expanded (i.e., ample \((s)\) is a proper subset of \(\operatorname{Act}(s))\), all ample actions \(\alpha \in \operatorname{ample}(s)\) are independent of \(\operatorname{Act}(s) \backslash \operatorname{ample}(s)\). Note that for \(n=0\), condition (A2) is false, as the existential quantification (over \(i\) ) ranges over an empty domain.

Condition (A2) guarantees that any finite execution in \(T S\) is of the form
\[
\varrho=s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\alpha} t \quad \text { with } \quad \alpha \in \operatorname{ample}(s)
\]
and \(\beta_{i}\) independent of \(\operatorname{ample}(s)\) for \(0<i \leqslant n\). If \(\alpha\) is a stutter action-which will be guaranteed by constraint (A3) - the execution obtained by shifting \(\alpha\) to the beginning is also an execution of \(T S\); see the transformation as described under case 1 above. That is to say, if \(\varrho\) is pruned in \(T S\), i.e., does not occur in \(\hat{T S}\) as \(\beta_{1} \notin \operatorname{ample}(s)\), then a stutterequivalent execution can be constructed by performing \(\alpha\) in \(s\). Infinite executions in \(T S\) are of the form
\[
s_{1} \xrightarrow{\beta_{1}} s_{2} \xrightarrow{\beta_{2}} \ldots \quad \text { with } \beta_{i} \text { independent of } \operatorname{ample}(s) \text { for } 0<i \leqslant n \text {. }
\]

For stutter action \(\alpha \in \operatorname{ample}(s)\), inserting the action \(\alpha\) at the beginning of this execution yields another execution of TS; see the transformation as described under case 2 above.

\section*{(A3) Stutter condition}

If \(\operatorname{ample}(s) \neq \operatorname{Act}(s)\) then any \(\alpha \in \operatorname{ample}(s)\) is a stutter action.

Stutter condition (A3) ensures that the transformations (cases 1 and 2) generate stutterequivalent executions; see Lemma 8.10 (page 604) and Lemma 8.11 (page 604). To be more precise, condition (A3) ensures that the switch from action sequence \(\beta_{1} \ldots \beta_{n} \alpha\) to \(\alpha \beta_{1} \ldots \beta_{n}\), and from \(\beta_{1} \beta_{2} \beta_{3} \ldots\) to \(\alpha \beta_{1} \beta_{2} \beta_{3} \ldots\) in state \(s\) with \(\alpha \in \operatorname{ample}(s)\) yields stutter-equivalent executions. Ample actions may thus be executed first.

\section*{(A4) Cycle condition}

For any cycle \(s_{0} s_{1} \ldots s_{n}\) in \(\hat{T S}\) and \(\alpha \in \operatorname{Act}\left(s_{i}\right)\), for some \(0<i \leqslant n\), there exists \(j \in\{1, \ldots, n\}\) such that \(\alpha \in \operatorname{ample}\left(s_{j}\right)\).

The cycle condition (A4) is the final condition that is needed to ensure that \(T S\) and \(\hat{T S}\) are stutter-equivalent. The justification of this condition is provided later.

The conditions imposed on ample sets are summarized in Figure 8.6.

\section*{(A1) Nonemptiness condition}
\[
\varnothing \neq \operatorname{ample}(s) \subseteq \operatorname{Act}(s)
\]

\section*{(A2) Dependency condition}

Let \(s \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\alpha} t\) be a finite execution fragment in TS. If \(\alpha\) depends on \(\operatorname{ample}(s)\), then \(\beta_{i} \in \operatorname{ample}(s)\) for some \(0<i \leqslant n\).

\section*{(A3) Stutter condition}

If \(\operatorname{ample}(s) \neq \operatorname{Act}(s)\) then any \(\alpha \in \operatorname{ample}(s)\) is a stutter action.

\section*{(A4) Cycle condition}

For any cycle \(s_{0} s_{1} \ldots s_{n}\) in \(\hat{T S}\) and \(\alpha \in \operatorname{Act}\left(s_{i}\right)\), for some \(0<i \leqslant n\), there exists \(j \in\{1, \ldots, n\}\) such that \(\alpha \in \operatorname{ample}\left(s_{j}\right)\).

Figure 8.6: Requirements on the ample set of state \(s\).

\section*{Example 8.12. Ample Set Conditions}

Consider the transition system TS in Figure 8.7 (left part) over \(A P=\{a\}\). Action \(\beta\) is a stutter action, and is independent of \(\{\alpha, \gamma, \delta\}\). Let \(\operatorname{ample}\left(s_{0}\right)=\{\beta\}\). This choice satisfies constraints (A1) through (A3). Consider now state \(s_{2}\). The choice \(\operatorname{ample}\left(s_{2}\right)=\{\alpha\}\) violates (A3), as \(\alpha\) is not a stutter action. ample \(\left(s_{2}\right)=\{\delta\}\) violates the cycle condition (A4): the reduced transition system \(\hat{T} S\) would contain the cycle \(s_{0} s_{2} s_{2}\) with \(\alpha \in \operatorname{Act}\left(s_{2}\right)\), but \(\alpha \notin \operatorname{ample}\left(s_{2}\right)\). Thus, we select ample \(\left(s_{2}\right)=\{\alpha, \delta\}\). The nonemptiness condition (A1) then leaves no freedom for \(s_{3}\) : ample \(\left(s_{3}\right)=\{\gamma\}\). The resulting reduced transition system \(\hat{T S}\) is depicted in Figure 8.7 (right part).

The traces of \(T S\) and \(\hat{T S}\) are either of the form \(\left(\varnothing^{+}\{a\}^{+}\right)^{\omega}\) or \(\left(\varnothing^{+}\{a\}^{+}\right)^{*} \varnothing^{\omega}\). Hence, \(T S \triangleq \hat{T} S\).

We now state the main result of this section. The proof of this result is provided by a series of lemmas, presented in the remainder of this section.

\section*{Theorem 8.13. Correctness of the Ample Set Approach}

Let TS be an action-deterministic, finite transition system without terminal states. Then if conditions (A1) through (A4) are satisfied, then \(\hat{T S} \triangleq T S\).

This theorem asserts that whenever \(\hat{T S}\) is constructed from \(T S\) using ample sets that


Figure 8.7: Transition sytem TS (left) and \(\hat{T S}\) (right)
satisfy all constraints (A1) through (A4), then \(\hat{T S}\) and \(T S\) are stutter trace equivalent. Since stutter trace equivalence is finer than \(\mathrm{LTL}_{\backslash \bigcirc}\) equivalence, conditions (A1) through (A4) guarantee that \(\hat{T S}\) and \(T S\) satisfy the same \(\operatorname{LTL}_{\backslash \bigcirc}\) formulae. The remainder of this subsection is devoted to the proof of Theorem 8.13. Since \(\hat{T S}\) is a fragment of \(T S\), every execution of \(\hat{T S}\) is an execution of \(T S\). It remains to show that, conversely, for every execution in TS there exists a stutter-equivalent execution in \(\hat{T S}\). The proof for this statement proceeds in a number of steps, and results from Lemmas 8.14 through 8.21.

The first two lemmas follow from simple observations that are based on the dependency condition (A2).

\section*{Lemma 8.14. Independence of Ample and Other Enabled Actions}

Let \(s \in \operatorname{Reach}(T S)\) and \(\alpha \in \operatorname{ample}(s)\). Then
(A2) implies that \(\alpha\) is independent of \(\operatorname{Act}(s) \backslash \operatorname{ample}(s)\).

Proof: Let \(s \in \operatorname{Reach}(T S), \alpha \in \operatorname{ample}(s)\), and \(\beta \in \operatorname{Act}(s) \backslash \operatorname{ample}(s)\). Then there exists an execution fragment starting in \(s\) which begins with action \(\beta\) :
\[
\rho=s \xrightarrow{\beta} s_{1} \xrightarrow{\gamma_{1}} s_{2} \xrightarrow{\gamma_{2}} s_{3} \xrightarrow{\gamma_{3}} \ldots
\]

Assume ample(s) satisfies (A2) and assume \(\alpha\) and \(\beta\) are dependent. The existence of the execution \(\rho\), however, violates condition (A2) since \(\beta\) is not preceded by an action in ample(s). Contradiction.
(In Remark 8.19 (page 614) it is shown that just imposing that any \(\alpha \in \operatorname{ample}(s)\) is independent of \(\operatorname{Act}(s) \backslash \operatorname{ample}(s)\) is too weak to establish that \(T S \triangleq \hat{T S}\).)

The following lemma asserts that for any execution fragment that starts in \(s\) it holds that any action in ample \((s)\) remains enabled as long as no action in ample( \(s\) ) has been executed.

\section*{Lemma 8.15. Enabledness of Ample Actions}

Let \(s \in \operatorname{Reach}(T S)\) and \(s=s_{0} \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{n}} s_{n}\) be a finite execution fragment in TS. If \(\operatorname{ample}(s)\) satisfies (A2) and \(\left\{\beta_{1}, \ldots, \beta_{n}\right\} \cap \operatorname{ample}(s)=\varnothing\), then for all actions \(\alpha \in \operatorname{ample}(s)\) are independent of \(\left\{\beta_{1}, \ldots, \beta_{n}\right\}\). In adddition, we have \(\alpha \in \operatorname{Act}\left(s_{i}\right)\) for \(0<i \leqslant n\).

Proof: Let \(s \in \operatorname{Reach}(T S)\) and \(s=s_{0} \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{n}} s_{n}\) be a finite execution fragment, \(\rho\) say, in TS. Consider \(\alpha \in \operatorname{ample}(s)\) and assume ample(s) satisfies constraints (A1) and (A2). The proof is by induction on \(i\).

Basis: \((i=1)\). By Lemma 8.14, \(\alpha\) and \(\beta_{1}\) are independent. But then also, \(\alpha \in \operatorname{Act}\left(\beta\left(s_{0}\right)\right)=\) Act \(\left(s_{1}\right)\).

Induction step: Suppose the claim holds for \(0<i<n\). Consider the execution fragment \(\rho_{i}=s_{0} \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{i}} s_{i}\). As \(\rho\) is an execution fragment in \(T S, \beta_{i+1} \in \operatorname{Act}\left(s_{i}\right)\). So, \(\rho_{i+1}=s_{0} \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{i}} s_{i} \xrightarrow{\beta_{i+1}} s_{i+1}\) is an execution fragment in TS. Then \(\alpha\) and \(\beta_{i+1}\) are independent, as otherwise \(\rho_{i+1}\) would violate (A2). The induction hypothesis yields \(\alpha \in \operatorname{Act}\left(s_{i}\right)\). As \(\alpha\) and \(\beta_{i+1}\) are independent, we get \(\alpha \in \operatorname{Act}\left(\beta_{i+1}\left(s_{i}\right)\right)=\operatorname{Act}\left(s_{i+1}\right)\).

\section*{Notation 8.16. Fully Expanded State}

A state \(s\) of \(T S\) is fully expanded if ample \((s)=\operatorname{Act}(s)\).

The following two lemmas show that the conditions (A1) through (A3) are necessary conditions to guarantee that for each execution in \(T S\) a stutter-equivalent execution in \(\hat{T S}\) can be determined.

\section*{Lemma 8.17. Constructing Stutter-Equivalent Executions (Case 1)}

Let \(\varrho\) be a finite execution fragment in Reach(TS) of the form
\[
s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\alpha} t
\]
where \(\beta_{i} \notin \operatorname{ample}(s)\), for \(0<i \leqslant n\), and \(\alpha \in \operatorname{ample(s).~If~ample(s)~satisfies~(A1)~through~}\) (A3), then there exists an execution fragment \(\varrho^{\prime}\) of the form
\[
s \xrightarrow{\alpha} t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n-1}} t_{n-1} \xrightarrow{\beta_{n}} t
\]
and \(\varrho \triangleq \varrho^{\prime}\).

Proof: Let \(\varrho=s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\alpha} t\) be a finite execution fragment in \(\operatorname{Reach}(T S)\) where \(\beta_{i} \notin \operatorname{ample}(s)\), for \(0<i \leqslant n\), and \(\alpha \in \operatorname{ample}(s)\). By Lemma 8.15, \(\alpha\) is independent of \(\left\{\beta_{1}, \ldots, \beta_{n}\right\}\). State \(s\) is not fully expanded, since \(\beta_{1} \in \operatorname{Act}(s) \backslash \operatorname{ample}(s)\). By condition (A3), \(\alpha\) is a stutter action. The claim now follows from Lemma 8.10 (page 604).
\(\operatorname{Provided} \beta_{1}, \ldots, \beta_{n} \notin \operatorname{ample}(s)\) and \(\alpha \in \operatorname{ample}(s)\), replacing the action sequence \(\beta_{1} \ldots \beta_{n} \alpha\) in \(T S\) by \(\alpha \beta_{1} \ldots \beta_{n}\) (as described in case 1) results in a stutter-equivalent execution fragment. The following lemma is similar in flavor but deals with infinitely many actions that are independent of \(\alpha \in \operatorname{ample}(s)\).

\section*{Lemma 8.18. Constructing Stutter-Equivalent Executions (Case 2)}

Let \(\rho=s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} s_{2} \xrightarrow{\beta_{3}} \ldots\) be an infinite execution fragment in Reach(TS) where \(\beta_{i} \notin \operatorname{ample}(s)\), for \(i>0\). If ample(s) satisfies (A1) through (A3), then there exists an execution fragment \(\rho^{\prime}\) of the form
\[
s \stackrel{\alpha}{\Longrightarrow} t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} t_{2} \xrightarrow{\beta_{3}} \ldots
\]
where \(\alpha \in \operatorname{ample}(s)\) and \(\rho \triangleq \rho^{\prime}\).

Proof: Let \(\rho=s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} s_{2} \xrightarrow{\beta_{3}} \ldots\) be an infinite execution fragment in \(\operatorname{Reach}(T S)\) where \(\beta_{i} \notin \operatorname{ample}(s)\), for \(i>0\). As \(\beta_{1} \in \operatorname{Act}(s) \backslash \operatorname{ample}(s)\), state \(s\) is not fully expanded. By condition (A3), all actions \(\alpha \in \operatorname{ample}(s)\) are stutter actions. Condition (A1) yields the existence of an action \(\alpha \in \operatorname{ample}(s)\). The claim now follows directly from Lemma 8.7 (page 602) and Lemma 8.11 (page 605).

Remark 8.19. An Alternative Dependency Condition (A2')
Consider the following variant of (A2): for any \(s \in \hat{S}\) with \(\operatorname{ample}(s) \neq \operatorname{Act}(s)\), any \(\alpha \in \operatorname{ample}(s)\) is independent of \(\operatorname{Act}(s) \backslash \operatorname{ample}(s)\). This variant, referred to as (A2 \(\left.2^{\prime}\right)\) in the sequel, seems-at first sight-a reasonable variant of (A2). It simply requires that any ample action of a nonfully expanded state \(s\) is independent of all enabled actions in \(s\) that are not in its ample set. The following example shows, however, that ( \(\mathrm{A}^{\prime}\) ) does not guarantee that \(T S \triangleq \hat{T} S\).

Consider the transition system depicted in Figure 8.8 (left part). Actions \(\alpha\) and \(\beta\) are independent and \(\alpha\) and \(\delta\) are stutter actions. The following ample sets satisfy the conditions (A1), (A2'), (A3) and (A4):
\[
\operatorname{ample}\left(s_{0}\right)=\{\alpha\} \quad \text { and } \quad \operatorname{ample}\left(s_{2}\right)=\{\beta\} \quad \text { and } \quad \operatorname{ample}\left(s_{3}\right)=\{\delta\} .
\]

The resulting reduced transition system \(\hat{T} S\) is depicted in Figure 8.8 (right part). Conditions (A1) and (A3) are obvious. Note that \(s_{1} \notin \hat{S}\), and thus no requirements are imposed
on \(\operatorname{ample}\left(s_{1}\right)\). Cycle condition (A4) is fulfilled, as there is only one cycle (the self-loop at \(s_{3}\) ), and \(\operatorname{ample}\left(s_{3}\right)=\operatorname{Act}\left(s_{3}\right)\). As state \(s_{0}\) is the only nonfully expanded state in \(\hat{T S}\), condition ( \(\mathrm{A} 2^{\prime}\) ) has only to be checked for \(s_{0}\). It holds as \(\alpha\) and \(\beta\) are independent.


Figure 8.8: A possible, though unsound reduction that satisfies (A2') but not (A2).

However, \(T S \neq \hat{T S}\) since, e.g., \(\hat{T S} \vDash \square \neg a\) (as it does not contain any \(a\)-state), while \(T S \not \vDash \square \neg a\). Thus, condition (A2') does not guarantee \(T S \triangleq \hat{T S}\). The ample sets defined above violate condition (A2). This can be seen as follows. Action \(\gamma\) depends on \(\alpha\), since both actions are enabled in state \(s_{1}\), but \(\alpha\) is not enabled in \(\gamma\left(s_{1}\right)=s_{4}\). Thus, the execution fragment \(s_{0} \xrightarrow{\beta} s_{1} \xrightarrow{\gamma} s_{4}\) violates (A2) as \(\alpha\) should occur before \(\gamma\). So, (A2) does not allow the reduction from \(T S\) to \(\hat{T S}\) as depicted in Figure 8.8.

Now consider the cycle condition (A4). The results we established so far ensure that the series of transformations \(\rho_{0} \mapsto \rho_{1} \mapsto \rho_{2} \mapsto \ldots\) (according to case 1 and 2) generate executions \(\rho_{i}\) that are stutter-equivalent to the execution \(\rho_{0}\) in TS. Moreover, at least the first \(i\) transitions in \(\rho_{i}\) are transitions in \(\hat{T S}\) and agree with the first \(i\) transitions in \(\rho_{j}\) for all \(j \geqslant i\). The idea is now to consider the "limit" \(\hat{\rho}_{0}\) of the executions \(\left(\rho_{i}\right)_{i \geqslant 0}\). That is, for all \(i>0\), the \(i\) th transition in \(\hat{\rho}_{0}\) is the \(i\) th transition in \(\rho_{i}\) (and \(\rho_{j}\) for all \(j \geqslant i\) ). Thus, \(\hat{\rho}_{0}\) is an execution in \(\hat{T} S\). Unfortunately, without the cycle condition (A4) it cannot be guaranteed that \(\rho_{i} \triangleq \hat{\rho}_{0}\). The intuitive reason is that some (nonstutter) action \(\beta\) might be delayed ad infinitum in the successive construction of \(\hat{\rho}_{0}\).

Let us consider this effect more in detail. Assume \(\alpha\) and \(\beta\) are independent, enabled in all states, and \(\alpha\) is a stutter action while \(\beta\) is not. The ample sets ample \((s)=\{\alpha\}\) for any state \(s\) fulfill conditions (A1) through (A3). Let \(\rho_{0}\) be an execution with the action sequence \(\beta \alpha^{\omega}\). Then the above transformation \(\rho_{0} \mapsto \rho_{1} \mapsto \rho_{2} \mapsto \ldots\) yields the executions \(\rho_{i}\) with the action sequence \(\alpha^{i} \beta \alpha^{\omega}\) (see case 1). Their limit is the execution \(\hat{\rho}_{0}\) with the action sequence \(\alpha^{\omega} \beta \equiv \alpha^{\omega}\). However, it is not guaranteed that \(\hat{\rho}_{0} \triangleq \rho_{i}\) as action \(\beta\) is
ignored forever in \(\hat{\rho}_{0}\).
In a similar way, in absence of condition (A4), the transformation according to case 2 may generate an execution in \(\hat{T S}\) that is not stutter-equivalent to the original execution \(\rho_{0}\). Let, e.g., \(\rho_{0}\) have the action sequence \(\beta_{1} \beta_{2} \ldots\). Then case 2 might generate a series of executions \(\rho_{1} \mapsto \rho_{2} \mapsto \ldots\) with action sequences
\[
\begin{array}{ccccccc}
\alpha & \beta_{1} & \beta_{2} & \beta_{3} & \beta_{4} & \beta_{5} & \ldots \\
\alpha & \alpha & \beta_{1} & \beta_{2} & \beta_{3} & \beta_{4} & \ldots \\
\alpha & \alpha & \alpha & \beta_{1} & \beta_{2} & \beta_{3} & \ldots \\
& & \vdots & & & &
\end{array}
\]
such that in the limit, action \(\beta_{1}\) is never performed.

The following example illustrates the necessity of cycle condition (A4).


Figure 8.9: Transition systems \(T S_{1}\) (left) and \(T S_{2}\) (right).
Example 8.20. Necessity of Cycle Condition (A4)
Consider the transition systems \(T S_{1}\) and \(T S_{2}\) depicted in Figure 8.9. The transition system \(T S=T S_{1} \| \mid T S_{2}\) is depicted in the left part of Figure 8.10. The reduced transition system \(\hat{T S}\), depicted on the right of Figure 8.10, results from choosing ample \(\left(\left\langle s_{0}, t_{i}\right\rangle\right)=\left\{\alpha_{i+1}\right\}\), for \(i=1,2,3\). Conditions (A1), (A2), and (A3) are fulfilled since \(\beta\) is independent of \(\left\{\alpha_{1}, \alpha_{2}, \alpha_{3}\right\}\). Consider the action sequence \(\beta\left(\alpha_{1} \alpha_{2} \alpha_{3}\right)^{\omega}\) and the associated execution \(\rho\) in \(T S\). The associated trace is
\[
\operatorname{trace}(\rho)=\varnothing\{a\}\{a\}\{a\} \ldots=\varnothing\{a\}^{\omega} \in \operatorname{Traces}(T S)
\]

In \(\hat{T S}\), however, there does not exist an execution that is stutter-equivalent to \(\rho\), since \(\hat{T S}\) has no states labeled with \(a\). In fact, \(\operatorname{Traces}(\hat{T S})=\left\{\varnothing^{\omega}\right\}\). Hence, \(T S \neq \hat{T S}\). (This can also be seen by considering the \(\mathrm{LTL}_{\backslash \bigcirc}\) formula \(\square \neg a\). Hence \(T S \not \vDash \square \neg a\) and \(\hat{T S} \vDash \square \neg a\).)

Let us explain why the above-mentioned replacement process, which should transform the executions of \(T S\) into a stutter-equivalent execution of \(\hat{T} S\), fails. In this example, case 1


Figure 8.10: Transition system \(T S\) (left) and unsound reduced transition system \(\hat{T S}\) (right).
continuously applies, and \(\rho_{0}\) is successively replaced with executions that are associated to the action sequences:
\[
\begin{aligned}
& \alpha_{1} \beta \alpha_{2} \alpha_{3} \alpha_{1} \alpha_{2} \alpha_{3} \ldots \\
& \alpha_{1} \alpha_{2} \beta \alpha_{3} \alpha_{1} \alpha_{2} \alpha_{3} \ldots \\
& \alpha_{1} \alpha_{2} \alpha_{3} \beta \alpha_{1} \alpha_{2} \alpha_{3} \ldots \\
& \alpha_{1} \alpha_{2} \alpha_{3} \alpha_{1} \beta \alpha_{2} \alpha_{3} \ldots \\
& \quad \vdots
\end{aligned}
\]

But the action \(\beta\) is never performed, since \(\beta\) does not occur in \(\hat{T S}\). Actually, the cycle con-


Figure 8.11: Cycle condition (A4) is violated.
dition (A4) is violated, since \(\beta\) is continuously enabled on the cycle \(\left\langle s_{0}, t_{0}\right\rangle\left\langle s_{0}, t_{1}\right\rangle\left\langle s_{0}, t_{2}\right\rangle\) \(\left\langle s_{0}, t_{0}\right\rangle\), see Figure 8.11, but not in any ample sets of these states.

The above considerations demonstrate that conditions (A1) through (A3) cannot guarantee \(T S \triangleq \hat{T S}\). The goal is now to show that (A1), (A2), and (A3) together with the cycle
condition (A4) suffice. In fact, with the cycle condition (A4) a situation described above is impossible.

\section*{Lemma 8.21. Stutter-Trace Inclusion of TS in \(\hat{T S}\)}

If (A1) through (A4) are satisfied, then \(T S \unlhd \hat{T S}\).

Proof: Let \(\rho_{0}\) be an execution in \(T S\) which starts in state \(s\) and is induced by the action sequence \(\beta_{1} \beta_{2} \beta_{3} \ldots\) where \(\beta_{1} \notin \operatorname{ample}(s)\). The execution \(\rho_{0}\) is successively replaced with stutter-equivalent executions \(\rho_{m}, m=1,2,3, \ldots\), by means of the transformations indicated in Lemmas 8.17 and 8.18. Each of these executions \(\rho_{m}\) starts in state \(s\) and is based on an action sequence of the form
\[
\alpha_{1} \ldots \alpha_{m} \beta_{1} \gamma_{1} \gamma_{2} \gamma_{3} \ldots
\]

The action sequence \(\alpha_{1} \alpha_{2} \ldots \alpha_{m}\) contains the actions of the ample sets, which are newly inserted according to Lemma 8.18, and all actions \(\beta_{n}\), which were shifted forward according to Lemma 8.17. \(\gamma_{1} \gamma_{2} \gamma_{3} \ldots\) denotes the remaining subsequence of \(\beta_{1}, \beta_{2}, \beta_{3} \ldots\). Thus, \(\rho_{m}\) is of the form
\[
s \stackrel{\alpha_{1}}{\Longrightarrow} t_{1} \xrightarrow{\alpha_{2}} \ldots \xrightarrow{\alpha_{m}} t_{m} \xrightarrow{\beta_{1}} t_{0}^{m} \xrightarrow{\gamma_{1}} t_{1}^{m} \xrightarrow{\gamma_{2}} t_{2}^{m} \xrightarrow{\gamma_{3}} \ldots
\]
where \(\alpha_{1}, \ldots, \alpha_{m}\) are stutter actions.
Actually, the case \(\beta_{1} \notin \operatorname{ample}\left(t_{m}\right)\) for all \(m \in \mathbb{N}\) is impossible, since - due to the finiteness of \(T S\) - the path fragment \(s t_{1} t_{2} \ldots t_{m}\) contains a cycle for sufficiently large \(m\). Moreover, \(\beta_{1} \in \operatorname{Act}\left(t_{m}\right)\) for all \(m\) by Lemma 8.15 on page 613. This contradicts condition (A4). Hence, \(\beta_{1} \in \operatorname{ample}\left(t_{m}\right)\) for some \(m \geqslant 1\). But then
\[
s \stackrel{\alpha_{1}}{\Longrightarrow} t_{1} \stackrel{\alpha_{2}}{\Longrightarrow} \ldots \stackrel{\alpha_{m}}{\Longrightarrow} t_{m} \stackrel{\beta_{1}}{\Longrightarrow} t_{0}^{m}
\]
is an execution fragment in \(\hat{T} S\) which is a prefix of \(\rho_{m+1}\) and all executions \(\rho_{j}\) for \(j>m\).
According to the outlined recipe, we obtain an execution \(\hat{\rho}_{0}\) in \(\hat{T S}\) (as the "limit" of \(\rho_{m}\), \(\rho_{m+1} \ldots\) ), where the induced action sequence contains all actions that occur in \(\rho_{0}\) (in TS). Let us assume that \(\rho_{0}\) has the form \(s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots\) and let \(0=k_{0}<k_{1}<k_{2}<\ldots\) such that \(\beta_{k_{1}} \beta_{k_{2}} \ldots\) results from \(\beta_{1} \beta_{2} \ldots\) by omitting all stutter actions in \(\beta_{1} \beta_{2}, \ldots\). (Note that the subsequence \(\beta_{k_{1}} \beta_{k_{2}} \ldots\) may be finite.) Then, \(\operatorname{trace}\left(\rho_{0}\right)\) has the form \(A_{0}^{+} A_{1}^{+} A_{2}^{+} \ldots\), where \(A_{i}\) is the label \(L\left(s_{k}\right)\) of all states \(s_{k}\) with \(k_{i} \leqslant k<k_{i+1}\). Since each of the nonstutter actions \(\beta_{k_{i}}\) is eventually "processed" when generating the executions \(\rho_{1}, \rho_{2}, \rho_{3}, \ldots\), for each index \(k_{i}\) there is some finite word \(w_{i}\) of the form \(A_{0}^{+} A_{1}^{+} \ldots A_{i}^{+}\)and some index \(\ell_{i}\) such that the traces of the executions \(\rho_{j}\) for all \(j \geqslant \ell_{i}\) start with \(w_{i}\). In particular, \(w_{i}\) is a proper prefix of \(w_{i+1}\) and the words \(w_{i}\) are prefixes of the trace associated with the "limit" execution \(\hat{\rho}_{0}\). Hence, \(\operatorname{trace}\left(\hat{\rho}_{0}\right)\) has the form \(A_{0}^{+} A_{1}^{+} A_{2}^{+} \ldots\), and \(\rho_{0} \triangleq \hat{\rho}_{0}\).

As \(\operatorname{Traces}(\hat{T S}) \subseteq \operatorname{Traces}(T S)\), the above lemma yields that \(T S \triangleq \hat{T S}\). This completes the proof of Theorem 8.13 (page 611).

\section*{Remark 8.22. Nonemptiness Condition (A1) and Terminal States}

Throughout this chapter, we assume a transition system without terminal states. This assumption has been made to be consistent with Chapter 5 where LTL formulae are interpreted as languages over infinite words. However, all concepts presented here (as well as those in Chapter 5) can also be applied to transition systems that have terminal states. The only difference from the approach presented here is that the nonemptiness condition has to be replaced with the following requirements (A1.1) and (A1.2):
\((\mathrm{A} 1.1) \mathrm{ample}(s) \subseteq \operatorname{Act}(s)\).
(A1.2) \(\operatorname{ample}(s)=\varnothing\) if and only if \(\operatorname{Act}(s)=\varnothing\).

Condition (A1.1) ensures that \(\hat{T S}\) is a subtransition system of \(T S\), while (A2.1) guarantees that any terminal state in \(\hat{T S}\) is a terminal state in TS. Theorem 8.13 holds for transition systems with terminal states, i.e., conditions (A1.1), (A1.2), and (A2) through (A4) yield \(T S \triangleq \hat{T} S\)

\subsection*{8.2.2 Dynamic Partial Order Reduction}

We now consider the integration of partial order reduction as part of LTL model checking. The basic strategy is to generate \(\hat{T S}\) during model checking. This is in contrast to static partial order reduction in which \(\hat{T S}\) is constructed prior to the verification. To simplify matters, we first treat on-the-fly, or dynamic, partial order reduction during invariant checking. As invariants can be checked by a depth-first search (DFS) technique, this boils down to integrating the ample set technique in a depth-first search. Subsequently, we show how nested depth-first search - the algorithm to find cycles containing an accept state in \(\hat{T S} \otimes \mathcal{A}_{\neg \varphi}\) - can be adapted such that partial order reduction is employed.

Partial Order Reduction in Depth-First Search Consider the invariant \(\square \Phi\) where \(\Phi\) is a propositional formula (i.e., not an LTL formula). In order to check whether \(T S \models\) \(\square \Phi\), a depth-first search algorithm can be applied, as explained in Chapter 3, that checks in every state of \(T S\) whether \(\Phi\) holds. This is done during the state-space generation. On exploring a state \(s\), all outgoing transitions of \(s\) are considered, i.e., all actions in \(\operatorname{Act}(s)\) are considered. When integrating partial order reduction into this procedure, only
the \(\alpha\)-successor states of \(s\) are considered for which \(\alpha \in \operatorname{ample}(s)\). The resulting depthfirst search algorithm is provided in Algorithm 38 on page 622. It generates the ample sets for any fresh encountered state of \(\hat{T S}\) and explores the successor states under \(\Longrightarrow\) according to a depth-first search strategy. The set mark \((s)\) keeps track of the actions in ample( \(s\) ) that have been considered during the search. As soon as this set equals ample( \(s\) ), all ample successors of \(s\) are considered, state \(s\) is popped from the stack, and \(\Phi\) is checked. In case an action \(\alpha\) is considered whose successor has not been encountered before, the state is marked as being reachable (under \(\Longrightarrow\) ) and pushed on the stack. For any freshly encountered state, an approximation of its ample set is generated that satisfies (A1) through (A3), but possibly not (A4). Later on, we will see that establishing these constraints can be done by means of local criteria. Roughly speaking, it is first attempted to determine ample \((s)\) as the set of actions of a single process that are enabled in state \(s\). To ensure the cycle condition (A4), the ample set is enlarged on demand. This is indicated by the statement ample \(\left(s^{\prime}\right):=\operatorname{Act}\left(s^{\prime}\right)\). This technique relies on replacing (A4) by the stronger condition:

\section*{(A4') Strong cycle condition}

Any cycle in \(\hat{T} S\) contains at least one state \(s\) with ample \((s)=\operatorname{Act}(s)\).

It is not difficult to assess that ( \(\mathrm{A} 4^{\prime}\) ) is a sufficient criterion for (A4), provided the other ample set conditions (A1) through (A3) hold. This is shown in the following lemma.

\section*{Lemma 8.23. Strong Cycle Condition}

If (A1) through (A3) hold, then condition (A4') implies (A4).

Proof: By contraposition. Assume \(s_{0} \ldots s_{n}(\) with \(n>0)\) is a cycle in \(\hat{T} S\) with \(\operatorname{Act}\left(s_{j}\right)=\) \(\operatorname{ample}\left(s_{j}\right)\) for some \(0<j \leqslant n\), i.e., state \(s_{j}\) is fully expanded. Assume (A4) is violated, i.e., for some \(i \neq j\) it holds that \(\beta \in \operatorname{Act}\left(s_{i}\right)\) and \(\beta \notin \operatorname{ample}\left(s_{k}\right)\) for all \(0<k \leqslant n\). Consider the transition \(s_{i} \xrightarrow{\alpha_{i}} s_{i+1}\) that is part of the cycle \(s_{0} \ldots s_{n}\). (Here, \(i+1\) should be read as \((i+1) \bmod n\).) As \(s_{0} \ldots s_{n}\) is a cycle in \(\hat{T S}\), it follows \(\alpha_{i} \in \operatorname{ample}\left(s_{i}\right)\). Condition (A2) yields that all actions in ample \(\left(s_{i}\right)\) are independent of those in \(\operatorname{Act}\left(s_{i}\right) \backslash \operatorname{ample}\left(s_{i}\right)\). In particular, \(\alpha_{i}\) is independent of \(\beta \notin \operatorname{ample}\left(s_{i}\right)\). But then, \(\beta \in \operatorname{Act}\left(\alpha_{i}\left(s_{i}\right)\right)=\operatorname{Act}\left(s_{i+1}\right)\). As \(\beta \notin \operatorname{ample}\left(s_{i+1}\right)\), it follows by a similar reasoning that \(\beta \in \operatorname{Act}\left(s_{i+2}\right)\). Continuing this reasoning, we obtain \(\beta \in \operatorname{Act}\left(s_{j}\right) \backslash \operatorname{ample}\left(s_{j}\right)\). This, however, contradicts that ample \(\left(s_{j}\right)=\) \(\operatorname{Act}\left(s_{j}\right)\).

The advantage of \(\left(\mathrm{A} 4^{\prime}\right)\) is that it can easily be integrated in the depth-first search algorithm. If the depth-first search finds a backward edge \(s^{\prime} \xrightarrow{\alpha} s^{\prime \prime}\), i.e., \(s^{\prime}\) is the current state and \(s^{\prime \prime}=\alpha\left(s^{\prime}\right)\) for the current action \(\alpha \in \operatorname{ample}\left(s^{\prime}\right)\) and \(s^{\prime \prime}\) is on the stack \(U\), then a cycle
\(s^{\prime \prime} \ldots s^{\prime} \ldots s^{\prime \prime}\) has been found that contains \(s^{\prime}\). Thus, we may simply enlarge ample( \(s^{\prime}\) ) by adding all actions \(\beta \in \operatorname{Act}\left(s^{\prime}\right)\) that are not in the current ample set of \(s\). This turns \(s^{\prime}\) into a fully expanded state and ensures ( \(\mathrm{A} 4^{\prime}\) ). Alternatively, we may fully expand \(\alpha\left(s^{\prime}\right)\), as it suffices to fully expand an arbitrary state on the cycle. (The check whether a state is on the stack \(U\) can be realized in constant time by, e.g., organizing \(R\) by a hash table where each entry is equipped with a bit indicating whether the state is in \(U\) or not.)

\section*{Example 8.24. POR (Reachability Analysis)}

Consider the following concurrent program consisting of two parallel processes:
\begin{tabular}{ll} 
Process \(0:\) \\
& while true \(\{\) \\
\(\ell_{0}:\) & skip; \\
\(m_{0}:\) & \(\quad\) wait until \((\neg b)\{\) \\
\(n_{0}:\) & \(\quad \ldots\) critical section \(\ldots\}\) \\
& \(\quad b:=\) true; \\
& \multirow{2}{c}{}
\end{tabular}

Process 1:
\[
\begin{array}{ll} 
& \text { while true }\{ \\
\ell_{1}: & \text { skip; } \\
m_{1}: & \text { wait until }(b)\{ \\
n_{1}: & \ldots \text { critical section } \ldots\} \\
& b:=\text { false } ; \\
& \}
\end{array}
\]
\[
\ell_{1}: \quad \text { skip; }
\]

The atomic proposition \(a\) holds in the states of \(T S\left(P G_{0} \| \mid P G_{1}\right)\), see Figure 8.12, for which at least one of the processes is in location \(n_{0}\) or \(n_{1}\). Action \(\delta_{i}\) denotes the execution of the skip statement by process \(i, \alpha_{i}\) denotes the action that corresponds to waiting at program location \(m_{i} ; \beta_{i}\) is the action denoting the exit of the busy-wait cycle by process \(i\), and \(\gamma_{i}\) the action to return to the beginning of the loop. The initial value of the shared boolean variable \(b\) is unspecified.

We apply Algorithm 38 with \(\Phi=\) true-just a state-space generation-to \(T S\left(P G_{0} \| P G_{1}\right)\). We start with state \(s_{0}=\left\langle\ell_{0}, \ell_{1}, \neg b\right\rangle\). The actions \(\delta_{0}\) and \(\delta_{1} \in \operatorname{Act}\left(s_{0}\right)\) are stutter actions and independent. The possibilities are
\[
\operatorname{ample}\left(s_{0}\right)=\left\{\delta_{0}\right\}, \operatorname{ample}\left(s_{0}\right)=\left\{\delta_{1}\right\} \text { or ample }\left(s_{0}\right)=\left\{\delta_{0}, \delta_{1}\right\}
\]

The first two alternatives are equivalent and lead to a better reduction than the last alternative. Let
\[
\operatorname{ample}\left(s_{0}\right)=\left\{\delta_{0}\right\}
\]

Intuitively, this corresponds to assigning a higher priority to process zero. The next encountered state is \(\delta_{0}\left(s_{0}\right)=s_{1}=\left\langle m_{0}, \ell_{1}, \neg b\right\rangle\). Actions \(\delta_{1}\) and \(\beta_{0}\) belong to \(\operatorname{Act}\left(s_{1}\right)\) and are independent. \(\beta_{0}\) (and \(\beta_{1}\) ) are not stutter actions. Let
\[
\operatorname{ample}\left(s_{1}\right)=\left\{\delta_{1}\right\}
\]

The next encountered state \(\delta_{1}\left(s_{1}\right)=s_{2}=\left\langle m_{0}, m_{1}, \neg b\right\rangle\). We have \(\operatorname{Act}\left(s_{2}\right)=\left\{\alpha_{1}, \beta_{0}\right\}\). The choice \(\operatorname{ample}\left(s_{2}\right)=\left\{\alpha_{1}\right\}\) violates condition (A4) as it closes the cycle \(s_{1} s_{1}\) on which the
```

Algorithm 38 Invariant checking using partial order reduction
Input: finite transition system $T S$ and propositional formula $\Phi$
Output: "yes" if $T S \models \square \Phi$ ", otherwise "no" plus a counterexample

```

```

procedure visit (state $s$ )
push $(s, U)$;
$R:=R \cup\{s\} ;$
compute ample(s) satisfying (A1)-(A3);
$\operatorname{mark}(s):=\varnothing$;
repeat
$s^{\prime}:=\operatorname{top}(U)$;
if $\operatorname{ample}\left(s^{\prime}\right)=\operatorname{mark}\left(s^{\prime}\right)$ then
$\operatorname{pop}(U) ; \quad$ (* all ample actions have been taken *)
$b:=b \wedge\left(s^{\prime} \models \Phi\right) ; \quad \quad\left(*\right.$ check validity of $\Phi$ in $\left.s^{\prime} *\right)$
else
let $\alpha \in \operatorname{ample}\left(s^{\prime}\right) \backslash \operatorname{mark}\left(s^{\prime}\right)$;
$\operatorname{mark}\left(s^{\prime}\right):=\operatorname{mark}\left(s^{\prime}\right) \cup\{\alpha\} ; \quad$ (* mark $\alpha$ as taken *)
if $\alpha\left(s^{\prime}\right) \notin R$ then
$\operatorname{push}\left(\alpha\left(s^{\prime}\right), U\right)$;
$R:=R \cup\left\{\alpha\left(s^{\prime}\right)\right\} ; \quad\left({ }^{*} \alpha\left(s^{\prime}\right)\right.$ is a new reachable state $\left.{ }^{*}\right)$
compute ample ( $\left.\alpha\left(s^{\prime}\right)\right)$ satisfying (A1)-(A3); (* see Section 8.2.3 *)
$\operatorname{mark}\left(\alpha\left(s^{\prime}\right)\right):=\varnothing ;$
else
if $\alpha\left(s^{\prime}\right) \in U$ then $\operatorname{ample}\left(s^{\prime}\right):=\operatorname{Act}\left(s^{\prime}\right) ;$ fi $\quad\left({ }^{*}\right.$ establish $\left.\left(\mathrm{A} 4^{\prime}\right){ }^{*}\right)$
fi
fi
until $((U=\varepsilon) \vee \neg b)$
endproc

```


Figure 8.12: Transition system for the program \(P G_{1}| | \mid P G_{2}\).
action \(\beta_{0}\) is enabled but is never chosen. As \(\beta_{0}\) is not a stutter action, the only reasonable choice is
\[
\operatorname{ample}\left(s_{2}\right)=\left\{\alpha_{1}, \beta_{0}\right\}
\]

By continuing such arguments the reduced transition system \(\hat{T S}\) depicted in Figure 8.13 is obtained. Thus, eight of the reachable 12 states in TS are obtained by partial order reduction, provided the ample sets are chosen as indicated above.

Partial Order Reduction in Nested Depth-First Search The next issue is to integrate partial order reduction in the \(\mathrm{LTL}_{\backslash \bigcirc}\) model-checking procedure. Rather than checking \(T S \models \varphi\) it is verified whether \(\hat{T S} \models \varphi\). The rough idea for verifying \(\hat{T S} \models \varphi\) is as follows. As usual, a Büchi automaton \(\mathcal{A}_{\neg \varphi}\) is constructed for the \(\operatorname{LTL}_{\backslash \bigcirc}\) formula \(\varphi\). The reachable states of the product transition system \(\hat{T S} \otimes \mathcal{A}_{\neg \varphi}\) are generated and during this state-space generation phase, the persistence property "eventually forever no accept state" is checked by means of the nested depth-first search described in Section 5.2. In the outer depth-first search, a reachable accept state is searched for, whereas in the inner depth-first search, it is checked whether a reachable accept state lies on a cycle. In order to obtain correct results it is evident that in both depth-first searches, the same ample sets need to be used. Otherwise, different transition systems are considered.

As described in Section 5.2, a cycle with an accept state in \(\mathcal{A}_{\neg \varphi}\) is found if the cycle check - the inner DFS which attempts to find a backward edge to an accept state - is started for state \(s\) once the outer DFS for \(s\) is completed, i.e., when all successors of \(s\) have been explored. When adopting the same approach as in Algorithm 38 for the nested


Figure 8.13: The reduced transition system \(\hat{T} S\) for \(P G_{1} \| \mid P G_{2}\).
depth-first search, the inner and outer depth-first search may dynamically change the ample set. This is due to the extension of the ample set to accomplish (A4'). Hence, we have to ensure that the inner DFS is started only if ample(s) cannot change anymore. To that end, a slightly modified version of the nested DFS procedure (indicated at the end of Section 5.2) is exploited. In this variant, the nested DFS is aborted as soon as the inner DFS visits a state \(t\) which is on the stack \(U\) for the outer DFS.

The main steps are outlined in Algorithm 39. The procedure cycle_check_por is a slight variant of the cycle check as given in Algorithm 7 (see page 210). This algorithm aborts as soon as a state \(t\) is visited which is on the stack \(U\) for the outer DFS. That is, this variant seeks an arbitrary backward edge rather than a backward edge from \(s^{\prime}\) to \(s\) (as is done in Algorithm 40 on page 626). The test whether the current state \(t\) of the inner DFS is on the stack \(U\) for the outer DFS can be performed in (expected) constant time if the elements in \(R\) (organized as a hash table) are augmented with a bit that yields the information whether or not a state is in \(U\).

To ensure that the inner DFS uses the same ample sets ample(s) as the outer DFS (and to avoid the recomputation of ample(s) in the inner DFS) one might use one bit for any enabled action \(\alpha\) which indicates whether \(\alpha\) belongs to ample \((s)\). A simple solution is to use hashing for representing \(T\) and \(R\), bits for \(U\) as well as the action bits. Thus, the entries in this hash table are tuples \(\left\langle s, b_{T}, b_{U},\left\langle\alpha_{1}, b_{1}\right\rangle, \ldots,\left\langle\alpha_{n}, b_{n}\right\rangle\right\rangle\) consisting of a state \(s \in R\), bit \(b_{T}\) indicating whether \(s \in T\), bit \(b_{U}\) indicating whether \(s \in U\), and a list \(\left\langle\alpha_{1}, b_{1}\right\rangle, \ldots,\left\langle\alpha_{n}, b_{n}\right\rangle\) for all actions \(\alpha_{i} \in \operatorname{Act}(s)\) where bit \(b_{i}\) equals one iff \(\alpha_{i} \in \operatorname{ample}(s)\).
```

Algorithm 39 Nested depth-first search with partial order reduction
Input: finite transition system $T S$ and propositional formula $\Phi$
Output: "yes" if $T S \models \square \Phi$ ", otherwise "no" plus a counterexample
set of states $R:=\varnothing$;
(* set of visited states in the outer DFS *)
stack of states $U:=\varepsilon$;
(* stack for the outer DFS *)
set of states $T:=\varnothing$;
(* set of visited states in the inner DFS *)
stack of states $V:=\varepsilon$;
(* stack for the inner DFS *)
boolean cycle_found := false;
while ( $I \backslash R \neq \varnothing \wedge \neg$ cycle_found) do
let $s \in I \backslash R$;
reachable_cycle( $s$ );
od
if $\neg$ cycle_found then
return ("yes") (*TS $\models$ "eventually for ever $\Phi$ " *)
else
return ("no", reverse( $V . U$ )) (* stack contents yield a counterexample *)
fi

```
procedure reachable_cycle (state \(s\) )
    \(\operatorname{push}(s, U)\); (* push \(s\) on the stack *)
    \(R:=R \cup\{s\} ; \quad\) (* mark \(s\) as reachable *)
    compute ample(s) satisfying (A1)-(A3); (* see Section 8.2.3 *)
    \(\operatorname{mark}(s):=\varnothing\);
    (* taken actions in \(s^{*}\) )
    repeat
        \(s^{\prime}:=\operatorname{top}(U)\);
        if ample \(\left(s^{\prime}\right)=\operatorname{mark}\left(s^{\prime}\right)\) then
            if \(s \not \models \Phi\) then
                cycle_found \(:=\) cycle_check_por \(\left(s^{\prime}\right)\); (* start the inner DFS in \(\left.s^{\prime *}\right)\)
            fi
            pop \((U)\);
        else
            let \(\alpha \in \operatorname{mark}\left(s^{\prime}\right) \backslash \operatorname{ample}\left(s^{\prime}\right)\);
            \(\operatorname{mark}\left(s^{\prime}\right):=\operatorname{mark}\left(s^{\prime}\right) \cup\{\alpha\} ; \quad\) (* mark \(\alpha\) as taken *)
            if \(\alpha\left(s^{\prime}\right) \notin R\) then
                \(\operatorname{push}\left(\alpha\left(s^{\prime}\right), U\right)\);
                    \(R:=R \cup\left\{\alpha\left(s^{\prime}\right)\right\} ; \quad \quad\left(* \alpha\left(s^{\prime}\right)\right.\) is a new reachable state \(\left.{ }^{*}\right)\)
                compute ample \(\left(\alpha\left(s^{\prime}\right)\right)\) satisfying (A1)-(A3); (* see Section 8.2.3 *)
                \(\operatorname{mark}\left(\alpha\left(s^{\prime}\right)\right):=\varnothing ;\)
            else
                if \(\alpha\left(s^{\prime}\right) \in U\) then \(\operatorname{ample}\left(s^{\prime}\right):=\operatorname{Act}\left(s^{\prime}\right)\); fi \(\quad\left(*\right.\) establish \(\left.\left(\mathrm{A} 4^{\prime}\right)^{*}\right)\)
            fi
        fi
    until \(((U=\varepsilon) \vee \neg b)\)
endproc
```

Algorithm 40 Cycle detection (inner DFS) using ample sets
Input: state $s$ in $\hat{T} S$ with $s \not \vDash \Phi$
Output: true if $s$ belongs to a cycle in $\hat{T S}$; otherwise false
(* $T$ organizes the set of states that have been visited in previous calls of
(* cycle_check_por(•). $V$ serves as DFS stack for cycle_check_por(•),
(* $U$ as DFS stack for the outer DFS (Algorithm 39).

```
procedure boolean cycle_check_por(state \(s\) )
    boolean cycle_found := false;
                                    (* no cycle found yet *)
    push \((s, V)\);
    \(T:=T \cup\{s\} ;\)
    repeat
            \(t:=\operatorname{top}(V) ; \quad \quad(*\) check whether \(t\) is still on the stack of the outer DFS *)
            if \(t \in U\) then
                cycle_found \(:=\) true; \(\quad\left(*\right.\) there is a cycle \(\left.t \ldots s \ldots t^{*}\right)\)
                \(\operatorname{push}(t, V)\);
            else
                if \(\operatorname{ample}(t)=\operatorname{mark}(t)\) then
                    \(\operatorname{pop}(V) ; \quad(*\) all successors of \(t\) in \(\hat{T S}\) have been explored *)
                    else
                        let \(\alpha \in \operatorname{ample}(t) \backslash \operatorname{mark}(t)\);
                            \(\operatorname{mark}(t):=\operatorname{mark}(t) \cup\{\alpha\} ;\)
                    push \((\alpha(t), V)\);
                    \(T:=T \cup\{\alpha(t)\} ;\)
                fi
            fi
        until \(((V=\varepsilon) \vee\) cycle_found \()\)
    return cycle_found
endproc

\subsection*{8.2.3 Computing Ample Sets}

This section is devoted to techniques to determine the ample sets by means of a static analysis of channel systems. The aim is to find criteria for selecting ample sets that can be checked efficiently by a syntactic analysis of the high-level formal description of the system provided as a channel system. As the cycle condition (A4) can be established in the way described before, the focus is on satisfying conditions (A1) through (A3). Recall that a channel system \(C S\) consists of a number of concurrent processes, \(\mathcal{P}_{1}\) through \(\mathcal{P}_{n}\) say, given as program graphs \(P G_{1}\) through \(P G_{n}\) that may have shared variables and may communicate with each other via channels, i.e., first-in-first-out buffers that can store messages. Communication via a channel of capacity zero corresponds to handshaking plus the exchange of some data. The transition system of \(C S\) is denoted as \(T S\), i.e., \(T S=T S(C S)\) where \(C S=\left[P G_{1}|\ldots| P G_{n}\right]\). A detailed introduction and formalization of channel systems is given in Chapter 2. When considering channel \(c\) as buffer, the communication action \(c!v\) puts value \(v\) (at the rear of) the buffer whereas \(c\) ? \(x\) retrieves an element from (the front of) the buffer while assigning it to \(x\).

Let \(A c t_{i}\) and \(L o c_{i}\) denote the action set and set of locations of \(P G_{i}\), respectively. Assume that any action \(\alpha \in A c t_{i}\) occurs in exactly one edge \(\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}\) in program graph \(P G_{i}\) and that the action sets \(A c t_{1}, \ldots, A c t_{n}\) are pairwise disjoint. (This can always be ensured by renaming of actions.) In order to treat all edges in a uniform manner, we assume communication actions to be preceded by guards that equal true, e.g., \(\ell \stackrel{c ? x}{\longrightarrow} \ell^{\prime}\) is written as \(\ell \stackrel{g: c ? x}{\longleftrightarrow} \ell^{\prime}\) where \(g\) equals true.

Let us first introduce some notations. For action \(\alpha\), let \(\operatorname{Var}(\alpha)\) denote the set of variables occurring in \(\alpha\), and \(\operatorname{Modify}(\alpha) \subseteq \operatorname{Var}(\alpha)\) the set of variables that are modified by \(\alpha\). For example:
- \(\operatorname{Var}(x:=x+y)=\{x, y\}\) and \(\operatorname{Modify}(x:=x+y)=\{x\}\),
- \(\operatorname{Var}(c ? x)=\operatorname{Modify}(c ? x)=\{x\}\), and
- \(\operatorname{Var}(c!v)=\operatorname{Modify}(c!v)=\varnothing\) if \(v\) is a value (in the domain of channel \(c\) ).

Variable \(x\) is local to process \(\mathcal{P}_{i}\) if no other process refers to \(x\), i.e., if \(x \notin \operatorname{Var}(\alpha)\) for any \(\alpha \in \bigcup_{\substack{1 \leqslant j \leqslant n \\ j \neq i}} A c t_{j}\).

Let \(\operatorname{Act}_{i}(s)=\operatorname{Act}(s) \cap \operatorname{Act}_{i}\) denote the set of actions of process \(\mathcal{P}_{i}\) which are enabled in (global) state \(s\) in \(T S=T S(C S)\). State \(s\) has the form \(\left\langle\ell_{1}, \ldots, \ell_{n}, \eta, \xi\right\rangle\) where \(\ell_{i}\) denotes
the current location (control point) of \(P G_{i}, \eta\) is the variable evaluation, and \(\xi\) the channel evaluation. As before, \(T S\) is assumed to have no terminal states, i.e., \(\operatorname{Act}_{i}(s) \neq \varnothing\) for some process \(\mathcal{P}_{i}\), for any state \(s .{ }^{1}\)

As a first step, we partition the set of processes \(\mathcal{P}_{1}\) through \(\mathcal{P}_{n}\) into two blocks. The first block contains processes \(\mathcal{P}_{i_{1}}\) through \(\mathcal{P}_{i_{k}}\); the remaining processes constitute the other block. A possible criterion to obtain such partitioning is to exploit the communication pattern of the processes, e.g., no process \(\mathcal{P}_{i_{j}}(0<j \leqslant k)\) is able to communicate to processes outside this set. The intuition is to let \(\operatorname{ample}(s)=\operatorname{Act}_{i_{1}}(s) \cup \ldots \cup \operatorname{Act}_{i_{k}}(s)\), for (global) state \(s\) in \(T S(C S)\). This guarantees that by executing actions not in ample(s), it is impossible that an action \(\beta\) that depends on ample(s) will become enabled in a global state, and therefore would possibly be executed before an action in ample(s)-violating (A2). For simplicity, suppose \(k=1\) and let ample \((s)=\operatorname{Act}_{i}(s)\), for some \(i\). If no other process \(\mathcal{P}_{j}\) with \(j \neq i\) can perform an action in \(s\), then \(\operatorname{Act}_{i}(s)=\operatorname{Act}(s)\), and \(s\) is fully expanded (and evidently satisfying (A1) through (A3)).

Consider that \(\operatorname{Act} t_{i}(s)\) is a proper subset of \(\operatorname{Act}(s)\). Checking the nonemptiness condition (A1) is trivial; it just amounts checking whether process \(\mathcal{P}_{i}\) can perform an action in state \(s\), i.e., \(\operatorname{Act}_{i}(s) \neq \varnothing\). Checking the stutter condition (A3) amounts to check whether all actions in \(A c t_{i}(s)\) are stutter actions. This step can be facilitated by determining the stutter actions by a static analysis - \(\alpha\) is a stutter action if the atomic propositions in \(T S(C S)\) refer neither to a variable in \(\operatorname{Modify}(\alpha)\) nor to the source or target location of edges of the form \(\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}\) nor to the content of channel \(c\) if \(\alpha\) is a receive or send action on c.

Conditions (A1) and (A3) are thus relatively easy to deal with. This does not hold for condition (A2) as this imposes a requirement on any execution in the full transition system \(T S(C S)\). (The cycle condition (A4) is also a global property, but refers to the reduced transition system \(\hat{T S}\) that needs to be analyzed anyway. For this reason the cycle condition is easier to handle.) It turns out that checking (A2) is as hard as checking a reachability property in the full transition system \(T S\).

\section*{Theorem 8.25. Algorithmic Difficulty of Checking (A2)}

The worst-case time complexity of checking (A2) in finite, action-deterministic transition system TS equals that of checking \(T S^{\prime} \models \exists \diamond a\), for some \(a \in A P\), where \(T S^{\prime}\) is a finite, action-deterministic transition system of the same size as TS.

Proof: Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a finite, action-deterministic transition system

\footnotetext{
\({ }^{1}\) In case \(T S=T S(C S)\) has terminal states, condition (A1) has to be replaced by (A1.1) and (A1.2), see Remark 8.22 on page 619.
}
and \(a \in A P\). Without loss of generality, assume \(I=\left\{s_{0}\right\}\). (In case there is more than one initial state, a new state \(s_{0}\) can be introduced such that \(s_{0}\) has no incoming transitions, and \(s_{0} \xrightarrow{\alpha} s^{\prime}\), if \(s \xrightarrow{\alpha} s^{\prime}\) for initial state \(s\).) Assume there is some (possibly unreachable) state \(t \in S\) with \(t \models a\). (If such a state does not occur in \(T S\), then this state \(t\) may just be added.)

The idea of the proof is to define transition system \(T S^{\prime}\) and ample sets for the states in \(T S^{\prime}\) such that \(t \in \operatorname{Reach}(T S)\) if and only if the ample sets defined for \(T S^{\prime}\) violate condition (A2). \(T S^{\prime}\) results from \(T S\) by adding the dependent actions \(\alpha\) and \(\beta\) to the actions of \(T S\), such that \(\beta\) is enabled in any state of \(T S\) and independent of Act while \(\alpha\) is only enabled in the \(a\)-states (and results in the new state trap). Formally:
\[
T S^{\prime}=\left(S \cup\{\operatorname{trap}\}, A c t^{\prime}, \rightarrow^{\prime},\left\{s_{0}\right\}, A P, L^{\prime}\right)
\]
where trap \(\notin S\), Act \(^{\prime}=\operatorname{Act} \cup\{\alpha, \beta, \tau\}\) for \(\alpha, \beta, \tau \notin\) Act. The labeling function \(L^{\prime}\) is irrelevant. The transition relation \(\rightarrow^{\prime}\) is defined by
\[
\frac{s \xrightarrow{\gamma} s^{\prime} \gamma \in A c t}{s \xrightarrow{\gamma} s^{\prime}} \text { and } \frac{s \models a}{s \xrightarrow{\alpha} \text { 'trap }} \quad \text { and } \frac{s \in S}{s \xrightarrow[\longrightarrow]{\beta} s} \text { and } \xrightarrow[\text { trap } \overbrace{}^{\prime}]{ } \text { trap }
\]

Accordingly, the transitions in \(T S\) are extended such that \(\alpha \in \operatorname{Act}(s)\) if \(s=a, \beta \in \operatorname{Act}(s)\) for all \(s \in S\), and trap is equipped with a self-loop labeled with the special action \(\tau\).

Let the ample sets for \(T S^{\prime}\) be defined by
\[
\operatorname{ample}(s)=\operatorname{Act}^{\prime}(s) \text { for all } s \in(S \cup\{\operatorname{trap}\}) \backslash\left\{s_{0}\right\} \text { and } \operatorname{ample}\left(s_{0}\right)=\{\beta\} .
\]

That is, all states are fully expanded except for the initial state \(s_{0}\). Actions \(\alpha\) and \(\beta\) are dependent in \(T S^{\prime}\), since \(\alpha\) and \(\beta\) are both enabled in any \(a\)-state \(s\), but \(\beta\) cannot be executed in state \(\alpha(s)=\) trap. Due to the \(\beta\)-loops inserted at all states \(s \in S, \beta\) is independent of Act.

Claim: \(T S \models \exists \diamond a\) if and only if the ample sets in \(T S^{\prime}\) violate (A2).
1. \((\Rightarrow)\) Let \(t \in \operatorname{Reach}(T S)\) for some \(t \in S\) with \(t \vDash a\). Then there exists an initial execution fragment in \(T S^{\prime}\) of the form
\[
\varrho=\underbrace{s_{0} \longrightarrow \ldots \rightarrow t}_{\substack{\text { initial execution } \\ \text { fragment in } T S}} \xrightarrow{\alpha} \text { trap. }
\]

Since \(\beta \in \operatorname{Act}\left(s_{0}\right)\) and \(\alpha\) and \(\beta\) are dependent in \(T S^{\prime}\), action \(\alpha\) depends on ample \(\left(s_{0}\right)=\) \(\{\beta\}\) in \(T S^{\prime}\). But since \(\alpha\) is only enabled in state \(t\) and not in any state visited in \(\varrho\) prior to \(t, \varrho\) violates (A2). (Note that \(\beta\) cannot occur in the execution fragment \(s_{0} \rightarrow \ldots \rightarrow t\), as \(\beta\) is not an action of \(T S\).)
2. \((\Leftarrow)\) Assume (A2) is violated in \(T S^{\prime}\). Then there exists a state \(v \in \operatorname{Reach}\left(T S^{\prime}\right)\) and an execution fragment
\[
v \xrightarrow{\gamma_{1}} s_{1} \xrightarrow{\gamma_{2}} \ldots \xrightarrow{\gamma_{n}} s_{n} \xrightarrow{\gamma} s^{\prime}
\]
in \(T S^{\prime}\) such that (i) \(\gamma\) depends on \(\operatorname{ample}(v)\) and (ii) \(\gamma_{1}, \ldots, \gamma_{n} \notin \operatorname{ample}(v)\). Since \(s_{0}\) is the only state that is not fully expanded, it follows from (ii) that \(v=s_{0}\). Due to \(\operatorname{ample}\left(s_{0}\right)=\{\beta\}\) and since \(\beta\) is independent of all actions in TS, (i) yields \(\gamma=\alpha\). Since \(\alpha\) is only enabled in the \(a\)-states, it follows \(s_{n} \models a\). Thus, \(T S^{\prime} \vDash \exists \diamond a\). By construction of \(T S^{\prime}\), we get \(v, s_{1}, \ldots, s_{n} \in \operatorname{Reach}(T S)\) - the only state in \(T S^{\prime}\) that is not in \(T S\) is the state trap that has no outgoing transitions. Hence, \(T S \models \exists \diamond a\).

\section*{Remark 8.26. A Technical Remark on the Proof of Theorem 8.25}

The above proof suggests that \(T S^{\prime}\) needs to be explicitly generated. This is, however, not true. Suppose \(T S=T S\left(P G_{1}\| \| \| P G_{n}\right)\) for program graphs \(P G_{1}, \ldots, P G_{n}\) and \(a\) is the atomic proposition stating that variable \(y\) evaluates to 0 . Introduce a new Boolean variable \(x\) with initial value 0 , strengthen all guards in any edge in \(P G_{1}, \ldots, P G_{n}\) with the conjunct \(x=0\), and introduce the program graphs \(P G_{\alpha}, P G_{\beta}\) and \(P G_{\tau}\) that are defined by
- the command "if \(x=0\) then \(\underbrace{\text { skip }}_{\beta} \mathrm{f}\) " in an infinite loop \(\left(P G_{\beta}\right)\),
- the command "if \(\underbrace{y=0}_{a}\) then \(\underbrace{x:=1}_{\alpha}\) f" \(\left(P G_{\alpha}\right)\),
- the skip-command in an infinite loop \(\left(P G_{\tau}\right)\).

Program graphs \(P G_{\beta}\) and \(P G_{\tau}\) consist of a single location with a self-loop (labeled with \(\beta\) and \(\tau\), respectively), while \(P G_{\alpha}\) consists of two locations connected by an edge with the guard \(y=0\) and the action \(\alpha\) for modeling the assignment \(x:=1\). We then consider
\[
T S^{\prime}=T S(\underbrace{P G_{1}\left|\|\ldots \mid\| P G_{n}\right.}_{\text {original program }} \| \underbrace{P G_{\alpha}\| \| P G_{\beta} \| \mid P G_{\tau}}_{\text {extension }})
\]

In this way, a representation of \(T S^{\prime}\) is obtained from the representation of \(T S\) in constant time.

As a result of Theorem 8.25, (A2) is not checked for an arbitrary set of transitions, but the structure of the program graph will be exploited to determine ample sets for which
condition (A2) is guaranteed to hold. In fact, by means of a static analysis of the program graphs it will be determined which actions will be dependent. Let us consider determining these dependencies in more detail.

Overapproximating Dependencies Let \(D \subseteq A c t \times A c t\) be a binary relation on actions, that is intended to be an overapproximation of the dependency relation. That is, if \((\alpha, \beta) \in D, \alpha\) and \(\beta\) are considered to be dependent. As \(D\) is an overapproximation, it follows that if \((\alpha, \beta) \notin D\), then \(\alpha\) and \(\beta\) are independent; on the other hand, \((\alpha, \beta) \in D\) does not exclude that \(\alpha\) and \(\beta\) might be independent. It is evident that if (A2) holds for \(D\), then it also holds for the proper dependency relation. Why do we not consider the proper dependency relation, i.e., the complement of the independence relation? The problem is that the notion of independency is a global property. For instance, it might seem reasonable to consider the actions \(x:=z+y\) and \(x:=z\) to be dependent as they modify the same variable. This is a local criterion. However, if there is no state in \(T S\) with \(y \neq 0\) in which both actions are enabled, then they could be considered to be independent. The latter is, however, a global property. Since the aim is to avoid the construction of \(T S\), and to determine ample sets prior to the state-space generation, we have to rely on criteria that can be considered during a static analysis of the program graphs.

Let us briefly sketch which simple syntactic criteria can be employed to determine \(D\). It seems reasonable that all pairs of actions belonging to the same process are dependent. A simple, though rather conservative strategy would be to consider all pairs of actions that refer to the same variable to be dependent. This is somewhat restrictive as, e.g., the assignments \(x:=y+1\) and \(z:=y+z\) are in fact independent since they do not modify the shared variable \(y\). A more refined strategy would consider the modified variables. A simple, conservative strategy could be to consider all communication actions acting on the same channel to be dependent (i.e., in \(D\) ). This is somewhat restrictive as, e.g., actions \(c!v\) and \(c ? x\) for channel \(c\) with capacity one can never be enabled in the same global state: either the channel is full or empty. Any action involving a synchronous channel (a channel with capacity zero) is a joint action of two processes, and may thus be considered to be dependent on all actions in both involved processes.

Local Criteria for (A2) Local criteria which ensure that the choice ample \((s)=\operatorname{Act}_{i}(s)\) satisfies condition (A2) can be obtained as follows. ( \(\operatorname{Recall}^{\text {that }} \operatorname{Act}(s)=\operatorname{Act}(s) \cap \operatorname{Act} t_{i}\).) As before, we assume all actions for the processes \(\mathcal{P}_{1}, \ldots, \mathcal{P}_{n}\) to be pairwise disjoint. To ensure condition (A2), we check whether the following two (somewhat conservative) conditions hold in global state \(s\) :
(A2.1) Any \(\beta \in \operatorname{Act}_{j}\) is independent of \(\operatorname{Act}_{i}(s)\) for \(i \neq j\).
(A2.2) Any \(\beta \in A c t_{i} \backslash A c t(s)\) may not become enabled through the activities of some process \(\mathcal{P}_{j}\) with \(i \neq j\).

Condition (A2.1) can be established by inspecting the program graphs of all processes \(\mathcal{P}_{j}\) and checking whether for any \(\alpha \in A_{c t}{ }_{i}\) and \(\beta \in \operatorname{Act}_{j},(\alpha, \beta) \notin D\). Note that all actions local to process \(\mathcal{P}_{i}\) are considered to be dependent by (A2.1). Condition (A2.2) considers a global state \(s=\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{n}, \eta, \xi\right\rangle\) and action \(\beta\) of process \(\mathcal{P}_{i}\) that might be enabled in \(\ell_{i}\), i.e., \(\ell_{i} \stackrel{g: \beta}{\longrightarrow} \ell_{i}^{\prime}\) occurs in \(P G_{i}\), but is disabled in \(s\). This may be due to the fact that the guard \(g\) does not hold, or the action \(\beta\) might be blocked (e.g., a send action to a full channel). In order to ensure that \(\beta\) will not become enabled by activities of some other process \(\mathcal{P}_{j}\), it should be the case that neither:
- the guard \(g\) is violated in \(s\) (as \(g\) may refer to shared variables that can be modified by other processes), nor that
- \(\beta\) is a disabled communication action for a channel of nonzero capacity (which may become enabled as another process either enters or removes a message from the channel), nor that
- \(\beta\) is a disabled handshaking action (which may become enabled due to an activity of another process),
nor that for some \(i \neq j\), there is an edge \(\ell_{j}^{\prime} \stackrel{h: \gamma}{\hookrightarrow} \ell_{j}^{\prime \prime}\) in \(P G_{j}\) such that
- \(\ell_{j}^{\prime}\) is reachable \(\left(\right.\) via \(\left.\hookrightarrow^{*}\right)\) from the current location \(\ell_{j}\) (in state \(s\) ) in \(P G_{j}\), and action \(\gamma\) modifies some variable that occurs in the guard \(g\), or \(\gamma\) and \(\beta\) are complementary communication actions, i.e., send and receive actions on the same channel.

Thus, (A2.2) ensures that for global state \(s=\left\langle\ell_{1}, \ldots, \ell_{i}, \ldots, \ell_{n}, \eta, \xi\right\rangle\) there do not exist actions \(\beta \in A c t_{i} \backslash A c t(s)\) and \(\gamma \in \bigcup_{j \neq i} A c t_{j}\) and states
\[
s^{\prime}=\left\langle\ldots, \ell_{j}^{\prime}, \ldots, \ell_{i}, \ldots, \eta^{\prime}, \xi^{\prime}\right\rangle, \text { and } s^{\prime \prime}=\left\langle\ldots, \ell_{j}^{\prime \prime}, \ldots, \ell_{i}, \ldots, \eta^{\prime \prime}, \xi^{\prime \prime}\right\rangle
\]
where process \(\mathcal{P}_{i}\) resides in location \(\ell_{i}\) (as in state \(s\) ) such that:
\[
\beta \notin A c t\left(s^{\prime}\right) \text { and } \underbrace{s \rightarrow \ldots \rightarrow s^{\prime}}_{\beta \text { is not enabled }} \xrightarrow{\gamma} s^{\prime \prime} \xrightarrow{\beta} \ldots
\]

Let us consider a small example. Condition (A2.2) is violated in state \(s=\left\langle\ldots, \ell_{i}, \ldots, x=\right.\) \(0\rangle\) if there are edges
\[
\ell_{i} \stackrel{y>0: \beta}{\longrightarrow} \ell_{i}^{\prime} \text { and } \ell_{j}^{\prime} \xrightarrow{x:=1} \ell_{j}^{\prime \prime}
\]
in \(P G_{i}\) and \(P G_{j}\) with \(i \neq j\) such that \(\ell_{j}^{\prime}\) is reachable (via \(\hookrightarrow^{*}\) ) from \(\ell_{j}\). Note that the action pairs \((\beta, \gamma)\) referred to in (A2.2), as well as the reachability relation \(\hookrightarrow^{*}\) in the program graphs \(P G_{1}, \ldots, P G_{n}\), can be determined by a static analysis. The same applies to (A2.1). As the size of program graphs is relatively small compared to their underlying transition system, the cost of analyzing the program graphs for checking (A2.1) and (A2.2) is negligible compared to the analysis of their transition systems.

The following lemma shows that conditions (A2.1) and (A2.2) are sufficient (though not necessary) for (A2):

\section*{Lemma 8.27. Sufficient Local Criteria for (A2)}

If (A2.1) and (A2.2) hold, then ample \((s)=\operatorname{Act}_{i}(s)\) satisfies (A2) for all execution fragments in TS that start in state \(s\).

Proof: By contraposition. Suppose that (A2.1) and (A2.2) hold for ample(s). Assume (A2) does not hold and global state \(s\) is of the form \(\left\langle\ldots, \ell_{i}, \ldots\right\rangle\), i.e., process \(\mathcal{P}_{i}\) is at location \(\ell_{i}\). Then, there exists a finite execution fragment
\[
s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\beta_{n+1}} \ldots
\]
where \(\beta_{1}, \ldots, \beta_{n} \notin \operatorname{Act}_{i}(s)\) and \(\beta_{n+1}\) depends on \(\operatorname{ample}(s)=\operatorname{Act}_{i}(s)\). Since (A2.1) holds, actions depending on \(\operatorname{Act}_{i}(s)\) are actions of process \(i\) :
\[
\beta_{n+1} \in A c t_{i} \backslash A c t_{i}(s)
\]

Let \(m\) be the largest index in \(\{1, \ldots, n\}\) such that \(\beta_{1}, \ldots, \beta_{m-1}\) are actions of other processes, i.e.:
\[
\beta_{1}, \ldots, \beta_{m-1} \in \bigcup_{j \neq i} A c t_{j} \backslash A c t_{i} \quad \text { and } \quad \beta_{m} \in A c t_{i}
\]

Since the actions \(\beta_{1}, \ldots, \beta_{m-1}\) cannot affect the location of process \(\mathcal{P}_{i}\), location \(\ell_{i}\) does not change in the first \(m-1\) steps, i.e., states \(s_{1}, \ldots, s_{m-1}\) are also of the form \(\left\langle\ldots, \ell_{i}, \ldots\right\rangle\). As \(\beta_{m} \notin \operatorname{Act}_{i}(s)\) and \(\beta_{m} \in \operatorname{Act}_{i}\left(s_{m-1}\right)\), action \(\beta_{m}\) becomes enabled in location \(\ell_{i}\) by executing one of the actions in the set \(\left\{\beta_{1}, \ldots, \beta_{m-1}\right\}\). Since \(\beta_{1}, \ldots, \beta_{m-1}\) are actions of processes different from \(\mathcal{P}_{i}\), this contradicts (A2.2).

Algorithm 41 on page 634 summarizes the main steps for computing the ample set of a given state \(s\) such that conditions (A1), (A2), and (A3) are fulfilled. We consider here the case where the candidates are the action sets \(\operatorname{Act} t_{i}(s)\), i.e., the actions of \(\mathcal{P}_{i}\) that are enabled in \(s\).
```

Algorithm 41 Computation of ample( $s$ )
Input: state $s=\left\langle\ell_{1}, \ldots, \ell_{n}, \eta, \xi\right\rangle$ in TS
Output: ample $(s) \subseteq \operatorname{Act}(s)$ such that (A1)-(A3) are fulfilled

```
(* let \(D \subseteq A c t \times\) Act such that \((\alpha, \beta) \in D\) if \(\alpha\) and \(\beta\) are dependent *)
determine \(\operatorname{Act}_{i}(s)\) for all \(0<i \leqslant n\);
if \(\left(\exists i . \operatorname{Act}_{i}(s)=\operatorname{Act}(s)\right)\) then return \(\operatorname{Act}(s) \mathbf{f}\);
for \(i=1\) to \(n\) do (* check whether \(\operatorname{ample}(s)=\operatorname{Act}_{i}(s)\) is possible \(\left.{ }^{*}\right)\)
if \(\left(A c t_{i} \neq \varnothing\right.\) and \(A c t_{i}(s)\) only contains stutter actions) then
        if \(\left(\exists j \neq i . \operatorname{Act}_{i}(s) \times \operatorname{Act}_{j}(s) \cap D=\varnothing\right)\) then
            \(b:=\) true; \(\quad\) (* (A2.1) holds *)
            if \(\exists \ell_{i} \stackrel{g: \beta}{\longleftrightarrow} \ell_{i}^{\prime}\) in \(P G_{i}\) where \(\beta\) is a handshaking action then
                \(b:=\) false;
                                    (* (A2.2) violated *)
            else
                for all \(\ell_{i} \stackrel{g: \beta}{\hookrightarrow} \ell_{i}^{\prime}\) in \(P G_{i}\) and \(\ell_{j}^{\prime} \stackrel{h: \gamma}{\hookrightarrow} \ell_{j}^{\prime \prime}\) in \(P G_{j}\) with \(j \neq i\) and \(\ell_{j} \hookrightarrow^{*} \ell_{j}^{\prime}\) do
                if ( \(\eta \not \vDash g\) and \(\gamma\) modifies some variable that occurs in \(g\) ) or
                    ( \(\beta\) and \(\gamma\) are complementary communication actions) then
                    \(b:=\) false;
                                    (* (A2.2) violated \({ }^{*}\) )
                fi
                od
            fi
            if \((b)\) then return \(\operatorname{Act}_{i}(s)\) fi (* (A1)-(A3) hold *)
        fi
fi
od
                    (* ample(s) cannot be defined as the action set of one process *)
return \(\operatorname{Act}(s)\)
(* ample(s) :=Act(s)*)

\subsection*{8.2.4 Static Partial Order Reduction}

Rather than realizing the partial order reduction criteria during the model-checking process, in static partial order reduction the reduced transition system \(\hat{T} S\), or preferably, a high-level formal description of \(\hat{T S}\), is constructed prior to the verification. The advantage of this approach is that it may be combined with other state-space reduction techniques such as symbolic approaches using binary decision diagrams, or bisimulation minimization. This section treats static partial order reduction starting from the high-level description \(P G_{1}\||\ldots|\| P G_{n}\), where, for the sake of simplicity, it is assumed that the program graphs have shared variables but do not communicate via channels.

The main quest for static partial order reduction is to determine ample sets that satisfy the conditions (A1) through (A4). As described in Section 8.2.3, the conditions (A1), (A2), and (A3) can be established by means of local criteria of transition systems. It turns out that these criteria can also be easily reformulated for program graphs and thus can be checked prior to model checking. In contrast to on-the-fly partial order reduction, however, (A4) cannot be checked by establishing the strong cycle condition (A4 \(4^{\prime}\) ), as this condition is tailored to the state space generation by means of a depth-first search, a step that is absent in the static approach. Instead, (A4) is established by inspection of the program graphs \(P G_{1}\) through \(P G_{n}\) by fixing a set \(A_{\text {sticky }}\) of sticky actions:
\[
A_{\text {sticky }} \subseteq A c t
\]

The purpose of actions in \(A_{\text {sticky }}\) is to serve to "break" any cycle in \(\hat{T} S\). For state \(s\), action \(\alpha \in A_{\text {sticky }} \cap \operatorname{ample}(s)\) "sticks", so to speak, all other enabled actions in \(s\) and forces their exploration.

\section*{Notation 8.28. Visible Action}

An action \(\alpha \in\) Act is visible if \(\alpha\) is not a stutter action, i.e., if there exists a state \(s\) in TS with \(\alpha \in \operatorname{Act}(s)\) and \(L(s) \neq L(\alpha(s))\). Let Vis denote the set of visible actions.

The following requirements are imposed on \(A_{\text {sticky }}\).
(S1) Visibility condition
\(V\) is \(\subseteq A_{\text {sticky }}\)

The visibility condition (S1) asserts that all visible actions are sticky, or equivalently, that all actions in Act \(\backslash A_{\text {sticky }}\) are stutter actions.

\section*{(S2) Cycle-breaking condition}
\[
\text { For any cycle } s_{0} \xlongequal{\beta_{1}} \ldots \xlongequal{\beta_{n}} s_{n} \text { in } \hat{T S}, A_{\text {sticky }} \cap\left\{\beta_{1}, \ldots, \beta_{n}\right\} \neq \varnothing \text {. }
\]

Condition (S2) thus asserts that any cycle in \(\hat{T S}\) contains at least one sticky action.
Consider now the following new condition on ample sets:

\section*{(A3/4) Sticky condition}

If \(\operatorname{ample}(s) \neq \operatorname{Act}(s)\), then \(\operatorname{ample}(s) \cap A_{\text {sticky }}=\varnothing\).

Condition (A3/4) asserts that ample actions of nonfully expanded states are not sticky. The condition (A3/4) implies both the stutter condition (A3) and the strong cycle condition ( \(\mathrm{A} 4^{\prime}\) ). This is stated by the following lemma.

\section*{Lemma 8.29. Sticky vs. Stutter and Cycle Condition}

Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a finite, action-deterministic transition system and \(A_{\text {sticky }} \subseteq\) Act satisfying (S1) and (S2). If the ample sets for \(s \in S\) satisfy (A3/4), then (A3) and (A4') hold.

Proof: Let \(s \in S\) with ample \((s) \neq \operatorname{Act}(s)\) and suppose (A3/4) holds.
1. The sticky condition (A3/4) ensures that ample \((s) \cap A_{\text {sticky }}=\varnothing\). By the visibility condition (S1), any \(\alpha \in \operatorname{ample}(s)\) is a stutter action. Thus, (A3) holds.
2. Consider the cycle \(s_{0} \stackrel{\beta_{1}}{\Longrightarrow} s_{1} \ldots \xlongequal{\beta_{n}} s_{n}\) in \(\hat{T} S\). By the cycle-breaking condition (S2), we have \(\beta_{i} \in A_{\text {sticky }}\) for some \(0<i \leqslant n\). So \(\beta_{i} \in \operatorname{ample}\left(s_{i-1}\right) \cap A_{\text {sticky }}\). The sticky condition (A3/4) yields that \(s_{i-1}\) is fully expanded, i.e., \(\operatorname{ample}\left(s_{i-1}\right)=\operatorname{Act}\left(s_{i-1}\right)\). Hence, (A4') holds.

As a consequence we have the following result. If \(A_{\text {sticky }}\) satisfies (S1) and (S2), and the ample sets satisfy (A1), (A2), and (A3/4), then \(T S \triangleq \hat{T S}\). In the sequel of this section, we will discuss in detail how to check that all conditions (S1), (S2), (A1), (A2) and (A3/4) are fulfilled by inspecting the program graphs of the concurrent processes involved.

Program Graph Reduction. First, we discuss how to establish the ample set conditions by a static analysis of the program graphs. It is assumed that \(A_{\text {sticky }}\) satisfies (S1) and (S2); later on it is explained how this can be guaranteed. For the sake of simplicity, processes cannot communicate via channels (but may have shared variables). That is, the full transition system \(T S\) has the form \(T S=T S(P G)\) where \(P G=P G_{1}\| \| \ldots \| P G_{n}\) for program graphs \(P G_{1}, \ldots, P G_{n}\) that do not contain any communication actions on channels. As before, \(A c t_{i}\) denotes the action set of \(P G_{i}\) and \(A c t_{i} \cap A c t_{j}=\varnothing\) if \(i \neq j\), and that no \(\alpha \in A c t_{i}\) occurs on more than one edge in \(P G_{i}\). Thus, there is a one-to-one correspondence between the actions in \(P G_{i}\) and the edges in \(P G_{i}\). It is assumed that an overapproximation \(D\) of the dependency relation has been computed (see page 631). Finally, we assume that \(P G_{i}\) has a single starting location, denoted \(\ell_{0, i}\), and an initial condition \(g_{0, i}\).

The idea is to transform \(P G=P G_{1}\| \|\| \| P G_{n}\) into \(\widehat{P G}=\widehat{P G_{1}}\| \| \ldots \| \mid \widehat{P G_{n}}\) such that \(T S(P G) \triangleq T S(\widehat{P G})\). The transformation of each program graph \(P G_{i}\) proceeds in a number of steps. First, the edges in \(P G_{i}\) are marked with the labels good and sticky and locations with the label ample in the following way:
1. For any action \(\alpha \in A_{\text {sticky }} \cap A c t_{i}\), the (unique) edge \(\ell \stackrel{g: \alpha}{\hookrightarrow} \ell^{\prime}\) in \(P G_{i}\) is marked as sticky.
2. An edge \(\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}\) in \(P G_{i}\) is marked good if and only if for any action \(\beta\) in some program graph \(P G_{j}, \quad i \neq j,(\alpha, \beta) \notin D\) and variables modified by \(\beta\) do not occur in guard \(g\).
3. Location \(\ell\) of \(P G_{i}\) is marked as ample if all its outgoing edges \(\ell \hookrightarrow \ell^{\prime}\) are marked good, but not sticky, and the disjunction of the guards of all outgoing edges of \(\ell\) is equivalent to true.

After this (partial) labeling phase, the program graphs \(P G_{i}\) are modified as follows. Let ample \(_{1}, \ldots\), ample \(_{n}\) be new Boolean variables. These variables are set to true on edges that lead to locations that are marked as ample, and are set to false on all other edges:
4. Edge \(\ell \stackrel{g: \alpha}{\hookrightarrow} \ell^{\prime}\) in \(P G_{i}\) is replaced by \(\ell \stackrel{g: \hat{\alpha}}{\longleftrightarrow} \ell^{\prime}\), where \(\hat{\alpha}\) is the atomic execution of \(\alpha\) followed by ample \(i_{i}:=\) true whenever \(\ell^{\prime}\) is marked as ample, and \(\alpha\) followed by \(\mathrm{ample}_{i}:=\) false (atomically) otherwise.

In the next step, the guards of the edges in program graph \(P G_{i}\) are strengthened with propositions about the \(\mathrm{ample}_{i}\) boolean variables. Let
\[
\begin{aligned}
h_{i} & =\bigwedge_{1 \leqslant j<i} \neg \text { ample }_{j} \wedge \text { ample }_{i}, \quad i=1, \ldots, n \text { and } \\
f & =\bigwedge_{1 \leqslant j \leqslant n} \neg \text { ample }_{j} .
\end{aligned}
\]

The intuitive meaning of proposition \(h_{i}\) is that the enabled actions of process \(\mathcal{P}_{i}\) will serve as an ample set, while \(f\) indicates that the state corresponding to the current location is fully expanded. The guards in \(P G_{i}\) are adapted in the following way:
5. Any edge \(\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}\) in \(P G_{i}\) is replaced with the edge \(\ell \stackrel{\widehat{g}: \hat{\alpha}}{\longleftrightarrow} \ell^{\prime}\) where \(\widehat{g}=g \wedge h_{i}\) if the source location \(\ell\) is marked as ample, and \(\widehat{g}=g \wedge f\) otherwise.

The initial condition of \(\hat{P G_{i}}\) is defined as follows:
6. \(\hat{g}_{0, i}=g_{0, i} \wedge\) ample \(_{i}\) if the starting location \(\ell_{0, i}\) in \(P G_{i}\) is marked as ample, and \(\hat{g}_{0, i}=g_{0, i} \wedge \neg\) ample \(_{i}\), otherwise.

These transformations of \(P G_{i}\) yield the program graph \(\hat{P G_{i}}\). The reduced transition system \(\hat{T S}\) is obtained as follows:
\[
\hat{T S}=T S(\hat{P G}) \quad \text { where } \quad \hat{P G}=\hat{P G_{1}}\| \| \ldots \| \hat{P G_{n}}
\]

Some remarks are in order. First, observe that \(\hat{P G_{i}}\) is not a reduced variant of \(P G_{i}\), i.e., it contains the locations of \(P G_{i}\) and extends its variables with the auxiliary variables ample \(_{i}\). State \(\hat{s}\) in \(\hat{T S}\) has the form \(\left\langle\ell_{1}, \ldots, \ell_{n}, \widehat{\eta}\right\rangle\) where \(\ell_{i}\) is a location in \(\hat{P G_{i}}\) (and \(P G_{i}\) ) and \(\widehat{\eta}\) is a variable evaluation for the variables in \(P G_{1}, \ldots, P G_{n}\) and the auxiliary variables ample \(e_{1}, \ldots\), ample \(_{n}\). In contrast, state \(s\) in \(T S\) has the form \(\left\langle\ell_{1}, \ldots, \ell_{n}, \eta\right\rangle\) where \(\eta\) assigns values to the variables in \(P G_{1}, \ldots, P G_{n}\). Nevertheless, \(\hat{T S}\) can be considered as a subsystem of \(T S\) that results from the transition relation \(\Longrightarrow\) defined by the ample sets: on entering state \(\left\langle\ell_{1}, \ldots, \ell_{n}, \widehat{\eta}\right\rangle\) in \(\hat{T S}\) it holds \(\widehat{\eta}\left(\right.\) ample \(\left._{i}\right)=\) true if and only if location \(\ell_{i}\) of \(P G_{i}\) is marked as ample. This also applies to the initial state. In particular, the reachable fragment in \(\hat{T S}\) will not be larger than the reachable fragment in TS.

\section*{Lemma 8.30. Invariant for \(\hat{T S}\)}

For any state \(\left\langle\ell_{1}, \ldots, \ell_{n}, \widehat{\eta}\right\rangle\) in \(\operatorname{Reach}(\hat{T} S)\) and \(0<i \leqslant n\) :
\[
\widehat{\eta} \models \text { ample }_{i} \text { if and only if } \ell_{i} \text { is marked as ample. }
\]

The above lemma allows us to identify state \(\hat{s}=\left\langle\ell_{1}, \ldots, \ell_{n}, \widehat{\eta}\right\rangle\) in \(\hat{T S}\) with state \(s=\) \(\left\langle\ell_{1}, \ldots, \ell_{n}, \eta\right\rangle\) in \(T S\) where \(\eta\) results from \(\widehat{\eta}\) by ignoring the vaulation of the auxiliary
variables ample \({ }_{1}, \ldots\), ample \(_{n}\). The action \(\hat{\alpha}\) in \(\hat{T S}\) can be identified with action \(\alpha\) in TS. Due to the transformation of \(P G_{i}\), ample ( \(s\) ) for global state \(s\) in \(T S\) consists of actions \(\widehat{\alpha}\) for \(\alpha \in \operatorname{Act}(s)\). This permits treating ample( \((\hat{s})\) as a subset of \(\operatorname{Act}(s)\).

\section*{Example 8.31. Program Graph Transformation}

Consider the program graphs \(P G_{1}\) and \(P G_{2}\) in Figure 8.14. Intuitively, \(P G_{1}\) and \(P G_{2}\) both describe a process which counts iteratively from 0 to \(N\) where \(N\) is a fixed positive integer. Initially, \(x=y=0\) and \(b=c=\) false, where these variables are all local.


Figure 8.14: Two counting processes as program graph \(P G_{1}\) (left) and \(P G_{2}\) (right).
Let \(A P=\{b, c\}\), i.e., the only observable part of the program graphs is the value of \(b\) and \(c\). It follows that the actions \(\alpha_{1}, \beta_{1}, \alpha_{2}\), and \(\beta_{2}\) are stutter actions, whereas \(\gamma_{1}\) and \(\gamma_{2}\) are visible. As all visible actions need to be sticky actions, a first choice for \(A_{\text {sticky }}\) is:
\[
A_{\text {sticky }}=\left\{\gamma_{1}, \gamma_{2}\right\} .
\]

This choice also satisfies (S2), since any cycle in \(\hat{T S}\) contains \(\gamma_{1}\) or \(\gamma_{2}\)-a cycle can only occur if both \(x\) and \(y\) have the same value, and this can only occur by first counting until \(N\) and resetting to zero (via a \(\gamma\) action). Actions \(\alpha_{1}\) and \(\beta_{1}\) are independent of \(A_{c t_{2}}=\left\{\alpha_{2}, \beta_{2}, \gamma_{2}\right\}\), and symmetrically, \(\alpha_{2}\) and \(\beta_{2}\) are independent of \(A c t_{1}=\left\{\alpha_{1}, \beta_{1}, \gamma_{1}\right\}\).

Consider the transformation from \(P G_{1}\) and \(P G_{2}\) into \(\hat{P G_{1}}\) and \(\hat{P G_{2}}\), respectively. This goes along the steps described before. We start by marking the \(\alpha_{i}\) - and \(\beta_{i}\)-edges as good as these actions are independent. The \(\gamma_{i}\)-edges are marked as sticky. Accordingly, the initial locations \(n_{1}\) and \(n_{2}\) are marked as ample (they only have edges marked as ample), while locations \(\ell_{1}\) and \(\ell_{2}\) are not. Completing steps (4) through (6) yields the program graphs \(\hat{P G} G_{1}\) and \(\hat{P G_{2}}\) in Figure 8.15. The initial condition for the modified program graphs is
\[
x=y=0 \wedge b=c=\text { false } \wedge \text { ample }_{1}=\text { ample }_{2}=\text { true }
\]



Figure 8.15: Transformed program graphs \(\hat{P G_{1}}\) (top) and \(\hat{P G_{2}}\) (bottom).
We now consider the reduced transition system \(\hat{T S}=T S\left(\hat{P G_{1}} \| \hat{P G_{2}}\right)\). Observe that \(\hat{T S}\) is smaller than \(T S=T S\left(P G_{1}| | P G_{2}\right)\) since it avoids the interleavings of the action sequences \(\alpha_{1}^{N}\) and \(\alpha_{2}^{N}\). Instead, if both processes are in their initial location \(n_{i}\), then the auxiliary variables ample \({ }_{1}\) and \(\mathrm{ample}_{2}\) evaluate to true. The added conjuncts \(h_{1}=\) ample \(_{1}\) on the edge \(n_{1} \hookrightarrow n_{1}\) and \(h_{2}=\neg\) ample \(_{1} \wedge\) ample \(_{2}\) on the edge \(n_{2} \hookrightarrow n_{2}\) give a higher priority to \(\hat{P G_{1}}\) and thus enforce performing the action sequence \(\widehat{\alpha}_{1}^{N} \beta_{1}\) prior to performing \(\widehat{\alpha}_{2}^{N}\). This can be seen as follows.

For the initial state of \(\hat{T} S\),
\[
\hat{s}=\left\langle n_{1}, n_{2}, x=y=0, b=c=\text { false }, \text { ample }_{1}=\text { ample }_{2}=\text { true }\right\rangle,
\]
we have
\[
\operatorname{Act}(\hat{s})=\left\{\widehat{\alpha}_{1}\right\} \quad \text { while } \quad \operatorname{Act}(s)=\left\{\alpha_{1}, \alpha_{2}\right\}
\]
where \(s=\left\langle n_{1}, n_{2}, x=y=0, b=c=\right.\) false \(\rangle\) is the state corresponding to \(\hat{s}\) in TS. The same applies to all reachable states \(s\) of \(T S\) where both processes are in their initial location and \(x<N\) and \(y<N\). Thus, in all these states, a significant reduction results. Consider
now a state \(t\) of the form
\[
t=\left\langle n_{1}, n_{2}, x=N, y=k, b=\ldots, c=\ldots\right\rangle \quad \text { where } 0 \leqslant k<N
\]
in TS. The corresponding state in \(\hat{T S}\) is
\[
\hat{t}=\left\langle n_{1}, n_{2}, x=N, y=k, b=\ldots, c=\ldots, \text { ample }_{1}=\text { ample }_{2}=\text { true }\right\rangle .
\]

We have \(\operatorname{Act}(\hat{t})=\left\{\widehat{\beta}_{1}\right\}\) which is a proper subset of \(\operatorname{Act}(t)=\left\{\beta_{1}, \alpha_{2}\right\}\). By symmetry, similar observations apply to the state where both program graphs are in their initial location, \(y=N\) and \(x<N\). State
\[
u=\left\langle\ell_{1}, n_{2}, x=N, y=k, b=\ldots, c=\ldots\right\rangle \quad \text { with } \quad \operatorname{Act}(u)=\left\{\gamma_{1}, \alpha_{2}\right\}
\]
in \(T S\) where \(0 \leqslant k<N\) corresponds to the following state in \(\hat{T S}\) :
\[
\hat{u}=\left\langle\ell_{1}, n_{2}, x=N, y=k, b=\ldots, c=\ldots, \text { ample } e_{1}=\text { false }, \text { ample } e_{2}=\text { true }\right\rangle
\]
with \(\operatorname{Act}(\hat{u})=\left\{\widehat{\alpha}_{2}\right\} \subset \operatorname{Act}(u)\). The only fully expanded states are the states of the form
\[
v=\left\langle\ell_{1}, \ell_{2}, x=y=N, b=\ldots, c=\ldots\right\rangle
\]
that correspond to
\[
\hat{v}=\left\langle\ell_{1}, \ell_{2}, x=y=N, b=\ldots, c=\ldots, \text { ample }_{1}=\text { ample }_{2}=\text { false }\right\rangle .
\]

Hence \(\operatorname{Act}(\hat{v})=\operatorname{Act}(v)=\left\{\gamma_{1}, \gamma_{2}\right\}\).

\section*{Theorem 8.32. Correctness of the Sticky Set Approach}

Let \(P G=P G_{1}\| \|\| \| P G_{n}\), and \(\hat{P G}=\hat{P G_{1}}\| \| \| \hat{P G_{n}}, T S=T S(P G)\) and \(\hat{T S}=\) TS \((\hat{P G})\). Then
\[
\text { if } A_{\text {sticky }} \text { satisfies (S1) and (S2). then } \hat{T S} \triangleq T S .
\]

Proof: Let \(P G=P G_{1}\| \| \ldots \| P G_{n}\) and \(\hat{P G}=\hat{P G_{1}}\| \| \ldots \| \hat{P G_{n}}\). Assume \(A_{\text {sticky }}\) satisfies (S1) and (S2). We prove that conditions (A1), (A2), and (A3/4) hold for all states in Reach \((\hat{T} S)\). Since (A3/4) implies (A3) and (A4 \({ }^{\prime}\) ), and (A4 \({ }^{\prime}\) ) implies (A4), it follows by Theorem 8.13 that \(T S \triangleq \hat{T S}\).

Let \(\hat{s}=\left\langle\ell_{1}, \ldots, \ell_{n}, \widehat{\eta}\right\rangle\) be a reachable state in \(\hat{T} S\) with corresponding state \(s=\left\langle\ell_{1}, \ldots, \ell_{n}, \eta\right\rangle\) in TS. Consider two cases.
1. None of the locations \(\ell_{i}\) is marked as ample. Then \(P G\) and \(\hat{P G}\) have the same outgoing edges in (global) location \(\left\langle\ell_{1}, \ldots, \ell_{n}\right\rangle\). Moreover, \(\eta \vDash g\) if and only if \(\widehat{\eta} \vDash g\) for any guard \(g\) that occurs in \(P G\). (Recall that \(\eta\) and \(\widehat{\eta}\) agree on the original program variables.) Hence, state \(\hat{s}\) is fully expanded, i.e.,
\[
\operatorname{Act}(\hat{s})=\{\widehat{\alpha} \mid \alpha \in \operatorname{Act}(s)\} .
\]

It follows that conditions (A1), (A2), and (A3/4) are then fulfilled.
2. \(\ell_{i}\) is marked as ample and \(\ell_{1}, \ldots, \ell_{i-1}\) are not marked as ample. By Lemma 8.30, \(\widehat{\eta} \models h_{i}\). For \(1 \leqslant j<i\), any outgoing edge from \(\ell_{j}\) in \(\hat{P G_{j}}\) is either guarded by \(h_{j}\) or \(f\). As \(\widehat{\eta}\left(\right.\) ample \(\left._{j}\right)=\) true, we get \(\widehat{\eta} \not \vDash h_{j}\). Similarly, as \(\widehat{\eta}\left(\right.\) ample \(\left._{i}\right)=\) true, we have \(\widehat{\eta} \not \vDash f\). Thus, for \(1 \leqslant j<i\), none of the outgoing edges from \(\ell_{j}\) in \(\hat{P G_{j}}\) is enabled in state \(\hat{s}\). Similarly, for \(i<j \leqslant n\) none of the outgoing edges from \(\ell_{j}\) in \(P G_{j}\) is enabled since these edges are guarded by \(h_{j}\) or \(f\) as both require ample \({ }_{i}\) to be false. Thus:
\[
\operatorname{ample}(\hat{s}) \subseteq\left\{\widehat{\alpha} \mid \alpha \in \operatorname{Act}_{i}(s)\right\}
\]

Since \(\ell_{i}\) is marked as ample, any action \(\alpha\) of an edge \(\ell_{i} \stackrel{g: \alpha}{\longrightarrow} \ell_{i}^{\prime}\) in \(P G_{i}\) is marked as good, but not marked as sticky. In particular, \(\operatorname{Act} t_{i}(s) \cap A_{\text {sticky }}=\varnothing\) which yields:
\[
\operatorname{ample}(\hat{s}) \cap A_{\text {sticky }}=\varnothing
\]
(where we identify \(\alpha\) and \(\widehat{\alpha}\) ). Thus, condition (A3/4) is fulfilled. Since all actions in ample ( \(\hat{s}\) ) are marked as good, conditions (A2.1) and (A2.2) are fulfilled. By Lemma 8.27 (page 633) it follows that the dependency condition (A2) holds. The nonemptiness condition (A1) is fulfilled since the mark ample ensures that the disjunction of the guards of all outgoing edges of \(\ell_{i}\) is true.

Computing Sticky Sets It remains to explain how a set \(A_{\text {sticky }}\) satisfying the conditions (S1) and (S2) can be determined by a static analysis of the program graph. The obvious choice \(A_{\text {sticky }}=\) Act would evidently suffice, but then (A3/4) would not allow for any reduction of the state space. The goal is to obtain a set \(A_{\text {sticky }}\) satisfying (S1) and (S2) that is as small as possible. Initially, we set \(A_{\text {sticky }}=\) Vis, the set of visible actions. This ensures the visibility condition (S1). In order to establish (S2), actions are added until any cycle in \(\hat{T S}\) contains an action in \(A_{\text {sticky }}\). This is done by a static analysis of the program graph, attempting to keep the cardinality of \(A_{\text {sticky }}\) minimal.

\section*{Notation 8.33. Control Path and Control Cycle}

A control path of \(P G_{i}\) is a sequence of edges in \(P G_{i}\) of the form
\[
\ell_{0} \stackrel{g_{1}: \alpha_{1}}{>} \ell_{1} \stackrel{g_{2}: \alpha_{2}}{>} \ldots \stackrel{{ }_{k}: \alpha_{k}}{>} \ell_{k} .
\]

A control path in \(P G_{i}\) is a control cycle if \(\ell_{k}=\ell_{0}\) and \(k>0\).

\section*{Notation 8.34. \(P G_{i}\)-Projection}

Let \(\varrho=s_{0} \xrightarrow{\alpha_{1}} \ldots \xrightarrow{\alpha_{k}} s_{k}\) be an execution fragment in \(T S\) where \(s_{j}=\left\langle\ell_{1}^{j}, \ldots, \ell_{n}^{j}, \eta^{j}\right\rangle\) for \(0 \leqslant j \leqslant k\). The \(P G_{i}\)-projection of \(\varrho\) is the control path of \(P G_{i}\) resulting from \(\varrho\) by (1) removing all transitions \(s_{j-1} \xrightarrow{\alpha_{j}} s_{j}\) where \(\alpha_{j} \notin A c t_{i}\) and (2) replacing the transitions \(s_{i-1} \xrightarrow{\alpha_{i}} s_{i}\) where \(\alpha_{j} \in A c t_{i}\) with the corresponding edge \(\ell_{i}^{j-1} \xlongequal{g_{j}: \alpha_{j}} \ell_{i}^{j}\).

The \(P G_{i}\)-projection of an execution fragment in \(T S\) is the control path in \(P G_{i}\) that has been taken in \(\varrho\). Vice versa, any execution fragment in \(T S\) arises by the combination of control paths in some of the program graphs \(P G_{1}, \ldots, P G_{n}\), possibly in an interleaved way.

\section*{Example 8.35. Projections}

Consider \(T S=T S\left(P G_{1} \| P G_{2}\right)\) where \(P G_{1}, P G_{2}\) are depicted in Figure 8.14 (page 639). For \(N=2\), the execution fragment
\[
\begin{array}{ll}
\left\langle n_{1}, n_{2}, x=0, y=0, b=c=\text { false }\right\rangle & \xrightarrow{\alpha_{1}} \\
\left\langle n_{1}, n_{2}, x=1, y=0, b=c=\text { false }\right\rangle & \xrightarrow{\alpha_{1}} \\
\left\langle n_{1}, n_{2}, x=2, y=0, b=c=\text { false }\right\rangle & \xrightarrow{\alpha_{2}} \\
\left\langle n_{1}, n_{2}, x=2, y=1, b=c=\text { false }\right\rangle & \xrightarrow{\beta_{1}} \\
\left\langle\ell_{1}, n_{2}, x=2, y=0, b=c=\text { false }\right\rangle &
\end{array}
\]
yields the \(P G_{1}\)-projection \(n_{1} \xrightarrow{x<2: \alpha_{1}} n_{1} \xrightarrow{x<2: \alpha_{1}} n_{1} \xrightarrow{x=2: \beta_{1}} \ell_{1}\) and the \(P G_{2^{-}}\) projection \(n_{2} \xrightarrow{y<2: \alpha_{1}} n_{2}\). Note that the \(P G_{i}\)-projection of \(\varrho\) might have length 0 (i.e., solely consists of the location \(\ell_{i}^{0}\) ) in which case \(\varrho\) does not take any edge of \(P G_{i}\).

The following lemma is obvious:

\section*{Lemma 8.36. Global Cycles vs. Control Cycle}

For each cycle \(s_{0} \xrightarrow{\alpha_{1}} \ldots \xrightarrow{\alpha_{m}} s_{m}\) in TS and for any \(0<j \leqslant n\) :
if \(\left\{\alpha_{1}, \ldots, \alpha_{m}\right\} \cap A c t_{j} \neq \varnothing\), then the \(P G_{j}\)-projection is a control cycle.

The converse of Lemma 8.36 does not hold, since, e.g., the control cycle \(n_{1} \xrightarrow{x<N: \alpha_{1}} n_{1}\) in the program graph \(P G_{1}\) of Figure 8.14 does not induce a cycle in \(T S\) since the value of \(x\) is strictly increasing in each iteration.

By Lemma 8.36, the cycle-breaking condition (S2) is obviously fulfilled if \(A_{\text {sticky }}\) contains at least one action of any control cycle in \(P G_{1}, \ldots, P G_{n}\). This is guaranteed as follows. Let \(A_{\text {sticky }}=\) Vis, the set of visible actions. This ensures (S1). As each visible action is in \(A_{\text {sticky }}\), it suffices for establishing ( S 2 ) that any cycle that does not contain a visible action is considered. To that end, all edges from \(P G_{i}\) with a visible action are removed. By means of a depth-first search of the resulting program graph \(P G_{i}^{\prime}\), say, control cycles are determined. Whenever a backward edge \(\ell \stackrel{g: \alpha}{\longleftrightarrow} \ell^{\prime}\) is found, \(\alpha\) is added to the sticky set. Applying this DFS-based approach to each program graph \(P G_{i}\) yields \(A_{\text {sticky }}\). From Lemma 8.36 together with the fact that for any control cycle in \(P G_{i}^{\prime}\) the depth-first search will find a backward edge on this cycle, it follows that \(A_{\text {sticky }}\) satisfies (S2). Hence, \(A_{\text {sticky }}\) contains at least one action of each control cycle in \(P G_{i}^{\prime}\). The remaining control cycles in \(P G_{i}\) contain at least one visible action, and thus, an action in Vis \(\subseteq A_{\text {sticky }}\).

This strategy often generates a rather large \(A_{\text {sticky }}\) set. For instance, for the program graphs in Figure 8.14, the set
\[
A_{\text {sticky }}=\underbrace{\left\{\gamma_{1}, \gamma_{2}\right\}}_{=\text {Vis }} \cup\left\{\alpha_{1}, \alpha_{2}\right\}
\]
is obtained because of the control cycles \(n_{1} \xrightarrow{x<N: \alpha_{1}} n_{1}\) and \(n_{2} \xrightarrow{y<N: \alpha_{2}} n_{2}\). However, these control cycles do not correspond to cycles in TS since each execution of \(\alpha_{1}\) strictly increases the value of \(x\). The same applies to \(\alpha_{2}\) and variable \(y\). In order to find control cycles that do not correspond to a global cycle, i.e., a cycle in \(T S\), we impose for each program variable \(x\) :
\[
\text { a transitive, irreflexive relation } \prec_{x} \text { on the domain of } x \text {. }
\]

For example, for integer variable \(x\), the relation \(\prec_{x}\) could be the natural order \(<\) or the reverse natural order \(>\). For Boolean variable \(x\), the order with false \(\prec_{x}\) true and true \(\nprec_{x}\) false is appropriate. The relation \(\prec_{x}\) should be chosen such that the classification of the actions into incrementing, decrementing, neutral, or complex actions (see below) is algorithmically simple. For variable \(x\) and order \(\prec_{x}\), the effect of action \(\alpha \in\) Act is said to be
- incrementing on \(x\) if \(\eta(x) \prec_{x} \operatorname{Effect}(\eta, \alpha)(x)\) for each valuation \(\eta\),
- decrementing on \(x\) if \(\operatorname{Effect}(\eta, \alpha)(x) \prec_{x} \eta(x)\) for each valuation \(\eta\),
- neutral on \(x\) if \(\operatorname{Effect}(\eta, \alpha)(x) \nprec_{x} \eta(x)\) and \(\eta(x) \nprec_{x} \operatorname{Effect}(\eta, \alpha)(x)\), for each valuation \(\eta\).
- complex in all other cases.

For instance, for integer variable \(x\) where \(\prec_{x}\) is the natural order \(<\) we have: \(x:=x+2\) has an incrementing effect on \(x\), action \(x:=x-5\) has a decrementing effect on \(x, y:=y-5\) is neutral on \(x\), and \(x:=y\) has a complex effect on \(x\).

A control path \(\ell_{0} \stackrel{g_{1}: \alpha_{1}}{\longrightarrow} \ldots \stackrel{g_{k}: \alpha_{k}}{\longrightarrow} \ell_{k}\) has incrementing effect on \(x\) if (1) the effect of \(\alpha_{j}\) on \(x\) is incrementing for some \(j \in\{1, \ldots, k\}\) and (2) for each \(j \in\{1, \ldots, r\}, j \neq k\), the effect of \(\alpha_{j}\) on \(x\) is either incrementing or neutral. Control paths with a decrementing effect on \(x\) are defined analogously. For an execution fragment
\[
s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{m}} s_{m}
\]
such that for some \(i \in\{1, \ldots, n\}\)
- the \(P G_{i}\)-projection is a control path \(\ell_{0} \stackrel{g_{1}: \alpha_{1}}{>} \ell_{1} \ldots \stackrel{g_{r}: \alpha_{r}}{\longrightarrow} \ell_{r}\) which has an incrementing effect on \(x\), and
- the \(P G_{j}\)-projection for \(j \neq i\) does not contain an action which has either a decrementing or complex effect on \(x\),
the value of \(x\) is strictly increasing with respect to the order \(\prec_{x}\). The transitivity of \(\prec_{x}\) yields \(v_{0} \prec_{x} v_{m}\), where \(v_{0}\) and \(v_{m}\) are the value of \(x\) in the state \(s_{0}\) and \(s_{m}\), respectively. Since \(\prec_{x}\) is irreflexive, \(v_{0} \neq v_{m}\). Thus, the given execution fragment cannot be a cycle in \(T S\), i.e., \(s_{0} \neq s_{m}\). A similar observation holds for control cycles with a decrementing effect on \(x\).

\section*{Example 8.37. Control Cycles vs. Global Cycles}

Consider the program graphs in Figure 8.16 where \(N>1\) is a fixed integer, \(x\) and \(y\) are nonnegative integer variables, and \(b, c\) Boolean variables. Program graph \(P G_{1}\) has the control cycle
\[
n_{1} \xrightarrow{x<N: x:=x+1} \ell_{1}>\overbrace{}^{b:=\neg b} n_{1} .
\]

For \(\prec_{x}=<\), this control cycle has an incrementing effect on \(x\). None of the actions \(\alpha_{2}, \beta_{2}, \gamma_{2}\) in \(P G_{2}\) modify the value of \(x\), i.e., these actions have a neutral effect on \(x\). Thus, there is no cycle in \(T S\) for which the \(P G_{1}\)-projection coincides with the above control cycle.


Figure 8.16: Program graphs \(P G_{1}\) (left) and \(P G_{2}\) (right).


Figure 8.17: Program graphs \(P G_{1}\) (left) and \(P G_{2}\) (right).

Now consider the program graphs in Figure 8.17. The control cycle
\[
n_{1} \stackrel{x<N: x:=x+1}{l_{1}>} \xrightarrow{b:=\neg b} n_{1}
\]
in \(P G_{1}\) has an incrementing effect on \(x\). However, action \(\alpha_{2}\) in \(P G_{2}\) has a decreasing effect on \(x\). In fact, there is a global cycle in \(\operatorname{TS}\left(P G_{1} \| P G_{2}\right)\) of the form
\[
\begin{aligned}
s= & \left\langle n_{1}, n_{2}, x=0, y=0, b=\text { false }, c=\text { false }\right\rangle \\
& \xrightarrow{\alpha_{1}} \\
& \left\langle\ell_{1}, n_{2}, x=1, y=0, b=\text { false }, c=\text { false }\right\rangle \\
& \xrightarrow{\beta_{1}} \\
& \left\langle n_{1}, n_{2}, x=1, y=0, b=\text { true }, c=\text { false }\right\rangle \\
& \xrightarrow{\alpha_{1}} \\
& \left\langle\ell_{1}, n_{2}, x=2, y=0, b=\text { true }, c=\text { false }\right\rangle \\
& \xrightarrow{\beta_{1}} \\
& \left.n_{1}, n_{2}, x=2, y=0, b=\text { false }, c=\text { false }\right\rangle \\
& \xrightarrow{\alpha_{2}} \\
& \left\langle n_{1}, \ell_{2}, x=0, y=0, b=\text { false }, c=\text { false }\right\rangle \\
& \xrightarrow{\beta_{2}} \\
\left\langle n_{1}, n_{2}, x=0, y=0, b=\text { false }, c=\text { false }\right\rangle & =s
\end{aligned}
\]
where the \(P G_{1}\)-projection yields the (duplication of the) above control cycle.

The goal is now to compute \(A_{\text {sticky }}\) by starting with \(A=V\) is (as before) and adding further actions to break any control cycle of program graph \(P G_{i}\), say, that only contains stutter actions, except for actions that have an incrementing or a decrementing effect on some variable \(x\) and for which there is no action in \(P G_{j}, i \neq j\) with an opposite effect on \(x\).

\section*{Notation 8.38. Opposite Actions}

Actions \(\alpha\) and \(\beta \in\) Act are opposite if there exists a variable \(x\) such that \(\alpha\) has either an incrementing or complex effect on \(x\) and \(\beta\) has either a decrementing or complex effect on \(x\), or vice versa. Let
\[
O p p(\alpha)=\{\beta \in A c t \mid \alpha \text { and } \beta \text { are opposite }\} .
\]

Note that any action \(\alpha\) with complex effect on some variable is opposite to itself, in which case \(\alpha \in \operatorname{Opp}(\alpha)\).

The computation of \(A_{\text {sticky }}\) proceeds as outlined in Algorithm 42 on page 648. Here, an action \(\alpha\) is called monotonic on \(x\) if it has either an incrementing or decrementing effect on \(x\).

\section*{Example 8.39. Computing \(A_{\text {sticky }}\)}

Reconsider the program graphs \(P G_{1}\) and \(P G_{2}\) in Figure 8.16 (page 646). Let \(\prec_{x}=<=\prec_{y}\) and false \(\prec\) true for the Boolean variables \(b\) and \(c\). Moreover, we assume that AP only refers to variable \(y\), but not to \(x, b\), or \(c\). Then, action \(\alpha_{2}\) is visible and all other actions are stutter actions. Hence, Algorithm 42 starts with \(A_{\text {sticky }}=\) Vis \(=\left\{\alpha_{2}\right\}\). Removal of all visible edges yields the program graphs in Figure 8.18.


Figure 8.18: Program graphs \(P G_{1}\) (left) and \(P G_{2}\) (right) after removal of visible edges.

The algorithm obtains \(M_{1}=\left\{\alpha_{1}\right\}\) since \(\alpha_{1}\) has an incrementing effect on \(x\), but there
```

Algorithm 42 Computation of $A_{\text {sticky }}$
Input: program graphs $P G_{1}, \ldots P G_{n}$
Output: $A_{\text {sticky }} \subseteq$ Act satisfying (S1) and (S2)
$A_{\text {sticky }}:=$ Vis; $\quad\left(*\right.$ establish $\left.(\mathrm{S} 1){ }^{*}\right)$
for $i=1$ to $n$ do
$P G_{i}:=P G_{i}$ where any edge $\ell \stackrel{g: \alpha}{\longrightarrow} \ell^{\prime}$ with $\alpha \in$ Vis is removed;
od
for $i=1$ to $n$ do
$M_{i}:=\left\{\alpha \in A c t_{i} \mid \alpha\right.$ is monotonic $\left.\wedge O p p(\alpha) \subseteq \bigcup_{j<k \leqslant n} A c t_{k}\right\} ;$
$P G_{i}:=P G_{i}$ where any edge $\ell \stackrel{g: \alpha}{\hookrightarrow} \ell^{\prime}$ with $\alpha \in M_{j}$ is removed;
od
for $i=1$ to $n$ do
perform a DFS on $P G_{i}$;
$A_{\text {sticky }}:=A_{\text {sticky }} \cup\left\{\alpha \mid \ell \stackrel{g: \alpha}{\hookrightarrow} \ell^{\prime}\right.$ is a backward edge in $\left.P G_{i}\right\} ;$
od
return $A_{\text {sticky }}$
(* $A_{\text {sticky }}$ fulfills (S1) and (S2) $\left.{ }^{*}\right)$

```
is no opposite action. Hence, the \(\alpha_{1}\)-edge from \(n_{1}\) is \(\ell_{1}\) will be removed. The resulting program graph is acyclic; no backward edge is found.

We obtain \(M_{2}=\varnothing\) as there is no monotonic action in the modified \(P G_{2}\). Hence, one of the actions \(\beta_{2}\) or \(\gamma_{2}\) becomes sticky (i.e., is added to \(A_{\text {sticky }}\) ), depending on the starting location for the depth-first search. The result of Algorithm 42 is thus either \(A_{\text {sticky }}=\left\{\alpha_{2}, \beta_{2}\right\}\) or \(A_{\text {sticky }}=\left\{\alpha_{2}, \gamma_{2}\right\}\).

\section*{Lemma 8.40. Soundness of Algorithm 42}

The set \(A_{\text {sticky }} \subseteq\) Act returned by Algorithm 42 satisfies (S1) and (S2).

Proof: The visibility condition (S1) obviously holds since initially \(A_{\text {sticky }}=\) Vis, and no actions are removed from \(A_{\text {sticky }}\). Now consider (S2). Let \(\varrho\) be a global cycle of the form
\[
s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{m}} s_{m}=s_{0} .
\]

If some of the actions \(\beta_{1}, \ldots, \beta_{m}\) are visible, then the cycle-breaking condition (S2) holds since Vis \(\subseteq A_{\text {sticky }}\). Assume \(\beta_{1}, \ldots, \beta_{m}\) are stutter actions. We prove that there exists some \(j\) such that (some of) the actions in \(\left(\left\{\beta_{1}, \ldots, \beta_{m}\right\} \cap A c t_{j}\right) \backslash M_{i}\) yield a control cycle \(\vartheta\) in \(P G_{j}^{\prime \prime}\) (the variant of \(P G_{j}\) obtained after removing all edges with action in \(M_{j}\) ). From this result, it follows that the depth-first search in \(P G_{j}^{\prime \prime}\) will classify one of the edges in \(\vartheta\) as a backward edge, and hence, adds at least one of the actions in \(\left\{\beta_{1}, \ldots, \beta_{m}\right\} \cap A_{c t}\) 的 \(A_{\text {sticky }}\). This yields \(A_{\text {sticky }} \cap\left\{\beta_{1}, \ldots, \beta_{m}\right\} \neq \varnothing\).

Let \(J\) be the set of indices \(j \in\{1, \ldots, n\}\) with \(\left\{\beta_{1}, \ldots, \beta_{m}\right\} \cap A c t_{j} \neq \varnothing\). Clearly, \(J \neq \varnothing\). Let \(j=\max J\) and assume there is no control cycle in \(P G_{j}^{\prime \prime}\) where the underlying action set is contained in
\[
\left(\left\{\beta_{1}, \ldots, \beta_{m}\right\} \cap A c t_{j}\right) \backslash M_{j} .
\]

Since the \(P G_{j}\)-projection of the given global cycle \(\varrho\) is a control cycle \(\vartheta\) in \(P G_{j}^{\prime}\) (Lemma 8.36 on page 643), the control cycle \(\vartheta\) in \(P G_{j}^{\prime}\) must contain an action \(\beta_{i} \in M_{j}\). Otherwise, \(\vartheta\) is a control cycle in \(P G_{j}^{\prime \prime}\) built by actions in \(\left(\left\{\beta_{1}, \ldots, \beta_{m}\right\} \cap A c t_{j}\right) \backslash M_{j}\).

By definition of \(M_{j}\), action \(\beta_{i}\) is monotonic. Since \(\beta_{1}, \ldots, \beta_{m}\) forms a global cycle in TS, there must be another action \(\beta_{h}, h \neq i\), on this cycle with an opposite effect:
\[
\beta_{h} \in O p p\left(\beta_{i}\right)
\]

Again, by definition of \(M_{j}\), the actions that are opposite to \(\beta_{i}\) belong to \(A c t_{k}\) for some \(k>j\). Thus, \(\beta_{h} \in\) Act \(_{k}\) for some \(k>j\). But then, \(k \in J\) and \(k>j\), which contradicts \(j=\max J\).

\subsection*{8.3 The Branching-Time Ample Set Approach}

The previous section presented conditions for the ample sets that ensure the stutter trace equivalence of \(T S\) and \(\hat{T} S\). Hence, these conditions are sound for verifying \(\mathrm{LTL}_{\backslash \bigcirc}\). This section treats partial order reduction for the branching-time temporal logics \(\mathrm{CTL}_{\backslash \bigcirc}\) and \(\mathrm{CTL}_{\uparrow \bigcirc}^{*}\). The aim is to adopt the ample set approach, and adapt it such that rather than establishing \(T S \triangleq \hat{T} S\) we obtain that \(T S \approx d i v \hat{T} S\). As \(\approx^{d i v}\) coincides with the logical equivalence of \(\mathrm{CTL}_{\backslash \bigcirc}\) and \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) (see Section 7.8.3 on page 560), this yields a sound approach for verifying these logics. We first argue by means of an example that the conditions (A1) through (A4) do not suffice for \(\mathrm{CTL}_{\backslash \mathrm{O}}\).

Example 8.41. Conditions (A1)-(A4) Are Insufficient for \(C T L_{\backslash 0}\)
Consider the transition system \(T S\) in Figure 8.19 (left) over \(A P=\{a, b, c\}\). Actions \(\alpha\) and \(\delta\) are independent of \(\{\beta, \gamma\}\) and \(\beta, \gamma\) are stutter actions. Let \(\operatorname{ample}\left(s_{0}\right)=\{\beta, \gamma\}\), and \(\operatorname{ample}(s)=\operatorname{Act}(s)\) for all states \(s \neq s_{0}\); see the reduced transition system \(\hat{T S}\) in Figure 8.19 (right).


Figure 8.19: Reduction satisfies (A1)-(A4), but \(T S \not \equiv_{C T L} \hat{T S}\).

Consider the \(\mathrm{CTL}_{\backslash \bigcirc}\) formula:
\[
\Phi=\forall \square(a \rightarrow(\forall \diamond b \vee \forall \diamond c)) .
\]

Formula \(\Phi\) asserts that for each reachable \(a\)-state \(s\) either any path from \(s\) eventually reaches a \(b\)-state or any path from \(s\) eventually reaches a \(c\)-state. Hence \(\hat{T} S \models \Phi\) as any \(a\)-state in \(T S\) has either a direct \(b\) - or \(c\)-successor, but \(T S \not \vDash \Phi\) as state \(u \not \vDash \forall \diamond b \vee \forall \diamond c\). The latter follows from the fact that
\[
\operatorname{Traces}_{T S}(u)=\left\{\{a\}\{a\}\{b\}^{\omega},\{a\}\{a\}\{c\}^{\omega}\right\} .
\]

As \(\hat{T S} \models \Phi\) and \(T S \not \vDash \Phi\), we get \(T S \not \approx \operatorname{div} \hat{T S}\). In fact, we have
\[
\beta\left(s_{0}\right) \not \overbrace{}^{d i v} \gamma\left(s_{0}\right) \not \overbrace{}^{d i v} \alpha\left(s_{0}\right)
\]
as \(\beta\left(s_{0}\right)\) can reach a \(b\)-state, while \(\gamma\left(s_{0}\right)\) cannot, and vice versa \(\gamma\left(s_{0}\right)\) can reach a \(c\)-state, while \(\beta\left(s_{0}\right)\) cannot. Moreover, \(\alpha\left(s_{0}\right)\) (and \(s_{0}\) ) can reach both a \(b\) - and a \(c\)-state. The same holds for the \(\alpha\)-successor \(s_{1}\) of \(s_{0}\). Since \(\hat{T S}\) does not contain \(s\) (or any equivalent state), TS \(\not \approx d i v \hat{T S}\).

Since \(s_{0} \approx \operatorname{div} \alpha\left(s_{0}\right)\), the reduction obtained by \(\operatorname{ample}\left(s_{0}\right)=\{\alpha\}\), and \(\operatorname{ample}(s)=\operatorname{Act}(s)\) for any state \(s \neq s_{0}\) yields the reduced transition system \(\hat{T S}\) in Figure 8.20). Now we have \(\hat{T S} \approx{ }^{d i v} T S\).


Figure 8.20: A sound partial order reduction for \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\).
The task is now to provide sufficient ample set conditions that ensure \(T S \approx{ }^{d i v} \hat{T S}\). Since \(\triangleq\) is strictly coarser than \(\approx^{d i v}\), the conditions (A1) though (A4) are extended with a condition ensuring that the branching structure of TS is preserved.

In Example 8.41, ample \(\left(s_{0}\right)=\{\alpha\}\) resulted in a correct reduction. The soundness of this reduction follows from the fact that \(s_{0} \approx \operatorname{div} \alpha\left(s_{0}\right)\). The \(\gamma\) - and \(\beta\)-transitions in state \(s_{0}\) can be mimicked in the reduced transition system by first performing the stutter step \(s_{0} \rightarrow \alpha\left(s_{0}\right)\) followed by \(\gamma\) or \(\beta\), respectively. It turns out that this strategy is generally applicable. Whenever (A1)-(A4) allow for choosing a singleton action set ample \((s)=\{\alpha\}\), then state \(s \approx^{\operatorname{div}} \alpha(s)\). A formal proof of this statement will be provided later. An intuitive explanation is that all other enabled actions in \(s\) (i.e., all actions in \(\operatorname{Act}(s) \backslash\{\alpha\})\) are enabled in \(\alpha(s)\) and lead to states that are equivalent to the corresponding successors of \(\alpha\). This suggests extending conditions (A1) through (A4) by

\section*{(A5) Branching condition}
\[
\text { If } \operatorname{ample}(s) \neq \operatorname{Act}(s), \text { then }|\operatorname{ample}(s)|=1
\]

\section*{Example 8.42.}

Consider again the transition system TS in Figure 8.19 (left). The reduction shown in Figure 8.20 fulfills conditions (A1) through (A5). The nonemptiness condition (A1) and the branching condition (A5) are obviously fulfilled, since ample \(\left(s_{0}\right)=\{\alpha\}\) is a singleton and all other states are fully expanded. The stutter condition (A3) holds since \(\alpha\) is a stutter action. The cycle condition (A4) is obvious since all states on a cycle (the states equipped with a self-loop) are fully expanded. Finally, the dependency condition (A2) is satisfied since \(\alpha\) is independent of \(\{\beta, \gamma\}\).


Figure 8.21: Partial order reduction that violates (A3).
Now consider the alternative selection of ample sets: \(\operatorname{ample}\left(s_{0}\right)=\{\alpha\}\) and \(\operatorname{ample}\left(\alpha\left(s_{0}\right)\right)=\) \(\{\delta\}\), while all other states are fully expanded. These ample sets yield the reduced system \(\hat{T S}\) shown in Figure 8.21. Observe that TS \(\not \approx\) div \(\hat{T S}\) since \(\hat{T S}\) does not contain a state that is equivalent to \(\beta\left(s_{0}\right)\) - there is no state labeled with \(\varnothing\) that can reach a \(c\)-state but

\section*{(A1) Nonemptiness condition}
\[
\varnothing \neq \operatorname{ample}(s) \subseteq \operatorname{Act}(s)
\]

\section*{(A2) Dependency condition}

Let \(s \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{n}} s_{n} \xrightarrow{\alpha} t\) be a finite execution fragment in TS. If \(\alpha\) depends on \(\operatorname{ample}(s)\), then \(\beta_{i} \in \operatorname{ample}(s)\) for some \(0<i \leqslant n\).

\section*{(A3) Stutter condition}

If \(\operatorname{ample}(s) \neq \operatorname{Act}(s)\), then any \(\alpha \in \operatorname{ample}(s)\) is a stutter action.

\section*{(A4) Cycle condition}

For any cycle \(s_{0} s_{1} \ldots s_{n}\) in \(\hat{T S}\) and \(\alpha \in \operatorname{Act}\left(s_{i}\right)\), for some \(0<i \leqslant n\), there exists \(j \in\{1, \ldots, n\}\) such that \(\alpha \in \operatorname{ample}\left(s_{j}\right)\).

\section*{(A5) Branching condition}

If \(\operatorname{ample}(s) \neq \operatorname{Act}(s)\), then \(|\operatorname{ample}(s)|=1\).

Figure 8.22: Requirements on the ample set of state \(s\) for \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\).
not a \(b\)-state. The same applies to \(\gamma\left(s_{0}\right)\). In fact, \(T S\) and \(\hat{T} S\) can be distinguished by the \(\mathrm{CTL}_{\backslash \bigcirc}\) formula
\[
\Phi=\forall \diamond(a \wedge \exists \diamond b \wedge \exists \diamond c)
\]
which holds for \(\hat{T S}\), but not for \(T S\). In fact, stutter condition (A3) is violated since the chosen action \(\delta\) in state \(\alpha\left(s_{0}\right)\) is not a stutter action.

The ample set conditions are summarized in Figure 8.22. The remainder of this section is concerned with the proof of the correctness theorem:

\section*{Theorem 8.43. Correctness of the Ample Set Approach for CTL \({ }_{\backslash \bigcirc}^{*}\)}

For action-deterministic, finite transition system TS without terminal states:
If conditions (A1) through (A5) are satisfied, then \(T S \approx\) div \(\hat{T S}\).

Since \(\approx{ }^{d i v}\) coincides with \(\mathrm{CTL}_{\stackrel{\text { }}{*}}\) equivalence (see Theorem 7.128, page 561), TS and \(\hat{T S}\) are \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\)-equivalent, provided that (A1)-(A5) hold.

For the remainder of this section, let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be an action-deterministic, finite transition system without terminal states and assume that the ample sets satisfy (A1)-(A5). As before, \(\hat{T S}\) denotes the reduced transition system that arises from the reachable fragment of \(T S\) under the transition relation \(\Longrightarrow\) given by
\[
\frac{s \xrightarrow{\alpha} s^{\prime} \wedge \alpha \in \operatorname{ample}(s)}{s \stackrel{\alpha}{\Longrightarrow} s^{\prime}}
\]

The initial states of \(\hat{T S}\) are the initial states in \(T S . \hat{S}\) denotes the state space of \(\hat{T S}\), i.e., all states in \(S\) that are reachable (via \(\Rightarrow\) ) from some initial state \(s_{0} \in I\).

It will be shown that (A1) through (A5) ensure that there exists a normed bisimulation that relates \(T S\) and \(\hat{T S}\). The concept of normed bisimulations has been introduced in Definition 7.120 (page 552). The correctness theorem then follows from the fact that normed bisimulation is strictly finer than \(\approx d i v\).

Let us briefly recall the main concepts of normed bisimulations. A normed bisimulation for (TS, \(\hat{T} S\) ) is a triple ( \(\mathcal{R}, \nu_{1}, \nu_{2}\) ) where \(\mathcal{R} \subseteq S \times \hat{S}\) and \(\nu_{1}, \nu_{2}: S \times \hat{S} \rightarrow \mathbb{N}\) are functions such that for all pairs \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) the following conditions are fulfilled:
(NI) \(L\left(s_{1}\right)=L\left(s_{2}\right)\), and
(NII) For all \(s_{1}^{\prime} \in \operatorname{Post}\left(s_{1}\right)\), at least one of the following three conditions holds:
(N1) there exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}^{\prime}, s_{2}^{\prime}\right) \in \mathcal{R}\), or
(N2) \(\left(s_{1}^{\prime}, s_{2}\right) \in \mathcal{R}\) and \(\nu_{1}\left(s_{1}^{\prime}, s_{2}\right)<\nu_{1}\left(s_{1}, s_{2}\right)\), or
(N3) there exists \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with \(\left(s_{1}, s_{2}^{\prime}\right) \in \mathcal{R}\) and \(\nu_{2}\left(s_{1}, s_{2}^{\prime}\right)<\nu_{2}\left(s_{1}, s_{2}\right)\),
and the analogous conditions for all \(s_{2}^{\prime} \in \operatorname{Post}\left(s_{2}\right)\) with the "swapped" norm functions \(\nu_{2}^{-}\left(\right.\)instead of \(\nu_{1}\) in (N2)) and \(\nu_{1}^{-}\left(\right.\)instead of \(\nu_{2}\) in (N3)) given by \(\nu_{i}^{-}(u, v)=\) \(\nu_{i}(v, u)\) for all \(u \in \hat{S}\) and \(v \in S\).

In addition, it is required that each initial state of \(T S\) is related by \(\mathcal{R}\) to some initial state of \(\hat{T S}\), and vice versa. Intuitively, \(\nu_{1}(s, \hat{s})\) is an upper bound on the "allowed" number of stutter steps from \(s\) that cannot be mimicked by transitions of \(\hat{s}\), Natural \(\nu_{2}(s, \hat{s})\) serves as a counter for the number of stutter steps that \(\hat{s}\) may perform to reach a state where the visible (nonstutter) steps of \(s\) can be simulated.

To define \(\mathcal{R}\) and the norm functions \(\nu_{1}, \nu_{2}\) we need some further notations.

\section*{Definition 8.44. Forming Path, Relation \(\triangleleft\)}

Let \(T S\) be a finite, action-deterministic transition system and \(s, s^{\prime} \in S\). A forming path from \(s\) to \(s^{\prime}\) is a finite execution fragment \(\varrho\) in \(T S\) of the form \(s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n}\) where
- \(s=s_{0}\) and \(s_{n}=s^{\prime}\),
- \(\beta_{1}, \ldots, \beta_{n}\) are stutter actions, and
- for \(1 \leqslant i<n\), the singleton action set \(\left\{\beta_{i+1}\right\}\) fulfills the dependence condition (A2) for state \(s_{i}\). That is, for any finite execution fragment \(s_{i} \xrightarrow{\alpha_{1}} t_{1} \xrightarrow{\alpha_{2}} \ldots \xrightarrow{\alpha_{m}} t_{m} \xrightarrow{\gamma} \ldots\) where \(\gamma\) is dependent on \(\beta_{i+1}\), there exists \(j \in\{1, \ldots, m\}\) such that \(\alpha_{j}=\beta_{i+1}\).

We write \(s \triangleleft s^{\prime}\) iff there exists a forming path from \(s\) to \(s^{\prime}\).

The following simple properties of forming paths and the relation \(\triangleleft\) will be used quite often in the following argumentation. First, the relation \(\triangleleft\) is transitive and reflexive (even though, in general, nonsymmetric). The reflexivity is clear as each execution fragment of length 0 yields a forming path. The transitivity follows from the fact that forming paths from \(s\) to \(s^{\prime}\) and from \(s^{\prime}\) to \(s^{\prime \prime}\) can be concatenated resulting in a forming path from \(s\) to \(s^{\prime \prime}\). Secondly, if \(s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n}\) is a forming path from \(s=s_{0}\) to \(s^{\prime}=s_{n}\) of length \(n\), then \(s_{i} \triangleleft s_{j}\) for \(0 \leqslant i \leqslant j \leqslant n\).

\section*{Lemma 8.45. Properties of Forming Paths (Part 1)}

Let \(s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} s_{n}\) be a forming path from \(s=s_{0}\) to \(s^{\prime}=s_{n}\) and \(\alpha\) be a stutter action such that \(s \xrightarrow{\alpha} t\) is a transition in TS. Then, we have:
(a) If \(\alpha \notin\left\{\beta_{1}, \ldots, \beta_{n}\right\}\), then \(\alpha\) is independent of \(\left\{\beta_{1}, \ldots, \beta_{n}\right\}\) and there exists a forming path \(t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} t_{n}\) from \(t=t_{0}\) to \(t_{n}\) such that \(s_{i} \xrightarrow{\alpha} t_{i}\) for all indices \(i \in\{1, \ldots, n\}\).
(b) If \(\alpha=\beta_{j}\) and \(\alpha \notin\left\{\beta_{1}, \ldots, \beta_{j-1}\right\}\) for some \(j \in\{1, \ldots, n\}\), then \(t \triangleleft s^{\prime}\) and there exists a forming path from \(t\) to \(s^{\prime}\) with the action sequence \(\beta_{1} \ldots \beta_{j-1} \beta_{j+1} \ldots \beta_{n}\).

Proof: We first prove (a). The independency of \(\alpha\) from \(\left\{\beta_{1}, \ldots, \beta_{n}\right\}\) is immediate from the definition of forming paths that require (A2) to hold on all states of a forming path. Lemma 8.15 (page 613) yields the existence of an execution fragment
\[
t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{n}} t_{n}
\]
from \(t=t_{0}\) to \(t_{n}\) such that \(s_{i} \xrightarrow{\alpha} t_{i}\) for all indices \(i \in\{1, \ldots, n\}\). It remains to show that this execution fragment is a forming path. Obviously, \(\beta_{1}, \ldots, \beta_{n}\) are stutter actions. We now have to show that the dependency condition (A2) holds for all states \(t_{i}\) and the singleton action sets \(\left\{\beta_{i+1}\right\}\). This is obvious since whenever
\[
t_{i} \xrightarrow{\gamma_{1}} v_{1} \xrightarrow{\gamma_{2}} \ldots \xrightarrow{\gamma_{k}} v_{k} \xrightarrow{\gamma} v
\]
is an execution fragment where \(\gamma\) depends on \(\beta_{i}\), then
\[
s_{i} \xrightarrow{\alpha} t_{i} \xrightarrow{\gamma_{1}} v_{1} \xrightarrow{\gamma_{2}} \ldots \xrightarrow{\gamma_{k}} v_{k} \xrightarrow{\gamma} v
\]
is an execution fragment from \(s_{i}\). Since (A2) holds for \(s_{i}\) and the singleton set \(\left\{\beta_{i+1}\right\}\) and \(\gamma\) depends on \(\beta_{i+1}\), one of the actions \(\alpha, \gamma_{1}, \ldots, \gamma_{k}\) agrees with \(\beta_{i+1}\). Since \(\alpha \notin\left\{\beta_{1}, \ldots, \beta_{n}\right\}\), we have \(\gamma_{j}=\beta_{i+1}\) for some \(j \in\{1, \ldots, k\}\). Hence, (A2) holds for \(t_{i}\) and the action set \(\left\{\beta_{i+1}\right\}\).

We now consider (b) and suppose that \(\alpha=\beta_{j}\) and \(\alpha \notin\left\{\beta_{1}, \ldots, \beta_{j-1}\right\}\). Applying (a) to the forming path
\[
s=s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{j-1}} s_{j-1}
\]
yields the existence of a forming path
\[
t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{j-1}} t_{j-1}
\]
from \(t=t_{0}\) to some state \(t_{j-1}\) such that \(s_{i} \xrightarrow{\alpha} t_{i}\) is a stutter step for \(i=1, \ldots, j-1\). Since \(\alpha=\beta_{j}\) we have
\[
t_{j-1}=\alpha\left(s_{j-1}\right)=\beta_{j}\left(s_{j-1}\right)=s_{j}
\]

Thus, by adding the forming path \(s_{j} \xrightarrow{\beta_{j+1}} s_{j+1} \xrightarrow{\beta_{j+2}} \ldots \xrightarrow{\beta_{n}} s_{n}\) from \(s_{j}\) to \(s_{n}=s^{\prime}\) a forming path
\[
t=t_{0} \xrightarrow{\beta_{1}} t_{1} \xrightarrow{\beta_{2}} \ldots \xrightarrow{\beta_{j-1}} t_{j-1}=s_{j} \xrightarrow{\beta_{j+1}} s_{j+1} \xrightarrow{\beta_{j+2}} \ldots \xrightarrow{\beta_{n}} s_{n}=s^{\prime}
\]
from \(t\) to \(s^{\prime}\) with the action sequence \(\beta_{1} \ldots \beta_{j-1} \beta_{j+1} \ldots \beta_{n}\) is obtained.

\section*{Lemma 8.46. Properties of Forming Paths (Part 2)}

Let \(s, s^{\prime}\) be two states in TS such that \(s \triangleleft s^{\prime}\) and let \(\alpha \in \operatorname{Act}(s)\).
(a) If there is a forming path from \(s\) to \(s^{\prime}\) in which \(\alpha\) does not occur, then \(\alpha \in \operatorname{Act}\left(s^{\prime}\right)\) and \(\alpha(s) \triangleleft \alpha\left(s^{\prime}\right)\).
(b) If \(\alpha\) is a stutter action with \(s \xrightarrow{\alpha} t\) and \(\neg\left(t \triangleleft s^{\prime}\right)\), then \(\alpha \in \operatorname{Act}\left(s^{\prime}\right)\) and \(s^{\prime} \xrightarrow{\alpha} t^{\prime}\) where \(t \triangleleft t^{\prime}\).

Proof: The proof for (a) can be provided using induction on the length \(n\) of a forming path from \(s\) to \(s^{\prime}\) where \(\alpha\) does not occur. The basis of induction \(n=0\) is obvious as we then have \(s=s^{\prime}\). In the induction step \(n-1 \Longrightarrow n(n \geqslant 1)\) we assume that
\[
s=s_{0} \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{1}} \ldots \xrightarrow{\beta_{n-1}} s_{n-1} \xrightarrow{\beta_{n}} s_{n}=s^{\prime}
\]
is a forming path from \(s\) to \(s^{\prime}\) such that \(\alpha \notin\left\{\beta_{1}, \ldots, \beta_{n}\right\}\). By induction hypothesis we have \(\alpha \in \operatorname{Act}\left(s_{n-1}\right)\) and
\[
\alpha(s) \triangleleft t
\]
where \(t=\alpha\left(s_{n-1}\right)\). As the dependence condition (A2) holds for state \(s_{n-1}\) and the singleton action set \(\left\{\beta_{n}\right\}\), actions \(\alpha\) and \(\beta_{n}\) are independent. Thus, \(\alpha \in \operatorname{Act}\left(s_{n}\right)\) and \(\beta_{n} \in \operatorname{Act}(t)\) and, with \(u=\beta_{n}(t)\),
\[
\alpha\left(s^{\prime}\right)=\alpha\left(s_{n}\right)=\alpha\left(\beta_{n}\left(s_{n-1}\right)\right)=\beta_{n}\left(\alpha\left(s_{n-1}\right)\right)=\beta_{n}(t)=u
\]

Condition (A2) also holds for \(\alpha(s)=t\) and the singleton action set \(\left\{\beta_{n}\right\}\), since \(\alpha\) and \(\beta_{n}\) are independent and (A2) holds for state \(s_{n-1}\) and the singleton action set \(\left\{\beta_{n}\right\}\). We get
\[
t \triangleleft u .
\]

Hence, \(\alpha(s) \triangleleft t \triangleleft u=\alpha\left(s^{\prime}\right)\) which yields \(\alpha(s) \triangleleft \alpha\left(s^{\prime}\right)\).
The proof for part (b) can be provided with similar arguments, also using induction of the length of a forming path from \(s\) to \(s^{\prime}\). See Exercise 8.15.

Note that part (a) of Lemma 8.46 applies to all actions \(\alpha \in \operatorname{Act}(s)\) which are nonstutter actions, since they cannot occur on a forming path. In addition, there might also be stutter actions \(\alpha\) enabled in \(s\) that do not occur on at least one forming path from \(s\) to the given state \(s^{\prime}\).

\section*{Notation 8.47. Relation \(\mathcal{R}_{f p}\)}

The relation \(\mathcal{R}_{f p}\) is given by \(\mathcal{R}_{f p}=\{(s, \hat{s}) \in S \times \hat{S} \mid s \triangleleft \hat{s}\}\) where, as before, \(S\) is the state space of \(T S\) and \(\hat{S}\) the state space of \(\hat{T S}\).

\section*{Notation 8.48. Forming Path in \(\hat{T S}\)}

A forming path in \(\hat{T} S\) means a forming path \(\hat{s}_{0} \xlongequal{\beta_{1}} \hat{s}_{1} \xlongequal{\beta_{2}} \ldots \xlongequal{\beta_{n}} \hat{s}_{n}\) as in Definition 8.44 consisting of transitions in \(\hat{T S}\). (Thus, \(\hat{s}_{0}, \hat{s}_{1}, \ldots, \hat{s}_{n} \in \hat{S}\) and \(\beta_{i+1} \in \operatorname{ample}\left(\hat{s}_{i}\right)\) for \(0 \leqslant i<n\).)

\section*{Lemma 8.49. Properties of Forming Paths in \(\hat{T} S\)}

Let \(\hat{s}\) be a state in \(\hat{T S}\).
(a) If \(\hat{\varrho}\) is a forming path in \(\hat{T} S\) starting in state \(\hat{s}\) and \((s, \hat{s}) \in \mathcal{R}_{f p}\), then \((s, \hat{u}) \in \mathcal{R}_{f p}\) for all states \(\hat{u}\) in \(\hat{\varrho}\).
(b) There exists a forming path from \(\hat{s}\) in \(\hat{T S}\) to some fully expanded state.

Proof: Part (a) is immediate by the transitivity of \(\triangleleft\). The statement in (b) follows from the fact that any execution fragment
\[
\hat{s}_{0} \stackrel{\beta_{1}}{\Longrightarrow} \hat{s}_{1} \stackrel{\beta_{2}}{\Longrightarrow} \ldots \xlongequal{\beta_{n}} \hat{s}_{n}
\]
in \(\hat{T S}\) where none of the states \(\hat{s}_{i}\) is fully expanded is a forming path in \(\hat{T S}\), because of the dependency condition (A2), the stutter condition (A3), and the branching condition (A5). As \(\hat{T} S\) is finite, the nonemptiness condition (A1) and the cycle condition (A4) ensure the existence of a forming path from \(\hat{s}\) to a fully expanded state.

\section*{Notation 8.50. Length of Shortest Forming Paths}

For \(s, s^{\prime} \in S\) with \(s \triangleleft s^{\prime}\) the length of a shortest forming path from \(s\) to \(s^{\prime}\) is denoted \(\left|s \triangleleft s^{\prime}\right|\). For \(\hat{s} \in \hat{S}\), we define \(\operatorname{dist}(\hat{s})\) as the length of a shortest forming path in \(\hat{T S}\) from \(\hat{s}\) to some fully expanded state.

Note that \(\operatorname{dist}(\hat{s})\) refers to the forming paths in \(\hat{T} S\). Thus, \(\operatorname{dist}(\hat{s})\) might be larger than \(\min _{\hat{v}}|\hat{s} \triangleleft \hat{v}|\) where \(\hat{v}\) ranges over all fully expanded states in \(\hat{T S}\).

\section*{Notation 8.51. Norm Functions \(\nu_{1}, \nu_{2}\)}

The functions \(\nu_{1}, \nu_{2}: S \times \hat{S} \rightarrow \mathbb{N}\) are defined as follows. For all \(s \in S\) and \(\hat{s} \in \hat{S}\) we put
\[
\nu_{1}(s, \hat{s})= \begin{cases}|s \triangleleft \hat{s}| & : \text { if }(s, \hat{s}) \in \mathcal{R}_{f p} \\ 0 & : \\ \text { otherwise }\end{cases}
\]
and \(\nu_{2}(s, \hat{s})=\operatorname{dist}(\hat{s})\).

Note that part (b) of Lemma 8.49 ensures that every state \(\hat{s} \in \hat{S}\) has a forming path to some fully expanded state. Thus, \(\nu_{2}\) is well-defined.

\section*{Lemma 8.52. Normed Bisimulation Equivalence of TS and \(\hat{T} S\)}
\(\left(\mathcal{R}_{f p}, \nu_{1}, \nu_{2}\right)\) is a normed bisimulation for (TS, \(\left.\hat{T} S\right)\).

Proof: We have \(\left(s_{0}, s_{0}\right) \in \mathcal{R}_{f p}\) for all initial states \(s_{0} \in I\), since we may consider a forming path of length 0 .

We show that for any pair \((s, \hat{s}) \in \mathcal{R}_{f p}\) conditions (NI) and (NII) hold for \((s, \hat{s})\) and for \((\hat{s}, s)\) when \(\nu_{1}\) and \(\nu_{2}\) and their arguments are exchanged. The labeling condition (NI) is obvious as all actions on a forming path are stutter actions. Thus, all states on a forming path have the same labeling.

We now consider a pair \((s, \hat{s}) \in \mathcal{R}_{f p}\) and an action \(\alpha \in \operatorname{Act}(s)\).
Case 1. \(\alpha\) does not occur on some forming path from sto \(\hat{s}\).
Then, \(\alpha \in \operatorname{Act}(\hat{s})\) and \(\alpha(s) \triangleleft \alpha(\hat{s})\) by part (a) of Lemma 8.46. Hence, if \(\alpha \in \operatorname{ample}(\hat{s})\), then case (N1) applies. If \(\alpha \notin \operatorname{ample}(\hat{s})\), then we choose the first action \(\beta\) of a shortest forming path in \(\hat{T S}\) from \(\hat{s}\) to some fully expanded state (see part (b) of Lemma 8.49 on page 658). Then,
\[
\hat{s} \stackrel{\beta}{\Longrightarrow} \beta(\hat{s})
\]
is a forming path in \(\hat{T} S\) of length 1 . Moreover, we have \((s, \beta(\hat{s})) \in \mathcal{R}_{f p}\) and
\[
\nu_{2}(s, \beta(\hat{s}))=\operatorname{dist}(\beta(\hat{s}))=\operatorname{dist}(\hat{s})-1=\nu_{2}(s, \hat{s})-1<\nu_{2}(s, \hat{s}) .
\]

Hence, case (N3) applies.
Case 2. \(\alpha\) occurs in some shortest forming path from sto \(\hat{s}\).
Since \(\alpha\) occurs is some forming path, \(\alpha\) is a stutter action. Let
\[
s_{0} \xrightarrow{\beta_{1}} s_{1} \ldots \xrightarrow{\beta_{j-1}} s_{j-1} \xrightarrow{\alpha} s_{j} \xrightarrow{\beta_{j+1}} \ldots \xrightarrow{\beta_{n}} s_{n}
\]
be a shortest forming path from \(s=s_{0}\) to \(\hat{s}=s_{n}\) with \(\alpha \notin\left\{\beta_{1}, \ldots, \beta_{j-1}\right\}\). By part (b) of Lemma 8.45 on page 655 there exists a forming path from \(t=\alpha(s)\) to \(\hat{s}\) with the action sequence \(\beta_{1} \ldots \beta_{j-1} \beta_{j+1} \ldots \beta_{n}\). Thus, \((t, \hat{s}) \in \mathcal{R}_{f p}\) and
\[
\nu_{1}(t, \hat{s})=|t \triangleleft \hat{s}| \leqslant n-1<n=|s \triangleleft \hat{s}|=\nu_{1}(s, \hat{s}) .
\]

Hence, case (N2) applies for the transition \(s \xrightarrow{\alpha} t\).
It remains to show that (NII) holds for \((\hat{s}, s) \in \mathcal{R}^{-1}\) with exchanged roles of \(\nu_{1}\) and \(\nu_{2}\). Let \(\alpha \in \operatorname{ample}(s)\). If \(s=\hat{s}\), then case (N1) applies. If \(s \neq \hat{s}\) then we take the first action
\(\beta\) of a shortest forming path from \(s\) to \(\hat{s}\). We then have \((\hat{s}, \beta(s)) \in \mathcal{R}_{f p}^{-1}\) and
\[
\nu_{1}(\beta(s), \hat{s})=|\beta(s) \triangleleft \hat{s}|=|s \triangleleft \hat{s}|-1=\nu_{1}(s, \hat{s})-1<\nu_{1}(s, \hat{s}) .
\]

This yields condition (N3).

\section*{Example 8.53. Normed Bisimulation Equivalence of TS and \(\hat{T S}\)}

We regard again the transition system \(T S\) in Figure 8.19 (page 650) and the ample sets \(\operatorname{ample}\left(s_{0}\right)=\{\alpha\}\), while all other states are fully expanded. This yields the reduced system shown in Figure 8.20 (page 651). We saw before that conditions (A1)-(A5) are fulfilled. The goal is now to provide a normed bisimulation for (TS, \(\hat{T S}\) ). Besides the trivial forming paths of length 0 , we have forming paths \(s_{0} \stackrel{\alpha}{\Longrightarrow} s_{1}\) in \(\hat{T} S\), and forming paths \(s_{\beta} \xrightarrow{\alpha} t_{\beta}\) and \(s_{\gamma} \xrightarrow{\alpha} t_{\gamma}\) in TS. Thus, according to Notation 8.47 we deal with the relation
\[
\mathcal{R}_{f p}=\operatorname{id} \cup\left\{\left(s_{0}, s_{1}\right),\left(s_{\gamma}, t_{\gamma}\right),\left(s_{\beta}, t_{\beta}\right)\right\}
\]
where id denotes the set of all pairs \((\hat{s}, \hat{s})\) with \(\hat{s}\) a state in \(\hat{T} S\). The norm functions \(\nu_{1}\) and \(\nu_{2}\) are defined as follows. For the pairs \((\hat{s}, \hat{s}) \in \operatorname{id}\) we have \(\nu_{1}(\hat{s}, \hat{s})=|\hat{s} \triangleleft \hat{s}|=0\). Moreover,
\[
\nu_{1}\left(s_{\gamma}, t_{\gamma}\right)=\nu_{1}\left(s_{\beta}, t_{\beta}\right)=\nu_{1}\left(s_{0}, s_{1}\right)=1
\]
and
\[
\nu_{2}\left(s_{0}, s_{1}\right)=\nu_{2}\left(s_{0}, s_{0}\right)=\operatorname{dist}\left(s_{0}\right)=1,
\]
while \(\nu_{i}(\cdot)=0\) in all remaining cases. Let us check that \(\left(\mathcal{R}_{f p}, \nu_{1}, \nu_{2}\right)\) is a normed bisimulation. Condition (NI) is obviously fulfilled. Moreover, (NII) is clear for the pairs \((\hat{s}, \hat{s}) \in \mathcal{R}_{f p}\). Let us regard the pair \(\left(s_{0}, s_{1}\right) \in \mathcal{R}_{f p}\). For the transitions \(s_{\gamma} \xrightarrow{\gamma} t_{\gamma}\) in \(T S\), we take the transition \(s_{0} \xrightarrow{\alpha} s_{1}\) in \(\hat{T S}\) where (N3) applies as we have \(\nu_{2}\left(s_{0}, s_{1}\right)=0<1=\nu_{2}\left(s_{0}, s_{0}\right)\). An analogy holds for \(s_{\beta} \xrightarrow{\beta} t_{\beta}\). For the transition \(s_{0} \xrightarrow{\alpha} s_{1}\) in \(T S\) the conditions of (N1) are fulfilled since \(s_{0} \xrightarrow{\alpha} s_{1}\) in \(\hat{T S}\) and \(\left(s_{1}, s_{1}\right) \in \mathcal{R}_{f p}\). Similarly, the transition \(s_{0} \xrightarrow{\alpha} s_{1}\) in \(\hat{T} S\) is matched by \(s_{0} \xrightarrow{\alpha} s_{1}\) in \(T S\) according to (N1). For the others pairs in \(\mathcal{R}_{f p}\), condition (NII) can be checked with similar arguments.

Lemma 8.52, together with Lemma 7.123 on page 555, yields that \(T S\) and \(\hat{T S}\) are equivalent under stutter bisimulation equivalence with divergence. This completes the proof for Theorem 8.43.

Clearly, the DFS-based technique explained in Sections 8.2.2 and 8.2.3 to generate the reduced system is also applicable to ensure conditions (A1)-(A5). The additional requirement (A5) is local and simply amounts checking whether the candidate action set \(A c t_{i}(s)\) for ample(s) of the chosen process \(\mathcal{P}_{i}\) is a singleton set. The static partial order approach is also applicable, but requires extending the obtained ample sets for all states where \(|\operatorname{ample}(s)| \geqslant 2\).

\subsection*{8.4 Summary}
- Partial order reduction attempts to analyze only a fragment \(\hat{T} S\) of the full transition system TS by ignoring several interleavings of independent actions.
- Swapping independent actions in an execution fragment yields an execution fragment that starts and ends in the same state.
- The ample set method relies on choosing ample \((s) \subseteq \operatorname{Act}(s)\) in state \(s\). If the ample sets satisfy two local conditions (the nonemptiness condition (A1) and the stutter condition (A3)), a global condition for TS (the dependency condition (A2)), and for \(\hat{T S}\) (the cycle condition (A4)), then \(T S\) and \(\hat{T S}\) are stutter trace equivalent, and thus, \(\mathrm{LTL}_{\backslash \bigcirc \text {-equivalent. }}\)
- Conditions (A1) and (A2) ensure that any execution \(\rho\) in \(T S\) can be turned into an execution \(\rho^{\prime}\) in \(\hat{T S}\) by successive permutations of independent actions and adding independent stutter actions. The stutter and the cycle conditions (A3) and (A4) ensure that \(\rho\) is stutter-equivalent to \(\rho^{\prime}\).
- In on-the-fly partial order reduction, the reduced transition system \(\hat{T} S\) is generated during (nested) depth-first search. To ensure the cycle condition (A4), states that yield a backward edge to a state on the stack are fully expanded.
- Checking the dependency condition (A2) has the same worst case time complexity as solving a reachability problem. Instead of ensuring (A2), some stronger conditions are imposed that can easily be checked by a static analysis and are stronger than (A2).
- Static partial order reduction generates the reduced transition system prior to the verification. To establish the cycle condition (A4), the static approach relies on determining a set of sticky actions. If the set of sticky actions fulfills the visibility condition (S1) and the cycle-breaking condition (S2), the reduction yields a stutterequivalent transition system.
- If the ample sets satisfy conditions (A1) through (A4) and the branching condition (A5), then \(T S\) and \(\hat{T S}\) are divergent-sensitive stutter-bisimilar, and thus, \(\mathrm{CTL}_{\mathrm{O}^{*} \text { - }}\) equivalent.

\subsection*{8.5 Bibliographic Notes}

Independence of actions. Partial order reduction has been inspired by earlier work on the commutativity of concurrent activities by Lipton [276] and Mazurkiewicz [287]. They
considered action sequences that are equivalent up to permuting independent actions. Equivalence classes of action sequences are called Mazurkiewicz traces. (Exercise 8.5 deals with a variant of Mazurkiewcz traces as in [253, 357, 125].) Other partial order models for concurrent systems that have been influential on partial order reduction are, among others, pomsets [343], partial orders [258], branching processes of Petri nets [144], and event structures [423]. The notion of independent actions (see Definition 8.3) has been introduced by Katz and Peled [234]. Deductive proof techniques for concurrent systems that are based on a partial order view have been proposed in the early eigthies by Apt, Francez, and de Roever [16], and Elrad and Francez [137]. These techniques have been extended and refined by several others; see e.g., [339, 225, 235, 380].

Partial order reduction. The concepts of partial order reduction for the algorithmic verification of concurrent asynchronous systems have been developed independently by Godefroid, Peled, and Valmari in the beginning of the nineties. This chapter is based on the ample set approach by Peled for \(\operatorname{LTL}_{\backslash \bigcirc}[324,211,325]\). Ample sets are similar to Godefroid's persistent sets [168, 171, 169] and Valmari's stubborn sets [398, 399, 400]. The ample set approach for branching-time properties presented in Section 8.3 originates from Gerth, Kuiper, Peled, and Penczek [165, 326]. Alternative partial order reduction approaches for branching-time properties have been provided by Willems and Wolper [422], and Ramakrishna and Smolka [352]. Alternative criteria for preserving linear and branching-time properties as well as sound criteria for the universal fragment of \(\mathrm{CTL}_{\backslash \bigcirc}^{*}\) have been presented by Penczek, Gerth, Kuiper, and Szreter [330] (see also Exercises 8.16-8.18). Partial order reduction techniques for equivalence checking under various implementation relations have been discussed, e.g., in [401, 218].

Computing the reduced model. A detailed discussion of nested depth-first search in combination with partial order reduction is given by Holzmann, Peled, and Yannakakis [212, 325]. For further reading on partial order reduction and its implementation in the model checker SPIN we refer to [169], [92, Chapter 10], and the monograph by Holzmann on SPIN [209]. In the last years, several variants have been discussed, see e.g. [169, 402, 310, 413, 135, \(64,328]\). Static partial order reduction (see Section 8.2.4) has been developed by Kurshan, Levin, Minea, Peled and Yenigün [251]. Other applications of partial order reduction include: integration with symbolic techniques (by Abdulla, Jonsson, Kindahl and Peled [3]), infinite-state systems (by Alur, Brayton, Henzinger, Qadeer and Rajamani [8]), and breadth-first search strategies (by Bosnacki and Holzmann [59]).

Related approaches. The concept of sleep sets \([168,172,169]\) is orthogonal to the ample, persistent, or stubborn sets. Sleep sets are aimed at decreasing the number of transitions rather than the number of states and are based on the (dynamic) knowledge of the graph structure that is obtained during the depth-first search traversal of the (possibly reduced) transition system. The concept of \(\tau\)-confluence, proposed by Groote and van de Pol [177],
is aimed at symbolic state-space reduction and obtains branching bisimilar transition systems. Rather than perform a partial order reduction of an interleaving representation of a concurrent system, one can attempt to directly perform model checking on partial orders. McMillan [289] has presented an algorithm to obtain an initial part of the (infinite) branching process of a 1-safe Petri net. The so-called complete finite prefix of the branching process contains all information on reachable states and transitions, and can be used as the basis for model-checking algorithms, see, e.g., [145, 416].

\subsection*{8.6 Exercises}

Exercise 8.1. Let \(T S\) be the transition system depicted in Figure8.23 with the action set Act \(=\{\alpha, \beta, \gamma, \delta, \tau\}\). Determine the pairs of independent actions in TS.


Figure 8.23: Transition system TS for Exercise 8.1.

EXERCISE 8.2. Let \(T S\) be a transition system and let \(\mathcal{I}\) be the set of action pairs \((\alpha, \beta)\) such that \(\alpha\) and \(\beta\) are independent of TS. Is \(\mathcal{I}\) transitive? reflexive? symmetric?

EXercise 8.3. Consider the transition system \(T S\left(P G_{1} \mid \| P G_{2}\right)\) for the program graphs \(P G_{1}\) (left) and \(P G_{2}\) (right) in Figure 8.24. Determine the pairs of independent actions.

EXERCISE 8.4. Consider the transition system \(T S_{\text {Sem }}\) for mutual exclusion with a semaphore (see Figure 2.8 on page 45). We deal with the action set \(A c t=\left\{\right.\) request \(_{i}\), rel \(_{i}\), enter \(\left.{ }_{i}: i=1,2\right\}\) and the proposition set \(A P=\left\{\right.\) crit \(_{1}\), crit \(\left._{2}\right\}\).


Figure 8.24: Program graphs \(P G_{1}\) (left) and \(P G_{2}\) (right) for Exercise 8.3.
(a) Consider the finite execution fragment \(\varrho\)
\[
\begin{aligned}
& \left\langle\text { noncrit }_{1}, \text { wait }_{2}, y=1\right\rangle
\end{aligned}
\]

Verify the statement of Lemma 8.10 (page 604) by investigating the three action sequences

> request \(_{1}\) enter \(_{1}\) request \(_{2}\) rel \(_{1}\), request \(_{1}\) request \(_{2}\) enter \(_{1}\) rel \(_{1}\), request \(_{2}\) request \(_{1}\) enter \(_{1}\) rel \(_{1}\),
and checking that they lead from \(s=\left\langle\right.\) noncrit \(_{1}\), noncrit \(\left._{2}, y=1\right\rangle\) to state \(t=\left\langle\right.\) noncrit \(_{1}\), wait \(_{2}, y=\) 1) and are stutter-equivalent to \(\varrho\).
(b) Consider the infinite execution \(\rho\)
\[
\left.\begin{array}{lll}
\left\langle\text { noncrit }_{1}, \text { noncrit }_{2}, y=1\right\rangle & \xrightarrow{\text { request }_{1}} \\
\left\langle\text { wait }_{1}, \text { noncrit }_{2}, y=1\right\rangle & \xrightarrow{\text { enter }_{1}} \\
\left\langle\text { crit }_{1}, \text { noncrit }_{2}, y=0\right\rangle & \xrightarrow{\text { rel }_{1}} & \\
\left\langle\text { noncrit }_{1},\right. \text { noncrit }
\end{array}, y=1\right\rangle \xrightarrow{\text { request }_{1}} \ldots \ldots .
\]
where process \(\mathcal{P}_{1}\) continuously passes through its three phases, while \(\mathcal{P}_{2}\) does nothing. Verify Lemma 8.11 (page 605) by investigating the action sequence request \({ }_{2}\left(\text { request }_{1} \text { enter }{ }_{1} \text { rel }_{1}\right)^{\omega}\) and checking that it yields an infinite execution that is stuttering equivalent to \(\rho\).

Exercise 8.5. Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be an action-deterministic transition system. and let \(\mathcal{I}_{s t}\) be the set of all pairs \((\alpha, \beta) \in\) Act \(\times\) Act of independent actions \(\alpha\) and \(\beta\) where \(\alpha\) or \(\beta\) (or
both) is a stutter action. Let stutter permutation equivalence \(\leftrightharpoons\) perm be the coarsest equivalence on Act* such that
\[
\bar{\gamma} \alpha \beta \bar{\delta} \leftrightharpoons{ }_{\text {perm }} \bar{\gamma} \beta \alpha \bar{\delta}
\]
if \(\bar{\gamma}, \bar{\delta} \in A c t^{*}\) and \((\alpha, \beta) \in \mathcal{I}_{s t}\).
(a) Let \(\varrho\) and \(\varrho^{\prime}\) be two finite execution fragments in \(T S\) that start in the same state and that rely on stutter permutation equivalent action sequences. Show that \(\varrho \triangleq \varrho^{\prime}\).
(b) Let \(\varrho=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} \ldots \xrightarrow{\alpha_{k}} s_{k}\) be a finite execution fragment in \(T S\) and let \(\beta_{1} \ldots \beta_{k} \in\) Act* \({ }^{*}\) such that \(\alpha_{1} \ldots \alpha_{k} \leftrightharpoons_{\text {perm }} \beta_{1} \ldots \beta_{k}\). Show that there exists a a finite execution fragment in \(T S\) which starts in \(s_{0}\) and relies on the action sequence \(\beta_{1} \ldots \beta_{k}\).

The extension of \(\leftrightharpoons_{\text {perm }}\) to an equivalence for infinite action sequences is defined as follows. If \(\widetilde{\alpha}=\alpha_{1} \alpha_{2} \alpha_{3} \ldots\) and \(\widetilde{\beta}=\beta_{1} \beta_{2} \beta_{3} \ldots\) are action sequences in Act \({ }^{\omega}\) then \(\widetilde{\alpha} \unlhd_{\text {perm }} \widetilde{\beta}\) if for all finite prefixes \(\alpha_{1} \ldots \alpha_{n}\) of \(\widetilde{\alpha}\) there exists a finite prefix \(\beta_{1} \ldots \beta_{m}\) of \(\widetilde{\beta}\) with \(m \geqslant n\) and a finite word \(\bar{\gamma} \in A c t^{*}\) such that
\[
\alpha_{1} \ldots \alpha_{n} \bar{\gamma} \leftrightharpoons \text { perm } \quad \beta_{1} \ldots \beta_{m} .
\]

We then define the binary relation \(\leftrightharpoons{ }_{\text {perm }}{ }^{\omega}\) on Act \(^{\omega}\) by
\[
\widetilde{\alpha} \leftrightharpoons_{\text {perm }} \omega \widetilde{\beta} \quad \text { iff } \quad \widetilde{\alpha} \unlhd_{\text {perm }} \widetilde{\beta} \text { and } \widetilde{\beta} \unlhd_{\text {perm }} \widetilde{\alpha} .
\]
(a) Show that \(\leftrightharpoons_{\text {perm }}{ }^{\omega}\) is an equivalence.
(b) Let \(\rho\) and \(\rho^{\prime}\) be two infinite execution fragments in \(T S\) starting in the same state \(s\) with the action sequences \(\widetilde{\alpha}\) and \(\widetilde{\beta}\), respectively. Show that if \(\widetilde{\alpha} \leftrightharpoons{ }_{\text {perm }}{ }^{\omega} \widetilde{\beta}\), then \(\rho \triangleq \rho^{\prime}\).
(c) Let \(\rho=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} \ldots\) be a infinite execution fragment in \(T S\) with the action sequence \(\widetilde{\alpha}=\alpha_{1} \alpha_{2} \ldots \in A c t^{\omega}\) and let \(\widetilde{\beta}=\beta_{1} \beta_{2} \ldots \in \operatorname{Act}{ }^{\omega}\) such that \(\widetilde{\alpha} \leftrightharpoons{ }_{\text {perm }}{ }^{\omega} \widetilde{\beta}\). Show that there exists an infinite execution fragment in \(T S\) which starts in \(s_{0}\) and relies on the action sequence \(\widetilde{\beta}\).
(d) Let \(\widetilde{\alpha}=\alpha_{1} \alpha_{2} \alpha_{3} \ldots \in \operatorname{Act}{ }^{\omega}\) and let \(\left(\widetilde{\alpha}_{i}\right)_{i \geqslant 1}\) be a sequence of infinite action sequences of the form
\[
\widetilde{\alpha}_{i}=\underbrace{\alpha_{1} \ldots \alpha_{i}}_{\text {as in } \widetilde{\alpha}} \widetilde{\beta}_{i}
\]
where \(\widetilde{\beta}_{i}\) is an element of \(A c t^{\omega}\) such that
\[
\widetilde{\alpha}_{1} \leftrightharpoons_{\text {perm }}{ }^{\omega} \widetilde{\alpha}_{2} \leftrightharpoons_{\text {perm }}{ }^{\omega} \widetilde{\alpha}_{3} \leftrightharpoons_{\text {perm }}{ }^{\omega} \ldots
\]

Can we conclude that \(\widetilde{\alpha}_{i} \leftrightharpoons_{\text {perm }}{ }^{\omega} \widetilde{\alpha}\) for all \(i \geqslant 1\) ? Provide a proof or a counterexample.

\section*{Exercise 8.6.}


The state labeling is as follows:
- \(L\left(s_{10}\right)=\varnothing\)
- \(L\left(s_{6}\right)=L\left(s_{7}\right)=\{a\}\)
- \(L\left(s_{3}\right)=L\left(s_{4}\right)=L\left(s_{5}\right)=L\left(s_{8}\right)=L\left(s_{9}\right)=\{b\}\)
- \(L\left(s_{1}\right)=L\left(s_{2}\right)=\{a, b\}\)

Indicate for each of the following ample sets whether they satisfy the requirements (A1) through (A3). Also check whether the requirement (A4) holds
- ample \(\left(s_{6}\right)=\{\gamma, \alpha\}\)
- \(\operatorname{ample}\left(s_{7}\right)=\{\beta\}\)
- \(\operatorname{ample}\left(s_{8}\right)=\{\alpha\}\)
- ample \(\left(s_{9}\right)=\{\alpha, \beta, \delta\}\)
- \(\operatorname{ample}\left(s_{10}\right)=\{\gamma, \eta\}\)

If the conditions (A1) through (A4) do not hold, provide a minimal extension of the ample sets to fix it. Clarify your adaptations.

Exercise 8.7. Consider the transition system \(T S_{P e t}\) for the Peterson mutual exclusion algorithm (see page 45).
(a) Which actions are independent?
(b) Apply the partial order reduction approach to \(T S_{P e t}\) with small ample sets according to
(i) Algorithm 38 (page 622) for checking the invariant \(\square \neg\left(\right.\) crit \(_{1} \wedge\) crit \(\left._{2}\right)\). Take \(A P=\) \(\left\{\right.\) crit \(_{1}\), crit \(\left._{2}\right\}\).
(ii) Algorithm 39 (page 625) for checking the liveness property \(\square \diamond\) crit \(_{1}\). Take \(A P=\) \{crit \(\left.{ }_{1}\right\}\).

Exercise 8.8. Consider the transition system \(T S=B C R\|B P\|\) Printer for the booking system considered in Example 2.29 on page 50.
(a) Which actions are independent?
(b) Apply the partial order reduction approach to \(T S\) with minimal ample sets according to Algorithm 38 (page 622) for checking the liveness property \(\varphi=\square\rangle\) "printer in location 1 " where the printer's locations serve as atomic propositions.


Figure 8.25: Transition system \(T S\) (left) and \(\hat{T S}\) (right) for Exercise 8.9.
EXERCISE 8.9. Figure 8.25 shows on its left a transition system \(T S\) and on its right a reduced system \(\hat{T} S\) that results from choosing ample \((s)=\{\alpha\}\). Check whether \(T S\) and \(\hat{T} S\) are stutter trace equivalent. If they are not, indicate which of the conditions (A1)-(A4) is (are) violated.

Answer the same question for the transition system in the reduction shown in Figures 8.26 and 8.27 , where different colors indicate different state labels.

Exercise 8.10. Consider the transition system TS depicted below. Show that conditions (A1)(A4) do not allow for any state reduction, although there is a smaller subsystem \(\hat{T} S\) that is stutter trace equivalent to \(T S\).


Figure 8.26: Transition system \(T S\) (left) and \(\hat{T S}\) (right) for Exercise 8.9.


Figure 8.27: Transition system \(T S\) (left) and \(\hat{T S}\) (right) for Exercise 8.9.


Exercise 8.11. Let \(T S_{i}=\left(S_{i}, \operatorname{Act}_{i}, \rightarrow_{i}, I_{i}, A P, L_{i}\right), i=1, \ldots, n\), be action-deterministic transition systems such that \(A c t_{i} \cap A c t_{j} \cap A c t_{k}=\varnothing\) if \(1 \leqslant i<j<k \leqslant n\). We consider the parallel composition with synchronization over common actions (see page 49), i.e., the transition system
\[
T S=T S_{1}\left\|T S_{2}\right\| \ldots \| T S_{n}
\]

For each state \(s=\left\langle s_{1}, \ldots, s_{n}\right\rangle\) of \(T S\), let \(\operatorname{Act}_{i}(s)=\operatorname{Act}_{i} \cap \operatorname{Act}(s)\) be the set of actions of \(T S_{i}\) that
are enabled in \(s\).

Show that the dependency condition (A2) holds if for each state \(s\) of \(T S\) the following conditions (i) and (ii) hold:
(i) If \(\operatorname{ample}(s) \neq \operatorname{Act}(s)\), then \(\operatorname{ample}(s)=\operatorname{Act}_{i}(s)\) for some \(i \in\{1, \ldots, n\}\).
(ii) If \(\operatorname{ample}(s)=\operatorname{Act}_{i}(s) \neq \operatorname{Act}(s)\), then \(\operatorname{ample}(s) \cap\left(\underset{\substack{1 \leqslant j \leqslant n \\ j \neq i}}{\bigcup} A c t_{j}\right)=\varnothing\).


Figure 8.28: Program graphs \(P G_{1}, P G_{2}, P G_{3}\) for Exercise 8.12.

EXERCISE 8.12. Apply the static partial order reduction approach to generate the action set \(A_{\text {sticky }}\) by means of Algorithm 42 (page 648 ) and the modified program graphs \(\hat{P G_{1}}, \hat{P G_{2}}\), and \(\hat{P G_{3}}\) for the three program graphs shown in Figure 8.28.

EXERCISE 8.13. Explain which modifications of the presented static partial order reduction approach are necessary to treat program graphs with channel-based message passing.

Exercise 8.14. Consider the transition system TS shown on the left of Figure 8.29 where the three black states are labeled by \(\{a\}\), while the white states are labeled by \(\varnothing\). Let ample \((\cdot)\) be the ample sets that yield the reduced system \(\hat{T} S\) shown on the right of Figure 8.29.
(a) Show that (A1)-(A5) are fulfilled.
(b) Provide a normed bisimulation for (TS, \(\hat{T S}\) ).


Figure 8.29: Transition system \(T S\) (left) and \(\hat{T S}\) (right) for Exercise 8.14.

Exercise 8.15. Provide the proof for part (b) of Lemma 8.46 on page 656 .

Exercise 8.16. Let \(T S=(S, A c t, \rightarrow, I, A P, L)\) be a finite, action-deterministic transition system without terminal states and let \(\mathcal{I}\) be a binary relation on Act \(\times\) Act such that \((\alpha, \beta) \notin \mathcal{I}\) if \(\alpha\) and \(\beta\) are dependent actions. Let Vis be the set of visible actions, i.e., Vis is the set of all actions \(\alpha \in\) Act that are not stutter actions. Furthermore, let ample sets ample(s) \(\subseteq \operatorname{Act}(s)\) be given such that the following conditions (A2 1 ), (A6), (A7) and (A8) hold for all states \(s \in S\) :
\(\left.\mathbf{( A 2} \boldsymbol{I}_{I}\right)\) If \(s \xrightarrow{\beta_{1}} s_{1} \xrightarrow{\beta_{2}} s_{2} \xrightarrow{\beta_{3}} \ldots \xrightarrow{\beta_{m}} s_{m} \xrightarrow{\gamma} t\) is a finite execution fragment in \(T S\) such that \(\gamma \notin \operatorname{ample}(s)\) and \((\alpha, \gamma) \notin \mathcal{I}\) for all \(\alpha \in \operatorname{ample}(s)\). then there exists an index \(n \in\{1, \ldots, m\}\) such that \(\beta_{n} \in \operatorname{ample}(s)\).
(A6) \((\) Vis \(\times\) Vis) \(\cap \mathcal{I}=\varnothing\).
(A7) If \(s\) is not fully expanded, then ample( \(s\) ) contains at least one visible action, i.e., ample \((s) \cap\) Vis \(\neq \varnothing\) and \(\operatorname{ample}(s) \backslash V i s \neq \varnothing\).
(A8) If \(s\) is not fully expanded, then ample(s) contains at least one stutter action, i.e., ample(s) \(\backslash\) Vis \(\neq \varnothing\).

Show that \(T S\) and \(\hat{T S}\) are \(L^{L T L}{ }_{\backslash \text {-equivalent. }}\)

Exercise 8.17. Let \(T S, \mathcal{I}\), and conditions (A2 \(\mathcal{I}\) ), (A6), (A7) be as in Exercise 8.16 and let ample

\[
\left.\begin{array}{rl}
\mathcal{R}=\{(s, \hat{s}) \in S \times \hat{S} \mid \quad & \text { there exists an execution fragment } \\
\qquad \xrightarrow[\alpha_{1}]{\alpha_{1}} s_{1} \xrightarrow[\alpha_{2}]{\alpha_{n}} \ldots \xrightarrow[s]{\alpha_{n}} \\
& \text { such that } \alpha_{1}, \ldots, \alpha_{n} \text { are stutter actions }
\end{array}\right\} .
\]

Show that \(\mathcal{R}\) is a stutter simulation for (TS, \(\hat{T S}\) ), defined as in Exercise 7.29 on page 592.

Exercise 8.18. Let \(T S\), \(\mathcal{I}\), and conditions (A2 \(\mathcal{I}\) ), (A6), (A7) be as in Exercise 8.16. Show that if conditions (A2 \(2_{\mathcal{I}}\) ), (A6), (A7), and the branching condition (A5) hold then \(T S\) and \(\hat{T S}\) are \(\forall C T L *\)-equivalent.
(Hint: Show that the relation \(\mathcal{R}\) as in Exercise 8.17 is a divergence-sensitive stutter simulation as in Exercise 7.29 (page 592) and apply the statement (h) of Exercise 7.29 and Theorem 7.76 (page 517)).

\section*{Chapter 9}

\section*{Timed Automata}

The logics we have encountered so far are interpreted over transition systems that describe how a reactive system may evolve from one state to another. Timing aspects are, however, not covered. That is, indications are given neither about the residence time of a state nor about the possibility of taking a transition within a particular time interval. However, reactive systems such as device drivers, coffee machines, communication protocols, and automatic teller machines, to mention a few, must react in time - they are time-critical. The behavior of time-critical systems is typically subject to rather stringent timing constraints. For a train crossing it is essential that on detecting the approach of a train, the gate is closed within a certain time bound in order to halt car and pedestrian traffic before the train reaches the crossing. For a radiation machine the time period during which a cancer patient is subjected to a high dose of radiation is extremely important; a small extension of this period is dangerous and can cause the patient's death.

To put it in a nutshell:
Correctness in time-critical systems not only depends on the logical result of the computation but also on the time at which the results are produced.

As timeliness is of vital importance to reactive systems, it is essential that the timing constraints of the system are guaranteed to be met. Checking whether timing constraints are met is the subject of this chapter. In order to express such timing constraints, the strategy will be to extend logical formalisms that allow expression of the ordering of events, with a notion of quantitative time. Such extensions allow expression of timing constraints such as:
"The traffic light will turn green within the next 30 seconds."

A first choice to be made is the time domain: is it discrete or continuous? A discrete time domain is conceptually simple. Transition systems are used to model timed systems where each action is assumed to last for a single time unit. More general delays can be modeled by using a dedicated unobservable action, \(\tau\) (for tick), say. The fact that action \(\alpha\) lasts \(k>1\) time units may be modeled by \(k-1\) tick actions followed (or preceded) by \(\alpha\). This approach typically leads to very large transition systems. Note that in such models the minimal time difference between any pair of actions is a multiple of an a priori, fixed, time unit. For synchronous systems, for instance, in which the involved processes proceed in a lockstep fashion, discrete time domains are appropriate: one time unit corresponds to one clock pulse. In this setting, traditional temporal logics can be used to express timing constraints. The next-step operator can be used to "measure" the discrete elapse of time, i.e., \(\bigcirc \Phi\) means that \(\Phi\) holds after exactly one time unit. By defining \(\bigcirc^{k+1} \Phi=\bigcirc^{k}(\bigcirc \Phi)\) and \(\bigcirc^{0} \Phi=\Phi\), general timing constraints can be specified. Using the shorthand \(\diamond^{\leqslant k} \Phi=\bigcirc^{0} \Phi \vee \bigcirc \Phi \vee \ldots \vee \bigcirc^{k} \Phi\), the above informally stated timing constraint on the traffic light may be expressed as
\[
\square\left(\text { red } \Rightarrow \diamond^{\leqslant 30} \text { green }\right) .
\]

For synchronous systems, transition systems and logics such as LTL or CTL can be used to express timing constraints, and traditional model checking algorithms suffice.

In this monograph we do not want to restrict ourselves to synchronous systems, and will consider-as in Newtonian physics - time of a continuous nature. That is to say, the nonnegative real numbers (the set \(\mathbb{R}_{\geqslant 0}\) ) will be used as time domain. A main advantage is that there is no need to fix a minimal time unit in advance as a continuous time model is invariant against changes of the time scale. This is more adequate for asynchronous systems, such as distributed systems, in which components may proceed at distinct speeds, and is more an intuitive than a discrete time model. Transition system representations of asynchronous systems without additional timing information are indeed too abstract to adequately model timing constraints, as illustrated by the following example.

\section*{Example 9.1. A Railroad Crossing}

Consider a railroad crossing, as discussed in Example 2.30 (page 51), see the schematic representation in Figure 9.1. For this railroad crossing a control system needs to be developed that closes the gate on receipt of a signal indicating that a train is approaching and only opens the gate once the train has signaled that it entirely crossed the road. The safety property that should be established by the control system is that the gates are always closed when the train is crossing the road. The complete system consists of the three components: Train, Gate, and Controller:

Train || Gate || Controller.


Figure 9.1: Railroad crossing (time abstract).


Figure 9.2: Transition systems for processes Train (left), Controller (middle), and Gate (right).

Recall that actions common to a pair of processes need to be performed jointly, while other actions are performed autonomously. The transition systems of these processes are depicted in Figure 9.2. It follows that the composite system Train || Gate || Controller does not guarantee that the gate is closed when the train is passing the crossing. This can easily be seen by inspecting an initial fragment of the composite transition system; see Figure 9.3-it cannot be deduced from the transition system whether, after sending the "approach" signal, the train reaches the road before or after the gate has been closed.

Under the assumption, though, that the train does not exceed a certain maximum speed, a lower bound for the duration between the signal "approach" and the time instant at which the train has reached the crossing can be indicated, see Figure 9.4. Let us assume that the train needs more than 2 minutes to reach the crossing after emission of the "approach" signal. Accordingly, timing assumptions are made for the controller and the gate. On receiving the "approach" signal, after exactly 1 minute the controller will signal the gate to be lowered. The actual closing of the gate is assumed not to exceed a minute. The branching in global state \(\langle\) near, 1 , up \(\rangle\) can now be labeled with timing information:


Figure 9.3: Initial fragment of the transition system Train || Controller || Gate.


Figure 9.4: The Train (left), Controller (middle), and Gate (right) with timing assumptions.


The train can only execute the local state change near \(\xrightarrow{\text { enter }}\) in after more than 2 minutes. On the other hand, the gate is closed at most 2 minutes after receiving the "approach" signal. Therefore, the global state change
\[
\langle\text { near }, 1, \text { up }\rangle \xrightarrow{\text { enter }}\langle\text { in, } 1, \text { up }\rangle
\]
never occurs. Thus, the gate is always closed before the train reaches the crossing. The fact that the gate remains closed as long as the train is on the crossing is ensured by the fact that action raise can only happen after the train has indicated exit.

As a modeling formalism for time-critical systems, the notion of timed automata has been developed, an extension of transition systems (in fact, program graphs) with clock variables that measure the elapse of time. This model includes means to impose constraints on the residence times of states, and on the timing of actions.

\subsection*{9.1 Timed Automata}

Timed automata model the behavior of time-critical systems. A timed automaton is in fact a program graph that is equipped with a finite set of real-valued clock variables, called clocks for short. In the sequel we assume that the set of clocks is denumerable, and we will use \(x, y\), and \(z\) as clocks. Clocks are different from the usual variables, as their access is limited: clocks may only be inspected, and reset to zero. Clocks can be reset to zero after which they start increasing their value implicitly as time progresses. All clocks proceed at rate one, i.e., after the elapse of \(d\) time units, all clocks advance by \(d\). The value of a clock thus denotes the amount of time that has been elapsed since its last reset. Clocks can intuitively be considered as stopwatches that can be started and checked independently of one another. Conditions on the values of the clocks are used as enabling conditions (i.e., guards) of actions: only if the condition is fulfilled is the action enabled and capable of being taken; otherwise, the action is disabled. Conditions which depend on clock values are called clock constraints. For the sake of simplicity, it is assumed that enabling conditions only depend on clocks and not on other data variables. Clock
constrains are also used to limit the amount of time that may be spent in a location. The following definition prescribes how constraints over clocks are to be defined.

\section*{Definition 9.2. Clock Constraint}

A clock constraint over set \(C\) of clocks is formed according to the grammar
\[
g::=x<c|x \leqslant c| x>c|x \geqslant c| g \wedge g
\]
where \(c \in \mathbb{N}\) and \(x \in C\). Let \(C C(C)\) denote the set of clock constraints over \(C\).
Clock constraints that do not contain any conjunctions are atomic. Let \(A C C(C)\) denote the set of all atomic clock constraints over \(C\).

Clock constraints are often written in abbreviated form, e.g., \(\left(x \geqslant c_{1}\right) \wedge\left(x<c_{2}\right)\) may be abbreviated by \(x \in\left[c_{1}, c_{2}\right)\) or \(c_{1} \leqslant x<c_{2}\). Clock difference constraints such as \(x-y<c\) can be added at the expense of a slightly more involved theory. For simplicity, they are omitted here and we restrict the discussion to atomic clock constraints that compare a clock with a constant \(c \in \mathbb{N}\). The decidability of the model-checking problem is not affected if \(c\) is allowed to be rational. In this case the rationals in each formula can be converted into natural numbers by suitable scaling. In general, we can multiply each constant by the least common multiple of denominators of all constants appearing in all clock constraints.

Intuitively, a timed automaton is a (slightly modified) program graph, whose variables are clocks. The clocks are used to formulate the real-time assumptions on system behavior. An edge in a timed automaton is labeled with a guard (when is it allowed to take an edge?), an action (what is performed when taking the edge?), and a set of clocks (which clocks are to be reset?). A location is equipped with an invariant that constrains the amount of time that may be spent in that location. The formal definition is:

\section*{Definition 9.3. Timed Automaton}

A timed automaton is a tuple \(T A=\left(\operatorname{Loc}, A c t, C, \hookrightarrow, \operatorname{Loc}_{0}, I n v, A P, L\right)\) where
- Loc is a finite set of locations, ,
- \(L o c_{0} \subseteq L o c\) is a set of initial locations,
- Act is a finite set of actions,
- \(C\) is a finite set of clocks,
- \(\hookrightarrow \subseteq \operatorname{Loc} \times C C(C) \times \operatorname{Act} \times 2^{C} \times \operatorname{Loc}\) is a transition relation,
- Inv : Loc \(\rightarrow C C(C)\) is an invariant-assignment function,
- \(A P\) is a finite set of atomic propositions, and
- \(L: L o c \rightarrow 2^{A P}\) is a labeling function for the locations.
\(A C C(T A)\) denotes the set of atomic clock constraints that occur in either a guard or a location invariant of \(T A\).

A timed automaton is a program graph with a finite set \(C\) of clocks. Edges are labeled with tuples \((g, \alpha, D)\) where \(g\) is a clock constraint on the clocks of the timed automaton, \(\alpha\) is an action, and \(D \subseteq C\) is a set of clocks. The intuitive interpretation of \(\ell \xrightarrow{g: \alpha, D} \ell^{\prime}\) is that the timed automaton can move from location \(\ell\) to location \(\ell^{\prime}\) when clock constraint \(g\) holds. Besides, when moving from location \(\ell\) to \(\ell^{\prime}\), any clock in \(D\) will be reset to zero and action \(\alpha\) is performed. Function Inv assigns to each location a location invariant that specifies how long the timed automaton may stay there. For location \(l\), \(\operatorname{Inv}(\ell)\) constrains the amount of time that may be spent in \(\ell\). That is to say, the location \(\ell\) should be left before the invariant \(\operatorname{Inv}(\ell)\) becomes invalid. If this is not possible - as there is no outgoing transition enabled-no further progress is possible. In the formal semantics of timed automata (see Definition 9.11) this situation causes time progress to halt. As time progress is no longer possible, this situation is also known as a timelock. This phenomenon will be discussed in more detail later. The function \(L\) has the same role as for transition systems and associates to any location the set of atomic propositions that are valid in that location.

Before considering the precise interpretation of timed automata, we give some simple examples.

For depicting timed automata we adopt the drawing conventions for program graphs. Invariants are indicated inside locations and are omitted when they equal true. Edges are labeled with the guard, the action, and the set of clocks to be reset. Empty sets of clocks are often omitted. The same applies to clock constraints that are constantly true. The resetting of set \(D\) of clocks is sometimes indicated by \(\operatorname{reset}(D)\). If the actions are irrelevant, they are omitted.

\section*{Example 9.4. Guards vs. Location Invariants}

Figure 9.5(a) depicts a simple timed automaton with one clock \(x\) and one location \(\ell\) equipped with a self-loop. The self-loop can be taken if clock \(x\) has at least the value 2 , and when being taken, clock \(x\) is reset. Initially, by default clock \(x\) has the value 0 .


Figure 9.5: Some timed automata with a single clock and one of their evolutions.

Figure 9.5(b) gives an example of an execution of this timed automaton, by depicting the value of clock \(x\) vs. the elapsed time since the start of the automaton. Each time the clock is reset to 0 , the automaton traverses the self-loop at location \(\ell\). As \(\operatorname{Inv}(\ell)=\) true, time can progress without any restriction while residing in \(\ell\). In particular, a legal behavior of this automaton is to stay in location \(\ell\) ad infinitum. Formally,
\[
\operatorname{Loc}=\operatorname{Loc}_{0}=\{\ell\}, C=\{x\}, \ell \stackrel{\text { true: } x \geqslant 2,\{x\}}{ } \quad \ell \text {, and } \operatorname{Inv}(\ell)=\text { true. }
\]

Labelings and actions are omitted.


Figure 9.6: Timed automaton for the Gate.

Changing the timed automaton of Figure 9.5 (a) slightly by incorporating a location invariant \(x \leqslant 3\) in location \(\ell\) leads to the effect that \(x\) cannot progress unboundedly anymore. Rather, if \(x \geqslant 2\) (guard) and \(x \leqslant 3\) (invariant) the outgoing transition must be taken. Note that it is not specified at which time instant in the interval \([2,3]\) the transition is taken, i.e., this is determined nondeterministically. The timed automaton and an example of its behavior are illustrated in Figure 9.5(c) and (d), respectively.

Observe that the same effect is not obtained when strengthening the guard in Figure 9.5(a) into \(2 \leqslant x \leqslant 3\) while keeping \(\operatorname{Inv}(\ell)=\) true. In that case, the outgoing transition can only be taken when \(2 \leqslant x \leqslant 3\)-as in the previous scenario-but is not forced to be taken, i.e., it can simply be ignored by letting time pass while staying in \(\ell\). This is illustrated in Figure 9.5(e) and (f).

Put in a nutshell, invariants are the only means to force transitions to be taken.

\section*{Example 9.5. Timed Automaton for the Gate}

Consider the gate for the railroad crossing (see Example 9.1, page 674). Assuming that lowering the gate takes at most a single time unit, and raising the gate takes at least one and at most two time units, the timed automaton for process Gate is given as in Figure 9.6. We have Act \(=\{\) lower, raise \(\}\) and
\[
\text { Loc }=\{\text { up, comingdown, down, goingup }\}
\]
with \(L o c_{0}=\{u p\}\). The transitions of the timed automaton are
\[
\begin{array}{ll}
\text { up } \stackrel{\text { true: lower, }\{x\}}{\Longrightarrow} \text { comingdown } & \text { comingdown } \xrightarrow{\text { true: } \tau, \varnothing} \text { down } \\
\text { down } \stackrel{\text { true: raise, }\{x\}}{\longrightarrow} \text { goingup } & \text { goingup } \xrightarrow{x \geqslant 1: \tau, \varnothing} \text { up }
\end{array}
\]

The location coming down with invariant \(x \leqslant 1\) has been added to model that the maximal delay between the occurrence of action lower and the change to location down is at most


Figure 9.7: Location diagram for the timed automaton Gate.
a single time unit. Clock \(x\) is set to zero on the occurrence of action lower and thus "measures" the elapse of time since that occurrence. By restricting the residence time of coming down to \(x \leqslant 1\), the switch to down must be made within one time unit. Note that this would not have been established by having a direct edge between locations up and down with guard \(x \leqslant 1\), as the value of \(x\) would not refer to the time of occurrence of lower. In a similar way, the purpose of location goingup with invariant \(x \leqslant 2\) is to model that raising the gate takes at most two time units. In the initial location up, no constraints are imposed on the residence time, i.e., \(\operatorname{Inv}(u p)=\) true. The same applies to location down. Let \(A P=\{u p\), down \(\}\) with the labeling function \(L(u p)=\{u p\}, L(\) down \()=\{\) down \(\}\), and \(L(\) comingdown \()=L(\) goingup \()=\varnothing\).

\section*{Remark 9.6. Location Diagram}

Every finite behavior of a timed automaton can be represented by a location diagram. This depicts for every time instant up to some a priori fixed upper bound, the location of the timed automaton during that behavior. For the timed automaton of the Gate a possible real-time behavior is indicated by the location diagram in Figure 9.7.

Parallel Composition of Timed Automata For modeling complex systems it is convenient to allow parallel composition of timed automata. This allows for the modeling of time-critical systems in a compositional manner. We consider a parallel composition operator, denoted \(\|_{H}\), that is parameterized with a set of handshaking actions \(H\). This operator is similar in spirit to the corresponding operator on transition systems; see Definition 2.26, page 48: actions in \(H\) need to be performed jointly by both involved timed
automata, whereas actions outside \(H\) are performed autonomously in an interleaved fashion.

\section*{Definition 9.7. Handshaking for Timed Automata}

Let \(T A_{i}=\left(\operatorname{Loc}_{i}, \operatorname{Act}_{i}, C_{i}, \hookrightarrow_{i}, \operatorname{Loc}_{0, i}, \operatorname{Inv}_{i}, A P_{i}, L_{i}\right), i=1,2\) be timed automata with \(H \subseteq A c t_{1} \cap A c t_{2}, C_{1} \cap C_{2}=\varnothing\) and \(A P_{1} \cap A P_{2}=\varnothing\). The timed automaton \(T A_{1} \|_{H} T A_{2}\) is defined as
\[
\left(L o c_{1} \times L o c_{2}, A c t_{1} \cup A c t_{2}, C_{1} \cup C_{2}, \hookrightarrow, L o c_{0,1} \times L o c_{0,2}, I n v, A P_{1} \cup A P_{2}, L\right)
\]
where \(L\left(\left\langle\ell_{1}, \ell_{2}\right\rangle\right)=L_{1}\left(\ell_{1}\right) \cup L_{2}\left(\ell_{2}\right)\) and \(\operatorname{Inv}\left(\left\langle\ell_{1}, \ell_{2}\right\rangle\right)=\operatorname{Inv}_{1}\left(\ell_{1}\right) \wedge \operatorname{Inv} v_{2}\left(\ell_{2}\right)\). The transition relation \(\hookrightarrow\) is defined by the following rules:
- for \(\alpha \in H\) :
\[
\frac{\ell_{1} \stackrel{g_{1}: \alpha, D_{1}}{\longrightarrow}{ }_{1} \ell_{1}^{\prime} \wedge \ell_{2} \stackrel{g_{2}: \alpha, D_{2}}{\longrightarrow}}{2} \ell_{2}^{\prime}
\]
- for \(\alpha \notin H\) :
\[
\frac{\ell_{1} \stackrel{g: \alpha, D}{\longrightarrow} \ell_{1}^{\prime}}{\left\langle\ell_{1}, \ell_{2}\right\rangle \stackrel{g: \alpha, D}{\longrightarrow}\left\langle\ell_{1}^{\prime}, \ell_{2}\right\rangle} \text { and } \frac{\ell_{2} \stackrel{\text { g: }, D}{ }{ }_{2} \ell_{2}^{\prime}}{\left\langle\ell_{1}, \ell_{2}\right\rangle \stackrel{g: \alpha, D}{\longrightarrow}\left\langle\ell_{1}, \ell_{2}^{\prime}\right\rangle}
\]

The location invariant of a composite location is simply the conjunction of the location invariants of its constituents. For \(\alpha \in H\), the transition in the resulting timed automaton is guarded by the conjunction of the guards of the individual timed automata. This entails that an action in \(H\) can only be taken when it is enabled in both timed automata. Besides, the clocks that are reset in the individual automata are all reset. As for transition systems, the operator \(\|_{H}\) is associative for a fixed set \(H\). Let \(T A_{1}\left\|_{H} T A_{2}\right\|_{H} \ldots \| T A_{n}\) denote the parallel composition of timed automata \(T A_{1}\) through \(T A_{n}\) where \(H \subseteq A c t_{1} \cap \ldots \cap A_{c t_{n}}\), assuming that all timed automata are compatible, i.e., each pair of \(T A_{i}\) and \(T A_{j}, i \neq j\) have disjoint sets of atomic propositions and disjoint clocks.

\section*{Example 9.8. Railroad Crossing}

Consider again the railroad crossing example. We extend the timed automaton for the gate (see Example 9.5) with timed automata for the controller and the train. The complete system is then given by
\[
\left(\text { Train } \|_{H_{1}} \text { Controller }\right) \|_{H_{2}} \text { Gate }
\]


Figure 9.8: Timed automata for the train and the controller .
where \(H_{1}=\{\) approach, exit \(\}\) and \(H_{2}=\{\) lower, raise \(\}\).
Let us assume that the train signals its approaching of the gate at least two time units before it enters the railroad crossing. Besides, it is assumed that the train has sufficient speed such that it leaves the crossing five time units after approaching it, at the latest. The timed automaton for the Train is depicted in Figure 9.8 (left). On approaching the gate, clock \(y\) is set to zero, and only if \(y>2\) is the train allowed to enter the crossing. The Controller is depicted in Figure 9.8 (right) and is forced to send the signal "lower" (to the Gate) exactly after one time unit after the Train has signaled its approaching.

Figure 9.9 shows the composite timed automaton. The prefix of a possible behavior of the complete system is sketched in the location diagram in Figure 9.10.

Note that this timed automaton contains the location \(\langle i n, 1, u p\rangle\). In this location, the train is at the crossing while the gate it still open. However, this location turns out to be unreachable. It can only be reached when \(y>2\), but as \(y\) and \(z\) are reset at the same time (on entrance of the preceding location), \(y>2\) implies \(z>2\), which is impossible due to the location invariant \(z \leqslant 1\).

\subsection*{9.1.1 Semantics}

The previous examples suggest that the state of a timed automaton is determined by its current location and the current values of all its clocks. In fact, any timed automaton canlike program graphs-be interpreted as a transition system. Due to the continuous time domain, these underlying transition systems have infinitely many states (even uncountably


Figure 9.9: The timed automaton (Train \(\|_{H_{1}}\) Controller) \(\|_{H_{2}}\) Gate.


Figure 9.10: Location diagram for a behavior of \(\left(\operatorname{Train} \|_{H_{1}}\right.\) Controller \() \|_{H_{2}}\) Gate.
many), and are infinitely branching. Timed automata can thus be considered as finite descriptions of infinite transition systems. The underlying transition system of a timed automaton results from unfolding. Its states consist of a control component, i.e., a location \(\ell\) of the timed automaton, together with a valuation \(\eta\) of the clocks. States are thus pairs of the form \(\langle\ell, \eta\rangle\). Let us first consider clock valuations.

\section*{Definition 9.9. Clock Valuation}

A clock valuation \(\eta\) for a set \(C\) of clocks is a function \(\eta: C \rightarrow \mathbb{R}_{\geqslant 0}\), assigning to each clock \(x \in C\) its current value \(\eta(x)\).

Let \(\operatorname{Eval}(C)\) denote the set of all clock valuations over \(C\). In the following, we often use notations like \(\left[x=v, y=v^{\prime}\right]\) to denote the clock evaluation \(\eta \in \operatorname{Eval}(\{x, y\})\) with \(\eta(x)=v\) and \(\eta(y)=v^{\prime}\).

We can now formally define what it means for a clock constraint to hold for a clock valuation or not. This is done in a similar way as characterizing the semantics of a temporal logic, namely by defining a satisfaction relation. In this case the satisfaction relation \(\models\) is a relation between clock valuations (over a set of clocks \(C\) ) and clock constraints (over C).

\section*{Definition 9.10. Satisfaction Relation for Clock Constraints}

For set \(C\) of clocks, \(x \in C, \eta \in \operatorname{Eval}(C), c \in \mathbb{N}\), and \(g, g^{\prime} \in C C(C)\), let \(=\subseteq \operatorname{Eval}(C) \times C C(C)\) be defined by
\[
\begin{array}{ll}
\eta \models \operatorname{true} & \\
\eta \models x<c & \text { iff } \eta(x)<c \\
\eta \models x \leqslant c & \text { iff } \eta(x) \leqslant c \\
\eta \models \neg g & \text { iff } \eta \not \models g \\
\eta \models g \wedge g^{\prime} & \\
\text { iff } \eta \models g \wedge \eta \models g^{\prime}
\end{array}
\]

Let \(\eta\) be a clock valuation on \(C\). For positive real \(d, \eta+d\) denotes the clock valuation where all clocks of \(\eta\) are increased by \(d\). Formally, \((\eta+d)(x)=\eta(x)+d\) for all clocks \(x \in C\). reset \(x\) in \(\eta\) denotes the clock valuation which is equal to \(\eta\) except that clock \(x\) reset. Formally:
\[
(\text { reset } x \text { in } \eta)(y)= \begin{cases}\eta(y) & \text { if } y \neq x \\ 0 & \text { if } y=x\end{cases}
\]

For the clock valuation \(\eta=[x=\pi, y=4]\), valuation \(\eta+9=[x=\pi+9, y=13]\), and reset \(x\) in \((\eta+9)=[x=0, y=13]\). Nested occurrences of reset are typically abbreviated. For instance, reset \(x\) in (reset \(y\) in \(\eta\) ) is denoted reset \(x, y\) in \(\eta\).

There are two possible ways in which a timed automaton can proceed: by taking a transition in the timed automaton, or by letting time progress while staying in a location. In the underlying transition system, the former is represented by a discrete transition and the latter by a delay transition. In the former case, the corresponding transition of the underlying transition system is labeled with the action of the transition in the timed automaton, in the latter case, it is labeled with a positive real number indicating the amount of time that has elapsed.

\section*{Definition 9.11. Transition System Semantics of a Timed Automaton}

Let \(T A=\left(L o c, A c t, C, \hookrightarrow, \operatorname{Loc}_{0}, \operatorname{Inv}, A P, L\right)\) be a timed automaton. The transition system \(T S(T A)=\left(S, A c t^{\prime}, \rightarrow, I, A P^{\prime}, L\right)\) with:
- \(S=\operatorname{Loc} \times \operatorname{Eval}(C)\)
- \(A c t^{\prime}=A c t \cup \mathbb{R}_{\geqslant 0}\)
- \(I=\left\{\left\langle\ell_{0}, \eta\right\rangle \mid \ell_{0} \in \operatorname{Loc}_{0} \wedge \eta(x)=0\right.\) for all \(\left.x \in C\right\}\)
- \(A P^{\prime}=A P \cup A C C(C)\)
- \(L^{\prime}(\langle\ell, \eta\rangle)=L(\ell) \cup\{g \in A C C(C) \mid \eta \models g\}\)
- the transition relation \(\rightarrow\) is defined by the following two rules:
- discrete transition: \(\langle\ell, \eta\rangle \xrightarrow{\alpha}\left\langle\ell^{\prime}, \eta^{\prime}\right\rangle\) if the following conditions hold:
(a) there is a transition \(\ell \xrightarrow{g: \alpha, D} \ell^{\prime}\) in \(T A\)
(b) \(\eta \models g\)
(c) \(\eta^{\prime}=\) reset \(D\) in \(\eta\)
(d) \(\eta^{\prime} \models \operatorname{Inv}\left(\ell^{\prime}\right)\)
- delay transition: \(\langle\ell, \eta\rangle \xrightarrow{d}\langle\ell, \eta+d\rangle\) for \(d \in \mathbb{R}_{\geqslant 0}\)
(e) if \(\eta+d \models \operatorname{Inv}(\ell)\)

For a transition that corresponds to (a) traversing a transition \(\ell \stackrel{g: \alpha, D}{\hookrightarrow} \ell^{\prime}\) in the timed automaton \(T A\) it must hold that (b) \(\eta\) satisfies the clock constraint \(g\) (ensuring the transition is enabled), and (c) the new clock valuation \(\eta^{\prime}\) is obtained by resetting all clocks \(D\) in \(\eta\) should (d) satisfy the location invariant of \(\ell^{\prime}\) (otherwise it is not allowed to be in \(\ell^{\prime}\) ). Idling in a location (second clause) for some non-negative amount of time is allowed (e) if the location invariant remains true while time progresses. For state \(\langle\ell, \eta\rangle\) such that \(\eta \models \operatorname{Inv}(\ell)\), there are typically uncountably many outgoing delay transitions of the form \(\langle\ell, \eta\rangle \xrightarrow{d}\) as \(d\) can be selected from a continuous domain.

\section*{Example 9.12. Light Switch}

The timed automaton Switch in Figure 9.11 models a switch that controls a light. When off, the switch may be turned on at any time instant. The user may switch off the light at least one time unit after the most recent time the light was switched on. After two time units the light automatically switches off. Clock \(x\) is used to keep track of the delay since the last time the light has been switched on. (The timed automaton does not distinguish between the switch_off action activated by the user and by the light. This could be made explicit by adding an edge from location on to off, with guard \(x=2\) and action \(\tau\).) The following location diagram indicates a possible behavior:



Figure 9.11: A simple light switch.

The transition system \(T S\) (Switch) has the state space
\[
S=\left\{\langle\text { off }, t\rangle \mid t \in \mathbb{R}_{\geqslant 0}\right\} \cup\left\{\langle\text { on }, t\rangle \mid t \in \mathbb{R}_{\geqslant 0}\right\}
\]
where \(t\) is a shorthand for the clock evaluation \(\eta\) with \(\eta(x)=t\). Uncountably many transitions emanate from the initial state \(\langle\) off, 0\(\rangle\). TS (Switch) has the following transitions for reals \(d\) and \(t\) :
\[
\begin{array}{lll}
\langle o f f, t\rangle & \xrightarrow{d} & \langle o f f, t+d\rangle \\
\langle o f f, t\rangle & \text { for all } t \geqslant 0 \text { and } d \geqslant 0 \\
\langle\text { switch_on } & \langle o n, 0\rangle & \text { for all } t \geqslant 0 \\
\langle o n, t\rangle & \xrightarrow{d} & \langle o n, t+d\rangle \\
\langle o n, t\rangle & \text { for all } t \geqslant 0 \text { and } d \geqslant 0 \text { with } t+d \leqslant 2 \\
\langle\text { switch_off } & \langle o f f, t\rangle & \text { for all } 1 \leqslant t \leqslant 2 .
\end{array}
\]

The set of reachable states in \(T S(\) Switch ) from state \(\langle o f f, 0\rangle\) is
\[
\left\{\langle o f f, t\rangle \mid t \in \mathbb{R}_{\geqslant 0}\right\} \cup\{\langle\text { on }, t\rangle \mid 0 \leqslant t \leqslant 2\}
\]
as the location invariant \(x \leqslant 2\) is violated in any state \(\langle o n, t\rangle\) with \(t>2\). A prefix of an example path of \(T S\) (Switch) is
\[
\begin{aligned}
& \langle o f f, 0\rangle \xrightarrow{0.57}\langle\text { off }, 0.57\rangle \xrightarrow{\text { switch_on }}\langle\text { on, } 0\rangle \xrightarrow{\sqrt{2}}\langle\text { on, } \sqrt{2}\rangle \xrightarrow{0.2} \\
& \langle\text { on, } \sqrt{2}+0.2\rangle \xrightarrow{\text { switch_off }}\langle\text { off }, \sqrt{2}+0.2\rangle \xrightarrow{\text { switch_on }}\langle\text { on, } 0\rangle \xrightarrow{1.7}\langle\text { on, 1.7 }\rangle \ldots
\end{aligned}
\]

\section*{Remark 9.13. Parallel Composition}

For timed automata we have
\[
T S\left(T A_{1}\right) \|_{H \cup \mathbb{R}_{>0}} T S\left(T A_{2}\right)=T S\left(T A_{1} \|_{H} T A_{2}\right)
\]
up to isomorphism. This is due to the fact that \(T A_{1}\) and \(T A_{2}\) do not have any shared variables. Synchronization over time passage actions reflects the natural fact that time proceeds equally fast in both components.

The paths in \(T S(T A)\) are discrete representations of the continuous-time "behavior" of TA. They indicate at least the states immediately before and after the execution of an action \(\alpha \in\) Act. However, due to the fact that, e.g., interval delays may be realized in uncountably many ways, different paths may describe the same behavior (i.e., location diagram). Consider, e.g., the behavior of the light switch of Example 9.12 where the light alternates between off and on while being off for exactly one time unit and on for two time units, i.e., a return to off takes place after exactly three time units. The following three example paths correspond to this continuous-time behavior:

The only difference between these paths is the delay transitions. In path \(\pi_{1}\), the one time unit residence in the initial state \(\langle o f f, 0\rangle\) is realized by means of the delay transition \(\langle o f f, 0\rangle \xrightarrow{1}\langle o f f, 1\rangle\). In contrast, the paths \(\pi_{2}\) and \(\pi_{3}\) realize this single time unit by two delay transitions:
\[
\begin{array}{llll}
\langle o f f, 0\rangle & \xrightarrow{0.5} & \langle o f f, 0.5\rangle \xrightarrow{0.5}\langle o f f, 1\rangle \quad \text { and } \\
\langle o f f, 0\rangle & \xrightarrow{0.1} & \langle o f f, 0.1\rangle \xrightarrow{0.9} & \langle o f f, 1\rangle .
\end{array}
\]

But the effect of the transition \(\langle\ell, \eta\rangle \xrightarrow{d_{1}+d_{2}}\left\langle\ell, \eta+d_{1}+d_{2}\right\rangle\) corresponds to the effect of the sequence of transitions:
\[
\langle\ell, \eta\rangle \xrightarrow{d_{1}}\left\langle\ell, \eta+d_{1}\right\rangle \xrightarrow{d_{2}}\left\langle\ell, \eta+d_{1}+d_{2}\right\rangle .
\]

In both cases, \(d_{1}+d_{2}\) time units pass without executing an action \(\alpha \in\) Act. Thereby, uncountably many states of the form \(\langle\ell, \eta+t\rangle\) with \(0 \leqslant t \leqslant d_{1}+d_{2}\) are passed through.

\section*{Remark 9.14. Multiple Actions in Zero Time}

The elapse of time in timed automata only takes place in locations. Actions \(\alpha \in\) Act take place instantaneously, i.e., they have a duration of zero time units. As a result, at a single time instant, several actions take place.

\subsection*{9.1.2 Time Divergence, Timelock, and Zenoness}

The semantics of a timed automaton is given by a transition system with uncountably many states (and transitions). Paths through this transition system correspond to possible
behaviors of the timed automaton. However, not every such path represents a realistic behavior. This subsection treats three essential phenomena for timed automata: time divergence, timelock, and zenoness.

Time Divergence Consider a location \(\ell\) such that for any \(t<d\), for fixed constant \(d \in \mathbb{R}_{>0}\), clock valuation \(\eta+t \models \operatorname{Inv}(\ell)\). A possible execution fragment starting from this location is
\[
\langle\ell, \eta\rangle \xrightarrow{d_{1}}\left\langle\ell, \eta+d_{1}\right\rangle \xrightarrow{d_{2}}\left\langle\ell, \eta+d_{1}+d_{2}\right\rangle \xrightarrow{d_{3}}\left\langle\ell, \eta+d_{1}+d_{2}+d_{3}\right\rangle \xrightarrow{d_{4}} \ldots
\]
where \(d_{i}>0\) and the infinite sequence \(d_{1}+d_{2}+\ldots\) converges toward \(d\). Such infinite path fragments are called time-convergent. A time-convergent path is counterintuitive as time advances only up to a certain value whereas by nature time always progresses. For example, the transition system of the timed automaton for the light switch (see Figure 9.11 on page 689) exhibits the time-convergent execution fragment
\[
\langle o f f, 0\rangle \xrightarrow{2^{-1}}\left\langle\text { off }, 1-2^{-1}\right\rangle \xrightarrow{2^{-2}}\left\langle\text { off }, 1-2^{-2}\right\rangle \xrightarrow{2^{-3}}\left\langle\text { off }, 1-2^{-3}\right\rangle \ldots \ldots
\]
which visits infinitely many states in the interval \(\left[\frac{1}{2}, 1\right]\). Time never proceeds beyond one. The corresponding path is time-convergent. As time-convergent paths are not realistic, they are not considered. That is to say, the analysis of timed automata is focused on time-divergent paths, i.e., paths in which time always progresses.

In order to formally define time-divergent paths, let us first define the elapsed time of a path. Intuitively, the elapsed time of a path is the total time that elapses along a path. The duration of an action \(\alpha \in\) Act is zero; the duration of a delay action \(d\) is \(d\).

\section*{Definition 9.15. Elapsed Time on a Path}

Let \(T A\) be a timed automaton with the set Act of actions. The function ExecTime : Act \(\cup \mathbb{R}_{>0} \rightarrow \mathbb{R}_{\geqslant 0}\) is defined as
\[
\operatorname{Exec} \operatorname{Time}(\tau)= \begin{cases}0 & \text { if } \tau \in A c t \\ d & \text { if } \tau=d \in \mathbb{R}_{>0} .\end{cases}
\]

For infinite execution fragment \(\rho=s_{0} \xrightarrow{\tau_{0}} s_{1} \xrightarrow{\tau_{1}} s_{2} \ldots\) in \(T S(T A)\) with \(\tau_{i} \in\) Act \(\cup \mathbb{R}_{>0}\) let
\[
\operatorname{ExecTime}(\rho)=\sum_{i=0}^{\infty} \operatorname{ExecTime}\left(\tau_{i}\right)
\]

The execution time of finite execution fragments is defined analogously. For the path fragment \(\pi\) in \(\operatorname{TS}(T A)\) induced by \(\rho\), \(\operatorname{ExecTime}(\pi)=\operatorname{Exec} T i m e(\rho)\).

Note that path fragment \(\pi\) may be induced by several execution fragments. However, every pair of execution fragments with the same path fragment are only distinguished by discrete transitions and not by delay transitions. Hence, \(\operatorname{Exec} \operatorname{Time}(\pi)\) is well-defined.

\section*{Definition 9.16. Time Divergence and Time Convergence}

The infinite path fragment \(\pi\) is time-divergent if \(\operatorname{Exec} \operatorname{Time}(\pi)=\infty\); otherwise, \(\pi\) is time-convergent.

\section*{Example 9.17. Light Switch}

For the light switch described in Example 9.12 (page 688), the path \(\pi\) in \(T S(\) Switch ) in which on and off periods of 1 minute alternate:
\[
\pi=\langle o f f, 0\rangle\langle o f f, 1\rangle\langle o n, 0\rangle\langle o n, 1\rangle\langle o f f, 1\rangle\langle o f f, 2\rangle\langle o n, 0\rangle\langle o n, 1\rangle\langle o f f, 1\rangle \ldots
\]
is time-divergent as \(\operatorname{Exec} \operatorname{Time}(\pi)=1+1+1+\ldots=\infty\). The path
\[
\pi^{\prime}=\langle o f f, 0\rangle\langle o f f, 1 / 2\rangle\langle o f f, 3 / 4\rangle\langle o f f, 7 / 8\rangle\langle o f f, 15 / 16\rangle \ldots
\]
in \(T S(\) Switch \()\) is time-convergent, as \(\operatorname{Exec} \operatorname{Time}\left(\pi^{\prime}\right)=\sum_{i=0}^{\infty}\left(\frac{1}{2}\right)^{i+1}=1<\infty\).

\section*{Definition 9.18. Time-Divergent Set of Paths}

For state \(s\) in \(T S(T A)\) let: \(\operatorname{Paths}_{\operatorname{div}}(s)=\{\pi \in \operatorname{Paths}(s) \mid \pi\) is time-divergent \(\}\).
That is, Paths \({ }_{d i v}(s)\) denotes the set of time-divergent paths in \(T S(T A)\) that start in \(s\). Although time-convergent paths are not realistic, their existence cannot be avoided. For the analysis of timed automata, time-convergent paths are simply ignored, e.g., a timed automaton satisfies an invariant when along all its time-divergent paths the invariant is satisfied.

Timelock. State \(s\) in \(T S(T A)\) contains a timelock if there is no time-divergent path starting in \(s\). Such states are unrealistic as time cannot progress for ever from these states. Timelocks are considered as undesired and need to be avoided when modeling a time-critical system by means of a timed automaton.

\section*{Definition 9.19. Timelock}

Let \(T A\) be a timed automaton. State \(s\) in \(T S(T A)\) contains a timelock if \(\operatorname{Paths}_{d i v}(s)=\varnothing\). \(T A\) is timelock-free if no state in \(\operatorname{Reach}(T S(T A))\) contains a timelock.


Figure 9.12: Timed automaton Switch \(_{1}\).


Figure 9.13: Timed automaton Switch \(_{2}\).

\section*{Example 9.20. Modified Light Switch}

We modify the light switch such that the light is on for a period with duration \(t \in[1,2)\), i.e., the light is always switched off within 2 minutes; see the timed automaton Switch \(_{1}\) in Figure 9.12. The state \(\langle o n, 2\rangle\) is reachable in transition system \(T S\left(\right.\) Switch \(\left._{1}\right)\), e.g., via the execution fragment:
\[
\langle o f f, 0\rangle \xrightarrow{\text { switch_on }}\langle\text { on, } 0\rangle \xrightarrow{2}\langle\text { on, } 2\rangle .
\]

As \(\langle o n, 2\rangle\) is a terminal state, Paths \(\operatorname{div}(\langle o n, 2\rangle)=\varnothing\), and the state contains a timelock. Timed automaton Switch \(_{1}\) is thus not timelock-free.

Any terminal state of a transition system that results from a timed automaton contains a timelock. Terminal states should not be confused with terminal locations, i.e., locations that have no outgoing edges. A terminal location \(\ell\) with \(\operatorname{Inv}(\ell)=\) true, e.g., does not result in a terminal state in the underlying transition system, as time may progress in \(\ell\) for ever. Terminal locations thus do not necessarily yield states with timelocks.

Not only terminal states may contain timelocks. Consider, e.g., another variant of the light switch where \(\operatorname{Inv}(o n)=x<3\), see timed automaton Switch \(_{2}\) in Figure 9.13. The reachable state \(\langle o n, 2\rangle\) is not terminal, , e.g., the time-convergent path in \(T S\left(\right.\) Switch \(\left._{2}\right)\) :
\[
\langle o n, 2\rangle\langle o n, 2.9\rangle\langle o n, 2.99\rangle\langle o n, 2.999\rangle\langle o n, 2.9999\rangle \ldots
\]
emanates from it. However, Paths \({ }_{d i v}(\langle o n, 2\rangle)=\varnothing\) as the state \(\langle o n, 2\rangle\) has no outgoing discrete transitions (as the guard \(1 \leqslant x<2\) is violated), and time cannot progress beyond three (due to \(\operatorname{Inv}(o n)=x<3\) ). State \(\langle o n, 2\rangle\) in \(T S\left(\right.\) Switch \(\left._{2}\right)\) contains a timelock. Timed automaton Switch \(_{2}\) is thus not timelock-free.

Zenoness As opposed to the presence of time-convergent paths, timelocks are considered as modeling flaws that should be avoided. The latter also applies to zenoness. Recall that the execution of actions \(\alpha \in\) Act is instantaneous, i.e., actions take no time. Without further restrictions, a timed automaton may perform infinitely many actions in a finite time interval. This phenomenon is also called zeno and represents nonrealizable behavior, since it would require infinitely fast processors.

\section*{Definition 9.21. Zeno Paths}

Let \(T A\) be a timed automaton. The infinite path fragment \(\pi\) in \(T S(T A)\) is zeno (or: a zeno path) if it is time-convergent and infinitely many actions \(\alpha \in\) Act are executed along \(\pi\).

\section*{Definition 9.22. Nonzenoness}

A timed automaton \(T A\) is non-zeno if there does not exist an initial zeno path in \(T S(T A)\).

Timed automaton \(T A\) is thus non-zeno if and only if for every path \(\pi\) in \(T S(T A)\), either \(\pi\) is time-divergent or \(\pi\) is time-convergent with almost only (i.e., all except for finitely many) delay transitions.

Note that non-zenoness, as well as timelock freedom, only refers to the reachable fragment of the transition system \(T S(T A)\). A non-zeno timed automaton may possess zeno paths starting in an unreachable state. Similarly, a timelock-free timed automaton may contain unreachable timelock states.

\section*{Example 9.23. Zeno Paths of a Light Switch}

Consider yet another variant of the light switch, in which the user has the possibility to push the on button while the light is on. While doing so, clock \(x\) is reset, and the light stays on for at most two units, unless the user pushes the on button again; see timed automaton Switch \(_{3}\) in Figure 9.14. The paths induced by the following execution fragments of \(T S\left(\right.\) Switch \(\left._{3}\right)\)
\[
\begin{aligned}
& \langle o f f, 0\rangle \xrightarrow{\text { switch_on }}\langle o n, 0\rangle \xrightarrow{\text { switch_on }}\langle o n, 0\rangle \xrightarrow{\text { switch_on }}\langle o n, 0\rangle \xrightarrow{\text { switch_on }} \ldots \\
& \langle o f f, 0\rangle \xrightarrow{\text { switch_on }}\langle o n, 0\rangle \xrightarrow{0.5}\langle\text { on, } 0.5\rangle \xrightarrow{\text { switch_on }}\langle\text { on, } 0\rangle \xrightarrow{\text { o.25 }}\langle\text { on, } 0.25\rangle \xrightarrow{\text { switch_on }} \ldots,
\end{aligned}
\]
are zeno paths during which the user presses the on button infinitely fast, or faster and faster, respectively.

This unrealizable behavior can be avoided by imposing a minimal non-zero delay, \(c\), say, between successive button pushings by the user. This is established by imposing \(\operatorname{Inv}(\) on \()=\)


Figure 9.14: Timed automaton Switch \(_{3}\).


Figure 9.15: Timed automaton Switch \(_{4}\).
\(x \geqslant c\) for \(c>0\). Note that \(c\) should be a natural number. In order to model a minimal response \(0<c<1\) where \(c\) is rational, say \(c=\frac{1}{100}\), all time constraints in the timed automaton Switch \(_{3}\) need to be rescaled (see Figure 9.15). Essentially, the timed automaton Switch \(_{4}\) computes with a modified time unit: one time unit for \(x\) in Switch \(_{4}\) corresponds to \(\frac{1}{100}\) minutes.

To check whether or not a timed automaton is non-zeno is algorithmically difficult. Instead, sufficient conditions are considered that are simple to check, e.g., by a static analysis of the timed automaton. The following criterion is based on the intuition that a timed automaton is non-zeno if on any of its control cycles, time advances with at least some constant amount (larger than zero). This yields:

\section*{Lemma 9.24. Sufficient Criterion for Nonzenoness}

Let TA be a timed automaton with set \(C\) of clocks such that for every control cycle in TA
\[
\ell_{0} \xrightarrow{g_{1}: \alpha_{1}, C_{1}} \ell_{1} \xrightarrow{g_{2}: \alpha_{2}, C_{2}} \ldots \xrightarrow{g_{n}: \alpha_{n}, C_{n}} \ell_{n} \text { with } \ell_{0}=\ell_{n} \text {, }
\]
there exists a clock \(x \in C\) such that
1. \(x \in C_{i}\) for some \(0<i \leqslant n\), and
2. for all clock evaluations \(\eta\) there exists \(c \in \mathbb{N}_{>0}\) such that
\[
\eta(x)<c \quad \text { implies } \quad\left(\eta \not \vDash g_{j} \text { or } \operatorname{Inv}\left(\ell_{j}\right)\right), \text { for some } 0<j \leqslant n .
\]

Then: TA is non-zeno.

Proof: Let \(T A\) be a timed automaton over \(C\) with \(x \in C\) satisfying the two constraints stated in the claim and \(i, j\) the corresponding indices. Let \(\pi\) be a path in \(T S(T A)\) that performs infinitely many actions \(\alpha \in\) Act. As TA contains finitely many states, \(\pi\) traverses some control cycle \(\ell_{0} \hookrightarrow \ell_{1} \hookrightarrow \ldots \hookrightarrow \ell_{n}=\ell_{0}\). Assume that \(i \leqslant j\). (As locations on cycles can be renumbered, this is not a restriction.) Consider the path fragment in \(T S(T A)\) that starts and ends in location \(\ell_{0}\) :
\[
\left\langle\ell_{0}, \eta_{0}\right\rangle \ldots\left\langle\ell_{i-1}, \eta_{i-1}\right\rangle\left\langle\ell_{i}, \eta_{i}\right\rangle \ldots\left\langle\ell_{j-1}, \eta_{j-1}\right\rangle\left\langle\ell_{j}, \eta_{j}\right\rangle \ldots\left\langle\ell_{0}, \eta_{0}^{\prime}\right\rangle .
\]

By the constraints satisfied by \(T A\), clock \(x\) is reset on the transition \(\ell_{i-1} \hookrightarrow \ell_{i}\), and the transition \(\ell_{j-1} \hookrightarrow \ell_{j}\) is only possible when \(\eta_{i-1}(x) \geqslant c\) (as for \(\eta_{i-1}(x)<c\), either the guard or the location invariant of \(\ell_{j}\) are violated). This implies that by traversing the cycle \(\ell_{0} \hookrightarrow \ldots \hookrightarrow \ell_{0}\) time advances with at least \(c>0\) time units. Hence, \(\pi\) is time-divergent and \(T A\) is non-zeno.

The condition in Lemma 9.24 is compositional, i.e., if \(T A\) and \(T A^{\prime}\) both satisfy the constraints, then the parallel composed timed automaton \(T A \| T A^{\prime}\) also satisfies the constraints. This follows directly from the fact that a control cycle in \(T A \| T A^{\prime}\) consists of a control cycle in \(T A\), or \(T A^{\prime}\), or both. If each control cycle in \(T A\) and in \(T A^{\prime}\) satisfies the constraint that time should advance with at least some positive amount, then this thus also applies to each control cycle in \(T A \| T A^{\prime}\). This property significantly simplifies to checking whether a composite timed automaton is non-zeno. In case a timed automaton is in fact untimed (as no clocks are used or all guards and invariants are vacuously true), it can be considered as non-zeno, and thus not affect the control cycles of other component timed automata in a composite system.

\section*{Example 9.25. Sufficient Condition for Nonzenoness}

The timed automaton in Figure 9.15 (page 695) satisfies the constraints in Lemma 9.24. In the control cycle off \(\hookrightarrow\) on \(\hookrightarrow\) off, clock \(x\) is reset on off \(\hookrightarrow\) on, and the guard \(x \geqslant 100\) ensures that when going from location off to on, time has advanced with at least 100 time units. On the control cycle on \(\hookrightarrow\) on, clock \(x\) is reset, and the guard \(x \geqslant 1\) ensures that at least one time unit elapses on traversing this control cycle.

The timed automata Train, Gate, and Controller of Example 9.8 (page 683) all satisfy for any control cycle the constraints in Lemma 9.24. This can be seen as follows. Timed
automaton Gate has one control cycle: up \(\hookrightarrow \ldots \hookrightarrow u p\). In that cycle, clock \(x\) is reset when moving from location down to going_up. Furthermore, for \(\eta(x)<1\), location up is not reachable due to the guard \(x \geqslant 1\) on going_up \(\hookrightarrow u p\). This ensures that on traversing the control cycle \(u p \hookrightarrow \ldots \hookrightarrow u p\), time advances with at least one time unit. The timed automaton Train contains the control cycle far \(\hookrightarrow \ldots \hookrightarrow\) far. However, clock \(y\) is reset on that cycle before reaching location near, and the guard \(y>2\) on near \(\hookrightarrow\) in guarantees that on traversing the control cycle at least one (in fact, more than two) time unit has elapsed. For Controller, the resetting of clock \(z\) and guard \(z=1\) ensure that also this timed automaton fulfills the constraints of Lemma 9.24. The timed automata Train, Gate and Controller are thus non-zeno. As the control cycles in the composed timed automaton (Train \(\|_{H_{1}}\) Gate) \(\|_{H_{2}}\) Controller are comprised of the constituting timed automata, this composed timed automaton is non-zeno.

The previous considerations indicate that a timed automaton is adequately modeling a time-critical system whenever it is non-zeno and does not contain any timelock. Timelockfree, non-zeno timed automata induce transition systems without terminal states such that along any path only finitely many actions are executed in finite time. In contrast to zeno paths and timelocks, time-convergent paths will be treated akin to unfair paths (in fair CTL) and are explicitly excluded for analysis purposes.

A delay of \(d>0\) time units can be realized in different ways, in general by \(n>0\) delay transitions of size \(d_{1}\) through \(d_{n}\) with \(d_{i}>0\) such that \(d=d_{1}+\ldots+d_{n}\). As we are only interested in the amount of time advancing, a sequence of delay transitions labeled with \(d_{1}\) through \(d_{n}\) and a sequence labeled with \(d_{1}^{\prime}\) through \(d_{k}^{\prime}\), say, such that \(\sum_{i=1}^{n} d_{i}=\sum_{i=1}^{k} d_{i}^{\prime}=\) \(d\), are considered equivalent and denoted by \(\stackrel{d}{\Longrightarrow}\). This relation is used later for defining the semantics of timed CTL.

\section*{Notation 9.26. Sets of Path Fragments}

Let \(T A\) be a timed automaton. For path fragments in \(T S(T A)\) along which infinitely many actions are performed, let
\[
s_{0} \xrightarrow{d_{0}} s_{1} \stackrel{d_{1}}{\Longrightarrow} s_{2} \xrightarrow{d_{2}} \ldots \quad \text { with } d_{0}, d_{1}, d_{2} \ldots \geqslant 0
\]
denote the equivalence class containing all infinite path fragments induced by execution fragments in \(T S(T A)\) of the form
\[
s_{0} \xlongequal[\substack{\text { time passage of } \\ d_{0} \text { time-units }}]{d_{0}^{1}} \ldots d_{0}^{d_{0}^{k_{0}}} s_{0} \xrightarrow{\alpha_{0}} s_{1} \underbrace{d_{1}^{1}}_{\substack{\text { time passage of } \\ d_{1} \text { time-units }}} \ldots \stackrel{d_{1}^{k_{1}}}{\rightleftarrows} s_{1}+d_{1} \xrightarrow{\alpha_{1}} s_{2} \underbrace{d_{2}^{1}}_{\substack{\text { time passage of } \\ d_{2} \text { time-units }}} \ldots d_{2}^{d_{2}^{k_{2}}} s_{2}+d_{2} \xrightarrow{\alpha_{2}} \ldots
\]
where \(k_{i} \in \mathbb{N}, d_{i} \in \mathbb{R}_{\geqslant 0}\) and \(\alpha_{i} \in\) Act such that \(\sum_{j=1}^{k_{i}} d_{i}^{j}=d_{i}\). Note that in the \(\Rightarrow\) notations, actions are abstracted away.

For infinite path fragment \(\pi \in s_{0} \xrightarrow{d_{0}} s_{1} \xrightarrow{d_{1}} \ldots\) that performs infinitely many actions, we have \(\operatorname{ExecTime}(\pi)=\sum_{i \geqslant 0} d_{i}\). Path fragments in \(s_{0} \xrightarrow{d_{0}} s_{1} \xrightarrow{d_{1}} \ldots\) are time-divergent if and only if \(\sum_{i} d_{i}\) diverges.

Time-divergent path fragments that perform finitely many actions \(\alpha \in\) Act (but contain infinitely many delay transitions) are represented in a similar way, except that after the execution of the last action \(\alpha \in\) Act the advance of time is represented by infinitely many \(\stackrel{1}{\Longrightarrow}\) transitions. That is, the set
\[
s_{0} \stackrel{d_{0}}{\Longrightarrow} s_{1} \stackrel{d_{1}}{\Longrightarrow} \ldots \stackrel{d_{n-1}}{\Longrightarrow} s_{n} \stackrel{1}{\Longrightarrow} s_{n}+1 \stackrel{1}{\Longrightarrow} s_{n}+2 \stackrel{1}{\Longrightarrow} \ldots
\]
contains all infinite path fragments induced by the execution fragment of the form
\[
s_{0} \underbrace{\rightarrow \ldots \rightarrow}_{\substack{\text { time passage of } \\
d_{0} \text { time-units }}} \xrightarrow{\alpha_{0}} \ldots \xrightarrow{\alpha_{n-2}} s_{n-1} \underbrace{\rightarrow \ldots \rightarrow}_{\begin{array}{c}
\text { time passage of } \\
d_{n}-1 \text { time-units }
\end{array}} \xrightarrow{\alpha_{n-1}} s_{n} \underbrace{\stackrel{1}{\rightarrow} s_{n}+1 \xrightarrow{1} s_{n}+2 \xrightarrow{1} \ldots \ldots}_{\text {infinite time passage }}
\]

Hence, \(s_{0} \xrightarrow{d_{0}} s_{1} \xrightarrow{d_{1}} \ldots\) is a uniform notation for all infinite time-divergent path fragments.

\subsection*{9.2 Timed Computation Tree Logic}

Timed CTL (TCTL, for short) is a real-time variant of CTL aimed to express properties of timed automata. In TCTL, the until modality is equipped with a time interval such that \(\Phi U^{J} \Psi\) asserts that a \(\Psi\)-state is reached within \(t \in J\) time units while only visiting \(\Phi\)-states before reaching the \(\Psi\)-state. The fact that a deadlock may be reached within thirty time units via legal states only, say, can be expressed as legal \(\cup^{[0,30]}\) deadlock, where the atomic propositions legal and deadlock have their obvious meaning. Timed CTL is sufficiently expressive to allow for the formulation of an important set of real-time system properties.

\section*{Definition 9.27. Syntax of Timed CTL}

Formulae in TCTL are either state or path formulae. TCTL state formulae over the set \(A P\) of atomic propositions and set \(C\) of clocks are formed according to the following grammar:
\[
\Phi::=\text { true }|a| g|\Phi \wedge \Phi| \quad \neg \Phi|\exists \varphi| \forall \varphi
\]
where \(a \in A P, g \in A C C(C)\) and \(\varphi\) is a path formula defined by:
\[
\varphi::=\Phi U^{J} \Phi
\]
where \(J \subseteq \mathbb{R}_{\geqslant 0}\) is an interval whose bounds are natural numbers.

Timed CTL extends CTL with atomic clock constraints over the clocks in \(C\), typically the set of clocks in the timed automaton under consideration. The propositional logic operators \(\vee, \rightarrow\), true, etc. are obtained in the usual way. The until operator is equipped with an interval \(J\) of real numbers. Timed variants of the modal operators \(\diamond\) and \(\square\) are obtained as follows: \(\diamond^{J} \Phi=\operatorname{true} U^{J} \Phi\) and
\[
\exists \square^{J} \Phi=\neg \forall \diamond^{J} \neg \Phi \quad \text { and } \quad \forall \square^{J} \Phi=\neg \exists \diamond^{J} \neg \Phi .
\]

The formula \(\exists \square^{J} \Phi\) asserts that there exists a path for which during the interval \(J, \Phi\) holds; \(\forall \square^{J} \Phi\) requires this to hold for all paths. As we will see later on when defining the formal semantics of TCTL, the path quantifiers quantify over time-divergent paths only. Accordingly, a state in \(T S(T A)\) satisfies \(\forall \diamond^{J} \Phi\) whenever all time-divergent paths starting in \(s\) satisfy \(\diamond^{J} \Phi\). The next-step operator is absent in TCTL. As the time domain is continuous there is no unique next time instant which makes it impossible to provide a suitable meaning to the next-step operator. Note that \(J \subseteq \mathbb{R}_{\geqslant 0}\) has natural bounds, i.e., the interval \(J\) is either of the form \([n, m],(n, m],[n, m)\) or \((n, m)\) for \(n, m \in \mathbb{N}\) and \(n \leqslant m\). For right-open intervals, \(m=\infty\) is allowed.

In the sequel, intervals are often denoted by shorthand, e.g., \(\diamond^{\leqslant 2}\) denotes \(\diamond^{[0,2]}\) and \(\square^{>8}\) denotes \(\square^{(8, \infty)}\). For the special case \(J=[0, \infty)\), the timing requirements are in fact trivially fulfilled. That is:
\[
\Phi \cup \cup^{[0, \infty)} \Psi=\Phi \cup \Psi \quad \text { and } \quad \diamond \Phi=\diamond^{[0, \infty)} \Phi \quad \text { and } \quad \square \Phi=\square^{[0, \infty)} \Phi
\]

The following examples illustrate the kind of timing properties that can be expressed in TCTL.

\section*{Example 9.28. Light Switch}

Consider the light switch of Example 9.12 (page 688). The property
"the light cannot be continuously switched on for more than 2 minutes"
is expressed by the TCTL formula:
\[
\forall \square\left(o n \longrightarrow \forall \diamond^{>2} \neg o n\right) .
\]

The property
"the light will stay on for at least 1 time unit and then switch off"
is expressed by the TCTL formula:
\[
\forall \square\left((o n \wedge(x=0)) \longrightarrow\left(\forall \square^{\leqslant 1} \text { on } \wedge \forall \diamond^{>1} o f f\right)\right)
\]

The clock \(x\) that occurs in the formula is used to specify the time instant at which the light is switched on.

\section*{Example 9.29. Railroad Crossing}

Consider the railroad crossing example. The safety property
"the gate is always closed when the train is at the crossing"
does not contain any timing aspects, and can be described as in CTL by the formula \(\forall \square(\) in \(\longrightarrow\) down \()\), where in and down are locations in the timed automata Train and Gate, respectively. The (timed liveness) property
"once the train is far, within 1 minute the gate is up for at least 1 minute" is expressed by the TCTL formula:
\[
\forall \square\left(f a r \longrightarrow \forall \diamond^{\leqslant 1} \forall \square \leqslant 1 \text { up }\right) .
\]

The fact that the train needs at least 2 minutes to reach the crossing after transmitting the "approach" signal is expressed by
\[
\forall \square((\text { near } \wedge(y=0)) \longrightarrow \forall \square \leqslant 2 \neg i n)
\]
where the atomic clock constraint \(y=0\) indicates the time instant at which the train signals its approach. Finally, the property that the train needs at most five minutes to pass the crossing since its approach is expressed by:
\[
\forall \square\left((n e a r \wedge(y=0)) \longrightarrow \forall \diamond^{\leqslant 5} f a r\right)
\]

The semantics of TCTL formulae is defined for states of the form \(\langle\ell, \eta\rangle\). The state formulae \(\forall \varphi\) and \(\exists \varphi\) are interpreted over all time-divergent paths. That is to say, time-convergent paths are not of any importance for the satisfaction of TCTL state formulae. This is similar to the treatment of unfair paths in the semantics of fair CTL; see Definition 6.33 on page 360 .

\section*{Definition 9.30. Satisfaction Relation for TCTL}

Let \(T A=\left(L o c, A c t, C, \hookrightarrow, L_{o c}\right.\), Inv, \(\left.A P, L\right)\) be a timed automaton, \(a \in A P, g \in A C C(C)\), and \(J \subseteq \mathbb{R}_{\geqslant 0}\). For state \(s=\langle\ell, \eta\rangle\) in \(T S(T A)\) and TCTL state formulae \(\Phi\) and \(\Psi\), and TCTL path formula \(\varphi\), the satisfaction relation \(\models\) is defined for state formulae by
\[
\begin{array}{llll}
s & =\text { true } & & \\
s \models a & & \text { iff } & \\
& a \in L(\ell) \\
s \models g & \text { iff } & \eta \models g \\
s \models \neg \Phi & & \text { iff } & \\
\text { not } s \models \Phi \\
s \models \Phi \wedge \Psi & & \text { iff } & \\
(s \models \Phi) \text { and }(s \models \Psi) \\
s \models \exists \varphi & & \text { iff } & \\
\pi \models \varphi \text { for some } \pi \in \operatorname{Paths}_{d i v}(s) \\
s \models \forall \varphi & & \text { iff } & \\
& \pi \models \varphi \text { for all } \pi \in \operatorname{Paths}_{d i v}(s) .
\end{array}
\]

For time-divergent path \(\pi \in s_{0} \stackrel{d_{0}}{\Longrightarrow} s_{1} \stackrel{d_{1}}{\Longrightarrow} \ldots\), the satisfaction relation \(\models\) for path formulae is defined by
\[
\begin{gathered}
\pi \models \Phi \cup^{J} \Psi \quad \text { iff } \quad \exists i \geqslant 0 . s_{i}+d \models \Psi \text { for some } d \in\left[0, d_{i}\right] \text { with } \\
\sum_{k=0}^{i-1} d_{k}+d \in J \quad \text { and } \\
\forall j \leqslant i . s_{j}+d^{\prime} \models \Phi \vee \Psi \text { for any } d^{\prime} \in\left[0, d_{j}\right] \text { with } \\
\sum_{k=0}^{j-1} d_{k}+d^{\prime} \leqslant \sum_{k=0}^{i-1} d_{k}+d
\end{gathered}
\]
where for \(s_{i}=\left\langle\ell_{i}, \eta_{i}\right\rangle\) and \(d \geqslant 0\) we have \(s_{i}+d=\left\langle\ell_{i}, \eta_{i}+d\right\rangle\).

The interpretations for atomic propositions, negation, and conjunction are as usual. Clock constraint \(g\) holds in \(\langle\ell, \eta\rangle\) whenever the values of the clocks in \(\eta\) satisfy \(g\). State formula \(\exists \varphi\) is true in state \(s\) if and only if there exists some time-divergent path starting in \(s\) that satisfies \(\varphi\). \(\forall \varphi\) holds in \(s\) whenever all time-divergent paths starting in \(s\) satisfy \(\varphi\). As stated before, path quantification is over time-divergent paths. Truth of \(\varphi\) on time-convergent paths of \(s\) is irrelevant. Let us now consider the semantics of the until operator. Time-divergent path \(\pi \in s_{0} \xrightarrow{d_{0}} s_{1} \xrightarrow{d_{1}} \ldots\) satisfies \(\Phi U^{J} \Psi\) whenever at some time point in \(J\), a state is reached satisfying \(\Psi\) and at all previous time instants \(\Phi \vee \Psi\) holds. The reader might wonder why it is not required-as in the temporal logics CTL and LTL before - that just \(\Phi\) holds at all preceding time instants. This is justified by the following example.


Figure 9.16: An example of a timed automaton .

\section*{Example 9.31. Semantics of Until}

Consider the timed automaton \(T A\) in Figure 9.16 and let \(\Phi=\forall\left(a U^{>1} b\right)\). Intuitively, we expect \(T A \models \Phi\), since \(\operatorname{Inv}(\ell)=x<1\), and location \(\ell^{\prime}\) cannot be left before three time units have elapsed. At time instant \(t=1.5\), e.g., \(T A\) always resides in location \(\ell^{\prime}\). Consider now the path
\[
\pi=\langle\ell, 0\rangle\langle\ell, 0.5\rangle\left\langle\ell^{\prime}, 0.5\right\rangle\left\langle\ell^{\prime}, 3\right\rangle\langle\ell, 0\rangle \ldots
\]

It follows that:
\[
\pi \in \underbrace{\langle\ell, 0\rangle}_{s_{0}} \stackrel{0.5}{\Longrightarrow} \underbrace{\left\langle\ell^{\prime}, 0.5\right\rangle}_{s_{1}} \stackrel{2.5}{\Longrightarrow} \underbrace{\langle\ell, 0\rangle}_{s_{2}} \ldots
\]

According to the TCTL semantics, \(\pi \models a \mathrm{U}^{>1} b\) since
\[
s_{1}+d \models b \text { for some } d \in[0,2.5] \text { such that } 0.5+d>1
\]
and
\[
s_{0}+d^{\prime} \models a \text { for all } d^{\prime} \in[0,0.5] \quad \text { and } \quad s_{1}+d^{\prime} \models a \vee b \text { for all } d^{\prime} \in[0, d]
\]

A semantics along the lines of CTL would require \(s_{1}+d^{\prime} \models a\) for all \(d^{\prime} \in[0, d]\) with \(d>0.5\). The event \(\neg a \wedge b\), however, occurs at a time instant that is before the required time bound \(] 1, \infty)\) is reached. As a result, the statement \(s_{1}+d^{\prime} \models a\) for all \(d^{\prime} \in[0, d]\) with \(d>0.5\) does not hold.

Note that in CTL (and LTL) it holds that \(\Phi \cup \Psi\) is equivalent to \((\Phi \vee \Psi) \cup \Psi\).

From the TCTLsemantics it follows for time-divergent path \(\pi \in s_{0} \xrightarrow{d_{0}} s_{1} \xrightarrow{d_{1}} \ldots\) that:
\[
\pi \models \diamond^{J} \Phi \text { iff } \exists i \geqslant 0 . s_{i}+d \models \Phi \text { for some } d \in\left[0, d_{i}\right] \text { with } \sum_{k=0}^{i-1} d_{k}+d \in J \text {. }
\]

As expected, a time-divergent path fragment satisfies \(\diamond^{J} \Phi\) whenever a \(\Phi\)-state is reached at some time instant \(t \in J\). For the \(\square^{J}\) operator we obtain that
\[
\pi \models \square^{J} \Phi \text { iff } \forall i \geqslant 0 . s_{i}+d \models \Phi \text { for any } d \in\left[0, d_{i}\right] \text { with } \sum_{k=0}^{i-1} d_{k}+d \in J .
\]


Figure 9.17: Timed automaton Switch \(_{3}\) (again).
Thus, \(\pi \models \square^{J} \Phi\) iff all states visited by \(\pi\) in the time interval \(J\) satisfy \(\Phi\).
Timed automaton \(T A\) satisfies TCTL state formula \(\Phi\) whenever all its initial states satisfy \(\Phi\).

\section*{Definition 9.32. TCTL Semantics for Timed Automata}

Let TA be a timed automaton with clocks \(C\) and locations Loc. For TCTL state formula \(\Phi\), the satisfaction set \(\operatorname{Sat}(\Phi)\) is defined by:
\[
\operatorname{Sat}(\Phi)=\{s \in \operatorname{Loc} \times \operatorname{Eval}(C) \mid s \models \Phi\} .
\]

The timed automaton \(T A\) satisfies TCTL state formula \(\Phi\) if and only if \(\Phi\) holds in all initial states of \(T A\) :
\[
T A \models \Phi \quad \text { if and only if } \quad \forall \ell_{0} \in \operatorname{Loc} 0 \cdot\left\langle\ell_{0}, \eta_{0}\right\rangle \models \Phi
\]
where \(\eta_{0}(x)=0\) for all \(x \in C\).

\section*{Example 9.33. TCTL Semantics}

Consider the timed automaton Switch \(_{3}\) depicted in Figure 9.17.
For the sake of convenience we only consider the reachable states in \(T S\left(\right.\) Switch \(\left._{3}\right)\). We then have:
\[
\begin{array}{ll}
\operatorname{Sat}(\forall \diamond<1 \text { off }) & =\{\langle\text { off }, t\rangle \mid t \geqslant 0\} \cup\{\langle\text { on, } t\rangle \mid 1<t \leqslant 2\} \\
\operatorname{Sat}(\exists \diamond<1 o f f) & =\{\langle\text { off }, t\rangle \mid t \geqslant 0\} \cup\{\langle\text { on, } t\rangle \mid 0<t \leqslant 2\} \\
\operatorname{Sat}(\forall \diamond(\text { on } \wedge(x=1))) & =\{\langle\text { on, } t\rangle \mid 0 \leqslant t \leqslant 1\} \\
\operatorname{Sat}(\forall \diamond(\text { on } \wedge(x=0))) & =\{\langle\text { on, } 0\rangle\} \\
\operatorname{Sat}(\forall \diamond(\text { on } \wedge(x \geqslant 3))) & =\varnothing
\end{array}
\]

Since \(S_{\text {witch }}^{3} 1=\Phi\) if and only if the initial state \(\langle o f f, 0\rangle\) of \(T S\left(S_{\text {witch }}^{3}\right.\) ) is in \(\operatorname{Sat}(\Phi)\), we have
\[
\text { Switch }_{3} \models \forall \diamond^{<1} \text { off } \text { and } \text { Switch }_{3} \models \exists \diamond^{<1} \text { off and } \text { Switch }_{3} \not \vDash \forall \diamond(\text { on } \wedge(x \in J))
\]
for any interval \(J \subseteq \mathbb{R}_{\geqslant 0}\). Consider
\[
\Phi=\forall \square((\text { on } \wedge(x=0)) \longrightarrow \forall \diamond(\text { on } \wedge(x=1))) .
\]

It follows that \(S_{\text {witch }}^{3} \boldsymbol{} \vDash \Phi\). It is, however, essential that universal path quantification only consider time-divergent paths. For example, the time-convergent path
\[
\langle o f f, 0\rangle\langle o n, 0\rangle\left\langle\text { on }, \frac{1}{2}\right\rangle\left\langle\text { on }, \frac{3}{4}\right\rangle\left\langle\text { on }, \frac{7}{8}\right\rangle\left\langle\text { on }, \frac{15}{16}\right\rangle \ldots
\]
does not visit a state satisfying on \(\wedge(x=1)\).
Note that a (time-divergent) path does not have to explicitly visit a state satisfying on \(\wedge\) \((x=1)\) to satisfy \(\Phi\), e.g., the path \(\langle o f f, 0\rangle\langle o n, 0\rangle\langle o n, 2\rangle\langle o f f, 2\rangle\langle o n, 0\rangle, \ldots\) satisfies \(\langle o n \wedge\) \((x=1)\) although a state with \(x=1\) does not appear in its representation. The state \(\langle\) on, 1\(\rangle\) is passed during the delay transition \(\langle o n, 0\rangle \xrightarrow{2}\langle o n, 2\rangle\).

Finally, note that \(S w_{i t c h} \nmid \vDash \forall \diamond\) on, as \(\operatorname{Inv}(o f f)=\) true, i.e., the timed automaton has a time-divergent path that resides in location off for ever.

\section*{Remark 9.34. TCTL vs. CTL}

Any TCTL formula \(\Phi\) in which all intervals are of the form \([0, \infty)\) may be considered as a CTL formula over the set of propositions \(A P\) and the atomic clock constraints in \(\Phi\). Due to time-divergent paths, though, there is a subtle difference between the interpretation of TCTL and CTL formulae. As a result, \(T S(T A) \models_{\text {тСте }} \forall \varphi\) and \(T S(T A) \not \vDash_{\text {СтL }} \forall \varphi\) is possible: whereas \(\models_{\text {тСтL }}\) ranges over all time-divergent paths, \(\models_{\text {СтL }}\) considers all paths, in particular also the time-convergent paths. Consider, e.g., the light switch described in Example 9.33 and the TCTL formula:
\[
\Phi=\forall \square(o n \longrightarrow \forall \diamond \text { off }) .
\]

It follows that
\[
\underbrace{T S(T A) \models \models_{\text {TСтL }} \Phi}_{\text {TCTL semantics }} \quad \text { and } \underbrace{T S(T A) \not \models_{\text {СTL }} \Phi .}_{\text {CTL semantics }} .
\]

The fact that \(T S(T A) \not \vDash_{\text {Сте }} \Phi\) results from the existence of time-convergent paths in which the location on is never left.

The semantics for TCTL is well-defined for timed automata that contain a timelock. Recall that timed automaton \(T A\) contains a timelock whenever there exists a state in \(T S(T A)\) from which no time-divergent path emanates. A state is timelock-free if and only if it
satisfies \(\exists \square\) true. The formula \(\exists \square\) true holds in state \(s\) whenever some time-divergent path satisfies \(\square\) true, i.e., whenever there is at least one time-divergent path. Note that for fair CTL, the states in which a fair path starts are also characterized by the formula \(\exists \square\) true. This yields the following characterization of timelock freeness:

\section*{Lemma 9.35. Characterizing Timelock}

Timed automaton TA is timelock-free iff \(\forall s \in \operatorname{Reach}(T S(T A)) . s \models_{T C T L} \exists \square\) true.
\(T A \models \forall \square \Phi\) if and only if all reachable states on all time-divergent paths satisfy \(\Phi\). In general, from \(T A \models \forall \square \Phi\) it may not be concluded that all reachable states in \(T S(T A)\) satisfy \(\Phi\). This holds for timelock-free timed automata, but not for others. In particular, the TCTL formula
\[
\forall \square \exists \square \text { true }
\]
is a tautology in TCTL (and not characteristic of timelock-free timed automata).

\subsection*{9.3 TCTL Model Checking}

The TCTL model-checking problem is to check for a given timed automaton \(T A\) and TCTL formula \(\Phi\) whether \(T A \models \Phi\). It is assumed that \(T A\) is non-zeno. The possible presence of timelocks is not relevant, as timelock freedom can be checked by a TCTL formula, see Remark 9.35. The main difficulty of the TCTL model-checking problem is that a transition system with uncountably many states has to be analyzed, since


A naive graph analysis in the state graph of \(T S(T A)\) is therefore not feasible. Instead, the basic idea is to consider a finite quotient of this transition system, the so-called region transition system, which is obtained from the timed automaton TA and the TCTL formula \(\Phi .{ }^{1}\) In essence, the region transition system \(R T S(T A, \Phi)\) is the quotient of \(T S(T A)\) with respect to a bisimulation relation. The states in the region transition system are equivalence classes of states in \(T S(T A)\) that all satisfy the same atomic clock constraints, and from which "similar" time-divergent paths emanate, i.e., such states are TCTL equivalent. As the number of equivalence classes is finite, this provides a basis for TCTL model checking. In fact, rather than checking the TCTL formula \(\Phi\), it is checked whether a derived CTL formula holds in \(\operatorname{RTS}(T A, \Phi)\).

\footnotetext{
\({ }^{1}\) In fact, the region transition system depends on the maximal constants with which clocks are compared in \(T A\) and the maximal timing constants in \(\Phi\).
}

To check whether a timed automaton satisfies a TCTL formula thus amounts to modelcheck its region transition system against a corresponding CTL formula. For the latter, traditional CTL model-checking algorithms can be exploited. Summarizing:
\[
T A \models_{\mathrm{TCTL}} \Phi \quad \text { iff } \underbrace{R T S(T A, \Phi)}_{\text {finite transition system }} \models_{\text {СTL }} \widehat{\Phi}
\]
where \(\widehat{\Phi}\) is a CTL formula that is obtained from the TCTL formula \(\Phi\) using the translation explained next. In summary, we obtain the scheme in Algorithm 43 where \(\cong\) denotes the equivalence used to obtain the quotient \(\operatorname{RTS}(T A, \Phi)\).
```

Algorithm 43 Basic recipe of TCTL model checking
Input: timed automaton $T A$ and TCTL formula $\Phi$ (both over $A P$ and $C$ )
Output: $T A \models \Phi$

```
    \(\widehat{\Phi}:=\) eliminate the timing parameters from \(\Phi ;\)
    determine the equivalence classes under \(\cong\);
    construct the region transition system \(T S=R T S(T A)\);
    apply the CTL model-checking algorithm to check \(T S \models \widehat{\Phi}\);
    \(T A \models \Phi\) if and only if \(T S \models \widehat{\Phi}\).

\subsection*{9.3.1 Eliminating Timing Parameters}

We first explain how intervals \(J \neq[0, \infty)\) that may appear in TCTL formulae as time bounds for path formulae are replaced by equivalent atomic clock constraints. Let TCTL \({ }_{\diamond}\) denote the set of TCTL formulae in which all intervals \(J\) are equal to \([0, \infty)\). That is, the only timing aspects that occur in \(\mathrm{TCTL}_{\diamond}\) formulae are atomic clock constraints. As such constraints can be considered as atomic propositions, in fact, TCTL \(\diamond\) is a subset of CTL. The resulting formulae provide the basis for the CTL formulae that are checked on the region transition system.

The basic idea of eliminating \(J \neq[0, \infty)\) from TCTL formula \(\Phi\) is to introduce a fresh clock, \(z\), say, that neither occurs in \(\Phi\) nor in the timed automaton under investigation,
and to enrich the formula \(\Phi\) with atomic clock constraints that may refer to \(z\). The clock \(z\) is used to measure the elapse of time until a certain property, i.e., sub formula of \(\Phi\), holds. For instance, to check the TCTL formula \(\exists \diamond^{J} \Phi\) in state \(s\), clock \(z\) is reset in state \(s\) and \(\Phi\) is checked whenever the current value of clock \(z\) lies in the interval \(J\). In order to formalize this idea, the following auxiliary notations are helpful.

\section*{Notation 9.36. Clock Evaluation \(\eta\{\ldots\}\)}

For clock evaluation \(\eta \in \operatorname{Eval}(C), z \notin C\) and \(d \in \mathbb{R}_{\geqslant 0}\), let \(\eta\{z:=d\}\) denote the clock valuation for \(C \cup\{z\}\) that extends \(\eta\) by setting \(z\) to \(d\) while keeping the value of all other clocks unchanged:
\[
\eta\{z:=d\}(x)= \begin{cases}\eta(x) & \text { if } x \in C \\ d & \text { if } x=z\end{cases}
\]

Let \(T A\) be a timed automaton over \(C\). For state \(s=\langle\ell, \eta\rangle\) in \(T S(T A)\) let \(s\{z:=d\}\) denote the state \(\langle\ell, \eta\{z:=d\}\rangle\). Note that \(s\{z:=d\}\) is a state in \(T S(T A \oplus z)\) where \(T A \oplus z\) is the timed automaton \(T A\) with the set of clocks \(C \cup\{z\}\).

The following theorem provides a recipe to transform any TCTL formula into a timing parameter-free TCTL formula.

\section*{Theorem 9.37. Elimination of Timing Parameters}

Let TA be timed automaton (Loc, Act, \(C, \hookrightarrow, \operatorname{Loc}_{0}, \operatorname{Inv}, A P, L\) ), and \(\Phi U^{J} \Psi\) a TCTL formula over \(C\) and \(A P\). For clock \(z \notin C\), let
\[
T A \oplus z=\left(\operatorname{Loc}, A c t, C \cup\{z\}, \hookrightarrow, \operatorname{Loc}_{0}, \operatorname{Inv}, A P, L\right) .
\]

For any state \(s\) of \(T S(T A)\) it holds that
\[
\begin{aligned}
& \text { 1. } s \not \models_{T C T L} \exists\left(\Phi \cup^{J} \Psi\right) \quad \text { iff } \underbrace{s\{z:=0\}}_{\text {state in TS(TA } \oplus z)} \models_{T C T L} \exists((\Phi \vee \Psi) \cup((z \in J) \wedge \Psi)) . \\
& \text { 2. } s \not \models_{T C T L} \forall\left(\Phi \cup^{J} \Psi\right) \quad \text { iff } \underbrace{s\{z:=0\}}_{\text {state in } T S(T A \oplus z)} \models_{T C T L} \forall((\Phi \vee \Psi) \cup((z \in J) \wedge \Psi)) .
\end{aligned}
\]

Proof: Since \(T A \oplus z\) just extends \(T A\) with a fresh clock \(z\) (which is not used in \(T A\) ), it follows that any path \(\pi\) in \(T S(T A)\) uniquely corresponds to a path \(\pi^{\prime}\) in \(T S(T A \oplus z)\) such that
\[
\pi \in s_{0} \xrightarrow{d_{0}} s_{1} \xrightarrow{d_{1}} s_{2} \xrightarrow{d_{2}}
\]
if and only if
\[
\pi^{\prime} \in s_{0}\{z:=0\} \stackrel{d_{0}}{\Longrightarrow} s_{1}\left\{z:=d_{0}\right\} \stackrel{d_{1}}{\Longrightarrow} s_{2}\left\{z:=d_{0}+d_{1}\right\} \stackrel{d_{2}}{\Longrightarrow} \ldots
\]

It is easy to see that \(\pi\) is time-divergent if and only if \(\pi^{\prime}\) is time-divergent. We now prove that \(\pi \models \Phi \cup^{J} \Psi\) iff \(\pi^{\prime} \models(\Phi \vee \Psi) \cup((z \in J) \wedge \Psi)\). We only consider the direction \(\Rightarrow\); the proof for the other direction is similar. Assume \(\pi \models \Phi U^{J} \Psi\). From the TCTL semantics this is equivalent to
\[
\begin{gathered}
\exists i \geqslant 0 . s_{i}+d \models \Psi \text { for some } d \in\left[0, d_{i}\right] \text { with } \sum_{k=0}^{i-1} d_{k}+d \in J \text { and } \\
\forall j \leqslant i . s_{j}+d^{\prime} \models \Phi \vee \Psi \text { for any } d^{\prime} \in\left[0, d_{j}\right] \text { with } \sum_{k=0}^{j-1} d_{k}+d^{\prime} \leqslant \sum_{k=0}^{i-1} d_{k}+d .
\end{gathered}
\]

As \(z\) is a fresh clock, this is equivalent to
\(\exists i \geqslant 0 . s_{i}^{\prime}+d \models \Psi\) for some \(d \in\left[0, d_{i}\right]\) with \(\sum_{k=0}^{i-1} d_{k}+d \in J\) and
\[
\forall j \leqslant i . s_{j}^{\prime}+d^{\prime} \models \Phi \vee \Psi \text { for any } d^{\prime} \in\left[0, d_{j}\right] \text { with } \sum_{k=0}^{j-1} d_{k}+d^{\prime} \leqslant \sum_{k=0}^{i-1} d_{k}+d .
\]
where \(s_{i}^{\prime}=s_{i}\left\{z:=\sum_{k=0}^{i-1} d_{k}\right\}\) and \(s_{j}^{\prime}=s_{j}\left\{z:=\sum_{k=0}^{j-1} d_{k}\right\}\). As clock \(z\) is never reset, the value of \(z\) in state \(s_{i}^{\prime}+d\) equals \(\sum_{k=0}^{i-1} d_{k}+d\). As this sum lies in \(J\), in the first conjunct \(\Psi\) may be strengthened by the atomic clock constraint \(z \in J\). This yields
\[
\begin{gathered}
\exists i \geqslant 0 . s_{i}^{\prime}+d \models(z \in J) \wedge \Psi \text { for some } d \in\left[0, d_{i}\right] \text { with } \underbrace{\sum_{k=0}^{i-1} d_{k}+d}_{=z} \in J \text { and } \\
\forall j \leqslant i . s_{j}^{\prime}+d^{\prime} \models \Phi \vee \Psi \text { for any } d^{\prime} \in\left[0, d_{j}\right] \text { with } \sum_{k=0}^{j-1} d_{k}+d^{\prime} \leqslant \underbrace{\sum_{k=0}^{i-1} d_{k}+d}_{=z} .
\end{gathered}
\]

The constraint \(\sum_{k=0}^{i-1} d_{k}+d \in J\) can now be omitted (as it is equivalent to \(z \in J\) ), whereas in the second part, we may weaken \(\Phi \vee \Psi\) into \((\Phi \vee \Psi) \vee(\Psi \wedge(z \in J))\) as for \(d^{\prime}=d\) and \(i=j\), \(\Psi \wedge(z \in J)\) holds. Applying the TCTL semantics yields that \(\pi^{\prime} \models(\Phi \vee \Psi) \cup((z \in J) \wedge \Psi)\).

\section*{Example 9.38. Eliminating Timing Parameters}

Let \(\Phi\) be a TCTL formula. According to the above mapping, the TCTL formula \(\exists \diamond \leqslant 2 \Phi\) is replaced by \(\exists \diamond((z \leqslant 2) \wedge \Phi)\). In a similar way, we replace: \(\exists \square \leqslant 2 \Phi=\neg \forall \diamond \leqslant 2 \neg \Phi\) by
\[
\begin{aligned}
\neg \forall \diamond((z \leqslant 2) \wedge \neg \Phi) & \equiv \exists \square(\neg(z \leqslant 2) \vee \Phi) \\
& =\exists \square((z \leqslant 2) \rightarrow \Phi) .
\end{aligned}
\]

Note that the resulting formulae are CTL formulae (or could be understood as such) provided \(\Phi\) does not contain intervals different from \([0, \infty)\).

In order to verify whether \(T A \models \Phi\) for TCTL formula \(\Phi\), the above result suggests equipping \(T A\) with a clock for each subformula of \(\Phi\) of the form \(\Psi U^{J} \Psi^{\prime}\) while replacing this subformula as indicated in Theorem 9.37. This yields \(\mathrm{TCTL}_{\diamond}\) formula \(\widehat{\Phi}\). As \(\widehat{\Phi}\) does not contain timing parameters, and any clock constraint can be considered as an atomic proposition, in fact, \(\widehat{\Phi}\) is a CTL formula! Verifying a timed CTL formula on a timed automaton TA thus reduces to checking a CTL formula on a TA extended with a clock whose sole purpose is to measure the elapse of time that is referred to in the formula.

\subsection*{9.3.2 Region Transition Systems}

Consider timed automaton \(T A\) and \(\mathrm{TCTL}_{\diamond}\) formula \(\Phi\). It is assumed that \(T A\) is equipped with an additional clock as explained in the previous section. The idea is impose an appropriate equivalence, denoted \(\cong\), on the clock valuations - and implicitly on the states of \(T S(T A)\) by letting \(\left\langle\ell^{\prime}, \eta^{\prime}\right\rangle \cong\left\langle\ell, \eta^{\prime}\right\rangle\) if \(\ell=\ell^{\prime}\) and \(\eta \cong \eta^{\prime}\)-such that:
(A) Equivalent clock valuations should satisfy the same clock constraints that occur in \(T A\) and \(\Phi\) :
\[
\eta \cong \eta^{\prime} \Rightarrow \quad\left(\eta \models g \quad \text { iff } \quad \eta^{\prime} \models g \text { for all } g \in A C C(T A) \cup A C C(\Phi)\right)
\]
where \(A C C(T A)\) and \(A C C(\Phi)\) denote the set of atomic clock constraints that occur in \(T A\) and \(\Phi\), respectively. These constraints are either of the form \(x \leqslant c\) or \(x<c\).
(B) Time-divergent paths emanating from equivalent states should be "equivalent". This property guarantees that equivalent states satisfy the same path formulae.
(C) The number of equivalence classes under \(\cong\) is finite.

In the sequel we adopt the following notation for clock values.

\section*{Notation 9.39. Integral and Fractional Part of Real Numbers}

Let \(d \in \mathbb{R}\). The integral part of \(d\) is the largest integer that is at most \(d\) :
\[
\lfloor d\rfloor=\max \{c \in \mathbb{N} \mid c \leqslant d\}
\]

The fractional part of \(d\) is defined by \(\operatorname{frac}(d)=d-\lfloor d\rfloor\). For example, \(\lfloor 17.59267\rfloor=17\), \(\operatorname{frac}(17.59267)=0.59267,\lfloor 85\rfloor=85\), and \(\operatorname{frac}(85)=0\).

The definition of clock equivalence is based on three observations that successively lead to a refined notion of equivalence. Let us discuss these observations in detail.

First observation. Consider atomic clock constraint \(g\), and let \(\eta\) be a clock valuation (both over the set \(C\) of clocks with \(x \in C)\). As \(g\) is an atomic clock constraint, \(g\) is either of the form \(x<c\) or \(x \leqslant c\) for \(c \in \mathbb{N}\). We have that \(\eta \models x<c\) whenever \(\eta(x)<c\), or equivalently, \(\lfloor\eta(x)\rfloor<c\). The fractional part of \(\eta(x)\) in this case is not relevant. Similarly, \(\eta \models x \leqslant c\) whenever either \(\lfloor\eta(x)\rfloor<c\), or \(\lfloor\eta(x)\rfloor=c\) and \(\operatorname{frac}(x)=0\). Therefore, \(\eta \models g\) only depends on the integral part \(\lfloor\eta(x)\rfloor\), and the fact whether \(\operatorname{frac}(\eta(x))=0\). This leads to the initial suggestion that clock valuations \(\eta\) and \(\eta^{\prime}\) are equivalent (denoted \(\cong_{1}\) ) whenever
\[
\begin{equation*}
\lfloor\eta(x)\rfloor=\left\lfloor\eta^{\prime}(x)\right\rfloor \quad \text { and } \quad \operatorname{frac}(\eta(x))=0 \text { iff } \operatorname{frac}\left(\eta^{\prime}(x)\right)=0 . \tag{9.1}
\end{equation*}
\]

This constraint ensures that equivalent clock valuations satisfy the clock constraint \(g\) provided \(g\) only contains atomic clock constraints of the form \(x<c\) or \(x \leqslant c\). (In case one would restrict all atomic clock constraints to be strict, i.e., of the form \(x<c\), the fractional parts would not be of importance and the second conjunct in the above equation may be omitted.) Note that it is crucial for this observation that only natural number constants are permitted in the clock constraints. This equivalence notion is rather simple, leads to a denumerable (but still infinite) number of equivalence classes, but is too coarse.

\section*{Example 9.40. A First Partitioning for Two Clocks}

To exemplify the kind of equivalence classes that one obtains, consider the set of clocks \(C=\) \(\{x, y\}\). The quotient space for \(C\) obtained by suggestion (9.1) is depicted in Figure 9.18) where the equivalence classes are
- the corner points \((q, p)\)
- the line segments \(\{(q, y) \mid p<y<p+1\}\) and \(\{(x, p) \mid q<x<q+1\}\), and
- the content of the squares \(\{(x, y) \mid q<x<q+1 \wedge p<y<p+1\}\)
where \(p, q \in \mathbb{N}\) and \(\{(x, p) \mid q<x<q+1\}\) is a shorthand for the set of all clock evaluations \(\eta\) with \(\eta(x) \in] q, q+1[\) and \(\eta(y)=p\).

Second observation. We demonstrate the fact that \(\cong_{1}\) is too coarse by means of a small example. Consider location \(\ell\) whose two outgoing transitions are guarded with \(x \geqslant 2\) (action \(\alpha\) ) and \(y>1(\operatorname{action} \beta)\), respectively; see also Figure 9.19. Let state \(s=\langle\ell, \eta\rangle\) with \(1<\eta(x)<2\) and \(0<\eta(y)<1\). Both transitions are disabled, so the only possibility is to let time advance. The transition that is enabled next depends on the ordering of the


Figure 9.18: Initial partitioning for two clocks .



Figure 9.19: Fragment of timed automaton and time passage of two clock valuations.
fractional parts of the clocks \(x\) and \(y\) : if \(\operatorname{frac}(\eta(x))<\operatorname{frac}(\eta(y))\), then \(\beta\) is enabled before \(\alpha\); if \(\operatorname{frac}(\eta(x)) \geqslant \operatorname{frac}(\eta(y))\), action \(\alpha\) is enabled first. Time-divergent paths in \(s\) may thus start with \(\alpha\) if \(\operatorname{frac}(\eta(x)) \geqslant \operatorname{frac}(\eta(y))\), and with \(\beta\) otherwise. This is represented by the fact that delaying leads to distinct successor classes depending on the ordering of the fractional parts of clock, see Figure 9.19 (right part).

Thus, besides \(\lfloor\eta(x)\rfloor\) and the fact whether \(\operatorname{frac}(\eta(x))=0\), apparently the order of the fractional parts of \(\eta(x), x \in C\) is important as well, i.e., whether for \(x, y \in C\) :
\[
\operatorname{frac}(\eta(x))<\operatorname{frac}(\eta(y)) \text { or } \operatorname{frac}(\eta(x))>\operatorname{frac}(\eta(y)) \text { or } \operatorname{frac}(\eta(x))=\operatorname{frac}(\eta(y))
\]

This suggests extending the initial proposal (9.1) for all \(x, y \in C\) by
\[
\begin{equation*}
\operatorname{frac}(\eta(x)) \leqslant \operatorname{frac}(\eta(y)) \quad \text { if and only if } \quad \operatorname{frac}\left(\eta^{\prime}(x)\right) \leqslant \operatorname{frac}\left(\eta^{\prime}(y)\right) \tag{9.2}
\end{equation*}
\]
i.e., \(\eta_{1} \cong_{2} \eta_{2}\) iff \(\eta_{1} \cong_{1} \eta_{2}\) and (9.2) holds. This strengthening will ensure that equivalent states \(\langle\ell, \eta\rangle\) and \(\left\langle\ell, \eta^{\prime}\right\rangle\) have similar time-divergent paths.


Figure 9.20: Refining the initial partitioning for two clocks.

\section*{Example 9.41. A Second Partitioning for Two Clocks}

This observation suggests to decompose the squares \(\{(x, y) \mid q<x<q+1 \wedge p<y<\) \(p+1\}\) into a line segment, an upper and lower triangle, i.e., the following three parts:
\[
\begin{aligned}
& \{(x, y) \mid q<x<q+1 \wedge p<y<p+1 \wedge x-y<q-p\}, \\
& \{(x, y) \mid q<x<q+1 \wedge p<y<p+1 \wedge x-y>q-p\}, \text { and } \\
& \{(x, y) \mid q<x<q+1 \wedge p<y<p+1 \wedge x-y=q-p\} .
\end{aligned}
\]

Figure 9.20 illustrates the resulting partitioning for two clocks.

Final observation. The above constraints on clock equivalence yield a denumerable though not finite quotient. To obtain an equivalence with a finite quotient, we exploit the fact that in order to decide whether \(T A \models \Phi\) only the clock constraints occurring in \(T A\) and \(\Phi\) are relevant. As there are only finitely many clock constraints, we can determine for each clock \(x \in C\) the maximal clock constraint, \(c_{x} \in \mathbb{N}\), say, with which \(x\) is compared in some clock constraint in either \(T A\) (as guard or location invariant) or \(\Phi\). Since \(c_{x}\) is the largest constant with which clock \(x\) is compared it follows that if \(\eta(x)>c_{x}\), the actual value of \(x\) is irrelevant. (Clock \(x\) that occurs neither in \(T A\) nor in \(\Phi\) is superfluous and can be omitted; for these clocks we set \(c_{x}=0\).) As a consequence, the constraints (9.1) are only relevant if \(\eta(x) \leqslant c_{x}\) and \(\eta^{\prime}(x) \leqslant c_{x}\), while for (9.2) in addition \(\eta(y) \leqslant c_{y}\) and \(\eta^{\prime}(y) \leqslant c_{y}\).

The above considerations suggest the following notion of clock equivalence.
\(y\)

\[
c_{y}=2 c_{x}=4
\]


Figure 9.21: Third (and final) partitioning for two clocks (for \(c_{x}=4\) and \(c_{y}=2\) ).

\section*{Definition 9.42. Clock Equivalence \(\cong\)}

Let \(T A\) be a timed automaton, \(\Phi\) a \(\mathrm{TCTL}_{\diamond}\) formula (both over set \(C\) of clocks), and \(c_{x}\) the largest constant with which \(x \in C\) is compared with in either \(T A\) or \(\Phi\). Clock valuations \(\eta, \eta^{\prime} \in \operatorname{Eval}(C)\) are clock-equivalent, denoted \(\eta \cong \eta^{\prime}\) if and only if either
- for any \(x \in C\) it holds that \(\eta(x)>c_{x}\) and \(\eta^{\prime}(x)>c_{x}\), or
- for any \(x, y \in C\) with \(\eta(x), \eta^{\prime}(x) \leqslant c_{x}\) and \(\eta(y), \eta^{\prime}(y) \leqslant c_{y}\) all the following conditions hold:
\[
\begin{aligned}
& -\lfloor\eta(x)\rfloor=\left\lfloor\eta^{\prime}(x)\right\rfloor \quad \text { and } \quad \operatorname{frac}(\eta(x))=0 \text { iff } \operatorname{frac}\left(\eta^{\prime}(x)\right)=0, \\
& -\operatorname{frac}(\eta(x)) \leqslant \operatorname{frac}(\eta(y)) \quad \text { iff } \quad \operatorname{frac}\left(\eta^{\prime}(x)\right) \leqslant \operatorname{frac}\left(\eta^{\prime}(y)\right) .
\end{aligned}
\]

As the clock equivalence \(\cong\) depends on \(T A\) and \(\Phi\), strictly speaking one should write \(\cong_{T A, \Phi}\) instead of \(\cong\). The dependency of \(\cong\) on \(T A\) and \(\Phi\) is limited to the largest constants \(c_{x}\); that is to say, neither the structure of \(T A\) nor that of \(\Phi\) is of relevance to clock equivalence. The equivalence \(\cong\) is lifted to states of the transition system \(T S(T A)\) as follows. For states \(s_{i}=\left\langle\ell_{i}, \eta_{i}\right\rangle, i=1,2\), in \(T S(T A)\) :
\[
s_{1} \cong s_{2} \quad \text { iff } \quad \ell_{1}=\ell_{2} \quad \text { and } \quad \eta_{1} \cong \eta_{2}
\]

Equivalence classes under \(\cong\) are called clock regions.

\section*{Definition 9.43. Clock and State Region}

Let \(\cong\) be a clock equivalence on \(C\). The clock region of \(\eta \in \operatorname{Eval}(C)\), denoted \([\eta]\), is defined by
\[
[\eta]=\left\{\eta^{\prime} \in \operatorname{Eval}(C) \mid \eta \cong \eta^{\prime}\right\}
\]

The state region of \(s=\langle\ell, \eta\rangle \in T S(T A)\), denoted [s], is defined by
\[
[s]=\langle\ell,[\eta]\rangle=\left\{\left\langle\ell, \eta^{\prime}\right\rangle \mid \eta^{\prime} \in[\eta]\right\}
\]

In the sequel, state and clock regions are often indicated as regions whenever it is clear from the context what is meant. Clock regions will be denoted by \(r, r^{\prime}\), and so forth. We often use casual notations to denote clock regions or clock valuations. For a timed automaton with two clocks, \(x\) and \(y\) say,
\[
\{(x, y) \mid 1<x<2,0<y<1, x-y<1\}
\]
denotes the clock region of all clock valuations \(\eta \in \operatorname{Eval}(\{x, y\})\) with
\[
1<\eta(x)<2 \quad \text { and } \quad 0<\eta(y)<1 \quad \text { and } \quad \operatorname{frac}(\eta(x))<\operatorname{frac}(\eta(y)) .
\]

\section*{Example 9.44. Light Switch}

Consider the timed automaton over \(C=\{x\}\) for the light switch and the \(\mathrm{TCTL}_{\diamond}\) formula \(\Phi=\) true. It follows that the largest constant with which \(x\) is compared is \(c_{x}=2\); this is due to the location invariant \(x \leqslant 2\).


We gradually construct the regions for this timed automaton by considering each of the constraints in Definition 9.42 separately. Clock valuations \(\eta, \eta^{\prime}\) are equivalent if \(\eta(x)\) and \(\eta^{\prime}(x)\) belong to the same equivalence class along the real line. (In general, for \(n\) clocks this amounts to considering an \(n\)-dimensional hyperspace on \(\mathbb{R}_{\geqslant 0}\).)
1. The requirement that \(\eta(x)>2\) and \(\eta^{\prime}(x)>2\) or \(\eta(x) \leqslant 2\) and \(\eta^{\prime}(x) \leqslant 2\) yields the partitioning into the intervals \([0,2]\) and \((2, \infty)\).


Figure 9.22: Clock regions for the light switch timed automaton.
2. The requirement that whenever \(\eta(x) \leqslant 2\) and \(\eta^{\prime}(x) \leqslant 2\), the integral parts of \(\eta(x)\) and \(\eta^{\prime}(x)\) agree and \(\operatorname{frac}(\eta(x))=0\) iff \(\operatorname{frac}\left(\eta^{\prime}(x)\right)=0\) yields the partitioning into the intervals
\[
[0,0],(0,1),[1,1],(1,2),[2],(2, \infty) .
\]
3. As there is only a single clock, the third constraint of Definition 9.42 trivially holds.

We thus obtain six clock regions (see Figure 9.22), and as there are two locations, twelve state regions.

\section*{Example 9.45. Two Clocks}

Consider the set of clocks \(C=\{x, y\}\) and assume \(c_{x}=2\) and \(c_{y}=1\). As in the previous example, we gradually construct the clock regions. Clock valuations \(\eta, \eta^{\prime} \in \operatorname{Eval}(\{x, y\})\) are equivalent if the real-valued pairs \((\eta(x), \eta(y))\) and \(\left(\eta^{\prime}(x), \eta^{\prime}(y)\right)\) are elements of the same clock region.
1. The requirement that \(\eta(x)>c_{x}\) and \(\eta^{\prime}(x)>c_{x}\) or \(\eta(x) \leqslant c_{x}\) and \(\eta^{\prime}(x) \leqslant c_{x}\) for any clock \(x \in C\) yields four classes:
\[
\begin{array}{ll}
\{(x, y) \mid 0 \leqslant x \leqslant 2,0 \leqslant y \leqslant 1\}, & \{(x, y) \mid 0 \leqslant x \leqslant 2, y>1\}, \\
\{(x, y) \mid x>2,0 \leqslant y \leqslant 1\}, & \text { and }\{(x, y) \mid x>2, y>1\}
\end{array}
\]
2. The second requirement of Definition 9.42 yields a refinement of the first three classes obtained in the previous step. For instance, the rectangle \([(0 \leqslant x \leqslant 2),(0 \leqslant y \leqslant 1)]\) is decomposed into the six corner points:
\[
(0,0),(0,1),(1,0),(1,1),(2,0) \text { and, }(2,1),
\]
the (open) line segments:
\[
\begin{array}{ll}
\{(0, y) \mid 0<y<1\}, & \{(1, y) \mid 0<y<1\}, \quad\{(2, y) \mid 0<y<1\}, \\
\{(x, 0) \mid 0<x<1\}, & \{(x, 0) \mid 1<x<2\}, \\
\{(x, 1) \mid 0<x<1\}, & \{(x, 1) \mid 1<x<2\},
\end{array}
\]
and the (open) squares:
\[
\{(x, y) \mid 0<x<1,0<y<1\} \quad \text { and } \quad\{(x, y) \mid 1<x<2,0<y<1\} .
\]

Similarly, \([(0 \leqslant x \leqslant 2),(y>1)]\) is decomposed into
\[
\begin{array}{lll}
\{(0, y) \mid y>1\} & \{(1, y) \mid y>1\} & \{(2, y) \mid y>1\} \\
\{(x, y) \mid 0<x<1, y>1\} & \{(x, y) \mid 1<x<2, y>1\} &
\end{array}
\]

In a similar way \(\{(x, y) \mid x>2,0 \leqslant y \leqslant 1\}\) is decomposed into three classes.
3. Finally, we apply the ordering constraint, see the third constraint of Definition 9.42 to \(\{(x, y) \mid 1<x<2,0<y<1\}\). Since the ordering of clocks now becomes important, this class is split into
\[
\begin{aligned}
& \{(x, y) \mid 1<x<2,0<y<1, \text { frac }(x)<\operatorname{frac}(y)\}, \\
& \{(x, y) \mid 1<x<2,0<y<1, \text { frac }(x)>\operatorname{frac}(y)\}, \\
& \{(x, y) \mid 1<x<2,0<y<1, x-y=1\} .
\end{aligned}
\]

A similar reasoning applies to \(\{(x, y) \mid 0<x<1,0<y<1\}\). The other classes are not further partitioned. For instance, \(\{(x, y) \mid 1<x<2, y>1\}\) is not further split as \(y>c_{y}\).

Summarizing, we obtain twenty-eight clock regions: six corner points, fourteen open line segments, four open triangles, and four open clock regions.

Even for apparently simple timed automata, a large number of regions can arise. For this reason, we abstain from indicating the regions in more complex examples such as the railroad crossing and real-time mutual exclusion examples. The number of clocks, as well as the constants \(c_{x}\), are essential factors that determine the number of regions. The number of clock regions and state regions is finite, i.e., consraint (C) holds. The following theorem contains an estimate for the number of clock regions. The number of state regions is a factor \(|L o c|\) larger.

\section*{Theorem 9.46. Number of Regions}

The number of clock regions is bounded from below and above as follows:
\[
|C|!* \prod_{x \in C} c_{x} \leqslant|E \operatorname{val}(C) / \cong| \leqslant|C|!* 2^{|C|-1} * \prod_{x \in C}\left(2 c_{x}+2\right)
\]
where for the upper bound it is assumed that \(c_{x} \geqslant 1\) for all \(x \in C\).

Proof: The lower and upper bounds are determined by considering a representation of clock regions such that there is a one-to-one relationship between the representation of a clock region and the clock region itself. This representation allows derivation of the bounds.

Let \(C\) be a set of clocks and \(\eta \in E v a l(C)\). Every clock region \(r\) can be represented by a tuple \(\langle J, \wp, D\rangle\) where \(J\) is a family of intervals, \(\wp\) is a permutation of a subset of clocks in \(C\), and \(D \subseteq C\) is a set of clocks such that
- \(J=\left(J_{x}\right)_{x \in C}\) is a family of intervals with
\[
J_{x} \in\{[0,0],] 0,1[,[1,1],] 1,2[, \ldots,] c_{x}-1, c_{x}\left[,\left[c_{x}, c_{x}\right],\right] c_{x}, \infty[ \}
\]
such that \(\eta(x) \in J_{x}\) for all clocks \(x \in C\) and clock evaluations \(\eta \in r\).
- Let \(C_{\text {open }}\) be the set of clocks \(x \in C\) such that \(J_{x}\) is an open interval, i,e,
\[
C_{\text {open }}=\left\{x \in C \mid J_{x} \in\{ ] 0,1[,] 1,2[, \ldots,] c_{x}-1, c_{x}[,] c_{x}, \infty[ \}\right\}
\]
\(\wp=\left\{x_{i_{1}}, \ldots, x_{i_{k}}\right\}\) is a permutation of \(C_{\text {open }}=\left\{x_{1}, \ldots, x_{k}\right\}\) such that for any \(\eta \in r\) the clocks are ordered according to their fractional parts, i.e.,
\[
i_{h}<i_{j} \quad \text { imples } \quad \operatorname{frac}\left(\eta\left(x_{i_{h}}\right)\right) \leqslant \operatorname{frac}\left(\eta\left(x_{i_{j}}\right)\right)
\]
- \(D \subseteq C_{\text {open }}\) contains all clocks in \(C_{\text {open }}\) such that for all clock evaluations \(\eta^{\prime} \in[\eta]\) the fractional part for clock \(x_{i_{j}}\) corresponds to the fractional part for its predecessor \(x_{i_{j-1}}\) in the permutation \(\wp\) :
\[
x_{i_{j}} \in D \text { implies } \operatorname{frac}\left(\eta\left(x_{i_{j-1}}\right)\right)=\operatorname{frac}\left(\eta\left(x_{i_{j}}\right)\right) .
\]

There is a one-to-one relation between the clock regions and triples \(\langle J, \wp, D\rangle\).
The indicated upper bound for the number of clock regions is obtained by the following combinatorial observation that there are
- exactly \(\prod_{x \in C}\left(2 c_{x}+2\right)\) different interval families \(J\),
- maximally \(\left|C_{\text {open }}\right|\) ! \(\leqslant|C|\) ! different permutations over \(C_{\text {open }}\), and
- maximally \(2^{\left|C_{\text {open }}\right|-1} \leqslant 2^{|C|-1}\) different choices for \(D \subseteq C \backslash\left\{x_{1}\right\}\).

The indicated lower bound is obtained when all clocks have a value in an open interval (though not the unbounded interval \(] c_{x}, \infty[\) ), and all have different fractional parts. In this case \(D=\varnothing\), and
\[
J_{x} \in\{ ] 0,1[,] 1,2[, \ldots,] c_{x}-1, c_{x}[ \} .
\]

As there are exactly \(\prod_{x \in C} c_{x}\) possibilities for \(J\) and maximally \(|C|\) ! different permutations, the lower bound follows.

\section*{Example 9.47. Number of Regions}

Let us illustrate the number of regions for a reasonable small timed automaton. Assume \(|C|=n\) such that \(c_{x}=2\) for all \(x \in C\). The lower bound for the number of clock regions indicated in Theorem 9.46 is \(n!\cdot 2^{n}\). The minimal number of clock regions for \(n=2\) equals 8; for \(n=3\) and \(n=4\) this rises to 48 and 384 respectively. For \(n=5\), there are at least 3840 clock regions.

\section*{Lemma 9.48.}

Let TA be a timed automaton and \(\Phi\) a \(T C T L \diamond\) formula both over the set \(C\) of clocks and \(\cong\) the clock equivalence induced by TA and \(\Phi\). Then:
1. For \(\eta, \eta^{\prime} \in \operatorname{Eval}(C)\) such that \(\eta \cong \eta^{\prime}\) :
\[
\eta \models g \quad \text { if and only if } \quad \eta^{\prime} \models g \text { for all } g \in A C C(T A) \cup A C C(\Phi)
\]
2. For \(s, s^{\prime} \in T S(T A)\) such that \(s \cong s^{\prime}\) :
\[
s \models a \quad \text { if and only if } s^{\prime} \models a \text { for any } a \in A P^{\prime} .
\]
where \(A P^{\prime}=A P \cup A C C(T A) \cup A C C(\Phi)\).

The first part of this lemma follows directly from the observations that justified the definition of clock equivalence. Using this result, the satisfaction relation of clock constraints
(see Definition 9.10) may now be used for clock regions; \([\eta] \models g\) denotes that \(\eta^{\prime} \models g\) for any \(\eta^{\prime} \in[\eta]\). As equivalent states have the same location, the second part of the lemma follows directly from the first part. All states of a state region thus satisfy the same clock constraints that occur in \(T A\) and \(\Phi\). This proves constraint (A) mentioned before.

It has been argued before that atomic clock constraints can in fact be considered as atomic propositions. Under this view, clock equivalence between states of \(T S(T A)\) is in fact a bisimulation. In the following, again let \(A P^{\prime}=A P \cup A C C(T A) \cup A C C(\Phi)\). We lift the notion of clock reset to regions as follows.

\section*{Notation 9.49. Region Reset Operator}

For \(r \in \operatorname{Eval}(C) / \cong\) and \(D \subseteq C\) let
\[
\text { reset } D \text { in } r=\{\text { reset } D \text { in } \eta \mid \eta \in r\} .
\]

Since for \(\eta, \eta^{\prime} \in \operatorname{Eval}(C)\) we have
\[
\eta \cong \eta^{\prime} \wedge D \subseteq C \Rightarrow \text { reset } D \text { in } \eta \cong \text { reset } D \text { in } \eta^{\prime},
\]
it follows that reset \(D\) in \(r \in \operatorname{Eval}(C) / \cong\). That is to say, resetting the clocks \(D\) in region \(r\) can be considered as a transition between state regions.

\section*{Theorem 9.50. Clock Equivalence is a Bisimulation}

Clock equivalence is a bisimulation equivalence over \(A P^{\prime}\).

Proof: We prove that \(\cong\) is a bisimulation (over \(A P^{\prime}\) ) by checking the conditions of a bisimulation (see Definition 7.1, page 451). Let \(s_{1}, s_{2} \in T S(T A)\) such that \(s_{1} \cong s_{2}\), that is, \(s_{1}=\left\langle\ell, \eta_{1}\right\rangle\) and \(s_{2}=\left\langle\ell, \eta_{2}\right\rangle\) such that \(\eta_{1} \cong \eta_{2}\).
1. From the second part of Lemma 9.48 (page 718), it follows that \(s_{1} \models a\) if and only if \(s_{2} \models a\) for any \(a \in A P^{\prime}\).
2. To show that any transition emanating from \(s_{1}\) can be mimicked by \(s_{2}\), distinguish between discrete and delay transitions.
(a) (Discrete transition). Assume \(\left\langle\ell, \eta_{1}\right\rangle=s_{1} \xrightarrow{\alpha} s_{1}^{\prime}=\left\langle\ell^{\prime}, \eta_{1}^{\prime}\right\rangle\). By the semantics of timed automata, this means that there is a transition \(\ell \stackrel{g: \alpha, D}{\longleftrightarrow} \ell^{\prime}\) in \(T A\) such that
\[
\eta_{1} \models g \quad \text { and } \quad \eta_{1}^{\prime}=\operatorname{reset} D \text { in } \eta_{1} \models \operatorname{Inv}\left(\ell^{\prime}\right) .
\]

Since \(\eta_{1} \cong \eta_{2}\) and \(\eta_{1} \models g\), it follows from the first part of Lemma 9.48 that \(\eta_{2} \models g\). Similarly, since \(\eta_{1} \cong \eta_{2}\), it follows that reset \(D\) in \(\eta_{1} \cong\) reset \(D\) in \(\eta_{2}\). As reset \(D\) in \(\eta_{1} \models \operatorname{Inv}\left(\ell^{\prime}\right)\) we have reset \(D\) in \(\eta_{2} \models \operatorname{Inv}\left(\ell^{\prime}\right)\). Thus:
\[
s_{2} \xrightarrow{\alpha} s_{2}^{\prime}=\left\langle\ell^{\prime}, \text { reset } D \text { in } \eta_{2}\right\rangle .
\]

As the states \(s_{1}^{\prime}\) and \(s_{2}^{\prime}\) are in the same state region, it follows that \(s_{1}^{\prime} \cong s_{2}^{\prime}\).
(b) (Delay transition). Assume \(s_{1} \xrightarrow{d} s_{1}^{\prime}=s_{1}+d\) for some \(d \in \mathbb{R}_{\geqslant 0}\). It is not difficult to see that for any \(d\) there exists \(d^{\prime}\) such that \(\eta_{1}+d \cong \eta_{2}+d^{\prime}\). From \(\eta_{1} \models \operatorname{Inv}(\ell)\) and \(\eta_{1}+d \models \operatorname{Inv}(\ell)\), it follows by Lemma 9.48 (page 718) that \(\eta_{2} \models \operatorname{Inv}(\ell)\) and \(\eta_{2}+d^{\prime} \models \operatorname{Inv}(\ell)\). But then \(s_{2} \xrightarrow{d^{\prime}} s_{2}+d^{\prime}=s_{2}^{\prime}\) and \(s_{1}^{\prime} \cong s_{2}^{\prime}\).

For transitions emanating from \(s_{2}^{\prime}\) an analogous reasoning applies.

Note that in the delay transitions, the amount of delaying is ignored. Instead, only the fact that some delay may take place is of importance. Such bisimulation is also called time abstract.

\section*{Remark 9.51. The Need for Ordering the Fractional Parts of Clocks}

For \(\eta_{1} \cong \eta_{2}\) it holds that whenever \(\eta_{1}(x), \eta_{2}(x) \leqslant c_{x}\) and \(\eta_{1}(y), \eta_{2}(y) \leqslant c_{y}\) then:
\[
\operatorname{frac}\left(\eta_{1}(x)\right) \leqslant \operatorname{frac}\left(\eta_{1}(y)\right) \quad \text { if and only if } \quad \operatorname{frac}\left(\eta_{2}(x)\right) \leqslant \operatorname{frac}\left(\eta_{2}(y)\right)
\]

Let us explain by means of a timed automaton with \(C=\{x, y\}\) and \(c_{x}=3, c_{y}=1\) that without this constraint, \(\cong\) would not be a bisimulation.

Assume for location \(\ell, \operatorname{Inv}(\ell)=y<1\). Consider state \(s_{1}=\left\langle\ell, \eta_{1}\right\rangle\) with
\[
1<\eta_{1}(x)<2, \quad 0<\eta_{1}(y)<1, \quad \eta_{1}(x)-\eta_{1}(y)>1
\]
and state \(s_{2}=\left\langle\ell, \eta_{2}\right\rangle\) with
\[
1<\eta_{2}(x)<2, \quad 0<\eta_{2}(y)<1, \quad \eta_{2}(x)-\eta_{2}(y)<1 .
\]

The only difference between \(s_{1}\) and \(s_{2}\) is the ordering of the clocks. According to the first two constraints of the definition of \(\cong\) (see Definition 9.42 on page 713 ), \(s_{1}\) and \(s_{2}\) would be equivalent. But the successor state regions of \(s_{1}\) and \(s_{2}\) after delaying are distinct.

There exists a delay transition from \(s_{1}\) to state \(s_{1}^{\prime}=\left\langle\ell, \eta_{1}^{\prime}\right\rangle\) with \(\eta_{1}^{\prime}(x)=2\) and \(\eta_{1}^{\prime}(y)<1\), i.e., the state region \(\langle\ell,[x=2, y<1]\rangle\). As clocks proceed at the same rate, clocks \(x\) and
\(y\) both advanced by an equal amount. Due to \(\operatorname{Inv}(\ell)=y<1\), any delay transition from state \(s_{2}\) yields a state in the state region:
\[
\langle\ell,\{(x, y) \mid 1<x<2,0<y<1, x-y<1\}\rangle .
\]

The state region \(\langle\ell,[x=2, y<1]\rangle\) cannot be reached. States \(s_{1}\) and \(s_{2}\) have no corresponding delay transitions, and thus are not bisimilar.

Due to the constraint on the ordering of the fractional parts of the clocks (see the third constraint in Definition 9.42), this is avoided and \(s_{1} \not \not s_{2}\).

Theorem 9.50 ensures that the partitioning given by the state regions represents a refinement of the bisimulation quotient and allows defining a quotient transition system in which any edge \(\langle\ell,[\eta]\rangle \rightarrow\left\langle\ell^{\prime},\left[\eta^{\prime}\right]\right\rangle\) is mimicked by \(\langle\ell, \eta\rangle \rightarrow\left\langle\ell^{\prime}, \eta^{\prime}\right\rangle\). States in the quotient transition system are state regions. Transitions between state regions are either delay or discrete transitions. The following example illustrates this by means of a small example.

\section*{Example 9.52. Region Transition System}

Consider the simple timed automaton in Figure 9.23 (left) and let \(\Phi=\) true. As the largest constant with which \(x\) is compared is \(2, c_{x}=2\). The region transition system is depicted in Figure 9.23 (lower part). Since there is only one location \(\ell\), in each state region the location is \(\ell\). All \(\tau\)-labeled transitions are delay transitions. There are two discrete transitions in the region automaton, both labeled with \(\alpha\), that lead to the initial state. The only state region equipped with a \(\tau\)-self-loop is called an unbounded region, as time may progress without bound while remaining in the same state region.

Let us now consider \(\Phi=\diamond(z \leqslant 2)\), i.e., \(c_{z}=2\). The region transition system for \(\Phi\) and the timed automaton of Figure 9.23 (upper part) is depicted in Figure 9.24. Note that it is in fact the region transition system before (see Figure 9.23 (lower part)) extended with two "copies" of it. These "copies" are introduced for the constraints \(x-z=2\) and \(z-x>2\). Note that the clock \(z\) is never reset. This is typical for clocks occurring in \(\Phi\) as there is no means in a formula to reset clocks.

To define the quotient transition system with respect to \(\cong\), some auxiliary notions will be introduced. A clock region is unbounded whenever all the value of any clock exceeds its maximum constant.

\section*{Definition 9.53. Unbounded Clock Region}

The clock region \(r_{\infty}=\left\{\eta \in \operatorname{Eval}(C) \mid \forall x \in C \cdot \eta(x)>c_{x}\right\}\) is unbounded.


Figure 9.23: Region transition system for a simple timed automaton with \(\Phi=\) true.


Figure 9.24: Region transition system for a simple timed automaton with \(\Phi\) with \(c_{z}=2\).

The following definition defines the successor region of a region that is obtained by delaying.

\section*{Definition 9.54. Successor Region}

Let \(r, r^{\prime} \in \operatorname{Eval}(C) / \cong\). \(r^{\prime}\) is the successor (clock) region of \(r\), denoted \(r^{\prime}=\operatorname{succ}(r)\), if either
1. \(r=r_{\infty}\) and \(r=r^{\prime}\), or
2. \(r \neq r_{\infty}, r \neq r^{\prime}\) and for all \(\eta \in r\) :
\[
\exists d \in \mathbb{R}_{>0} .\left(\eta+d \in r^{\prime} \text { and } \forall 0 \leqslant d^{\prime} \leqslant d . \eta+d^{\prime} \in r \cup r^{\prime}\right)
\]

The successor state region is defined as \(\operatorname{succ}(\langle\ell, r\rangle)=\langle\ell, \operatorname{succ}(r)\rangle\).

Stated in words, the progress of time in the unbounded region yields the unbounded region. The clock region \(r^{\prime}\) is the delay successor of \(r \neq r_{\infty}\) if any clock valuation in \(r\) can delay to a clock valuation in \(r^{\prime}\), without having the possibility of leaving the regions \(r\) and \(r^{\prime}\) at any earlier time instant. Observe that the delay successor of a region is unique; this corresponds to the deterministic nature of progressing time. For any delay, the progress of time in an unbounded region is possible. This is not applicable to the other regions: after delaying sufficiently long, the region is left. Therefore, \(\operatorname{succ}(r) \neq r\) for \(r \neq r_{\infty}\).

Note that \(\operatorname{succ}(\langle\ell, r\rangle)\) just defines the possible successor regions of a state region, and does not consider the location invariant of \(\ell\).

\section*{Example 9.55. Successor Clock Regions}

For \(C=\{x\}\) with \(c_{x}=2\), the successor region are given by
\[
\begin{array}{ll}
\operatorname{succ}(\{0\}) & =] 0,1[ \\
\operatorname{succ}(] 0,1[) & =\{1\}, \\
\operatorname{succ}(\{1\}) & =] 1,2[ \\
\operatorname{succ}(] 1,2[) & =\{2\}, \\
\operatorname{succ}(\{2\}) & =] 2, \infty\left[=r_{\infty},\right. \text { and } \\
\operatorname{succ}(] 2, \infty[) & =] 2, \infty[.
\end{array}
\]

For \(C=\{x, y\}\) with \(c_{x}=2\) and \(c_{y}=1, r_{\infty}=\{(x, y) \mid x>2, y>1\}\). The following regions are unbounded with respect to \(y\), but bounded for \(x\) :
\[
\begin{aligned}
& r_{1}=\{(x, y) \mid 0<x<1, y>1\} \quad r_{2}=\{(1, y) \mid y>1\} \\
& r_{3}=\{(x, y) \mid 1<x<2, y>1\} \quad r_{4}=\{(2, y) \mid y>1\}
\end{aligned}
\]

It follows that \(\operatorname{succ}\left(r_{i}\right)=r_{i+1}\) for \(0<i \leqslant 3\) and \(\operatorname{succ}\left(r_{4}\right)=r_{\infty}\). For the bounded clock regions, the successor regions are defined by
\[
\begin{aligned}
& \operatorname{succ}(\{(x, y) \mid 0<x<1,0<y<1, x<y\})=\{(x, 1) \mid 0<x<1\}, \\
& \operatorname{succ}(\{(x, y) \mid 0<x<1,0<y<1, x>y\})=\{(1, y) \mid 0<y<1\}, \\
& \operatorname{succ}(\{(x, y) \mid 0<x<1,0<y<1, x=y\})=\{(1,1)\}, \\
& \operatorname{succ}(\{(x, y) \mid 1<x<2,0<y<1, x<y\})=\{(x, 1) \mid 1<x<2\} \text {, } \\
& \operatorname{succ}(\{(x, y) \mid 1<x<2,0<y<1, x>y\})=\{(2, y) \mid 0<y<1\}, \\
& \operatorname{succ}(\{(x, y) \mid 1<x<2,0<y<1, x=y\})=\{(2,1)\} .
\end{aligned}
\]

Besides, \(\operatorname{succ}(\{(0,0)\})=\{(x, y) \mid 0<x<1,0<y<1, x=y\})\) and \(\operatorname{succ}(\{(2,1)\})=r_{\infty}\).

\section*{Remark 9.56. Time Passage}

The successor region \(\operatorname{succ}(r)\) denotes the region that is obtained from \(r\) by delaying for some amount. Not any delay in \(r\), however, results in \(\operatorname{succ}(r)\). A delay within a region or to some (not necessarily direct) successor region of \(\operatorname{succ}(r)\) is possible too. Delaying within a region is only possible if there is no clock that has a fixed value. This applies to regions like \(] 0,1\) ( for \(|C|=1\) ) or \(\{(x, y) \mid 0<x<1,0<y<1, x-y=0\}\) for \(C=\{x, y\}\). Delaying within e.g., the regions \(\{0\}\) or \(\{(x, 1) \mid 0<x<1\}\) is impossible-any delay in these regions implies changing region.

The passage of time over several regions implies a change from \(r\) to \(\operatorname{succ}^{n}(r)\) for \(n>1\). For example, the delay transition in a timed automaton
\[
\langle o f f, 0\rangle \xrightarrow{1.578}\langle o f f, 1.578\rangle
\]
can be decomposed into three transitions, in which any delay is made to a state of the successor region
\[
\underbrace{\langle o f f, 0\rangle}_{\text {region }\{0\}} \stackrel{ }{0.5} \underbrace{\langle o f f, 0.5\rangle}_{\text {region }] 0,1[ } \stackrel{0.5}{{ }^{0}[\underbrace{\langle o f f, 1\rangle}_{\text {region }\{1\}}} \xrightarrow{0.578} \underbrace{\langle o f f, 1.578\rangle}_{\text {region }] 1, \infty[ } .
\]

These considerations show that any time passage can be described by successor regions.

Recall that the interpretation of TCTL formulae was provided by considering time-divergent paths. The following result provides a characterization of time-convergent paths in a nonzeno timed automaton in terms of state regions. Given that the timed automaton is non-zeno, any time-convergent path contains finitely many delay transitions. That is to say, from some point on, a time-convergent path stays within a certain state region. Vice versa, a path that from some point on does not change state region, is time-convergent (except for paths that reside in the unbounded region \(r_{\infty}\) ).

\section*{Lemma 9.57. Time Convergence}

For non-zeno TA and \(\pi=s_{0} s_{1} s_{2} \ldots\) an initial, infinite path fragment in \(T S(T A)\) :
(a) if \(\pi\) is time-convergent, then there exists a state region \(\langle\ell, r\rangle\) such that for some \(j\)
\[
s_{i} \in\langle\ell, r\rangle \text { for all } i \geqslant j .
\]
(b) if there exists a state region \(\langle\ell, r\rangle\) with \(r \neq r_{\infty}\) and an index \(j\) such that
\[
s_{i} \in\langle\ell, r\rangle \text { for all } i \geqslant j,
\]
then \(\pi\) is time-convergent.

Proof:
(a) Assume \(\pi=s_{0} s_{1} s_{2} \ldots\) is a time-convergent path in \(T S(T A)\). Since \(\pi\) is timeconvergent, finitely many actions occur in \(\pi\), i.e., from some point, say from index \(k\) on, only delays occur. Thus there exists a state \(\langle\ell, \eta\rangle\) and a sequence of non-negative real numbers \(d_{k+1}, d_{k+2} \ldots\) such that
\[
s_{k}=\langle\ell, \eta\rangle \quad \text { and } \quad s_{i} \in\left\langle\ell, \eta+d_{k+1}+d_{k+2}+\ldots+d_{i}\right\rangle \text { for all } i>k .
\]

Recall that the number of clock regions is finite, and observe that apart from the self-loop at \(r_{\infty}\), cycles (of delay successors) of clocks, regions do not exist. Thus, as \(\pi\) is infinite, and from some point only delays occur in \(\pi\), some region must be visited continuously from some index on. That is, there exists a clock region \(r\) and an index \(j \geqslant k\) such that
\[
s_{i}=\left\langle\ell, \eta+d_{k+1}+d_{k+2}+\ldots+d_{i}\right\rangle \in\langle\ell, r\rangle \text { for all } i \geqslant j .
\]
(b) Assume there exists a state region \(\langle\ell, r\rangle\) with \(r \neq r_{\infty}\) and some \(j\) such that \(s_{i} \in\langle\ell, r\rangle\) for all \(i \geqslant j\). We show that \(\pi=s_{0} s_{1} \ldots\) is time-convergent by contraposition.

Assume \(\pi\) is time-divergent. That is, infinitely many delay transitions \(d_{i}\) occur in \(\pi\) and \(\sum_{i} d_{i}\) is not converging. In particular, infinitely many delay transitions occur in \(\langle\ell, r\rangle\). Thus, no evaluation \(\eta \in r\) exists with \(\eta(x)=0\) for some \(x\). Distinguish two cases.
(i) Infinitely many actions \(\alpha \in\) Act occur in \(\pi\). We show that this is impossible. Assume action \(\alpha\) occurs infinitely often in \(\pi\). Then \(\pi\) traverses some loop \(\ell \stackrel{g: \alpha, D}{\longrightarrow} \ell\) in \(T A\) infinitely often. As no evaluation \(\eta \in r\) exists with \(\eta(x)=0\) for some \(x\), it follows that \(D=\varnothing\). As \(r \models g\), the action \(\alpha\) can be executed infinitely often in every state \(\langle\ell, \eta\rangle\) of \(\langle\ell, r\rangle\), i.e., the transition system \(T S(T A)\) contains a loop \(\langle\ell, \eta\rangle \xrightarrow{\alpha}\langle\ell, \eta\rangle\) for any state \(\langle\ell, \eta\rangle\) of \(\langle\ell, r\rangle\). This, however, contradicts the assumption that \(T A\) is non-zeno.
(ii) Finitely many actions \(\alpha \in\) Act are executed along \(\pi\). Then there is an index \(j\) such that only delay transitions \(d_{j+1}, d_{j+2}, \ldots\) occur in the path fragment \(s_{j} s_{j+1} \ldots\) Let \(s_{j}=\langle\ell, \eta\rangle\). Since \(\pi\) is time-divergent, \(\sum_{i=j+1}^{\infty} d_{i}\) does not converge. That is, there exists an index \(k \geqslant j\) such that \(\sum_{i=j+1}^{k} d_{i}+\eta \in r_{\infty}\). This, however, contradicts the assumption that \(r \neq r_{\infty}\) is not left.

Lemma 9.57 suggests that in order to indicate time-divergent paths it suffices to ignore delay transitions within a region.

\section*{Definition 9.58. Region Transition System}

Let \(T A=\left(L o c, A c t, C, \hookrightarrow, L^{\prime} c_{0}, \operatorname{Inv}, A P, L\right)\) be a non-zeno timed automaton and let \(\Phi\) be a \(\mathrm{TCTL}_{\diamond}\) formula. Then the region transition system of \(T A\) for \(\Phi\) is defined as follows.
\[
R T S(T A, \Phi)=\left(S^{\prime}, \operatorname{Act} \cup\{\tau\}, \rightarrow^{\prime}, I^{\prime}, A P^{\prime}, L^{\prime}\right) \quad \text { where }
\]
- if \(S\) is the set of all states in \(T S(T A)\), then the state space of \(R T S(T A, \Phi)\) is
\[
S^{\prime}=S / \cong=\{[s] \mid s \in S\}
\]
the set of all state regions,
- \(I^{\prime}=\{[s] \mid s \in I\}\),
- \(A P^{\prime}=A C C(T A) \cup A C C(\Phi) \cup A P\),
- \(L^{\prime}(\langle\ell, r\rangle)=L(\ell) \cup\left\{g \in A P^{\prime} \backslash A P \mid r \models g\right\}\),
- the transition relation \(\rightarrow^{\prime}\) is defined by:
\[
\begin{aligned}
& \frac{\ell \stackrel{g: \alpha, D}{\hookrightarrow} \ell^{\prime} \wedge r \models g \wedge \text { reset } D \text { in } r \models \operatorname{Inv}\left(\ell^{\prime}\right)}{\langle\ell, r\rangle \xrightarrow{\alpha}\left\langle\ell^{\prime}, \text { reset } D \text { in } r\right\rangle} \quad \text { and } \\
& \frac{r \models \operatorname{Inv}(\ell) \wedge \operatorname{succ}(r) \models \operatorname{Inv}(\ell)}{\langle\ell, r\rangle \xrightarrow{\tau^{\prime}}\langle\ell, \operatorname{succ}(r)\rangle} .
\end{aligned}
\]

The dependence on formula \(\Phi\) is implicit; its maximal constants are of relevance to the clock equivalence only. In case the region transition system does not depend on (the maximal constants occurring in) \(\Phi\), we simply write \(R T S(T A)\). This applies, e.g., when \(\Phi=\) true or the constants in \(\Phi\) for clock \(x\) do not exceed the maximal constants for \(x\) in \(T A\), for any \(x\).

\section*{Example 9.59. Light Switch}

Consider the timed automaton modeling a light switch:


Let \(\Phi=\) true. Then \(c_{x}=2\), and the region transition system \(R T S(\) Switch \()=R T S(\) Switch, true \()\) is as follows, where the \(\tau\) action labels are omitted:


Although there exists a timelock in state \(\langle o n,[x>2]\rangle\), the timed automaton Switch is timelock-free, since \(\langle o n,[x>2]\rangle\) is unreachable.

Any path in the (infinite) transition system of timed automaton \(S\) witch has a corresponding path in \(R T S(\) Switch \()\). The following path in \(T S(\) Switch \()\) :
\[
\langle o f f, 0\rangle\langle o f f, 2.5\rangle\langle o f f, 2.7\rangle\langle o n, 0\rangle\langle o n, 1.456\rangle\langle o n, 1.7\rangle\langle o n, 2\rangle \ldots
\]
corresponds to the path in \(\operatorname{RTS}(T A)\) :
\[
\begin{gathered}
\langle o f f,[x=0]\rangle\langle o f f,[0<x<1]\rangle\langle o f f,[x=1]\rangle\langle o f f,[1<x<2]\rangle \\
\langle o f f,[x=2]\rangle\langle o f f,[x>2]\rangle\langle o n,[x=0]\rangle\langle o n,[0<x<1]\rangle \\
\langle o n,[x=1]\rangle\langle o n,[1<x<2]\rangle\langle o n,[x=2]\rangle \ldots
\end{gathered}
\]
(In fact, both paths are stutter trace equivalent with respect to the set of propositions \(A P^{\prime}\).) Conversely, every path in \(R T S(\) Switch ) represents a set of paths in \(T S(\) Switch ). For example, the path fragment
\[
\langle o f f,[x=0]\rangle\langle o f f,[0<x<1]\rangle\langle o n,[x=0]\rangle\langle o n,[0<x<1]\rangle \ldots
\]
in \(R T S\) (Switch) is representative for all path fragments in \(T S\) (Switch) which reside in location off for \(t\) time units with \(0<t<1\) and then change to location on; e.g.,
\begin{tabular}{llllll}
\(\langle o f f, 0\rangle\) & \(\langle o f f, 0.231\rangle\) & \(\langle o f f, 0.5788\rangle\) & \(\langle\) off, 0.98\(\rangle\) & \(\langle o n, 0\rangle\) & \(\ldots\) and \\
\(\langle o f f, 0\rangle\) & \(\langle o f f, 0.001\rangle\) & \(\langle o n, 0\rangle\) & \(\langle o n, 0.789\rangle\) & \(\langle o n, 0.79\rangle\) & \(\ldots\)
\end{tabular}

Although these path fragments change from off to on at different time instants, they are stutter trace equivalent with respect to the set \(A P^{\prime}\) of propositions.

The time-divergent path
\[
\pi=\langle o f f, 0\rangle\langle o f f, 1\rangle\langle o f f, 2\rangle\langle o f f, 3\rangle \ldots,
\]
in which the light is never switched on, finally reaches the unbounded state region \(\langle\) off, \([x\rangle\) \(2]\rangle\), in which time passage is represented by the self-loop:
\[
\langle\text { off },[x>2]\rangle \xrightarrow{\tau}\langle\text { off },[x>2]\rangle .
\]

Thus, in the region transition system, \(\pi\) is represented by the path
\[
\begin{array}{lll}
\langle\text { off, }[x=0]\rangle & \langle\text { off, }[0<x<1]\rangle & \langle\text { off, }[x=1]\rangle \\
\langle\text { off, }[1<x<2]\rangle & \langle\text { off, }[x=2]\rangle & \langle\text { off, }[x>2]\rangle
\end{array} \quad\langle\text { off, }[x>2]\rangle \ldots .
\]

Example 9.60. Region Transition System with Two Clocks
Consider the timed automaton \(T A\) :


Figure 9.25: Region transition system for example of a example timed automaton.

where \(c_{x}=c_{y}=1\). The reachable part of \(R T S(T A)\) is indicated in Figure 9.25. It is not difficult to see that the region transition system exhibits a path satisfying \(\square(y<1)\). The following result asserts that this yields \(T A \models \exists \square(y<1)\).

The following result establishes the correctness of the model-checking approach of timed automata via region transition systems.

\section*{Theorem 9.61. Correctness of TCTL Model Checking}

For non-zeno timed automaton \(T A\) and \(T C T L_{\diamond}\) formula \(\Phi\) :
\[
\underbrace{T A \models \Phi}_{\text {TCTL semantics }} \text { if and only if } \underbrace{R T S(T A, \Phi) \models \Phi}_{C T L ~ s e m a n t i c s ~} \text {. }
\]

Proof: Let TA be a non-zeno timed automaton. By structural induction over the structure of \(\Phi\), it is proven that any state subformula \(\Psi\) (of \(\Phi\) ) and any state \(s\) of \(T S(T A)\) :
\[
s \not \models_{\text {тСтL }} \Psi \quad \text { if and only if }[s] \models_{\text {СтL }} \Psi .
\]

The base cases are straightforward. For the induction step, it is assumed that the maximal state subformulae of \(\Psi\) have been replaced by atomic propositions. (This assumption is justified by the fact that CTL model checking is a recursive descent procedure over the parse tree of \(\Psi\).) Consider the proof for \(\Psi=\forall(a \cup b)\) for \(a, b \in A P^{\prime}\); the proof for the existentially quantified until-operator goes along similar lines.
- \((\Rightarrow)\) : Assume \(s \models_{\text {тстL }} \forall(a \cup b)\) and \(s \in T S(T A)\). Then \(\pi \models_{\text {тстL }} a \cup b\) for any time-divergent path \(\pi \in \operatorname{Paths}_{\operatorname{div}}(s)\). Let
\[
\pi^{\prime}=\underbrace{\left\langle\ell_{0}, r_{0}\right\rangle}_{[s]}\left\langle\ell_{1}, r_{1}\right\rangle\left\langle\ell_{2}, r_{2}\right\rangle \ldots
\]
be a path in \(R T S(T A, \forall \diamond a)\) starting in the state region \([s]\). Then there exists a corresponding path \(\pi=s_{0} s_{1} s_{2} \ldots\) in \(T S(T A)\) such that \(s_{0}=s\) and \(s_{i}=\left\langle\ell_{i}, \eta_{i}\right\rangle\) where \(\eta_{i} \in r_{i}\) for all \(i \geqslant 0\). As none of the regions \(r_{i} \neq r_{\infty}\) is equipped with a selfloop, \(\pi^{\prime}\) either traverses a cycle (of length at least two) in \(R T S(T A)\), or the self-loop at \(r_{\infty}\), infinitely often. That is, there does not exist a state region \(\left\langle\ell_{i}, r_{i}\right\rangle\) such that for some \(j\) :
\[
s_{i} \in\left\langle\ell_{i}, r_{i}\right\rangle \text { for all } i \geqslant j .
\]

By Lemma 9.57(a) (page 725), it follows that \(\pi\) is time-divergent.
Since \(s \models_{\text {тстL }} \forall(a \cup b)\), it follows that \(\pi \models a \cup b\). In particular, there exists a state \(s_{j}\) with \(s_{j} \models b\) such that \(s_{i} \models a \vee b\) for all \(i \leqslant j\). As in \(\pi\) each visited region in \(\pi^{\prime}\) is represented by a state, it follows that \(\left\langle\ell_{j}, r_{j}\right\rangle \models b\) and \(\left\langle\ell_{i}, r_{i}\right\rangle \models a \vee b\) for any \(i \leqslant j\). Hence, \(\pi^{\prime} \models_{\text {Сте }} a \cup b\).
- \((\Leftarrow)\) : Assume \([s] \models_{\text {Сте }} \forall(a \cup b)\). Let \(\pi=s_{0} s_{1} s_{2} \ldots\) be a time-divergent path in \(T S(T A)\) with \(s=s_{0}\) and
\[
s_{i}=\left\langle\ell_{i}, \eta_{i}\right\rangle, \quad i=0,1,2, \ldots
\]

We show that \(\pi \models_{\text {тстL }} \diamond a\). Assume w.l.o.g. that delay transitions in \(\pi\) are between successor regions. (Any delay transition can be decomposed into delay transitions between successor regions.) As \(\pi\) is time-convergent, it follows from Lemma 9.57(b) (page 725), that delay transitions within a state region \(\langle\ell, r\rangle\) with \(r \neq r_{\infty}\) do not occur in \(\pi\). Then
\[
\pi^{\prime}=\left[s_{0}\right]\left[s_{1}\right]\left[s_{2}\right] \ldots=\left\langle\ell_{0},\left[\eta_{0}\right]\right\rangle\left\langle\ell_{1},\left[\eta_{1}\right]\right\rangle\left\langle\ell_{2},\left[\eta_{2}\right]\right\rangle \ldots
\]
is a path in \(\operatorname{RTS}(T A, \Phi)\) starting in the state region \(\left[s_{0}\right]=[s]\). As \([s]=_{\text {СтL }} \forall(a \cup b)\) it follows that \(\pi^{\prime} \models_{\text {Сте }} a \cup b\). Thus, there exists an index \(j\) such that \(\left[s_{j}\right] \models b\) and \(\left[s_{i}\right] \vDash a\) for \(i \leqslant j\). Hence, \(s_{j} \models b\) and \(s_{i} \models a \vee b\) for any \(i \leqslant j\), and \(\pi \models_{\text {тстц }} \diamond a\). It follows that \(s \not \models_{\text {тСтL }} \forall(a \cup b)\).

The region transition system \(R T S(T A)\) provides a simple and effective criterion to check whether non-zeno timed automaton \(T A\) is timelock-free.

\section*{Theorem 9.62. Timelock Freedom}

Non-zeno timed automaton TA is timelock-free iff RTS(TA) does not have any reachable terminal states.

Proof: Essentially, the proposition follows from Lemma 9.57 (page 725).

Example 9.63. Modified Light Switch (Timelock)
Consider the timed automaton Switch \(_{\text {timelock }}\) :

which has a timelock in state \(\langle o n, \eta\rangle\) with \(\eta(x)>1\). The transition system \(R T S\left(\right.\) Switch \(\left._{\text {timelock }}\right)\) is of the following form.


The state region \(\langle o n,[x=2]\rangle\) is reachable and does not have any outgoing transitions. Thus, Switch \(_{\text {timelock }}\) is not timelock-free.

The light switch described in Example 9.12 (page 688) is timelock-free, since the associated region transition system does not contain a reachable terminal state, see Example 9.59 (page 727).

Timelock freedom can thus be checked by reachability analysis of the region transition system.

\section*{Remark 9.64. Bisimulation, Time Divergence, and Fairness}

The region transition system \(R T S(T A)\) is not bisimilar to \(T S(T A)\), since neither delay transitions within a state region nor delays over several regions are explicitly represented in \(R T S(T A) . R T S(T A)\) and \(T S(T A)\) are, however, stutter-equivalent for the set of atomic propositions \(A P^{\prime}\). This follows from the fact that delay transitions over several state regions in \(T S(T A)\) can be simulated in \(R T S(T A)\) by a sequence of delay transitions between successor regions, and that delay transitions in \(T S(T A)\) within a state region may be considered as stutter steps. Equivalent states satisfy the same \(a \in A P^{\prime}\) (see Lemma 9.48 , page 718 ) and every path \(\pi^{\prime}=r_{0} r_{1} r_{2} \ldots\) in \(R T S(T A, \Phi)\) can be lifted to a path \(\pi=s_{0} s_{1} s_{2} \ldots\) with \(s_{i} \in r_{i}\) for all \(i \geqslant 0\).

The reader may wonder whether region transition systems can be enriched with self-loops (for bounded regions), and by incorporating delay transitions that jump over successor regions. Let us discuss both enrichments. Self-loops could be added to region transition systems to mimic the delay within a region. In order to only consider time-divergent paths, fairness assumptions need to be imposed, however, to rule out infinite (time-convergent) behaviors that take a self-loop of a bounded region infinitely often. Incorporating delay transitions is more problematic, though. This is due to the fact that in order to determine the truth of a TCTL path formula in a given path, states that are not explicitly on that path may be relevant (and thus should not be skipped). For example, any path corresponding to an execution fragment of the form
\[
\langle\ell,[x=0]\rangle \underbrace{d_{1}}_{\text {delay of } 2 \text { time units }} \ldots \stackrel{d_{1}}{\longrightarrow}\langle\ell,[x=2]\rangle \longrightarrow \ldots
\]
satisfies \(\diamond(x=1)\), since the interim state \(\langle\ell,[x=1]\rangle\) (which does not explicitly occur in the path representation) is passed and satisfies the clock constraint \(x=1\). Delay transitions that take over successor regions are therefore not represented in region transition systems.

\subsection*{9.3.3 The TCTL Model-Checking Algorithm}

The previous sections have described a technique to reduce the model-checking problem \(T A \models \Phi\) for non-zeno timed automaton TA and TCTL formula \(\Phi\) to a CTL model-checking problem on the region transition system for the clocks and their maximal constants in \(T A\) and \(\Phi\). TCTL formulae have been assumed to be of the form \(\exists\left(\Phi U^{J} \Psi\right)\) with \(J \neq[0, \infty)\) and such that \(\Phi\) and \(\Psi\) do not contain any until formulae equipped with intervals different from \([0, \infty)\). Such formulae are transformed into CTL formulae using a fresh clock \(z\), say. Before explaining on how this approach can be generalized toward nested formulae of the form \(\mathrm{U}^{J}\) with \(J \neq[0, \infty)\), we summarize the approach so far by means of a small example.


Figure 9.26: Region transition system for example of a light switch timed automaton.

\section*{Example 9.65. Formulae with a Single Time Bound}

Consider the following timed automaton \(T A\) and the TCTL formula \(\Phi=\exists \diamond \leqslant 1\) on .


As a first step, \(\Phi\) is replaced by \(\widehat{\Phi}=\exists \diamond((z \leqslant 1) \wedge o n)\) and \(T A\) is equipped with an additional clock \(z\). The maximal constants for the clocks \(x\) and \(z\) are \(c_{x}=1\) and \(c_{z}=1\). The region transition system \(T S=R T S(T A \oplus z, \widehat{\Phi})\) is depicted in Figure 9.26.

The state region \(\langle o n,[x=0, z=1]\rangle \vDash(z \leqslant 1) \wedge o n\) and is reachable from the initial state region. Therefore \(T S \models_{C T L} \exists \diamond((z \leqslant 1) \wedge\) on \()\), and thus \(T A \models \exists \diamond \leqslant 1\) on.

It remains to clarify how formulae with nested time bounds, such as, e.g.,
\[
\Phi=\forall \square^{\geqslant 3} \exists \diamond^{] 1,2]} \text { on }
\]
are treated. As before, we assume a non-zeno timed automaton \(T A=(\) Loc, Act, \(C, \hookrightarrow\), \(\left.L o c_{0}, \operatorname{Inv}, A P, L\right)\). Without loss of generality, TA is assumed to be timelock-free. A simple way of treating formulae with nested time bounds is to introduce a fresh clock for each
subformula. This amounts to, e.g., transforming the example formula \(\Phi\) just above into:
\[
\left.\left.\widehat{\Phi}=\forall \square\left(\left(z_{1} \geqslant 3\right) \rightarrow \exists \diamond\left(z_{2} \in\right] 1,2\right] \wedge o n\right)\right)
\]

To check whether \(T A \models \Phi\) amounts to applying a CTL model checker to \(R T S(T A \oplus\) \(\left.\left\{z_{1}, z_{2}\right\}\right)\) and \(\widehat{\Phi}\). Although this is a straightforward approach, the number of clocks grows linearly with the number of time bounds. As the size of the region automaton is exponential in the number of clocks, it is beneficial to keep the number of (additional) clocks minimal. In the sequel it will be shown that a single additional clock suffices for TCTL formulae with arbitrary many (nested) time bounds.

As before, the timed automaton \(T A\) is equipped with a single fresh clock \(z\), say, that does not occur in TA. The principle of the model-checking algorithm is similar in spirit to that of CTL-a recursive descent procedure over the parse tree of TCTL formula \(\Phi\). The clock \(z\) is of interest only for subformulae \(\Psi\) of \(\Phi\) that contain a time bound. Once \(\operatorname{Sat}(\Psi)\) is determined, the clock \(z\) is available for checking another subformula of \(\Phi\) which contains a time bound.

Let \(T S=T S(T A \oplus z)\), the transition system of \(T A\) equipped with the additional clock \(z\) which will be used to evaluate the time-constrained formulae. The state space of TS is
\[
S_{t s}=\operatorname{Loc} \times \operatorname{Eval}(C \cup\{z\})
\]

For state \(s=\langle\ell, \eta\rangle \in S_{t s}\) and \(d \in \mathbb{R}_{\geqslant 0}\), let state \(s\{z:=d\} \in S_{t s}\) which originates from \(s\) by assigning value \(d\) to clock \(z\) and leaving all other clocks unaffected:
\[
s\{z:=d\}=\langle\ell, \eta\{z:=d\}\rangle \quad \text { where } \quad \eta\{z:=d\}(x)= \begin{cases}d & \text { if } z=x \\ \eta(x) & \text { otherwise }\end{cases}
\]

Let \(R=R T S(T A \oplus z, \Phi)\) and let \(S_{r t s}\) be the state space of \(R\). For state region \([s]=\langle\ell, r\rangle \in\) \(S_{r t s}\), let \([s]\{z:=0\}\) denote the state region \(\langle\ell, r\{z:=0\}\rangle\) in \(S_{r t s}\) where
\[
r\{z:=0\}=\{\eta\{z:=0\} \mid \eta \in r\} .
\]

Example 9.66. Regions in \(\operatorname{RTS}(T A \oplus z, \Phi)\)
The states in the region transition systems \(R=R T S(T A \oplus z, \Phi)\) and \(R T S(T A, \Phi)\) are strongly related; in fact, their only difference is the clock evaluation for clock \(z\). Let \(C=\{x\}\). The state region \(\langle\ell,[0<x<1]\rangle\) in \(\operatorname{RTS}(T A, \Phi)\) corresponds to the following state regions in \(S_{r t s}\) :
\[
\text { - }\langle\ell,[0<x<1, z=c]\rangle \text { for } c=0,1, \ldots, c_{z},
\]
- \(\left\langle\ell,\left[0<x<1, z>c_{z}\right]\right\rangle\),
- \(\langle\ell,[0<x<1, c<z<c+1, \operatorname{frac}(x) \bowtie \operatorname{frac}(z)]\rangle\) for \(c=0,1, \ldots, c_{z}\) and \(\bowtie \in\{<,>,=\}\).

In the sequel, we present an algorithm to check whether \(T A \models \Phi\) for TCTL formula \(\Phi\) that may contain several (possibly nested) time bounds. The basic idea is to exploit the CTL model-checking algorithm on the region transition system of \(T A \oplus z\) and the CTL formula \(\widehat{\Phi}\). This algorithm computes the satisfaction sets:
\[
\operatorname{Sat}_{R}(\widehat{\Phi})=\left\{[s] \in S_{r t s} \mid[s] \models \widehat{\Phi}\right\} .
\]

Due to the bottom-up nature of the CTL model-checking algorithm, on computing \(\operatorname{Sat}_{R}(\Psi)\) for subformula \(\Psi\) of \(\widehat{\Phi}\), the satisfaction sets \(\operatorname{Sat}_{R}\left(\Psi_{i}\right)\) for any subformula \(\Psi_{i}\) of \(\Psi\) have been determined and are at our disposal. Any subformula \(\Psi_{i}\) can thus be replaced by (fresh) atomic propositions \(a_{\Psi_{i}}\). Accordingly, states in the region transition system \(R\) are either labeled with atomic propositions that occur in \(T A\), atomic clock constraints in \(\widehat{\Phi}\), or propositions \(a_{\Psi_{i}}\) for any subformula \(\Psi_{i}\) of \(\Phi\). States in \(R\) may, in addition, be labeled with propositions of the form \(z \in J\) in case \(\Psi_{i} \cup^{J} \Psi_{j}\) is a subformula of \(\widehat{\Phi}\). These propositions are the atomic clock constraints of clock \(z\). The following example illustrates this.

\section*{Example 9.67. Propositions for the Region Transition System}

Consider the TCTL formula:


The set of propositions of \(R\) contains the propositions \(a\) and \(b\), the clock constraint \(x=9\), and the propositions \(a_{\Psi_{1}}\) through \(a_{\Psi_{5}}\), and \(a_{\Phi}\). In addition, the clock constraints \(z \leqslant 3\), \(z \in[2,6], z \in] 2,5[\) and \(z \geqslant 3\) are included; these clock constraints correspond to the time bounds of the subformulae of \(\Phi\).

As stated above, verifying whether \(T A \models \Phi\) boils down to applying a CTL model checker to the region transition system \(R\) and \(\widehat{\Phi}\). For subformula \(\Psi\) of \(\Phi\) we have
\[
\underbrace{s \in \operatorname{Sat}(\Psi)}_{\text {satisfaction relation in } T S(T A)} \text { iff } \underbrace{a_{\Psi} \in L_{r t s}([s])}_{\text {label in } R T S(T A, \Phi)} .
\]

Algorithm 44 (page 737) summarizes the essential steps of the TCTL model-checking algorithm. First, the region transition system for \(T A \oplus z\) and \(\Phi\) is determined. By means of a recursive descent procedure over the parse tree of \(\Phi\), the satisfaction set \(\operatorname{Sat}_{R}(\Psi)=\) \(\left\{[s] \in S_{r t s} \mid[s] \models \Psi\right\}\) is determined for each subformula \(\Psi\) of \(\Phi\). For propositional logical formula \(\Psi\), the treatment is straightforward. The interesting cases are the path formulae. Let us explain the treatment for \(\Psi=\exists\left(a \cup^{J} b\right)\); for universally quantified formulae, the procedure is similar. A CTL model checker is applied on the region transition system and the CTL formula \(\widehat{\Psi}=\exists(a \cup((z \in J) \wedge b))\). Note that \(\widehat{\Psi}\) is a CTL formula over the set of atomic propositions that occur in \(\Phi\), clock constraints in \(\Phi\), and clock constraints on \(z\) (like \(z \in J\) ). Theorems 9.37 (page 707) and 9.61 (page 729) yield
\[
s \models_{\mathrm{TCTL}} \Psi \quad \text { iff } \quad s\{z:=0\} \models_{\text {тСтL }} \widehat{\Psi} \quad \text { iff } \quad[s\{z:=0\}] \models_{\text {тСТL }} \widehat{\Psi}
\]
where \([s\{z:=0\}]\) is a state in the region transition system \(R\). Accordingly, all states \([s]\) in \(R\) for which
\[
[s\{z:=0\}] \models_{\text {тСтL }} \widehat{\Psi}
\]
are labeled with proposition \(a_{\Psi}\) once \(\operatorname{Sat}_{R}(\Psi)\) has been determined. Once all subformulae have been treated, \(T A \models \Phi\) if and only if all initial states in the region transition system \(R\) are labeled with \(a_{\Phi}\). In case \(T A\) refutes \(\Phi\), the path fragment that is returned as a counterexample by the CTL model checker for the region transition system can be returned. The same applies to the generation of witnesses.

Given the fact that \(T A \models \Phi\) can be decided by means of a CTL-like model-checking algorithm on the region transition system, we obtain:

\section*{Theorem 9.68. Time Complexity of TCTL Model Checking}

For timed automaton TA and TCTL formula \(\Phi\), the TCTL model-checking problem TA \(\models\) \(\Phi\) can be determined in time \(\mathcal{O}((N+K) \cdot|\Phi|)\), where \(N\) and \(K\) are the number of states and transitions in the region transition system \(\operatorname{RTS}(T A, \Phi)\), respectively.

The worst-case time complexity of TCTL model checking is thus linear in the size of \(\Phi\) due to the recursive descent over the parse tree of \(\Phi\)-and in the size of the region transition system. As the state-space size of the region transition system grows exponentially in the number of clocks (and maximal constants \(c_{x}\) ) (see Theorem 9.46, page 717), the time complexity of TCTL model checking is exponential in the number of clocks. Although this does not affect the worst-case time complexity, it provides ample means to significantly reduce the number of states in the transition system (like the region transition system) that symbolically represent the behavior of the timed automaton.

We state here without proof that the verification problem for real-time systems is computationally hard, as even for TCTL there is a PSPACE lower bound.
```

Algorithm 44 TCTL model checking (basic idea)

```

Input: non-zeno, timelock-free timed automaton \(T A\) and TCTL formula \(\Phi\) Output: "yes" if \(T A \models \Phi\), "no" otherwise.
```

    \(R:=R T S(T A \oplus z, \Phi) ; \quad\) (* with state space \(S_{r t s}\) and labeling \(\left.L_{r t s}{ }^{*}\right)\)
    for all \(i \leqslant|\Phi|\) do
        for all \(\Psi \in \operatorname{Sub}(\Phi)\) with \(|\Psi|=i\) do
        switch \((\Psi)\) :
            true \(\quad: \quad \operatorname{Sat}_{R}(\Psi):=S_{\text {rts }} ;\)
            \(a \quad: \quad \operatorname{Sat}_{R}(\Psi):=\left\{s \in S_{r t s} \mid a \in L_{r t s}(s)\right\} ;\)
            \(\Psi_{1} \wedge \Psi_{2} \quad: \quad \operatorname{Sat}_{R}(\Psi):=\left\{s \in S_{r t s} \mid\left\{a_{\Psi_{1}}, a_{\Psi_{2}}\right\} \subseteq L_{r t s}(s)\right\} ;\)
                \(\neg \Psi^{\prime} \quad: \quad \operatorname{Sat}_{R}(\Psi):=\left\{s \in S_{r t s} \mid a_{\Psi^{\prime}} \notin L_{r t s}(s)\right\} ;\)
                \(\exists\left(\Psi_{1} \cup^{J} \Psi_{2}\right) \quad: \quad \operatorname{Sat}_{R}(\Psi):=\operatorname{Sat}_{\text {Сте }}\left(\exists\left(\left(a_{\Psi_{1}} \vee a_{\Psi_{2}}\right) \cup\left((z \in J) \wedge a_{\Psi_{2}}\right)\right)\right) ;\)
                \(\forall\left(\Psi_{1} \cup^{J} \Psi_{2}\right) \quad: \quad \operatorname{Sat}_{R}(\Psi):=\operatorname{Sat}_{\mathrm{CTL}}\left(\forall\left(\left(a_{\Psi_{1}} \vee a_{\Psi_{2}}\right) \cup\left((z \in J) \wedge a_{\Psi_{2}}\right)\right)\right) ;\)
            end switch
    ```
                    (* add \(a_{\Psi}\) to the labeling of all state regions where \(\Psi\) holds *)
            forall \(s \in S_{r t s}\) with \(s\{z:=0\} \in \operatorname{Sat}_{R}(\Psi)\) do \(L_{r t s}(s):=L_{r t s}(s) \cup\left\{a_{\Psi}\right\}\) od;
        od
    od
    if \(I_{\text {rts }} \subseteq \operatorname{Sat}_{R}(\Phi)\) then return "yes" else return "no" fi

\section*{Theorem 9.69. Complexity of TCTL Model Checking}

The TCTL model-checking problem is PSPACE-complete.

The proof of the PSPACE-completeness is not provided here; it can be found in [9].

\subsection*{9.4 Summary}
- Timed automata are program graphs in which clock variables are used to measure the elapse of time. Guards on edges determine when an edge may be taken, whereas (location) invariants describe how long the system may reside in a location.
- A timed automaton describes an infinite transition system, and thus is a model to finitely model a continuous-time system.
- A path of a timed automaton is time-divergent if time may pass without bound. A timed automaton is timelock-free whenever, from all states in its transition system, at least one time-divergent path can start. That is, all its states satisfy \(\exists \square\) true (under the TCTL semantics).
- A path of a timed automaton is zeno if it is not time-divergent and infinitely many actions are performed on it. A non-zeno timed automaton does not have initial zeno paths. A sufficient criterion for non-zenoness is that in all control cycles at least a single time unit elapses.
- Timed CTL is a real-time variant of CTL in which clock constraints may act as atomic propositions and the until operator is equipped with a time interval. The semantics of universal and existential path formulae is defined in terms of timedivergent paths. This treatment is similar to fair CTL where only fair paths are considered.
- Model-checking whether timed automaton TA satisfies TCTL formula \(\Phi\) can be reduced to checking a CTL formula derived from \(\Phi\) on a finite transition system (determined by both \(T A\) and \(\Phi\) ), the so-called region transition system. The key to this concept is a clock equivalence relation, a bisimulation relation on propositions and clock constraints.
- The size of the region transition system is exponential in the number of clocks and the constants with which clocks are compared.
- A timed automaton is timelock-free whenever its region transition system does not have reachable terminal states.

\subsection*{9.5 Bibliographic Notes}

Timed automata. The distinction between instantaneous activities (like changing a state) and delays originates from real-time process algebras such as timed CSP [366], timed CCS [427], and ATP [315]. The resulting two-phase behavior in which discrete phases - a state change due to some activity - and continuous phases - passage of time - alternate has been adopted in timed automata. Timed automata have been introduced by Alur and Dill [10] (the journal version is published as [11]), Dill [131] and H. R. Lewis [271]. Henzinger et al. [200] introduced the idea of invariants and called this safety timed automata. Extensions of timed automata include, e.g., timed automata with deadlines (as opposed to invariants) by Bornot and Sifakis [58], timed automata with drifting clocks - associating an interval to each clock that specifies the relative speed with respect to an exact reference clock - by Olivero, Sifakis, and Yovine [316], and hybrid automata for describing a mixture of discrete and continous behavior of a more general nature by Maler, Manna, and Pnueli [281]. D'Argenio and Brinksma [111] defined a process algebra for describing safety timed automata.

Model-checking timed automata. The decidability of model-checking timed CTL has been shown by Alur, Courcoubetis, and Dill [9], using clock equivalence and region transition systems. The symbolic manipulation of timed automata by means of sets of linear inequations over pairs of clocks has been brought up by Henzinger et. al. [200]. Bouyer has shown that symbolic backward reachability in the setting where timed automata are equipped with clock difference constraints (i.e., \(x-y \prec c\) ) calls for a special treatment [61]. Survey papers on model-checking timed automata have been provided by Bengtsson and Yi [43] and Yovine [430]. Time-abstract bisimulation originates from Larsen and Yi [265] who studied this equivalence in the context of a real-time process algebra. Decidability of time-abstract bisimulation has been shown by Cerans [78]. Tripakis and Yovine [394] introduced the use of time-abstract bisimulation for timed automata verification. In the journal paper [395], these authors describe linear-time (using timed Büchi automata) and branching-time (timed CTL) verification algorithms based on time-abstract bisimulation. They also present a quotienting algorithm with respect to time-abstract bisimulation. Difference bound matrices have been proposed for timed automata by Dill [131]. Berthomieu and Menasche [48] have used these data structures for the analysis of time Petri nets and Bellman [40] exploited these structures for constraint graphs. Larsen et al. [263] introduced a symbolic data structure akin to binary decision diagrams for dealing with zones in a compact manner. The edited volume [294] contains an overview of techniques and tools for the verification of real-time systems.

Linear-time properties. Timed automata with final locations accept timed languages, i.e., sets of infinite sequences of pairs consisting of a symbol and a real value representing its time of occurrence. Alur and Dill [11] proved that the class of languages accepted
by timed automata is not closed under complementation, and hence no simple logical characterization of this class exists. Alur and Dill [12] present an algorithm to check the emptiness of the language accepted by a timed automaton. Alur, Fix, and Henzinger [13] proposed event-recording automata which are closed under all Boolean operations, including complementation. An event-recording automaton is a timed automaton that contains for every event (i.e., action) a clock that records the time of the last occurrence of the event. Asarin, Caspi, and Maler [20] defined timed regular expressions and proved that, à la Kleene's theorem for finite automata, its expressive power is equivalent to timed automata (without location invariants). Intersection and renaming are essential operators for timed regular expressions to obtain this result. As in classical automata theory, the construction of automata from expressions is rather straightforward, but the reverse direction is much more involved. The authors extend their results also to timed \(\omega\)-regular expressions.

Model checkers for timed automata. Behrmann, Larsen, Petterson, and Yi and colleagues have developed the model checker UppaAL [35] since the mid-nineties. It supports a safety fragment of timed CTL, and allows bounded liveness properties to be checked using test automata. Around the same time, Yovine et al. developed the model checker Kronos [429] which supports timed CTL. During recent years, UppaAL has been extended with branch-and-bound algorithms for dealing with priced timed automata where locations are equipped with a cost rate [353]. Wang [417] has developed a timed automata model checker using a symbolic data structure akin to binary decision diagrams.

\subsection*{9.6 Exercises}

Exercise 9.1. Consider the timed automaton LightSwitch illustrated below.


Questions:
(a) Determine the transition system \(T S(\) LightSwitch ).
(b) Determine the region transition system \(R T S\) (LightSwitch, true).
(c) Check whether LightSwitch is timelock-free and non-zeno.

Exercise 9.2. Consider the following two timed automata:

(a)

(b)

As these timed automata have a single location only, the state of these timed automata can be considered as just a point in the real plane. A point \((d, e)\) (with \(d, e \geqslant 0)\) thus represents that clock \(x\) has value \(d\) and clock \(y\) has value \(e\). Determine the reachable state space of each of these timed automata. Justify your answers.

Exercise 9.3. (Modeling exercise). A control system must ensure the safe and correct functioning of a set of traffic lights at a T-junction between a major and a minor road. The lights will be set on green on the major road and red on the minor road unless a vehicle is detected by a sensor in the road just before the lights on the minor road. In this case the lights will be switchable in the standard manner and allow traffic to leave the minor road. After a suitable interval the lights will revert to their default position to allow traffic to flow on the major road again. Once a vehicle is detected the sensor will be disabled until the minor-road lights are set to red again. A sketch of the T-junction is provided below.


Questions:
1. First we ignore all timing issues involved and concentrate on the qualitative aspects of the behavior of the traffic lights. Model the above system as a network of (timed) automata. For convenience, you may assume that the two major-road lights are fully synchronized and can be modeled as a single light. Complement your system model by adding a process that regulates the arrival of cars in the minor road.
2. Adapt your model so as to incorporate the following timing constraints. Deal with each timing constraint separately so as reduce the complexity. Indicate for each timing constraint the necessary adaptations to your un-timed model:
(a) a minor-road light stays on green for 30 seconds,
(b) all interim lights stay on for 5 seconds,
(c) there is a 1 second delay between switching one light off and another on (e.g., switching from red to yellow),
(d) the major-road lights must be on green for at least 30 seconds in each cycle,
(e) (more involved) but must respond to the sensor immediately after that.
3. We extend the T-junction in the following way. Suppose there is a pedestrian crossing a short distance down the minor road but beyond the sensor. There is a button on each side of the road for pedestrians to indicate they wish to cross. The crossing should only allow people to cross when the "minor lights" are set to red in order to minimize waiting times for traffic on the minor road. The new situation is sketched below.


Extend your timed model of the previous question in order to cope with this new situation.
4. Does the crossing indeed only allow pedestrians to cross when the "minor lights" are set to red?
(This exercise is inspired by [151].)

Exercise 9.4. Consider the following six timed automata:

(a)

(b)

(c)

(d)

(e)

(f)

Give for each individual timed automaton a TCTL formula that distinguishes this timed automaton from all other ones. It is allowed to use only the atomic propositions \(a, b\), and \(c\) and clock constraints in the TCTL formulae. Location identifiers are not allowed. (This exercise is due to Pedro D'Argenio.)

Exercise 9.5. Given the following timed automaton TA: \(x>1\)


Questions:
(a) Determine the transition system \(T S(T A)\).
(b) Determine the set of states \(\operatorname{Sat}(\exists \diamond<4 a)\).
(c) Determine the region transition system \(\operatorname{RTS}(T A\), true).

\section*{Chapter 10}

\section*{Probabilistic Systems}

Whereas model-checking techniques focus on the absolute guarantee of correctness "it is impossible that the system fails" - in practice such rigid notions are hard, or even impossible, to guarantee. Instead, systems are subject to various phenomena of a stochastic nature, such as message loss or garbling and the like, and correctness - "with \(99 \%\) chance the system will not fail" - is becoming less absolute. This chapter considers the automated verification of probabilistic systems, i.e., systems that exhibit probabilistic aspects \({ }^{1}\). Probabilistic aspects are essential for, among others:
- Randomized algorithms. Typical examples are distributed algorithms like leader election or consensus algorithms where coin-tossing experiments are used to break the "symmetry" between the processes such that, e.g., consensus will eventually be reached with probability 1.
- Modeling unreliable and unpredictable system behavior. Phenomena like message loss, processor failure, and the like may be modeled by nondeterminism. This is often appropriate in early system design phases where systems are considered at a high level of abstraction and where information about the likelihood is (sometimes deliberately) left unspecified. In later design stages, though, where the internal system characteristics become more dominant, probabilities are a useful vehicle to quantify and thus refine this information.
- Model-based performance evaluation. As performance evaluation is aimed at forecasting system performance and dependability, probabilistic information - what is

\footnotetext{
\({ }^{1}\) Note: verifying probabilistic systems should not be confused with probabilistic verification, a modelchecking technique that is based on a partial state-space exploration.
}
the distribution of the message transmission delay or what is the failure rate of a processor? - needs to be present in order to evaluate quantitative properties like waiting time, queue length, time between failure, and so on.

In order to model random phenomena, transition systems are enriched with probabilities. This can be done in different ways. In discrete-time Markov chains (MCs), all choices are probabilistic. Markov chains are the most popular operational model for the evaluation of performance and dependability of information-processing systems. Roughly speaking, Markov chains are transition systems with probability distributions for the successors of each state. That is, instead of a nondeterministic choice, the next state is chosen probabilistically. Markov chains are not appropriate for modeling randomized distributed systems, since they cannot model the interleaving behavior of the concurrent processes in an adequate manner. For this purpose, Markov decision processes (MDPs) are used. In MDPs, both nondeterministic and probabilistic choices coexist. Put in a nutshell, MDPs are transition systems in which in any state a nondeterministic choice between probability distributions exists. Once a probability distribution has been chosen nondeterministically, the next state is selected in a probabilistic manner-as in MCs. Any MC is thus an MDP in which in any state the probability distribution is uniquely determined. Randomized distributed algorithms are typically appropriately modeled by MDPs, as probabilities affect just a small part of the algorithm and nondeterminism is used to model concurrency between processes by means of interleaving.

The verification of probabilistic systems can be focused on either quantitative properties or qualitative properties (or both). Quantitative properties typically put constraints on the probability or expectation of certain events. Instances of quantitative properties are, e.g., the requirement that the probability for delivering a message within the next \(t\) time units is at least 0.98 , or that the expected number of unsuccessful attempts to find a leader in a concurrent system is at most seven. Qualitative properties, on the other hand, typically assert that a certain (good) event will happen almost surely, i.e., with probability 1, or dually, that a certain (bad) event almost never occurs, i.e., with zero probability. Typical qualitative properties for Markov models are reachability, persistence (does eventually an event always hold?), and repeated reachability (can certain states be repeatedly reached?).

The purpose of this chapter is to present the main verification principles for qualitative and quantitative properties of MCs and MDPs. These range from techniques to analyze reachability, persistence, and repeated reachability properties, to model-checking algorithms for a probabilistic variant of CTL, called Probabilistic Computation Tree Logic, PCTL for short. This logic is appropriate for expressing a large class of properties in a rather elegant manner. For instance, the property "eventually a leader will be elected with probability at least \(\frac{4}{5}\) " for a randomized leader election protocol is expressed in PCTL by
\[
\mathbb{P}_{\geqslant 0.8}(\diamond \text { leader }) .
\]

Alternatively,
\[
\mathbb{P}_{\leqslant 0.015}(\neg c . e m p t y ~ U \leqslant 6 \text { c.full })
\]
asserts that the probability of a channel \(c\) being fully occupied within the next six steps, while in all intermediate configurations \(c\) is nonempty, is bounded above by 0.015 .

Apart from branching-time properties in PCTL, this chapter also covers linear-time properties. As opposed to PCTL where probabilities are expressed syntactically by means of the \(\mathbb{P}\) operator, in the linear-time setting probabilistic concepts only appear on the semantic level. That is, in the probabilistic linear time setting, LTL formulae serve to specify the desired or bad behaviors and the aim is to establish the probability that a given LTL formula holds. Thus, the satisfaction relation over states is no longer Boolean-a formula holds for a state or not-but assigns probability values to states. This chapter deals with the verification of linear-time properties such as regular safety properties and \(\omega\)-regular safety properties.

In line with the rest of this monograph, we adopt a state-based view of probabilistic models. This means that Markov chains and Markov decision processes are treated as variants of directed graphs (read: transition systems) where the edges (i.e., transitions) are augmented with randomness information. This is in contrast to many textbooks where MCs are defined as sequences of random variables. The state-based approach with atomic propositions for the states appears to be more natural, when viewing MCs and MDPs as operational models for reactive systems and structures for temporal logics (as we do). As compositional approaches for Markov models are outside the scope of this monograph, actions are irrelevant in this chapter and are therefore omitted.

\subsection*{10.1 Markov Chains}

Markov chains behave as transition systems with the only difference that nondeterministic choices among successor states are replaced by probabilistic ones. That is to say, the successor state of state \(s\), say, is chosen according to a probability distribution. This probability distribution only depends on the current state \(s\), and not on, e.g., the path fragment that led to state \(s\) from some initial state. Accordingly, the system evolution does not depend on the history (i.e., the path fragment that has been executed so far), but only on the current state \(s\). This is known as the memoryless property.

\section*{Definition 10.1. (Discrete-Time) Markov Chain (MC)}

A (discrete-time) Markov chain is a tuple \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) where
- \(S\) is a countable, nonempty set of states,
- \(\mathbf{P}: S \times S \rightarrow[0,1]\) is the transition probability function such that for all states \(s\) :
\[
\sum_{s^{\prime} \in S} \mathbf{P}\left(s, s^{\prime}\right)=1
\]
- \(\iota_{\text {init }}: S \rightarrow[0,1]\) is the initial distribution, such that \(\sum_{s \in S} \iota_{\text {init }}(s)=1\), and
- AP is a set of atomic propositions and \(L: S \rightarrow 2^{A P}\) a labeling function.
\(\mathcal{M}\) is called finite if \(S\) and \(A P\) are finite. For finite \(\mathcal{M}\), the size of \(\mathcal{M}\), denoted size \((\mathcal{M})\), is the number of states plus the number of pairs \(\left(s, s^{\prime}\right) \in S \times S\) with \(\mathbf{P}\left(s, s^{\prime}\right)>0\).

The transition probability function \(\mathbf{P}\) specifies for each state \(s\) the probability \(\mathbf{P}\left(s, s^{\prime}\right)\) of moving from \(s\) to \(s^{\prime}\) in one step, i.e., by a single transition. The constraint imposed on \(\mathbf{P}\) ensures that \(\mathbf{P}\) is a distribution. For the mathematical treatment of Markov chains, it is irrelevant whether the nonzero transition probabilities are rational or not. However, for algorithmic purposes the values \(\mathbf{P}\left(s, s^{\prime}\right)\) are supposed to be rational for all states \(s, s^{\prime} \in S\).

The value \(\iota_{\text {init }}(s)\) specifies the probability that the system evolution starts in state \(s\). The states \(s\) with \(\iota_{\text {init }}(s)>0\) are considered as the initial states. In a similar way, the states \(s^{\prime}\) for which \(\mathbf{P}\left(s, s^{\prime}\right)>0\) are viewed as the possible successors of \(s\). For state \(s\) and \(T \subseteq S\), let \(\mathbf{P}(s, T)\) denote the probability of moving from \(s\) to some state \(t \in T\) in a single step. That is,
\[
\mathbf{P}(s, T)=\sum_{t \in T} \mathbf{P}(s, t)
\]

In the sequel, we often identify the transition probability function \(\mathbf{P}: S \times S \rightarrow[0,1]\) with the matrix \((\mathbf{P}(s, t))_{s, t \in S}\). The row \(\mathbf{P}(s, \cdot)\) for state \(s\) in this matrix contains the probabilities of moving from \(s\) to its successors, while the column \(\mathbf{P}(\cdot, s)\) for state \(s\) specifies the probabilities of entering state \(s\) from any other state. Similarly, the initial distribution \(\iota_{\text {init }}\) is often viewed as a vector \(\left(\iota_{\text {init }}(s)\right)_{s \in S}\).

The use of atomic propositions and the labeling function \(L\) is the same as for transition systems. In the remainder of this chapter, we will often treat the state names as atomic propositions, i.e., \(A P=S\) and \(L(s)=\{s\}\).

A Markov chain induces an underlying digraph where states act as vertices and there is an edge from \(s\) to \(s^{\prime}\) if and only if \(\mathbf{P}\left(s, s^{\prime}\right)>0\). Paths in Markov chains are maximal (i.e.,
infinite) paths in the underlying digraph. They are defined as infinite state sequences \(\pi=s_{0} s_{1} s_{2} \cdots \in S^{\omega}\) such that \(\mathbf{P}\left(s_{i}, s_{i+1}\right)>0\) for all \(i \geqslant 0\). For path \(\pi \operatorname{in} \mathcal{M}, \inf (\pi)\) denotes the set of states that are visited infinitely often in \(\pi\). For finite Markov chains, \(\inf (\pi)\) is nonempty for all paths \(\pi\).

Markov chains are depicted by their underlying digraph where edges are equipped with the transition probabilities in \(] 0,1]\). If a state \(s\) has a unique successor \(s^{\prime}\), i.e., \(\mathbf{P}\left(s, s^{\prime}\right)=1\), the transition probability may be omitted.

\section*{Example 10.2. A Simple Communication Protocol}

Consider a simple communication protocol operating with a channel. It is error-prone in the sense that messages may be lost, see the Markov chain depicted in Figure 10.1. Here, \(\iota_{\text {init }}(s t a r t)=1\) and \(\iota_{\text {init }}(s)=0\) for \(s \neq s t a r t\), i.e., start is the unique initial state. In the state start, a message is generated that is senf off along the channel in its unique successor state try. The message is lost with probability \(\frac{1}{10}\), in which case the message will be sent off again, until it is eventually delivered. As soon as the message has been delivered correctly, the system returns to its initial state.


Figure 10.1: Markov chain for a simple communication protocol.
Using the enumeration start, try, lost, delivered for the states, the transition probability function \(\mathbf{P}\) viewed as a \(4 \times 4\) matrix and the initial distribution viewed as a column vector are
\[
\mathbf{P}=\left(\begin{array}{cccc}
0 & 1 & 0 & 0 \\
0 & 0 & \frac{1}{10} & \frac{9}{10} \\
0 & 1 & 0 & 0 \\
1 & 0 & 0 & 0
\end{array}\right) \quad \iota_{\mathrm{init}}=\left(\begin{array}{l}
1 \\
0 \\
0 \\
0
\end{array}\right)
\]

An example of a path is
\[
\pi=(\text { start try lost try lost try delivered })^{\omega} .
\]

Along this path each message has to be retransmitted two times before delivery. It follows that \(\inf (\pi)=S\). For \(T=\{\) lost, delivered \(\}\), we have \(\mathbf{P}(\) try,\(T)=1\).

\section*{Example 10.3. Simulating a Die by a Fair Coin}

Consider simulating the behavior of a standard six-sided die by a fair coin, as originally proposed by Knuth and Yao [242], see the Markov chain depicted in Figure 10.2.


Figure 10.2: Markov chain for simulating a die by a fair coin.

The computation starts in the initial state \(s_{0}\), i.e., we have \(\iota_{\text {init }}\left(s_{0}\right)=1\) and \(\iota_{\text {init }}(s)=0\) for all states \(s \neq s_{0}\). The states \(1,2,3,4,5\), and 6 at the bottom stand for the possible die outcomes. Each inner node stands for tossing a fair coin. If the outcome is heads, the left branch determines the next state; if the outcome is tails, the right branch determines the next state.

If the coin-tossing experiment in state \(s_{0}\) yields heads, the system moves to state \(s_{1,2,3}\). Tossing the coin again leads with equal probability to either state \(s_{2,3}\) (from which the die-outcomes 2 or 3 are possible with equal probability) or to state \(s_{1,2,3}^{\prime}\). From the latter state, a coin flipping yields with probability \(\frac{1}{2}\) the outcome 1 , or with probability \(\frac{1}{2}\) a return to state \(s_{1,2,3}\). The behavior for outcome tails in the initial state is symmetric. We will establish later that, in fact, this Markov chain indeed adequately models a die, i.e., the outcomes are equally likely.

\section*{Example 10.4. The Craps Gambling Game}

The game craps is based on betting on the outcome of the roll of two dice. The outcome of the first roll-the "come-out" roll-determines whether there is a need for any further rolls. On outcome 7 or 11 , the game is over and the player wins. The outcomes 2,3 , or 12 , however, are "craps"; the player loses. On any other outcome, the dice are rolled again, but the outcome of the come-out roll is remembered (the "point"). If the next roll yields

7 or the point, the game is over. On 7, the player loses, on point the player wins. In any other case, the dice are rolled until eventually either 7 or the point is obtained. Figure 10.3 depicts the Markov chain that describes the behavior of the craps game. State start is the unique initial state. We have \(\mathbf{P}\) (start, won \()=\frac{2}{9}\) as there are eight combinations of the come-out roll that are successful: \((1,6),(2,5),(3,4),(5,6)\) and all symmetric counterparts. The other transition probabilities are determined in a similar way. The self-loops at the states \(4,5,6,8,9\), and 10 model the rerolling of the dice. For \(T=\{4,5,6\}\), we have \(\mathbf{P}\left(s_{0}, T\right)=\frac{1}{3}\).


Figure 10.3: Markov chain for the behavior of the craps game.

\section*{Example 10.5. Zeroconf Protocol}

The \(\operatorname{IPv} 4\) zeroconf protocol is designed for a home local network of appliances (microwave oven, laptop, VCR, DVD player, etc.) each of which is supplied with a network interface to enable mutual communication. Such adhoc networks must be hot-pluggable and selfconfiguring. Among others, this means that when a new appliance (interface) is connected to the network, it must be configured with a unique IP address automatically. The zeroconf protocol solves this task in the following way. A host that needs to be configured randomly selects an IP address, \(U\) say, out of the 65,024 available addresses and broadcasts a message (called probe) saying "Who is using the address U?" If the probe is received by a host that is already using the address \(U\), it replies by a message indicating that \(U\) is in use. After receiving this message the host to be configured will restart: it randomly selects a new address, broadcasts a probe, etc.

Due to message loss or a busy host, a probe or a reply message may not arrive at some (or all) other hosts. In order to increase the reliability of the protocol, a host is required


Figure 10.4: Markov chain of the \(\operatorname{IPv} 4\) zeroconf protocol (for \(n=4\) probes).
to send \(n\) probes, each followed by a listening period of \(r\) time units. Therefore, the host can start using the selected IP address only after \(n\) probes have been sent and no reply has been received during \(n \cdot r\) time units. Note that after running the protocol a host may still end up using an IP address already in use by another host, e.g., because all probes were lost. This situation, called address collision, is highly undesirable since it may force a host to kill active TCP/IP connections.

The protocol behavior of a single host is modeled by a Markov chain consisting of \(n+5\) states (see Figure 10.4 for \(n=4\) ) where \(n\) is the maximal number of probes needed (as above). The initial state is \(s_{0}\) (labeled start). In state \(s_{n+4}\) (labeled \(o k\) ) the host finally ends up with an unused IP address; in state \(s_{n+2}\) (labeled error) it ends up with an address that is already in use, i.e., an address collision. State \(s_{i}(0<i \leqslant n)\) is reached after issuing the \(i\) th probe. In state \(s_{0}\) the host randomly chooses an IP address. With probability \(q=m / 65024\), where \(m\) is the number of hosts in the network when connecting the host to the network, this address is already in use. With probability \(1-q\) the host chooses an unused address and ends up in state \(s_{n+3}\). Then it issues \(n-1\) probes and waits \(n \cdot r\) time units before using this address. (The sending of these probes and the waiting time are abstracted from in the MC.) If the chosen IP address is already in use, state \(s_{1}\) is reached. Now two situations are possible. With probability \(p\), no reply is received during \(r\) time units (as either the probe or its reply has been lost), and a next probe is sent, resulting in state \(s_{2}\). If, however, a reply has arrived in time, the host returns to the initial state and restarts the protocol. The behavior in state \(s_{i}(2 \leqslant i<n)\) is similar. If in state \(s_{n}\), however, no reply has received within \(r\) time units after sending the \(n\)th probe, an address collision occurs.

We adopt the notions of direct successor and direct predecessor from transition systems. Let \(\operatorname{Paths}(\mathcal{M})\) denote the set of paths in \(\mathcal{M}\), and \(\operatorname{Paths}_{f i n}(\mathcal{M})\) denote the set of finite
path fragments \(s_{0} s_{1} \ldots s_{n}\) where \(n \geqslant 0\) and \(\mathbf{P}\left(s_{i}, s_{i+1}\right)>0\) for \(0 \leqslant i<n\). Paths(s) denotes the set of all paths in \(\mathcal{M}\) that start in state \(s\). Similarly, Paths \(_{f i n}(s)\) denotes the set of all path fragments \(s_{0} s_{1} \ldots s_{n}\) such that \(s_{0}=s\). The set of direct successors and direct predecessors are defined as follows. Let Post \((s)\) denote the set of successors of \(s\), i.e., \(\operatorname{Post}(s)=\left\{s^{\prime} \in S \mid \mathbf{P}\left(s, s^{\prime}\right)>0\right\}\). Similarly, \(\operatorname{Pre}(s)=\left\{s^{\prime} \in S \mid \mathbf{P}\left(s^{\prime}, s\right)>0\right\}\). \(\operatorname{Post}^{*}(s)\) denotes the set of all states that are reachable from \(s\) via a finite path fragment and \(\operatorname{Pre}^{*}(s)=\left\{s^{\prime} \in S \mid s \in \operatorname{Post}^{*}\left(s^{\prime}\right)\right\}\). For \(B \subseteq S\), let
\[
\operatorname{Post}^{*}(B)=\bigcup_{s \in B} \operatorname{Post}^{*}(s) \text { and } \operatorname{Pre}^{*}(B)=\bigcup_{s \in B} \operatorname{Pre}^{*}(s)
\]

\section*{Notation 10.6. Absorbing State}

State \(s\) of MC \(\mathcal{M}\) is called absorbing if \(\operatorname{Post}^{*}(s)=\{s\}\). Since \(\mathbf{P}\) is a stochastic matrix, i.e., each row sum equals \(1, s\) is an absorbing state if and only if \(\mathbf{P}(s, s)=1\), and \(\mathbf{P}(s, t)=0\) for all states \(t \neq s\).

\section*{Remark 10.7. Discrete-Time Markov Chains}

Markov chains as in Definition 10.1 are often called discrete-time Markov chains. This has mainly historical reasons. In many cases, Markov chains are used as a time-abstract model, like transition systems provide a time-abstract operational model for reactive systems. In fact, as for transition systems, a timed interpretation of Markov chains is only adequate when the underlying time domain is discrete and each transition is assumed to take a single time unit.

The underlying graph of a Markov chain forms the basis for reasoning about qualitative properties, such as LTL or CTL formulae. To that end, define for \(\mathrm{MC} \mathcal{M}=\) \(\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) the transition system \(T S(\mathcal{M})=(S,\{\tau\}, \rightarrow, I, A P, L)\) where \(I=\{s \in\) \(\left.S \mid \iota_{\text {init }}(s)>0\right\}\) and \(s \xrightarrow{\tau} t\) iff \(\mathbf{P}(s, t)>0\). LTL or CTL formulae abstract away from the probabilities; e.g., the CTL formula \(\exists \diamond a\) may hold even if it is very unlikely that a state where \(a\) holds will eventually be visited. For instance, for the communication protocol of Example 10.2 (page 749), the event to reach state delivered within the next 100 steps, formalized in LTL by \(\varphi=\bigvee_{0 \leqslant i \leqslant 100} \bigcirc^{i}\) delivered, does not hold for the starting state, but gives no quantitative information about the fraction of paths for which \(\varphi\) is violated. In fact, the chance that \(\varphi\) does not hold is extremely small (namely \(10^{-50}\) ), and hence, might be classified as negligible. It even holds that
\[
\text { start } \vDash \exists \square \neg \text { delivered, }
\]
although it seems to be impossible that a message never will be delivered. In fact, the event \(\square \neg\) delivered holds with zero probability, as we will see later.

The goal is now to specify and check quantitative properties. Such properties may, e.g., assert that the probability of reaching a certain bad state is sufficiently small, or that the probability of achieving a certain desired system behavior is above a given threshold. Qualitative properties arise as a special case of quantitative properties where the probability bounds are the trivial bounds zero or 1 . That is, typical qualitative properties require that the probability of reaching a bad state is zero, or dually, that a certain desired system behavior appears with probability 1 .

Although Markov chains yield a rather intuitive probabilistic model, reasoning about quantitative or qualitative properties requires a formalization of the probabilities for sets of paths. This formalization is based on measure theory; in particular probability spaces and \(\sigma\)-algebras. Readers familiar with the main concepts of this field may skip the following short summary of the main concepts. Further details can be found, e.g., in [21, 150, 319].

Excursus on Probability Spaces A \(\sigma\)-algebra is a pair (Outc, \(\mathfrak{E}\) ) where Outc is a nonempty set and \(\mathfrak{E E} \subseteq 2^{\text {Outc }}\) a set consisting of subsets of Outc that contains the empty set and is closed under complementation and countable unions, i.e.,
- \(\varnothing \in \mathfrak{E}\),
- if \(E \in \mathfrak{E}\), then \(\bar{E}=\) Outc \(\backslash E \in \mathfrak{E}\),
- if \(E_{1}, E_{2}, \ldots \in \mathfrak{E}\), then \(\bigcup_{n \geqslant 1} E_{n} \in \mathfrak{E}\).

Note that the conditions of \(\sigma\)-algebras yield that Outc \(\in \mathfrak{E}\), as Outc \(=\bar{\varnothing}\), and that \(\mathfrak{E}\) is closed under countable intersections, as
\[
\bigcap_{n \geqslant 0} E_{n}=\overline{\bigcup_{n \geqslant 0} \overline{E_{n}}}
\]

Occasionally, the set Outc is supposed to be fixed and \(\mathfrak{E}\) is called a \(\sigma\)-algebra. The elements of Outc are often called outcomes, while the elements of \(\mathfrak{E}\) are called events.

For any set Outc, the powerset \(\mathfrak{E}=2^{\text {Outc }}\) yields a \(\sigma\)-algebra over Outc. In this \(\sigma\)-algebra, all subsets of Outc are events. The other extreme is the \(\sigma\)-algebra consisting of the empty set and Outc, i.e., \(\mathfrak{E}=\{\varnothing\), Outc \(\}\). Here, no nonempty proper subset of Outc is an event.

A probability measure on (Outc, \(\mathfrak{E})\) is a function \(\operatorname{Pr}: \mathfrak{E} \rightarrow[0,1]\) such that \(\operatorname{Pr}(\) Outc \()=1\), and if \(\left(E_{n}\right)_{n \geqslant 1}\) is a family of pairwise disjoint events \(E_{n} \in \mathfrak{E}\), then:
\[
\operatorname{Pr}\left(\bigcup_{n \geqslant 1} E_{n}\right)=\sum_{n \geqslant 1} \operatorname{Pr}\left(E_{n}\right) .
\]

A probability space is a \(\sigma\)-algebra equipped with a probability measure, i.e., it is a triple (Outc, \(\mathfrak{E}, \operatorname{Pr}\) ) where (Outc, \(\mathfrak{E})\) is a \(\sigma\)-algebra and \(\operatorname{Pr}\) a probability measure on (Outc, \(\mathfrak{E}\) ). The value \(\operatorname{Pr}(E)\) is called the probability measure of \(E\), or simply the probability of \(E\). In the context of probability measures, the events (i.e., the elements of \(\mathfrak{E}\) ) are often said to be measurable. That is, measurability of a set \(E \subseteq\) Outc means that \(E \in \mathfrak{E}\), and hence, it makes sense to speak about the probability measure of \(E\).

\section*{Example 10.8. Tossing a Fair Coin}

Consider the experiment in which a fair coin is tossed once. The possible outcomes are heads and tails, i.e., Outc \(=\{\) heads, tails \(\}\). For the events to be considered let us assume the singleton events \(\{\) heads \(\}\) and \(\{\) tails \(\}\) suffice. The smallest \(\sigma\)-algebra that contains these events is the powerset of \(\{\) heads, tails \(\}\). So, let \(\mathfrak{E}=2^{\text {Outc }}\). Since the coin is assumed to be fair, the probability measure \(\operatorname{Pr}\) is given by
\[
\operatorname{Pr}(\varnothing)=0, \quad \operatorname{Pr}(\{\text { heads }\})=\operatorname{Pr}(\{\text { tails }\})=\frac{1}{2}, \quad \operatorname{Pr}(\{\text { heads }, \text { tails }\})=1 .
\]

In general, whenever Outc is countable, then a probability measure on the powerset of Outc can be obtained by fixing a function \(\mu:\) Outc \(\rightarrow[0,1]\) such that
\[
\sum_{e \in \text { Outc }} \mu(e)=1
\]

Such functions \(\mu\) are called distributions on Outc. Any distribution \(\mu\) induces a probability measure on the \(\sigma\)-algebra \(\mathfrak{E}=2^{\text {Outc }}\) in the following way. For subset \(E\) of Outc, \(\operatorname{Pr}_{\mu}(E)\) is defined by \(\sum_{e \in E} \mu(e)\). In fact, it is easy to check that \(\mu\) satisfies the conditions of a probability measure. In the sequel, \(\operatorname{Pr}_{\mu}(E)\) is often abbreviated by \(\mu(E)\) and \(\operatorname{Distr}(\) Outc) is used to denote the set of distributions on Outc.

Let us summarize some fundamental properties of probability measures. Since \(E \cup \bar{E}=\) Outc and \(E\) and \(\bar{E}\) are disjoint, the above conditions imply
\[
\operatorname{Pr}(\bar{E})=1-\operatorname{Pr}(E)
\]

In particular, \(\operatorname{Pr}(\varnothing)=\operatorname{Pr}(\overline{O u t c})=1-\operatorname{Pr}(\) Outc \()=1-1=0\). Probability measures are monotonic, i.e., for events \(E\) and \(E^{\prime}\) such that \(E \subseteq E^{\prime}\) it holds that
\[
\operatorname{Pr}\left(E^{\prime}\right)=\operatorname{Pr}(E)+\operatorname{Pr}\left(E^{\prime} \backslash E\right) \geqslant \operatorname{Pr}(E)
\]

Furthermore, if \(\left(E_{n}\right)_{n \geqslant 1}\) is a family of events, possibly not pairwise disjoint, then: \(\bigcup_{n \geqslant 1} E_{n}=\) \(\bigcup_{n \geqslant 1} E_{n}^{\prime}\) where \(E_{1}^{\prime}=E_{1}\) and \(E_{n}^{\prime}=E_{n} \backslash\left(E_{1} \cup \ldots \cup E_{n-1}\right)\) for \(n \geqslant 2\). Since \(E_{n}^{\prime} \cap E_{m}^{\prime}=\varnothing\)
if \(n \neq m\) we get
\[
\operatorname{Pr}\left(\bigcup_{n \geqslant 1} E_{n}\right)=\operatorname{Pr}\left(\bigcup_{n \geqslant 1} E_{n}^{\prime}\right)=\sum_{n \geqslant 1} \operatorname{Pr}\left(E_{n}^{\prime}\right) .
\]

If \(E_{1} \subseteq E_{2} \subseteq E_{3} \subseteq \ldots\) and \(E_{n}^{\prime}\) is as above, then we have \(E_{n}^{\prime}=E_{n} \backslash E_{n-1}\) for \(n \geqslant 2\), which yields
\[
\begin{aligned}
\operatorname{Pr}\left(\bigcup_{n \geqslant 1} E_{n}\right) & =\overbrace{\operatorname{Pr}\left(E_{1}\right)}^{=\operatorname{Pr}\left(E_{1}^{\prime}\right)}+\sum_{n=2}^{\infty} \overbrace{\left(\operatorname{Pr}\left(E_{n}\right)-\operatorname{Pr}\left(E_{n-1}\right)\right)}^{=\operatorname{Pr}\left(E_{n}^{\prime}\right)} \\
& =\lim _{N \rightarrow \infty}\left(\operatorname{Pr}\left(E_{1}\right)+\sum_{n=2}^{N}\left(\operatorname{Pr}\left(E_{n}\right)-\operatorname{Pr}\left(E_{n-1}\right)\right)\right) \\
& =\lim _{N \rightarrow \infty} \operatorname{Pr}\left(E_{N}\right) .
\end{aligned}
\]

Note that this limit exists and agrees with the supremum of \(\left\{\operatorname{Pr}\left(E_{1}\right), \operatorname{Pr}\left(E_{2}\right), \ldots\right\}\), as the monotonicity of \(\operatorname{Pr}\) yields \(\operatorname{Pr}\left(E_{1}\right) \leqslant \operatorname{Pr}\left(E_{2}\right) \leqslant \ldots \leqslant 1\). For countable intersections analogous results apply. That is, if \(E_{1} \supseteq E_{2} \supseteq \ldots\), then
\[
\operatorname{Pr}\left(\bigcap_{n \geqslant 1} E_{n}\right)=\lim _{n \rightarrow \infty} \operatorname{Pr}\left(E_{n}\right)=\inf _{n \geqslant 1} \operatorname{Pr}\left(E_{n}\right) .
\]

This follows from the fact that the sequence \(\left(\overline{E_{n}}\right)_{n \geqslant 1}\) of the complements \(\overline{E_{n}}=\operatorname{Outc} \backslash E_{n}\) is decreasing. Thus, the above yields
\[
\begin{aligned}
\operatorname{Pr}\left(\bigcap_{n \geqslant 1} E_{n}\right) & =\operatorname{Pr}\left(\overline{\bigcup_{n \geqslant 1} \overline{E_{n}}}\right) \\
& =1-\operatorname{Pr}\left(\bigcup_{n \geqslant 1} \overline{E_{n}}\right) \\
& =1-\lim _{n \rightarrow \infty} \operatorname{Pr}\left(\overline{E_{n}}\right) \\
& =1-\lim _{n \rightarrow \infty}\left(1-\operatorname{Pr}\left(E_{n}\right)\right) \\
& =\lim _{n \rightarrow \infty} \operatorname{Pr}\left(E_{n}\right)
\end{aligned}
\]

Any event \(E\) with \(\operatorname{Pr}(E)=1\) is said to hold almost surely. Note that if \(E\) holds almost surely, then \(\operatorname{Pr}(D)=\operatorname{Pr}(E \cap D)\) for all events \(D\) as \(\operatorname{Pr}(D \backslash E)=0\) since \(D \backslash E\) is a subset of \(\bar{E}\) and \(\operatorname{Pr}(\bar{E})=1-\operatorname{Pr}(E)=1-1=0\). This yields
\[
\operatorname{Pr}(D)=\operatorname{Pr}(E \cap D)+\underbrace{\operatorname{Pr}(D \backslash E)}_{=0}=\operatorname{Pr}(E \cap D) .
\]

In particular, the event \(E_{1} \cap E_{2}\) of events \(E_{1}\) and \(E_{2}\) that hold almost surely, holds almost surely. As can be shown by induction, this carries over to any event that can be written as a finite intersection \(\bigcap_{1 \leqslant i \leqslant n} E_{i}\) of events \(E_{1}, \ldots, E_{n}\) that hold almost surely. By taking the limit of such (finite) intersections, we obtain that \(\bigcap_{i \geqslant 1} E_{i}\) holds almost surely if \(\operatorname{Pr}\left(E_{i}\right)=1\) for all \(i \geqslant 0\).

To define an appropriate \(\sigma\)-algebra for a given Markov chain \(\mathcal{M}\) (see below), we will use the fact that for each set Outc and each subset \(\Pi\) of \(2^{\text {Outc }}\) there exists a smallest \(\sigma\)-algebra that contains \(\Pi\). This is due to the observations that
- the powerset \(2^{\text {Outc }}\) of Outc is a \(\sigma\)-algebra, and
- the intersection of \(\sigma\)-algebras is a \(\sigma\)-algebra.

Thus, the intersection \(\mathfrak{E}_{\Pi}=\bigcap_{\mathfrak{E}} \mathfrak{E}\) where \(\mathfrak{E}\) ranges over all \(\sigma\)-algebras on Outc that contain \(\Pi\) is a \(\sigma\)-algebra and is contained in any \(\sigma\)-algebra \(\mathfrak{E}\) such that \(\Pi \subseteq \mathfrak{E} . \mathfrak{E}_{\Pi}\) is called the \(\sigma\)-algebra generated by \(\Pi\), and \(\Pi\) the basis for \(\mathfrak{E}_{\Pi}\).

Probability Measure of a Markov Chain In order to be able to associate probabilities to events in Markov chains, the intuitive notion of probabilities in the Markov chain \(\mathcal{M}\) is formalized by associating a probability space with \(\mathcal{M}\). The infinite paths of \(\mathcal{M}\) play the role of the outcomes. That is, \(\operatorname{Outc}^{\mathcal{M}}=\operatorname{Paths}(\mathcal{M})\). Recall that \(\operatorname{Paths}(\mathcal{M})\) denotes the set of all infinite sequences \(s_{0} s_{1} s_{2} \ldots \in S^{\omega}\) such that \(\mathbf{P}\left(s_{i}, s_{i+1}\right)>0\) for all \(i \geqslant 0\). (The requirement that \(\iota_{\text {init }}\left(s_{0}\right)>0\) could be added, but is irrelevant here.) The \(\sigma\)-algebra associated with \(\mathcal{M}\) is generated by the cylinder sets spanned by the finite path fragments in \(\mathcal{M}\).

\section*{Definition 10.9. Cylinder Set}

The cylinder set of \(\widehat{\pi}=s_{0} \ldots s_{n} \in \operatorname{Paths}_{f i n}(\mathcal{M})\) is defined as
\[
\operatorname{Cyl}(\widehat{\pi})=\{\pi \in \operatorname{Paths}(\mathcal{M}) \mid \widehat{\pi} \in \operatorname{pref}(\pi)\}
\]

The cylinder set spanned by the finite path \(\widehat{\pi}\) thus consists of all infinite paths that start with \(\widehat{\pi}\). The cylinder sets serve as basis events of the \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\) associated with \(\mathcal{M}\).

Definition 10.10. \(\sigma\)-Algebra of a Markov Chain
The \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\) associated with \(\mathrm{MC} \mathcal{M}\) is the smallest \(\sigma\)-algebra that contains all cylinder sets \(\operatorname{Cyl}(\widehat{\pi})\) where \(\widehat{\pi}\) ranges over all finite path fragments in \(\mathcal{M}\).

From classical concepts of probability theory (see e.g. [21, 150]), it follows that there exists a unique probability measure \(\operatorname{Pr}^{\mathcal{M}}\) (or, briefly, \(\operatorname{Pr}\) ) on the \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\) associated with \(\mathcal{M}\) where the probabilities for the cylinder sets (i.e., the events) are given by
\[
\operatorname{Pr}^{\mathcal{M}}\left(\operatorname{Cyl}\left(s_{0} \ldots s_{n}\right)\right)=\iota_{\text {init }}\left(s_{0}\right) \cdot \mathbf{P}\left(s_{0} \ldots s_{n}\right)
\]
where
\[
\mathbf{P}\left(s_{0} s_{1} \ldots s_{n}\right)=\prod_{0 \leqslant i<n} \mathbf{P}\left(s_{i}, s_{i+1}\right)
\]

For path fragments of length zero, let \(\mathbf{P}\left(s_{0}\right)=1\).
For paths starting in a certain (possibly noninitial) state \(s\), the same construction is applied to the \(\mathrm{MC} \mathcal{M}_{s}\) that results from \(\mathcal{M}\) by declaring \(s\) as the unique initial state. Formally, for \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) and \(s \in S\), the MC \(\mathcal{M}_{s}\) is defined by \(\mathcal{M}_{s}=\left(S, \mathbf{P}, \iota_{s}^{1}, A P, L\right)\) where
\[
\iota_{s}^{1}(t)= \begin{cases}1 & \text { if } s=t \\ 0 & \text { otherwise }\end{cases}
\]

We often write \(\operatorname{Pr}_{s}^{\mathcal{M}}\) for \(\operatorname{Pr}^{\mathcal{M}}\). If \(\mathcal{M}\) is clear from the context, then \(\mathcal{M}\) is omitted and we simply write \(\operatorname{Pr}_{s}\).

\section*{Example 10.11. Cylinder Sets}

Consider the Markov chain for the craps game, see Figure 10.3. The state sequence \(\widehat{\pi}=\operatorname{start} 666 \in\) Paths \(_{f i n}\left(s_{0}\right)\). Its probability is given by
\[
\iota_{s}(\text { start }) \cdot \mathbf{P}(\text { start }, 6) \cdot \mathbf{P}(6,6) \cdot \mathbf{P}(6,6)=\frac{5}{36} \cdot\left(\frac{25}{36}\right)^{2}
\]

The cylinder set of \(\widehat{\pi}\) is given by
\[
\left\{\operatorname{start} 6^{n} \text { won }^{\omega} \mid n>2\right\} \cup\left\{\operatorname{start} 6^{n} \text { lost }^{\omega} \mid n>2\right\} \cup\left\{\operatorname{start} 6^{\omega}\right\}
\]

\section*{Notation 10.12. LTL-Style Notations for Events}

In the remainder of this chapter, LTL-like notations are frequently used to describe events in Markov chains. For example, for set \(B \subseteq S\) of states, \(\Delta B\) denotes the event to reach
(some state in) \(B\) eventually, and \(\square \diamond B\) describes the event that \(B\) should be visited infinitely often. For singleton sets, the set brackets are typically omitted; e.g., \(\square \diamond t\) means that \(\{t\}\) is visited infinitely often. In line with the notations for LTL, we often write \(\pi \models \varphi\) rather than \(\pi \in \varphi\) if \(\varphi\) is an LTL-style notation for an event (i.e., an LTL formula with sets of states as atomic propositions) and \(\pi\) a path in \(\mathcal{M}\). The probability for \(\varphi\) to hold in state \(s\) is denoted by \(\operatorname{Pr}^{\mathcal{M}}(s \models \varphi)\), i.e.,
\[
\operatorname{Pr}^{\mathcal{M}}(s \models \varphi)=\operatorname{Pr}_{s}^{\mathcal{M}}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \varphi\} .
\]

If \(\mathcal{M}\) is clear from the context the superscript \(\mathcal{M}\) is omitted.

\subsection*{10.1.1 Reachability Probabilities}

One of the elementary questions for the quantitative analysis of systems modeled by Markov chains is to compute the probability of reaching a certain set \(B\) of states. The set \(B\) may represent a set of certain bad states which should be visited only with some small probability, or dually, a set of good states which should rather be visited frequently.

In the sequel, let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov chain and \(B \subseteq S\) a set of states. The event of interest is denoted \(\diamond B\). In order to be able to reason about the probability of this event, we need to characterize the event \(\diamond B\) as a measurable set of paths. In fact, this is not difficult because the event \(\diamond B\) agrees with the union of all basic cylinders \(\operatorname{Cyl}\left(s_{0} \ldots s_{n}\right)\) where \(s_{0} \ldots s_{n}\) is an initial path fragment in \(\mathcal{M}\) such that \(s_{0}, \ldots, s_{n-1} \notin B\) and \(s_{n} \in B\). The set of all such paths is given by \(\operatorname{Paths}_{f i n}(\mathcal{M}) \cap(S \backslash B)^{*} B\). As the set of finite path fragments is countable, this set is measurable, i.e., it is an element of the \(\sigma\)-algebra \(E^{\mathcal{M}}\) on \(\mathcal{M}\). Moreover, since these cylinder sets are pairwise disjoint, the probability of eventually reaching \(B\) is given by
\[
\begin{aligned}
\operatorname{Pr}^{\mathcal{M}}(\diamond B) & =\sum_{s_{0} \ldots s_{n} \in{\operatorname{Paths} s_{f n}(\mathcal{M}) \cap(S \backslash B)^{*} B} \operatorname{Pr}^{\mathcal{M}}\left(\operatorname{Cyl}\left(s_{0} \ldots s_{n}\right)\right)} \\
& =\sum_{s_{0} \ldots s_{n} \in \operatorname{Paths}_{f i n}(\mathcal{M}) \cap(S \backslash B)^{* B}} \iota_{\text {init }}\left(s_{0}\right) \cdot \mathbf{P}\left(s_{0} \ldots s_{n}\right) .
\end{aligned}
\]

\section*{Example 10.13. Computing Reachability Probabilities by Infinite Series}

Consider the Markov chain in Figure 10.1 on page 749, and suppose our interest is to compute the probability of reaching the state delivered. For this event, the finite initial path fragments \(s_{0} \ldots s_{n}\) with \(s_{i} \neq\) delivered for \(0 \leqslant i<n\) and \(s_{n}=\) delivered are of
interest. These path fragments are of the form
\[
\widehat{\pi}_{n}=\text { start try }(\text { lost try })^{n} \text { delivered }
\]
where \(n\) is an arbitrary natural number. The probability of the cylinder set spanned by \(\widehat{\pi_{n}}\) is \(\left(\frac{1}{10}\right)^{n} \cdot \frac{9}{10}\). Hence,
\[
\operatorname{Pr}^{\mathcal{M}}(\diamond \text { delivered })=\sum_{n=0}^{\infty}\left(\frac{1}{10}\right)^{n} \cdot \frac{9}{10}=\frac{\frac{9}{10}}{1-\frac{1}{10}}=\frac{\frac{9}{10}}{\frac{9}{10}}=1
\]

The event \(\diamond\) delivered thus holds almost surely. This is, in fact, rather intuitive since in the absence of a bound on the number of (re)transmissions it is to be expected that any message is delivered eventually. Let us now impose an upper bound, say three, for the number of transmission of a message. In order to determine the probability of the event to deliver the message within three trials, we sum the probabilities for the cylinder sets of the path fragments \(\widehat{\pi}_{0}, \widehat{\pi}_{1}\), and \(\widehat{\pi}_{2}\) which yields
\[
\frac{9}{10}+\frac{1}{10} \cdot \frac{9}{10}+\frac{1}{10} \cdot \frac{1}{10} \cdot \frac{9}{10}=0.999
\]

This example shows how the probability of reaching a certain set of states can be calculated by means of infinite sums. In general, this technique is rather involved and cumbersome. For instance, the computation of the event \(\diamond\) won in the craps game (see Figure 10.3, page 751 ) is much more involved. Let us now explain how probabilities of reaching a certain set \(B\) of states in a finite MC can be computed in a more efficient way, i.e., without considering infinite sums.

Let variable \(x_{s}\) denote the probability of reaching \(B\) from \(s\), for arbitrary \(s \in S\). The goal is to compute \(x_{s}=\operatorname{Pr}(s \models \diamond B)\) for all states \(s\). Clearly, if \(B\) is not reachable from \(s\) in the underlying directed graph of \(\mathcal{M}\), then \(x_{s}=0\). Also the converse holds, i.e., if \(x_{s}>0\), then \(B\) is reachable from \(s\). Moreover, \(x_{s}=1\) if \(s \in B\). For the states \(s \in S \backslash B\) for which \(B\) is reachable, it holds that
\[
x_{s}=\sum_{t \in S \backslash B} \mathbf{P}(s, t) \cdot x_{t}+\sum_{u \in B} \mathbf{P}(s, u) .
\]

This equation states that either \(B\) is reached within one step, i.e., by a finite path fragment \(s u\) with \(u \in B\) (second summand), or first a state \(t \in S \backslash B\) is reached from which \(B\) is reached - this corresponds to the path fragments \(s t \ldots u\) of length \(\geqslant 2\) where all states (except the last one) do not belong to \(B\) (first summand). Let \(\tilde{S}=\operatorname{Pre}^{*}(B) \backslash B\) denote
the set of states \(s \in S \backslash B\) such that there is a path fragment \(s_{0} s_{1} \ldots s_{n}(n>0)\) with \(s_{0}=s\) and \(s_{n} \in B\). For the vector \(\mathbf{x}=\left(x_{s}\right)_{s \in \tilde{S}}\), we have
\[
\mathbf{x}=\mathbf{A x}+\mathbf{b}
\]
where the matrix \(\mathbf{A}\) contains the transition probabilities for the states in \(\tilde{S}\), i.e., \(\mathbf{A}=\) \((\mathbf{P}(s, t))_{s, t \in \tilde{S}}\), and the vector \(\mathbf{b}=\left(b_{s}\right)_{s \in \tilde{S}}\) contains the probabilities of reaching \(B\) from \(\tilde{S}\) within one step, i.e., \(b_{s}=\mathbf{P}(s, B)=\sum_{u \in B} \mathbf{P}(s, u)\). The above equation system can be rewritten into a (heterogeneous) linear equation system
\[
(\mathbf{I}-\mathbf{A}) \cdot \mathbf{x}=\mathbf{b}
\]
where \(\mathbf{I}\) is the identity matrix of cardinality \(|\tilde{S}| \times|\tilde{S}|\).

\section*{Example 10.14. Simple Communication Protocol}

Consider the simple communication protocol in Figure 10.1 on page 749 and the event \(\diamond B\) for \(B=\{\) delivered \(\}\). Then \(x_{s}>0\) for all states \(s\), since delivered is reachable from all states. In this case, \(\tilde{S}=\{\) start, try, lost \(\}\) and we obtain the equations
\[
\begin{aligned}
x_{\text {start }} & =x_{\text {try }} \\
x_{\text {try }} & =\frac{1}{10} \cdot x_{\text {lost }}+\frac{9}{10} \\
x_{\text {lost }} & =x_{\text {try }} .
\end{aligned}
\]

These equations can be rewritten as
\[
\left(\begin{array}{rrc}
1 & -1 & 0 \\
0 & 1 & -\frac{1}{10} \\
0 & -1 & 1
\end{array}\right) \cdot \mathbf{x}=\left(\begin{array}{c}
0 \\
\frac{9}{10} \\
0
\end{array}\right)
\]
which yields the (unique) solution \(x_{\text {start }}=x_{\text {try }}=x_{\text {lost }}=1\). Thus, the event of eventually reaching the state delivered is almost sure for any state.

The above technique yields the following two-phase algorithm to compute reachability probabilities in finite Markov chains: first, perform a graph analysis to compute the set \(\tilde{S}\) of all states that can reach \(B\) (e.g., by a backward DFS- or BFS-based search from \(B\) ), then generate the matrix \(\mathbf{A}\) and the vector \(\mathbf{b}\), and solve the linear equation system \((\mathbf{I}-\mathbf{A}) \cdot \mathbf{x}=\mathbf{b}\). However, \((\mathbf{I}-\mathbf{A}) \cdot \mathbf{x}=\mathbf{b}\) might have more than one solution. This is the case if \(\mathbf{I}-\mathbf{A}\) is singular, i.e., does not have an inverse. This problem is addressed below by characterizing the desired probability vector as the least solution in \([0,1]^{\tilde{S}}\). This characterization enables computing the probability vector by an iterative approximation method. (To apply direct methods, like Gaussian elimination, several rows and columns of the matrix \(\mathbf{I}-\mathbf{A}\) have to be removed such that the remaining linear equation system
has a unique solution.) In fact, we present a characterization for a slightly more general problem, viz. constrained reachability (i.e., until properties). The obtained characterization also partly applies to infinite Markov chains.

Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a (possibly infinite) MC and \(B, C \subseteq S\). Consider the event of reaching \(B\) via a finite path fragment which ends in a state \(s \in B\), and visits only states in \(C\) prior to reaching \(s\). Using LTL-like notations, this event is denoted by \(C \cup B\). The event \(\diamond B\) considered above agrees with \(S \cup B\). For \(n \geqslant 0\), the event \(C \cup \leqslant n B\) has the same meaning as \(C \cup B\), except that it is required to reach \(B\) (via states in \(C\) ) within \(n\) steps. Formally, \(C U \leqslant n B\) is the union of the basic cylinders spanned by path fragments \(s_{0} s_{1} \ldots s_{k}\) such that \(k \leqslant n\) and \(s_{i} \in C\) for all \(0 \leqslant i<k\) and \(s_{k} \in B\).

Let \(S_{=0}, S_{=1}, S_{\text {? }}\) a partition of \(S\) such that
- \(B \subseteq S_{=1} \subseteq\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=1\}\),
- \(S \backslash(C \cup B) \subseteq S_{=0} \subseteq\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=0\}\), and
- \(S_{?}=S \backslash\left(S_{=1} \cup S_{=0}\right)\).

For all states in the set \(S_{=1}\), the event \(C \cup B\) almost surely holds. All states in \(S_{\text {? }}\) belong to \(C \backslash B\). Let the matrix \(\mathbf{A}\) be a quadratic matrix with rows and columns for the states in \(S_{\text {? }}\). This matrix is obtained from the transition probability matrix \(\mathbf{P}\) by omitting the rows and columns for the states \(s \in S_{=0} \cup S_{=1}\) from \(\mathbf{P}\). That is
\[
\mathbf{A}=(\mathbf{P}(s, t))_{s, t \in S_{?}}
\]

Similarly, let vector be defined as \(\left(b_{s}\right)_{s \in S \text { ? }}\) where \(b_{s}=\mathbf{P}\left(s, S_{=1}\right)\).
We now provide a least fixed point characterization for the probability vector \((\operatorname{Pr}(s \models\) \(C \cup B))_{s \in S_{?}}\). Note that the set \([0,1]^{S_{?}}\) consists of all vectors \(\mathbf{y}=\left(y_{s}\right)_{s \in S_{?}}\) with \(y_{s} \in[0,1]\) for all \(s \in S_{\text {? }}\). To obtain the least fixed point characterization, the set \([0,1]^{S_{?}}\) is equipped with the partial order \(\leqslant\) given by \(\mathbf{y} \leqslant \mathbf{y}^{\prime}\) if and only if \(y_{s} \leqslant y_{s}^{\prime}\) for all \(s \in S_{\text {? }}\), where \(\mathbf{y}=\left(y_{s}\right)_{s \in S_{?}}\) and \(\mathbf{y}^{\prime}=\left(y_{s}^{\prime}\right)_{s \in S_{?}}\).

\section*{Theorem 10.15. Least Fixed Point Characterization}

The vector \(\mathbf{x}=(\operatorname{Pr}(s \models C \cup B))_{s \in S ?}\) is the least fixed point of the operator \(\Upsilon:[0,1]^{S_{?}} \rightarrow\) \([0,1]^{S ?}\) which is given by
\[
\Upsilon(\mathbf{y})=\mathbf{A} \cdot \mathbf{y}+\mathbf{b}
\]

Furthermore, if \(\mathbf{x}^{(0)}=\mathbf{0}\) is the vector consisting of zeros only, and \(\mathbf{x}^{(n+1)}=\Upsilon\left(\mathbf{x}^{(n)}\right)\) for \(n \geqslant 0\), then
- \(\mathbf{x}^{(n)}=\left(x_{s}^{(n)}\right)_{s \in S_{?}}\) where \(x_{s}^{(n)}=\operatorname{Pr}\left(s \models C U^{\leqslant n} S_{=1}\right)\) for each state \(s \in S_{\text {? }}\),
- \(\mathbf{x}^{(0)} \leqslant \mathbf{x}^{(1)} \leqslant \mathbf{x}^{(2)} \leqslant \ldots \leqslant \mathbf{x}\), and
- \(\mathbf{x}=\lim _{n \rightarrow \infty} \mathbf{x}^{(n)}\).

Before proving Theorem 10.15, let us first comment on the well-definedness of \(\Upsilon\) as a function from \([0,1]^{S_{?}}\) to \([0,1]^{S_{?}}\). For \(\mathbf{y}=\left(y_{s}\right)_{s \in S_{?}}\), the vector \(\Upsilon(\mathbf{y})=\left(y_{s}^{\prime}\right)_{s \in S_{?}}\) has the entries
\[
y_{s}^{\prime}=\sum_{t \in S_{?}} \mathbf{P}(s, t) \cdot y_{t}+\mathbf{P}\left(s, S_{=1}\right)
\]

Since \(0 \leqslant y_{t} \leqslant 1\) for all \(t \in S_{?}, \mathbf{P}\left(s, s^{\prime}\right) \geqslant 0\) and \(\sum_{s^{\prime} \in S} \mathbf{P}\left(s, s^{\prime}\right)=1\), this implies \(0 \leqslant y_{s}^{\prime} \leqslant 1\). Therefore, \(\Upsilon(\mathbf{y}) \in[0,1]^{S_{?}}\).

Proof: Let \(x_{s}=\operatorname{Pr}(s \mid=C \cup B)\) for each state \(s \in S\). Then, it follows from the definitions of \(S_{=0}\) and \(S_{=1}\) that \(\mathbf{x}=\left(x_{s}\right)_{s \in S_{?}}, x_{s}=0\) for \(s \in S_{=0}\), and \(x_{s}=1\) for \(s \in S_{=1}\).
1. Let us first show that \(\mathbf{x}\) agrees with the limit of the vectors \(\mathbf{x}^{(n)}=\left(x_{s}^{(n)}\right)_{s \in S ?}\). By induction on \(n\), one can show that \(x_{s}^{(n)}=\operatorname{Pr}\left(s \models C \mathrm{U}^{\leqslant n} S_{=1}\right)\). Since \(C \cup S_{=1}\) is the countable union of the events \(C \mathrm{U}^{\leqslant n} S_{=1}\), we obtain
\[
\lim _{n \rightarrow \infty} x_{s}^{(n)}=\operatorname{Pr}\left(s \models C \cup S_{=1}\right)=x_{s}
\]
2. The fixed point property \(\mathbf{x}=\Upsilon(\mathbf{x})\) is a consequence of the fact that \(x_{s}=\operatorname{Pr}(s \models\) \(C \cup B)=0\) if \(s \in S_{=0}\) and \(x_{s}=\operatorname{Pr}(s \models C \cup B)=1\) if \(s \in S_{=1}\). For \(s \in S_{\text {? }}\) we derive that
\[
\begin{aligned}
x_{s} & =\sum_{t \in S} \mathbf{P}(s, t) \cdot x_{t} \\
& =\sum_{t \in S_{=0}} \mathbf{P}(s, t) \cdot \underbrace{x_{t}}_{=0}+\sum_{t \in S_{?}} \mathbf{P}(s, t) \cdot x_{t}+\sum_{t \in S_{=1}} \mathbf{P}(s, t) \cdot \underbrace{x_{t}}_{=1} \\
& =\sum_{t \in S_{?}} \mathbf{P}(s, t) \cdot x_{t}+\underbrace{\sum_{t \in S_{=1}} \mathbf{P}(s, t)}_{=b_{s}}
\end{aligned}
\]
is the component for state \(s\) in the vector \(\Upsilon(\mathbf{x})\).
3. It remains to show that \(\mathbf{x} \leqslant \mathbf{y}\) for each fixed point \(\mathbf{y}\) of \(\Upsilon\). This follows from the fact that \(\mathbf{x}^{(n)} \leqslant \mathbf{y}\) for all \(n\), which can be shown by induction on \(n\). But then \(\mathbf{x}=\lim _{n \rightarrow \infty} \mathbf{x}^{(n)} \leqslant \mathbf{y}\).

This completes the proof.

\section*{Remark 10.16. Expansion Law}

The statement of the above theorem with \(S_{=1}=B\) and \(S_{=0}=S \backslash(C \cup B)\) can be considered as the probabilistic counterpart to the characterization of the CTL formula \(\exists(C \cup B)\) as the least solution of the expansion law:
\[
\exists(C \cup B) \equiv B \vee(C \wedge \exists \bigcirc \exists(C \cup B))
\]

To make this clear, let us rewrite this expansion law in the following way. The set \(X=\) \(\operatorname{Sat}(\exists(C \cup B))\) is the least set such that
\[
B \cup\{s \in C \backslash B \mid \operatorname{Post}(s) \cap X \neq \varnothing\} \subseteq X
\]

In Theorem 10.15, this set-based least fixed point characterization is phrased for the quantitative setting where instead of truth values for " \(s \in X\) " we deal with values \(x_{s}\) in \([0,1]\). If \(s \in B\), then \(x_{s}=1\). This corresponds to the statement that \(s \in B\) implies \(s \in X\). If \(s \in C \backslash B\), then \(x_{s}=\sum_{t \in C \backslash B} \mathbf{P}(s, t) \cdot x_{t}+\sum_{t \in B} \mathbf{P}(s, t)\). This corresponds to the statement that \(s \in C \backslash B\) and \(\operatorname{Post}(s) \cap X \neq \varnothing\) implies \(s \in X\). Finally, if \(s \in S \backslash(C \cup B)\), then \(x_{s}=0\). This corresponds to the statement that \(s \notin X\) if \(s \notin C \cup B\).

The last part of Theorem 10.15 in fact provides a recipe for computing an approximation of the desired probability vector \(\mathbf{x}\). It suggests to use the following iterative approach:
\[
\mathbf{x}^{(0)}=\mathbf{0} \quad \text { and } \quad \mathbf{x}^{(n+1)}=\mathbf{A} \mathbf{x}^{(n)}+\mathbf{b} \quad \text { for } n \geqslant 0
\]

This method, often called power method, provides a simple iterative algorithm which computes the vectors \(\mathbf{x}^{(0)}, \mathbf{x}^{(1)}, \mathbf{x}^{(2)}, \ldots\) by matrix-vector multiplication and vector addition. It aborts as soon as \(\max _{s \in S_{?}}\left|x_{s}^{(n+1)}-x_{s}^{(n)}\right|<\varepsilon\) for some small user-defined tolerance \(\varepsilon\). Although the convergence of the power method is guaranteed, it is often less efficient than other iterative methods for solving linear equation systems of large dimensions, such as the Jacobi or Gauss-Seidel method. Explanations of such numerical methods for linear equation systems go beyond the scope of this monograph and can be found in textbooks such as [196, 248, 344].

Remark 10.17. Choosing \(S_{=0}\) and \(S_{=1}\)
The constraints on \(S_{=0}\) and \(S_{=1}\)
\[
B \subseteq S_{=1} \subseteq\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=1\}
\]
and
\[
S \backslash(C \cup B) \subseteq S_{=0} \subseteq\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=0\},
\]
do not uniquely characterize \(S_{=0}\) and \(S_{=1}\). For instance, \(S_{=0}=S \backslash(C \cup B)\) and \(S_{=1}=B\) would suffice. For efficiency reasons, it is advantageous to deal with the largest possible sets. The larger the sets \(S_{=0}\) and \(S_{=1}\), the smaller is their complement \(S_{\text {? }}\), and the smaller the linear equation system that needs to be solved-as there is a variable for each \(s \in S\) ? . A reasonable choice is
\[
\begin{aligned}
& S_{=0}=\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=0\} \quad \text { and } \\
& S_{=1}=\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=1\} .
\end{aligned}
\]

These sets can be computed by simple graph algorithms that have a linear-time complexity in the size of the \(\mathrm{MC} \mathcal{M}\), i.e., the number of states plus the number of nonzero entries in the transition probability matrix. The computation of \(S_{=0}\) is straightforward since
\[
\operatorname{Pr}(s \models C \cup B)=0 \text { if and only if } s \not \models \exists(C \cup B)
\]
where \(\exists(C \cup B)\) is viewed as a CTL formula. Thus, \(S_{=0}\) can be calculated by means of a backward search starting from \(B\) in time \(\mathcal{O}(\operatorname{size}(\mathcal{M}))\). The computation of \(S_{=1}\) can also be realized by means of simple graph traversal techniques, as will be explained in the next section; see page 775 ff .

The results we have established so far permit computing approximations of the probability \(\operatorname{Pr}(s \models C \cup B)\) by iterative methods. In order to apply direct methods for a linear equation system, like Gaussian elimination, the uniqueness of the linear equation system to be solved is required. Without any additional assumptions on the choice of \(S_{=0}\), the uniqueness can, however, not be guaranteed. This even applies to finite Markov chains as shown by the following example.

\section*{Remark 10.18. Several Fixed Points}

Under the assumption that \(S_{=0}\) is a proper subset of \(\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=0\}\), the operator \(\Upsilon\) may have more than one fixed point. Consider, e.g., the Markov chain:


Consider the event \(\Delta s\). A possible choice is \(S_{=0}=\varnothing\) and \(S_{=1}=\{s\}\), i.e., \(S_{?}=\{t\}\). Matrix \(\mathbf{A}\) is the identity matrix of cardinality one and \(\mathbf{b}\) is the vector with the single entry 0 for state \(t\). Thus, the equation system \(\mathbf{x}=\mathbf{A x}+\mathbf{b}\) represents the trivial equation \(x_{t}=x_{t}\) and the operator \(\Upsilon:[0,1] \rightarrow[0,1]\) is given by \(\Upsilon\left(y_{t}\right)=y_{t}\). Clearly, \(\Upsilon\) has infinitely many fixed points, namely all values \(y_{t} \in[0,1]\). However, the probabilities for the event \(\diamond s\) are given by the least fixed point \(x_{t}=0\) of \(\Upsilon\).

Unique fixed points are guaranteed if \(\mathcal{M}\) is finite and if \(S_{=0}\) covers all states \(s\) with \(s \not \vDash \exists(C \cup B)\). As mentioned before, these are exactly the states \(s\) for which \(\operatorname{Pr}(s \neq\) \(C \cup B)=0\). This fact is captured by the following theorem:

\section*{Theorem 10.19. Unique Solution}

Let \(\mathcal{M}\) be a finite Markov chain with state space \(S\), and \(B, C \subseteq S\),
\[
S_{=0}=\operatorname{Sat}(\neg \exists(C \cup B)) \text { and } B \subseteq S_{=1} \subseteq\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=1\}
\]
and \(S_{\text {? }}=S \backslash\left(S_{=0} \cup S_{=1}\right)\). Then, the vector \((\operatorname{Pr}(s \models C \cup B))_{s \in S_{?}}\) is the unique solution of the equation system \(\mathbf{x}=\mathbf{A x}+\mathbf{b}\) where \(\mathbf{A}=(\mathbf{P}(s, t))_{s, t \in S_{?}}\), and \(\mathbf{b}=\left(\mathbf{P}\left(s, S_{=1}\right)\right)_{s \in S_{?}}\).

Proof: Suppose there are two solutions of the equation system, say \(\mathbf{x}=\mathbf{A x}+\mathbf{b}\) and \(\mathbf{y}=\mathbf{A y}+\mathbf{b}\). Thus \(\mathbf{x}-\mathbf{y}=\mathbf{A}(\mathbf{x}-\mathbf{y})\). Assume \(\mathbf{A x}=\mathbf{x}\) implies \(\mathbf{x}=\mathbf{0}\), where \(\mathbf{0}\) is a vector just consisting of zeros. Then \(\mathbf{x}-\mathbf{y}=\mathbf{A}(\mathbf{x}-\mathbf{y})\) yields \(\mathbf{x}-\mathbf{y}=\mathbf{0}\), and therefore \(\mathbf{x}=\mathbf{y}\).

We now prove that indeed \(\mathbf{A x}=\mathbf{x}\) implies \(\mathbf{x}=\mathbf{0}\). This is proven by contraposition. Assume \(\mathbf{x}=\left(x_{s}\right)_{s \in S ?}\) is a vector such that \(\mathbf{A x}=\mathbf{x}\) and \(\mathbf{x} \neq \mathbf{0}\). Since \(\mathcal{M}\) is finite, the maximum of the values \(\left|x_{s}\right|\) is well-defined. Let \(x\) be this maximum, and \(T\) the set of states for which \(x_{s}=x\), i.e.:
\[
x=\max \left\{\left|x_{s}\right| \mid s \in S_{?}\right\} \quad \text { and } \quad T=\left\{s \in S_{?}| | x_{s} \mid=x\right\} .
\]

As \(\mathbf{x} \neq \mathbf{0}\), it follows that \(x>0\). Furthermore, \(T \neq \varnothing\). Since the values \(\mathbf{P}(s, t)\) are non-negative and \(\sum_{t \in S_{?}} \mathbf{P}(s, t) \leqslant 1\), we obtain for each \(s \in T\) :
\[
x=\left|x_{s}\right| \leqslant \sum_{t \in S_{?}} \mathbf{P}(s, t) \cdot \underbrace{\left|x_{t}\right|}_{\leqslant x} \leqslant x \cdot \sum_{t \in S_{?}} \mathbf{P}(s, t) \leqslant x .
\]

This yields
\[
x=\left|x_{s}\right|=\sum_{t \in S_{?}} \mathbf{P}(s, t) \cdot\left|x_{t}\right|=x \cdot \sum_{t \in S_{?}} \mathbf{P}(s, t)
\]

As \(x>0\), this implies \(\sum_{t \in S_{?}} \mathbf{P}(s, t)=1\) and \(\left|x_{t}\right|=x\) for all states in \(\operatorname{Post}(s) \cap S_{\text {? }}\). But then for any \(s \in T\) :
\[
\operatorname{Post}(s)=\{t \in S \mid \mathbf{P}(s, t)>0\} \subseteq T
\]

We conclude that \(\operatorname{Post}^{*}(s) \subseteq T\) for all \(s \in T\). In particular, \(\operatorname{Post}^{*}(s) \subseteq S_{\text {? }}\) and \(\operatorname{Post}^{*}(s) \cap\) \(B=\varnothing\) (recall that \(B \subseteq S_{=1}\) and therefore \(B \cap S_{?}=\varnothing\) ). Thus, none of the states \(s \in T\) can reach \(B\). Therefore,
\[
T \subseteq\{s \in S \mid s \not \models \exists(C \cup B)\} \subseteq S_{=0}
\]

But, as \(T \subseteq S_{\text {? }}\) and \(S_{=0} \cap S_{?}=\varnothing, T\) must be empty. Contradiction.

\section*{Remark 10.20. Nonsingularity of Matrix I-A}

A short remark for readers familiar with matrix norms. By roughly the same arguments as in the proof of Theorem 10.15 (page 762) it follows that the matrix \(\mathbf{A}\) does not have an Eigenvalue \(\lambda\) with \(|\lambda| \geqslant 1\). Then, the spectral norm of \(\mathbf{A}\), defined as
\[
\max \{|\lambda| \mid \lambda \text { is a (complex) Eigenvalue of } \mathbf{A}\}
\]
is strictly less than 1 . Thus, the infinite series:
\[
\mathbf{I}+\mathbf{A}+\mathbf{A}^{2}+\mathbf{A}^{3}+\ldots=\sum_{n \geqslant 0} \mathbf{A}^{n}
\]
converges and its limit is the inverse of \(\mathbf{I}-\mathbf{A}\) where \(\mathbf{I}\) is the identity matrix (of the same cardinality as A). Note that
\[
\begin{aligned}
& (\mathbf{I}-\mathbf{A}) \cdot\left(\mathbf{I}+\mathbf{A}+\mathbf{A}^{2}+\mathbf{A}^{3}+\ldots\right) \\
= & \left(\mathbf{I}+\mathbf{A}+\mathbf{A}^{2}+\mathbf{A}^{3}+\ldots\right)-\left(\mathbf{A}+\mathbf{A}^{2}+\mathbf{A}^{3}+\mathbf{A}^{4}+\ldots\right) \\
= & \mathbf{I}
\end{aligned}
\]

This yields another argument why \(\mathbf{x}=(\mathbf{I}-\mathbf{A})^{-1} \mathbf{b}\) is a unique solution of \(\mathbf{A x}+\mathbf{b}=\mathbf{x}\).

As a result, Theorem 10.15 yields an iterative method for computing step-bounded constrained reachability probabilities. The probabilities of the events \(C \mathrm{U}{ }^{\leqslant n} B\) can be computed by the following iterative scheme: \(\mathbf{x}^{(0)}=\mathbf{0}\) and \(\mathbf{x}^{(i+1)}=\mathbf{A} \mathbf{x}^{(i)}+\mathbf{b}\) for \(0 \leqslant i<n\). The matrix \(\mathbf{A}\) and the vector \(\mathbf{b}\) are defined based on \(S_{=0}=S \backslash(C \cup B), S_{=1}=B\) and \(S_{?}=C \backslash B\). That is:
\[
\mathbf{A}=(\mathbf{P}(s, t))_{s, t \in C \backslash B} \quad \text { and } \quad \mathbf{b}=(\mathbf{P}(s, B))_{s \in C \backslash B}
\]

For \(s \in C \backslash B\), entry \(\mathbf{x}^{(n)}(s)\) equals \(\operatorname{Pr}\left(s \models C \mathbf{U}^{\leqslant n} B\right)\).

Example 10.21. Constrained Reachability in the Craps Game
Consider the Markov chain modeling the craps game, see Figure 10.3 (page 751). The event of interest is \(C \mathrm{U}^{\leqslant n} B\) where \(B=\{\) won \(\}, C=\{\) start, \(4,5,6\}\). The bounded constrained reachability probability \(\operatorname{Pr}\left(\right.\) start \(\left.\models C \cup^{\leqslant n} B\right)\) is the likelihood of winning the game by either throwing 7 or 11 (the player wins directly), or one or more times only 4,5 , or 6 . It follows that
\[
S_{=0}=\{8,9,10, \text { lost }\} \quad S_{=1}=\{\text { won }\} \quad S_{?}=\{\text { start }, 4,5,6\}
\]

Using the state order start \(<4<5<6\), the matrix \(\mathbf{A}\) and the vector \(\mathbf{b}\) are given by
\[
\mathbf{A}=\frac{1}{36}\left(\begin{array}{cccc}
0 & 3 & 4 & 5 \\
0 & 27 & 0 & 0 \\
0 & 0 & 26 & 0 \\
0 & 0 & 0 & 25
\end{array}\right) \quad \mathbf{b}=\frac{1}{36}\left(\begin{array}{l}
8 \\
3 \\
4 \\
5
\end{array}\right)
\]

The least fixed point characterization suggests the following iterative scheme:
\[
\mathbf{x}^{(0)}=\mathbf{0} \quad \text { and } \quad \mathbf{x}^{(i+1)}=\mathbf{A} \mathbf{x}^{(i)}+\mathbf{b} \text { for } 0 \leqslant i<n .
\]
where \(\mathbf{x}^{i}\) records for any states in \(S_{\text {? }}\) the probability of the event \(C \mathrm{U} \leqslant^{n} B\). Applying this iterative scheme to the example yields \(\mathbf{x}^{(1)}=\mathbf{b}\) and
\[
\mathbf{x}^{(2)}=\frac{1}{36}\left(\begin{array}{cccc}
0 & 3 & 4 & 5 \\
0 & 27 & 0 & 0 \\
0 & 0 & 26 & 0 \\
0 & 0 & 0 & 25
\end{array}\right) \cdot \frac{1}{36}\left(\begin{array}{l}
8 \\
3 \\
4 \\
5
\end{array}\right)+\frac{1}{36}\left(\begin{array}{l}
8 \\
3 \\
4 \\
5
\end{array}\right)=\left(\frac{1}{36}\right)^{2}\left(\begin{array}{l}
338 \\
189 \\
248 \\
305
\end{array}\right)
\]

For instance, we have \(\operatorname{Pr}(\) start \(\models C \cup \leqslant 2 B)=\frac{338}{36^{2}}\). In a similar way, one obtains \(\mathbf{x}^{(3)}, \mathbf{x}^{(4)}\), and so forth.

\section*{Remark 10.22. Transient State Probabilities}

The \(n\)th power of \(\mathbf{A}\), i.e., the matrix \(\mathbf{A}^{n}\), contains the state probabilities after exactly \(n\) steps (i.e., transitions) inside \(S_{\text {? }}\). More precisely, matrix entry \(\mathbf{A}^{n}(s, t)\) equals the sum of the probabilities \(\mathbf{P}\left(s_{0} s_{1} \ldots s_{n}\right)\) of all path fragments \(s_{0} s_{1} \ldots s_{n}\) with \(s_{0}=s, s_{n}=t\) and \(s_{i} \in S_{\text {? }}\) for \(0 \leqslant i \leqslant n\). That is:
\[
\mathbf{A}^{n}(s, t)=\operatorname{Pr}\left(s \models S_{?} \mathbf{U}^{=n} t\right)
\]

If \(B=\varnothing\) and \(C=S\), then \(S_{=1}=S_{=0}=\varnothing\) and \(S_{\text {? }}=S\), yielding \(\mathbf{A}=\mathbf{P}\). The entry \(\mathbf{P}^{n}(s, t)\) (of the \(n\)th power of \(\mathbf{P}\) ) thus equals the probability of being in state \(t\) after \(n\) steps given that the computation starts in state \(s\), i.e., \(\mathbf{P}^{n}(s, t)=\operatorname{Pr}\left(s \models S \mathrm{U}^{=n} t\right)\). The probability of \(\mathcal{M}\) being in state \(t\) after exactly \(n\) transitions
\[
\Theta_{n}^{\mathcal{M}}(t)=\sum_{s \in S} \mathbf{P}^{n}(s, t) \cdot \iota_{\mathrm{init}}(s)
\]
is called the transient state probability for state \(t\). The function \(\Theta_{n}^{\mathcal{M}}\) is the transient state distribution. When considering \(\Theta_{n}^{\mathcal{M}}\) as the vector \(\left(\Theta_{n}^{\mathcal{M}}(t)\right)_{t \in S}\), the above equation can be rewritten as
\[
\Theta_{n}^{\mathcal{M}}=\underbrace{\mathbf{P} \cdot \mathbf{P} \cdot \ldots \cdot \mathbf{P}}_{n \text { times }} \cdot \iota_{\text {init }}=\mathbf{P}^{n} \cdot \iota_{\text {init }},
\]
where the initial distribution is viewed as a column vector. Due to the numerical instability of computing the \(n\)th power of a matrix, using, e.g., iterative squaring, it is recommended to compute \(\Theta_{n}^{\mathcal{M}}\) by successive matrix-vector multiplication:
\[
\Theta_{0}^{\mathcal{M}}=\iota_{\text {init }} \quad \text { and } \quad \Theta_{n+1}^{\mathcal{M}}=\mathbf{P} \cdot \Theta_{n}^{\mathcal{M}} \text { for } n \geqslant 0
\]

Transient state probabilities are thus special instances of constrained reachability probabilities. In fact, constrained reachability probabilities in \(\mathrm{MC} \mathcal{M}\) coincide with transient state probabilities in a slightly modified Markov chain. Let us first illustrate this by means of simple, i.e., unconstrained, step-bounded reachability probabilities in \(\mathcal{M}\), say the event \(\diamond^{\leqslant n} B\). The MC \(\mathcal{M}\) is modified such that all states \(s \in B\) are made absorbing. That is, all outgoing transitions of \(s \in B\) are replaced by a self-loop. This yields the MC \(\mathcal{M}_{B}\). The intuition of this transformation is that once a path reaches a state in \(B\), its subsequent states are of no importance to \(\operatorname{Pr}^{\mathcal{M}}\left(s \models \diamond^{\leqslant n} B\right)\). Formally, \(\mathcal{M}_{B}=\left(S, \mathbf{P}_{B}, L\right)\) with \(S\) and \(L\) as for \(\mathcal{M}\), and \(\mathbf{P}_{B}\) is defined by \(\mathbf{P}_{B}(s, t)=\mathbf{P}(s, t)\) if \(s \notin B, \mathbf{P}_{B}(s, s)=1\), and \(\mathbf{P}_{B}(s, t)=0\) for \(s \in B\) and \(s \neq t\). It then follows for any \(s \in S\) :
\[
\operatorname{Pr}^{\mathcal{M}}\left(s \models \diamond^{\leqslant n} B\right)=\operatorname{Pr}^{\mathcal{M}_{B}}\left(s \models \diamond^{=n} B\right) .
\]

The probability of reaching a \(B\)-state within \(n\) steps in \(\mathcal{M}\) is now given by:
\[
\operatorname{Pr}^{\mathcal{M}}\left(\diamond^{\leqslant n} B\right)=\sum_{t \in B} \Theta_{n}^{\mathcal{M}_{B}}(t)
\]

The reachability probability \(\operatorname{Pr}^{\mathcal{M}}\left(\diamond^{\leqslant n} B\right)\) in the MC \(\mathcal{M}\) thus agrees with the cumulative transient state probability for the \(B\)-states in the MC \(\mathcal{M}_{B}\).

Using similar arguments, one can show that the computation of step-bounded constrained reachability probabilities \(\operatorname{Pr}^{\mathcal{M}}\left(C \mathrm{U}^{\leqslant n} B\right)\) can be reduced to the problem of computing transient state probabilities in a slightly modified Markov chain. As for the simple reachability probabilities, all states in \(B\) are made absorbing. In addition, all states in \(S \backslash(C \cup B)\) are made absorbing. This is justified by the fact that paths that visit some state in \(S \backslash(C \cup B)\) contribute probability zero to \(\operatorname{Pr}^{\mathcal{M}}(s \neq C \cup \leqslant n B)\). We thus consider the \(\mathrm{MC} \mathcal{M}^{\prime}=\mathcal{M}_{B \cup(S \backslash(C \cup B))}\) that results from \(\mathcal{M}\) by making all states in \(B \cup(S \backslash(C \cup B))\) absorbing. It then follows that for any state \(s \in S\) :
\[
\operatorname{Pr}^{\mathcal{M}}\left(s \models C \mathrm{U}^{\leqslant n} B\right)=\operatorname{Pr}^{\mathcal{M}^{\prime}}\left(s \models \diamond^{=n} B\right) .
\]

The probability of reaching a \(B\)-state within \(n\) steps in \(\mathcal{M}\) via \(C\)-states only is now given by
\[
\operatorname{Pr}^{\mathcal{M}}(C \mathrm{U} \leqslant n B)=\sum_{t \in B} \Theta_{n}^{\mathcal{M}^{\prime}}(t)
\]

\section*{Example 10.23. Reachability by Transient Probabilities}

Consider again the Markov chain modeling the craps game (see Figure 10.3, page 751). Let \(C \mathrm{U}^{\leqslant n} B\) be the event of interest where \(B=\{\) won \(\}, C=\{\) start, \(4,5,6\}\). According to the described procedure just above, all states in \(B\) and all states that are neither in \(B\) nor in \(C\) are made absorbing. This yields the MC depicted in Figure 10.5. The constrained


Figure 10.5: Markov chain for the craps game for \(C \mathrm{U}{ }^{\leqslant n} B\).
reachability probabilities for \(C \mathrm{U} \leqslant n B\) are obtained as follows. As before, let \(\iota_{\text {init }}\) (start) \(=1\) and \(\iota_{\text {init }}(s)=0\) for any other state. For \(n=0\), we have
\[
\operatorname{Pr}\left(\text { start } \models C U^{\leqslant 0} B\right)=\Theta_{0}(\text { won })=\iota_{\text {init }}(\text { won })=0
\]

For \(n=1\), one obtains
\[
\operatorname{Pr}\left(\text { start } \models C \mathrm{U}^{\leqslant 1} B\right)=\iota_{\text {init }}(\text { start }) \cdot \mathbf{P}(\text { start }, \text { won })=\frac{2}{9} .
\]

For \(n=2\), one obtains \(\operatorname{Pr}\left(\right.\) start \(\left.\models C \mathrm{U}^{\leqslant 2} B\right)=\iota_{\text {init }}(\) start \() \cdot \mathbf{P}^{2}\) (start, won) which equals \(\frac{338}{36^{2}}\). For other \(n\), the probabilities are obtained in a similar way.

\subsection*{10.1.2 Qualitative Properties}

In the previous section, we have shown different ways to compute reachability probabilities in Markov chains. This section is focused on qualitative properties of Markov chains. Such properties typically require certain events to happen with probability 1 , or more generally, to compute all states where a certain event holds almost surely. Dually, the problem
to check whether an event occurs with zero probability is a qualitative property. This section will show that qualitative properties for events such as reachability, constrained reachability, repeated reachability - can a certain set of states be visited repeatedly? - and persistence - can only a certain set of states be visited from some moment on?-can all be verified using graph analysis, i.e., by just considering the underlying digraph of the Markov chain and ignoring the transition probabilities. This is due to a fundamental result for finite Markov chains stating that almost surely the computations will enter a bottom strongly connected component (an SCC that once entered cannot be left anymore) and visit each of its states infinitely often. This result will be proven in detail. Finally, it will be shown that this result does not apply to infinite Markov chains.

The first thing to be settled is that the events to be considered, such as repeated reachability and persistence, are measurable.

\section*{Remark 10.24. Measurability of Repeated Reachability and Persistence}

Let \(B \subseteq S\) be a set of states in a Markov chain \(\mathcal{M}\). The set of all paths that visit \(B\) infinitely often is measurable. This holds in arbitrary (possibly infinite) Markov chains and follows from the fact that the event \(\square \diamond B\) can be written as the countable intersection of countable unions of cylinder sets:
\[
\square \diamond B=\bigcap_{n \geqslant 0} \bigcup_{m \geqslant n} C y l("(m+1) \text { st state is in } B ")
\]
where \(C y l\left(\right.\) " \((m+1)\) st state is in \(B\) ") denotes the union of all cylinder sets \(C y l\left(t_{0} t_{1} \ldots t_{m}\right)\) spanned by a finite path fragment of length \(m\) that ends in a \(B\)-state. That is, \(t_{0} t_{1} \ldots t_{m} \in\) \(\operatorname{Paths}_{f i n}(\mathcal{M})\) and \(t_{m} \in B\). Let us prove the equality in the above equation.
1. \(\subseteq\) : Let \(\pi=s_{0} s_{1} s_{2} \ldots\) be a path in \(\mathcal{M}\) such that \(\pi \models \square \diamond B\). Thus, for all indices \(n\) there exists some \(m \geqslant n\) such that \(s_{m} \in B\). But then \(s_{0} s_{1} \ldots \ldots s_{m}\) is a finite path fragment which ends in some state in \(T\) and whose cylinder set contains \(\pi\). Thus, for all \(n \geqslant 0\) there exists \(m \geqslant n\) such that
\[
\pi \in \operatorname{Cyl}\left(s_{0} \ldots s_{m}\right) \subseteq \operatorname{Cyl}("(m+1) \text {-st state is in } B ")
\]

This shows that \(\pi\) belongs to the set on the right.
2. \(\supseteq\) : Let \(\pi=s_{0} s_{1} s_{2} \ldots\) be a path in \(\bigcap_{n \geqslant 0} \bigcup_{m \geqslant n} C y l\) (" \((m+1)\) st state \(\in B\) "). Thus, for each \(n\) there exists some \(m \geqslant n\) and a finite path fragment \(t_{0} t_{1} \ldots t_{m}\) such that \(t_{m} \in B\) and \(\pi \in \operatorname{Cyl}\left(t_{0} t_{1} \ldots t_{m}\right)\). But then \(t_{0} t_{1} \ldots t_{m}\) is a prefix of \(\pi\) and \(s_{m}=t_{m} \in B\). As this holds for any \(n\), it follows that \(\pi \models \square \diamond B\).

These arguments can be generalized to establish the measurability of events stating that finite path fragments, like \(\widehat{\pi}\), appear infinitely often. This can be seen as follows. Let
\(\widehat{\pi}=t_{0} \ldots t_{k}\) be a finite path fragment in \(\mathcal{M}\) and \(C y l\) (" \(\widehat{\pi}\) is taken from the \(m\) th state") the union of all cylinder sets \(C y l\left(s_{0} \ldots s_{m-1} t_{0} \ldots t_{k}\right)\). Then, the set of paths satisfying \(\square \diamond \widehat{\pi}\), i.e., the set of paths that contain infinitely many occurrences of \(\widehat{\pi}\) is measurable, since it is given by:
\[
\bigcap_{n \geqslant 0} \bigcup_{m \geqslant n} C y l(" \widehat{\pi} \text { is taken from the } m \text { th state"). }
\]

Now let \(\Pi\) be a set of finite path fragments. The above yields that the event \(\Lambda_{\hat{\pi} \in \Pi} \square \diamond \widehat{\pi}\), stating that each finite path fragment \(\widehat{\pi} \in \Pi\) is taken infinitely often, is measurable too. Note that since the state space \(S\) of a Markov chain is countable (see Definition 10.1, page 747), Paths \(_{f i n}(\mathcal{M})\) is countable.

Let us now consider persistence properties, i.e., events of the form \(\diamond \square B\). Such events consist of all paths \(\pi=s_{0} s_{1} s_{2} \ldots\) such that, for some \(n \geqslant 0\), the suffix \(s_{n} s_{n+1} \ldots\) only contains states in \(B\). The event \(\nabla \square B\) is measurable as it is the complement of the measurable event \(\square \diamond(S \backslash B)\).

While the infinite repetition of a nondeterministic choice in a transition system does not impose any restrictions on the sequence of selected alternatives, randomization (as in Markov chains) somehow implies strong fairness for all transitions. This follows from the following theorem which states that under the assumption that a certain state \(t\), say, of a Markov chain is visited infinitely often, then almost surely all finite path fragments \(t_{0} t_{1} \ldots t_{n}\) that start in \(t\) (i.e., \(t_{0}=t\) ) will be taken infinitely often too. Here, the notion "almost surely" refers to the conditional probabilities, under the condition that \(t\) is visited infinitely often. Formally, an event \(E\) is said to hold almost surely under the condition of another event \(D\) if \(\operatorname{Pr}(D)=\operatorname{Pr}(E \cap D)\).

\section*{Theorem 10.25. Probabilistic Choice as Strong Fairness}

For (possibly infinite) Markov chain \(\mathcal{M}\) and \(s\), \(t\) states in \(\mathcal{M}\) :
\[
\operatorname{Pr}^{\mathcal{M}}(s \models \square \diamond t)=\operatorname{Pr}_{s}^{\mathcal{M}}\left(\bigwedge_{\widehat{\pi} \in{\operatorname{Paths} f_{f i}}(t)} \square \diamond \widehat{\pi}\right)
\]
where \(\bigwedge_{\widehat{\pi} \in \operatorname{Paths}_{f n}(t)} \square \diamond \widehat{\pi}\) denotes the set of all paths \(\pi\) such that any path fragment \(\widehat{\pi} \in\) Paths \(_{\text {fin }}(t)\) occurs infinitely often in \(\pi\).

In particular, for each state \(t \in S\) and \(u \in \operatorname{Post}(t)\) the event "transition \(t \rightarrow u\) is taken infinitely often" holds almost surely given that \(t\) is visited infinitely often. In this sense, executions of Markov chains are strongly fair with respect to all probabilistic choices.

Proof: The proof is provided in three steps.
1. We first prove that for any \(\widehat{\pi} \in \operatorname{Paths}_{f i n}(t)\) it holds that:
\[
\operatorname{Pr}(s \models \square \diamond t)=\operatorname{Pr}(s \models \diamond \widehat{\pi}) .
\]

Let \(p=\mathbf{P}(\widehat{\pi})\). As \(\widehat{\pi} \in \operatorname{Paths}_{f i n}(t)\), it follows that \(0<p \leqslant 1\). Let \(E_{n}(\widehat{\pi})\) be the event "visit \(t\) at least \(n\) times, but never take the path fragment \(\widehat{\pi}\) ". Note that \(E_{1}(\widehat{\pi}) \supseteq E_{2}(\widehat{\pi}) \supseteq \ldots\) Furthermore:
\[
\operatorname{Pr}_{s}\left(E_{n}(\widehat{\pi})\right) \leqslant(1-p)^{n}
\]

Let the event \(E(\widehat{\pi})=\bigcap_{n \geqslant 1} E_{n}(\widehat{\pi})\). That is, \(E(\widehat{\pi})\) is the event that \(t\) is visited infinitely often, but the path fragment \(\widehat{\pi}\) is never taken. As \(E_{1}(\widehat{\pi}) \supseteq E_{2}(\widehat{\pi}) \supseteq \ldots\) and
\[
\operatorname{Pr}_{s}(E(\widehat{\pi}))=\lim _{n \rightarrow \infty} \operatorname{Pr}_{s}\left(E_{n}(\widehat{\pi})\right) \leqslant \lim _{n \rightarrow \infty}(1-p)^{n}=0
\]
we have that:
\[
\operatorname{Pr}(s \models \square \diamond t \wedge \text { "path fragment } \widehat{\pi} \text { is never taken" })=0
\]
2. Now consider the event \(F_{n}(\widehat{\pi})\), for \(n \geqslant 0\), that represents the fact that state \(t\) is visited infinitely often, while the path fragment \(\widehat{\pi}\) will not be taken anymore from the \(n\)th state on. The probability for the event \(F_{n}(\widehat{\pi})\) is given by
\[
\operatorname{Pr}_{s}\left(F_{n}(\widehat{\pi})\right)=\sum_{s^{\prime} \in S} \operatorname{Pr}(\underbrace{s \models \bigcirc^{n} s^{\prime}}_{\text {the } n \text {-th state is } s^{\prime}}) \cdot \underbrace{\operatorname{Pr}_{s^{\prime}}(E(\widehat{\pi}))}_{=0}=0 .
\]

Now consider the event
\[
F(\widehat{\pi})=\square \diamond t \wedge \text { "from some moment on, the path fragment } \widehat{\pi} \text { is never taken". }
\]

It follows that \(F(\widehat{\pi})=\bigcup_{n \geqslant 1} F_{n}(\widehat{\pi})\). Since \(F_{1} \subseteq F_{2} \subseteq \ldots\) it follows that
\[
\operatorname{Pr}_{s}(F(\widehat{\pi}))=\lim _{n \rightarrow \infty} \operatorname{Pr}_{s}\left(F_{n}(\widehat{\pi})\right)=0
\]

Hence:
\[
\begin{aligned}
& \operatorname{Pr}(s \models \square \diamond t \wedge \square \diamond \text { "path fragment } \widehat{\pi} \text { is taken" }) \\
= & \operatorname{Pr}(s \models \square \diamond t)-\operatorname{Pr}_{s}(F(\widehat{\pi})) \\
= & \operatorname{Pr}(s \models \square \diamond t) .
\end{aligned}
\]
3. We now generalize this result to all finite path fragments starting in state \(t\). Let the event
\[
F=\bigcup_{\widehat{\pi} \in P_{a t h s f i n}(t)} F(\widehat{\pi})
\]

As the set of finite path fragments is countable we have
\[
\operatorname{Pr}_{s}(F) \leqslant \sum_{\widehat{\pi}} \operatorname{Pr}_{s}(F(\widehat{\pi}))=0
\]
where \(\widehat{\pi}\) ranges over all finite path fragments starting in \(t\). Hence, \(\operatorname{Pr}_{s}(F)=0\).

Thus, under the condition that state \(t\) is visited infinitely often, almost surely any finite path fragment starting in \(t\) will be taken infinitely often.

As a direct consequence of Theorem 10.25, it follows that any successor of \(t\) is visited infinitely often:
\[
\operatorname{Pr}^{\mathcal{M}}(s \models \square \diamond t)=\operatorname{Pr}^{\mathcal{M}}\left(s \models \bigwedge_{u \in \operatorname{Post*}^{*}(t)} \square \diamond u\right) .
\]

Hence, for any state \(s\) it holds that
\[
\operatorname{Pr}\left(s \models \bigwedge_{t \in S} \bigwedge_{u \in \operatorname{Post}^{*}(t)}(\square \diamond t \rightarrow \square \diamond u)\right)=1
\]

\section*{Notation 10.26. Graph Notations for Markov Chains}

Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite MC. In the sequel, we often use graph-theoretical notations for MCs which refer to the underlying digraph of \(\mathcal{M}\). For example, a subset \(T\) of \(S\) is called strongly connected if for each pair \((s, t)\) of states in \(T\) there exists a path fragment \(s_{0} s_{1} \ldots s_{n}\) such that \(s_{i} \in T\) for \(0 \leqslant i \leqslant n\), \(s_{0}=s\) and \(s_{n}=t\). A strongly connected component (SCC, for short) of \(\mathcal{M}\) denotes a strongly connected set of states such that no proper superset of \(T\) is strongly connected. A bottom SCC (BSCC, for short) of \(\mathcal{M}\) is an SCC \(T\) from which no state outside \(T\) is reachable, i.e., for each state \(t \in T\) it holds that \(\mathbf{P}(t, T)=1\). (Recall that \(\mathbf{P}(s, T)=\sum_{t \in T} \mathbf{P}(s, t)\).) Let \(B S C C(\mathcal{M})\) denote the set of all BSCCs of the underlying digraph of \(\mathcal{M}\).

Let us now apply Theorem 10.25 to finite MCs. They enjoy the property that at least one state is visited infinitely often on each infinite path. The following result asserts that the set of states that is visited infinitely often on a path almost surely forms a BSCC. Before presenting this result, we have to ensure that the event "being in a given BSCC" is measurable. This can be seen as follows. Recall that \(\inf (\pi)\) denotes the set of states that are visited infinitely often along \(\pi\). The event \(\inf (\pi)=T\) for some BSCC \(T\) is measurable as it can be written as a finite intersection of measurable sets (cf. Remark 10.24 on page 771):
\[
\bigwedge_{t \in T} \square \diamond t \wedge \diamond \square T
\]

The set of paths for which \(\inf (\pi)\) is a BSCC agrees with the measurable event
\[
\bigvee_{T \in B S C C(\mathcal{M})}(\inf (\pi)=T)
\]

\section*{Theorem 10.27. Limit Behavior of Markov Chains}

For each state s of a finite Markov chain \(\mathcal{M}\) :
\[
\operatorname{Pr}_{s}^{\mathcal{M}}\{\pi \in \operatorname{Paths}(s) \mid \inf (\pi) \in B S C C(\mathcal{M})\}=1
\]

Proof: For each path \(\pi\), the set \(\inf (\pi)\) is strongly connected, and thus contained in some SCC of \(\mathcal{M}\). Hence, for each state \(s\) :
\[
\begin{equation*}
\sum_{T} \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \inf (\pi)=T\}=1 \tag{*}
\end{equation*}
\]
where \(T\) ranges over all nonempty \(\operatorname{SCCs}\) of \(\mathcal{M}\). Assume that \(\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \inf (\pi)=\) \(T\}\) is positive. By Theorem 10.25, almost all paths \(\pi\) with \(\inf (\pi)=T\) fulfill
\[
\operatorname{Post}^{*}(T)=\operatorname{Post}^{*}(\inf (\pi)) \subseteq \inf (\pi)=T
\]

Hence, \(T=\operatorname{Post}^{*}(T)\), i.e., \(T\) is a BSCC. The claim now follows from (*).

Stated in words, Theorem 10.27 states that almost surely any finite Markov chain eventually reaches a BSCC and visits all its states infinitely often.

Example 10.28. Zeroconf Protocol (Revisited)
Consider the Markov chain for the zeroconf protocol; see Example 10.5 (page 751). This MC has two BSCCs, namely \(\left\{s_{8}\right\}\) and \(\left\{s_{6}\right\}\). According to the previous theorem, any infinite path will almost surely lead to one of these BSCCs. This means in particular that the probability of infinitely often attempting to acquire a new address (after receiving no response on a probe) equals zero.

The previous theorem is a central result for the analysis of many types of properties of finite MCs. In the following, we consider three important almost sure properties: reachability, repeated reachability, and persistence properties. The problem of almost sure reachability amounts to determining the set of states that reach a certain given set of states \(B\), say, almost surely. The following result provides the foundations for such a computation.

\section*{Theorem 10.29. Almost Sure Reachability}

Let \(\mathcal{M}\) be a finite Markov chain with state space \(S, s \in S\) and \(B \subseteq S\) a set of absorbing states. Then, the following statements are equivalent:
(a) \(\operatorname{Pr}(s \models \diamond B)=1\).
(b) \(\operatorname{Post}^{*}(t) \cap B \neq \varnothing\) for each state \(t \in \operatorname{Post}^{*}(s)\).
(c) \(s \in S \backslash \operatorname{Pre}^{*}\left(S \backslash \operatorname{Pre}^{*}(B)\right)\).

In particular, \(\{s \in S \mid \operatorname{Pr}(s \mid \diamond B)=1\}=S \backslash \operatorname{Pre}^{*}\left(S \backslash \operatorname{Pre}^{*}(B)\right)\).

\section*{Proof:}
1. \((\mathrm{a}) \Longrightarrow(\mathrm{b}):\) Let \(t \in \operatorname{Post}^{*}(s)\) such that \(\operatorname{Post}^{*}(t) \cap B=\varnothing\), i.e., \(t\) is a successor of \(s\) from which \(B\) cannot be reached. Then: \(\operatorname{Pr}(s \models \diamond B) \leqslant 1-\operatorname{Pr}(s \models \diamond t)<1\). This shows that if (b) is violated, (a) is so too.
2. \((\mathrm{b}) \Longleftrightarrow(\mathrm{c})\) : By definition of Post and Pre it follows that for each \(u \in S\) and \(C \subseteq S\), \(\operatorname{Post}^{*}(u) \cap C \neq \varnothing\) if and only if \(u \in \operatorname{Pre}^{*}(C)\). Thus:
\[
\begin{array}{ll} 
& \operatorname{Post}^{*}(t) \cap B \neq \varnothing \text { for any state } t \in \operatorname{Post}^{*}(s) \\
\text { iff } & \operatorname{Post}^{*}(s) \subseteq \operatorname{Pre}^{*}(B) \\
\text { iff } & \operatorname{Post}^{*}(s) \cap\left(S \backslash \operatorname{Pre}^{*}(B)\right)=\varnothing \\
\text { iff } & s \notin \operatorname{Pre}^{*}\left(S \backslash \operatorname{Pre}^{*}(B)\right) \\
\text { iff } & s \in S \backslash \operatorname{Pre}^{*}\left(S \backslash \operatorname{Pre}^{*}(B)\right) .
\end{array}
\]
3. \((\mathrm{b}) \Longrightarrow(\mathrm{a})\) : Assume \(\operatorname{Post}^{*}(t) \cap B \neq \varnothing\) for any successor \(t\) of \(s\). By Theorem 10.27, almost all paths \(\pi \in \operatorname{Paths}(s)\) reach a BSCC. Since each state in \(B\) is absorbing, each BSCC \(T\) of \(\mathcal{M}\) is either of the form \(T=\{t\}\) with \(t \in B\) or it satisfies \(T \cap B=\varnothing\). Let us show that the latter case cannot occur. Consider \(T \cap B=\varnothing\) for BSCC \(T\). As \(T\) is a BSCC, \(\operatorname{Post}^{*}(u) \cap B=\varnothing\) for each \(u \in T\). However, as \(B\) is reachable from any successor of \(s\), there is no \(\operatorname{BSCC} T\) with \(T \cap B=\varnothing\) that is reachable from \(s\). Hence, from state \(s\) almost surely a state in \(B\) will be reached.

Thus, if all states in a finite Markov chain can reach a set \(B\) of absorbing states, then \(B\) will be reached almost surely from each state. This, however, also implies that almost surely \(B\) will be visited infinitely often, as stated by the following corollary:

\section*{Corollary 10.30. Global Almost Sure Reachability}

Let \(\mathcal{M}\) be a finite Markov chain with state space \(S\), and \(B \subseteq S\). Then:
\[
s \models \exists \diamond B \text { for any } s \in S \quad \text { implies } \operatorname{Pr}(s \models \square \diamond B)=1 \text { for all } s \in S
\]

Theorem 10.29 suggests the following algorithm to determine the set of states that reach a certain set \(B\) of states almost surely in a finite \(\mathrm{MC} \mathcal{M}\) :
1. Make all states in \(B\) absorbing. This yields MC \(\mathcal{M}_{B}\).
2. Determine the set \(S \backslash \operatorname{Pre}^{*}\left(S \backslash \operatorname{Pre}^{*}(B)\right)\) by a graph analysis in \(\mathcal{M}_{B}\). This can be done by a backward search from \(B\) to compute \(\operatorname{Pre}^{*}(B)\) followed by a backward search from \(S \backslash \operatorname{Pre}^{*}(B)\) to determine \(\operatorname{Pre}^{*}\left(S \backslash \operatorname{Pre}^{*}(B)\right)\). Both searches are carried out on \(\mathcal{M}_{B}\).

Thus, the time complexity of determining the set of states \(s\) for which \(\operatorname{Pr}(s \models \diamond B)=1\) is linear in the size of \(\mathcal{M}\).

Now consider a constrained reachability condition \(C \cup B\) where \(C\) and \(B\) are subsets of the state space of a finite MC \(\mathcal{M}\).

\section*{Corollary 10.31. Qualitative Constrained Reachability}

For finite Markov chain \(\mathcal{M}\) with state space \(S\) and \(B, C \subseteq S\), the sets
\[
S_{=0}=\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=0\} \quad \text { and } \quad S_{=1}=\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=1\}
\]
can be computed in time \(\mathcal{O}(\operatorname{size}(\mathcal{M}))\).

Proof: Let \(\mathcal{M}\) be a finite MC and \(B, C \subseteq S\).
1. The set \(S_{=0}=\{s \in S \mid \operatorname{Pr}(s \models C \cup B)=0\}\) agrees with the complement of the satisfaction set of the CTL formula \(\exists(C \cup B)\). This set can be computed by means of a backward analysis starting from the \(B\)-states in time linear in the size of \(\mathcal{M}\).
2. A linear-time algorithm for the computation of \(S_{=1}=\{s \in S \mid \operatorname{Pr}(s \mid C \cup B)=1\}\) is obtained as follows. In fact, the problem of calculating \(S_{=1}\) can be solved by a reduction to the problem of computing the set of states that almost surely eventually
reach \(B\) in a slightly modified Markov chain. The idea is to make all \(B\)-states absorbing and do the same for all states in \(S \backslash(C \cup B)\). To that end, \(\mathcal{M}\) is changed into the Markov chain \(\mathcal{M}^{\prime}\) with state space \(S\) and transition probability function \(\mathbf{P}^{\prime}\) given by
\[
\mathbf{P}^{\prime}(s, t)= \begin{cases}1 & : \quad \text { if } s \in B \cup S \backslash(C \cup B) \\ \mathbf{P}(s, t) & : \text { otherwise }\end{cases}
\]

The justification for the transformation of \(\mathcal{M}\) into \(\mathcal{M}^{\prime}\) is given by
- \(\operatorname{Pr}^{\mathcal{M}}(s \models C \cup B)=\operatorname{Pr}^{\mathcal{M}}(s \models \diamond B)\) for all states \(s \in C \backslash B\),
- \(\operatorname{Pr}^{\mathcal{M}}(s \models C \cup B)=\operatorname{Pr}^{\mathcal{M}^{\prime}}(s \models \Delta B)=1\) for all states \(s \in B\),
- \(\operatorname{Pr}^{\mathcal{M}}(s \models C \cup B)=\operatorname{Pr}^{\mathcal{M}^{\prime}}(s \models \diamond B)=0\) for all states \(s \in S \backslash(C \cup B)\).

Thus, the probabilities for the constrained reachability property \(C \cup B\) in \(\mathcal{M}\) can be determined by computing the reachability probabilities \(\diamond B\) in \(\mathcal{M}^{\prime}\). The latter can be done in time linear in the size of \(\mathcal{M}^{\prime}\) (see above), which is bounded from above by that of \(\mathcal{M}\). (Recall that \(\operatorname{size}(\mathcal{M})\) is the number of states and transitions in \(\mathcal{M}\).)

\section*{Example 10.32. Qualitative Constrained Reachability}

Consider again the craps game, see Example 10.4 (page 750). As before, let \(B=\{\) won \(\}\) and \(C=\{\) start, \(4,5,6\}\) and consider the event \(C \cup B\). We have
\[
S_{=0}=S \backslash \operatorname{Sat}(\exists(C \cup B))=\{\text { lost, } 8,9,10\}
\]

To determine the set \(S_{=1}\) all states in \(B\) and in \(S \backslash(B \cup C)\) are made absorbing. This yields the MC in Figure 10.5 (page 770). The states that almost surely reach the state won in this MC are \(\{4,5,6\), won \(\}\). These are the states in \(S_{=1}\).

Using the above result, the qualitative model-checking problem to determine whether \(\operatorname{Pr}^{\mathcal{M}}(C \cup B)=1\) can be solved by a graph analysis in time linear in the size of \(\mathcal{M}\). In fact, the same holds for repeated reachability events:

\section*{Corollary 10.33. Qualitative Repeated Reachability}

Let \(\mathcal{M}\) be a finite Markov chain with state space \(S, B \subseteq S\), and \(s \in S\). Then, the following statements are equivalent:
(a) \(\operatorname{Pr}(s \models \square \diamond B)=1\).
(b) \(T \cap B \neq \varnothing\) for each \(B S C C T\) that is reachable from \(s\).
(c) \(s \models \forall \square \exists \diamond B\).

Proof: The equivalence of (a) and (b) follows immediately from the fact that from state \(s\) almost surely a BSCC \(T\) will be reached and all states of \(T\) will be visited infinitely often (see Theorem 10.27). The equivalence of (a)/(b) and (c) is left as an exercise to the reader (see Exercise 10.5 on page 901).

As a consequence, to check whether \(\operatorname{Pr}^{\mathcal{M}}(\square \diamond B)=1\), it suffices to analyze the BSCCs in \(\mathcal{M}\), which can be done in linear time. The following result asserts that repeated reachability probabilities can be computed in time polynomial in the size of \(\mathcal{M}\). This is achieved by a reduction to the problem of computing the probabilities for eventually visiting certain BSCCs:

\section*{Corollary 10.34. Quantitative Repeated Reachability}

Let \(\mathcal{M}\) be a finite Markov chain with state space \(S, B \subseteq S\), and \(s \in S\), and \(U\) be the union of all BSCCs \(T\) (in \(\mathcal{M}\) ) with \(T \cap B \neq \varnothing\). Then:
\[
\operatorname{Pr}(s \models \square \diamond B)=\operatorname{Pr}(s \models \diamond U)
\]

In a similar way, by an analysis of the BSCCs it can also be checked whether a strong or weak fairness constraint (or other \(\omega\)-regular liveness property) holds almost surely. For persistence property \(\diamond \square B\) with \(B \subseteq S\), for instance, we have
- \(\operatorname{Pr}(s \models \diamond \square B)=1\) iff \(T \subseteq B\) for each BSCC \(T\) that is reachable from \(s\).
- \(\operatorname{Pr}(s \models \diamond \square B)=\operatorname{Pr}(s \models \diamond V)\) where \(V\) is the union of all BSCCs \(T\) with \(T \subseteq B\).

The conclusion of this section is that checking a qualitative property, such as reachability, repeated reachability, and persistence on a finite Markov chain can be achieved by a graph analysis. The transition probabilities thus are of no importance for this purpose! As the following example indicates, this does not hold for infinite Markov chains.

\section*{Remark 10.35. Qualitative Properties in Infinite Markov Chains}

We state without proof that the above considerations on the limiting behavior of finite Markov chains do not hold for infinite Markov chains. In fact, even for strongly connected infinite Markov chains it is possible that the probability of visiting a certain state infinitely
often is zero. Thus, it is possible that \(\operatorname{Pr}(s \models \diamond T)>0\) for all states \(s\), while \(\operatorname{Pr}(s \models\) \(\square \diamond T)=0\). Typical examples are so-called one-dimensional random walks. A concrete example is the infinite Markov chain \(\mathcal{M}_{p}\) where \(p\) is a rational number in \(] 0,1[\). The state space of \(\mathcal{M}_{p}\) is \(S=\mathbb{N}\) with \(L(0)=\{0\}\) and \(L(n)=\varnothing\) for \(n>0\). The transition probabilities in \(\mathcal{M}_{p}\) are given by (see Figure 10.6):
\[
\begin{array}{llll}
\mathbf{P}(n, n+1)=p & \text { and } & \mathbf{P}(n, n-1)=1-p & \text { for all } n>0 \\
\mathbf{P}(0,0)=1-p & \text { and } & \mathbf{P}(0,1)=p
\end{array}
\]

It can be shown that for \(p \leqslant \frac{1}{2}\), the leftmost state is almost surely eventually reached, and almost surely infinitely often visited, i.e.:
\[
\operatorname{Pr}(n \models \diamond 0)=\operatorname{Pr}(n \models \square \diamond 0)=1 \quad \text { for all } n \geqslant 0 .
\]

However, for \(p>\frac{1}{2}\), it is more likely to move to the right (i.e., to move from state \(n\) to state \(n+1)\) than to the left, and it can be shown that \(\operatorname{Pr}(n \models \diamond 0)<1\) and \(\operatorname{Pr}(n \models \square \diamond 0)=0\) for all \(n>0\). (Further details on infinite Markov chains and random walks can be found in most textbook on Markov chains; see e.g., [63, 238, 248].)


Figure 10.6: Infinite Markov chain for a one-dimensional random walk.

\subsection*{10.2 Probabilistic Computation Tree Logic}

Probabilistic computation tree logic (PCTL, for short) is a branching-time temporal logic, based on the logic CTL (see Chapter 6). A PCTL formula formulates conditions on a state of a Markov chain. The interpretation is Boolean, i.e., a state either satisfies or violates a PCTL formula. The logic PCTL is defined like CTL with one major difference. Instead of universal and existential path quantification, PCTL incorporates, besides the standard propositional logic operators, the probabilistic operator \(\mathbb{P}_{J}(\varphi)\) where \(\varphi\) is a path formula and \(J\) is an interval of \([0,1]\). The path formula \(\varphi\) imposes a condition on the set of paths, whereas \(J\) indicates a lower bound and/or upper bound on the probability. The intuitive meaning of the formula \(\mathbb{P}_{J}(\varphi)\) in state \(s\) is: the probability for the set of paths satisfying \(\varphi\) and starting in \(s\) meets the bounds given by \(J\). The probabilistic operator can be considered as the quantitative counterpart to the CTL path quantifiers \(\exists\) and \(\forall\). The CTL formulae \(\exists \varphi\) and \(\forall \varphi\) assert the existence of certain paths and the absence of paths where a certain condition does not hold respectively. They, however, do not impose any
constraints on the likelihood of the paths that satisfy the condition \(\varphi\). Later on in this section, the relationship between the operator \(\mathbb{P}_{J}(\varphi)\) and universal and existential path quantification is elaborated in detail.

The path formulae \(\varphi\) are defined as for CTL, except that a bounded until operator is additionally incorporated. The intuitive meaning of the path formula \(\Phi U^{\leqslant n} \Psi\) for a natural number \(n\) is that a \(\Psi\) state should be reached within \(n\) transitions, and that all states prior to reaching the \(\Psi\)-state satisfy \(\Phi\).

\section*{Definition 10.36. Syntax of PCTL}

PCTL state formulae over the set AP of atomic propositions are formed according to the following grammar:
\[
\Phi::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\neg \Phi| \mathbb{P}_{J}(\varphi)
\]
where \(a \in A P, \varphi\) is a path formula and \(J \subseteq[0,1]\) is an interval with rational bounds. PCTL path formulae are formed according to the following grammar:
\[
\varphi::=\bigcirc \Phi\left|\Phi_{1} \cup \Phi_{2}\right| \Phi_{1} \cup \cup^{\leqslant n} \Phi_{2}
\]
where \(\Phi, \Phi_{1}\), and \(\Phi_{2}\) are state formulae and \(n \in \mathbb{N}\).

As in CTL, the linear temporal operators \(\bigcirc\) and \(U\) (and its bounded variant) are required to be immediately preceded by \(\mathbb{P}\). Rather than writing the intervals explicitly, often abbreviations are used; e.g., \(\mathbb{P}_{\leqslant 0.5}(\varphi)\) denotes \(\mathbb{P}_{[0,0.5]}(\varphi), \mathbb{P}_{=1}(\varphi)\) stands for \(\mathbb{P}_{[1,1]}(\varphi)\), and \(\mathbb{P}_{>0}(\varphi)\) denotes \(\mathbb{P}_{] 0,1]}(\varphi)\).

The propositional logic fragment of PCTL, as well as the path formulae \(\bigcirc \Phi\) and \(\Phi_{1} \cup \Phi_{2}\) has the same meaning as in CTL. Path formula \(\Phi_{1} U^{\leqslant n} \Phi_{2}\) is the step-bounded variant of \(\Phi_{1} \cup \Phi_{2}\). It asserts that the event specified by \(\Phi_{2}\) will hold within at most \(n\) steps, while \(\Phi_{1}\) holds in all states that are visited before a \(\Phi_{2}\)-state has been reached. Other Boolean connectives are derived in the usual way, e.g., \(\Phi_{1} \vee \Phi_{2}\) is obtained by \(\neg\left(\neg \Phi_{1} \wedge \neg \Phi_{2}\right)\). The eventually operator \((\diamond)\) can be derived as usual: \(\diamond \Phi=\) true \(U \Phi\). Similarly, for step-bounded eventually we have:
\[
\diamond \leqslant n \Phi=\operatorname{true} U{ }^{\leqslant n} \Phi
\]

A path satisfies \(\diamond^{\leqslant n} \Phi\) if it reaches a \(\Phi\)-state within \(n\) steps.
The always operator can be derived using the duality of eventually and always (as in CTL and LTL) and the duality of lower and upper bounds. The latter means that an event \(E\)
holds with probability at most \(p\) if and only if the dual event \(\bar{E}=\operatorname{Paths}(\mathcal{M}) \backslash E\) holds with probability at least \(1-p\). Thus, it is possible to define, e.g.,
\[
\mathbb{P}_{\leqslant p}(\square \Phi)=\mathbb{P}_{\geqslant 1-p}(\diamond \neg \Phi) \quad \text { and } \quad \mathbb{P}_{] p, q]}\left(\square^{\leqslant n} \Phi\right)=\mathbb{P}_{[1-q, 1-p[ }\left(\diamond^{\leqslant n} \neg \Phi\right)
\]

Other temporal operators, like weak until \(W\) or release \(R\) (see Section 5.1.5, page 252) can be derived in an analogous way. This is left as an exercise to the reader; see Exercise 10.9 (page 902).

\section*{Example 10.37. Specifying Properties in PCTL}

Consider the simulation of a six-sided die by a fair coin as described in Example 10.3 (page 750). The PCTL formula
\[
\bigwedge_{1 \leqslant i \leqslant 6} \mathbb{P}_{=\frac{1}{6}}(\diamond i)
\]
expresses that each of the six possible outcomes of the die should occur with equal probability.

Consider a communication protocol that uses an imperfect channel that might lose messages. The PCTL formula
\[
\mathbb{P}_{=1}(\diamond \text { delivered }) \wedge \mathbb{P}_{=1}\left(\square\left(\text { try_to_send } \rightarrow \mathbb{P}_{\geqslant 0.99}(\diamond \leqslant 3 \text { delivered })\right)\right)
\]
asserts that almost surely some message will be delivered (first conjunct) and that almost surely for any attempt to send a message, with probability at least 0.99 , the message will be delivered within three steps.

Consider the craps game, see Example 10.4. The property "the probability of winning without ever rolling 8,9 , or 10 is at least 0.32 " is expressed by the PCTL formula:
\[
\mathbb{P}_{\geqslant 0.32}(\neg(8 \vee 9 \vee 10) \cup \text { won }) .
\]

Impatient players might be interested in this property but with a bounded number of rolls, say, five. This is expressed by
\[
\mathbb{P}_{\geqslant 0.32}\left(\neg(8 \vee 9 \vee 10) \mathrm{U}^{\leqslant 5} \text { won }\right)
\]

Finally, the PCTL formula
\[
\mathbb{P}_{\geqslant 0.32}\left(\neg(8 \vee 9 \vee 10) \mathrm{U}^{\leqslant 5} \mathbb{P}_{=1}(\square \text { won })\right)
\]
expresses in addition that almost surely the player will always win.

PCTL formulae are interpreted over the states and paths of a Markov chain \(\mathcal{M}\). For the state formulae, the satisfaction relation \(\vDash\) is a relation between states in \(\mathcal{M}\) and state
formulae. As usual, we write \(s \models \Phi\) rather than \((s, \Phi) \in \models\). The interpretation is as usual, e.g., \(s \models \Phi\) if and only if \(\Phi\) holds in \(s\). For the path formulae, \(\models\) is a relation between infinite path fragments in \(\mathcal{M}\) and path formulae.

\section*{Definition 10.38. Satisfaction Relation for PCTL}

Let \(a \in A P\) be an atomic proposition, \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov chain, state \(s \in S, \Phi, \Psi\) be PCTL state formulae, and \(\varphi\) be a PCTL path formula. The satisfaction relation \(\models\) is defined for state formulae by
\[
\begin{array}{lll}
s \models a & \text { iff } & a \in L(s), \\
s \models \neg \Phi & \text { iff } & s \not \models \Phi, \\
s \models \Phi \wedge \Psi & \text { iff } & s \models \Phi \text { and } s \models \Psi, \\
s \models \mathbb{P}_{J}(\varphi) & \text { iff } & P r(s \models \varphi) \in J .
\end{array}
\]

Here, \(\operatorname{Pr}(s \models \varphi)=\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \varphi\}\).
Given a path \(\pi\) in \(\mathcal{M}\), the satisfaction relation is defined (as for CTL):
\[
\begin{array}{lll}
\pi \models \bigcirc \Phi & \text { iff } & \pi[1] \models \Phi, \\
\pi \models \Phi \cup \Psi & \text { iff } & \exists j \geqslant 0 .(\pi[j] \models \Psi \wedge(\forall 0 \leqslant k<j . \pi[k] \models \Phi)), \\
\pi \models \Phi \cup \leqslant n \Psi & \text { iff } & \exists 0 \leqslant j \leqslant n .(\pi[j] \models \Psi \wedge(\forall 0 \leqslant k<j . \pi[k] \models \Phi))
\end{array}
\]
where for path \(\pi=s_{0} s_{1} s_{2} \ldots\) and integer \(i \geqslant 0, \pi[i]\) denotes the \((i+1)\)-st state of \(\pi\), i.e., \(\pi[i]=s_{i}\).

Let \(\operatorname{Sat}_{\mathcal{M}}(\Phi)\), or briefly \(\operatorname{Sat}(\Phi)\), denote \(\{s \in S \mid s \models \Phi\}\).
The semantics of the probability operator \(\mathbb{P}\) refers to the probability for the sets of paths for which a path formula holds. To ensure that this is well-defined, we need to establish that the events specified by PCTL path formulae are measurable, i.e., elements of the \(\sigma\) algebra \(\mathfrak{E}^{\mathcal{M}}\) (see Definition 10.10, page 758). As the set \(\{\pi \in \operatorname{Paths}(s)|\pi|=\varphi\}\) for PCTL path formula \(\varphi\) can be considered as a countable union of cylinder sets, its measurability is ensured. This follows from the following lemma.

\section*{Lemma 10.39. Measurability of PCTL Events}

For each PCTL path formula \(\varphi\) and states of a Markov chain \(\mathcal{M}\), the set \(\{\pi \in \operatorname{Paths}(s) \mid\) \(\pi \models \varphi\}\) is measurable.

Proof: For any PCTL path formula \(\varphi\), it will be proven that the set \(\operatorname{Paths}(s, \varphi)=\{\pi \in\)
\(\operatorname{Paths}(s) \mid \pi \models \varphi\}\) is measurable, i.e., belongs to the \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\). This is done by showing that this set can be considered as the countable union of cylinder sets in \(\mathfrak{E}^{\mathcal{M}}\).

There are three possibilities for \(\varphi\). If \(\varphi=\bigcirc \Phi\), then \(\operatorname{Paths}(s, \varphi)\) agrees with the union of the cylinder sets \(C y l(s t)\) where \(t \models \Phi\). For \(\varphi=\Phi U^{\leqslant n} \Psi\), the set \(\operatorname{Paths}(s, \varphi)\) is obtained by the union of all cylinder sets \(\operatorname{Cyl}\left(s_{0} s_{1} \ldots s_{k}\right)\) where \(k \leqslant n, s_{k} \models \Psi, s_{0}=s\), and \(s_{i} \models \Phi\) for \(0 \leqslant i<k\). For unbounded until, i.e., \(\varphi=\Phi \cup \Psi\), we have:
\[
\operatorname{Paths}(s, \varphi)=\bigcup_{n \geqslant 0}\left\{\pi \in \operatorname{Paths}(s) \mid \pi \models \Phi U^{\leqslant n} \Psi\right\}
\]

The equivalence of PCTL formulae is defined as for the other logics encountered so far: two state formulae are equivalent whenever their semantics are equal. Formally, for PCTL state formulae \(\Phi\) and \(\Psi\) :
\[
\begin{gathered}
\Phi \equiv \Psi \quad \text { iff } \quad \text { for all Markov chains } \mathcal{M} \text { and states } s \text { of } \mathcal{M} \text { it holds that: } \\
\qquad s \models \Phi \Longleftrightarrow s \models \Psi
\end{gathered}
\]
iff
\[
\operatorname{Sat}_{\mathcal{M}}(\Phi)=\operatorname{Sat}_{\mathcal{M}}(\Psi) \text { for all Markov chains } \mathcal{M}
\]

In addition to the equivalence rules for propositional logic, we have, e.g.,
\[
\mathbb{P}_{<p}(\varphi) \equiv \neg \mathbb{P}_{\geqslant p}(\varphi)
\]
where \(p \in] 0,1]\) is a rational number and \(\varphi\) an arbitrary PCTL path formula. This equivalence follows directly from the PCTL semantics. Thus, the expressiveness of PCTL does not change when only permitting the upper bounds on the path probabilities ( \(<p\) and \(\leqslant p\) ) and one of the (qualitative) bounds \(=0\) or \(=1\). Note that, e.g.,
\[
\mathbb{P}_{] 0.3,0.7]}(\varphi) \equiv \neg \mathbb{P}_{\leqslant 0.3}(\varphi) \wedge \mathbb{P}_{\leqslant 0.7}(\varphi)
\]

Another example is the following duality law:
\[
\mathbb{P}_{>0}\left(\bigcirc \mathbb{P}_{>0}(\diamond \Phi)\right) \equiv \mathbb{P}_{>0}\left(\diamond \mathbb{P}_{>0}(\bigcirc \Phi)\right)
\]

This is proven as follows. First, consider \(\Rightarrow\). Let \(s\) be a state of Markov chain \(\mathcal{M}\) with \(s \models \mathbb{P}_{>0}\left(\bigcirc \mathbb{P}_{>0}(\diamond \Phi)\right)\). Then, for some direct successor \(t\) of \(s\), it holds that \(\left.t \models \mathbb{P}_{>0}(\diamond \Phi)\right)\). This implies that there exists a finite path fragment \(t_{0} t_{1} \ldots t_{n}\) which starts in \(t_{0}=t\) and ends in state \(t_{n} \in \operatorname{Sat}(\Phi)\). So, \(t_{n-1} \models \mathbb{P}_{>0}(\bigcirc \Phi)\). Since \(s t_{0} t_{1} \ldots t_{n-1}\) is a path fragment starting in \(s\), it follows that \(s \models \mathbb{P}_{>0}\left(\diamond \mathbb{P}_{>0}(\bigcirc \Phi)\right)\).

Consider \(\Leftarrow\). Let \(s \models \mathbb{P}_{>0}\left(\diamond \mathbb{P}_{>0}(\bigcirc \Phi)\right)\). From the PCTL semantics it follows that
\[
\operatorname{Pr}\left(s \models \diamond \operatorname{Sat}\left(\mathbb{P}_{>0}(\bigcirc \Phi)\right)\right)>0
\]

Thus, there exists a path fragment \(s_{0} s_{1} \ldots s_{n}\) starting in \(s_{0}=s\) and ending in a state \(s_{n} \in \operatorname{Sat}\left(\mathbb{P}_{>0}(\bigcirc \Phi)\right)\). Hence, \(s_{n}\) has a successor \(t\) with \(t \models \Phi\). But then \(s_{1} \ldots s_{n} t\) is a witness for \(s_{1} \models \mathbb{P}_{>0}(\diamond \Phi)\). Since \(s_{1} \in \operatorname{Post}(s)\), it follows that \(s \models \mathbb{P}_{>0}\left(\bigcirc \mathbb{P}_{>0}(\diamond \Phi)\right)\).

\subsection*{10.2.1 PCTL Model Checking}

The PCTL model-checking problem is the following decision problem. Given a finite Markov chain \(\mathcal{M}\), state \(s\) in \(\mathcal{M}\), and PCTL state formula \(\Phi\), determine whether \(s \models \Phi\). As for CTL model checking, the basic procedure is to compute the satisfaction set \(\operatorname{Sat}(\Phi)\). This is done recursively using a bottom-up traversal of the parse tree of \(\Phi\); see Algorithm 13 (page 342). The nodes of the parse tree represent the subformulae of \(\Phi\). For each node of the parse tree, i.e., for each subformula \(\Psi\) of \(\Phi\), the set \(\operatorname{Sat}(\Psi)\) is determined. For the propositional logic fragment of PCTL this is performed in exactly the same way as for CTL. The most interesting part is the treatment of subformulae of the form \(\Psi=\mathbb{P}_{J}(\varphi)\). In order to determine whether \(s \in \operatorname{Sat}(\Psi)\), the probability \(\operatorname{Pr}(s \models \varphi)\) for the event specified by \(\varphi\) needs to be established. Then
\[
\operatorname{Sat}\left(\mathbb{P}_{J}(\varphi)\right)=\{s \in S \mid \operatorname{Pr}(s \models \varphi) \in J\} .
\]

For several path formulae \(\varphi\), the computation of the probability \(\operatorname{Pr}(s \models \varphi)\) has, in fact, already been treated in previous sections.

Let us first consider the next-step operator. For \(\varphi=\bigcirc \Psi\), the following equality holds:
\[
\operatorname{Pr}(s \models \bigcirc \Psi)=\sum_{s^{\prime} \in \operatorname{Sat}(\Psi)} \mathbf{P}\left(s, s^{\prime}\right)
\]
where \(\mathbf{P}\) is the transition probability function of \(\mathcal{M}\). In matrix-vector notation we thus have that the vector \((\operatorname{Pr}(s \models \bigcirc \Psi))_{s \in S}\) can be computed by multiplying \(\mathbf{P}\) with the characteristic vector for \(\operatorname{Sat}(\Psi)\), i.e., bit vector \(\left(b_{s}\right)_{s \in S}\) where \(b_{s}=1\) if and only if \(s \in \operatorname{Sat}(\Psi)\). Checking the next-step operator thus reduces to a single matrix-vector multiplication.

The probability \(\operatorname{Pr}(s \models \varphi)\) for until formulae \(\varphi=\Phi U^{\leqslant n} \Psi\) or \(\varphi=\Phi U \Psi\) can be obtained using the techniques explained in Section 10.1.1 (see page 759ff). The events \(C \mathrm{U}^{\leqslant n} B\) or \(C \cup B\), respectively, should be taken with \(C=\operatorname{Sat}(\Phi)\) and \(B=\operatorname{Sat}(\Psi)\). For the bounded until operator \(\mathrm{U}{ }^{\leqslant n}\), the vector \((\operatorname{Pr}(s \models \varphi))_{s \in S}\) can be obtained by \(\mathcal{O}(n)\) vector-matrix multiplications. For the until operator, a linear equation system needs to be solved. In both cases, the dimension of the involved matrix is bounded by \(N \times N\) where \(N=|S|\) is the number of states in \(\mathcal{M}\). This yields

\section*{Theorem 10.40. Time Complexity of PCTL Model Checking for MCs}

For finite Markov chain \(\mathcal{M}\) and PCTL formula \(\Phi\), the PCTL model-checking problem \(\mathcal{M} \models \Phi\) can be solved in time
\[
\mathcal{O}\left(\operatorname{poly}(\operatorname{size}(\mathcal{M})) \cdot n_{\max } \cdot|\Phi|\right)
\]
where \(n_{\max }\) is the maximal step bound that appears in a subpath formula \(\Psi_{1} \mathrm{U}^{\leqslant n} \Psi_{2}\) of \(\Phi\) (and \(n_{\max }=1\) if \(\Phi\) does not contain a step-bounded until operator).

For efficiency reasons, in order to check the qualitative PCTL properties, such as \(\mathbb{P}_{=1}(a \cup b)\) or \(\mathbb{P}_{>0}(a \cup b)\), graph-based techniques are employed as described in the next subsection. This avoids solving a system of linear equations.

For CTL, counterexamples and witnesses can be provided as an additional diagnostic feedback when checking a transition system against a CTL formula \(\exists \varphi\) or \(\forall \varphi\). Alternatively, a CTL model checker may provide an output file containing the information on which states fulfill the subformulae. Recall that the satisfaction sets of all subformulae are computed by the CTL model checker anyway. The probabilistic analogue is to provide the information on the probabilities \(\operatorname{Pr}(s \models \varphi)\) for all subformulae \(\mathbb{P}_{J}(\varphi)\). Note that these values \(\operatorname{Pr}(s \models \varphi)\) have been computed by the PCTL model checker. Although this information can be very helpful, for large state spaces it might be hard to extract any relevant information. To provide more comprehensive diagnostic information, the PCTL model-checking procedure can be extended to return sets of finite path fragments that constitute a witness for the satisfaction or refutation of subformulae \(\mathbb{P}_{J}(\varphi)\). Let us explain this by means of (unconstrained) reachability properties. If \(s \not \vDash \mathbb{P}_{\leqslant p}(\diamond \Psi)\), then \(\operatorname{Pr}(s \models \diamond \Psi)>p\). A proof for the latter is given by a finite set \(\Pi\) of finite path fragments \(s_{0} s_{1} \ldots s_{n}\) such that \(s_{0}=s\), \(s_{i} \not \vDash \Psi\) for \(0 \leqslant i<n\) and \(s_{n} \neq \Psi\) and \(\sum_{\widehat{\pi} \in \Pi} \mathbf{P}(\widehat{\pi})>p\). A witness of \(s \not \vDash \mathbb{P}_{\leqslant p}(\diamond \Psi)\) thus is a finite set of paths, all satisfying \(\diamond \Psi\), such that their total probability mass exceeds \(p\).

\section*{Example 10.41. Counterexample}

Consider the (abstract) Markov chain depicted in Figure 10.7 and assume the PCTL formula of interest is \(\mathbb{P}_{\leqslant \frac{1}{2}}(\diamond b)\). The fact that \(s_{0} \not \vDash \mathbb{P}_{\leqslant \frac{1}{2}}(\diamond b)\) can be witnessed by, e.g., the set of finite paths:
\[
\{\underbrace{s_{0} s_{1} t_{1}}_{\text {probability }}, \underbrace{0.2}_{\text {probability }} \underbrace{s_{0} s_{1} s_{2} t_{1}}_{\text {probability }}, \underbrace{s_{0} s_{2} t_{1}}_{0.15}\}
\]
whose total probability exceeds the probability bound \(\frac{1}{2}\). Note that counterexamples are not unique; e.g., replacing the finite path \(s_{0} s_{2} t_{1}\) by \(s_{0} s_{1} s_{2} t_{1}\) in the above set of paths also yields a counterexample.

An evidence of \(s \not \vDash \mathbb{P}_{\geqslant p}(\diamond \Psi)\)-note that the only difference with the previous case is that


Figure 10.7: An example of a Markov chain.
\(p\) is an upper bound rather than a lower bound - is obtained by considering the set of finite paths that refute \(\diamond \Psi\) and showing that these paths occur with a probability that exceeds \(1-p\). More precisely, consider the finite set \(\Pi\) of finite paths \(s_{0} s_{1} \ldots s_{n}\) such that \(s_{0}=s\), \(s_{i} \not \vDash \Psi\) for \(0 \leqslant i<n\) and \(s_{n}\) belongs to some BSCC \(C\) of \(\mathcal{M}\) such that \(C \cap \operatorname{Sat}(\Psi)=\varnothing\). Moreover, it is required that \(\sum_{\widehat{\pi} \in \Pi} \mathbf{P}(\widehat{\pi})>1-p\). Note that all paths that start with a prefix \(s_{0} s_{1} \ldots s_{n} \in \Pi\) never visit a state for which \(\Psi\) holds. The cylinder sets \(C y l(\widehat{\pi})\) of all path fragments \(\widehat{\pi} \in \Pi\) are contained in \(\{\pi \in \operatorname{Paths}(s) \mid \pi \models \square \neg \Psi\}\). Hence:
\[
\operatorname{Pr}(s \models \diamond \Psi)=1-\operatorname{Pr}(s \models \square \neg \Psi) \leqslant 1-\sum_{\widehat{\pi} \in \Pi} \mathbf{P}(\widehat{\pi})<1-(1-p)=p .
\]

Vice versa, \(\operatorname{Pr}(s \models \square \neg \Psi)\) equals the probability of reaching a BSCC \(T\) via a path fragment through \(\neg \Psi\)-states such that \(T \cap S a t(\Psi)=\varnothing\). Hence, \(\operatorname{Pr}(s \mid \diamond \Psi)<p\) if and only if there exists such a finite set \(\Pi\). Such a set \(\Pi\) can be obtained by successively increasing \(n\) and collecting all finite path fragments \(s_{0} s_{1} \ldots s_{k}\) up to length \(k \leqslant n\) that fulfill the above conditions (i.e., \(s_{0}, \ldots, s_{k-1} \not \models \Psi\) and \(s_{k} \in T\) for some BSCC \(T\) with \(T \cap \operatorname{Sat}(\Psi)=\varnothing\) ), until the probabilities \(\mathbf{P}\left(s_{0} s_{1} \ldots s_{k}\right)\) sum up to a value \(>1-p\).

\subsection*{10.2.2 The Qualitative Fragment of PCTL}

The logic PCTL has been introduced as a variant of CTL where the path quantifiers \(\exists\) and \(\forall\) are replaced by the probabilistic operator \(\mathbb{P}_{J}\). It is the purpose of this section to compare the expressiveness of CTL and PCTL in more detail. As PCTL provides the possibility to specify lower (or upper) bounds in the likelihood that they differ from zero and 1 , as in \(\mathbb{P}_{\geqslant \frac{1}{2}}(\varphi)\), it is evident that there exist properties that can be defined in PCTL but not in CTL. It remains to investigate how the qualitative fragment of PCTL-only allowing bounds with \(p=0\) or \(p=1\)-relates to CTL. As we will see, these logics are not equally expressive, but are incomparable (though their common fragment is large).

\section*{Definition 10.42. Qualitative Fragment of PCTL}

State formulae in the qualitative fragment of PCTL (over AP) are formed according to the following grammar:
\[
\Phi::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\quad \neg \Phi| \mathbb{P}_{>0}(\varphi) \mid \mathbb{P}_{=1}(\varphi)
\]
where \(a \in A P, \varphi\) is a path formula formed according to the following grammar:
\[
\varphi::=\bigcirc \Phi \mid \Phi_{1} \cup \Phi_{2}
\]
where \(\Phi, \Phi_{1}\), and \(\Phi_{2}\) are state formulae.

The formulae of the qualitative fragment of PCTL are referred to as qualitative PCTL formulae. Although the only allowed probability bounds are \(>0\) and \(=1\), the bounds \(=0\) and \(<1\) can be derived, as
\[
\mathbb{P}_{=0}(\varphi) \equiv \neg \mathbb{P}_{>0}(\varphi) \quad \text { and } \quad \mathbb{P}_{<1}(\varphi) \equiv \neg \mathbb{P}_{=1}(\varphi)
\]

Thus, e.g., \(\mathbb{P}_{=1}\left(\diamond \mathbb{P}_{>0}(\bigcirc a)\right)\) and \(\mathbb{P}_{<1}\left(\mathbb{P}_{>0}(\diamond a) \cup b\right)\) are qualitative PCTL formulae, while \(\mathbb{P}_{<0.5}(\diamond a)\) and \(\mathbb{P}_{=1}\left(a U^{\leqslant 5} b\right)\) are not. Note that the bounded until operator is not part of the qualitative fragment of PCTL.

In the sequel, let \(\mathcal{M}\) be a Markov chain. PCTL formulae will be interpreted over \(\mathcal{M}\). CTL formulae are interpreted over the induced transition system of \(\mathcal{M}\), i.e., \(\operatorname{TS}(\mathcal{M})\) (see page 753). Recall that \(s \xrightarrow{\tau} s^{\prime}\) in this transition system if and only if \(\mathbf{P}\left(s, s^{\prime}\right)>0\); the exact transition probabilities are thus abstracted away in \(\operatorname{TS}(\mathcal{M})\).

\section*{Definition 10.43. Equivalence of PCTL and CTL Formulae}

The PCTL formula \(\Phi\) is equivalent to the CTL formula \(\Psi\), denoted \(\Phi \equiv \Psi\), if \(\operatorname{Sat}_{\mathcal{M}}(\Phi)=\) \(\operatorname{Sat}_{T S}(\mathcal{M})(\Psi)\) for each Markov chain \(\mathcal{M}\).

Consider first the PCTL formula \(\mathbb{P}_{=1}(\varphi)\) and the CTL formula \(\forall \varphi\). The former asserts \(\varphi\) to hold almost surely. This allows for certain exceptional executions where \(\varphi\) is violated. The formula \(\forall \varphi\), however, requires \(\varphi\) to hold for all paths. No exceptions are allowed. In fact, we have:
\[
s \models \forall \diamond a \quad \text { implies } \quad s \models \mathbb{P}_{=1}(\diamond a) .
\]

The reverse direction, however, does not hold. For example, consider the following Markov chain:


Formula \(\mathbb{P}_{=1}(\diamond a)\) holds in state \(s\) as the probability of visiting state \(s\) infinitely often is zero. On the other hand, the path \(s^{\omega}\) is possible and clearly violates \(\diamond a\). Thus:
\[
s \neq \mathbb{P}_{=1}(\diamond a) \quad \text { but } \quad s \not \models \forall \diamond a \text {. }
\]

However, for certain path formulae \(\varphi\), the probability bounds \(>0\) and \(=1\) correspond to existential and universal path quantification in CTL. The simplest such cases are path formulae involving the next-step operator:
\[
\begin{array}{llll}
s & =\mathbb{P}_{=1}(\bigcirc a) & \text { iff } & s \models \forall \bigcirc a \\
s \models \mathbb{P}_{>0}(\bigcirc a) & \text { iff } & s \models \exists \bigcirc a
\end{array}
\]
where \(s\) is a state in an arbitrary (possibly infinite) Markov chain. The same applies to reachability conditions \(\diamond a\) that hold with positive probability if and only if \(\forall a\) holds for some path, and to invariants \(\square a\) that hold on all paths if and only if they hold almost surely:
\[
\begin{array}{lll}
s \models \mathbb{P}_{>0}(\diamond a) & \text { iff } & s \models \exists \diamond a \\
s \models \mathbb{P}_{=1}(\square a) & \text { iff } & s \models \forall \square a .
\end{array}
\]

Let us illustrate how to conduct the formal proof of these statements. Consider the first statement. Assume \(s \models \mathbb{P}_{>0}(\diamond a)\). By the PCTL semantics, it follows that \(\operatorname{Pr}(s \models \diamond a)>0\). Thus, \(\{\pi \in \operatorname{Paths}(s) \mid \pi \models \diamond a\} \neq \varnothing\), and hence, \(s \models \exists \diamond a\). Vice versa, assume \(s \vDash \exists \diamond a\), i.e., there exists a finite path fragment \(s_{0} s_{1} \ldots s_{n}\) with \(s_{0}=s\) and \(s_{n}=a\). It follows that all paths in the cylinder set \(\operatorname{Cyl}\left(s_{0} s_{1} \ldots s_{n}\right)\) fulfill \(\diamond a\). Thus:
\[
\operatorname{Pr}(s \models \diamond a) \geqslant \operatorname{Pr}_{s}\left(\operatorname{Cyl}\left(s_{0} s_{1} \ldots s_{n}\right)\right)=\mathbf{P}\left(s_{0} s_{1} \ldots s_{n}\right)>0
\]

Therefore, \(s \models \mathbb{P}_{>0}(\diamond a)\).
The second statement follows by duality. First, observe that
\[
s \models \mathbb{P}_{=1}(\square a)=\mathbb{P}_{=0}(\diamond \neg a) \equiv \neg \mathbb{P}_{>0}(\diamond \neg a)
\]

We can now apply the equivalence of \(\mathbb{P}_{>0}(\diamond a)\) and \(\exists \diamond a\) as follows:
\[
s \models \neg \mathbb{P}_{>0}(\diamond \neg a) \text { iff } s \not \models \mathbb{P}_{>0}(\diamond \neg a) \text { iff } s \not \models \exists \diamond \neg a \text { iff } s \models \underbrace{\neg \exists \diamond \neg a}_{=\forall \square a} .
\]

Thus, the equivalence of \(\mathbb{P}_{>0}\left(\bigcirc \mathbb{P}_{>0}(\diamond \Phi)\right)\) and \(\mathbb{P}_{>0}\left(\diamond \mathbb{P}_{>0}(\bigcirc \Phi)\right)\) (stated on page 784) is the probabilistic analogue of the CTL duality law \(\exists \bigcirc \exists \diamond \Phi \equiv \exists \diamond \exists \bigcirc \Phi\).

The arguments for the equivalence of \(\mathbb{P}_{>0}(\diamond a)\) and \(\exists \diamond a\) can be generalized toward constrained reachability:
\[
s \models \mathbb{P}_{>0}(a \cup b) \quad \text { iff } \quad s \models \exists(a \cup b)
\]

A generalization that, however, does not work is swapping the probability bound \(>0\) and \(=1\) in the above two equivalences, as \(\mathbb{P}_{=1}(\diamond a) \not \equiv \forall \diamond a\) and \(\mathbb{P}_{>0}(\square a) \not \equiv \exists \square a\). In fact, the PCTL formula \(\mathbb{P}_{=1}(\diamond a)\) cannot be expressed in CTL.

\section*{Lemma 10.44.}
1. There is no CTL formula that is equivalent to \(\mathbb{P}_{=1}(\diamond a)\).
2. There is no CTL formula that is equivalent to \(\mathbb{P}_{>0}(\square a)\).

Proof: We provide the proof of the first statement; the second statement follows by duality, i.e., \(\mathbb{P}_{=1}(\diamond a)=\neg \mathbb{P}_{>0}(\square \neg a)\). The proof is by contraposition. Assume there exists a CTL formula \(\Phi\) such that \(\Phi \equiv \mathbb{P}_{=1}(\diamond a)\). Consider the infinite Markov chain \(\mathcal{M}_{p}\) modeling the random walk; see Figure 10.6 (page 780). The state space of \(\mathcal{M}_{p}\) is \(S=\mathbb{N}\) and there are transitions from state \(n\) to state \(n-1\) with probability \(1-p\) and to state \(n+1\) with probability \(p\) for \(n \geqslant 1\), while \(\mathbf{P}(0,0)=1-p\) and \(\mathbf{P}(0,1)=p\). Let the labeling function \(L\) be such that \(a\) only holds in state 0 . That is, \(0 \models a\) and \(n \not \vDash a\) for all \(n>0\). Recall that for \(p<\frac{1}{2}\), it holds that for any state \(n\) almost surely state 0 will be visited, while for \(p>\frac{1}{2}\), the Markov chain \(\mathcal{M}_{p}\) drifts to the right and the probability of reaching state 0 from any other state \(n>0\) is strictly smaller than 1 . Thus, e.g., in \(\mathcal{M}_{\frac{1}{4}}\) the formula \(\mathbb{P}_{=1}(\diamond a)\) holds for all states, while in \(\mathcal{M}_{\frac{3}{4}}\), the formula \(\mathbb{P}_{=1}(\diamond a)\) does not hold in, e.g., state \(n=1\). Hence:
\[
1 \in \operatorname{Sat}_{\mathcal{M}_{\frac{1}{4}}}\left(\mathbb{P}_{=1}(\diamond a)\right) \quad \text { but } \quad 1 \notin \operatorname{Sat}_{\mathcal{M}_{\frac{3}{4}}}\left(\mathbb{P}_{=1}(\diamond a)\right) .
\]

Since \(\operatorname{TS}\left(\mathcal{M}_{\frac{1}{4}}\right)=\operatorname{TS}\left(\mathcal{M}_{\frac{3}{4}}\right)\), it follows that
\[
\operatorname{Sat}_{\mathcal{M}_{\frac{1}{4}}}(\Phi)=\operatorname{Sat}_{\mathcal{M}_{\frac{3}{4}}}(\Phi)
\]

Hence, state 1 either fulfills the CTL formula \(\Phi\) in both structures or in none of them. This, however, contradicts the assumption that \(\mathbb{P}_{=1}(\diamond a)\) and \(\Phi\) are equivalent.

The proof relies on the fact that the satisfaction of \(\mathbb{P}_{=1}(\diamond a)\) for infinite Markov chains may depend on the precise value of the transition probabilities, while CTL just refers to the underlying graph of a Markov chain. In case one is restricted to finite Markov chains, the statement in Lemma 10.44 no longer holds. For each finite Markov chain \(\mathcal{M}\) and state \(s\) of \(\mathcal{M}\) it holds that
\[
s \models \mathbb{P}_{=1}(\diamond a) \quad \text { iff } \quad s \models \forall((\exists \diamond a) \mathrm{W} a)
\]
where W is the weak until operator (see Remark 6.9 on page 327 ) defined by \(\Phi \mathrm{W} \Psi=\) \(\Phi U \Psi \vee \square \Phi\). The proof of (a slight generalization of) this statement is the subject of Exercise 10.11 (page 903). For finite Markov chains, the qualitative fragment of PCTL can be embedded into CTL.

Whereas the previous lemma stated that some qualitative PCTL cannot be expressed in CTL, the following result states that for some CTL formulae no equivalent qualitative PCTL exists. As a result, CTL and the qualitative fragment of PCTL are incomparable.

\section*{Lemma 10.45.}
1. There is no qualitative PCTL formula that is equivalent to \(\forall \diamond\) a.
2. There is no qualitative PCTL formula that is equivalent to \(\exists \square a\).

Proof: We provide the proof of the first claim; the second claim follows by duality, since \(\forall \diamond a=\neg \exists \square \neg a\). For \(n \geqslant 1\), let the MCs \(\mathcal{M}_{n}\) and \(\mathcal{M}_{n}^{\prime}\) be defined as follows. The state spaces of \(\mathcal{M}_{n}\) and \(\mathcal{M}_{n}^{\prime}\) are \(S_{n}=\left\{t_{0}, t_{1}, \ldots, t_{n-1}\right\} \cup\left\{s_{n}\right\}\) and \(S_{n}^{\prime}=\left\{t_{0}, t_{1}, \ldots, t_{n}\right\}\), respectively. \(\mathcal{M}_{n}\) and \(\mathcal{M}_{n}^{\prime}\) agree on the fragment consisting of their common states, i.e., state \(t_{0}\) through \(t_{n-1}\). State \(s_{n}\) in \(\mathcal{M}_{n}\) has a self-loop and a transition to \(t_{n-1}\), both with probability \(\frac{1}{2}\). In contrast, state \(t_{n}\) in \(\mathcal{M}_{n}^{\prime}\) only has a transition \(t_{n} \rightarrow t_{n-1}\) with probability 1. Both Markov chains are depicted in Figure 10.8. The transition probabilities in \(\mathcal{M}_{n}^{\prime}\) are given by:
\[
\mathbf{P}_{n}^{\prime}\left(t_{i}, t_{i-1}\right)=1 \text { for } 1 \leqslant i \leqslant n \quad \text { and } \quad \mathbf{P}_{n}^{\prime}\left(t_{0}, t_{0}\right)=1
\]
and \(\mathbf{P}_{n}^{\prime}(\cdot)=0\) otherwise. The transition probabilities in \(\mathcal{M}_{n}\) are given by
\[
\begin{aligned}
& \mathbf{P}_{n}\left(s_{n}, t_{n-1}\right)=\mathbf{P}_{n}\left(s_{n}, s_{n}\right)=\frac{1}{2} \\
& \mathbf{P}_{n}\left(t_{i}, t_{i-1}\right)=1 \text { for } 1 \leqslant i<n \\
& \mathbf{P}_{n}\left(t_{0}, t_{0}\right)=1
\end{aligned}
\]
and \(\mathbf{P}_{n}(\cdot)=0\) in the remaining cases. Let \(A P=\{a\}\) and labeling functions \(L_{n}\) for \(\mathcal{M}_{n}\) and \(L_{n}^{\prime}\) for \(\mathcal{M}_{n}^{\prime}\) given by \(L_{n}\left(s_{n}\right)=L_{n}\left(t_{i}\right)=L_{n}^{\prime}\left(t_{j}\right)=\varnothing\) for \(1 \leqslant i<n\) and \(1 \leqslant j \leqslant n\) and \(L_{n}\left(t_{0}\right)=L_{n}^{\prime}\left(t_{0}\right)=\{a\}\).

It can now be shown that for any qualitative PCTL formula \(\Phi\) of nesting depth smaller than \(n\) :
\[
s_{n} \models \Phi \quad \text { iff } \quad t_{n} \models \Phi .
\]


Figure 10.8: The Markov chains \(\mathcal{M}_{n}\) and \(\mathcal{M}_{n}^{\prime}\).

This can be proven by induction on the nesting depth of \(\Phi\). (The proof is omitted here and left as an exercise.) The nesting depth of qualitative PCTL formulae is defined by
\[
\begin{array}{ll}
\operatorname{nd}(\text { true }) & =0 \\
\operatorname{nd}(a) & =0 \\
\operatorname{nd}(\Phi \wedge \Psi) & =\max \{\operatorname{nd}(\Phi), \operatorname{nd}(\Psi)\} \\
\operatorname{nd}(\neg \Phi) & =\operatorname{nd}(\Phi) \\
\operatorname{nd}\left(\mathbb{P}_{J}(\varphi)\right) & =\operatorname{nd}(\varphi)+1 \\
\operatorname{nd}(\bigcirc \Phi) & =\operatorname{nd}(\Phi) \\
\operatorname{nd}(\Phi \cup \Psi) & =\max \{\operatorname{nd}(\Phi), \operatorname{nd}(\Psi)\}
\end{array}
\]

Suppose now that there exists a qualitative PCTL formula \(\Phi\) such that \(\Phi \equiv \forall \vee\). Let \(n=n d(\Phi)+1\). Since \(s_{n} \models \Phi\) iff \(t_{n} \models \Phi\), states \(s_{n}\) and \(t_{n}\) either both fulfill \(\Phi\) or none of them. On the other hand, \(s_{n} \not \vDash \forall \diamond a\) (because \(s_{n} s_{n} s_{n} \ldots \not \vDash \diamond a\) ), while \(t_{n} \vDash \forall \diamond a\). Contradiction.

These results indicate that extending the syntax of PCTL by adding universal and existential path quantification increases its expressiveness. An essential aspect in the proof of Lemma 10.45 is the existence of a state \(s\), say, that has a self-loop and (at least) one outgoing transition to another state. (In the proof, the state of interest is \(s_{n}\).) As the probability of taking this loop infinitely often is zero, the validity of a qualitative PCTL formula in \(s\) in unaffected by the self-loop. However, the (zero probable) path \(s^{\omega}\) may be decisive for the validity of CTL formulae. The path \(s^{\omega}\), however, can be considered as unfair, as there are infinitely many opportunities to take the other outgoing transition. The existence of such unfair computations turns out to be vital for the validity of Lemma 10.45. In fact, under appropriate fairness constraints, the equivalence of \(\mathbb{P}_{=1}(\diamond a)\) and \(\forall \diamond a\) can be established. This can be seen as follows. Assume \(\mathcal{M}\) is a finite Markov chain and that any state \(s\) in \(\mathcal{M}\) is uniquely characterized by an atomic proposition, say \(s\). We
regard the strong fairness constraint sfair defined by
\[
\text { sfair }=\bigwedge_{s \in S} \bigwedge_{t \in \operatorname{Post}(s)}(\square \diamond s \rightarrow \square \diamond t) \text {. }
\]

It asserts that when a state \(s\) is visited infinitely often, then any of its direct successors is visited infinitely often too. Using Theorem 10.25 (page 772), we obtain (see Exercise 10.8 on page 902 )
\[
\begin{array}{lll}
s \models \mathbb{P}_{=1}(a \cup b) & \text { iff } & s \models_{\text {sfair }} \forall(a \cup b) \quad \text { and } \\
s \models \mathbb{P}_{>0}(\square a) & \text { iff } & s \models_{\text {sfair }} \exists \square a .
\end{array}
\]

As sfair is a realizable fairness constraint-from every reachable state at least one fair path starts-we obtain
\[
\begin{array}{lllll}
s \models_{\text {sfair }} \exists(a \cup b) & \text { iff } & s \models \exists(a \cup b) & \text { iff } & s \models \mathbb{P}_{>0}(a \cup b) \\
s \models_{\text {sfair }} \forall \bigcirc a & \text { iff } & s \models \forall \bigcirc a & \text { iff } & s \models \mathbb{P}_{=1}(\bigcirc a) \\
s \models_{\text {sfair }} \exists \bigcirc a & \text { iff } & s \models \exists \bigcirc a & \text { iff } & s \models \mathbb{P}_{>0}(\bigcirc a) .
\end{array}
\]

Thus, for finite Markov chains the qualitative fragment of PCTL can be viewed as a variant of CTL with some kind of strong fairness. A similar result holds for infinite Markov chains, but then the outermost conjunction in sfair has an infinite range. The same applies to the innermost conjunction in case infinite branching occurs.

Repeated Reachability and Persistence Properties. We now show that two properties that cannot be expressed in CTL (but which can be in CTL*) can be expressed in the qualitative fragment of PCTL.

For CTL, universal repeated reachability properties can be formalized by the combination of the modalities \(\forall \square\) and \(\forall \diamond\) :
\[
s \models \forall \square \forall \diamond a \quad \text { iff } \quad \pi \models \square \diamond a \text { for all } \pi \in \operatorname{Paths}(s) .
\]

More details can be found in Remark 6.8 on page 326. For finite Markov chains, a similar result holds for (the qualitative fragment of) PCTL.

\section*{Lemma 10.46. Almost Sure Repeated Reachability is PCTL-definable}

Let \(\mathcal{M}\) be a finite Markov chain and s a state of \(\mathcal{M}\). Then:
\[
s \models \mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right) \quad \text { iff } \quad \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \square \diamond a\}=1
\]

Proof: " \(\Longrightarrow "\) : Assume \(s \models \mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right)\). As \(\mathbb{P}_{=1}(\square a)\) is equivalent to \(\forall \square a\), we have \(s \vDash \forall \square \operatorname{Sat}\left(\mathbb{P}_{=1}(\diamond a)\right)\). Hence, all states \(t\) that are reachable from \(s\) satisfy \(\left.\mathbb{P}_{=1}(\diamond a)\right)\). Consider a BSCC \(T \subseteq S\) of \(\mathcal{M}\) such that \(\operatorname{Pr}(s \models \diamond T)>0\). Then, each state \(t\) in \(T\) is reachable from \(s\) and \(t \models \mathbb{P}_{=1}(\diamond a)\). As \(\operatorname{Post}^{*}(t)=T, T \cap \operatorname{Sat}(a) \neq \varnothing\). Hence, almost all paths \(\pi \in \operatorname{Paths}(s)\) that fulfill \(\diamond T\) satisfy \(\square \diamond a\). Theorem 10.27 (page 775) then yields the claim.
\(" \Longleftarrow ":\) Assume \(\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \square \diamond a\}=1\). Theorem 10.27 (page 775) yields
\[
\sum_{\substack{T \in B S C C(\mathcal{M}) \\ T \cap S a t(a) \neq \varnothing}} \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \Delta T\}=1
\]

On the other hand, \(\operatorname{Pr}(t \models \diamond a)=1\) for each state \(t\) of a BSCC \(T\) in \(\mathcal{M}\) with \(T \cap \operatorname{Sat}(a) \neq \varnothing\). Therefore, the union of all BSCCs \(T\) with \(T \cap \operatorname{Sat}(a) \neq \varnothing\) is a subset of \(\operatorname{Sat}\left(\mathbb{P}_{=1}(\diamond a)\right)\). This yields
\[
\operatorname{Pr}\left(s \models \diamond \operatorname{Sat}\left(\mathbb{P}_{=1}(\diamond a)\right)\right)=1
\]
and \(s \models \mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right)\).

According to this result, the PCTL formula \(\mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right)\) denotes that almost surely a state for which \(a\) holds is repeatedly reachable. In the sequel, let \(\mathbb{P}_{=1}(\square \diamond a)\) abbreviate this formula.

There is no CTL formula that is equivalent to the CTL* formula \(\exists \square \diamond a\). This is different for PCTL, as the requirement that the probability for the event \(\square \diamond a\) is positive can be described in PCTL. For finite Markov chains, even arbitrary rational lower or upper bounds for the likelihood of \(\square \diamond a\) are PCTL-definable. In fact, the argument given in the proof of Lemma 10.46 can be generalized for arbitrary probability bounds. The crucial point is that almost surely a BSCC \(T \in \mathcal{M}\) will be reached and each of its states will be visited infinitely often. Thus, the probabilities for \(\square \diamond a\) agree with the probability of reaching a BSCC \(T\) where \(a\) holds for some state in \(T\). Hence, for finite Markov chain \(\mathcal{M}\) and probability interval \(J\), the PCTL formula \(\mathbb{P}_{J}(\square \diamond a)\) defined by
\[
\mathbb{P}_{J}(\square \diamond a)=\mathbb{P}_{J}(\diamond \underbrace{\mathbb{P}_{=1}(\square \diamond a)}_{=\mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right)})
\]
characterizes all states \(s\) in \(\mathcal{M}\) for which \(\operatorname{Pr}(s \models \square \diamond a) \in J\). This is expressed by the following theorem:

Theorem 10.47. Repeated Reachability Probabilities Are PCTL-definable
Let \(\mathcal{M}\) be a finite Markov chain, s a state of \(\mathcal{M}\) and \(J \subseteq[0,1]\) an interval. Then:
\[
s \models \underbrace{\mathbb{P}_{J}\left(\diamond \mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right)\right.}_{=\mathbb{P}_{J}(\square \diamond a)} \quad \text { iff } \quad \operatorname{Pr}(s \models \square \diamond a) \in J .
\]

Proof: As \(\mathcal{M}\) is finite, Theorem 10.27 (page 775) permits computing the probability for visiting an \(a\)-state infinitely often, as follows:
\[
\begin{aligned}
& \operatorname{Pr}(s \models \square \diamond a) \\
= & \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \square \diamond a\} \\
= & \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \inf (\pi) \in B S C C(\mathcal{M}) \wedge \inf (\pi) \cap \operatorname{Sat}(a) \neq \varnothing\} \\
= & \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \diamond T \text { for some } T \in B S C C(\mathcal{M}) \text { with } T \cap \operatorname{Sat}(a) \neq \varnothing\} \\
= & \operatorname{Pr}\left(s \models \diamond \operatorname{Sat}\left(\mathbb{P}_{=1}(\square \diamond a)\right)\right) .
\end{aligned}
\]

Hence, \(\operatorname{Pr}(s \models \square \diamond a) \in J\) if and only if \(\operatorname{Pr}\left(s \models \diamond \operatorname{Sat}\left(\mathbb{P}_{=1}(\square \diamond a)\right)\right) \in J\) if and only if \(s \models \mathbb{P}_{J}\left(\diamond \mathbb{P}_{=1}(\square \diamond a)\right)=\mathbb{P}_{J}(\square \diamond a)\).

Recall that universal persistence properties cannot be expressed in CTL; see Lemma 6.19 (page 335). For finite Markov chains, PCTL allows specifying almost sure persistence and, moreover, persistence properties with arbitrary lower or upper bounds on the probability. This is stated by the following theorem.

\section*{Theorem 10.48. Persistence Probabilities are PCTL-Definable}

For finite Markov chain \(\mathcal{M}\), state s of \(\mathcal{M}\) and interval \(J \subseteq[0,1]\) :
\[
s \models \mathbb{P}_{J}\left(\diamond \mathbb{P}_{=1}(\square a)\right) \quad \text { iff } \quad \operatorname{Pr}(s \models \diamond \square a) \in J
\]

Proof: This result follows by Theorem 10.27 (page 775) and the observation that for each BSCC \(T\) of \(\mathcal{M}\) : (i) if \(T \subseteq \operatorname{Sat}(a)\), then \(\operatorname{Pr}(t \models \square a)=1\) for all states \(t \in T\), and (ii) if \(T \backslash \operatorname{Sat}(a) \neq \varnothing\), then \(\operatorname{Pr}(t \models \diamond \square a)=0\) for all states \(t \in T\).

In particular, the requirement that a persistence property holds almost surely for a certain state \(s\) (i.e., \(\operatorname{Pr}(s \models \diamond \square a)=1\) ) is given by the PCTL formula \(\mathbb{P}_{=1}\left(\diamond \mathbb{P}_{=1}(\square a)\right)\).

Let \(\mathbb{P}_{J}(\diamond \square \Phi)\) denote the PCTL formula to express a probability bound on persistence properties.

\subsection*{10.3 Linear-Time Properties}

The previous section has introduced the branching-time temporal logic PCTL, and has presented a model-checking algorithm for this logic for finite Markov chains. This section deals with linear-time properties; see Chapter 3. Recall that an LT property is a set of infinite traces. The quantitative model-checking problem that we are confronted with is: given a finite Markov chain \(\mathcal{M}\) and an \(\omega\)-regular property \(P\), compute the probability for the set of paths in \(\mathcal{M}\) for which \(P\) holds. Some special cases, like properties of the form \(C \cup B\) or \(\square \diamond B\), where \(B\) and \(C\) are sets of states in \(\mathcal{M}\), have been discussed before. The purpose of this section is to generalize these results toward arbitrary \(\omega\)-regular properties.

\section*{Definition 10.49. Probability of LT Properties}

Let \(\mathcal{M}\) be a Markov chain and \(P\) an \(\omega\)-regular property (both over \(A P\) ). The probability for \(\mathcal{M}\) to exhibit a trace in \(P\), denoted \(\operatorname{Pr}^{\mathcal{M}}(P)\), is defined by
\[
\operatorname{Pr}^{\mathcal{M}}(P)=\operatorname{Pr}^{\mathcal{M}}\{\pi \in \operatorname{Paths}(\mathcal{M}) \mid \operatorname{trace}(\pi) \in P\}
\]

This definition, of course, requires the measurability of the set of paths \(\pi\) with trace \((\pi) \in\) \(P\); see Remark 10.57 on page 804. For state \(s\) in \(\mathcal{M}\) let \(\operatorname{Pr}^{\mathcal{M}}(s \models P)\), or briefly, \(\operatorname{Pr}(s \models P)\), for \(\operatorname{Pr}^{\mathcal{M}_{s}}(P)\), i.e.,
\[
\operatorname{Pr}(s \models P)=\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{trace}(\pi) \in P\} .
\]

Similarly, for the LTL formula \(\varphi\) we write \(\operatorname{Pr}^{\mathcal{M}}(\varphi)\) for \(\operatorname{Pr}^{\mathcal{M}}(\operatorname{Words}(\varphi))=\operatorname{Pr}^{\mathcal{M}}\{\pi \in\) \(\operatorname{Paths}(\mathcal{M}) \mid \pi \models \varphi\}\), where \(\models\) is the standard LTL satisfaction relation, i.e., \(\pi \models \varphi\) iff \(\operatorname{trace}(\pi) \in \operatorname{Words}(\varphi)\). For state \(s\) of \(\mathcal{M}, \operatorname{Pr}^{\mathcal{M}}(s \models \varphi)\) denotes \(\operatorname{Pr}^{\mathcal{M}}(\varphi)\), i.e.,
\[
\operatorname{Pr}(s \models \varphi)=\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \varphi\} .
\]

Given an \(\omega\)-regular property \(P\) over \(A P\) and finite Markov chain \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\), the goal is to compute \(\operatorname{Pr}^{\mathcal{M}}(P)\). As for verifying \(\omega\)-regular properties, we adopt an automata-based approach. The main steps of this approach are as follows. The (complement of the) LT property \(P\) is represented by means of an automaton \(\mathcal{A}\), say. For
regular safety properties this is an automaton for the bad prefixes, whereas for \(\omega\)-regular properties this is a Büchi automaton for the complement of \(P\). It then suffices to check a reachability and persistence property, respectively, on the product \(\mathcal{M} \otimes \mathcal{A}\).

In order to guarantee that \(\mathcal{M} \otimes \mathcal{A}\) is a Markov chain, however, the automaton \(\mathcal{A}\) needs to be deterministic. This is a main difference with the traditional setting of transition systems where nondeterministic (finite-state or Büchi) automata do suffice. For regular safety properties we therefore assume \(\mathcal{A}\) to be a DFA for the bad prefixes. For the \(\omega\)-regular properties, \(\mathcal{A}\) is assumed to be a deterministic Rabin automaton (DRA, for short). DRAs are equally expressive as \(\omega\)-regular languages. (Recall that deterministic Büchi automata are strictly less expressive than NBAs and therefore do not suffice for our purpose).

Rather than checking a reachability or persistence property, determining \(\operatorname{Pr}^{\mathcal{M}}(P)\) is reduced to computing the probability of accepting runs in the product Markov chain \(\mathcal{M} \otimes \mathcal{A}\). This is first discussed in detail for regular safety properties, and subsequently for \(\omega\)-regular properties.

Regular Safety Properties We first consider regular safety properties. Recall that a safety property is regular whenever all bad prefixes constitute a regular language. Let \(\mathcal{A}=\left(Q, 2^{A P}, \delta, q_{0}, F\right)\) be a deterministic finite automaton (DFA) for the bad prefixes of a regular safety property \(P_{\text {safe }}\). That is,
\[
P_{\text {safe }}=\left\{A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega} \mid \forall n \geqslant 0 . A_{0} A_{1} \ldots A_{n} \notin \mathcal{L}(\mathcal{A})\right\}
\]

Without loss of generality, the transition function \(\delta\) is assumed to be total, i.e., \(\delta(q, A)\) is defined for each \(A \subseteq A P\) and each state \(q \in Q\). Furthermore, let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite Markov chain. Our interest is to compute the probability
\[
\operatorname{Pr}^{\mathcal{M}}\left(P_{\text {safe }}\right)=1-\sum_{s \in S} \iota_{\text {init }}(s) \cdot \operatorname{Pr}(s \models \mathcal{A})
\]
for \(\mathcal{M}\) to generate a trace in \(P_{\text {safe }}\), i.e., a trace such that none of its prefixes is accepted by \(\mathcal{A}\). The probability \(\operatorname{Pr}(s \models \mathcal{A})\) is given by
\[
\begin{aligned}
\operatorname{Pr}(s \models \mathcal{A}) & =\operatorname{Pr}_{s}^{\mathcal{M}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{pref}(\operatorname{trace}(\pi)) \cap \mathcal{L}(\mathcal{A}) \neq \varnothing\} \\
& =\operatorname{Pr}_{s}^{\mathcal{M}}\left\{\pi \in \operatorname{Paths}(s) \mid \operatorname{trace}(\pi) \notin P_{\text {safe }}\right\}
\end{aligned}
\]
where \(\operatorname{pref}\left(A_{0} A_{1} \ldots\right)\) denotes the set of all finite prefixes of the infinite word \(A_{0} A_{1} \ldots \in\) \(\left(2^{A P}\right)^{\omega}\). The value \(\operatorname{Pr}(s \models \mathcal{A})\) can be written as the (possibly infinite) sum:
\[
\operatorname{Pr}(s \models \mathcal{A})=\sum_{\widehat{\pi}} \mathbf{P}(\widehat{\pi})
\]
where \(\widehat{\pi}\) ranges over all finite path fragments \(s_{0} s_{1} \ldots s_{n}\) starting in \(s_{0}=s\) such that (1) \(\operatorname{trace}\left(s_{0} s_{1} \ldots s_{n}\right)=L\left(s_{0}\right) L\left(s_{1}\right) \ldots L\left(s_{n}\right) \in \mathcal{L}(\mathcal{A})\), and (2) the length \(n\) of \(\widehat{\pi}\) is minimal according to (1), i.e., trace \(\left(s_{0} s_{1} \ldots s_{i}\right) \notin \mathcal{L}(\mathcal{A})\) for all \(0 \leqslant i<n\). Condition (2) is equivalent to the requirement that \(\operatorname{trace}(\widehat{\pi})\) is a minimal bad prefix of \(P_{\text {safe }}\).

Computing the values \(\operatorname{Pr}(s \models \mathcal{A})\) by using these sums may be difficult. Instead, we adapt the techniques for checking regular safety properties of transition systems to the probabilistic case. This involves the product of \(\mathcal{M}\) and \(\mathcal{A}\) which is defined as follows.

\section*{Definition 10.50. Product Markov Chain}

Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov chain and \(\mathcal{A}=\left(Q, 2^{A P}, \delta, q_{0}, F\right)\) be a DFA. The product \(\mathcal{M} \otimes \mathcal{A}\) is the Markov chain:
\[
\mathcal{M} \otimes \mathcal{A}=\left(S \times Q, \mathbf{P}^{\prime}, \iota_{\text {init }}^{\prime},\{\text { accept }\}, L^{\prime}\right)
\]
where \(L^{\prime}(\langle s, q\rangle)=\{\) accept \(\}\) if \(q \in F\) and \(L^{\prime}(\langle s, q\rangle)=\varnothing\) otherwise, and
\[
\iota_{\text {init }}^{\prime}(\langle s, q\rangle)= \begin{cases}\iota_{\text {init }}(s) & \text { if } q=\delta\left(q_{0}, L(s)\right) \\ 0 & \text { otherwise }\end{cases}
\]

The transition probabilities in \(\mathcal{M} \otimes \mathcal{A}\) are given by
\[
\mathbf{P}^{\prime}\left(\langle s, q\rangle,\left\langle s^{\prime}, q^{\prime}\right\rangle\right)= \begin{cases}\mathbf{P}\left(s, s^{\prime}\right) & \text { if } q^{\prime}=\delta\left(q, L\left(s^{\prime}\right)\right) \\ 0 & \text { otherwise }\end{cases}
\]

Since \(\mathcal{A}\) is deterministic, \(\mathcal{M} \otimes \mathcal{A}\) can be viewed as the unfolding of \(\mathcal{M}\) where the automaton component \(q\) of the states \(\langle s, q\rangle\) in \(\mathcal{M} \otimes \mathcal{A}\) records the current state in \(\mathcal{A}\) for the path fragment taken so far. More precisely, for each (finite or infinite) path fragment \(\pi=s_{0} s_{1} s_{2} \ldots\) in \(\mathcal{M}\) there exists a unique run \(q_{0} q_{1} q_{2} \ldots\) in \(\mathcal{A}\) for \(\operatorname{trace}(\pi)=L\left(s_{0}\right) L\left(s_{1}\right) L\left(s_{2}\right) \ldots\) and
\[
\pi^{+}=\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle\left\langle s_{2}, q_{3}\right\rangle \ldots
\]
is a path fragment in \(\mathcal{M} \otimes \mathcal{A}\). Vice versa, every path fragment in \(\mathcal{M} \otimes \mathcal{A}\) which starts in state \(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle\) arises from the combination of a path fragment in \(\mathcal{M}\) and a corresponding run in \(\mathcal{A}\). Note that the DFA \(\mathcal{A}\) does not affect the probabilities. That is, for each measurable set \(\Pi\) of paths in \(\mathcal{M}\) and state \(s\),
\[
\operatorname{Pr}_{s}^{\mathcal{M}}(\Pi)=\operatorname{Pr}_{\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle}^{\mathcal{M} \otimes \mathcal{A}} \underbrace{\left\{\pi^{+} \mid \pi \in \Pi\right\}}_{\Pi^{+}}
\]
where the superscripts \(\mathcal{M}\) and \(\mathcal{M} \otimes \mathcal{A}\) are used to indicate the underlying Markov chain. In particular, if \(\Pi\) is the set of paths that start in \(s\) and refute \(P_{\text {safe }}\), i.e.,
\[
\Pi=\left\{\pi \in \operatorname{Paths}^{\mathcal{M}}(s) \mid \operatorname{pref}(\operatorname{trace}(\pi)) \cap \mathcal{L}(\mathcal{A}) \neq \varnothing\right\}
\]
the set \(\Pi^{+}\)is the set of paths in \(\mathcal{M} \otimes \mathcal{A}\) that start in \(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle\) and eventually reach an accept state of \(\mathcal{A}\) :
\[
\Pi^{+}=\left\{\pi^{+} \in \operatorname{Paths}^{\mathcal{M} \otimes \mathcal{A}}\left(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle\right) \mid \pi^{+} \models \diamond \text { accept }\right\}
\]

Recall that the atomic proposition accept characterizes the set of states \(\langle s, q\rangle\) where \(q\) is an accept state of \(\mathcal{A}\). Thus, the paths satisfying \(\rangle\) accept agree with the event \(\diamond B\) where \(B=S \times F\). This shows that \(\operatorname{Pr}^{\mathcal{M}}\left(s \models P_{\text {safe }}\right)\) can be derived from the probability for the event \(\diamond\) accept in \(\mathcal{M} \otimes \mathcal{A}\). This is formally stated in the following theorem.

\section*{Theorem 10.51. Quantitative Analysis for Safety Properties}

Let \(P_{\text {safe }}\) be a regular safety property, \(\mathcal{A}\) a DFA for the set of bad prefixes of \(P_{\text {safe }}, \mathcal{M}\) a Markov chain, and s a state in \(\mathcal{M}\). Then:
\[
\begin{aligned}
\operatorname{Pr}^{\mathcal{M}}\left(s \models P_{\text {safe }}\right) & =\operatorname{Pr}^{\mathcal{M} \otimes \mathcal{A}}\left(\left\langle s, q_{s}\right\rangle \not \models \diamond \text { accept }\right) \\
& =1-\operatorname{Pr}^{\mathcal{M} \otimes \mathcal{A}}\left(\left\langle s, q_{s}\right\rangle \models \diamond \text { accept }\right)
\end{aligned}
\]
where \(q_{s}=\delta\left(q_{0}, L(s)\right)\).

Thus, the computation of the probability for a regular safety property in a Markov chain is reducible to computing the reachability probability in a product Markov chain. For finite Markov chains, the latter problem can be solved by means of the techniques discussed in Section 10.1.1 (see page 759ff).

For the special case of qualitative regular safety properties, i.e., whether \(P_{\text {safe }}\) almost surely holds in (finite) \(\mathcal{M}\), a graph analysis suffices. By means of a DFS- or BFS-based search algorithm we can check whether a state \(\langle s, q\rangle\) with \(q \in F\) is reachable in \(\mathcal{M} \otimes \mathcal{A}\). That is, the CTL formula \(\exists \diamond\) accept is checked. For the dual qualitative constraint, i.e., whether \(P_{\text {safe }}\) holds for \(\mathcal{M}\) with probability zero, the graph-based techniques suggested in Corollary 10.29 (page 775) can be exploited to check whether \(\diamond\) accept holds almost surely in \(\mathcal{M} \otimes \mathcal{A}\).
\(\omega\)-Regular Properties Let us now consider the wider class of LT properties, i.e., \(\omega\) regular properties. Recall that \(P\) is \(\omega\)-regular whenever \(P\) defines an \(\omega\)-regular language.
Let \(P\) be an \(\omega\)-regular property. In case the (complement of) \(P\) can be described by a deterministic Büchi automaton \(\mathcal{A}\), say, the technique for regular safety properties can
be roughly adopted. Consider the product Markov chain \(\mathcal{M} \otimes \mathcal{A}\) (see Definition 10.50 on page 798). It can now be shown, using similar arguments as for regular safety properties, that the probability of the event \(\square \diamond\) accept in the product MC \(\mathcal{M} \otimes \mathcal{A}\) coincides with the probability of refuting \(P\) by \(\mathcal{M}\), i.e.,
\[
\operatorname{Pr}^{\mathcal{M}}(s \models \mathcal{A})=\operatorname{Pr}_{s}^{\mathcal{M}}\left\{\pi \in \operatorname{Paths}(s) \mid \operatorname{trace}(\pi) \in \mathcal{L}_{\omega}(\mathcal{A})\right\}
\]

The probability of \(\square \diamond\) accept can be obtained in polynomial time in the following way. First, determine the BSCCs of \(\mathcal{M} \otimes \mathcal{A}\) (by a standard graph analysis). For each BSCC \(B\) that contains a state \(\langle s, q\rangle\) with \(q \in F\), determine the probability of eventually reaching \(B\). This goes as indicated in Corollary 10.34 on page 779. Thus:

\section*{Theorem 10.52. Quantitative Analysis for DBA-Definable Properties}

Let \(\mathcal{A}\) be a \(D B A\) and \(\mathcal{M}\) a Markov chain. Then, for all states \(s\) in \(\mathcal{M}\) :
\[
\operatorname{Pr}^{\mathcal{M}}(s \models \mathcal{A})=\operatorname{Pr}^{\mathcal{M} \otimes \mathcal{A}}\left(\left\langle s, q_{s}\right\rangle \models \square \diamond \text { accept }\right)
\]
where \(q_{s}=\delta\left(q_{0}, L(s)\right)\).

Proof: The argument is similar as for regular safety properties. The connection between a path \(\pi\) in \(\mathcal{M}\) and the corresponding path \(\pi^{+}\)in \(\mathcal{M} \otimes \mathcal{A}\) (which arises by augmenting the states \(s_{i}\) in \(\pi=s_{0} s_{1} s_{2} \ldots\) with the automaton states of the unique run for \(\operatorname{trace}(\pi)\) in \(\left.\mathcal{A}\right)\) is as follows:
\[
\operatorname{trace}(\pi) \in \mathcal{L}_{\omega}(\mathcal{A}) \quad \text { iff } \quad \pi^{+} \models \square \diamond \text { accept }
\]

Since \(\mathcal{A}\) is deterministic, \(\mathcal{A}\) does not affect the probabilities in \(\mathcal{M} \otimes \mathcal{A}\). Thus, the probability for path \(\pi\) in \(\mathcal{M}\) with \(\operatorname{trace}(\pi) \in \mathcal{L}_{\omega}(\mathcal{A})\) agrees with the probability of generating a path \(\pi^{+}\)in \(\mathcal{M} \otimes \mathcal{A}\) which arises by the lifting of path \(\pi\) in \(\mathcal{M}\) where \(\operatorname{trace}(\pi) \in \mathcal{L}_{\omega}(\mathcal{A})\). The latter agrees with the probability for the paths \(\pi^{+}\)in \(\mathcal{M} \otimes \mathcal{A}\) with \(\pi^{+} \models \square \diamond\) accept.

Since DBAs do not have the full power of \(\omega\)-regular languages (see Section 4.3.3 on page 188 ff ), this approach is not capable of handling arbitrary \(\omega\)-regular properties. To overcome this deficiency, Büchi automata will be replaced by an alternative automaton model for which their deterministic counterparts are as expressive as \(\omega\)-regular languages. Such automata have the same components as NBAs (finite set of states, and so on) except for the acceptance condition. We consider Rabin automata. The acceptance condition of a Rabin automaton is given by a set of pairs of states:
\[
\left\{\left(L_{i}, K_{i}\right) \mid 0<i \leqslant k\right\} \quad \text { with } \quad L_{i}, K_{i} \subseteq Q
\]

A run of a Rabin automaton is accepting if for some pair \(\left(L_{i}, K_{i}\right)\) the states in \(L_{i}\) are visited finitely often and the states in \(K_{i}\) infinitely often. That is, an accepted run should
satisfy the following LTL formula:
\[
\bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)
\]

A deterministic Rabin automaton is deterministic in the usual sense: it has a single initial state, and for each state and each input symbol, there is at most one successor state.

\section*{Definition 10.53. Deterministic Rabin Automaton (DRA)}

A deterministic Rabin automaton (DRA) is a tuple \(\mathcal{A}=\left(Q, \Sigma, \delta, q_{0}, A c c\right)\) where \(Q\) is a finite set of states, \(\Sigma\) an alphabet, \(\delta: Q \times \Sigma \rightarrow Q\) the transition function, \(q_{0} \in Q\) the starting state, and
\[
A c c \subseteq 2^{Q} \times 2^{Q}
\]

A run for \(\sigma=A_{0} A_{1} A_{2} \ldots \in \Sigma^{\omega}\) denotes an infinite sequence \(q_{0} q_{1} q_{2} \ldots\) of states in \(\mathcal{A}\) such that \(q_{i} \xrightarrow{A_{i}} q_{i+1}\) for \(i \geqslant 0\). The run \(q_{0} q_{1} q_{2} \ldots\) is accepting if there exists a pair \((L, K) \in A c c\) such that
\[
\left(\exists n \geqslant 0 . \forall m \geqslant n . q_{m} \notin L\right) \wedge\left(\exists n \geqslant 0 . q_{n} \in K\right) .
\]

The accepted language of \(\mathcal{A}\) is
\[
\mathcal{L}_{\omega}(\mathcal{A})=\left\{\sigma \in \Sigma^{\omega} \mid \text { the run for } \sigma \text { in } \mathcal{A} \text { is accepting }\right\} .
\]

Any DBA can be considered as an DRA in the following way. Assume a DBA is given with accept set \(F\), i.e., an accepting run should visit some state in \(F\) infinitely often. The DRA with the same states and transitions and with the singleton acceptance condition Acc \(=\{(\varnothing, F)\}\) is evidently equivalent to this DBA. DRAs are thus at least as expressive as DBAs.

Example 10.54. \(\quad D R A\) for \(\square \diamond a\)
Consider the DBA for the LTL formula \(\square \diamond a\); see Figure 10.9. The alphabet of this DBA is \(2^{\{a\}}=\{\{a\}, \varnothing\}\). Each accepting run has to visit \(q_{1}\) infinitely often. The DRA with Acc \(=\left\{\left(\varnothing,\left\{q_{1}\right\}\right)\right\}\) is equivalent to this DBA: a run is accepting if and only if state \(q_{1}\) is visited infinitely often. As the first component of the accept condition is \(\varnothing\), the requirement that the first set is only visited finitely many times is vacuously true.

Recall that some \(\omega\)-regular properties cannot be expressed by a DBA. This applies, e.g., to persistence properties such as the LTL formula \(\diamond \square a\). DRAs are more expressive than


Figure 10.9: A DBA for \(\square \diamond a\).


Figure 10.10: A deterministic Rabin automaton for \(\diamond \square a\).

DBAs. For instance, consider the DRA in Figure 10.10 with the acceptance condition Acc \(=\left\{\left(\left\{q_{0}\right\},\left\{q_{1}\right\}\right)\right\}\). The accepted language of this DRA is the set of infinite words whose runs end with a suffix that never visits \(q_{0}\), and, thus stays forever in state \(q_{1}\). These are exactly the words \(A_{0} A_{1} \ldots\) in \(\operatorname{Words}(\diamond \square a)\). As \(\diamond \square a\) cannot be described by a DBA, the class of languages accepted by a DRA is strictly larger than the class of languages that are recognizable by a DBA. In fact, the following fundamental result asserts that DRAs are as expressive as \(\omega\)-regular properties.

\section*{Theorem 10.55. DRAs and \(\omega\)-Regular Languages}

The class of languages accepted by DRAs agrees with the class of \(\omega\)-regular languages.

Proof: The proof of this result is outside the scope of this monograph. The interested reader is referred to [174] for more details.

Recall that NBAs are also as expressive as \(\omega\)-regular properties, cf. Theorem 4.32, and thus DRAs and NBAs are equally expressive. In fact, there exists an algorithm that takes as input an NBA \(\mathcal{A}\) and generates an equivalent DRA of size \(2^{\mathcal{O}(n \log n)}\) where \(n\) is the size of \(\mathcal{A}\).

Let us now return to our original problem: the verification of quantitative \(\omega\)-regular properties over Markov chains. Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite Markov chain and let \(\mathcal{A}=\left(Q, 2^{A P}, \delta, q_{0}, A c c\right)\) be a DRA which accepts the complement of the \(\omega\)-regular LT property \(P\). The probability of \(\mathcal{M}\) to generate traces in \(\mathcal{L}_{\omega}(\mathcal{A})\) can be computed-as for regular safety properties - on the basis of a product construction. The product of \(\mathcal{M}\) and \(\mathcal{A}\) is defined as the product of a Markov chain and a DFA; see Definition 10.50 (page 798).

If the acceptance condition of \(\mathcal{A}\) is
\[
A c c=\left\{\left(L_{1}, K_{1}\right), \ldots,\left(L_{k}, K_{k}\right)\right\}
\]
then the sets \(L_{i}, K_{i}\) serve as atomic propositions in \(\mathcal{M} \otimes \mathcal{A}\). The labeling function is the obvious one, i.e., if \(H \in\left\{L_{1}, \ldots, L_{k}, K_{1}, \ldots, K_{k}\right\}\), then \(H\) holds in state \(\langle s, q\rangle\) of \(\mathcal{M} \otimes \mathcal{A}\) iff \(q \in H\).

According to the following result, the computation of probabilities for satisfying \(\omega\)-regular properties boils down to computing the reachability probabilities for certain bottom strongly connected components (BSCCs) in \(\mathcal{M} \otimes \mathcal{A}\). These BSCCs are called accepting. A BSCC \(T\) in \(\mathcal{M} \otimes \mathcal{A}\) is accepting if it fulfills the acceptance condition Acc. More precisely, \(T\) is accepting if and only if there exists some index \(i \in\{1, \ldots, k\}\) such that
\[
T \cap\left(S \times L_{i}\right)=\varnothing \quad \text { and } \quad T \cap\left(S \times K_{i}\right) \neq \varnothing
\]

Stated in words, there is no state \(\langle s, q\rangle \in T\) such that \(q \in L_{i}\) and for some state \(\left\langle t, q^{\prime}\right\rangle \in T\) it holds that \(q \in K_{i}\). Thus, once such an accepting BSCC \(T\) is reached in \(\mathcal{M} \otimes \mathcal{A}\), the acceptance criterion for the DRA \(\mathcal{A}\) is fulfilled almost surely.

\section*{Theorem 10.56. DRA-Based Analysis of Markov Chains}

Let \(\mathcal{M}\) be a finite Markov chain, s a state in \(\mathcal{M}, \mathcal{A}\) a \(D R A\), and let \(U\) be the union of all accepting BSCCs in \(\mathcal{M} \otimes \mathcal{A}\). Then:
\[
\operatorname{Pr}^{\mathcal{M}}(s \mid \mathcal{A})=\operatorname{Pr}^{\mathcal{M} \otimes \mathcal{A}}\left(\left\langle s, q_{s}\right\rangle \models \diamond U\right)
\]
where \(q_{s}=\delta\left(q_{0}, L(s)\right)\).

Proof: Let \(\Pi\) be the set of paths \(\pi \in \operatorname{Paths}(s)\) in the Markov chain \(\mathcal{M}\) such that trace \((\pi) \in\) \(\mathcal{L}_{\omega}(\mathcal{A})\), and \(\Pi^{+}\)be the set of paths in \(\mathcal{M} \otimes \mathcal{A}\) that are obtained by augmenting the paths \(\pi \in \Pi\) with their corresponding runs in \(\mathcal{A}\). Note that the automaton \(\mathcal{A}\) is deterministic. Hence, for any path \(\pi\) in \(\mathcal{M}\) there is a unique run in \(\mathcal{A}\) for \(\operatorname{trace}(\pi)\). As the transition probabilities in \(\mathcal{M} \otimes \mathcal{A}\) are not affected by the DRA \(\mathcal{A}\), it follows that
\[
\operatorname{Pr}_{s}^{\mathcal{M}}(\Pi)=\operatorname{Pr}_{\left\langle s, q_{s}\right\rangle}^{\mathcal{M} \otimes \mathcal{A}}\left(\Pi^{+}\right)
\]
with \(q_{s}=\delta\left(q_{0}, L(s)\right)\).

Since the traces of the paths \(\pi \in \Pi\) belong to \(\mathcal{L}_{\omega}(\mathcal{A})\), their runs are accepting, i.e., for some acceptance pair ( \(L_{i}, K_{i}\) ) of \(\mathcal{A}, K_{i}\) is visited infinitely often, and from some moment on, \(L_{i}\) is not visited anymore. Vice versa, \(\operatorname{trace}(\pi) \in L_{\omega}(\mathcal{A})\) for any path \(\pi\) in \(\mathcal{M}\) where
the extended path \(\pi^{+}\)in \(\mathcal{M} \otimes \mathcal{A}\) fulfills \(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\) for some acceptance pair ( \(L_{i}, K_{i}\) ) of \(\mathcal{A}\). Hence:
\[
\operatorname{Pr}^{\mathcal{M}}(s \models \mathcal{A})=\operatorname{Pr}^{\mathcal{M} \otimes \mathcal{A}}\left\{\left\langle s, q_{s}\right\rangle \models \bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)\right\}
\]

Whether a run \(\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots\) satisfies the acceptance condition of a DRA only depends on the states that are visited infinitely often. By Theorem 10.27 (page 775), such states almost surely form a BSCC in \(\mathcal{M} \otimes \mathcal{A}\). Hence, the probability for \(\bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)\) in \(\mathcal{M} \otimes \mathcal{A}\) coincides with the probability of reaching an accepting BSCC in \(\mathcal{M} \otimes \mathcal{A}\).

This theorem yields that the probability of \(\mathcal{M}\) to generate a trace in \(\mathcal{L}_{\omega}(\mathcal{A})\) is given by
\[
\operatorname{Pr}^{\mathcal{M}}(\mathcal{A})=\sum_{s \in S} \iota_{\text {init }}(s) \cdot \operatorname{Pr}^{\mathcal{M} \otimes \mathcal{A}}\left(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle \models \diamond U\right)
\]

This result suggests determining the BSCCs in the product Markov chain \(\mathcal{M} \otimes \mathcal{A}\) to check which BSCC is accepting (i.e., determine \(U\) ), and then to compute for each of the accepting BSCCs the reachability probability. The first stage of this algorithm can be performed by a standard graph analysis. To check whether a BSCC is accepting amounts to checking all pairs \(\left(L_{i}, K_{i}\right) \in\) Acc. Finally, reachability probabilities can be determined by solving a set of linear equations, as indicated earlier in this chapter (see Section 10.1.1). The size of this linear equation system is linear in the size of the Markov chain \(\mathcal{M}\) and the DRA \(\mathcal{A}\). The overall time complexity of this procedure is
\[
\mathcal{O}(\operatorname{poly}(\operatorname{size}(\mathcal{M}), \operatorname{size}(\mathcal{A})))
\]

To check whether almost all traces of \(\mathcal{M}\) are accepted by the DRA \(\mathcal{A}\), it suffices to check whether all BSCCs of \(\mathcal{M} \otimes \mathcal{A}\) that are reachable from some initial state \(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle\) in \(\mathcal{M} \otimes \mathcal{A}\) for which \(\iota_{\text {init }}(s)>0\) are accepting.

\section*{Remark 10.5\%. Measurability of \(\omega\)-Regular Properties}

So far, we have implicitly assumed the measurability of \(\omega\)-regular properties. The fact that \(\omega\)-regular properties are indeed measurable follows by the following argument.

Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov chain and \(P\) an \(\omega\)-regular property which is represented by the DRA \(\mathcal{A}=\left(Q, 2^{A P}, \delta, q_{0}, A c c\right)\) with \(A c c=\left\{\left(L_{1}, K_{1}\right), \ldots,\left(L_{k}, K_{k}\right)\right\}\). We need to prove that:
\[
\Pi=\{\pi \in \operatorname{Paths}(\mathcal{M}) \mid \operatorname{trace}(\pi) \in \underbrace{\mathcal{L}_{\omega}(\mathcal{A})}_{=P}\}
\]
is measurable. Each path \(\pi=s_{0} s_{1} \ldots \in \Pi\) is lifted to a path \(\pi^{+}\)in the product Markov chain \(\mathcal{M} \otimes \mathcal{A}\) such that \(\pi^{+}=\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots\). The path \(\pi^{+}\)results from \(\pi\) by "adding" the states in \(\mathcal{A}\) of the (unique) run \(q_{0} q_{1} q_{2} \ldots\) in \(\mathcal{A}\) for trace \((\pi)\). Then:
\[
\pi \in \Pi \text {, i.e., } \operatorname{trace}(\pi) \in \mathcal{L}_{\omega}(\mathcal{A})
\]
iff the (unique) run \(q_{0} q_{1} q_{2} \ldots\) for \(\operatorname{trace}(\pi)\) in \(\mathcal{A}\) is accepting
iff the path \(\pi^{+}=\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots\) in \(\mathcal{M} \otimes \mathcal{A}\) satisfies
\[
\underset{1 \leqslant i \leqslant k}{\bigvee}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)
\]

Let \(\varphi_{i}=\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\), for \(0<i \leqslant k\), and \(\Pi_{i}\) the set of all paths \(\pi\) in \(\mathcal{M}\) such that \(\pi^{+} \models \varphi_{i}\). Clearly, \(\Pi=\Pi_{1} \cup \ldots \cup \Pi_{k}\). To show the measurability of \(\Pi\) it thus suffices to prove the measurability of \(\Pi_{i}\). Let us consider \(\Pi_{i}\) in somewhat more detail. It follows that \(\Pi_{i}=\Pi_{i}^{\diamond \square} \cap \Pi_{i}^{\square \diamond}\) where \(\Pi_{i}^{\diamond \square}\) is the set of paths \(\pi\) in \(\mathcal{M}\) such that \(\pi^{+} \models \diamond \square \neg L_{i}\), and \(\Pi_{i}^{\square \diamond}\) is the set of paths \(\pi\) in \(\mathcal{M}\) such that \(\pi^{+} \models \square \diamond K_{i}\). It remains to show that \(\Pi_{i}^{\diamond \square}\) and \(\Pi_{i}^{\square \diamond}\) are measurable. The set \(\Pi_{i}^{\diamond \square}\) can be written as
\[
\bigcup_{n \geqslant 0} \bigcap_{m \geqslant n} \bigcup_{s_{0} \ldots s_{n} \ldots s_{m}} C y l\left(s_{0} \ldots s_{n} \ldots s_{m}\right)
\]
where \(s_{0} \ldots s_{n} \ldots s_{m}\) ranges over all finite path fragments in \(\mathcal{M}\) such that \(q_{j} \notin L_{i}\) for \(n<\) \(j \leqslant m+1\) for the induced run \(q_{0} q_{1} \ldots q_{n+1} \ldots q_{m+1}\) in DRA \(\mathcal{A}\). Thus, \(\Pi_{i}^{\diamond \square}\) is measurable.

Similarly, the set \(\Pi_{i}^{\square \diamond}\) can be written as
\[
\bigcap_{n \geqslant 0} \bigcup_{m \geqslant n} \bigcup_{s_{0} \ldots s_{n} \ldots s_{m}} C y l\left(s_{0} \ldots s_{n} \ldots s_{m}\right)
\]
where \(s_{0} \ldots s_{n} \ldots s_{m}\) ranges over all finite path fragments in \(\mathcal{M}\) such that \(q_{m+1} \in K_{i}\) for the induced run \(q_{0} q_{1} \ldots q_{n+1} \ldots q_{m+1}\) in \(\mathcal{A}\). Hence, both sets \(\Pi_{i}^{\diamond \square}\) and \(\Pi_{i}^{\square \diamond}\) arise through countable unions and intersections of cylinder sets and are therefore measurable.

Although the approach with DRA is conceptually simple, it has the drawback of a double exponential blowup when starting with an LTL formula for the LT property. We mention without proof that there exist LTL formulae \(\varphi_{n}\) of size \(\mathcal{O}(\operatorname{poly}(n))\) for which the smallest DRA representation for \(\varphi_{n}\) has \(2^{2^{n}}\) states. Using alternative techniques, the double exponential blowup can be reduced to a single exponential blowup. The details of these advanced techniques fall outside the scope of this monograph. Algorithms with a better asymptotic worst-case complexity cannot be expected, as the qualitative model-checking problem for finite Markov chains "given a finite Markov chain \(\mathcal{M}\) and an LTL formula \(\varphi\), does \(\operatorname{Pr}(\mathcal{M} \models \varphi)=1\) hold?" is PSPACE-complete. This result is due to Vardi [407] and stated here without proof.

\section*{Theorem 10.58.}

The qualitative model-checking problem for finite Markov chains is PSPACE-complete.

\section*{10.4 \(\mathrm{PCTL}^{*}\) and Probabilistic Bisimulation}

This section introduces probabilistic bisimulation for Markov chains, and shows that this notion of bisimulation coincides with PCTL equivalence. That is to say, PCTL equivalence serves as a logical characterization of probabilistic bisimulation. Vice versa, probabilistic bisimulation serves as an operational characterization of PCTL equivalence. It is shown that the same applies to PCTL*, a logic that results from the state formulae in PCTL and allowing LTL formulae as path formulae. These results may thus be considered as the quantitative analogue to the result that bisimulation on transition systems coincides with CTL and CTL* equivalence; see Theorem 7.20 (page 469).

\subsection*{10.4.1 PCTL*}

The logic PCTL* extends PCTL by dropping the requirement that any temporal operator must be proceeded by a state formula. In addition, it allows for boolean combinations of path formulae. Thus, the logic PCTL* permits Boolean combinations of formulae of the form \(\mathbb{P}_{J}(\varphi)\) where the interval \(J\) specifies a probability bound, whereas \(\varphi\) is an LTL formula whose substate formulae are PCTL* state formulae. The logic PCTL* is strictly more expressive than LTL (with probability bounds) and PCTL.

\section*{Definition 10.59. Syntax of PCTL*}

PCTL* state formulae over the set AP of atomic propositions are formed according to the following grammar:
\[
\Phi::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\quad \neg \Phi| \mathbb{P}_{J}(\varphi)
\]
where \(a \in A P, \varphi\) is a path formula and \(J \subseteq[0,1]\) is an interval with rational bounds. PCTL* path formulae are formed according to the following grammar:
\[
\varphi::=\Phi\left|\varphi_{1} \wedge \varphi_{2}\right| \neg \varphi|\bigcirc \varphi| \varphi_{1} \cup \varphi_{2}
\]
where \(\Phi\) is a PCTL* state formula.

Other Boolean operators and the temporal modalities \(\diamond, \square, W\), and R can be derived as in CTL*. The step-bounded until operator has been omitted from the syntax of PCTL* path formulae for the sake of simplicity. It can be defined in terms of the other operators in the following way:
\[
\varphi_{1} \mathrm{U} \leqslant n \varphi_{2}=\bigvee_{0 \leqslant i \leqslant n} \psi_{i} \quad \text { where } \psi_{0}=\varphi_{2} \text { and } \psi_{i+1}=\varphi_{1} \wedge \bigcirc \psi_{i} \text { for } i \geqslant 0
\]

The logic PCTL can thus be regarded as a sublogic of PCTL*.
For a given Markov chain \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\), the satisfaction relation for PCTL \(^{*}\) state- and path formulae is defined as for PCTL and CTL*, respectively. For example,
\[
s \models \mathbb{P}_{J}(\varphi) \quad \text { iff } \quad \operatorname{Pr}(s \models \varphi) \in J
\]
where \(\operatorname{Pr}(s \models \varphi)=\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \varphi\}\). The satisfaction relation for the path formulae is exactly as in CTL*. Let \(\operatorname{Sat}_{\mathcal{M}}(\Phi)\), or briefly \(\operatorname{Sat}(\Phi)\), denote the set \(\{s \in S \mid s \models \Phi\}\). A PCTL* state formula \(\Phi\) is said to hold for \(\mathcal{M}\), denoted \(\mathcal{M} \vDash \Phi\), if \(s \models \Phi\) for all states \(s \in S\) with \(\iota_{\text {init }}(s)>0\).

The equivalence of PCTL* state formulae is defined as for PCTL, i.e.,
\[
\Phi \equiv \Psi \quad \text { iff } \quad \operatorname{Sat}_{\mathcal{M}}(\Phi)=\operatorname{Sat}_{\mathcal{M}}(\Psi) \text { for all Markov chains } \mathcal{M}
\]
(It is assumed that the set \(A P\) of atomic propositions is fixed.) In case we are restricted to finite Markov chains, we write \(\equiv_{f}\), i.e.,
\[
\Phi \equiv_{f} \Psi \quad \text { iff } \quad \operatorname{Sat}_{\mathcal{M}}(\Phi)=\operatorname{Sat}_{\mathcal{M}}(\Psi) \text { for all finite Markov chains } \mathcal{M}
\]

The results established for repeated reachability and persistence properties (see Theorem 10.47, page 795 and Theorem 10.48, page 795, respectively, can now be rephrased by
\[
\begin{array}{ll}
\mathbb{P}_{J}(\square \diamond \Phi) & \equiv_{f} \quad \mathbb{P}_{J}\left(\diamond \mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond \Phi)\right)\right) \\
\mathbb{P}_{J}(\diamond \square \Phi) & \equiv_{f} \quad \mathbb{P}_{J}\left(\diamond \mathbb{P}_{=1}(\square \Phi)\right)
\end{array}
\]

Let us now consider the model-checking problem for PCTL*, i.e., for a given finite Markov chain \(\mathcal{M}\) and PCTL formula \(\Phi\), does \(\mathcal{M} \vDash \Phi\) hold? This problem can be tackled by combining the procedure for PCTL model checking and the techniques discussed before for the quantitative analysis of LTL formulae. The main procedure goes along the lines of the model-checking algorithm for CTL* and relies on a bottom-up treatment of the parse tree of \(\Phi\). For each inner node (corresponding to state subformula \(\Psi\) of \(\Phi\) ), the satisfaction set \(\operatorname{Sat}(\Psi)=\{s \in S \mid s \models \Psi\}\) is computed. For the propositional logic fragment this computation is obvious and is as for CTL. The interesting case is a state formula of the form \(\Psi=\mathbb{P}_{J}(\varphi)\). In order to treat such formulae, first all maximal state subformulae of \(\varphi\) are replaced by new atomic propositions. Intuitively, these atomic propositions represent the satisfaction sets for these subformulae. Due to the bottom-up nature of the model-checking algorithm, these satisfaction sets have been determined already, and this replacement amounts to labeling the states in these sets. Applying this replacement procedure to \(\varphi\) yields an LTL formula \(\varphi^{\prime}\). Using the techniques explained in Section 10.3, one computes for each state \(s\) the probability \(\operatorname{Pr}\left(s \models \varphi^{\prime}\right)\) of satisfying \(\varphi^{\prime}\) (and thus \(\varphi\) ). The returned result is then:
\[
\operatorname{Sat}(\Psi)=\{s \in S \mid \operatorname{Pr}(s \models \varphi) \in J\}
\]

Due to the double-exponential transformation of an LTL formula \(\varphi^{\prime}\) into a deterministic Rabin automaton, the time complexity of model-checking PCTL* is double exponential in \(|\varphi|\) and polynomial in the size of the Markov chain \(\mathcal{M}\). Using alternative techniques may yield a single exponential time complexity in \(|\varphi|\). A further improvement cannot be expected due to the PSPACE-completeness of the qualitative model-checking problem for LTL; see Theorem 10.58, page 805.

\subsection*{10.4.2 Probabilistic Bisimulation}

To compare the behavior of transition systems, bisimulation and simulation relations have been extensively treated in Chapter 7. Bisimulation relations are equivalences requiring two bisimilar states to be equally labeled and exhibit identical stepwise behavior. It can be lifted to transition systems by comparing their initial states. Bisimilar states thus need to be able to mimic each individual step of each other. This section considers a notion of bisimulation on Markov chains that takes the transition probabilities into account. They can be viewed as a quantitative variant of bisimulation for transition systems. The crucial idea is to require bisimulation-equivalent states to have the same transition probability for each equivalence class (under bisimulation). As for transition systems, bisimulationequivalent states must be equally labeled.

\section*{Definition 10.60. Bisimulation for Markov Chains}

Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov chain. A probabilistic bisimulation on \(\mathcal{M}\) is an equivalence relation \(\mathcal{R}\) on \(S\) such that for all states \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) :
1. \(L\left(s_{1}\right)=L\left(s_{2}\right)\).
2. \(\mathbf{P}\left(s_{1}, T\right)=\mathbf{P}\left(s_{2}, T\right)\) for each equivalence class \(T \in S / \mathcal{R}\).

States \(s_{1}\) and \(s_{2}\) are bisimulation-equivalent (or bisimilar), denoted \(s_{1} \sim_{\mathcal{M}} s_{2}\), if there exists a bisimulation \(\mathcal{R}\) on \(\mathcal{M}\) such that \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\).

A probabilistic bisimulation is often referred to as bisimulation in the sequel. The first condition states that the states are equally labeled. The last condition requires that for bisimilar states the probability of moving by a single transition to some equivalence class is equal. Recall that \(\mathbf{P}(s, T)=\sum_{t \in T} \mathbf{P}(s, t)\) denotes the probability of moving from state \(s\) directly to some state in \(T\). As opposed to bisimulation on states in transition systems (see Definition 7.7 on page 456), any probabilistic bisimulation is required to be
an equivalence - otherwise it would be senseless to refer to equivalence classes in the last constraint. This is in contrast to the nonprobabilistic case where bisimulation relations for a single transition system \(T S\) need not be symmetric (although the coarsest bisimulation \(\sim_{T S}\) is, in fact, an equivalence).

As for transition systems, the above notion of (probabilistic) bisimulation equivalence for the states of a single Markov chain can be extended to compare two Markov chains \(\mathcal{M}_{1}\), \(\mathcal{M}_{2}\). Let \(\mathcal{M}_{1}, \mathcal{M}_{2}\) be Markov chains over the same set of atomic propositions with initial distributions \(\iota_{\text {init }}^{1}\) and \(\iota_{\text {init }}^{2}\), respectively. Consider the Markov chain \(\mathcal{M}=\mathcal{M}_{1} \uplus \mathcal{M}_{2}\) that results from the disjoint union of \(\mathcal{M}_{1}\) and \(\mathcal{M}_{2}\). Then \(\mathcal{M}_{1}\) and \(\mathcal{M}_{2}\) are bisimilar if
\[
\iota_{\text {init }}^{1}(T)=\iota_{\text {init }}^{2}(T)
\]
for each bisimulation equivalence class \(T\) of \(\mathcal{M}=\mathcal{M}_{1} \uplus \mathcal{M}_{2}\). Here, \(\iota_{\text {init }}(T)\) denotes \(\sum_{t \in T} \iota_{\text {init }}(t)\).

\section*{Example 10.61. Bisimulation for Markov Chains}

Consider the Markov chain in Figure 10.11. (Note that it consists of two mutually unreachable parts.) The reflexive, symmetric, and transitive closure of the relation
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right),\left(u_{1}, u_{3}\right),\left(u_{2}, u_{3}\right),\left(v_{1}, v_{3}\right),\left(v_{2}, v_{3}\right)\right\}
\]
is a probabilistic bisimulation. This can be seen as follows. The equivalence classes are \(T_{1}=\left[s_{1}\right]_{\sim}=\left\{s_{1}, s_{2}\right\}, T_{2}=\left[u_{1}\right]_{\sim}=\left\{u_{1}, u_{2}, u_{3}\right\}\) and \(T_{3}=\left[v_{1}\right]_{\sim}=\left\{v_{1}, v_{2}, v_{3}\right\}\). It follows that
\[
\mathbf{P}\left(s_{1}, T_{1}\right)=\mathbf{P}\left(s_{2}, T_{1}\right)=0, \mathbf{P}\left(s_{1}, T_{2}\right)=\mathbf{P}\left(s_{2}, T_{2}\right)=\frac{2}{3}, \mathbf{P}\left(s_{1}, T_{3}\right)=\mathbf{P}\left(s_{2}, T_{3}\right)=\frac{1}{3}
\]

In a similar way, it can be established that for all states in \(T_{2}\) and \(T_{3}\) the probabilities of transition to each of the equivalence classes are the same.

\section*{Definition 10.62. Bisimulation Quotient}

Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov chain. The quotient Markov chain \(\mathcal{M} / \sim_{\mathcal{M}}\) is defined by
\[
\mathcal{M} / \sim_{\mathcal{M}}=\left(S / \sim_{\mathcal{M}}, \mathbf{P}^{\prime}, \iota_{\text {init }}^{\prime}, A P, L^{\prime}\right)
\]
where \(\mathbf{P}^{\prime}\left([s]_{\sim},[t]_{\sim}\right)=\mathbf{P}\left(s,[t]_{\sim}\right), \iota_{\text {init }}^{\prime}\left([s]_{\sim}\right)=\sum_{s^{\prime} \in[s]_{\text {init }}}(s)\) and \(L^{\prime}\left([s]_{\sim}\right)=L(s)\).

The state space of the quotient Markov chain \(\mathcal{M} / \sim_{\mathcal{M}}\) is the set of equivalence classes under \(\sim_{\mathcal{M}}\). The transition probability from equivalence class \([s]_{\sim}\) to \([t]_{\sim}\) equals \(\mathbf{P}\left(s,[t]_{\sim}\right)\).


Figure 10.11: Two examples of Markov chains for which \(s_{1} \sim_{\mathcal{M}} s_{2}\).

Note that this is well-defined as \(\mathbf{P}(s, T)=\mathbf{P}\left(s^{\prime}, T\right)\) for all \(s \sim s^{\prime}\) and all bisimulation equivalence classes \(T\).

\section*{Example 10.63. The Craps Game}

Consider the Markov chain modeling the craps game; see Figure 10.3 (page 751). Assume all states are labeled with the empty set, except for \(L(\) won \()=\{\) won \(\}\). The equivalence relation that induces the following partitioning of the state space
\[
\{\text { start }\},\{\text { won }\},\{\text { lost }\},\{4,10\},\{5,9\},\{6,8\}
\]
is a probabilistic bisimulation. This can be seen as follows. The fact that state won is not bisimilar to any other state is clear as it is labeled differently. State lost is not bisimilar to any other state, as it is the only absorbing state labeled with \(\varnothing\). As state start is the only state that can move to \(\{\) won \(\}\) with probability \(\frac{2}{9}\), it is not bisimilar to any other state. Consider now, e.g., states 5 and 9 . Both states can move to won with the same probability. The same applies to any equivalence class \(T\). States 5 and 9 are thus bisimilar. A similar reasoning applies to the other states. The quotient Markov chain is depicted in Figure 10.12 .

The remainder of this section is focused on establishing that bisimulation-equivalent states satisfy the same PCTL* formulae. This means in particular that \(s\) and any state in \([s]_{\sim}\) satisfy the same PCTL* formulae. Checking whether \(\mathcal{M} \vDash \Phi\) for PCTL* formula \(\Phi\) may thus be established by checking \(\mathcal{M} / \sim \models \Phi\). In fact, it will be shown that PCTL equivalence, \(\mathrm{PCTL}^{*}\) equivalence, and probabilistic bisimulation coincide. This means that in order to show that two states are not bisimilar it suffices to indicate a PCTL (or PCTL*) formula that distinguishes them.

The key to establishing that bisimulation-equivalent states satisfy the same PCTL* formula is the following observation. Let \(s \sim_{\mathcal{M}} s^{\prime}\) where \(s\) and \(s^{\prime}\) are states in the Markov


Figure 10.12: Bisimulation quotient for the craps game.
chain \(\mathcal{M}\). Then, the probability measures \(\operatorname{Pr}_{s}^{\mathcal{M}}\) and \(\operatorname{Pr}_{s^{\prime}}^{\mathcal{M}}\) agree for all measurable sets of paths that are closed under statewise bisimulation equivalence. In order to formally prove this property, we need some additional notations and some standard concepts from measure theory. Let us first lift the notion of bisimulation to paths.

\section*{Definition 10.64. Bisimulation-Equivalent Paths}

The infinite paths \(\pi_{1}=s_{0,1} s_{1,1} s_{2,1} \ldots\) and \(\pi_{2}=s_{0,2} s_{1,2} s_{2,2} \ldots\) in a Markov chain \(\mathcal{M}\) are bisimulation equivalent, denoted \(\pi_{1} \sim_{\mathcal{M}} \pi_{2}\), if they are statewise bisimilar:
\[
\pi_{1} \sim_{\mathcal{M}} \pi_{2} \quad \text { if and only if } \quad s_{i, 1} \sim_{\mathcal{M}} s_{i, 2} \text { for all } i \geqslant 0
\]

\section*{Definition 10.65. Bisimulation-Closed \(\sigma\)-Algebra}

Let \(\mathcal{M}\) be a Markov chain with state space \(S\) and \(T_{0}, T_{1}, \ldots, T_{n} \in S / \sim_{\mathcal{M}}\) be equivalence classes of \(\sim_{\mathcal{M}}\). The bisimulation-closed \(\sigma\)-algebra \(\mathfrak{E}_{\sim \mathcal{M}}^{\mathcal{M}}\) denotes the \(\sigma\)-algebra generated by the sets \(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\) where \(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\) is the set of all paths \(t_{0} t_{1} \ldots t_{n} t_{n+1} t_{n+2} \ldots\) with \(t_{i} \in T_{i}\) for \(0 \leqslant i \leqslant n\).

All sets in the bisimulation-closed \(\sigma\)-algebra \(\mathfrak{E}_{\sim}^{\mathcal{M}}\) are measurable with respect to the stan-
dard \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\) associated with \(\mathcal{M}\) (see Definition 10.10). That is to say:
\[
\mathfrak{E}_{\sim}^{\mathcal{M}} \subseteq \mathfrak{E}^{\mathcal{M}}
\]

This inclusion is due to the fact that the basic events \(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\) of the \(\sigma\)-algebra \(\mathfrak{E}_{\sim}^{\mathcal{M}}\) are countable unions of basic elements-cylinder sets spanned by the finite path fragments in \(\mathcal{M}\)-of the \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\). Formally:
\[
\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)=\bigcup_{\substack{ \\t_{0} t_{1} \ldots t_{n} \in \operatorname{Paths}_{f i n}(\mathcal{M}) \\ t_{i} \in T_{i}, 0 \leqslant i \leqslant n}} \operatorname{Cyl}\left(t_{0} t_{1} \ldots t_{n}\right)
\]
where \(T_{0}, T_{1}, \ldots, T_{n} \in S / \sim_{\mathcal{M}}\) are bisimulation-equivalence classes. Thus, the bisimulationclosed \(\sigma\)-algebra \(\mathfrak{E}_{\sim}^{\mathcal{M}}\) is a so-called sub- \(\sigma\)-algebra of the \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\). We refer to the elements of \(\mathfrak{E}_{\sim}^{\mathcal{M}}\) as bisimulation-closed events.

In fact, the events in the bisimulation-closed \(\sigma\)-algebra \(\mathfrak{E}_{\sim}^{\mathcal{M}}\) are bisimulation-closed measurable sets of paths in \(\mathfrak{E}^{\mathcal{M}}\). Let \(\Pi \in \mathfrak{E}^{\mathcal{M}}\) be a measurable set of paths. Note that measurability is understood with respect to the standard \(\sigma\)-algebra \(\mathfrak{E}^{\mathcal{M}}\) on the paths of \(\mathcal{M}\). The set \(\Pi\) is bisimulation-closed if for any \(\pi_{1} \in \Pi\) and \(\pi_{2}\) such that \(\pi_{1} \sim_{\mathcal{M}} \pi_{2}\) it holds that \(\pi_{2} \in \Pi\). That is:
\[
\mathfrak{E}_{\sim}^{\mathcal{M}}=\left\{\Pi \in \mathfrak{E}^{\mathcal{M}} \mid \Pi \text { is bisimulation-closed }\right\} .
\]

The following result asserts that the probability measure \(\operatorname{Pr}_{s}^{\mathcal{M}}\) for bisimulation-closed sets of paths agree for bisimilar states.

Lemma 10.66. Preservation of Probabilities of Bisimulation-Closed Events Let \(\mathcal{M}\) be a Markov chain. For all states \(s_{1}, s_{2}\) in \(\mathcal{M}\) :
\[
s_{1} \sim \mathcal{M} s_{2} \quad \text { implies } \quad \operatorname{Pr}_{s_{1}}^{\mathcal{M}}(\Pi)=\operatorname{Pr}_{s_{2}}^{\mathcal{M}}(\Pi)
\]
for all bisimulation-closed events \(\Pi \subseteq \operatorname{Paths}(\mathcal{M})\).

Proof: Since the bisimulation-closed \(\sigma\)-algebra \(\mathfrak{E}_{\sim}^{\mathcal{M}}\) is closed under intersection, by standard results of measure theory it holds that: for fixed function \(f:\left(S / \sim_{\mathcal{M}}\right)^{+} \rightarrow[0,1]\), there exists at most one probability measure \(\mu\) on the bisimulation-closed \(\sigma\)-algebra \(\mathfrak{E}_{\sim}^{\mathcal{M}}\) such that
\[
\mu\left(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\right)=f\left(T_{0} T_{1} \ldots T_{n}\right)
\]
for all bisimulation equivalence classes \(T_{i}, 0 \leqslant i \leqslant n\).

This result yields that it suffices to show that \(s_{1} \sim_{\mathcal{M}} s_{2}\) implies that \(\operatorname{Pr}_{s_{1}}^{\mathcal{M}}\) and \(\operatorname{Pr}_{s_{2}}^{\mathcal{M}}\) agree on the basic events \(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\) of \(\mathfrak{E} \mathcal{\sim}\). For bisimulation equivalence classes \(T, U\) of \(\mathcal{M}\), let \(\mathbf{P}(T, U)\) denote the value \(\mathbf{P}(t, U)\) for all (some) \(t \in T\). Then we have
\[
\begin{aligned}
& \operatorname{Pr}_{s_{1}}^{\mathcal{M}}\left(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\right) \\
= & \mathbf{P}\left(T_{0}, T_{1}\right) \cdot \mathbf{P}\left(T_{1}, T_{2}\right) \ldots \cdot \mathbf{P}\left(T_{n-1}, T_{n}\right) \\
= & \operatorname{Pr}_{s_{2}}^{\mathcal{M}}\left(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\right)
\end{aligned}
\]
provided that \(s_{1}, s_{2} \in T_{0}\). Otherwise, i.e., if \(s_{1}, s_{2} \notin T_{0}\), then:
\[
\operatorname{Pr}_{s_{1}}^{\mathcal{M}}\left(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\right)=0=\operatorname{Pr}_{s_{2}}^{\mathcal{M}}\left(\operatorname{Cyl}\left(T_{0} T_{1} \ldots T_{n}\right)\right)
\]

For finitely-branching transition systems, CTL-equivalence, CTL*-equivalence, and bisimulation equivalence all coincide. An analogous result holds for PCTL, PCTL*, and probabilistic bisimulation equivalence. In fact, the restriction to finitely-branching systems is not necessary. That is, PCTL, PCTL*, and probabilistic bisimulation equivalence coincide for any arbitrary, possibly infinite, Markov chain. In the probabilistic setting, nonbisimilar states can even be distinguished by a formula of a (small) fragment of PCTL that just consists of atomic propositions, conjunction, and the operator \(\mathbb{P}_{\leqslant p}(\bigcirc \cdot)\). As in the nonprobabilistic setting, the until operator is thus not necessary for logically characterizing bisimulation. As opposed to the nonprobabilistic case where full propositional logic (containing conjunction and negation) is necessary, negation is not needed in the probabilistic setting.

Let \(\mathrm{PCTL}^{-}\)denote the following fragment of PCTL. State formulae in \(\mathrm{PCTL}^{-}\)are formed according to
\[
\Phi::=a\left|\Phi_{1} \wedge \Phi_{2}\right| \mathbb{P}_{\leqslant p}(\bigcirc \Phi)
\]
where \(a \in A P\) and \(p\) is a rational number in \([0,1]\). Negation is present neither as an operator in \(\mathrm{PCTL}^{-}\)nor can it be expressed. The results indicated above are summarized in the following theorem:

Theorem 10.67. PCTL/PCTL* and Bisimulation Equivalence
Let \(\mathcal{M}\) be a Markov chain and \(s_{1}\), \(s_{2}\) states in \(\mathcal{M}\). Then, the following statements are equivalent:
(a) \(s_{1} \sim_{\mathcal{M}} s_{2}\).
(b) \(s_{1}\) and \(s_{2}\) are PCTL*-equivalent, i.e., fulfill the same PCTL* formulae.
(c) \(s_{1}\) and \(s_{2}\) are PCTL-equivalent, i.e., fulfill the same PCTL formulae.
(d) \(s_{1}\) and \(s_{2}\) are PCTL--equivalent, i.e., fulfill the same PCTL- formulae.

Proof: \((\mathrm{a}) \Longrightarrow(\mathrm{b})\) : Using Lemma 10.66 on page 812 it can be shown by structural induction on the syntax of PCTL* state formula \(\Phi\) and PCTL* path formula \(\varphi\) :
(1) For states \(s_{1}, s_{2}\) in \(\mathcal{M}\), if \(s_{1} \sim_{\mathcal{M}} s_{2}\), then:
(1.1) \(s_{1} \models \Phi\) if and only if \(s_{2} \models \Phi\).
(1.2) \(\operatorname{Pr}\left(s_{1} \models \varphi\right)=\operatorname{Pr}\left(s_{2} \models \varphi\right)\).
(2) For paths \(\pi_{1}, \pi_{2}\) in \(\mathcal{M}\), if \(\pi_{1} \sim_{\mathcal{M}} \pi_{2}\), then \(\pi_{1} \models \varphi\) if and only if \(\pi_{2} \models \varphi\).

The proof of the first part is fairly similar to the nonprobabilistic case (see Theorem 7.20 on page 469). We only consider statement (1.2) and assume as an induction hypothesis that (2) holds for \(\varphi\). Let \(s_{1} \sim_{\mathcal{M}} s_{2}\) and \(\Pi\) be the set of paths in \(\mathcal{M}\) satisfying \(\varphi\), i.e., \(\Pi=\{\pi \in \operatorname{Paths}(\mathcal{M}) \mid \pi \vDash \varphi\}\). The induction hypothesis applied to \(\varphi\) (see item (2)) yields that \(\Pi\) is bisimulation-closed. Hence, by Lemma 10.66 it follows that
\[
\operatorname{Pr}\left(s_{1} \models \varphi\right)=\operatorname{Pr}_{s_{1}}^{\mathcal{M}}(\Pi)=\operatorname{Pr}_{s_{2}}^{\mathcal{M}}(\Pi)=\operatorname{Pr}\left(s_{2} \models \varphi\right) .
\]
\((\mathrm{b}) \Longrightarrow(\mathrm{c})\) and \((\mathrm{c}) \Longrightarrow(\mathrm{d})\) : Obvious, since PCTL is a sublogic of PCTL* and \(\mathrm{PCTL}^{-}\)is a sublogic of PCTL.
\((\mathrm{c}) \Longrightarrow(\mathrm{a})\) : Prior to proving that (a) is a consequence of (d), we first treat a simpler case and prove that PCTL-equivalent states in a finite Markov chain are probabilistically bisimilar. The aim of this step is to show that roughly the same arguments as in the nonprobabilistic case can be applied. We have to show that the relation
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid s_{1} \equiv_{P C T L} s_{2}\right\}
\]
is a probabilistic bisimulation. (Here, \(\equiv_{P C T L}\) denotes PCTL equivalence of states.) Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\). As \(s_{1}\) and \(s_{2}\) fulfill the same atomic propositions, it follows that \(L\left(s_{1}\right)=\) \(L\left(s_{2}\right)\). It remains to prove that \(\mathbf{P}\left(s_{1}, T\right)=\mathbf{P}\left(s_{2}, T\right)\) for each equivalence class \(T\) under \(\mathcal{R}\). Let \(T, U\) be \(\mathcal{R}\)-equivalence classes with \(T \neq U\), and PCTL formula \(\Phi_{T, U}\) be such that \(\operatorname{Sat}\left(\Phi_{T, U}\right) \subseteq T\) and \(\operatorname{Sat}\left(\Phi_{T, U}\right) \cap U=\varnothing\). Define:
\[
\Phi_{T}=\bigwedge_{\substack{U \in S / \sim \mathcal{M} \\ U \neq T}} \Phi_{T, U}
\]

It is evident that \(\operatorname{Sat}\left(\Phi_{T}\right)=T\). That is to say, \(\Phi_{T}\) is a PCTL master formula for the equivalence class \(T\). Without loss of generality, assume \(\mathbf{P}\left(s_{1}, T\right) \leqslant \mathbf{P}\left(s_{2}, T\right)\). Let \(s_{1} \vDash\) \(\mathbb{P}_{\leqslant p}\left(\Phi_{T}\right)\) where \(p=\mathbf{P}\left(s_{1}, T\right)\). As \(s_{1} \equiv\) PCTL \(s_{2}\), it follows that \(s_{2} \models \mathbb{P}_{\leqslant p}\left(\Phi_{T}\right)\). But then \(\mathbf{P}\left(s_{1}, T\right)=p=\mathbf{P}\left(s_{2}, T\right)\). This yields condition (2) for \(\mathcal{R}\).
\((\mathrm{d}) \Longrightarrow(\mathrm{a})\) : Assume \(\mathcal{M}\) is an arbitrary (possibly infinite) Markov chain. The goal is to show that
\[
\mathcal{R}=\left\{\left(s_{1}, s_{2}\right) \in S \times S \mid s_{1} \equiv_{P C T L^{-}} s_{2}\right\}
\]
is a probabilistic bisimulation. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\). The fact that \(s_{1}\) and \(s_{2}\) are equally labeled is established as in the previous part of the proof (i.e., (c) implies (a)). However, the above argument to prove that \(\mathbf{P}\left(s_{1}, T\right)=\mathbf{P}\left(s_{2}, T\right)\) for any \(\mathcal{R}\)-equivalence class \(T\) is not applicable in case there are infinitely many bisimulation equivalence classes. (Note that then \(\Phi_{T}\) would be defined by an infinite conjunction.) Let the satisfaction sets \(\operatorname{Sat}(\Phi)\) for \(\mathrm{PCTL}^{-}\)state formula \(\Phi\) be basic events on the state space \(S\) of \(\mathcal{M}\) and \(\mathfrak{E}_{S}\) be the smallest \(\sigma\)-algebra on \(S\) that contains the sets \(\operatorname{Sat}(\Phi)\). Since the set of all \(\mathrm{PCTL}^{-}\)formulae is countable, any \(\mathrm{PCTL}^{-}\)-equivalence class \(T \in S / \mathcal{R}\) can be written as the countable intersection of the satisfaction sets \(\operatorname{Sat}(\Phi)\) where \(\Phi\) is a \(\mathrm{PCTL}^{-}\)formula and \(T \subseteq \operatorname{Sat}(\Phi)\). Thus, all PCTL \({ }^{-}\)-equivalence classes \(T \in S / \mathcal{R}\) belong to \(\mathfrak{E}_{S}\).

As \(\mathrm{PCTL}^{-}\)permits conjunction, the set of all satisfaction sets \(\operatorname{Sat}(\Phi)\) is closed under finite intersections. By a standard result of measure theory, we have that for every probability measure \(\mu_{1}, \mu_{2}\) on \(\mathfrak{E}_{S}\) :
\[
\mu_{1}(\operatorname{Sat}(\Phi))=\mu_{2}(\operatorname{Sat}(\Phi)) \text { for any } \mathrm{PCTL}^{-} \text {formula } \Phi \quad \text { implies } \quad \mu_{1}=\mu_{2}
\]

In the remainder of the proof, this result is exploited to show that \(\operatorname{Pr}\left(s_{1}, T\right)=\operatorname{Pr}\left(s_{2}, T\right)\) for all states \(s_{1}, s_{2}\) with \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and all \(T \in S / \mathcal{R}\).

For state \(s\) in \(\mathcal{M}\), let the probability measure \(\mu_{s}\) on \(\mathfrak{E}_{S}\) be defined by
\[
\mu_{s}(T)=\mathbf{P}(s, T)=\sum_{t \in T} \mathbf{P}(s, t) \quad \text { for } T \in \mathfrak{E}_{S}
\]

Clearly, \(\mu_{s}(\operatorname{Sat}(\Phi))=\mathbf{P}(s, \operatorname{Sat}(\Phi))\) for any \(\mathrm{PCTL}^{-}\)formula \(\Phi\). The goal is now to show that \(\mu_{s_{1}}=\mu_{s_{2}}\) for PCTL \({ }^{-}\)-equivalent states \(s_{1}\) and \(s_{2}\). This is established as follows. Let \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) and \(\Phi\) be a \(\mathrm{PCTL}^{-}\)formula such that \(\mathbf{P}\left(s_{1}, \operatorname{Sat}(\Phi)\right)\) equals \(p\), say, and \(\mathbf{P}\left(s_{2}, \operatorname{Sat}(\Phi)\right)\) equals \(q\). Without loss of generality, assume \(p \leqslant q\). Then:
\[
s_{1} \models \mathbb{P}_{\leqslant p}(\bigcirc \Phi)
\]

As \(\mathbb{P}_{\leqslant p}(\bigcirc \Phi)\) is a \(\mathrm{PCTL}^{-}\)formula and \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\), it follows that \(s_{2} \models \mathbb{P}_{\leqslant p}(\bigcirc \Phi)\). But then \(q=\mathbf{P}\left(s_{2}, \operatorname{Sat}(\Phi)\right) \leqslant p \leqslant q\), and hence:
\[
\mu_{s_{1}}(\operatorname{Sat}(\Phi))=p=q=\mu_{s_{2}}(\operatorname{Sat}(\Phi))
\]

The probability measures \(\mu_{s_{1}}\) and \(\mu_{s_{2}}\) on \(\mathfrak{E}_{S}\) thus coincide for all basic events \(\operatorname{Sat}(\Phi)\) of \(\mathfrak{E}_{S}\). As the set of basic events \(\operatorname{Sat}(\Phi)\) is closed under intersection, any measure on \(\mathfrak{E}_{S}\) is uniquely determined by its values on the basic events. Thus, \(\mu_{s_{1}}(T)=\mu_{s_{2}}(T)\) for all \(T \in \mathfrak{E}_{S}\). All PCTL--equivalence classes \(T \in S / \mathcal{R}\) belong to \(\mathfrak{E}_{S}\), so:
\[
\mathbf{P}\left(s_{1}, T\right)=\mu_{s_{1}}(T)=\mu_{s_{2}}(T)=\mathbf{P}\left(s_{2}, T\right)
\]

Thus, \(\mathcal{R}\) is a probabilistic bisimulation.

The importance of Theorem 10.67 is manifold. First, it states that bisimulation equivalence preserves all quantitative PCTL*-definable properties. This justifies considering bisimulation-equivalent states to be "equal". Second, Theorem 10.67 asserts that bisimulation equivalence is the coarsest equivalence enjoying this property. That is to say, any strictly coarser equivalence identifies states with different probabilistic behavior in the sense of PCTL*-definable properties, and even some relatively simple properties that can be stated in PCTL \({ }^{-}\). Moreover, Theorem 10.67 can be used to prove that two states are not bisimulation equivalent. In order to do so, it suffices to provide a PCTL* formula (or \(\mathrm{PCTL}^{-}\)or PCTL formula) that distinguishes the given states. Finally, observe that probabilistic bisimulation equivalence of states \(s_{1}\) and \(s_{2}\) in a Markov chain \(\mathcal{M}\) implies bisimulation equivalence of \(s_{1}\) and \(s_{2}\) (as considered in Chapter 7) in the transition system associated with \(\mathcal{M}\).

\section*{Example 10.68. Craps Game}

Consider the probabilistic bisimulation quotient for the craps game, see Figure 10.12. It follows that \(s_{4,10} \chi_{\mathcal{M}} s_{6,8}\), since there exists a PCTL formula, e.g., \(\mathbb{P}_{<\frac{1}{6}}(\bigcirc\) won), which holds in \(s_{4,10}\) but not in \(s_{6,8}\).

\subsection*{10.5 Markov Chains with Costs}

In addition to the probability of certain events, it is natural to analyze the average behavior of executions in a Markov chain. For instance, for a communication system where a sender and a receiver can transfer messages via an unreliable channel, an interesting measure of interest is the expected number of attempts to send a message until correct delivery. Another example is a multiprocessor system where one might be interested in the average number of steps between two successive failures - the mean "time" between failures. For a battery-powered embedded system, a measure of interest is the expected power consumption during operation.

The aim of this section is to consider an extension of Markov chains, called Markov reward chains, and to consider expected measures. A Markov reward chain is a Markov chain in which states (or transitions) are augmented with rewards, natural numbers that can be interpreted as bonuses, or dually as costs. We consider equipping states with rewards. The idea is that whenever a state \(s\) is left, the reward associated with \(s\) is earned.

\section*{Definition 10.69. Markov Reward Model (MRM)}

A Markov reward model (MRM) is a tuple ( \(\mathcal{M}\), rew) with \(\mathcal{M}\) a Markov chain with state space \(S\) and rew : \(S \rightarrow \mathbb{N}\) a reward function that assigns to each state \(s \in S\) a non-negative integer reward rew \((s)\).

Intuitively, the value \(\operatorname{rew}(s)\) stands for the reward earned on leaving state \(s\). Formally, the cumulative reward for a finite path \(\widehat{\pi}=s_{0} s_{1} \ldots s_{n}\) is defined by
\[
\operatorname{rew}(\widehat{\pi})=\operatorname{rew}\left(s_{0}\right)+\operatorname{rew}\left(s_{1}\right)+\ldots+\operatorname{rew}\left(s_{n-1}\right)
\]

Note that the reward of the last state \(s_{n}\) in the path \(\widehat{\pi}\) is not considered.

\section*{Example 10.70. Zeroconf Protocol}

Consider the Markov chain modeling the behavior of a single station in the zeroconf protocol 10.5 (page 751). For convenience, the Markov chain is depicted in Figure 10.13. We consider three reward functions for this model:


Figure 10.13: Markov chain of the \(\operatorname{IPv4}\) zeroconf protocol (for \(n=4\) probes).
- The first reward assignment (denoted rew \({ }_{1}\) ) represents waiting time - recall that on transmitting a probe, an acknowledgment is awaited for exactly \(r\) time units. It is defined by \(\operatorname{rew}_{1}\left(s_{i}\right)=r\) for \(0<i \leqslant n \operatorname{rew}_{1}\left(s_{0}\right)=0\) assuming that the host randomly selects an address promptly, \(\operatorname{rew}_{1}\left(s_{n+3}\right)=n \cdot r\), rew \(\left(s_{n+2}\right)=\operatorname{rew}_{1}\left(s_{n+4}\right)=0\), and
\(\operatorname{rew}_{1}\left(s_{n+1}\right)=E\), where \(E\) denotes some large number that represents the highly undesirable situation of an address collision.
- The second reward assignment (denoted rew r \(_{2}\) ) is used to keep track of the number of probes that are sent in total. It is defined by \(\operatorname{rew}_{2}\left(s_{i}\right)=1\) for \(0<i \leqslant n\), \(\operatorname{rew}_{2}\left(s_{n+3}\right)=n\) and 0 otherwise.
- Finally, the third reward assignment (denoted rew \({ }_{3}\) ) is used to keep track of the number of failed attempts to acquire an unused address. It is defined by \(\operatorname{rew}_{3}\left(s_{1}\right)=1\) and 0 otherwise.

Consider the finite path \(\widehat{\pi}=s_{0} s_{1} s_{0} s_{1} s_{2} s_{0} s_{7} s_{8}\). It follows that rew \((\widehat{\pi})=7 r\), rew \(W_{2}(\widehat{\pi})=7\) and \(\operatorname{rew}_{3}(\widehat{\pi})=2\).

\subsection*{10.5.1 Cost-Bounded Reachability}

Several quantitative properties of Markov reward models can be defined on the basis of cumulative rewards for finite paths. This section considers cost-bounded reachability, i.e., the expected reward before reaching a given set of states, and the probability of reaching these states within a given bound on the cumulative reward. Let us first consider expected rewards. Let ( \(\mathcal{M}\), rew) be an MRM with state space \(S\) and \(B \subseteq S\) the set of target states. For an infinite path \(\pi=s_{0} s_{1} s_{2} \ldots\) in \(\mathcal{M}\) let
\[
\operatorname{rew}(\pi, \diamond B)= \begin{cases}\operatorname{rew}\left(s_{0} s_{1} \ldots s_{n}\right) & \text { if } s_{i} \notin B \text { for } 0 \leqslant i<n \text { and } s_{n} \in B \\ \infty & \text { if } \pi \not \vDash \diamond B .\end{cases}
\]

Stated in words, \(\operatorname{rew}(\pi, \diamond B)\) denotes the cumulative reward earned along an infinite path \(\pi\) until reaching a \(B\)-state for the first time. The expected reward until reaching \(B\) is now defined as the expectation of the function \(\operatorname{rew}(\pi, \diamond B)\) :

\section*{Definition 10.71. Expected Reward for Reachability Properties}

For state \(s\) and \(B \subseteq S\), the expected reward until reaching \(B\) from \(s\) is defined as follows. If \(\operatorname{Pr}(s \models \diamond B)<1\), then \(\operatorname{ExpRew}(s \models \diamond B)=\infty\). Otherwise, i.e., if \(\operatorname{Pr}(s \models \diamond B)=1\), then:
\[
\operatorname{ExpRew}(s \models \diamond B)=\sum_{r=0}^{\infty} r \cdot \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \diamond B \wedge \operatorname{rew}(\pi, \diamond B)=r\}
\]

The infinite series for \(\operatorname{Pr}(s \models \diamond B)=1\) converges for any reward function rew. The intuitive argument for this is that the rewards along paths are added, while the probabilities of the transitions are multiplicated. The formal proof is left as an exercise to the reader; see Exercise 10.18 (page 904).

If \(\operatorname{Pr}(s \models \diamond B)=1\), then an equivalent characterization of the expected reward earned until \(B\) is reached can be provided by means of a weighted sum of the rewards earned along minimal path fragments from \(s\) to \(B\) :
\[
\operatorname{ExpRew}(s \models \diamond B)=\sum_{\widehat{\pi}} \mathbf{P}(\widehat{\pi}) \cdot \operatorname{rew}(\widehat{\pi})
\]
where \(\widehat{\pi}\) ranges over all finite paths \(s_{0} \ldots, s_{n}\) with \(s_{n} \in B, s_{0}=s\) and \(s_{0}, \ldots, s_{n-1} \notin B\).

\section*{Example 10.72. Simulating a Die by a Coin (Revisited)}

Consider again the Markov chain in Example 10.3 (page 750) which describes how a sixsided die can be simulated by a fair coin. For convenience, the Markov chain is illustrated again in Figure 10.14. In order to reason about the number of coin flips that are required


Figure 10.14: Markov chain for simulating a die by a fair coin.
to obtain a certain die outcome, the Markov chain is extended to an MRM. The set \(B\) of target states is thus \(B=\{1,2,3,4,5,6\}\). The reward function assigns reward 1 to all states except the states in \(B\), and reward zero to the states in \(B\). The cumulative reward of the infinite path \(\pi=s_{0} s_{1,2,3} s_{1,2,3}^{\prime} 1^{\omega}\) is \(\operatorname{rew}(\pi, \diamond B)=3\). The cumulative reward of \(\pi^{\prime}=s_{0}\left(s_{1,2,3} s_{1,2,3}^{\prime}\right)^{\omega}\) is \(\infty\).

In order to determine the number of rounds that are performed until a die outcome is obtained, an alternative reward function is used. A round is a visit to either state \(s_{1,2,3}\) or
\(s_{4,5,6}\). To reason about the number of rounds, reward 1 is associated with the states \(s_{1,2,3}\) and \(s_{4,5,6}\). All other states are equipped with reward zero. Using this reward function, we obtain for the expected number of rounds before reaching a target state in \(B\) :
\[
\operatorname{ExpRew}\left(s_{0} \mid \diamond B\right)=\sum_{r=1}^{\infty} r \cdot \frac{3}{4} \cdot\left(\frac{1}{4}\right)^{r-1}=\frac{4}{3}
\]

Note that in every round a target state is reached (in this round) with probability \(\frac{3}{4}\), while the probability for yet another round is \(\frac{1}{4}\). Thus, the probability for exactly \(r\) rounds agrees with the probability of reaching \(B\) from the initial state \(s_{0}\) while entering \(\left\{s_{1,2,3}, s_{3,4,5}\right\}\) exactly \(r\) times, i.e., \(\frac{3}{4} \cdot\left(\frac{1}{4}\right)^{r-1}\). The average number of rounds that is performed until an outcome is thus \(\frac{4}{3}\).

Just a brief comment on the computation of \(\sum_{r=1}^{\infty} r \cdot \frac{3}{4} \cdot\left(\frac{1}{4}\right)^{r-1}\). One possible way to compute the value of this infinite series is the observation that \(\sum_{r=1}^{\infty} r \cdot\left(\frac{1}{4}\right)^{r-1}\) can be viewed as the value of the first derivative of the function \(f:]-1,1[\rightarrow \mathbb{R}\) :
\[
f(x)=\sum_{r=0}^{\infty} x^{r}=\frac{1}{1-x}
\]
for \(x=\frac{1}{4}\). Then, \(f^{\prime}(x)=\sum_{r=1}^{\infty} r \cdot x^{r-1}=\frac{1}{(1-x)^{2}}\). Hence,
\[
\sum_{r=1}^{\infty} r \cdot\left(\frac{1}{4}\right)^{r-1}=f^{\prime}\left(\frac{1}{4}\right)=\frac{1}{\left(1-\frac{1}{4}\right)^{2}}=\frac{1}{\left(\frac{3}{4}\right)^{2}}
\]

Therefore: \(\sum_{r=1}^{\infty} r \cdot \frac{3}{4} \cdot\left(\frac{1}{4}\right)^{r-1}=\frac{3}{4} \cdot f^{\prime}\left(\frac{1}{4}\right)=\frac{3}{4} \cdot \frac{1}{\left(\frac{3}{4}\right)^{2}}=\frac{4}{3}\).

Let us now discuss in detail how to compute the expected rewards \(\operatorname{Exp} \operatorname{Rew}(s \vDash \diamond B)\), for finite MRMs. Using the graph-based techniques explained in Section 10.1.2 (page 770ff), we first determine the set \(S_{=1}\) of states \(s\) that reach \(B\) almost surely, i.e.:
\[
S_{=1}=\{s \in S \mid \operatorname{Pr}(s \models \diamond B)=1\} .
\]

The task is to compute \(x_{s}=\operatorname{Exp} \operatorname{Rew}(s \models \diamond B)\) for each state \(s \in S_{=1}\). For all states \(s \in S_{=1} \backslash B\), i.e., \(s \in S \backslash B\) for which \(\operatorname{Pr}(s \models \diamond B)=1\), we have \(\operatorname{Pr}(u \models \diamond B)=1\) for all direct successors \(u\) of \(s\). Hence, if \(s \in S_{=1}\), then either \(s \in B\) or \(\operatorname{Post}(s) \subseteq S_{=1}\). Moreover, the values \(x_{s}=\operatorname{Exp} \operatorname{Rew}(s \models \diamond B)\) provide a solution of the following equation system:
\[
x_{s}= \begin{cases}0 & \text { if } s \in B \\ \operatorname{rew}(s)+\sum_{u \in \operatorname{Post}(s)} \mathbf{P}(s, u) \cdot x_{u} & \text { if } s \in S_{=1} \backslash B\end{cases}
\]

In fact, the vector \(\left(x_{s}\right)_{s \in S=1}\) is the unique solution of the above linear equation system. This follows from the results established in the proof of Theorem 10.19 on page 766 for the subchain consisting of the states in \(S_{=1}\). Notice that the proof of Theorem 10.19 yields that
\[
\mathbf{A x}=\mathbf{x} \quad \text { implies } \quad \mathbf{x}=\mathbf{0}
\]
where \(\mathbf{A}=(\mathbf{P}(s, u))_{s, u \in S=1 \backslash B}\). Since \(\mathbf{A}\) is a quadratic matrix, this implication is equivalent to the nonsingularity of \(\mathbf{I}-\mathbf{A}\). That is, for any vector \(\mathbf{b}\), the linear equation system \(\mathbf{x}=\mathbf{A x}+\mathbf{b}\), which can be rewritten as \((\mathbf{I}-\mathbf{A}) \mathbf{x}=\mathbf{b}\), has a unique solution, namely \(\mathbf{x}=(\mathbf{I}-\mathbf{A})^{-1} \mathbf{b}\). In fact, the above equation system for the expected rewards can be rewritten in the form \(\mathbf{x}=\mathbf{A x}+\mathbf{b}\) where \(\mathbf{x}\) stands for the vector \(\left(x_{s}\right)_{s \in S_{=1} \backslash B}\) and \(\mathbf{b}=\left(b_{s}\right)_{s \in S=1 \backslash B}\) is the vector with the entries \(b_{s}=\operatorname{rew}(s)\).

Summarizing, computing the expected reward earned until a certain set of states will be reached in a finite MRM has a polynomial time complexity of the size of \(\mathcal{M}\). The necessary techniques are a graph analysis to determine \(S_{=1}\), and solving a linear equation system.

\section*{Example 10.73. Simulating a Die by a Coin (Revisited)}

Consider again Knuth and Yao's example of simulating a die by a fair coin. As in Example 10.72, consider the MRM where rewards serve as counters for the number of rounds. That is, \(\operatorname{rew}\left(s_{1,2,3}\right)=\operatorname{rew}\left(s_{4,5,6}\right)=1\), and \(\operatorname{rew}(s)=0\) for all other states. Let us compute the average number of rounds performed until an outcome is obtained, i.e., until a state in \(B=\{1,2,3,4,5,6\}\) is reached, by the above linear equation system. Clearly, \(\operatorname{Pr}(s \models\) \(\diamond B)=1\) for each state \(s\). Hence, all states are contained in \(S_{=1}\). Let \(x_{1,2,3}\) denote \(x_{s_{1,2,3}}\) and \(x_{1,2,3}^{\prime}\) denote \(x_{s_{1,2,3}^{\prime}}\). The values \(x_{s}=\operatorname{ExpRew}(s \models \diamond B)\) are a solution of the following equation system:
\[
\begin{aligned}
& x_{0}=\frac{1}{2} \cdot x_{1,2,3}+\frac{1}{2} \cdot x_{4,5,6} \\
& x_{1,2,3}=1+\frac{1}{2} \cdot x_{1,2,3}^{\prime}+\frac{1}{2} \cdot x_{2,3} \\
& x_{2,3}=\frac{1}{2} \cdot x_{2}+\frac{1}{2} \cdot x_{3} \\
& x_{1,2,3}^{\prime}=\frac{1}{2} \cdot x_{1,2,3}+\frac{1}{2} \cdot x_{1} \\
& x_{4,5,6}=1+\frac{1}{2} \cdot x_{4,5,6}^{\prime}+\frac{1}{2} \cdot x_{4,5} \\
& x_{4,5}=\frac{1}{2} \cdot x_{4}+\frac{1}{2} \cdot x_{5} \\
& x_{4,5,6}^{\prime}=\frac{1}{2} \cdot x_{4,5,6}+\frac{1}{2} \cdot x_{6} \\
& x_{1}=x_{2}=x_{3}=x_{4}=x_{5}=x_{6}=0
\end{aligned}
\]

Using the values \(x_{i}=0\) for \(0<i \leqslant 6\), we obtain \(x_{2,3}=x_{4,5}=0\). The equations for the
other states simplify to
\[
\begin{aligned}
x_{0} & =\frac{1}{2} \cdot x_{1,2,3}+\frac{1}{2} \cdot x_{4,5,6} \\
x_{1,2,3} & =1+\frac{1}{2} \cdot x_{1,2,3}^{\prime} \\
x_{1,2,3}^{\prime} & =\frac{1}{2} \cdot x_{1,2,3} \\
x_{4,5,6} & =1+\frac{1}{2} \cdot x_{4,5,6}^{\prime} \\
x_{4,5,6}^{\prime} & =\frac{1}{2} \cdot x_{4,5,6}
\end{aligned}
\]

This yields the linear equation system:
\[
\left(\begin{array}{rrrrr}
1 & -\frac{1}{2} & 0 & -\frac{1}{2} & 0 \\
0 & 1 & -\frac{1}{2} & 0 & 0 \\
0 & -\frac{1}{2} & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & -\frac{1}{2} \\
0 & 0 & 0 & -\frac{1}{2} & 1
\end{array}\right) \cdot\left(\begin{array}{l}
x_{0} \\
x_{1,2,3} \\
x_{1,2,3}^{\prime} \\
x_{4,5,6} \\
x_{4,5,6}^{\prime}
\end{array}\right)=\left(\begin{array}{l}
0 \\
1 \\
0 \\
1 \\
0
\end{array}\right)
\]

The unique solution of this linear equation system is \(x_{0}=x_{1,2,3}=x_{4,5,6}=\frac{4}{3}\) and \(x_{1,2,3}^{\prime}=\) \(x_{4,5,6}^{\prime}=\frac{2}{3}\). Thus, we obtain-as before- \(\operatorname{ExpRew}\left(x_{0} \models \diamond B\right)=x_{0}=\frac{4}{3}\) for the average number of rounds.

\section*{Remark 10.74. Alternative Definitions of the Expected Reward}

According to Definition 10.71 (page 818), \(\operatorname{Exp} \operatorname{Rew}(s \models \diamond B)=\infty\) if \(B\) is not almost surely reached from \(s\). This choice is consistent with the fact that \(\operatorname{rew}(\pi, \Delta B)=\infty\) whenever \(\pi \not \models \diamond B\) for some path \(\pi\). Let us discuss two variants of this definition.

For certain applications, it might be more reasonable to define the cumulative reward for an infinite path fragment which never visits \(B\) as zero (rather than \(\infty\) ). In this case, the expected reward ExpRew \((s \models \diamond B)\) is defined as
\[
\sum_{r=0}^{\infty} r \cdot \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \diamond B \wedge \operatorname{rew}(\pi, \diamond B)=r\}
\]
for all cases. (That is, no exceptional case for \(\operatorname{Pr}(s \models \diamond B)<1\).) The proposed algorithm for computing \(\operatorname{ExpRew}(s \models \diamond B)\) can easily be adapted to deal with this adapted definition by adding the constraint \(x_{s}=0\) for \(\operatorname{Pr}(s \models \diamond B)=0\) to the linear equation system. The equation \(x_{t}=0\) for \(t \in B\) remains unchanged. For all other states \(s\), i.e., all states \(s\) where \(s \notin B\) and \(\operatorname{Pr}(s \models \diamond B)>0\), we deal with the equation:
\[
x_{s}=\operatorname{rew}(s)+\sum_{u \in \operatorname{Post}(s)} \mathbf{P}(s, u) \cdot x_{u}
\]

Another way of treating the paths that never reach a \(B\)-state is to consider conditional expectations. That is, we consider the expected reward to reach \(B\) under the condition
that \(B\) is eventually reached. The corresponding definition of this conditional expectation, denoted \(\operatorname{CExpRew}(s \models \diamond B)\), is as follows. If \(\operatorname{Pr}(s \models \diamond B)>0\), then
\[
\operatorname{CExpRew}(s \models \diamond B)=\sum_{r=0}^{\infty} r \cdot \frac{\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \diamond B \wedge \operatorname{rew}(\pi, \diamond B)=r\}}{\operatorname{Pr}(s \models \diamond B)}
\]

If \(\operatorname{Pr}(s \models \diamond B)=0\), i.e., \(B\) is not reachable from \(s\), then \(\operatorname{CxpRew}(s \models \diamond B)\) can be defined as undefined or some fixed value. The computation of the conditional expected rewards goes as follows. By means of a backward graph analysis starting from the \(B\)-states, the set
\[
\operatorname{Pre}^{*}(B)=\{s \in S \mid \operatorname{Pr}(s \models \diamond B)>0\} .
\]
is determined. This is a standard backward reachability analysis. Let \(\mathcal{M}^{\prime}\) be a new Markov chain with state space \(\operatorname{Pr}^{*}(B)\). The transition probabilities of \(\mathcal{M}^{\prime}\) are defined by
\[
\mathbf{P}^{\prime}\left(s, s^{\prime}\right)=\frac{\mathbf{P}\left(s, s^{\prime}\right)}{\mathbf{P}\left(s, \operatorname{Pre}^{*}(B)\right)} \text { if } s \in \operatorname{Pr}^{*}(B) \backslash B \text { and } s^{\prime} \in \operatorname{Pr}^{*}(B)
\]

For \(t \in B\) let \(\mathbf{P}^{\prime}(t, t)=1\) and \(\mathbf{P}^{\prime}\left(t, s^{\prime}\right)=0\) for all states \(s^{\prime} \in \operatorname{Pr} e^{*}(B) \backslash\{t\}\). Then \(\operatorname{Pr}^{\mathcal{M}^{\prime}}(s \models \diamond B)=1\) for all \(s\) in \(\mathcal{M}^{\prime}\) and
\[
\operatorname{CExpRew}^{\mathcal{M}}(s \models \diamond B)=\operatorname{ExpRew}^{\mathcal{M}^{\prime}}(s \models \Delta B)
\]

The conditional expected reward in \(\mathcal{M}\) is thus equal to the expected reward in \(\mathcal{M}^{\prime}\). Thus, the problem of computing the conditional expected rewards \(C \operatorname{Exp} \operatorname{Rew}(s \vDash \diamond B)\) is reducible to the problem of computing (unconditional) expected rewards.

Other important quantitative measures for MRMs are cost-bounded reachability probabilities, i.e., the probability of reaching a certain given set of states within a certain bound on the cumulative reward. Paths that reach one of the target states whose cumulative reward exceeds the bound are considered too costly. Let \(B \subseteq S, r \in \mathbb{N}\) and \(\diamond_{\leqslant r} B\) denote the event of reaching the set \(B\) with cumulative reward at most \(r\). Then:
\[
\operatorname{Pr}\left(s \models \diamond_{\leqslant r} B\right)=\operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \diamond B \wedge \operatorname{rew}(\pi, \diamond B) \leqslant r\}
\]
denotes the probability of reaching \(B\) from \(s\) while the cumulative reward earned until \(B\) is reached is at most \(r\). For example, when \(B\) stands for the set of good states, \(\operatorname{Pr}\left(s \models \diamond_{\leqslant r} B\right)\) denotes the probability of reaching a good state with cost at most \(r\). We have
\[
\operatorname{Pr}\left(s \models \diamond_{\leqslant r} B\right)=\sum_{\widehat{\pi}} \mathbf{P}(\widehat{\pi})
\]
where \(\widehat{\pi}\) ranges over all finite paths \(s_{0} \ldots s_{n}\) with \(s_{n} \in B, s_{0}=s\) and \(s_{0}, \ldots, s_{n-1} \notin B\) such that \(\operatorname{rew}(\widehat{\pi}) \leqslant r\).

For finite MRMs, the values \(x_{s, r}=\operatorname{Pr}\left(s \mid=\diamond_{\leqslant r} B\right)\) can be computed via the following equation system:
- if \(s \in B\), then \(x_{s, r}=1\);
- if \(s \notin \operatorname{Pre}^{*}(B)\) or \(s \in \operatorname{Pre}^{*}(B) \backslash B \wedge(\operatorname{rew}(s)>r)\); then \(x_{s, r}=0\)
- in all other cases, i.e., if \(s \in \operatorname{Pre}^{*}(B) \backslash B\) and \(\operatorname{rew}(s) \leqslant r\) :
\[
x_{s, r}=\sum_{u \in S} \mathbf{P}(s, u) \cdot x_{u, r-\operatorname{rew}(s)} .
\]

The above equation system can be regarded as a linear equation system with variables \(x_{s, \rho}\) where \((s, \rho) \in S \times\{0,1, \ldots, r\}\). If all rewards are positive, i.e., \(\operatorname{rew}(s)>0\) for any state \(s\), then \(x_{s, r}\) is completely determined by the values \(x_{u, \rho}\) for \(u \in S\) and \(\rho<r\). Hence, in this case the above equation system can be solved by successively computing the vectors \(\left(x_{s, \rho}\right)_{s \in S}\) for \(\rho=0,1, \ldots, r\). In the presence of zero-reward states, the above equation for \(x_{s, r}\) may contain some variables \(x_{u, r}\) for the same reward bound \(r\). However, \(x_{s, r}\) only depends on the values \(x_{u, \rho}\) for \(\rho \leqslant r\). Thus, the vectors \(\mathbf{x}_{0}, \mathbf{x}_{1}, \mathbf{x}_{2}, \ldots, \mathbf{x}_{r}\) can be determined in this order, where the vector \(\mathbf{x}_{\rho}=\left(x_{s, \rho}\right)_{s \in S}\) is obtained from \(\mathbf{x}_{i}\) for \(0 \leqslant i<\rho\) as follows. If \(s \in B\) or \(s \notin \operatorname{Pre}^{*}(B)\), the value of \(x_{s, \rho}\) is 1 or 0 (see the first two items above). For \(s \in \operatorname{Pre}^{*}(B) \backslash B\) and rew \((s)>0\), the value of \(x_{s, \rho}\) is obtained from the previously computed values \(x_{u, \rho-\operatorname{rew}(s)}\) (see the sum in the third item). For the remaining states, i.e., states in \(S_{0}=\left\{s \in \operatorname{Pre}^{*}(B) \backslash B \mid \operatorname{rew}(s)=0\right\}\), the values \(x_{s, \rho}\) are obtained as the unique solution of the following linear equation system:
\[
\mathbf{x}=\mathbf{A}_{0} \cdot \mathbf{x}+\mathbf{b}
\]

Vector \(\mathbf{x}\) stands for \(\left(x_{s, \rho}\right)_{s \in S_{0}}\). The matrix \(\mathbf{A}_{0}\) contains the transition probabilities between the states in \(S_{0}\), i.e., \(\mathbf{A}_{0}=(\mathbf{P}(s, u))_{s, u \in S_{0}}\). The entries of the vector \(\mathbf{b}=\left(b_{s, \rho}\right)_{s \in S_{0}}\) are given by
\[
\begin{aligned}
& b_{s, \rho}=\quad \sum_{u \in B} \mathbf{P}(s, u) \cdot \overbrace{x_{u, \rho}}^{=1}+\sum_{u \in S \backslash \operatorname{Pre}^{*}(B)} \mathbf{P}(s, u) \cdot \overbrace{x_{u, \rho}}^{=0} \\
& +\sum_{\substack{u \in \operatorname{Pre}^{*}(B) \backslash B \\
\operatorname{rew}(u)>\rho}} \mathbf{P}(s, u) \cdot \overbrace{x, \rho}^{=0}+\sum_{\substack{u \in \operatorname{Pre}^{*}(B) \backslash B \\
\rho \geqslant \operatorname{rew}(u)>0}} \mathbf{P}(s, u) \cdot x_{u, \rho} \\
& =\mathbf{P}(s, B)+\sum_{\substack{u \in \operatorname{Pre}(B) \backslash B \\
\rho \geqslant \operatorname{rew}(u)>0}} \mathbf{P}(s, u) \cdot \underbrace{x_{u, \rho}}_{\begin{array}{c}
\text { already } \\
\text { computed }
\end{array}}
\end{aligned}
\]

The above equation system is linear and has a unique solution. Recall that \(\mathbf{P}(s, B)\) stands for the probability of moving from state \(s\) within one transition to a state \(t \in B\), i.e., \(\mathbf{P}(s, B)=\sum_{t \in B} \mathbf{P}(s, t)\).

Summarizing, the probabilities \(\operatorname{Pr}\left(s \models \diamond_{\leqslant r} B\right)\) can be computed with a time complexity which is polynomial in the size of \(\mathcal{M}\) and linear in the reward bound \(r\). To compute \(\operatorname{Pr}\left(s_{0} \models \diamond_{\leqslant r} B\right)\) for some designated state \(s_{0}\), in fact not all variables \(x_{s, \rho}\) might be relevant. The relevant variables are obtained by starting with the equation for \(x_{s_{0}, r}\) and then adding only those equations \(x_{s, \rho}\) that appeared in an already generated equation. This observation has no influence on the worst-case time complexity, but may yield drastic speedups.

\section*{Example 10.75. Simulation of a Die by a Coin (Revisited)}

Consider again the MRM for the simulation of a die by a fair coin where the reward function serves to count the number of rounds; see Examples 10.72 (page 819) and 10.73 (page 821). Consider the probability of obtaining a die outcome within the first two rounds. This amounts to computing the values \(\left.x_{s, \rho}=\operatorname{Pr}(s \models\rangle_{\leqslant \rho} B\right)\) for \(\rho=0,1,2\) and all states \(s\), where \(B=\{1,2,3,4,5,6\}\). We have
\[
\begin{aligned}
& x_{s_{0}, 0}=x_{s_{1,2,3}, 0}=x_{s_{4,5,6}, 0}=0 \\
& x_{s_{1,2,3}^{\prime}, 0}=x_{s_{4,5,6}^{\prime}, 0}=\frac{1}{2} \\
& x_{s_{2,3}, 0}=x_{s_{4,5}, 0}=x_{i}=1, \quad i=1, \ldots, 6
\end{aligned}
\]

The probability of reaching a \(B\)-state from the initial state in the first round is given by \(x_{s_{0}, 1}\) where
\[
\begin{aligned}
& x_{s_{0,1}}=\frac{1}{2} \cdot x_{s_{1,2,3}, 1}+\frac{1}{2} \cdot x_{s_{4,5,6,1}} \\
& x_{s_{1,2,3}, 1}=\frac{1}{2} \cdot \underbrace{x_{s_{2,3}, 0}}_{=1}+\frac{1}{2} \cdot \underbrace{x_{s_{1,2,3}^{\prime}, 0}}_{=\frac{1}{2}}=\frac{3}{4} \\
& x_{s_{4,5,6}, 1}=\frac{1}{2} \cdot \underbrace{x_{s_{4,5}, 0}}_{=1}+\frac{1}{2} \cdot \underbrace{x_{s_{4,5,6}^{\prime}, 0}^{\prime}}_{=\frac{1}{2}}=\frac{3}{4}
\end{aligned}
\]

This yields \(x_{s_{0}, 1}=\frac{3}{4}\). For the other states, we have \(x_{s_{2,3}, 1}=x_{s_{4,5}, 1}=x_{i}=1\) for \(1 \leqslant i \leqslant 6\) and
\[
x_{s_{1,2,3}^{\prime}, 1}=\frac{1}{2} \cdot x_{s_{1,2,3}, 1}+\frac{1}{2} \cdot x_{s_{2,3}, 1}=\frac{1}{2} \cdot \frac{3}{4}+\frac{1}{2} \cdot 1=\frac{7}{8},
\]
and similarly \(x_{s_{4,5,6}^{\prime}, 1}=\frac{7}{8}\). The probability \(x_{s_{0}, 2}\) of reaching \(B\) from the initial state within two rounds is given by
\[
\begin{aligned}
& x_{s_{0}, 2}=\frac{1}{2} \cdot x_{s_{1,2,3}, 2}+\frac{1}{2} \cdot x_{s_{4,5,6,2}} \\
& x_{s_{1,2,3}, 2}=\frac{1}{2} \cdot \underbrace{x_{s_{2,3,1}}}_{=1}+\frac{1}{2} \cdot \underbrace{x_{s_{1,2,3}, 1}}_{=\frac{7}{8}}=\frac{15}{16} \\
& x_{s_{4,5,6}, 1}=\frac{1}{2} \cdot \underbrace{x_{s_{4,5,1}}}_{=1}+\frac{1}{2} \cdot \underbrace{x_{s_{4,5,6}^{\prime}, 1}}_{=\frac{7}{8}}=\frac{15}{16} .
\end{aligned}
\]

Thus, \(x_{s_{0}, 2}=\frac{15}{16}\). In fact, we have \(x_{s_{0}, r}=1-\frac{1}{4^{r}}\) for all integers \(r \geqslant 0\); see Exercise 10.19 (page 904).

The techniques discussed so far for cost-bounded reachability probabilities can be adapted to cost-bounded constrained reachability probabilities in a fairly straightforward manner. For \(C, B \subseteq S\), let
\[
\operatorname{Pr}\left(s \models C \mathrm{U}_{\leqslant r} B\right)=\operatorname{Pr}_{s}\left\{\pi \in \operatorname{Paths}(s) \mid \pi \models C \mathrm{U}_{\leqslant r} B\right\}
\]
where \(s_{0} s_{1} s_{2} \ldots \vDash C \mathrm{U}_{\leqslant r} B\) if and only if there exists a finite prefix \(s_{0} \ldots s_{n}\) such that \(s_{i} \in C\) for \(0 \leqslant i<n, s_{n} \in B\) and \(r e w\left(s_{0} s_{1} \ldots s_{n}\right) \leqslant r\). In fact, the values \(\operatorname{Pr}\left(s \models C \mathrm{U}_{\leqslant r} B\right)\) are obtained by an analogous linear equation system as for \(\left.\operatorname{Pr}(s \models\rangle_{\leqslant r} B\right)\). The equations are
- if \(s \in B\), then \(x_{s, r}=1\);
- if \(s \not \vDash \exists(C \cup B)\) or \(s \notin B \wedge \operatorname{rew}(s)>r\), then \(x_{s, r}=0\);
- in all other cases, i.e., if \(s \in \exists(C \cup B)\) and \(r e w(s) \leqslant r\) :
\[
x_{s, r}=\sum_{u \in S} \mathbf{P}(s, u) \cdot x_{u, r-\operatorname{rew}(s)} .
\]

The techniques discussed so far are key ingredients for the model-checking of a variant of PCTL that incorporates rewards. This logic is called PRCTL, which is short for Probabilistic Reward CTL. It is defined as follows.

\section*{Definition 10.76. Syntax of PRCTL}

PRCTL state formulae over the set AP of atomic propositions are formed according to the following grammar:
\[
\Phi::=\text { true }|a| \Phi_{1} \wedge \Phi_{2}|\quad \neg \Phi| \mathbb{P}_{J}(\varphi) \mid \mathbb{E}_{R}(\Phi)
\]
where \(a \in A P, \varphi\) is a path formula, \(J \subseteq[0,1]\) and \(R\) are intervals with rational bounds. PCTL path formulae are formed according to the following grammar:
\[
\varphi:: \left.=\bigcirc \Phi\left|\Phi_{1} \cup \Phi_{2}\right| \underbrace{\Phi_{1} \mathrm{U} \leqslant n \Phi_{2}}_{\begin{array}{c}
\text { step-bounded } \\
\text { until }
\end{array}} \right\rvert\, \underbrace{\Phi_{1} \mathrm{U}_{\leqslant r} \Phi_{2}}_{\begin{array}{c}
\text { reward-bounded } \\
\text { until }
\end{array}}
\]
where \(\Phi, \Phi_{1}\), and \(\Phi_{2}\) are state formulae and \(n, r \in \mathbb{N}\).

The semantics of the propositional logic fragment and probability operator is defined as for PCTL. For the expectation operator \(\mathbb{E}_{R}(\cdot)\), the semantics is defined by:
\[
s \models \mathbb{E}_{R}(\Phi) \quad \text { iff } \quad \operatorname{ExpRew}(s \models \diamond \operatorname{Sat}(\Phi)) \in R .
\]

\section*{Example 10.77.}

The logic PRCTL may be used to specify the simulation of a die by a coin by the requirements
\[
\mathbb{E}_{\leqslant \frac{3}{2}}(\text { outcome }) \wedge \bigwedge_{1 \leqslant i \leqslant 6} \mathbb{P}_{=\frac{1}{6}}(\diamond i) \wedge \mathbb{P}_{\geqslant \frac{15}{16}}\left(\diamond_{\leqslant 2} \text { outcome }\right)
\]
where outcome is an atomic proposition that labels the six outcome states \(1,2,3,4,5\), and 6. The above PRCTL formula asserts that (1) the average number of rounds to obtain an outcome is bounded by \(\frac{3}{2},(2)\) the correctness of the outcomes in the sense that each of the six possible outcomes has probability \(\frac{1}{6}\), and (3) with probability at least \(\frac{15}{16}\) the outcome is obtained within at most two rounds.

\section*{Example 10.78. Zeroconf Protocol}

Consider again the zeroconf protocol as introduced in Example 10.5 (page 751). Consider the reward function that represents waiting times. The property "the probability of ending up with an unused address within \(n\) steps exceeds \(p^{\prime \prime \prime}\) can be expressed as PRCTL formula
\[
\mathbb{P}_{>p^{\prime}}\left(\diamond_{\leqslant n} o k\right)
\]
where ok uniquely indicates the state in which an unused address has been selected.
Consider now the reward function that keeps track of the number of probes sent. The property "the probability of ending up with an unused address after at most \(n\) probes exceeds \(p^{\prime \prime \prime}\) is expressed by the PRCTL formula
\[
\mathbb{P}_{>p^{\prime}}\left(\diamond_{\leqslant n} o k\right)
\]

Although the formula is identical to the previous one, its interpretation is somewhat different due to the different reward function.

\subsection*{10.5.2 Long-Run Properties}

This section is concerned with a different class of properties, viz. long-run averages. As opposed to the measures considered in the previous section, long-run averages are based
on the limiting behavior of Markov chains. Before considering long-run averages, let us first consider the long-run distribution of a Markov chain.

The long-run distribution is a limit defined on the basis of transient distributions. Recall that the transient state distribution \(\Theta_{n}=\Theta_{n}^{\mathcal{M}}\) is a function that assigns to each state \(t \in S\) the probability of being in state \(t\) after exactly \(n\) steps given the initial distribution \(\iota_{\text {init }}\); see Remark 10.22 (page 768). The limiting behavior of \(\mathcal{M}\) is obtained when \(n\) tends to go to infinity.

For a given Markov chain \(\mathcal{M}\) and states \(s, t\) in \(\mathcal{M}\), let \(\theta_{n}^{\mathcal{M}}(s, t)\) (briefly \(\theta_{n}(s, t)\) ) denote the probability of being in state \(t\) after exactly \(n\) steps when starting in state \(s\), i.e.:
\[
\theta_{n}(s, t)=\operatorname{Pr}_{s}\left\{s_{0} s_{1} s_{2} \ldots \in \operatorname{Paths}(s) \mid s_{0}=s \wedge s_{n}=t\right\}
\]

The transient state distribution \(\Theta_{n}\) arises by the values \(\theta_{n}(s, t)\) by taking the initial distribution into account:
\[
\Theta_{n}(t)=\sum_{s \in S} \iota_{\text {init }}(s) \cdot \theta_{n}(s, t)
\]

For fixed state \(s\), the function \(t \mapsto \theta_{n}(s, t)\) agrees with the transient state distribution \(\Theta_{n}^{\mathcal{M} s}\) in the Markov chain \(\mathcal{M}_{s}\) that is obtained from \(\mathcal{M}\) by declaring \(s\) to be the unique starting state.

\section*{Definition 10.79. Long-Run Distributions}

Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite Markov chain and \(s, t \in S\). The long-run average probability \(\theta^{\mathcal{M}}(s, t)\) (briefly \(\theta(s, t)\) ) is given by
\[
\theta(s, t)=\lim _{n \rightarrow \infty} \frac{1}{n} \cdot \sum_{i=1}^{n} \theta_{i}(s, t)
\]

The function \(t \mapsto \theta(s, t)\) is the long-run distribution for starting state \(s\). The long-run distribution for \(\mathcal{M}\) is given by
\[
\Theta^{\mathcal{M}}(t)=\sum_{s \in S} \iota_{\text {init }}(s) \cdot \theta(s, t)
\]

The limit in the above definition exists in every finite Markov chain. We will not provide a formal proof of this fact and refer to textbooks on Markov chains, e.g., [248]. Intuitively, the long-run probability \(\theta(s, t)\) is the fraction of (discrete) time to be in state \(t\) when starting in state \(s\). For example, in the following two-state Markov chain:


When starting in \(s\), then after any even number of steps the current state is \(s\), while after any odd number of steps the current state is \(t\). Hence:
\[
\theta_{2 k}(s, s)=\theta_{2 k+1}(s, t)=1 \text { for all } k \geqslant 0
\]
while \(\theta_{2 k}(s, t)=\theta_{2 k+1}(s, s)=0\). Therefore:
\[
\frac{1}{2 k} \sum_{i=1}^{2 k} \theta_{i}(s, t)=\frac{k}{2 k}, \quad \frac{1}{2 k+1} \sum_{i=1}^{2 k+1} \theta_{i}(s, t)=\frac{k+1}{2 k+1} .
\]

Hence, \(\lim _{n \rightarrow \infty} \theta_{n}(s, t)\) does not exist, but
\[
\theta(s, t)=\lim _{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^{n} \theta_{i}(s, t)=\frac{1}{2}
\]

This example illustrates that the sequence of the transient probabilities \(\theta_{n}(s, t)\) may not converge. Hence, the definition of long-run probabilities cannot be simplified by just taking the limit of the transient probabilities \(\theta_{n}(s, t)\).

However, if \(\lim _{n \rightarrow \infty} \theta_{n}(s, t)\) exists, then it agrees with \(\theta(s, t)\). This is illustrated by the Markov chain:


Whenever the current state is \(s\), then both successors \(s\) and \(t\) are equally likely. If, however, the current state is \(t\), then almost surely the next state is \(s\). This provides an intuitive explanation that the fraction of the time being in state \(s\) is \(\frac{2}{3}\), and for state \(t\) is \(\frac{1}{3}\). For this example, we have
\[
\theta_{n+1}(s, s)=\frac{1}{2} \theta_{n}(s, s)+\theta_{n}(s, t) \quad \text { and } \quad \theta_{n+1}(s, t)=\frac{1}{2} \theta_{n}(s, s)
\]

The following table shows the values for \(0 \leqslant n \leqslant 4\) :
\begin{tabular}{c||c|c}
\(n\) & \(\theta_{n}(s, s)\) & \(\theta_{n}(s, t)\) \\
\hline \hline 0 & 1 & 0 \\
1 & \(\frac{3}{4}\) & \(\frac{1}{4}\) \\
2 & \(\frac{5}{8}\) & \(\frac{3}{8}\) \\
3 & \(\frac{11}{16}\) & \(\frac{5}{16}\) \\
4 & \(\frac{21}{32}\) & \(\frac{11}{31}\) \\
\(\vdots\) & \(\vdots\) & \(\vdots\) \\
limit & \(\frac{2}{3}\) & \(\frac{1}{3}\)
\end{tabular}

In fact, \(\lim _{n \rightarrow \infty} \theta_{n}(s, s)=\frac{2}{3}=\theta(s, s)\) and \(\lim _{n \rightarrow \infty} \theta_{n}(s, t)=\frac{1}{3}=\theta(s, t)\).
Long-run probabilities provide the basis for several interesting measures. The following definition considers the expected long-run reward that is earned between two successive visits to a state in the set \(B\).

\section*{Definition 10.80. Expected Long-run Reward between \(B\)-States}

Let \(\mathcal{M}\) be a finite MRM with state space \(S, B \subseteq S\) and \(s \in S\) such that \(\operatorname{Pr}(s \models \square \diamond B)=1\). The expected long-run reward between two \(B\)-states for \(s\) is given by
\[
\operatorname{LongRunER}_{s}(B)=\sum_{t \in B} \theta(s, t) \cdot \operatorname{ExpRew}(t \models \bigcirc \diamond B)
\]
where \(\operatorname{Exp} \operatorname{Rew}(t \models \bigcirc \diamond B)\) is defined as \(\operatorname{Exp} \operatorname{Rew}(t \models \diamond B)\), except that only paths of length \(\geqslant 1\) are considered, i.e.,
\[
\operatorname{ExpRew}(t \models \bigcirc \diamond B)=\operatorname{rew}(t)+\sum_{u \in S} \mathbf{P}(t, u) \cdot \operatorname{ExpRew}(u \models \diamond B)
\]

The value \(\operatorname{Exp} \operatorname{Rew}(t \models \bigcirc \diamond B)\) can be interpreted as the average reward earned by moving from \(t\) to \(B\) within one or more steps. The intuitive meaning of \(\operatorname{LongRunER}{ }_{s}(B)\) is the average reward earned between two successive visits to a \(B\)-state, when considering the Markov chain \(\mathcal{M}\) on the long run. For example, if \(B\) characterizes certain failure states, and the reward function is defined to be one for all states outside \(B\), then \(\operatorname{LongRunER}_{s}(B)\) yields the average number of steps that are performed between two failures.

The computation of the expected long-run rewards \(\operatorname{LongRunER}(B)\) relies on techniques for computing the long-run probabilities. We will discuss now how the values \(\theta(s, t)\) can be obtained-once again - by solving linear equation systems.

In the sequel, let \(\mathcal{M}\) be a finite Markov chain with state space \(S\) and \(s \in S\). Let us first establish that the function \(t \mapsto \theta(s, t)\) is, in fact, a distribution over \(S\). Indeed, we have \(0 \leqslant \theta(s, t) \leqslant 1\), since \(0 \leqslant \theta_{i}(s, t) \leqslant 1\) for all \(i\) and therefore \(\sum_{1 \leqslant i \leqslant n} \theta_{i}(s, t) \leqslant n\). Moreover, for each \(i \geqslant 1\) :
\[
\sum_{t \in S} \theta_{i}(s, t)=1
\]

Therefore:
\[
\sum_{t \in S} \theta(s, t)=\frac{1}{n} \sum_{t \in S} \sum_{i=1}^{n} \theta_{i}(s, t)=\frac{1}{n} \sum_{i=1}^{n} \underbrace{\sum_{t \in S} \theta_{i}(s, t)}_{=1}=\frac{1}{n} \cdot n=1
\]

Furthermore, we have the following balance equation:
\[
\theta(s, u)=\sum_{t \in S} \theta(s, t) \cdot \mathbf{P}(t, u)
\]

This follows from
\[
\begin{aligned}
& \sum_{t \in S} \theta(s, t) \cdot \mathbf{P}(t, u) \\
= & \lim _{n \rightarrow \infty} \frac{1}{n} \sum_{t \in S} \sum_{i=1}^{n} \theta_{i}(s, t) \cdot \mathbf{P}(t, u) \\
= & \lim _{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^{n} \sum_{t \in S} \theta_{i}(s, t) \cdot \mathbf{P}(t, u) \\
= & \lim _{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^{n} \theta_{i+1}(s, u) \\
= & \lim _{n \rightarrow \infty} \underbrace{\frac{n+1}{n}}_{\text {tends to } 1} \cdot(\underbrace{\frac{1}{n+1} \cdot \sum_{j=1}^{n+1} \theta_{j}(s, u)}_{\text {tends to } \theta(s, u)}-\underbrace{\frac{\theta_{1}(s, u)}{n+1}}_{\text {tends to } 0})=\theta(s, u) .
\end{aligned}
\]

As almost surely a BSCC will be reached, \(\theta(s, t)=0\) if \(t\) does not belong to a BSCC that is reachable from \(s\). Note that then \(\theta_{i}(s, t)\) tends to zero if \(i\) tends to infinity. In particular, if \(s \in T\) and \(T \in \operatorname{BSCC}(\mathcal{M})\), then:
\[
\sum_{t \in T} \theta(s, t)=1
\]

This and the balance equation (applied to the sub-Markov chain consisting of the states in \(B\) ) yields that the values \(x_{s, t}=\theta(s, t)\) for \(s, t \in T\) and \(T \in B S C C(\mathcal{M})\) yield a solution
of the equation system:
\[
\sum_{t \in T} x_{s, t}=1 \quad \text { and } \quad x_{s, u}=\sum_{t \in T} x_{s, t} \cdot \mathbf{P}(t, u)
\]
where \(s\) and \(u\) range over all states of \(T\). In fact, the above equation system has a unique solution. In particular:
\[
\theta(s, t)=x_{s, t}=x_{s^{\prime}, t}=\theta\left(s^{\prime}, t\right)
\]
for all states \(s, s^{\prime}, t\) of a BSCC. Thus, the long-run distributions \(t \mapsto \theta(s, t)\) for the states \(s\) that are contained in some BSCC \(T\) can be solved by means of the linear equation system with variables \(x_{t}\) for \(t \in T\) and the equations
- \(\sum_{t \in T} x_{t}=1 ;\)
- \(\sum_{t \in T} x_{t} \cdot \mathbf{P}(t, u)=x_{u}\) for all states \(u \in T\).

This equation system has a unique solution \(\left(x_{t}\right)_{t \in T}\) and \(x_{t}=\theta(s, t)\) for all states \(s, t \in T\). Furthermore, \(\theta(s, u)=0\) for all states \(s \in T\) and \(u \notin T\). For all states \(s\) which do not belong to a BSCC, the long-run probabilities \(\theta(s, \cdot)\) are obtained as follows:
- \(\theta(s, t)=0\) if \(t\) is not contained in BSCC, and
- \(\theta(s, t)=\operatorname{Pr}(s \models \Delta T) \cdot x_{t}\) if \(t \in T\) and \(T \in B S C C(\mathcal{M})\), where \(x_{t}=\theta(s, t)\) for all/some \(s \in T\).

A corresponding operator to specify lower or upper bounds for the expected long-run rewards, say \(\mathbb{L}_{R}(\Phi)\) where \(R\) is a reward interval, may be added to the logic PRCTL with the semantics:
\[
s \models \mathbb{L}_{R}(\Phi) \text { if and only if } \operatorname{LongRunER}(S a t(\Phi)) \in R .
\]

The above-sketched technique can then be used to compute the satisfaction set of \(\mathbb{L}_{R}(\Phi)\).

\subsection*{10.6 Markov Decision Processes}

Nondeterminism is absent in Markov chains. Markov decision processes (MDPs, for short) can be viewed as a variant of Markov chains that permits both probabilistic and nondeterministic choices. As in Markov chains, the probabilistic choices may serve to model and
quantify the possible outcomes of randomized actions such as tossing a coin or sending a message over a lossy communication channel. Probabilistic choices may also be adequate for modeling the interface of a system with its environment. For example, for a vending machine it might be reasonable to assign probability \(\frac{9}{10}\) for the option chocolate bar, and probability \(\frac{1}{10}\) for the option apple. This, however, requires statistical experiments to obtain adequate distributions that model the average behavior of the environment, i.e., the user of the vending machine. In cases where this information is not available, or where it is needed to guarantee system properties which should hold for all potential environments, a natural choice is to model the interface with the environment by nondeterminism.

Another important motivation for the incorporation of nondeterminism in probabilistic models such as Markov chains is provided by the field of randomized distributed algorithms. Such algorithms are concurrent and hence nondeterministic in nature. This is due to the interleaving of the behavior of the distributed processes involved, i.e., the nondeterministic choice to determine which of the concurrent processes performs the next step (see Chapter 2). Besides, they are probabilistic in the sense that typically a rather restricted set of actions (like tossing a coin or selecting a number from a certain range) has a random nature. A simple example is a two-process mutual exclusion protocol where access to the critical section is governed by an arbiter that on the basis of tossing a coin decides which process acquires access.

Finally, nondeterminism is crucial for abstraction techniques of Markov chains. Abstraction is typically based on the grouping of states. If this abstraction is based on probabilistic bisimulation, then there is no need for nondeterminism, as the transition probabilities between groups (i.e., equivalence classes) of states is uniquely determined. If, however, a coarser abstraction is considered where e.g., states are grouped on the basis of their atomic propositions, one obtains a range of probabilities for the transition probabilitiesnondeterminism, so to speak. In the case of data abstraction, e.g., one might replace probabilistic branching by a nondeterministic choice.

Several operational models have been studied in the literature that all can be viewed as variants of Markov chains with both nondeterministic choices and discrete probabilities for the transition relation. We adopt the traditional notion of Markov decision processes (see e.g., the textbook [346]), extended by atomic propositions:

\section*{Definition 10.81. Markov Decision Process (MDP)}

A Markov decision process is a tuple \(\mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) where
- \(S\) is a countable set of states,
- Act is a set of actions,
- \(\mathbf{P}: S \times \operatorname{Act} \times S \rightarrow[0,1]\) is the transition probability function such that for all states \(s \in S\) and actions \(\alpha \in\) Act:
\[
\sum_{s^{\prime} \in S} \mathbf{P}\left(s, \alpha, s^{\prime}\right) \in\{0,1\}
\]
- \(\iota_{\text {init }}: S \rightarrow[0,1]\) is the initial distribution such that \(\sum_{s \in S} \iota_{\text {init }}(s)=1\),
- \(A P\) is a set of atomic propositions and \(L: S \rightarrow 2^{A P}\) a labeling function.

An action \(\alpha\) is enabled in state \(s\) if and only if \(\sum_{s^{\prime} \in S} \mathbf{P}\left(s, \alpha, s^{\prime}\right)=1\). Let \(\operatorname{Act}(s)\) denote the set of enabled actions in \(s\). For any state \(s \in S\), it is required that \(\operatorname{Act}(s) \neq \varnothing\). Each state \(s^{\prime}\) for which \(\mathbf{P}\left(s, \alpha, s^{\prime}\right)>0\) is called an \(\alpha\)-successor of \(s\).

The transition probabilities \(\mathbf{P}(s, \alpha, t)\) can be arbitrary real numbers in \([0,1]\) (that sum up to either zero or 1 for fixed \(s\) and \(\alpha\) ). For algorithmic purposes, the transition probabilities are supposed to be rational. Furthermore, the verification algorithms are restricted to finite MDPs. An MDP is finite whenever the state space \(S\), the set Act of actions, and the set \(A P\) of atomic propositions are finite.

An MDP has a unique initial distribution \(\iota_{\text {init }}\). In fact, this could be generalized by allowing a set of initial distributions. A computation then starts by choosing one of these initial distributions nondeterministically. For the sake of simplicity, we just consider a single initial distribution.

The intuitive operational behavior of an MDP \(\mathcal{M}\) is as follows. A stochastic experiment according to the initial distribution \(\iota_{\text {init }}\), yields a starting state \(s_{0}\) such that \(\iota_{\text {init }}\left(s_{0}\right)>0\). On entering state \(s\), say, first a nondeterministic choice between the enabled actions needs to be resolved. That is to say, it needs to be determined which action in \(\operatorname{Act}(s)\) is to be performed next. In the absence of any information about the frequency of actions-given actions \(\alpha\) and \(\beta\) in \(\operatorname{Act}(s)\) it is unknown how often \(\alpha\) needs to be selected-this choice is purely nondeterministic. Suppose action \(\alpha \in \operatorname{Act}(s)\) has been selected. On performing \(\alpha\) in state \(s\), one of the \(\alpha\)-successors of \(s\) is selected randomly according to the distribution \(\mathbf{P}(s, \alpha, \cdot)\). That is, with probability \(\mathbf{P}(s, \alpha, t)\) the next state is \(t\). If \(t\) is the unique \(\alpha\) successor of \(s\), then almost surely \(t\) is the successor after selecting \(\alpha\), i.e., \(\mathbf{P}(s, \alpha, t)=1\). In this case, \(\mathbf{P}(s, \alpha, u)=0\) for all states \(u \neq t\).

Any Markov chain is an MDP in which for any state \(s, \operatorname{Act}(s)\) is just a singleton set. Vice versa, any MDP with this property is a Markov chain. The action names are then irrelevant and can be omitted. Markov chains are thus a proper subset of MDPs.

The direct successors and predecessors of a state are defined as follows. For \(s \in S, \alpha \in\) Act and \(T \subseteq S\), let \(\mathbf{P}(s, \alpha, T)\) denote the probability of moving to a state in \(T\) via \(\alpha\), i.e.,
\[
\mathbf{P}(s, \alpha, T)=\sum_{t \in T} \mathbf{P}(s, \alpha, t)
\]
\(\operatorname{Post}(s, \alpha)\) denotes the set of \(\alpha\)-successors of \(s\), i.e.,
\[
\operatorname{Post}(s, \alpha)=\{t \in S \mid \mathbf{P}(s, \alpha, t)>0\} .
\]

Note that \(\operatorname{Post}(s, \alpha)=\varnothing\) if and only if \(\alpha \notin \operatorname{Act}(s)\). \(\operatorname{Pre}(t)\) denotes the set of pairs \((s, \alpha)\) with \(s \in S\) and \(\alpha \in \operatorname{Act}(s)\) such that \(t \in \operatorname{Post}(s, \alpha)\), i.e.,
\[
\operatorname{Pre}(t)=\{(s, \alpha) \in S \times \operatorname{Act} \mid \mathbf{P}(s, \alpha, t)>0\}
\]

Example 10.82. An Example of an MDP
Consider the MDP \(\mathcal{M}\) depicted below:


State \(s\) is the only initial state, i.e., \(\iota_{\text {init }}(s)=1\) and \(\iota_{\text {init }}(t)=\iota_{\text {init }}(u)=0\). The sets of enabled actions are
- \(\operatorname{Act}(s)=\{\alpha, \beta\}\) with \(\mathbf{P}(s, \alpha, t)=1, \mathbf{P}(s, \beta, u)=\mathbf{P}(s, \beta, s)=\frac{1}{2}\), and
- \(\operatorname{Act}(t)=\operatorname{Act}(u)=\{\gamma\}\) with \(\mathbf{P}(t, \gamma, s)=\mathbf{P}(u, \gamma, s)=1\).

In state \(s\), a nondeterministic choice between actions \(\alpha\) and \(\beta\) exists. On selecting action \(\alpha\), the next state is \(t\); on selecting action \(\beta\), the successor states \(s\) and \(u\) are equally probable. Some successor and predecessor sets are \(\operatorname{Post}(s, \alpha)=\{t\}, \operatorname{Post}(s, \beta)=\{s, u\}\), and \(\operatorname{Pre}(s)=\{(s, \beta),(t, \gamma),(u, \gamma)\}\).

\section*{Example 10.83. A Randomized Mutual Exclusion Protocol}

Consider a simple randomized mutual exclusion protocol for two concurrent processes \(P_{1}\)
and \(P_{2}\). The coordination of accessing their critical sections is provided by a randomized arbiter which permits process \(P_{i}\) to enter its critical section if the other process is in its noncritical section. If both processes acquire access to the critical section, the arbiter tosses a fair coin to decide which of the two processes has to wait and which process may enter the critical section.


Figure 10.15: MDP for a randomized mutual exclusion protocol.

The composite behavior of the two concurrent processes and the arbiter can be modeled by the MDP depicted in Figure 10.15. All states of the MDP where at least one of the processes is in its noncritical location exhibit a nondeterministic choice between the enabled actions of \(P_{1}\) and \(P_{2}\). The corresponding req, enter or rel action does not have a proper probabilistic effect, since it yields a unique successor. The transition probabilities that equal 1 are omitted from the figure. Only in state \(\left\langle\right.\) wait \(_{1}\), wait \(\left.{ }_{2}\right\rangle\) is there a proper probabilistic choice performed by the arbiter to select the next process to enter the critical section.

Remark 10.84. Actions in an MDP
The action names \(\alpha \in\) Act are needed here only for technical reason, viz. to group all edges belonging to the same probabilistic choice. When using MDPs in a compositional framework, where the MDP for a complex system arises through the parallel composition of several other MDPs, then it is more appropriate to allow for sets of distributions per action and state rather than just a single distribution. Note that, e.g., the interleaving of two equally named actions performed by different processes yields a global state that has two outgoing transitions with the same action label. This more general approach can be formalized by replacing \(\mathbf{P}\) with a transition relation of the form
\[
\rightarrow \subseteq S \times \operatorname{Act} \times \operatorname{Distr}(S)
\]
where \(\operatorname{Distr}(S)\) denotes the set of distributions over \(S\), i.e., functions \(\mu: S \rightarrow[0,1]\) such that \(\sum_{s \in S} \mu(s)=1\). Labeled transition systems appear as special case of this type of transition relation. This follows from the fact that any transition \(s \xrightarrow{\alpha} t\) in a transition system can be identified with the transition \(s \xrightarrow{\alpha} \mu_{t}^{1}\) where \(\mu_{t}^{1} \in \operatorname{Distr}(S)\) denotes the distribution with \(\mu_{t}^{1}(t)=1\) and \(\mu_{t}^{1}(u)=0\) for any \(u \neq t\). The notion of MDP in this monograph (see Definition 10.81) is obtained by considering transitions \(s \xrightarrow{\alpha} \mu_{s, \alpha}\) for the actions \(\alpha \in \operatorname{Act}(s)\), where \(\mu_{s, \alpha}\) denotes the distribution \(t \mapsto \mathbf{P}(s, \alpha, t)\). Since compositional approaches for MDPs fall outside the scope of this monograph, the action names are irrelevant. The actions may be assumed to be renamed such that per state and action there is at most one distribution.

\section*{Remark 10.85. Probmela}

In a similar way as transition systems for nonprobabilistic (concurrent) programs can be described by higher-level modeling languages such as (nano)Promela, (see Section 2.2.5 on page 63), probabilistic systems can be specified by various higher-level description techniques. We consider the main features of a probabilistic variant of nanoPromela, called Probmela. As in nanoPromela, a Probmela model consists of finitely many concurrent processes that may communicate via either shared variables or channels (or both). The processes are described by statements of a probabilistic guarded command language. The core of this language is as in nanoPromela and consists of assignments, conditional commands (i.e., statements built by the keywords if-if), loops (i.e., statements built by do-od), communication actions \(c\) ? \(x\) (for input) and \(c\) !expr (for output), and atomic regions. The language features three probabilistic features: randomized assignments, probabilistic choice, and lossy channels. Let us discuss these features in somewhat more detail.
- A random assignment has the form \(x:=\operatorname{random}(V)\) where \(x\) is a variable and \(V\) is a finite nonempty subset of \(\operatorname{dom}(x)\). On executing this assignment, a value \(v \in V\) is chosen probabilistically according to a uniform distribution over \(V\) and assigned to \(x\). The probability that value \(v \in V\) is assigned to \(x\) is thus \(\frac{1}{|V|}\).
- The probabilistic choice operator is a probabilistic variant of if-fi-statements where the resolution of the choices is by probabilities rather than by guards. The syntax is
\[
\text { pif }\left[p_{1}\right] \Rightarrow s t m t_{1} \ldots\left[p_{n}\right] \Rightarrow s t m t_{n} \text { fip }
\]
where \(p_{i}\) is a non-negative real number such that \(\sum_{i} p_{i}=1\). The above statement models the probabilistic choice between the statements stmt \(t_{1}\) through stmt \({ }_{n}\), where stmt \({ }_{i}\) is selected with probability \(p_{i}\).
- Communication channels may either be perfect (as in nanoPromela) or lossy. This is indicated upon declaring a channel. A perfect channel never loses a message,
\[
\begin{aligned}
& c:=s_{0} ; \\
& \text { do } \quad:: c=s_{0} \quad \Rightarrow \text { pif }\left[\frac{1}{2}\right] \Rightarrow c:=s_{1,2,3} \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=s_{4,5,6} \quad \text { fip } \\
& :: c=s_{1,2,3} \quad \Rightarrow \text { pif } \quad\left[\frac{1}{2}\right] \Rightarrow c:=s_{1,2,3}^{\prime} \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=s_{2,3} \quad \text { fip } \\
& :: c=s_{2,3} \quad \Rightarrow \text { pif }\left[\frac{1}{2}\right] \Rightarrow c:=2 \quad\left[\frac{1}{2}\right] \Rightarrow c:=3 \quad \text { fip } \\
& :: c=s_{1,2,3}^{\prime} \quad \Rightarrow \quad \text { pif } \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=s_{1,2,3} \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=1 \quad \text { fip } \\
& :: c=s_{4,5,6} \quad \Rightarrow \quad \text { pif } \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=s_{4,5,6}^{\prime} \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=s_{4,5} \quad \text { fip } \\
& :: c=s_{4,5} \quad \Rightarrow \quad \text { pif } \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=4 \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=5 \quad \text { fip } \\
& :: c=s_{4,5,6}^{\prime} \quad \Rightarrow \quad \text { pif } \quad\left[\frac{1}{2}\right] \Rightarrow c:=s_{4,5,6} \quad\left[\frac{1}{2}\right] \Rightarrow \quad c:=6 \quad \text { fip } \\
& \text { od }
\end{aligned}
\]

Figure 10.16: Probmela specification of Knuth and Yao's simulation of a die by a fair coin.

\begin{abstract}
whereas via a lossy channel a message is lost with probability \(p\), a fixed probability that is defined on declaring the channel. The effect of the communication action \(c!v\) along lossy channel \(c\) is that \(v\) is written into the buffer for \(c\) with probability \(1-p\), and it fails (i.e., the content of the buffer for \(c\) remains unchanged while performing the action \(c!v)\) with probability \(p\).
\end{abstract}

The stepwise behavior of a Probmela program can be formalized by means of an MDP. The states are tuples \(\left\langle\ell_{1}, \ldots, \ell_{n}, \eta, \xi\right\rangle\) where \(\ell_{i}\) is the location of process \(i, \eta\) is a variable evaluation, and \(\xi\) is a channel evaluation. This is the same as for nanoPromela. The transitions in the MDP specify a probability distribution over the successor states (i.e., an enabled action of a state and its probabilistic effect). As Probmela is only used for the examples, we refrain from presenting these inference rules here.

To illustrate the language Probmela, consider the simulation of a die by a coin; see Figure 10.2 on page 750. The Probmela code is shown in Figure 10.16. Here, \(c\) is a variable whose domain is the state space of the MC in Figure 10.2. The statements
\[
\operatorname{pif}\left[\frac{1}{2}\right] \Rightarrow c:=\ldots\left[\frac{1}{2}\right] \Rightarrow c:=\ldots \text { fip }
\]
represent the coin-tossing experiments that are performed in the inner nodes. The final value of \(c\) stands for the die outcome. Note that the Probmela model just consists of a single process which does not behave nondeterministically. The semantics is thus a Markov chain.

\section*{Example 10.86. Alternating Bit Protocol with Lossy Channel}

The alternating bit protocol (see Example 2.32 on page 57 ) can be modeled by an MDP,
assuming that the probability \(p\) for a channel to lose a message is known. An MDP of the alternating bit protocol can be obtained by specifying the behavior of the sender, the receiver, and the timer by Probmela statements. As in Example 2.32, the asynchronous channels \(c\) and \(d\) are used for connecting the sender and the receiver. The synchronous channel \(e\) is used for activating the timer and the timeout signal. Channel \(c\) is declared to be lossy with failure probability \(p\), while channel \(d\) is supposed to be perfect. The sender is described by the following Probmela fragment:
\[
\begin{array}{llll}
x:=0 & & & \\
\text { do }:: \text { true } \Rightarrow & c!x ; e!o n ; & \\
& \text { if }:: & d!y \Rightarrow & \Rightarrow:=\neg x ; \text { e!timeroff } \\
& & :: & e ? y \Rightarrow c!x ; e!o n \\
& & \text { fi } & \\
\text { od } & & &
\end{array}
\]

The behavior of the receiver and the timer can be specified by analogous Probmela statements. As \(c\) is a lossy channel with probability \(p\), the statement \(c!x\) will not result in inserting message \(x\) into \(c\).

\section*{Example 10.87. Randomized Dining Philosophers}

Example 3.2 (page 90) treats the dining philosophers problem. We have seen that a naive symmetric protocol fails to ensure deadlock freedom as all philosophers might, e.g., pick up their left chopsticks simultaneously (or in any order). A deadlock-free solution is obtained when using variables that control the sticks and make them available to a single philosopher only. In order to break the symmetry, the initial values of these variables have to be chosen in such a way that initially at least one philosopher can pick up both sticks.

In the probabilistic setting, there are simpler solutions which are fully symmetric and that need neither global control nor additional shared variables. Consider, e.g., the proposal by Lehmann and Rabin [268]. The idea of their algorithm is that the philosophers toss a fair coin to decide which of the sticks they pick up first. If they fail to take the chosen stick, they then repeat the random choice. Otherwise they try to get the other stick. If the missing stick is not available, they return the taken stick and repeat the coin-tossing experiment. The Probmela code of philosopher i, shown in Figure 10.17 on page 840, uses Boolean variables \(s t i c k{ }_{i}\) for the sticks that indicate whether stick \(i\) is available, where initially stick \(_{i}=\) true for all sticks.

This randomized algorithm is deadlock-free as whenever a philosopher attempts to access an occupied stick, then he does not wait but repeats the coin-tossing experiment. Starvation freedom is guaranteed in the sense that almost surely any hungry philosopher who tries infinitely often to get the sticks will eventually eat.
```

mode $_{i}:=$ think;
do :: true $\Rightarrow$
mode $_{i}:=$ try;
pif $\left[\frac{1}{2}\right] \Rightarrow$
if $::$ stick $_{i} \quad \Rightarrow \quad$ stick $_{i}:=$ false;
if $:: s t i c k_{i+1} \Rightarrow \quad$ stick $_{i+1}:=$ false;
mode $_{i}:=$ eat;
stick $_{i}:=$ true;
stick $_{i+1}:=$ true;
mode $_{i}:=$ think;
$:: \neg$ stick $_{i+1} \Rightarrow$ stick $_{i}:=$ true;
fi
:: Stick $_{i} \quad \Rightarrow$ skip
fi
$\left[\frac{1}{2}\right] \Rightarrow$
if $::$ stick $_{i+1} \quad \Rightarrow$ stick $_{i+1}:=$ false;
if $\quad::$ stick $_{i} \Rightarrow \quad$ stick $_{i}:=$ false;
mode $_{i}:=$ eat;
stick $_{i}:=$ true;
stick $_{i+1}:=$ true;
mode $_{i}:=$ think;
:: $\neg$ stick $_{i} \Rightarrow \quad$ stick $_{i+1}:=$ true;
fi
:: $\neg$ stick $_{i+1} \Rightarrow$ skip
fi
fip
od

```

Figure 10.17: Probmela code for philosopher \(i\).

The semantics of the Probmela specification for \(n\) philosophers is an MDP. In each state, each of the philosophers has an enabled action. Most of the actions are nonprobabilistic in the sense that they yield a unique successor state. Only when the current location of some philosopher is the pif-fip-statement does its enabled action stand for the cointossing experiment to decide which stick to pick up first. This yields two equally likely successors.

\section*{Notation 10.88. Finiteness, Size, and Graph of an MDP}

Let \(\mathcal{M}\) be an MDP as in Definition 10.81 (page 833). \(\mathcal{M}\) is called finite if the state space \(S\), the action set \(A c t\), and the set \(A P\) of atomic propositions are finite. The size of \(\mathcal{M}\) is
defined as the number of edges in the underlying graph of \(\mathcal{M}\), i.e., the number of triples \((s, \alpha, t)\) such that \(\mathbf{P}(s, \alpha, t)>0\). Here, the underlying graph of \(\mathcal{M}\) denotes the directed graph \((S, E)\) where the states of \(\mathcal{M}\) act as vertices and \((s, t) \in E\) if and only if there exists an action \(\alpha\) such that \(\mathbf{P}(s, \alpha, t)>0\).

\section*{Notation 10.89. LTL/CTL-like Notations}

As earlier in this chapter, we will often use LTL- or CTL-like notations with states or sets of states as atomic propositions. The satisfaction relation then refers to the underlying graph of the MDP, e.g., for \(s \in S\) and \(B \subseteq S\), the statement \(s \models \exists \diamond B\) means that some state in \(B\) is reachable from \(s\) in the underlying graph of the MDP \(\mathcal{M}\).

The paths in an MDP \(\mathcal{M}\) describe the potential computations that arise by resolving both the nondeterministic and probabilistic choices in \(\mathcal{M}\). They are obtained by traversing the underlying graph via consecutive edges. Paths and path fragments in an MDP are defined as alternating sequences of states and actions. \({ }^{2}\) More precisely,

\section*{Definition 10.90. Path in an MDP}

An infinite path fragment in an \(\operatorname{MDP} \mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) is an infinite sequence \(s_{0} \alpha_{1} s_{1} \alpha_{2} s_{2} \alpha_{3} \ldots \in(S \times A c t)^{\omega}\), written as
\[
\pi=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} s_{2} \xrightarrow{\alpha_{3}} \ldots
\]
such that \(\mathbf{P}\left(s_{i}, \alpha_{i+1}, s_{i+1}\right)>0\) for all \(i \geqslant 0\). Any finite prefix of \(\pi\) that ends in a state is a finite path fragment. Paths (s) denotes the set of infinite path fragments that start in state \(s ; \operatorname{Paths}_{\text {fin }}(s)\) denotes the set of finite path fragments that start in \(s\). Let \(\operatorname{Paths}(\mathcal{M})=\) \(\bigcup_{s \in S} \operatorname{Paths}(s)\) and \(\operatorname{Paths}_{f i n}(\mathcal{M})=\bigcup_{s \in S} \operatorname{Paths}_{f i n}(s)\).

For Markov chains, the set of paths is equipped with a \(\sigma\)-algebra and a probability measure that reflects the intuitive notion of probabilities for (measurable sets of) paths. For MDPs, this is slightly different since no constraints are imposed on the resolution of the nondeterministic choices. Let us explain this phenomenon by means of a simple example. Suppose that \(\mathcal{M}\) is an MDP with a single initial state \(s_{0}\) with two enabled actions \(\alpha\) and \(\beta\) both representing a coin-tossing experiment. The coin used for action \(\alpha\) is supposed to be fair and yields heads and tails with equal probability, while action \(\beta\) represents flipping an unfair coin that provides the outcome heads with probability \(\frac{1}{6}\) and tails with probability \(\frac{5}{6}\). What is the probability for obtaining the outcome tails for the initial state \(s_{0}\) ? This question cannot be answered; in fact, it is senseless, as this probability depends on whether

\footnotetext{
\({ }^{2}\) As opposed to transition systems, no distinction is made between an execution and a path. The action names in the paths are used to keep track of the stepwise probabilities.
}
action \(\alpha\) or action \(\beta\) will be chosen. This, however, is not specified! It is guaranteed that heads will appear with probability at least \(\frac{1}{6}\) and at most \(\frac{1}{2}\).

Also under the assumption that after performing \(\alpha\) and \(\beta\) we return to state \(s_{0}\) and repeat the selection of one of the actions \(\alpha\) or \(\beta\) ad infinitum, the nondeterminism in \(s_{0}\) can be resolved arbitrarily. Any sequence of the actions \(\alpha\) and \(\beta\) constitutes a legal behavior of the MDP. Thus, it might even be the case that action \(\beta\) is never taken, in which case the probability of obtaining heads within the first \(n\) coin tosses is \(1-\left(\frac{1}{2}\right)^{n}\). For the other extreme, if \(\beta\) is chosen the first \(n\) times, then the probability of obtaining the outcome heads at least once in the first \(n\) rounds is \(1-\left(\frac{5}{6}\right)^{n}\). However, several other values are possible for the probability of the event "at least once outcome heads in the first \(n\) rounds". For instance, the value \(1-\frac{5^{k}}{6^{k} \cdot 2^{n-k}}\) is obtained if during the first \(k \leqslant n\) steps, action \(\beta\) is chosen and from the \((k+1)\) st step on action \(\alpha\) is selected.

This example illustrates that MDPs are not augmented with a unique probability measure. Instead, reasoning about probabilities of sets of paths of an MDP relies on the resolution of nondeterminism. This resolution is performed by a scheduler. A scheduler chooses in any state \(s\) one of the enabled actions \(\alpha \in \operatorname{Act}(s)\). (Recall that \(\operatorname{Act}(s)\) is nonempty for any state \(s\).) It does not impose any constraint on the probabilistic choice that is resolved once \(\alpha\) has been chosen.

\section*{Definition 10.91. Scheduler}

Let \(\mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be an MDP. A scheduler for \(\mathcal{M}\) is a function \(\mathfrak{S}: S^{+} \rightarrow\) Act such that \(\mathfrak{S}\left(s_{0} s_{1} \ldots s_{n}\right) \in \operatorname{Act}\left(s_{n}\right)\) for all \(s_{0} s_{1} \ldots s_{n} \in S^{+}\).

The path (fragment)
\[
\pi=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} s_{2} \xrightarrow{\alpha_{3}} \ldots
\]
is called a \(\mathfrak{S}\)-path (fragment) if \(\alpha_{i}=\mathfrak{S}\left(s_{0} \ldots s_{i-1}\right)\) for all \(i>0\).

In the literature, a scheduler is sometimes referred to as adversary, policy, or strategy. Note that for any scheduler, the actions are dropped from the history \(s_{0} s_{1} \ldots s_{n}\). This is not a restriction as for any sequence \(s_{0} s_{1} \ldots s_{n}\) the relevant actions \(\alpha_{i}\) are given by \(\alpha_{i+1}=\) \(\mathfrak{S}\left(s_{0} s_{1} \ldots s_{i}\right)\). Hence, the scheduled action sequence can be constructed from prefixes of the path at hand. Any path fragment \(s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} \ldots \xrightarrow{\alpha_{n}} s_{n}\) where \(\alpha_{i} \neq \mathfrak{S}\left(s_{0} s_{1} \ldots s_{i}\right)\) for some \(i\) does not describe a path fragment that can be obtained from \(\mathfrak{S}\).

As a scheduler resolves all nondeterministic choices in an MDP, it induces a Markov chain. That is to say, the behavior of an MDP \(\mathcal{M}\) under the decisions of scheduler \(\mathfrak{S}\) can be formalized by a Markov chain \(\mathcal{M}_{\mathfrak{S}}\). Intuitively, this Markov chain arises by unfolding \(\mathcal{M}\) into a tree, or forest if there are two or more states \(s\) with \(\iota_{\text {init }}(s)>0\). The paths in the

Markov chain represent the \(\mathfrak{S}\)-paths; the states in the Markov chain are state sequences of the Markov decision process \(\mathcal{M}\). Formally:

Definition 10.92. Markov Chain of an MDP Induced by a Scheduler
Let \(\mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be an MDP and \(\mathfrak{S}\) a scheduler on \(\mathcal{M}\). The Markov chain \(\mathcal{M}_{\mathfrak{S}}\) is given by
\[
\mathcal{M}_{\mathfrak{S}}=\left(S^{+}, \mathbf{P}_{\mathfrak{S}}, \iota_{\mathrm{init}}, A P, L^{\prime}\right)
\]
where for \(\sigma=s_{0} s_{1} \ldots s_{n}\) :
\[
\mathbf{P}_{\mathfrak{S}}\left(\sigma, \sigma s_{n+1}\right)=\mathbf{P}\left(s_{n}, \mathfrak{S}(\sigma), s_{n+1}\right)
\]
and \(L^{\prime}(\sigma)=L\left(s_{n}\right)\).

Note that \(\mathcal{M}_{\mathfrak{S}}\) is infinite, even if the MDP \(\mathcal{M}\) is finite. Intuitively, state \(s_{0} s_{1} \ldots s_{n}\) of \(\mathcal{M}_{\mathfrak{S}}\) represents the configuration where the MDP \(\mathcal{M}\) is in state \(s_{n}\) and \(s_{0} s_{1} \ldots s_{n-1}\) stands for the history, i.e., the path fragment that leads from the starting state \(s_{0}\) to the current state \(s_{n}\). Since \(\mathfrak{S}\) might select different actions for path fragments that end in the same state \(s\), a scheduler as in Definition 10.91 is also referred to as history-dependent.

\section*{Example 10.93. Markov Chain Induced by a Scheduler}

Consider the MDP of Example 10.82 (page 835), illustrated again below:


Let us consider some examples of schedulers for this MDP. Scheduler \(\mathfrak{S}_{\alpha}\) always selects action \(\alpha\) in state \(s\). It is defined by \(\mathfrak{S}_{\alpha}(\sigma)=\alpha\) if \(\operatorname{last}(\sigma)=s\), otherwise \(\mathfrak{S}_{\alpha}(\sigma)=\gamma\). In a similar way, \(\mathfrak{S}_{\beta}\) is defined by \(\mathfrak{S}_{\beta}(\sigma)=\beta\) if \(\operatorname{last}(\sigma)=s\), otherwise \(\mathfrak{S}_{\beta}(\sigma)=\gamma\). The only \(\mathfrak{S}_{\alpha}\)-path in \(\mathcal{M}\) is \(s \xrightarrow{\alpha} t \xrightarrow{\gamma} s \xrightarrow{\alpha} \ldots\) The path \(s \xrightarrow{\beta} s \xrightarrow{\beta} s \xrightarrow{\beta} u \xrightarrow{\gamma} s \xrightarrow{\beta} u \ldots\) is a \(\mathfrak{S}_{\beta}\)-path. Likewise \(s \xrightarrow{\beta} u \xrightarrow{\gamma} s \xrightarrow{\beta} u \ldots\)

Finally, let \(\mathfrak{S}\) be a scheduler that selects \(\alpha\) in \(s\) when just returning from \(u\), and \(\beta\) otherwise. Thus: \(\mathfrak{S}\left(s_{0} \ldots s_{n} s\right)=\alpha\) if \(s_{n}=u\) and \(\mathfrak{S}\left(s_{0} \ldots s_{n} s\right)=\beta\) otherwise. Let \(\mathfrak{S}(s)=\alpha\). Note that this scheduler decides on the basis of the one-but-last visited state. In the states \(u\) and \(t\), the only enabled action \(\gamma\) is chosen.

The Markov chain \(\mathcal{M}_{\mathfrak{S}_{\alpha}}\) is the infinite chain:


An initial fragment of the Markov chain \(\mathcal{M}_{\mathfrak{S}_{\beta}}\) is depicted in Figure 10.18 on page 844.


Figure 10.18: Initial fragment of the Markov chain \(\mathcal{M}_{\mathfrak{S}_{\beta}}\).

There is a one-to-one correspondence between the \(\mathfrak{S}\)-paths of the MDP \(\mathcal{M}\) and the paths in the Markov chain \(\mathcal{M}_{\mathfrak{S}}\). For a \(\mathfrak{S}\)-path
\[
\pi=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} s_{2} \xrightarrow{\alpha_{3}} \ldots
\]
the corresponding path in the Markov chain \(\mathcal{M}_{\mathfrak{S}}\) is given by
\[
\pi^{\mathfrak{S}}=\widehat{\pi}_{0} \widehat{\pi}_{1} \widehat{\pi}_{2} \ldots
\]
where \(\widehat{\pi}_{n}=s_{0} s_{1} \ldots s_{n}\). Vice versa, for a path \(\widehat{\pi}_{0} \widehat{\pi}_{1} \widehat{\pi}_{2} \ldots\) in the Markov chain \(\mathcal{M}_{\mathfrak{S}}\), \(\widehat{\pi}_{0}=s_{0}\) for some state \(s_{0}\) such that \(\iota_{\text {init }}\left(s_{0}\right)>0\) and, for each \(n>0, \widehat{\pi}_{n}=\widehat{\pi}_{n-1} s_{n}\) for some state \(s_{n}\) in the MDP \(\mathcal{M}\) such that \(\mathbf{P}\left(s_{n-1}, \mathfrak{S}\left(\widehat{\pi}_{n-1}\right), s_{n}\right)>0\). Hence:
\[
s_{0} \xrightarrow{\mathfrak{S}\left(\hat{\pi}_{0}\right)} s_{1} \xrightarrow{\mathfrak{S}\left(\hat{\pi}_{1}\right)} s_{2} \xrightarrow{\mathfrak{S}\left(\hat{\pi}_{2}\right)} \ldots
\]
is a \(\mathfrak{S}\)-path in \(\mathcal{M}\). In the sequel, we often identify the paths in \(\mathcal{M}_{\mathfrak{S}}\) with the corresponding \(\mathfrak{S}\)-paths in \(\mathcal{M}\).

When we consider an ordinary transition system as an MDP (where any action taken in a state yields a unique successor), the concept of a scheduler is just another formalization of a path. In the general setting, i.e., for MDPs with proper probabilistic actions, however, any scheduler induces a set of paths.

As \(\mathcal{M}_{\mathfrak{S}}\) is a Markov chain, one can now reason about the probabilities for measurable sets of \(\mathfrak{S}\)-paths. Let \(\operatorname{Pr}_{\mathfrak{S}}^{\mathcal{M}}\), or simply \(\operatorname{Pr}^{\mathfrak{S}}\), denote the probability measure \(\operatorname{Pr}^{\mathcal{M}}{ }^{\mathfrak{S}}\) associated with the Markov chain \(\mathcal{M}_{\mathfrak{\subseteq}}\). This measure is the basis for associating probabilities with events in the MDP \(\mathcal{M}\). Let, e.g., \(P \subseteq\left(2^{A P}\right)^{\omega}\) be an \(\omega\)-regular property. Then \(\operatorname{Pr}^{\mathscr{G}}(P)\) is defined (see Notation 10.49 , page 796) as the probability measure of the set of \(\mathfrak{S}\)-paths \(\pi\) in the Markov chain \(\mathcal{M}_{\mathfrak{S}}\) such that \(\operatorname{trace}(\pi) \in P\) :
\[
\operatorname{Pr}^{\mathfrak{S}}(P)=\operatorname{Pr}^{\mathcal{M}_{\mathfrak{G}}}(P)=\operatorname{Pr}_{\mathcal{M}_{\mathfrak{S}}}\left\{\pi \in \operatorname{Paths}\left(\mathcal{M}_{\mathfrak{S}}\right) \mid \operatorname{trace}(\pi) \in P\right\}
\]

Similarly, for fixed state \(s\) of \(\mathcal{M}\), which is considered as the unique starting state,
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models P)=\operatorname{Pr}_{s}^{\mathcal{M}}\{\{\pi \in \operatorname{Paths}(s) \mid \operatorname{trace}(\pi) \in P\}
\]
where we identify the paths in \(\mathcal{M}_{\mathfrak{S}}\) with the corresponding \(\mathfrak{S}\)-paths in \(\mathcal{M}\). This explains the notation above where the paths \(\pi \in \operatorname{Paths}(s)\) in \(\mathcal{M}\) are identified with the associated paths \(\pi^{\mathfrak{S}}\) in the Markov chain \(\mathcal{M}_{\mathfrak{S}}\).

The quantitative analysis of an MDP \(\mathcal{M}\) against \(\omega\)-regular specifications amounts to establishing the best lower and/or upper probability bounds that can be guaranteed, when ranging over all schedulers. This corresponds to computing
\[
\inf _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}(s \models P) \quad \text { and } \quad \sup _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}(s \models P)
\]
where the infimum and the supremum are taken over all schedulers \(\mathfrak{S}\) for \(\mathcal{M}\). Later in this section, we will establish that the infimum and supremum may be replaced by minimum and maximum, respectively. Ranging over all schedulers and considering minimal or maximal probabilities corresponds to a worst-case analysis. This is due to the fact that the full class of schedulers covers all possible resolutions of the nondeterminism that is present.

\section*{Example 10.94. A Randomized Mutual Exclusion Protocol (Revisited)}

Consider again the MDP modeling the mutual exclusion with a randomized arbiter; see Example 10.83 (page 835). Consider the \(\omega\)-regular property stating that the first process infinitely often enters its critical section, i.e., \(\square \diamond\) crit \(_{1}\). Let scheduler \(\mathfrak{S}_{1}\) be such that in any state it only selects one of the enabled actions of the first process, and never an action of the second process. There is a single \(\mathfrak{S}_{1}\)-path where the first process successively passes
its three phases (noncritical, waiting, and critical), while the second process does nothing. Hence,
\[
\operatorname{Pr}_{\mathfrak{S}_{1}}\left(\square \diamond c r i t_{1}\right)=1
\]

On the other extreme, for the scheduler \(\mathfrak{S}_{2}\) which always selects one of the actions of the second process, but ignores the first process, we have
\[
\operatorname{Pr}_{\mathfrak{S}_{2}}\left(\square \diamond c r i t_{1}\right)=0 .
\]

Let us now consider the probability for the event that the first process enters its critical section within at most three waiting rounds from state \(\left\langle\right.\) wait \(_{1}\), noncrit \(\left._{2}\right\rangle\). A waiting round is interpreted as any path fragment of length three where the first process does nothing, while the second process passes through all three phases. For scheduler \(\mathfrak{S}_{1}\), this event almost surely holds. The maximal probability is thus 1 . The minimal probability is \(1-\left(\frac{1}{2}\right)^{3}\) and is obtained by the scheduler \(\mathfrak{S}_{3}\) which selects the action of the second process whenever entering state \(\left.\left\langle\text { wait }_{1} \text {, noncrit }\right\rangle_{2}\right\rangle\). For the states \(\left\langle\right.\) wait \(_{1}\), wait \(\left.{ }_{2}\right\rangle\) and \(\left\langle\right.\) wait \(_{1}\), crit \(\left._{2}\right\rangle\), there is no proper choice and \(\mathfrak{S}_{3}\) selects the uniquely enabled action. For the other states, the decisions of \(\mathfrak{S}_{3}\) are irrelevant.

\section*{Example 10.95. Randomized Leader Election}

Many communication protocols rely on the presence of a certain node (process) that acts as a leader. In order to select such a leader, a distributed algorithm is employed. We consider here the symmetric leader election protocol proposed by Itai and Rodeh [223]. It considers \(n\) processes in a unidirectional ring topology. Each node can act in either of two modes: active or passive. Process \(i\) is connected to its neighbors by FIFO channels \(c_{i}\) and \(c_{i+1}\), each of capacity 1. Addition should be considered modulo the ring size \(n\).

In the active mode, process \(i\) randomly chooses a bit and sends it via channel \(c_{i+1}\) to its unique successor on the ring. Process \(i\) receives the randomly chosen bit by process \(i-1\) along channel \(c_{i}\). If the bit of process \(i\) is 1 , while the obtained bit is zero, then process \(i\) switches to the passive mode. Otherwise, process \(i\) takes part in the next round.

In the passive mode, process \(i\) just acts as a relay node: it passes each bit obtained from its predecessor \(i-1\) to its successor \(i+1\). The leader has been elected as soon as there is only one active process left. The Probmela-specification for process \(i\) is shown in Figure 10.19 on page 847 .

The semantics of the composite Probmela program obtained by the specifications for the \(n\) processes is an MDP. The correctness of the algorithm can formally be established by showing that under each scheduler almost surely eventually the number of active processes (as indicated by the variable \#active) equals one.
```

mode $_{i}:=$ active;
do $::$ mode $_{i}=$ active $\Rightarrow$
$x_{i}:=\operatorname{random}(0,1) ;$
$c_{i+1}!x_{i} ; c_{i} ? y_{i}$;
if $:: y_{i}=1 \wedge x_{i}=0 \Rightarrow$ mode $_{i}:=$ passive;
\#active := \#active - 1
$:: y_{i}=0 \vee x_{i}=1 \Rightarrow$ skip
fi
$::$ mode $_{i}=$ passive $\Rightarrow c_{i} ? y_{i} ; c_{i+1}!y_{i}$
od

```

Figure 10.19: Probmela code for the randomized leader election protocol.

Definition 10.91 provides a rather general notion of scheduler and does not impose any restrictions on the decisions of a scheduler. The function from histories (i.e., finite path fragments) to actions does not require any consistency in the decisions and even allows for schedulers for which the decisions are not computable. For the verification of a large class of qualitative and quantitative properties of finite MDPs, however, a simple subclass of schedulers suffices. This means, e.g., that the maximal and minimal reachability probabilities (when ranging over the full class of schedulers) are obtained by schedulers of this subclass. The full generality of schedulers, as in Definition 10.91, is thus not needed.

\section*{Definition 10.96. Memoryless Scheduler}

Let \(\mathcal{M}\) be an MDP with state space \(S\). Scheduler \(\mathfrak{S}\) on \(\mathcal{M}\) is memoryless (or: simple) iff for each sequence \(s_{0} s_{1} \ldots s_{n}\) and \(t_{0} t_{1} \ldots t_{m} \in S^{+}\)with \(s_{n}=t_{m}\) :
\[
\mathfrak{S}\left(s_{0} s_{1} \ldots s_{n}\right)=\mathfrak{S}\left(t_{0} t_{1} \ldots t_{m}\right)
\]

In this case, \(\mathfrak{S}\) can be viewed as a function \(\mathfrak{S}: S \rightarrow\) Act.

Stated in words, scheduler \(\mathfrak{S}\) is memoryless if it always selects the same action in a given state. This choice is independent of what has happened in the history, i.e., which path led to the current state.

For instance, the schedulers \(\mathfrak{S}_{1}, \mathfrak{S}_{2}\) and \(\mathfrak{S}_{3}\) described informally in Example 10.94 (page 845) are all memoryless. The schedulers \(\mathfrak{S}_{\alpha}\) and \(\mathfrak{S}_{\beta}\) in Example 10.93 (page 843) are also memoryless. The scheduler \(\mathfrak{S}\) in Example 10.93 is not memoryless as it bases its decisions on the one-but-last state.

Memoryless schedulers are somehow extreme as they simply select one alternative (i.e., action) per state while ignoring all others. For instance, for the mutual exclusion example,
memoryless schedulers always select the same process to proceed in the states where both processes are enabled to perform an action. A variant of memoryless schedulers are socalled finite-memory schedulers, briefly fm-schedulers. The behavior of a fm-scheduler is described by a deterministic finite automaton (DFA). The selection of the action to be performed in the MDP \(\mathcal{M}\) depends on the current state of \(\mathcal{M}\) (as before) and the current state (called mode) of the scheduler, i.e., the DFA.

\section*{Definition 10.97. Finite-Memory Scheduler}

Let \(\mathcal{M}\) be an MDP with state space \(S\) and action set Act. A finite-memory scheduler \(\mathfrak{S}\) for \(\mathcal{M}\) is a tuple \(\mathfrak{S}=(Q\), act, \(\Delta\), start \()\) where
- \(Q\) is a finite set of modes,
- \(\Delta: Q \times S \rightarrow Q\) is the transition function,
- act : \(Q \times S \rightarrow\) Act is a function that selects an action \(\operatorname{act}(q, s) \in \operatorname{Act}(s)\) for any mode \(q \in Q\) and state \(s\) of \(\mathcal{M}\),
- start : \(S \rightarrow Q\) is a function that selects a starting mode for state \(s\) of \(\mathcal{M}\).

The behavior of an MDP \(\mathcal{M}\) under a finite-memory scheduler \(\mathfrak{S}=(Q\), act, \(\Delta\), start \()\) is as follows. Initially, a starting state \(s_{0}\) is randomly determined according to the initial distribution \(\iota_{\text {init }}\), i.e., \(\iota_{\text {init }}\left(s_{0}\right)>0\). The fm-scheduler \(\mathfrak{S}\) initializes its DFA to the mode \(q_{0}=\operatorname{start}\left(s_{0}\right) \in Q\). Assume that \(\mathcal{M}\) is in state \(s\) and the current mode of \(\mathfrak{S}\) is \(q\). The decision of \(\mathfrak{S}\), i.e., the selected action, is now given by \(\alpha=\operatorname{act}(q, s) \in \operatorname{Act}(s)\). The scheduler subsequently changes to mode \(\Delta(q, s)\), while \(\mathcal{M}\) performs the selected action \(\alpha\) and randomly moves to the next state according to the distribution \(\mathbf{P}(s, \alpha, \cdot)\).

Let us briefly explain how finite-memory schedulers are related to the (general) notion of a scheduler, see Definition 10.91 (page 842). A finite-memory scheduler \(\mathfrak{S}=\) \(\left(Q\right.\), act, \(\Delta\), start) is identified with the function, i.e., scheduler, \(\mathfrak{S}^{\prime}:\) Paths \(f_{i n} \rightarrow\) Act which is defined as follows. For the starting state \(s_{0}\), let \(\mathfrak{S}^{\prime}\left(s_{0}\right)=\operatorname{act}\left(\operatorname{start}\left(s_{0}\right), s_{0}\right)\). For path fragment \(\widehat{\pi}=s_{0} s_{1} \ldots s_{n}\) let
\[
\mathfrak{S}^{\prime}(\widehat{\pi})=\operatorname{act}\left(q_{n}, s_{n}\right)
\]
where \(q_{0}=\operatorname{start}\left(s_{0}\right)\) and \(q_{i+1}=\Delta\left(q_{i}, s_{i}\right)\) for \(0 \leqslant i \leqslant n\).

Finite-memory schedulers enjoy the property that the Markov chain \(\mathcal{M}_{\mathfrak{S}}\) can be identified with a Markov chain where the states are just pairs \(\langle s, q\rangle\) where \(s\) is a state in the MDP
\(\mathcal{M}\) and \(q\) a mode of \(\mathfrak{S}\). Formally, \(\mathcal{M}_{\mathfrak{S}}^{\prime}\) is the Markov chain with state space \(S \times Q\), labeling \(L^{\prime}(\langle s, q\rangle)=L(s)\), the starting distribution \(\iota_{\text {init }}\), and the transition probabilities:
\[
\mathbf{P}_{\mathfrak{S}}^{\prime}(\langle s, q\rangle,\langle t, p\rangle)=\mathbf{P}(s, \operatorname{act}(q, s), t)
\]

It can be shown that the infinite Markov chain \(\mathcal{M}_{\mathfrak{S}}\) is probabilistic bisimulation-equivalent to \(\mathcal{M}_{\mathfrak{S}}^{\prime}\). This justifies to identifying \(\mathcal{M}_{\mathfrak{S}}\) with \(\mathcal{M}_{\mathfrak{S}}^{\prime}\). Hence, if \(\mathcal{M}\) is a finite MDP, then we consider \(\mathcal{M}_{\mathfrak{S}}\) as a finite MC.

Memoryless schedulers can be considered as finite-memory schedulers with just a single mode. That is, the Markov chain \(\mathcal{M}_{\mathfrak{S}}\) induced by the memoryless scheduler \(\mathfrak{S}\) can be viewed as a Markov chain with state space \(S\). In particular, for finite \(\mathcal{M}, \mathcal{M}_{\mathfrak{S}}\) can be viewed as a finite Markov chain which results by selecting a single enabled action per state (and discarding the other enabled actions). For finite MDP \(\mathcal{M}\) the number of memoryless schedulers is finite, albeit sometimes very large. Assuming, e.g., there are exactly two enabled actions per state, then the total number of memoryless schedulers is \(2^{n}\) where \(n=|S|\).

\section*{Example 10.98. Memoryless vs. Finite-Memory Schedulers}

Consider the MDP \(\mathcal{M}\) depicted below:


We have
- \(\operatorname{Act}\left(s_{0}\right)=\{\alpha, \beta\}, \mathbf{P}\left(s_{0}, \alpha, t\right)=\mathbf{P}\left(s_{0}, \beta, u\right)=1\), and
- \(\operatorname{Act}(t)=\operatorname{Act}(u)=\{\gamma\}\) with \(\mathbf{P}\left(t, \gamma, s_{0}\right)=\mathbf{P}\left(u, \gamma, s_{0}\right)=1\).

The MDP is deterministic apart from state \(s_{0}\) where a nondeterministic choice between actions \(\alpha\) and \(\beta\) exists. There are only two memoryless schedulers for \(\mathcal{M}\) : scheduler \(\mathfrak{S}_{\alpha}\), which always chooses \(\alpha\) in \(s_{0}\), and scheduler \(\mathfrak{S}_{\beta}\), which always chooses \(\beta\) for \(s_{0}\).

The \(\beta\)-successor \(u\) of \(s_{0}\) is not reachable from \(s_{0}\) in the Markov chain \(\mathcal{M}_{\mathfrak{S}_{\alpha}}\) while by symmetry the \(\alpha\)-successor \(t\) of \(s_{0}\) is not accessible from \(s_{0}\) under scheduler \(\mathfrak{S}_{\beta}\). For both memoryless schedulers, the probability of satisfying the \(\omega\)-regular property \(\diamond a \wedge \diamond b\) is thus zero:
\[
\operatorname{Pr}_{\mathfrak{S}_{\alpha}}\left(s_{0} \models \diamond a \wedge \Delta b\right)=\operatorname{Pr}_{\mathfrak{S}_{\beta}}\left(s_{0} \models \Delta a \wedge \diamond b\right)=0 .
\]

However, for the - nonmemoryless-scheduler \(\mathfrak{S}_{\alpha \beta}\), which alternates between selecting \(\alpha\) and \(\beta\) (starting with \(\alpha\), say) on visiting \(s_{0}\), the event \(\diamond a \wedge \diamond b\) holds almost surely. In fact, since \(\mathcal{M}\) does not contain a proper probabilistic choice there is exactly one \(\mathfrak{S}_{\alpha \beta}\)-path starting in \(s_{0}\) :
\[
\pi=s_{0} \xrightarrow{\alpha} t \xrightarrow{\gamma} s_{0} \xrightarrow{\beta} u \xrightarrow{\gamma} s_{0} \xrightarrow{\alpha} t \xrightarrow{\gamma} s_{0} \xrightarrow{\beta} \ldots,
\]
and \(\pi \models \diamond a \wedge \diamond b\). Thus: although the event \(\diamond a \wedge \diamond b\) holds with probability zero under all memoryless schedulers, there is a nonmemoryless scheduler for which this event almost surely holds. This shows that the class of memoryless schedulers is insufficiently powerful to characterize the minimal (or, dually, maximal) probability for \(\omega\)-regular events.

The precise definition of the finite-memory scheduler \(\mathfrak{S}_{\alpha \beta}=(Q\), act, \(\Delta\), start \()\) is as follows. It has two modes: one in which it is only able to select \(\alpha\), while in the other mode only \(\beta\) can be selected. The scheduler switches mode whenever visiting \(s_{0}\). Formally, the state space is \(Q=\left\{q_{\alpha}, q_{\beta}\right\}\). The action function is given by
\[
\operatorname{act}\left(q_{\beta}, s_{0}\right)=\beta \quad \text { and } \quad \operatorname{act}\left(q_{\alpha}, s_{0}\right)=\alpha
\]
and \(\operatorname{act}(q, t)=\operatorname{act}(q, u)=\gamma\) for \(q \in Q\). Swapping modes is formalized by \(\Delta\left(q_{\beta}, s_{0}\right)=q_{\alpha}\) and \(\Delta\left(q_{\alpha}, s_{0}\right)=q_{\beta}\). If \(\mathcal{M}\) is in state \(t\) or \(u\), then \(\mathfrak{S}\) stays in its current mode. This is formalized by \(\Delta\left(q_{\beta}, t\right)=\Delta\left(q_{\beta}, u\right)=q_{\beta}\) and \(\Delta\left(q_{\alpha}, t\right)=\Delta\left(q_{\alpha}, u\right)=q_{\alpha}\). As we assume that on the first visit of \(s_{0}\), the action \(\alpha\) is selected, the starting mode is defined by \(\operatorname{start}(s)=q_{\alpha}\) for all \(s \in S\).

\section*{Remark 10.99. Randomized Schedulers}

Schedulers, as in Definition 10.91 (page 842), are deterministic since they select a unique action for the current state. This can be generalized by allowing schedulers to select enabled actions probabilistically. That is, given a history a randomized scheduler returns a probability for each action. (The reader should not confuse this aspect of randomness with the random selection of the successor state within an MDP.) Mathematically, this means that randomized schedulers are functions \(\mathfrak{S}: S^{+} \rightarrow \operatorname{Distr}(A c t)\), where \(\operatorname{Distr}(A c t)\) is a distribution over the set \(A c t\). It is required that any action \(\alpha\) for which \(\mathfrak{S}\left(s_{0} \ldots s_{n}\right)(\alpha)>0\) is enabled in the state \(s_{n}\). We do not consider randomized schedulers any further and just mention that they can be approximated by deterministic schedulers and yield the same extreme probabilities for \(\omega\)-regular properties as deterministic schedulers. Thus, although they are more general, randomized schedulers do not yield any extra power for our purposes. This justifies not treating them any further here.

\subsection*{10.6.1 Reachability Probabilities}

This section treats the computation of (constrained) reachability probabilities in MDPs. That is, we are concerned with the following problem. Let \(\mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite MDP and \(B \subseteq S\) a set of target states. The measure of interest is the maximal, or dually, the minimal probability of reaching a state in \(B\) when starting in state \(s \in S\). For maximal probabilities this amounts to determining
\[
\operatorname{Pr}^{\max }(s \models \diamond B)=\sup _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}(s \models \diamond B)
\]

Note that the supremum ranges over all, potentially infinitely many, schedulers for \(\mathcal{M}\). This section will show that these maximal probabilities can be computed by solving a linear program. (Recall that reachability probabilities in MCs can be determined by solving a linear equation system.) Furthermore, it will be shown that rather than considering all schedulers-including history-dependent, finite-memory schedulers, and so forth-it suffices to only consider the subclass of memoryless schedulers. That is, there exists a memoryless scheduler that maximizes the probabilities to reach \(B\). This holds for any state \(s\). The supremum can thus be replaced by a maximum.

Reasoning about the maximal probabilities for \(\diamond B\) is needed, e.g., for showing that \(\operatorname{Pr}^{\mathfrak{S}}(s \models \diamond B) \leqslant \varepsilon\) for all schedulers \(\mathfrak{S}\) and some small upper bound \(0<\varepsilon \leqslant 1\). Then:
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models \square \neg B) \geqslant 1-\varepsilon \quad \text { for all schedulers } \mathfrak{S}
\]

The task to compute \(\operatorname{Pr}^{\max }(s \models \diamond B)\) can thus be understood as showing that a safety property (namely \(\square \neg B\) ) holds with sufficiently large probability, viz. \(1-\varepsilon\), regardless of the resolution of nondeterminism.

\section*{Theorem 10.100. Equation System for Max Reachability Probabilities}

Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S, s \in S\) and \(B \subseteq S\). The vector \(\left(x_{s}\right)_{s \in S}\) with \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\) yields the unique solution of the following equation system:
- If \(s \in B\), then \(x_{s}=1\).
- If \(s \not \vDash \exists \diamond B\), then \(x_{s}=0\).
- If \(s \notin B\) and \(s \models \exists \diamond B\), then
\[
x_{s}=\max \left\{\sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t} \mid \alpha \in \operatorname{Act}(s)\right\}
\]

The CTL-like notations in the last two items refer to the underlying digraph of the MDP \(\mathcal{M}\). That is, \(s \models \exists \diamond B\) asserts that \(B\) is reachable from \(s\). Obviously, \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\) is a solution of the above equation system. The proof of its uniqueness is rather technical and omitted here. It uses similar arguments as for Markov chains, see Theorem 10.19 on page 766. As for Markov chains, the second item could be omitted and replaced by the requirement that the equations for \(x_{s}\) in the third item hold for all states \(s \in S \backslash B\). The uniqueness of \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\) is then, however, no longer guaranteed. As for Markov chains, one can prove that \(\left(x_{s}\right)_{s \in S}\) is the least solution in \([0,1]^{S}\).

\section*{Example 10.101. Equation System for Max Reachability Probabilities}

Consider the MDP \(\mathcal{M}\) depicted in Figure 10.20 on page 852. Assume we are interested in \(\operatorname{Pr}^{\max }(s \models \diamond B)\) with \(B=\left\{s_{2}\right\}\). The vector \(\left(x_{s}\right)_{s \in S}\) with \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\) yields the unique solution of the following equation system: \(x_{3}=0, x_{2}=1\) and
\[
x_{0}=\max \left\{\frac{3}{4} x_{2}+\frac{1}{4} x_{3}, \frac{1}{2} x_{2}+\frac{1}{2} x_{1}\right\} \quad \text { and } \quad x_{1}=\frac{1}{2} x_{0}+\frac{1}{2} x_{3}=\frac{1}{2} x_{0}
\]
where \(x_{i}\) denotes \(x_{s_{i}}\). The unique solution of this set of linear equations is: \(\left(x_{s}\right)_{s \in S}=\) \(\left(\frac{3}{4}, \frac{3}{8}, 1,0\right)\).


Figure 10.20: Example MDP for max reachability.

The following result asserts that there exists a memoryless scheduler that maximizes the probabilities of reaching \(B\) eventually. This holds for any state \(s\) in MDP \(\mathcal{M}\).

Lemma 10.102. Existence of Optimal Memoryless Schedulers
Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S\), and \(B \subseteq S\). There exists a memoryless scheduler \(\mathfrak{S}\) such that for any \(s \in S\)
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models \diamond B)=\operatorname{Pr}^{\max }(s \models \diamond B) .
\]

Proof: Let \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\). The proof is by constructing a memoryless scheduler \(\mathfrak{S}\) such that \(\operatorname{Pr}^{\mathfrak{G}}(s \models \diamond B)=\operatorname{Pr}^{\max }(s \vDash \diamond B)\). This goes as follows. For any state \(s\), let \(\operatorname{Act}{ }^{\max }(s)\) be the set of actions \(\alpha \in \operatorname{Act}(s)\) such that
\[
x_{s}=\sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t}
\]

We first observe that it does not suffice to select arbitrary actions from the set \(A c t^{\max }(s)\). For example, consider a state with \(A c t^{\max }(s)=\{\alpha, \beta\}\) where \(\mathbf{P}(s, \beta, t)=1\) for some \(t \in B\), while via \(\alpha\) the set \(B\) cannot be reached, e.g., since \(\mathbf{P}(s, \alpha, s)=1\). A selection of actions is thus needed that ensures reachability of \(B\) in the induced Markov chain under \(\mathfrak{S}\).

Consider the MDP \(\mathcal{M}^{\max }\) that results from \(\mathcal{M}\) by removing the actions \(\beta \in \operatorname{Act}(s) \backslash\) \(\operatorname{Act}{ }^{\max }(s)\) from \(\operatorname{Act}(s)\) for any state \(s\) for which \(B \cap \operatorname{Post}^{*}(s) \neq \varnothing\). This simplification of \(\mathcal{M}\) does not change the maximal probabilities to reach \(B\). For \(s \models \exists \diamond B\), let \(\|s\|\) be the length of a shortest path fragment from \(s\) to \(B\) in the MDP \(\mathcal{M}^{\max }\). It follows that \(\|s\|=0\) if and only if \(s \in B\). By induction on \(n \geqslant 1\) we define actions \(\mathfrak{S}(s)\) for the states \(s\) with \(s \models \exists \diamond B\) and \(\|s\|=n\). If \(\|s\|=n \geqslant 1\), then choose an action \(\mathfrak{S}(s) \in A c t^{\max }(s)\) such that \(\mathbf{P}(s, \mathfrak{S}(s), t)>0\) for some state \(t\) with \(t \models \exists \diamond B\) and \(\|t\|=n-1\). For the states \(s\) that cannot reach \(B\), we select an arbitrary action \(\mathfrak{S}(s) \in \operatorname{Act}(s)\). This yields a memoryless scheduler \(\mathfrak{S}\). The induced Markov chain \(\mathcal{M}_{\mathfrak{S}}\) is finite with state space \(S\). Moreover, the probabilities for \(\diamond B\) provide the unique solution of the linear equation system:
- If \(s \in B\), then \(y_{s}=1\).
- If \(s \not \vDash \exists \diamond B\), then \(y_{s}=0\).
- If \(s \notin B\) and \(s \models \exists \diamond B\), then \(y_{s}=\sum_{t \in S} \mathbf{P}(s, \mathfrak{S}(s), t) \cdot y_{t}\).

Since the vector \(\left(x_{s}\right)_{s \in S}\) also solves the above equation system we obtain:
\[
\operatorname{Pr}^{\mathfrak{G}}(s \models \Delta B)=y_{s}=x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)
\]

An optimal memoryless scheduler for the MDP in Figure 10.20 and the event \(\diamond B\) is the scheduler that selects \(\alpha\) in any state.

Theorem 10.100 suggests an iterative approximation technique, called value iteration, to calculate the values \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\). By means of a backward reachability analysis, the set of states \(s\) is determined such that \(s \neq \exists \diamond B\). These are precisely the states for which \(x_{s}>0\). For the states \(s \in \operatorname{Pre}^{*}(B) \backslash B\) we have
\[
x_{s}=\lim _{n \rightarrow \infty} x_{s}^{(n)}
\]
where
\[
x_{s}^{(0)}=0 \quad \text { and } \quad x_{s}^{(n+1)}=\max \left\{\sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t}^{(n)} \mid \alpha \in \operatorname{Act}(s)\right\}
\]

In the above equation also states in \(B \cup\left(S \backslash \operatorname{Pre}^{*}(B)\right)\) might appear. For these states we have
\[
x_{s}^{(n)}=1 \text { if } s \in B \quad \text { and } \quad x_{s}^{(n)}=0 \text { if } s \notin \operatorname{Pre}^{*}(B) .
\]

Note that \(x_{s}^{(0)} \leqslant x_{s}^{(1)} \leqslant x_{s}^{(2)} \leqslant \ldots\). Thus, the values \(\operatorname{Pr}^{\max }(s \models \diamond B)\) can be approximated by successively computing the vectors
\[
\left(x_{s}^{(0)}\right),\left(x_{s}^{(1)}\right),\left(x_{s}^{(2)}\right), \ldots
\]
until \(\max _{s \in S}\left|x_{s}^{(n+1)}-x_{s}^{(n)}\right|\) is below a certain (typically very small) threshold.

\section*{Example 10.103. Value Iteration}

To illustrate the value iteration, we consider the MDP depicted in Figure 10.21 on page 855. As the atomic propositions are irrelevant, they have been omitted. Let \(B=\left\{s_{3}\right\}\). It follows that \(x_{3}^{(i)}=1\) and \(x_{2}^{(i)}=0\) for any \(i\). The latter follows from the fact that \(s_{2} \not \vDash \exists \diamond B\). For the remaining states we obtain
\[
\begin{aligned}
x_{0}^{(i+1)} & =\max \left\{\frac{2}{3} x_{4}^{(i)}, x_{1}^{(i)}\right\} & x_{5}^{(i+1)} & =\max \left\{\frac{2}{3} x_{7}^{(i)}, x_{6}^{(i)}\right\} \\
x_{1}^{(i+1)} & =\frac{1}{2} x_{1}^{(i)}+\frac{1}{9} & x_{6}^{(i+1)} & =\frac{3}{5} x_{6}^{(i)}+\frac{2}{5} x_{5}^{(i)} \\
x_{4}^{(i+1)} & =\frac{1}{4} x_{5}^{(i)}+\frac{3}{4} x_{6}^{(i)} & x_{7}^{(i+1)} & =\frac{1}{2} .
\end{aligned}
\]

The successive computation of the vectors \(\left(x_{s}\right)_{s \in S}\) yields
\[
\left(x^{(0)}\right)=(0,0,0,1,0,0,0,0) \quad \text { and } \quad\left(x^{(1)}\right)=\left(0, \frac{1}{9}, 0,1,0,0,0, \frac{1}{2}\right)
\]
and
\[
\left(x^{(2)}\right)=\left(\frac{1}{9}, \frac{1}{6}, 0,1,0, \frac{1}{3}, 0, \frac{1}{2}\right) \quad \text { and } \quad\left(x^{(3)}\right)=\left(\frac{1}{6}, \frac{7}{36}, 0,1, \frac{1}{12}, \frac{1}{3}, \frac{2}{15}, \frac{1}{2}\right) \ldots
\]


Figure 10.21: Example of an MDP.

Remark 10.104. Value Iteration for Step-Bounded Reachability Properties
The value iteration approach also yields a method to compute the maximal probabilities for the event \(\Delta \leqslant n B\). In fact, we have
\[
x_{s}^{(n)}=\sup _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}\left(s \models \diamond^{\leqslant n} B\right)
\]
where \(\mathfrak{S}\) ranges over all schedulers. Furthermore, there exists a finite-memory scheduler \(\mathfrak{S}\) that maximizes the probabilities for \(\diamond^{\leqslant n} B\). (Thus, the supremum can be replaced by maximum.) Such an optimal fm-scheduler can be composed by using modes \(0,1, \ldots, n\). The starting mode is 0 for each state in \(\mathcal{M}\). The next-mode function changes from mode \(i-1\) to \(i\) for \(1 \leqslant i \leqslant n\). As soon as mode \(n\) has been reached, \(\mathfrak{S}\) stays forever in mode \(n\). For mode \(i \in\{0,1, \ldots, n-1\}\) and current state \(s\) in \(\mathcal{M}, \mathfrak{S}\) selects an action \(\alpha \in \operatorname{Act}(s)\) that maximizes the value \(\sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t}^{(i-1)}\). The actions \(\mathfrak{S}(n, s)\) are arbitrary. Then, \(\operatorname{Pr}^{\mathscr{G}}\left(s \models \diamond^{\leqslant n} B\right)=x_{s}^{(n)}\).

An alternative method to compute \(\operatorname{Pr}^{\max }(s \models \diamond B)\) is obtained by rewriting the equation system in Theorem 10.100 into a linear program.

\section*{Theorem 10.105. Linear Program for Max Reachability Probabilities}

Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S\), and \(B \subseteq S\). The vector \(\left(x_{s}\right)_{s \in S}\) with \(x_{s}=\) \(\operatorname{Pr}^{\max }(s \models \diamond B)\) yields the unique solution of the following linear program:
- If \(s \in B\), then \(x_{s}=1\).
- If \(s \not \vDash \exists \diamond B\), then \(x_{s}=0\).
- If \(s \notin B\) and \(s \models \exists \diamond B\), then \(0 \leqslant x_{s} \leqslant 1\) and for all actions \(\alpha \in \operatorname{Act}(s)\) :
\[
x_{s} \geqslant \sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t}
\]
where \(\sum_{s \in S} x_{s}\) is minimal.

Proof: It is not difficult to see that the vector \(\left(x_{s}\right)_{s \in S}\) with \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\) is a solution to the equations in the first two items, and satisfies the inequalities in the last item. Hence, there exists a solution \(\left(y_{s}\right)_{s \in S}\) of the above linear program. Since the sum of the elements of \(y_{s}\) is minimal under all vectors that satisfy the linear program, it follows that \(\sum_{s} x_{s} \geqslant \sum_{s} y_{s}\).

It remains to show the uniqueness of the solution. In fact, any solution \(\left(y_{s}\right)_{s \in S}\) of the linear program is also a solution of the equation system in Theorem 10.100. Due to the minimality of \(\sum_{s \in S} y_{s}\), the value \(y_{s}\) in the third item agrees with \(\max \left\{\sum_{t} \mathbf{P}(s, \alpha, t) \cdot y_{t} \mid \alpha \in \operatorname{Act}(s)\right\}\).

This can be seen as follows. If this would not hold, then we could apply the value iteration process starting with \(x_{s}^{(0)}=y_{s}\) for all \(s\). This yields a decreasing sequence of vectors \(\left(x_{s}^{(n)}\right)_{n \geqslant 0}\). The limit \(\left(y_{s}^{\prime}\right)_{s \in S}\) solves both the first three items of the linear program and the equations in Theorem 10.100. As \(y_{s}^{\prime}=\lim _{n \rightarrow \infty} x_{s}^{(n)} \leqslant y_{s}\), the minimality of \(\sum_{s \in S} y_{s}\) as a solution of the inequalities in Theorem 10.105 yields \(y_{s}^{\prime}=y_{s}\).

Thus, Theorem 10.100 yields that \(y_{s}=x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\) for all states \(s\).

\section*{Example 10.106. Linear Program}

Consider again the MDP in Figure 10.21 and let \(B=\left\{s_{3}\right\}\). The equations of the linear
program for \(\operatorname{Pr}^{\max }(s \models \diamond B)\) are:
\[
\begin{array}{lll}
x_{0} \geqslant \frac{2}{3} x_{4} & x_{0} \geqslant x_{1} & x_{1} \geqslant \frac{1}{2} x_{1}+\frac{1}{9} \\
x_{2}=0 & x_{3}=1 & x_{4} \geqslant \frac{1}{4} x_{5}+\frac{3}{4} x_{6} \\
x_{5} \geqslant \frac{2}{3} x_{7} & x_{5} \geqslant x_{6} & x_{6} \geqslant \frac{3}{5} x_{6}+\frac{2}{5} x_{5} \\
x_{7} \geqslant \frac{1}{2} & &
\end{array}
\]

Note that the third item in Theorem 10.105 can be rewritten into
\[
\sum_{t \in S_{?} \backslash\{s\}} \mathbf{P}(s, \alpha, t) \cdot x_{t}+(1-\mathbf{P}(s, \alpha, s)) \cdot x_{s} \geqslant \mathbf{P}(s, \alpha, B)
\]
where \(\mathbf{P}(s, \alpha, B)=\sum_{t \in B} \mathbf{P}(s, \alpha, t)\). The set \(S_{\text {? }}\) contains all states such that the value \(x_{s}=\operatorname{Pr}^{\max }(s \models \diamond B)\) is not fixed to 0 or 1 by the first two items, i.e., \(S_{\text {? }}=\{s \in S \backslash B \mid\) \(s \vDash \exists \diamond B\}\). Hence:
\[
s \in S_{?} \quad \text { if and only if } \quad s \notin B \text { and } \operatorname{Pr}^{\max }(s \mid=\Delta B)>0
\]

Thus, the third item in the above theorem can be read as a linear inequality \(\mathbf{A} \cdot \mathbf{x} \geqslant \mathbf{b}\) where \(\mathbf{x}\) is the vector \(\left(x_{s}\right)_{s \in S_{?}}\) and \(\mathbf{A}\) is a matrix with a row for each pair \((s, \alpha)\) with \(s \in S_{\text {? }}\) and \(\alpha \in \operatorname{Act}(s)\), and two extra rows for each state \(s \in S_{\text {? }}\) and a column for each state \(t \in S_{\text {? }}\). The entry of \(\mathbf{A}\) in the row for \((s, \alpha)\) and column for state \(t\) equals \(\mathbf{P}(s, \alpha, t)\) provided \(s \neq t\). For \(s=t\) the entry of \(\mathbf{A}\) equals \(1-\mathbf{P}(s, \alpha, s)\). The two extra rows represent the inequality \(0 \leqslant x_{s} \leqslant 1\), which can be split into the two constraints \(x_{s} \geqslant 0\) and \(-x_{s} \geqslant-1\). Similarly, \(\mathbf{b}\) is a vector with a component for each pair \((s, \alpha)\) where \(s \in S_{\text {? }}\) and \(\alpha \in \operatorname{Act}(s)\) and two extra components per state \(s \in S_{?}\). The value of \(\mathbf{b}\) for \((s, \alpha)\) is the probability of moving from \(s\) via action \(\alpha\) to a state in \(B\) (i.e., the value \(\mathbf{P}(s, \alpha, B)\) ).

In this sense, Theorem 10.105 yields a characterization of the values \(\operatorname{Pr}(s \models \diamond B)\) as the linear optimization problem which asks for the unique solution of \(\mathbf{A} \cdot \mathbf{x} \geqslant \mathbf{b}\) under the side condition that \(\sum_{s \in S} x_{s}\) is minimal under all solutions of \(\mathbf{A x} \geqslant \mathbf{b}\). The precise values for \(\operatorname{Pr}^{\max }(s \models \diamond B)\) can thus be computed by standard algorithms to solve linear programs, e.g., the simplex algorithm or polytime methods [367].

Corollary 10.107. Complexity of Computing Max Reachability Probabilities
For finite \(M D P \mathcal{M}\) with state space \(S, B \subseteq S\) and \(s \in S\), the values \(\operatorname{Pr}^{\max }(s \models \diamond B)\) can be computed in time polynomial in the size of \(\mathcal{M}\).

As a consequence, the question whether \(\operatorname{Pr}^{\mathfrak{G}}(s \models \diamond B) \leqslant p\) for some upper bound \(p \in[0,1[\) is decidable in polynomial time. This result, however, is more of theoretical interest
than of practical relevance. In practice, the simplex method often outperforms polytime algorithms for linear programs, although its worst-case time complexity is exponential. Experiments with randomized distributed algorithms (modeled as MDPs) indicates that the value iteration method is often faster than the simplex method.

The value iteration or linear program approach can be improved by first computing the set of states \(s\) with \(\operatorname{Pr}^{\max }(s \models \diamond B)=1\). This can be efficiently done by standard graph algorithms. For these states, the corresponding equation and inequalities, respectively, for \(x_{s}\) in the third item of Theorem 10.100 and 10.105 respectively can be omitted. This simplifies the value iteration procedure and decreases the size of the linear program.

Let us now consider the qualitative reachability analysis of a finite MDP \(\mathcal{M}\). The goal is to compute the set of states \(s\) for which \(\operatorname{Pr}^{\max }(s \neq \diamond B)=1\). Similar to the case with Markov chains, the states in \(B\) are first made absorbing. This means that any state \(s \in B\) is equipped with a single enabled action \(\alpha_{s}\) with \(\mathbf{P}\left(s, \alpha_{s}, s\right)=1\). Since there exists a memoryless scheduler that maximizes the probabilities for the event \(\diamond B\), it suffices to compute the set of states \(s\) such that \(\operatorname{Pr}^{\mathfrak{G}}(s \models \diamond B)=1\) for some memoryless scheduler \(\mathfrak{S}\), say. The latter is equivalent to the requirement that no state \(t \not \vDash \exists \diamond B\) is reachable from \(s\) in the Markov chain \(\mathcal{M}^{\mathfrak{G}}\) via a path fragment in \(S \backslash B\). As the number of memoryless schedulers may be exponential (in the size of \(\mathcal{M}\) ), considering all such schedulers is not an adequate solution.

However, the analysis of the underlying graph of \(\mathcal{M}\) suffices to determine all states \(s\) where \(\operatorname{Pr}^{\max }(s \models \diamond B)=1\). This approach is iterative and successively removes all vertices \(u\) with \(\operatorname{Pr}^{\max }(u \vDash \diamond B)<1\). Initially, each vertex \(u \in U_{0}\) with \(U_{0}=S \backslash \operatorname{Sat}(\exists \diamond B)\) is removed. The set \(U_{0}\) can simply be determined by a graph analysis. Then, for all states \(t\), all actions \(\alpha\) are deleted from \(\operatorname{Act}(t)\) that satisfy \(\operatorname{Post}(t, \alpha) \cap U_{0} \neq \varnothing\). If after deleting these actions, \(A c t(t)=\varnothing\), then \(t\) will be removed. This procedure is repeated as long as possible, and yields the MDP \(\mathcal{M}_{1}\). Then, we restart the whole procedure with \(\mathcal{M}_{1}\) rather than \(\mathcal{M}\), etc., until all states in the obtained \(\operatorname{MDP} \mathcal{M}_{i}\) can reach \(B\).

This approach is outlined in Algorithm 45. This algorithm treats the MDP as a directed graph where the successors of any vertex \(t \in S\) are the pairs \((t, \alpha)\) with \(\alpha \in \operatorname{Act}(t)\). The outgoing edges of the auxiliary vertices \((t, \alpha)\) lead to the vertices \(u \in \operatorname{Post}(t, \alpha)=\{u \in\) \(S \mid \mathbf{P}(t, \alpha, u)>0\}\). For \(u \in S\), \(\operatorname{Pre}(u)\) denotes the set of state action pairs \((t, \alpha) \in S \times\) Act such that \(\mathbf{P}(t, \alpha, u)>0\). Removing \(\alpha\) from \(\operatorname{Act}(t)\) means removing the auxiliary vertex \((t, \alpha)\), the edge from vertex \(t\) to vertex \((t, \alpha)\), and the outgoing edges from \((t, \alpha)\).
```

Algorithm 45 Computing the set of states $s$ with $\operatorname{Pr}^{\max }(s \models \diamond B)=1$
Input: MDP $\mathcal{M}$ with finite state space $S, B \subseteq S$ for $s \in B: \operatorname{Act}(s)=\left\{\alpha_{s}\right\}$ and $\mathbf{P}\left(s, \alpha_{s}, s\right)=1$
(i.e., $B$ is absorbing)
Output: $\left\{s \in S \mid \operatorname{Pr}^{\max }(s \models \diamond B)=1\right\}$
$U:=\{s \in S \mid s \not \models \exists \diamond B\} ;$
repeat
$R:=U$;
while $R \neq \varnothing$ do
let $u \in R$;
$R:=R \backslash\{u\} ;$
for all $(t, \alpha) \in \operatorname{Pre}(u)$ such that $t \notin U$ do
remove $\alpha$ from $\operatorname{Act}(t)$;
if $\operatorname{Act}(t)=\varnothing$ then
$R:=R \cup\{t\} ;$
$U:=U \cup\{t\} ;$
fi
od
(* all incoming edges of $u$ have been removed *)
remove $u$ and its outgoing edges from $\mathcal{M}$
od
(* determine the states $s$ that cannot reach $B$ in the modified MDP *)
$U:=\{s \in S \backslash U \mid s \not \models \exists \diamond B\} ;$
until $U=\varnothing$
(* all states can reach $B$ in the generated sub-MDP of $\mathcal{M}^{*}$ )
return all states in the remaining MDP

```

\section*{Lemma 10.108. Correctness of Algorithm 45}

For finite MDP \(\mathcal{M}\) and \(B\), a set of states in \(\mathcal{M}\), Algorithm 45 returns the set of all states \(s\) in \(\mathcal{M}\) such that \(\operatorname{Pr}^{\max }(s \mid \diamond B)=1\).

Proof: The correctness of Algorithm 45 relies on the following two facts: (i) for any state \(t\) which is removed, \(\operatorname{Pr}^{\max }(t \models \diamond B)<1\), and (ii) for any action \(\alpha\) removed from \(\operatorname{Act}(t)\) there is no memoryless scheduler \(\mathfrak{S}\) with \(\mathfrak{S}(t)=\alpha\) and \(\operatorname{Pr}^{\mathfrak{S}}(t \mid=\diamond B)=1\).

Now let \(\mathcal{M}_{0}=\mathcal{M}, \mathcal{M}_{1}, \mathcal{M}_{2}, \ldots, \mathcal{M}_{i}\) be the sequence of MDPs that are generated by Algorithm 45. More precisely, \(\mathcal{M}_{j}\) is the MDP obtained by the first \(j\) iterations of the repeat loop. The final MDP \(\mathcal{M}_{i}\) still contains all states \(s\) where \(\operatorname{Pr}^{\max }(s \models \diamond B)=1\) and all actions \(\alpha \in \operatorname{Act}(s)\) that can be used by an optimal (memoryless) scheduler for \(\diamond B\).

Let \(\mathfrak{S}\) be a finite-memory scheduler for \(\mathcal{M}_{i}\) that treats all enabled actions in a fair way (i.e., if \(t\) is visited infinitely often, then any action that is enabled in \(t\) is taken infinitely often). Then the (finite) Markov chain \(\mathcal{M}_{i}^{\mathfrak{S}}\) induced by \(\mathfrak{S}\) enjoys the property that all states can reach \(B\). Corollary 10.30 on page 777 then yields that \(\operatorname{Pr}^{\mathscr{S}}(s \models \diamond B)=1\) for all states \(s\) in \(\mathcal{M}_{i}\). Thus, the set of states in \(\mathcal{M}_{i}\) agrees with the set of states where \(\operatorname{Pr}^{\max }(s \models \diamond B)=1\).

The worst-case complexity of Algorithm 45 is quadratic in the size of \(\mathcal{M}\). This can be seen as follows. The maximal number of iterations of the outermost loop is \(N=|S|\), as in each iteration at least one state is eliminated. In each iteration, the set of states that can reach \(B\) needs to be computed. This takes \(\mathcal{O}(\operatorname{size}(\mathcal{M}))\) time. The other operations cause the overall cost \(\mathcal{O}(\operatorname{size}(\mathcal{M}))\), since any state \(t\) and state action pair \((t, \alpha)\) can be removed at most once.

So far, this section has addressed the problem of computing the maximal probabilities for reaching a certain set \(B\) of states in an MDP. Typically, \(B\) represents a set of bad states and the aim is to show that, regardless of the scheduling policy, the probability of entering a \(B\)-state is sufficiently small. We will now show that similar techniques are applicable to compute the minimal probabilities for reaching a certain set of states. When, e.g., \(B\) stands for a set of good states then
\[
\operatorname{Pr}^{\min }(s \models \diamond B)=\inf _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}(s \models \diamond B)
\]
yields the best lower bound that can be guaranteed for the probability of eventually reaching \(B\). In analogy to Theorem 10.100 , the minimal probabilities can be characterized by an equation system. To ensure the uniqueness of the solution of this equation system it is required that all states \(s\) such that \(\operatorname{Pr}^{\min }(s \models \diamond B)=0\) have the value \(x_{s}=0\).

Theorem 10.109. Equation System for Min Reachability Probabilities
Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S\) and \(B \subseteq S\). The vector \(\left(x_{s}\right)_{s \in S}\) with \(x_{s}=\) \(\operatorname{Pr}^{\min }(s \models \diamond B)\) yields the unique solution of the following equation system:
- If \(s \in B\), then \(x_{s}=1\).
- If \(\operatorname{Pr}^{\min }(s \models \diamond B)=0\), then \(x_{s}=0\).
- If \(\operatorname{Pr}^{\min }(s \models \diamond B)>0\) and \(s \notin B\), then:
\[
x_{s}=\min \left\{\sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t} \mid \alpha \in \operatorname{Act}(s)\right\}
\]

Theorem 10.109 suggests that we first compute the set containing all states \(s\) satisfying \(\operatorname{Pr}^{\min }(s \models \diamond B)=0\), followed by a value iteration method to obtain an approximation of the values \(x_{s}=\operatorname{Pr}^{\min }(s \vDash \diamond B)\) for the states with \(x_{s}>0\) and \(s \notin B\). For this, we put \(x_{s}^{(i)}=0\) if \(\operatorname{Pr}^{\min }(s \models \diamond B)=0\), and \(x_{s}^{(i)}=1\) if \(s \in B\), for all \(i\). For the remaining states, the iteration is started with \(x_{s}^{(0)}=0\). For successive iterations, let
\[
x_{s}^{(n+1)}=\min \left\{\sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t}^{(n)} \mid \alpha \in \operatorname{Act}(s)\right\}
\]

Then: \(x_{s}^{(0)} \leqslant x_{s}^{(1)} \leqslant x_{s}^{(2)} \leqslant \ldots\) and
\[
\lim _{n \rightarrow \infty} x_{s}^{(n)}=\operatorname{Pr}^{\min }(s \models \diamond B)
\]

Moreover, \(x_{s}^{(n)}\) agrees with the minimal probability of reaching \(B\) within at most \(n\) steps, where the minimum is taken over all schedulers. That is:
\[
x_{s}^{(n)}=\min _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}\left(s \models \diamond^{\leqslant n} B\right) .
\]

In fact, this minimum always exists and can be established by a finite-memory scheduler. This scheduler has \(n\) modes, 0 through \(n-1\). The starting mode is 0 , the next-mode function switches from mode \(i-1\) to mode \(i\) for \(0<i \leqslant n\) and stays forever in mode \(n\) after the \(n\)th step. \(\mathfrak{S}(i, s)\) is an action \(\alpha \in \operatorname{Act}(s)\) that minimizes the value \(\sum_{t \in S} \mathbf{P}(s, \alpha, t)\). \(x_{t}^{(i-1)}\). The actions \(\mathfrak{S}(n, s) \in \operatorname{Act}(s)\) are irrelevant.

The preprocessing required to compute the set
\[
S_{=0}^{\min }=\left\{s \in S \mid \operatorname{Pr}^{\min }(s \models \diamond B)\right\}=0
\]
can be performed by graph algorithms. The set \(S_{=0}^{\min }\) is given by \(S \backslash T\) where
\[
T=\bigcup_{n \geqslant 0} T_{n}
\]
and \(T_{0}=B\) and, for \(n \geqslant 0\) :
\[
T_{n+1}=T_{n} \cup\left\{s \in S \mid \forall \alpha \in \operatorname{Act}(s) \exists t \in T_{n} . \mathbf{P}(s, \alpha, t)>0\right\}
\]

As \(T_{0} \subseteq T_{1} \subseteq T_{2} \subseteq \ldots \subseteq S\) and \(S\) is finite, the sequence \(\left(T_{n}\right)_{n \geqslant 0}\) eventually stabilizes, i.e., there exists some \(n \geqslant 0\) such that \(T_{n}=T_{n+1}=\ldots=T\).

\section*{Lemma 10.110. Characterization of \(S_{=0}^{\min }\)}

Let \(\mathcal{M}\) and \(T\) be as above. Then, \(S_{=0}^{\min }=S \backslash T\), i.e., for all states \(s \in S\) :
\[
\operatorname{Pr}^{\min }(s \models \Delta B)>0 \quad \text { iff } \quad s \in T
\]

Proof: We show by induction on \(n\) that \(\operatorname{Pr}^{\min }(s \models \diamond B)>0\) for all \(s \in T_{n}\). Base case: For \(n=0\) the claim trivially holds as \(T_{0}=B\) and \(\operatorname{Pr}^{\mathfrak{S}}(s \mid \forall B)=1\) for each scheduler \(\mathfrak{S}\) and state \(s \in B\). Induction step: Assume the claim holds for 0 through \(n-1\). Let \(n \geqslant 1\). Consider \(s \in T_{n} \backslash T_{n-1}\) and let \(\mathfrak{S}\) be a scheduler. Let \(\alpha=\mathfrak{S}(s)\) be the action selected by \(\mathfrak{S}\) when \(\mathcal{M}\) starts in \(s\). By definition of \(T_{n}, T_{n-1} \cap \operatorname{Post}(s, \alpha) \neq \varnothing\). Let \(t \in T_{n-1} \cap \operatorname{Post}(s, \alpha)\). By induction hypothesis, it holds that \(\operatorname{Pr}^{\min }(t \vDash \diamond B)>0\). For \(s \in T_{n}\), it then holds that
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models \diamond B) \geqslant \mathbf{P}(s, \alpha, t) \cdot \operatorname{Pr}^{\min }(t \models \diamond B)>0
\]

Therefore, \(\operatorname{Pr}^{\min }(s \models \diamond B)>0\).
Vice versa, if \(s \in S \backslash T\), then \(\operatorname{Post}(s, \alpha) \cap T=\varnothing\) for some action \(\alpha \in \operatorname{Act}(s)\). Hence, we may consider a memoryless scheduler \(\mathfrak{S}\) which selects for each state \(s \in S \backslash T\) such action \(\alpha \in \operatorname{Act}(s)\). Then, in the Markov chain \(\mathcal{M}_{\mathfrak{S}}\) the states in \(S \backslash T\) cannot reach any state in \(T\). Therefore, \(\operatorname{Pr}^{\mathscr{G}}(s \models \diamond B)=0\) for all \(s \in S \backslash T\). This yields \(\operatorname{Pr}^{\min }(s \models \diamond B)=0\) for \(s \in S \backslash T\).

The computation of \(S \backslash S_{=0}^{\min }=T=\bigcup_{n \geqslant 0} T_{n}\) can be performed in time linear in the size of \(\mathcal{M}\). The main steps of such a linear-time algorithm are summarized in Algorithm 46.

To speed up the convergence of the value iteration process, one might also compute \(S_{=1}^{\min }=\left\{s \in S \mid \operatorname{Pr}^{\min }(s \mid \diamond B)=1\right\}\) using graph analysis techniques. Note that the computation of \(S_{=1}^{\min }\) also solves the qualitative verification problem which asks whether \(\diamond B\) almost surely holds under all schedulers.
```

Algorithm 46 Computing the set of states $s$ with $\operatorname{Pr}^{\min }(s \models \diamond B)=0$
Input: finite MDP $\mathcal{M}$ with state space $S$ and $B \subseteq S$
Output: $\left\{s \in S \mid \operatorname{Pr}^{\min }(s \models \diamond B)=0\right\}$

```
```

T:= B;
R:=B;
while R\not=\varnothing do
let }t\inR\mathrm{ ;
R:= R\{t};
for all (s,\alpha)\in\operatorname{Pre}(t)\mathrm{ with }s\not\inT\mathrm{ do}
remove \alpha from Act(s)
if Act(s)=\varnothing then
add s to R and T
fi
od
od
return T

```

\section*{Lemma 10.111. Characterization of \(S_{=1}^{\min }\)}

Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S, B \subseteq S\) and \(s \in S\). The following statements are equivalent:
(a) \(\operatorname{Pr}^{\min }(s \models \diamond B)<1\).
(b) There exists a memoryless scheduler \(\mathfrak{S}\) such that \(\operatorname{Pr}^{\mathfrak{S}}(s \models \square \neg B)>0\).
(c) \(s \models \exists((\neg B) \cup t)\) for some state \(t\) such that \(\operatorname{Pr}^{\mathfrak{S}}(t \models \square \neg B)=1\) for some memoryless scheduler \(\mathfrak{S}\).

Proof: \((\mathrm{a}) \Longrightarrow(\mathrm{b})\) : If \(\operatorname{Pr}^{\min }(s \models \diamond B)<1\) then, by Lemma 10.113 (page 865), \(\operatorname{Pr}^{\mathscr{G}}(s \models\) \(\diamond B)<1\) for some memoryless scheduler \(\mathfrak{S}\). But then
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models \square \neg B)=1-\operatorname{Pr}^{\mathfrak{S}}(s \models \diamond B)>0 .
\]
\((\mathrm{b}) \Longrightarrow(\mathrm{c})\) : Let \(\mathfrak{S}\) be a memoryless scheduler such that \(\operatorname{Pr}^{\mathfrak{S}}(s \vDash \square \neg B)>0\). Consider the Markov chain \(\mathcal{M}_{\mathfrak{S}}\). As \(\mathfrak{S}\) is memoryless, \(\mathcal{M}_{\mathfrak{S}}\) is finite. By Theorem 10.27 on page 775, almost all paths in \(\mathcal{M}_{\mathfrak{S}}\) eventually enter a BSCC and visit all its states infinitely often. Hence, there is a BSCC \(C\) of \(\mathcal{M}_{\mathfrak{S}}\) such that \(C \cap B=\varnothing\) and
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models \exists((\neg B) \cup C))>0 .
\]

Furthermore, \(\operatorname{Pr}^{\mathscr{G}}(t \models \square \neg B)=1\) for all states \(t \in C\).
(c) \(\Longrightarrow\left(\right.\) a): Assume \(s \vDash \exists((\neg B) \cup t)\) for some state \(t\) such that \(\operatorname{Pr}^{\mathfrak{G}}(t \models \square \neg B)=1\) for some memoryless scheduler \(\mathfrak{S}\). Let
\[
\widehat{\pi}=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} \ldots \xrightarrow{\alpha_{n}} s_{n}
\]
such that \(s_{0}=s, s_{i} \notin B\) for \(0 \leqslant i<n\) and \(s_{n}=t\). Let scheduler \(\mathfrak{S}^{\prime}\) generate this path fragment with positive probability, i.e., \(\mathfrak{S}^{\prime}\left(s_{0} s_{1} \ldots s_{i}\right)=\alpha_{i+1}\) for \(0 \leqslant i<n\). Moreover, \(\mathfrak{S}^{\prime}\) behaves as \(\mathfrak{S}\) for all state sequences that extend \(s_{0} s_{1} \ldots s_{n}\). Then:
\[
\operatorname{Pr}^{\mathfrak{S}^{\prime}}(s \models \square \neg B) \geqslant \prod_{1 \leqslant i \leqslant n} \mathbf{P}\left(s_{i-1}, \alpha_{i}, s_{i}\right)>0
\]
and therefore \(\operatorname{Pr}^{\mathfrak{S}^{\prime}}(s \models \diamond B)<1\).

Lemma 10.111 suggests computing the set \(S_{=1}^{\min }\) as the complement of the set of states that can reach \(T\) with
\[
T=\left\{s \in S \mid \operatorname{Pr}^{\max }(t \mid=\square \neg B)=1\right\}
\]
via a path fragment through \(S \backslash B\). The set \(T\) can be obtained as follows. Initially, let \(A(s)=\operatorname{Act}(s)\) for any \(s \in S \backslash B\) and \(A(t)=\varnothing\) for \(t \in B\). Any state \(t \in B\) is removed from \(T\). This entails that for all state action pairs \((s, \alpha) \in \operatorname{Pre}(t)\) such that \(\alpha \in A(s)\) and \(\alpha\) is removed from \(A(s)\). If in this way \(A(s)\) becomes empty for some state \(s \in T\), then \(s\) is removed from \(T\) by the same procedure. That is, all state action pairs \((u, \beta) \in \operatorname{Pre}(s)\) with \(\beta \in A(u)\) are considered and \(\beta\) is removed from \(\operatorname{Act}(u)\). This elimination process is repeated as long as there are states \(s\) with \(A(s)=\varnothing\). This results in a sub-MDP with state space \(T \subseteq S \backslash B\) and nonempty action sets \(A(t) \subseteq A c t(t)\) such that \(\operatorname{Post}(t, \alpha) \subseteq T\) for all \(t \in T\) and \(\alpha \in A(t)\).

Clearly, \(\operatorname{Pr}^{\mathscr{G}}(t \models \square \neg B)=1\) for any state \(t \in T\) and any scheduler \(\mathfrak{S}\) which only selects actions from \(A(t)\) for state sequences \(t_{0} t_{1} \ldots t_{n}\) with \(t_{n}=t\). Vice versa, by induction on the number of iterations, it can be shown that \(\operatorname{Pr}^{\max }(s \models \square \neg B)<1\) for all states \(s\) that have been removed during the elimination process. Hence, indeed:
\[
T=\left\{s \in S \mid \operatorname{Pr}^{\max }(t \mid=\square \neg B)=1\right\}
\]

We conclude that
\[
S_{=1}^{\min }=\{s \in S \mid s \not \vDash \exists(\neg B) \cup T\}=S \backslash \operatorname{Sat}(\exists((\neg B) \cup T))
\]
and obtain the following theorem:

\section*{Theorem 10.112. Qualitative Analysis for Min Reachability Problems}

The sets \(S_{=0}^{\min }\) and \(S_{=1}^{\min }\) can be computed using graph-based algorithms in time \(\mathcal{O}(\) size \((\mathcal{M}))\).

Using similar techniques as in Lemma 10.102 (page 852), the above equation system can be used to show the existence of a memoryless scheduler \(\mathfrak{S}\) that minimizes the probabilities of reaching \(B\) for all states.

\section*{Lemma 10.113. Existence of Optimal Memoryless Schedulers}

Let \(\mathcal{M}\) be a finite \(M D P, B \subseteq S\), and \(s \in S\). There exists a memoryless scheduler \(\mathfrak{S}\) that minimizes the probabilities of eventually reaching \(B\), i.e., for all states s:
\[
\operatorname{Pr}^{\mathfrak{G}}(s \models \diamond B)=\operatorname{Pr}^{\min }(s \models \diamond B) .
\]

Furthermore, similar to the case for maximal reachability probabilities, the above equation system can be rewritten into a linear program. This linear program is defined by
- If \(\operatorname{Pr}^{\min }(s \models \diamond B)=1\), then \(x_{s}=1\).
- If \(\operatorname{Pr}^{\min }(s \models \diamond B)=0\), then \(x_{s}=0\).
- If \(0<\operatorname{Pr}(s \mid \forall B)<1\), then \(0<x_{s}<1\) and for all actions \(\alpha \in \operatorname{Act}(s)\) :
\[
x_{s} \leqslant \sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{t}
\]
where \(\sum_{s \in S} x_{s}\) is maximal.

\section*{Remark 10.114. Constrained Reachability}

The techniques for the events \(\diamond B\) or \(\diamond^{\leqslant n} B\) are also applicable to treat the constrained reachability properties \(C \cup B\) or \(C \cup \leqslant n B\). This works as follows. As a first step, all states in \(s \in S \backslash(C \cup B)\) are made absorbing. That is, their enabled actions are replaced by a single action, \(\alpha_{s}\), say, with \(\mathbf{P}\left(s, \alpha_{s}, s\right)=1\). In the thus obtained \(\operatorname{MDP}, \operatorname{Pr}^{*}(s \vDash \diamond B)\) or \(\operatorname{Pr}^{*}\left(s \models \diamond^{\leqslant n} B\right)\), respectively, are determined in the described manner. (Here, \(*\) is either \(\max\) or min.)

By Lemmas 10.102 and 10.113, there exists a memoryless scheduler that optimizes (i.e., maximizes or minimizes) the unbounded constrained reachability probabilities \(C \cup B\). For
the step-bounded event \(C \mathrm{U}^{\leqslant n} B\), finite-memory schedulers exist that yield the extreme probabilities.

In fact, the techniques sketched in this section provide the key ingredients for PCTL model checking of finite MDPs.

\subsection*{10.6.2 PCTL Model Checking}

In Section 10.2, the probabilistic computation tree logic (PCTL) has been introduced as a CTL-like branching time logic for Markov chains. Without changing the syntax, PCTL can also serve as a temporal logic to specify important properties of finite MDPs. The main (and only) difference with the setting for Markov chains is that the probabilistic operator \(\mathbb{P}_{J}(\cdot)\) ranges over all schedulers. Thus, \(\mathbb{P}_{J}(\varphi)\) asserts that the probability bounds given by \(J\) for the event specified by \(\varphi\) are satisfied, regardless of the resolution of the nondeterminism.

The syntax of PCTL as a logic for MDPs is the same as for Markov chains, see Definition 10.36 on page 781. For an \(\operatorname{MDP} \mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\), the satisfaction relation for PCTL state- and path formulae is defined as follows:
\[
\begin{array}{llll}
s & =\text { true } & & \\
s & \models a & \text { iff } & a \in L(s) \\
s & =\neg \Phi & & \text { iff } \\
& s \not \models \Phi \\
s & =\Phi_{1} \wedge \Phi_{2} & & \text { iff }
\end{array} \quad s \models \Phi_{1} \text { and } s \models \Phi_{2} .
\]

Here, \(\operatorname{Pr}^{\mathfrak{S}}(s \models \varphi)\) is a shorthand for \(\operatorname{Pr}_{s}^{\mathfrak{S}}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \varphi\}\). We write \(\operatorname{Sat}_{\mathcal{M}}(\Phi)\), or briefly \(\operatorname{Sat}(\Phi)\), for the satisfaction set of \(\Phi\) in \(\mathcal{M}\), i.e.,
\[
\operatorname{Sat}_{\mathcal{M}}(\Phi)=\{s \in S \mid s \models \Phi\} .
\]

The semantics of path formulae is exactly the same as for PCTL interpreted over Markov chains. Thus, the probabilistic operator \(\mathbb{P}_{J}(\cdot)\) imposes probability bounds for all schedulers. In particular, we have
- \(s \models \mathbb{P}_{\leqslant p}(\varphi)\) if and only if \(\operatorname{Pr}^{\max }(s \models \varphi) \leqslant p\), and
- \(s \models \mathbb{P}_{\geqslant p}(\varphi)\) if and only if \(\operatorname{Pr}^{\min }(s \models \varphi) \geqslant p\)
where \(\operatorname{Pr}^{\max }(s \models \varphi)=\sup _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{G}}(s \models \varphi)\) with \(\mathfrak{S}\) ranging over all possible schedulers for \(\mathcal{M}\). Similarly, \(\operatorname{Pr}^{\min }(s \models \varphi)\) denotes the infimum of the probability for the event specified by \(\varphi\) under all schedulers. For finite MDPs, the same holds for strict lower and upper probability bounds (i.e., \(<p\) and \(>p\) ), since for any PCTL path formula \(\varphi\) there exists a finite-memory scheduler that maximizes or minimizes the probabilities for \(\varphi\) (see Remark 10.114 on page 865). Thus for finite MDPs:
\[
\operatorname{Pr}^{\max }(s \models \varphi)=\max _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}(s \models \varphi) \text { and } \operatorname{Pr}^{\min }(s \models \varphi)=\min _{\mathfrak{S}} \operatorname{Pr}^{\mathfrak{S}}(s \models \varphi)
\]

The always operator can be derived as in the setting of Markov chains. For instance, \(\mathbb{P}_{\leqslant p}(\square \Phi)\) can be defined as \(\mathbb{P}_{\geqslant 1-p}(\diamond \neg \Phi)\). Then:
\[
s \models \mathbb{P}_{\leqslant p}(\square \Phi) \quad \text { iff } \quad \operatorname{Pr}^{\mathfrak{G}}(s \models \square S a t(\Phi)) \leqslant p \text { for all schedulers } \mathfrak{S} .
\]

For example, for the mutual exclusion protocol with a randomized arbiter, the PCTL specification
\[
\mathbb{P}_{=1}\left(\square\left(\neg \text { crit }_{1} \vee \neg c r i t_{2}\right)\right) \wedge \bigwedge_{i=1,2} \mathbb{P}_{=1}\left(\square\left(\text { wait }_{i} \rightarrow \mathbb{P}_{\geqslant \frac{7}{8}}\left(\diamond \leqslant 9 \text { crit }_{i}\right)\right)\right)
\]
asserts the mutual exclusion property (first conjunct) and that every waiting process will enter its critical section within the next nine steps with at least probability \(\frac{7}{8}\) (regardless of the scheduling policy).

Equivalence of PCTL Formulae To distinguish the equivalence relations on PCTL state formulae that result from the Markov chain semantics and the MDP semantics, we use the notations \(\equiv_{\text {MDP }}\) and \(\equiv_{\text {MC }}\), respectively. The relation \(\equiv_{\text {MDP }}\) denotes the equivalence of PCTL formulae when interpreted over Markov decision processes. That is, \(\Phi \equiv_{\text {MDP }} \Psi\) if and only if for all MDPs \(\mathcal{M}\), it holds that \(\operatorname{Sat}_{\mathcal{M}}(\Phi)=\operatorname{Sat}_{\mathcal{M}}(\Psi)\). Similarly, \(\Phi \equiv_{\text {MC }} \Psi\) if and only if for all Markov chains \(\mathcal{M}, \operatorname{Sat}_{\mathcal{M}}(\Phi)=\operatorname{Sat}_{\mathcal{M}}(\Psi)\). Since any Markov chain can be viewed as an MDP where each state has a single enabled action, \(\equiv_{\text {MDP }}\) is finer than \(\equiv_{\mathrm{MC}}\). That is:
\[
\Phi \equiv_{\mathrm{MDP}} \Psi \quad \text { implies } \quad \Phi \equiv_{\mathrm{MC}} \Psi
\]

The converse, however, does not hold. This can be shown as follows. Consider a formula of the form \(\mathbb{P}_{\leqslant p}(\varphi)\) with \(0 \leqslant p<1\) and assume \(\varphi\) is a satisfiable, but not valid, path formula such as \(\bigcirc a\) or \(a \cup b\). Then:
\[
\mathbb{P}_{\leqslant p}(\varphi) \equiv_{\mathrm{MC}} \neg \mathbb{P}_{>p}(\varphi)
\]

This equivalence is trivial since the probability for an event \(E\) in a Markov chain is \(\leqslant p\) if and only if the probability for \(E\) is not \(>p\). An analogous argument is not applicable to Markov decision processes, as universal quantification is inherent in the semantics of the probabilistic operator \(\mathbb{P}_{J}(\cdot)\) :
\[
\begin{array}{lll}
s \models \mathbb{P}_{\leqslant p}(\varphi) & \text { iff } \quad \operatorname{Pr}^{\mathfrak{G}}(s \models \varphi) \leqslant p \text { for all schedulers } \mathfrak{S} \\
s \models \neg \mathbb{P}_{>p}(\varphi) & \text { iff } & \operatorname{not}\left(\operatorname{Pr}^{\mathfrak{S}}(s \models \varphi)>p \text { for all schedulers } \mathfrak{S}\right) \\
& \text { iff } & \operatorname{Pr}^{\mathfrak{S}}(s \models \varphi) \leqslant p \text { for some scheduler } \mathfrak{S}
\end{array}
\]

Hence, \(\mathbb{P}_{\leqslant p}(\varphi) \not \equiv_{\mathrm{MDP}} \neg \mathbb{P}_{>p}(\varphi)\). Nonetheless, several equivalences established for MCs also hold for MDPs, such as, e.g.,
\[
\mathbb{P}_{] p, q]}(\varphi) \equiv_{\mathrm{MDP}} \mathbb{P}_{>p}(\varphi) \wedge \mathbb{P}_{\leqslant q}(\varphi)
\]

Let us briefly consider the relationship between the qualitative fragment of PCTL and CTL. The qualitative fragment for MDPs contains four qualitative properties built by the probabilistic operators:
\[
\mathbb{P}_{=1}(\varphi) \text { and } \mathbb{P}_{>0}(\varphi) \text { and } \mathbb{P}_{<1}(\varphi) \text { and } \mathbb{P}_{=0}(\varphi)
\]

As opposed to the setting for Markov chains, these operators cannot be derived from each other. Thus, the syntax of the qualitative fragment of PCTL has to be extended with \(\mathbb{P}_{<1}(\varphi)\) and \(\mathbb{P}_{=0}(\varphi)\). As for Markov chains, the formulae \(\mathbb{P}_{=1}(\square a)\) and \(\forall \square a\) are equivalent. The same applies to \(\mathbb{P}_{>0}(a \cup b)\) and \(\exists(a \cup b)\). The qualitative PCTL formulae \(\mathbb{P}_{>0}(\square a)\) and \(\mathbb{P}_{=1}(\diamond a)\) are not definable in CTL and, vice versa, the CTL formulae \(\exists \square a\) and \(\forall \diamond a\) cannot be specified by a qualitative PCTL formula.

PCTL Model Checking Given a finite MDP \(\mathcal{M}\) and a PCTL state formula \(\Phi\), to check whether all initial states \(s\) satisfy \(\Phi\), an adapted version of the PCTL model-checking algorithm for Markov chains can be employed. That is, as for other CTL-like branching time logics, one successively computes the satisfaction sets \(S a t(\Psi)\) for the state subformulae of \(\Phi\). The only difference from PCTL model checking for MCs is the treatment of the probabilistic operator. Consider, e.g., \(\Psi=\mathbb{P}_{\leqslant p}\left(\bigcirc \Psi^{\prime}\right)\). The model checker determines the values
\[
x_{s}=\operatorname{Pr}^{\max }\left(s \models \bigcirc \Psi^{\prime}\right)=\max \left\{\sum_{t \in \operatorname{Sat}\left(\Psi^{\prime}\right)} \mathbf{P}(s, \alpha, t) \mid \alpha \in \operatorname{Act}(s)\right\}
\]
and returns \(\operatorname{Sat}(\Psi)=\left\{s \in S \mid x_{s} \leqslant p\right\}\). For until formulae \(\Psi=\mathbb{P}_{\leqslant p}\left(\Psi_{1} \cup \Psi_{2}\right)\) and \(\Psi=\) \(\mathbb{P}_{\leqslant p}\left(\Psi_{1} \mathrm{U} \leqslant n \Psi_{2}\right)\), we apply the techniques explained in the previous section to compute \(x_{s}=\operatorname{Pr}^{\max }(s \vDash C \cup B)\) or \(x_{s}=\operatorname{Pr}^{\max }\left(s \vDash C \cup \cup^{\leqslant n} B\right)\). This can be done by either solving a linear program or by means of value iteration. The model checker returns \(\operatorname{Sat}(\Psi)=\left\{s \in S \mid x_{s} \leqslant p\right\}\). The treatment of strict upper probability bounds \(<p\) or lower probability bounds \((\geqslant p\) or \(>p)\) is similar. Formulae \(\mathbb{P}_{J}(\cdot)\) with probability intervals \(J \subseteq[0,1]\) different from \([0, p[,[0, p]] p, 1],,[p, 1]\) can be treated by splitting \(\mathbb{P}_{J}(\cdot)\) into a conjunction of a formula with an upper bound and a formula with a lower probability bound.

The overall time complexity of this approach is linear in the length of the formula \(\Phi\) and the maximal step-bound \(n_{\max }\) of a subformula \(\mathbb{P}_{J}\left(\Psi_{1} \mathrm{U}^{\leqslant n} \Psi_{2}\right)\) of \(\Phi\) and polynomial in the size of \(\mathcal{M}\), provided that a linear program solver with polynomial worst-case running time is used.

\section*{Theorem 10.115. Time Complexity of PCTL Model Checking for MDPs}

For finite MDP \(\mathcal{M}\) and PCTL formula \(\Phi\), the PCTL model-checking problem \(\mathcal{M} \vDash \Phi\) can be determined in time
\[
\mathcal{O}\left(\operatorname{poly}(\operatorname{size}(\mathcal{M})) \cdot n_{\max } \cdot|\Phi|\right)
\]
where \(n_{\max }\) is the maximal step bound that appears in a sub-path formula \(\Psi_{1} \cup \leqslant n \Psi_{2}\) of \(\Phi\) (and \(n_{\max }=1\) if \(\Phi\) does not contain a step-bounded until operator).

For CTL we explained the concept of witnesses and counterexamples for state formulae \(\exists \varphi\) and \(\forall \varphi\); see Section 6.6 (page 373). Witnesses and counterexamples have been defined as "sufficiently long" initial finite path fragments \(\widehat{\pi}\) that can be extended to a path \(\pi\) where \(\pi \vDash \varphi\) (witness) and \(\pi \not \vDash \varphi\) (counterexample). In the probabilistic case, the situation is analogous, but now memoryless (or finite-memory) schedulers play the role of the finite path fragments. If, e.g., \(s \not \vDash \mathbb{P}_{\leqslant p}\left(\Psi_{1} \cup \Psi_{2}\right)\), then \(\operatorname{Pr}^{\max }\left(s \neq \Psi_{1} \cup \Psi_{2}\right)>p\) and there exists a memoryless scheduler \(\mathfrak{S}\) such that \(\operatorname{Pr}^{\mathfrak{G}}\left(s \models \Psi_{1} \cup \Psi_{2}\right)>p\). Note that such a memoryless scheduler is computed implicitely by the algorithms presented for the computation of \(\operatorname{Pr}^{\max }\left(s \vDash \Psi_{1} \cup \Psi_{2}\right)>p\). This scheduler \(\mathfrak{S}\) can be viewed as a counterexample for \(s \not \vDash \mathbb{P}_{\leqslant p}\left(\Psi_{1} \cup \Psi_{2}\right)\). Of course, it suffices to represent only some information on \(\mathfrak{S}\), e.g., S's decisions on states where \(\Psi_{1} \wedge \neg \Psi_{2}\) holds. Counterexamples for other probability bounds ( \(\geqslant p,>p\) or \(<p\) ) and other path formulae can be obtained in an analogous way. Memoryless schedulers are sufficient as long as the step-bounded until operator is not used. For \(s \not \vDash \mathbb{P}_{J}\left(\Psi_{1} U^{\leqslant n} \Psi_{2}\right)\) a counterexample can be obtained by a finite-memory scheduler \(\mathfrak{S}\) where \(\operatorname{Pr}^{\mathfrak{S}}\left(s \models \Psi_{1} \cup^{\leqslant n} \Psi_{2}\right) \notin J\). The concept of witnesses can be realized for formulae of the form \(\neg \mathbb{P}_{J}(\varphi)\), since they require the existence of a scheduler where the probability for \(\varphi\) is not in \(J\). In fact, any memoryless (or finite-memory) scheduler \(\mathfrak{S}\) with \(\operatorname{Pr}^{\mathfrak{S}}(s \models \varphi) \notin J\) constitutes a proof for \(s \models \neg \mathbb{P}_{J}(\varphi)\). Again, memoryless schedulers are sufficient for PCTL without the step-bounded until operator.

\subsection*{10.6.3 Limiting Properties}

Recall that for analyzing finite MCs against liveness properties, the key observation is that almost surely a BSCC will be reached whose states are visited infinitely often; see Theorem 10.27 (page 775). An analogous result can be established for finite MDPs by means of so-called end components. An end component is a sub-MDP that is closed under probabilistic choices and that is strongly connected.

\section*{Definition 10.116. Sub-MDP}

Let \(\mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov decision process. A sub-MDP of \(\mathcal{M}\) is a pair \((T, A)\) where \(\varnothing \neq T \subseteq S\) and \(A: T \rightarrow 2^{A c t}\) is a function such that:
- \(\varnothing \neq A(s) \subseteq \operatorname{Act}(s)\) for all states \(s \in T\),
- \(s \in T\) and \(\alpha \in A(s)\) implies \(\operatorname{Post}(s, \alpha)=\{t \in S \mid \mathbf{P}(s, \alpha, t)>0\} \subseteq T\).

A sub-MDP \(\left(T^{\prime}, A^{\prime}\right)\) is said to be contained in a sub-MDP \((T, A)\) if \(T^{\prime} \subseteq T\) and \(A^{\prime}(t) \subseteq A(t)\) for all states \(t \in T\).

The digraph \(G_{(T, A)}\) induced by a sub-MDP \((T, A)\) is the directed graph with the vertex set
\[
T \cup\{\langle s, \alpha\rangle \in T \times A c t \mid \alpha \in A(s)\}
\]
and the edges \(s \rightarrow\langle s, \alpha\rangle\) for \(s \in S\) and \(\alpha \in A(s)\) and \(\langle s, \alpha\rangle \rightarrow t\) for all \(t \in \operatorname{Post}(s, \alpha)\).

\section*{Definition 10.117. End Component}

An end component of MDP \(\mathcal{M}\) is a sub-MDP \((T, A)\) such that the digraph \(G_{(T, A)}\) induced by \((T, A)\) is strongly connected.

Let \(\operatorname{EC}(\mathcal{M})\) denote the set of end components of \(\mathcal{M}\).

Example 10.118. End Component
Consider the MDP \(\mathcal{M}\) of Figure 10.21. The sub-MDP \((T, A)\) with \(T=\left\{s_{5}, s_{6}\right\}\) and \(A\left(s_{6}\right)=A\left(s_{5}\right)=\{\alpha\}\) is an end component of \(\mathcal{M}\). The digraph \(G_{(T, A)}=(V, E)\) where \(V=\left\{s_{5},\left\langle s_{5}, \alpha\right\rangle, s_{6},\left\langle s_{6}, \alpha\right\rangle\right\}\) where \(s_{5} \rightarrow\left\langle s_{5}, \alpha\right\rangle,\left\langle s_{5}, \alpha\right\rangle \rightarrow s_{6}, s_{6} \rightarrow\left\langle s_{6}, \alpha\right\rangle\) and \(\left\langle s_{6}, \alpha\right\rangle \rightarrow\) \(s_{6}\) and \(\left\langle s_{6}, \alpha\right\rangle \rightarrow s_{5}\).

For each end component \((T, A)\) there is a scheduler, even a finite-memory one, that almost surely enforces staying forever in \(T\) while visiting all states in \(T\) infinitely often.

Lemma 10.119. Recurrence Property of End Components
For end component \((T, A)\) of finite \(M D P \mathcal{M}\), there exists a finite-memory scheduler \(\mathfrak{S}\) such that for any \(s \in S\) :
\[
\operatorname{Pr}^{\mathfrak{S}}\left(s \mid \square T \wedge \bigwedge_{t \in T} \square \diamond t\right)=1 .
\]

Proof: The goal is to define an fm-scheduler \(\mathfrak{S}\) that schedules all actions \(\alpha \in A(s)\) infinitely often, but never an action \(\beta \notin A(s)\). For \(s \in T\) let
\[
A(s)=\left\{\alpha_{0}^{s}, \ldots, \alpha_{k_{s}-1}^{s}\right\}
\]

The modes of \(\mathfrak{S}\) are the functions \(q: T \rightarrow \mathbb{N}\) such that \(0 \leqslant q(s)<k_{s}\). The scheduler \(\mathfrak{S}\) selects the action in the current state \(s\) according to a round-robin policy. That is, if \(\alpha_{i}\) is the last selected action in state \(s\), the next action to be selected is \(\alpha_{i+1}\) (modulo \(k_{s}\) ). Formally:
\[
\operatorname{act}(q, s)=\alpha_{i} \quad \text { where } \quad i=q(s) \bmod k_{s}
\]

The next mode function is given by \(\Delta(q, s)=p\) where \(p(t)=q(t)\) for \(t \in T \backslash\{s\}\) and \(p(s)=(q(s)+1) \bmod k_{s}\). The starting modes are given by start \((\cdot)=q_{0}\) where \(q_{0}(s)=0\) for all \(s \in T\). The decisions of \(\mathfrak{S}\) for states outside \(T\) are irrelevant. For instance, \(\operatorname{act}(q, s)=\alpha_{s}, \Delta(q, s)=q\) for all modes \(q\) and states \(s \notin T\).

Since \(\mathfrak{S}\) only selects actions in \(A(s)\) for \(s \in T\), all \(\mathfrak{S}\)-paths that start in a state \(s \in T\) never visit a state outside \(T\). As the digraph induced by \((T, A)\) is strongly connected, the (finite) MC induced by \(\mathfrak{S}\) consists of a single BSCC. By Theorem 10.27 (page 775), almost surely any state in \(\mathcal{M}_{\mathfrak{S}}\) will be visited infinitely often.

For an infinite path
\[
\pi=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} s_{2} \xrightarrow{\alpha_{3}} \ldots
\]
in a finite MDP \(\mathcal{M}\), the limit of \(\pi\), denoted \(\operatorname{Limit}(\pi)\), is defined as the pair \((T, A)\) where \(T\) is the set of states that are visited infinitely often in \(\pi\) :
\[
T=\inf \left(s_{0} s_{1} s_{2} \ldots\right)=\left\{s \in S \mid \not{ }_{\exists} n \geqslant 0 . s_{n}=s\right\}
\]
and \(A: T \rightarrow\) Act, the function that assigns to each state \(s \in T\) the set of actions that are taken infinitely often in \(s\), i.e., for \(s \in T\) :
\[
A(s)=\left\{\alpha \in \operatorname{Act}(s) \mid \exists n \geqslant 0 . s_{n}=s \wedge \alpha_{n+1}=\alpha\right\} .
\]

The following theorem is the MDP analogue to Theorem 10.27 (page 775). It states that under each scheduler the limit of almost all paths constitutes an end component.

\section*{Theorem 10.120. Limiting Behavior of MDPs}

For each state s of a finite MDP \(\mathcal{M}\) and scheduler \(\mathfrak{S}\) for \(\mathcal{M}\) :
\[
\operatorname{Pr}_{s}^{\mathscr{G}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi) \in E C(\mathcal{M})\}=1
\]

Proof: Let \(\mathcal{M}\) be a finite \(\operatorname{MDP}\) and \(\pi\) an infinite path starting in \(s\). Given that \(\mathcal{M}\) is finite, there is a state \(t\), say, and an action \(\alpha\), say, such that \(\pi\) visits \(t\) and leaves it via \(\alpha\) infinitely often. That is, the event \(\square \diamond(t, \alpha)\) holds. Consider an \(\alpha\)-successor \(u\) of \(t\), i.e., \(p=\mathbf{P}(t, \alpha, u)>0\). Given \(\square \diamond(t, \alpha)\), the probability of entering \(u\) finitely often is zero. This follows from the fact that the probability of not taking the transition \(t \xrightarrow{\alpha} u\) from some moment on is bounded by \(\lim _{n \rightarrow \infty}(1-p)^{n}=0\). Thus, almost surely, \(u\) is entered infinitely often via the edge \(t \xrightarrow{\alpha} u\). Thus, for each scheduler \(\mathfrak{S}\) and almost all \(\mathfrak{S}\)-paths \(\pi\) it holds that: if \(\operatorname{Limit}(\pi)=(T, A), t \in T, \alpha \in A(t)\) and \(\mathbf{P}(t, \alpha, u)>0\), then \(u \in T\). The claim now follows from the fact that the underlying graph of \(\operatorname{Limit}(\pi)\) is strongly connected.

Let \(\mathcal{M}\) be a finite MDP and \(P \subseteq\left(2^{A P}\right)^{\omega}\) an LT property which only depends on the labelings that are repeated infinitely often. For example, \(P\) could stand for a repeated reachability property \(\square \diamond a\), or Boolean combinations of repeated reachability properties, such as a persistence property \(\diamond \square b\), a strong fairness condition \(\bigwedge_{1 \leqslant i \leqslant k}\left(\square \diamond a_{i} \rightarrow \square \diamond b_{i}\right)\), or a Rabin condition \(\bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square a_{i} \wedge \square \diamond b_{i}\right)\) where \(a, a_{i}, b\), and \(b_{i}\) are atomic propositions. LT properties \(P\) for which the satisfaction only depends on the labelings that appear infinitely often, but not on their order, are called limit LT properties. Let \(\inf \left(A_{0} A_{1} A_{2} \ldots\right)=\{A \subseteq\) \(\left.A P \mid \forall i \geqslant 0 . \exists j \geqslant i . A_{i}=A\right\}\) denote the set of labelings that appear infinitely often in the word \(A_{0} A_{1} A_{2} \ldots \in\left(2^{A P}\right)^{\omega}\).

\section*{Notation 10.121. Limit LT Property}

LT property \(P\) over \(A P\) is a limit \(L T\) property if for all words \(\sigma, \sigma^{\prime} \in\left(2^{A P}\right)^{\omega}\) :
\[
\sigma \in P \text { and } \inf (\sigma)=\inf \left(\sigma^{\prime}\right) \quad \text { implies } \quad \sigma^{\prime} \in P
\]

For limit LT properties, the satisfaction relation for the paths in a finite MDP \(\mathcal{M}\) can be expressed by conditions on sets of states. Let \(S\) be the state space of \(\mathcal{M}\) and \(T \subseteq S\). Let
\[
T \models P \quad \text { iff } \quad \forall \sigma \in\left(2^{A P}\right)^{\omega} \cdot \inf (\sigma)=L(T) \text { implies } \quad \sigma \in P
\]
where \(L(T)=\left\{L(t) \in 2^{A P} \mid t \in T\right\}\). If \(P\) is given by an LTL formula \(\varphi\), i.e., \(P=\) \(\operatorname{Words}(\varphi)\), then we write \(T \models \varphi\) instead of \(T \models P\).

Note that
\[
T \models P \quad \text { implies } \quad \text { for all paths } \pi: \inf (\pi)=T \text { implies } \operatorname{trace}(\pi) \in P
\]

By Theorem 10.120, only the sets \(\inf (\pi)=T\) where \(T\) is the state set of an end component of \(\mathcal{M}\) are relevant when analyzing the probabilities for a limit LT property \(P\). An end components \((T, A)\) is said to be accepting (for \(P\) ) iff \(T \models P\).

This permits establishing a reduction from the probabilistic reachability problem to the problem of computing extreme probabilities for limit LT properties. Let \(U_{P}\) denote the union of the sets \(T\) of all end components \((T, A)\) of \(\mathcal{M}\) such that \(T \models P\). The set \(U_{P}\) is also called the success set of \(P\) in \(\mathcal{M}\). Similarly, let \(V_{P}\) be the union of the sets \(T\) of all end components \((T, A)\) of \(\mathcal{M}\) such that \(\neg(T \models P)\).

\section*{Theorem 10.122. Verifying Limit LT Properties}

Let \(\mathcal{M}\) be a finite \(M D P\) and \(P\) a limit LT property. For any state s of \(\mathcal{M}\) :
(a) \(\operatorname{Pr}^{\max }(s \models P)=\operatorname{Pr}^{\max }\left(s \models \diamond U_{P}\right)\), and
(b) \(\operatorname{Pr}^{\min }(s \models P)=1-\operatorname{Pr}^{\max }\left(s \models \diamond V_{P}\right)\).

Furthermore, there exist finite-memory schedulers \(\mathfrak{S}_{\max }\) and \(\mathfrak{S}_{\text {min }}\) such that for any state s of \(\mathcal{M}\) :
(c) \(\operatorname{Pr}^{\max }(s \models P)=\operatorname{Pr}^{\mathfrak{S}_{\text {max }}}(s \models P)\) and \((d) \operatorname{Pr}^{\min }(s \models P)=\operatorname{Pr}^{\mathfrak{S}_{\text {min }}}(s \models P)\).

Proof: First consider the statement for maximal probabilities. For each scheduler \(\mathfrak{S}\) we have \(\operatorname{Pr}^{\mathfrak{S}}(s \models P) \leqslant \operatorname{Pr}^{\mathfrak{S}}\left(s \models \diamond U_{P}\right)\). By Theorem 10.120 it holds that
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models P)=\operatorname{Pr}^{\mathfrak{S}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi) \in E C(\mathcal{M}) \wedge \inf (\pi) \models P\}
\]

By definition of \(U_{P}, \pi \vDash \diamond U_{P}\) for each path \(\pi\) with \(\operatorname{Limit}(\pi)\) is an end component and \(\inf (\pi) \models P\).

Vice versa, there exists a finite-memory scheduler such that \(\operatorname{Pr}^{\mathscr{G}}(s \models P)=\operatorname{Pr}^{\max }(s \models\) \(\left.\diamond U_{P}\right)\). To see this, consider a memoryless scheduler \(\mathfrak{S}_{0}\) that maximizes the probabilities of reaching \(U_{P}\) for all states \(s\) in \(\mathcal{M}\). In addition, for each end component \((T, A)\) let the fm-scheduler \(\mathfrak{S}_{(T, A)}\) be such that it ensures staying forever in \(T\) while visiting all states \(t \in T\) infinitely often, once started in some state in \(T\). By Lemma 10.119 (page 870), such an fm-scheduler does exist. Furthermore, for each state \(u \in U_{P}\), select an end component \(E C(u)=(T, A)\) with \(u \in T\) and \(T \models P\).

Let \(\mathfrak{S}\) be the scheduler which first mimics \(\mathfrak{S}_{0}\) until a state \(u\) in \(U_{P}\) has been reached. From this moment on, \(\mathfrak{S}\) behaves as \(\mathfrak{S}_{E C(u)}\). For this scheduler \(\mathfrak{S}\), almost all paths that eventually enter \(U_{P}\) will visit all states of an end component \((T, A)\) with \(T \models P\) infinitely
often. In particular, the condition \(\inf (\pi) \models P\) holds for almost all \(\mathfrak{S}\)-paths, provided they eventually reach \(U_{P}\). This yields
\[
\begin{aligned}
\operatorname{Pr}^{\mathfrak{S}}(s \models P) & =\sum_{u \in U_{P}} \operatorname{Pr}^{\mathfrak{S}_{1}}\left(s \models\left(\neg U_{P}\right) \cup u\right) \cdot \underbrace{\operatorname{Pr}^{\mathfrak{S}_{2}}(u \models P)}_{=1} \\
& =\operatorname{Pr}^{\text {max }}\left(s \models \diamond U_{P}\right) .
\end{aligned}
\]

Since \(\operatorname{Pr}^{\max }\left(s \models \diamond U_{P}\right)\) is an upper bound for the probabilities for \(P\) and starting state \(s\) under all schedulers, this yields the claim.

Statement (b) for minimal probabilities can be derived from (a) using the fact that the class of limit LT properties is closed under negation (i.e., with \(P\) also \(\bar{P}=\left(2^{A P}\right)^{\omega} \backslash P\) is a limit property) and that \(\operatorname{Pr}^{\mathfrak{S}}(s \models P)=1-\operatorname{Pr}^{\mathfrak{G}}(s \models \bar{P})\) for all schedulers \(\mathfrak{S}\). Hence:
\[
\operatorname{Pr}^{\min }(s \models P)=1-\operatorname{Pr}^{\max }(s \models \bar{P})
\]
and any fm-scheduler \(\mathfrak{S}\) that maximizes the probabilities for \(\bar{P}\) minimizes the probabilities for \(P\). For \(T \subseteq S\) we have \(T \models \bar{P}\) iff \(\neg(T \models P)\). Hence, the set \(V_{P}\) of all states \(t\) that are contained in some end component \((T, A)\) with \(\neg(T \models P)\) agrees with the set \(U_{\bar{P}}\) that arises by the union of all end components \((T, A)\) where \(T \models \bar{P}\). But then, (a) applied to \(\bar{P}\) yields
\[
\operatorname{Pr}^{\max }(s \models \bar{P})=\operatorname{Pr}^{\max }\left(s \models \diamond U_{\bar{P}}\right)=\operatorname{Pr}^{\max }\left(s \models \diamond V_{P}\right) .
\]

Hence, for finite MDPs the PCTL formula \(\mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right)\) asserts that under all schedulers the event \(\square \diamond a\) holds almost surely. This follows from the fact that
\[
s \models \mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\diamond a)\right)
\]
iff for all end components \((T, A)\) reachable from \(s\), it holds that \(T \cap \operatorname{Sat}(a) \neq \varnothing\)
iff \(\quad s \not \vDash \exists \diamond V_{P}\) with \(P=\square \diamond a\)
iff \(\operatorname{Pr}^{\text {min }}(s \models \square \diamond a)=1\).
Hence, almost sure repeated reachability is PCTL-definable for finite MDPs.
For certain limit LT properties, memoryless schedulers suffice to provide extreme probabilities. This holds, e.g., for repeated reachability properties \(\square \diamond B\); see Exercise 10.23 (page 905).

As for reachability properties, graph-based methods suffice to check qualitative limit LT properties. The simplest case is the question whether a limit LT property can be fulfilled with positive probability under some scheduler.

\section*{Corollary 10.123. Qualitative Limit LT Properties (Positive Probability)}

Let \(P\) be a limit LT property and \(\mathcal{M}\) a finite \(M D P\) and \(s\) a state in \(\mathcal{M}\). Then, the following statements are equivalent:
(a) \(\operatorname{Pr}^{\mathfrak{S}}(s \models P)>0\) for some scheduler \(\mathfrak{S}\),
(b) \(\operatorname{Pr}^{\max }(s \models P)>0\), and
(c) \(s \models \exists \diamond U_{P}\).

Checking whether \(\operatorname{Pr}^{\mathscr{S}}(s \models P)=1\) for some scheduler \(\mathfrak{S}\) (i.e., \(\operatorname{Pr}^{\max }(s \models P)=1\) ) amounts to verifying whether \(\operatorname{Pr}^{\max }\left(s \models \diamond U_{P}\right)=1\). This can be done using the techniques in Section 10.6.1 (see Algorithm 45 on page 859), provided \(U_{P}\) is given. The same holds for qualitative limit LT properties that refer to \(\operatorname{Pr}^{\min }(s \neq P)\). They can be treated either by algorithms to compute \(V_{P}\) in combination with the algorithms for minimal reachability probabilities, or-as shown in the proof of Theorem 10.122 (page 873)-by using the duality of maximal and minimal probabilities for limit LT properties.

The remainder of this section is focused on the computation of the success set \(U_{P}\) for limit LT property \(P\) in a finite MDP \(\mathcal{M}\). Recall that \(U_{P}\) is the union of the sets \(T\) of all end components \((T, A)\) such that \(T \models P\). Clearly, \(U_{P}\) results from an analysis of the set of all end components of \(\mathcal{M}\). However, the number of end components can be exponential in the size of \(\mathcal{M}\). This is due to the fact that end components may overlap, i.e., it is possible to have two end components \(\left(T_{1}, A_{1}\right)\) and \(\left(T_{2}, A_{2}\right)\), say, such that \(\left(T_{1}, A_{1}\right) \neq\left(T_{2}, A_{2}\right)\) and \(T_{1} \cap T_{2} \neq \varnothing\). However, for certain limit LT properties \(P\), the set \(U_{P}\) can be characterized by means of maximal end components, i.e., end components that are not properly contained in any other end component:

\section*{Notation 10.124. Maximal end components}

An end component \((T, A)\) of a finite MDP \(\mathcal{M}\) is called maximal if there is no end component \(\left(T^{\prime}, A^{\prime}\right)\) such that \((T, A) \neq\left(T^{\prime}, A^{\prime}\right)\) and \(T \subseteq T^{\prime}\) and \(A(s) \subseteq A^{\prime}(s)\) for all \(s \in T\).

Let \(\operatorname{MEC}(\mathcal{M})\) denote the set of all maximal end components in \(\mathcal{M}\).

Any end component is contained in exactly one maximal end component. This follows from the fact that the union of end components \(\left(T_{1}, A_{1}\right)\) and \(\left(T_{2}, A_{2}\right)\) with \(\left(T_{1}, A_{1}\right) \neq\left(T_{2}, A_{2}\right)\) and \(T_{1} \cap T_{2} \neq \varnothing\) is an end component. Here, the union of sub-MDPs \(\left(T_{1}, A_{1}\right)\) and \(\left(T_{2}, A_{2}\right)\) is the sub-MDP \(\left(T_{1} \cup T_{2}, A_{1} \cup A_{2}\right)\) where \(A_{1} \cup A_{2}\) denotes the function \(T_{1} \cup T_{2} \rightarrow 2^{\text {Act }}\) such that \(t \mapsto A_{1}(t) \cup A_{2}(t)\) if \(t \in T_{1} \cap T_{2}\) and \(t \mapsto A_{1}(t)\) if \(t \in T_{1} \backslash T_{2}\) and \(t \mapsto A_{2}(t)\)
if \(t \in T_{2} \backslash T_{1}\). Furthermore, maximal end components are pairwise disjoint. Hence, the number of maximal end components is bounded above by the number of states in \(\mathcal{M}\).

The set \(U_{P}\) for the limit LT property \(P\) results from the end components \((T, A)\) where \(T \models P\). In case \(P\) is a repeated reachability property, say \(\square \diamond B\) for some \(B \subseteq S\), then \(T \models P\) is equivalent to the requirement that \(T\) contains a \(B\)-state. If \(\left(T^{\prime}, A^{\prime}\right)\) is an end component with \(T^{\prime} \models \square \diamond B\), then \(T \models \square \diamond B\), where \((T, A)\) is the unique maximal end component that contains \(\left(T^{\prime}, A^{\prime}\right)\). Therefore, the success set \(U_{P}\) for the event \(P=\square \diamond B\) arises by the union of all maximal end components \((T, A)\) that contain at least one \(B\)-state. Formally:
\[
U_{\square \diamond B}=\bigcup_{\substack{(T, A) \in M E C(\mathcal{M}) \\ T \cap B \neq \varnothing}} T
\]

Note that an analogous result does not hold for persistence properties \(\diamond \square B\), since a non-maximal end component \(\left(T^{\prime}, A^{\prime}\right)\) with \(T^{\prime} \subseteq B\) may be contained in a maximal end component which also contains some states not in \(B\). However, \(U_{\Delta \square P}\) agrees with the union of all maximal end components \((T, A)\) where \(T \subseteq B\) in a slightly modified MDP \(\mathcal{M}_{\square B}\). The MDP \(\mathcal{M}_{\square B}\) results from \(\mathcal{M}\) by removing all states \(s \notin B\). More precisely, \(\mathcal{M}_{\square B}\) has the state space \(B \cup\{\) no \(\}\). The fresh state no is added as a trap state. The transition probabilities are defined by
\[
\mathbf{P}_{\square B}(s, \alpha, t)=\mathbf{P}(s, \alpha, t) \text { if } \alpha \in A c t, s \in B \text { and } \operatorname{Post}(s, \alpha) \subseteq B
\]

Let \(\mathbf{P}_{\square B}(s, \alpha, \cdot)=0\) for \(\alpha \in \operatorname{Act}(s)\) such that \(\operatorname{Post}(s, \alpha) \backslash B \neq \varnothing\). If state \(s\) in \(\mathcal{M}\) only has transitions to \(S \backslash B\), it has no enabled actions left over. For such states \(s\), let \(\mathbf{P}_{\square B}(s, \tau, n o)=1\). Furthermore, \(\mathbf{P}_{\square B}(n o, \tau, n o)=1\). Here, \(\tau\) is a pseudo action that has no further importance. The maximal end components of \(\mathcal{M}_{\square B}\) which do not contain no are end components \((T, A)\) of \(\mathcal{M}\) with \(T \subseteq B\). And vice versa, any end component \((T, A)\) of \(\mathcal{M}\) with \(T \subseteq B\) is contained in a maximal end component of \(\mathcal{M}_{\square B}\).

In fact, a similar technique is applicable for Rabin acceptance conditions:

\section*{Lemma 10.125. Success Set for Rabin Conditions}

Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S, B_{i}, C_{i} \subseteq S\). For a limit LT property \(P\) which-when interpreted on \(\mathcal{M}\)-is given by
\[
\bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square B_{i} \wedge \square \diamond C_{i}\right)
\]
it holds that
\[
U_{P}=\bigcup_{1 \leqslant i \leqslant k} U_{\square \diamond C_{i}}^{\mathcal{M}_{\square B_{i}}}
\]
where \(U_{\square \diamond B_{i}}^{\mathcal{M}_{\square}}\) is the success set of the event \(\square \diamond C_{i}\) in the MDP \(\mathcal{M}_{\square B_{i}}\), i.e., the set of all states \(t \in S\) that are contained in some maximal end component \((T, A)\) of \(\mathcal{M}_{\square B_{i}}\) such that no \(\notin T\) and \(T \cap C_{i} \neq \varnothing\).

The computation of the maximal end components of a finite MDP \(\mathcal{M}\) can be performed by means of iterative computations of SCCs. The idea is to successively remove all states and actions that are not contained in some end component. In the first iteration, the nontrivial SCCs \(T_{1}, \ldots, T_{k}\) are determined in the underlying graph of \(\mathcal{M}\). (A nontrivial SCC is an SCC which contains at least one edge, i.e., a cycle.) Then, for each state \(s \in T_{i}\), any action \(\alpha \in \operatorname{Act}(s)\) for which \(\operatorname{Post}(s, \alpha) \backslash T_{i} \neq \varnothing\), is removed from \(\operatorname{Act}(s)\). If \(\operatorname{Act}(s)\) becomes empty, state \(s\) is removed together with the actions \(\beta\) from \(\operatorname{Act}(t)\) with \((t, \beta) \in \operatorname{Pre}(s)\). This yields a sub-MDP \(\mathcal{M}_{1}\), say, of \(\mathcal{M}\) where the action set \(\operatorname{Act}(s)\) of each state \(s\) solely consists of actions \(\alpha\) such that all \(\alpha\)-successors of \(s\) belong to the SCC \(T_{i}\) of \(\mathcal{M}\) with \(s \in T_{i}\).

Due to the removal of actions, however, the strong connectivity of \(T_{i}\) as a vertex set of the underlying graph of \(\mathcal{M}_{1}\) might be lost. We therefore have to repeat the whole procedure. That is, we compute the nontrivial SCCs \(T_{1}^{1}, \ldots, T_{k_{1}}^{1}\) of \(\mathcal{M}_{1}\) and repeat the procedure described above for these SCCs. This yields a sub-MDP \(\mathcal{M}_{2}\), say of \(\mathcal{M}_{1}\). This procedure is repeated until a sub-MDP \(\mathcal{M}_{i}=\mathcal{M}^{\prime}\) of \(\mathcal{M}\) is obtained for which nontrivial SCCs agree with the maximal end components of \(\mathcal{M}\).

These steps are summarized in Algorithm 47. Here, if \(T \subseteq S\) and \(A: S \rightarrow 2^{\text {Act }}\) is a function, then \(\left.A\right|_{T}\) denotes the restriction of \(A\) to \(T\), i.e., \(\left.A\right|_{T}: T \rightarrow 2^{\text {Act }}\) is given by \(\left.A\right|_{T}(t)=A(t)\) for all \(t \in T\).

\section*{Lemma 10.126. Correctness of Algorithm 47}

For finite MDP \(\mathcal{M}\) with state space \(S\), Algorithm 47 returns \(\operatorname{MEC}(\mathcal{M})\) and requires at most \(|S|\) (outermost) iterations.

Proof: The termination of Algorithm 47 follows from the fact that in each iteration (except the last) of the repeat-loop, the partition induced by MEC is refined and covers at most the elements of the previous iteration. More precisely, let \(\mathrm{MEC}_{0}=\{S\}\) and let \(\mathrm{MEC}_{i}\) be the set MEC immediately after the \(i\) th iteration of the repeat loop. For fixed \(i\), the sets in \(\mathrm{MEC}_{i}\) are nonempty and pairwise disjoint. They constitute a partition of some subset \(S_{i}\) of \(S\). The sets \(S_{i}\) are decreasing, i.e., \(S_{0} \supseteq S_{1} \supseteq S_{2} \supseteq \ldots\), as for each \(i \geqslant 1\) and \(T \in \mathrm{MEC}_{i}\) there exists \(U \in \mathrm{MEC}_{i-1}\) such that \(T \subseteq U\). Furthermore, if \(T=U\), then in the \(i\) th iteration no action \(\alpha \in A(t)\) has been removed from the action-sets of any state \(t \in T\). This follows from the fact that the pairs \(\left(T,\left.A\right|_{T}\right)\) for \(T \in \mathrm{MEC}_{i}\) are sub-MDPs of \(\mathcal{M}\), i.e., \(\operatorname{Post}(t, \alpha) \subseteq T\) for all \(\alpha \in A(t)\) and \(t \in T\). Since the repeat loop terminates as soon as
```

Algorithm 47 Computing the maximal end components of a finite MDP
Input: finite MDP $\mathcal{M}$ with state space $S$
Output: the set $\operatorname{MEC}(\mathcal{M})$
for all $s \in S$ do $A(s):=\operatorname{Act}(s) ;$ od
MEC $:=\varnothing ; \quad$ MEC $_{\text {new }}:=\{S\} ;$
repeat
MEC := MEC ${ }_{\text {new }} ; \quad$ MEC $_{\text {new }}:=\varnothing ;$
for all $T \in \mathrm{MEC}$ do
$R:=\varnothing ; \quad \quad$ (* set of states to be removed *)
compute the nontrivial SCCs $T_{1}, \ldots, T_{k}$ of the digraph $G_{\left(T,\left.A\right|_{T}\right)}$;
for $i=1, \ldots, k$ do
for all states $s \in T_{i}$ do
$A(s):=\left\{\alpha \in A(s) \mid \operatorname{Post}(s, \alpha) \subseteq T_{i}\right\} ;$
if $A(s)=\varnothing$ then
$R:=R \cup\{s\} ;$
fi
od
od
while $R \neq \varnothing$ do
let $s \in R$
remove $s$ from $R$ and from $T$;
for all $(t, \beta) \in \operatorname{Pre}(s)$ with $t \in T$ do
$A(t):=A(t) \backslash\{\beta\} ;$
if $A(t)=\varnothing$ then
$R:=R \cup\{t\} ;$
fi
od
od
for $i=1, \ldots, k$ do
if $T \cap T_{i} \neq \varnothing$ then
MEC $_{\text {new }}:=$ MEC $_{\text {new }} \cup\left\{T \cap T_{i}\right\}$
${ }^{*}\left(T \cap T_{i},\left.A\right|_{T \cap T_{i}}\right)$ is a sub-MDP of $\left.\mathcal{M}^{*}\right)$
fi
od
od
until $\left(\right.$ MEC $=$ MEC $\left._{\text {new }}\right)$
return $\left\{\left(T,\left.A\right|_{T}\right) \mid T \in \operatorname{MEC}\right\}$

```
\(\mathrm{MEC}=\mathrm{MEC}_{\text {new }}\), for all iterations (except the last) there is at least one set \(U \in \mathrm{MEC}_{i-1}\) such that some \(u \in U\) is removed in the \(i\) th iteration.

After \(|S|\) iterations, \(\mathrm{MEC}_{|S|}\) would consist of singleton sets only and no further refinements would be possible. Hence, the maximal number of iterations of the repeat loop equals \(|S|\).

For the (partial) correctness of Algorithm 47, let us first observe that Algorithm 47 never removes states or actions that belong to some end component. That is, whenever \(\left(T^{\prime}, A^{\prime}\right)\) is an end component of \(\mathcal{M}\) then in each iteration of Algorithm 47, there exists \(T \in\) MEC such that \(\left(T^{\prime}, A^{\prime}\right)\) is contained in \(\left(T,\left.A\right|_{T}\right)\). This is a consequence of the following observation.

Whenever \((T, A)\) is a sub-MDP such that each end component \(\left(T^{\prime}, A^{\prime}\right)\) of \(\mathcal{M}\) with \(T \cap T^{\prime} \neq\) \(\varnothing\) is contained in \((T, A)\), then:
- For SCC \(C\) in \(G_{(T, A)}\) it holds that for any state \(s \in T\) and \(\alpha \in A(s)\) such that \(\operatorname{Post}(s, \alpha) \backslash C \neq \varnothing\), there is no end component \(\left(T^{\prime}, A^{\prime}\right)\) of \(\mathcal{M}\) such that \(s \in T^{\prime}\) and \(\alpha \in A^{\prime}(s)\). In particular, any end component \(\left(T^{\prime}, A^{\prime}\right)\) of \(\mathcal{M}\) with \(T^{\prime} \cap T \neq \varnothing\) is contained in the sub-MDP that results from \((T, A)\) by removing \(\alpha\) from \(A(s)\).
- If \(s \in T\) and \(A(s)\) becomes empty by removing all actions, then there is no end component of \(\mathcal{M}\) containing \(s\). In particular, any end component ( \(T^{\prime}, A^{\prime}\) ) of \(\mathcal{M}\) with \(T^{\prime} \cap T \neq \varnothing\) is contained in the sub-MDP that results from \((T, A)\) by removing state \(s\) and all actions \(\beta\) from \(A(t)\) where \(t \in T\) and \(\mathbf{P}(t, \beta, s)>0\).

Hence, the output of Algorithm 47 is the set of sub-MDPs \(\left(T_{1}, A_{1}\right), \ldots,\left(T_{k}, A_{k}\right)\) of \(\mathcal{M}\) where each end component \(\left(T^{\prime}, A^{\prime}\right)\) is contained in some \(\left(T_{i}, A_{i}\right)\).

On the other hand, since any \(\left(T_{i}, A_{i}\right)\) remains unchanged in the last iteration of the repeat loop, the graph \(G_{\left(T_{i}, A_{i}\right)}\) is strongly connected. Thus, \(\left(T_{i}, A_{i}\right)\) is an end component of \(\mathcal{M}\), and therefore a maximal end component of \(\mathcal{M}\). Hence, Algorithm 47 returns \(\operatorname{MEC}(\mathcal{M})\).

Let us now consider the worst-case time complexity of Algorithm 47. The SCCs of a digraph with \(N\) vertices and \(M\) edges can be computed in time \(\mathcal{O}(N+M)\). The cost of each iteration of the outermost loop is thus linear in the size of \(\mathcal{M}\). The number of iterations is bounded by \(|S|\) as shown above. Hence, the worst-case time complexity of Algorithm 47 is quadratic in the size of the MDP. More precisely, it is bounded above by
\[
\mathcal{O}(|S| \cdot(|S|+M))
\]
where \(M\) is the number of triples \((s, \alpha, t)\) such that \(\mathbf{P}(s, \alpha, t)>0\). This shows that the success set \(U_{P}\) of a limit LT property \(P\) which is given by a Rabin acceptance condi-
tion \(\bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square B_{i} \wedge \square \diamond C_{i}\right)\) (as in Lemma 10.125 on page 876 ) can be computed in time \(\mathcal{O}\left(\operatorname{size}(\mathcal{M})^{2} \cdot k\right)\). Thus:

\section*{Theorem 10.127. Time Complexity of Verifying Limit Rabin Properties}

Let \(\mathcal{M}\) be a finite \(M D P\) and \(P\) a limit LT property specified by a Rabin condition:
\[
\bigvee_{0<i \leqslant k}\left(\diamond \square B_{i} \wedge \square \diamond C_{i}\right)
\]

Then: the values \(\operatorname{Pr}^{\max }(s \models P)\) can be computed in time \(\mathcal{O}(\operatorname{poly}(\operatorname{size}(\mathcal{M})) \cdot k)\).

By duality, the same holds for limit LT properties \(P\) which are given by a strong fairness condition
\[
\bigwedge_{1 \leqslant i \leqslant k}\left(\square \diamond C_{i} \rightarrow \square \diamond D_{i}\right) \equiv \neg \bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square \neg D_{i} \wedge \square \diamond C_{i}\right)
\]
and the values \(\operatorname{Pr}^{\min }(s \neq P)\). Recall that by part (b) of Theorem 10.122 (page 873), we have:
\[
\operatorname{Pr}^{\min }(s \models P)=1-\operatorname{Pr}^{\max }\left(s \models \diamond V_{P}\right)
\]
where \(V_{P}\) is the union of the sets \(T\) of all end components \((T, A)\) such that \(\neg(T \models P)\). For a strong fairness condition as above we have \(\neg(T \models P)\) if and only if there exists \(i\) such that \(T \cap C_{i} \neq \varnothing\) and \(T \cap B_{i}=\varnothing\). Hence, \(V_{P}\) arises as the union of the maximal end components \((T, A)\) of \(\mathcal{M}_{\square \neg B_{i}}\) where \(T \cap C_{i} \neq \varnothing\). Here, \(\mathcal{M}_{\square \neg B_{i}}\) is the MDP obtained from \(\mathcal{M}\) by removing the states \(s \in S \backslash B_{i}\) (see page 876 for the precise definition).

\subsection*{10.6.4 Linear-Time Properties and PCTL*}

This section treats the quantitative verification of \(\omega\)-regular property \(P\) against an MDP \(\mathcal{M}\). This entails the computation of the values \(\operatorname{Pr}^{\min }(s \neq P)\) or, dually, \(\operatorname{Pr}^{\max }(s \models P)\) for state \(s\) in \(\mathcal{M}\). For example, let \(P\) describe the "good" behaviors and assume it is required to establish whether \(P\) holds in \(\mathcal{M}\) under all schedulers with some sufficiently large probability \(1-\varepsilon\), say. More precisely, the requirement is to show that
\[
\sum_{s \in S} \iota_{\text {init }}(s) \cdot \operatorname{Pr}^{\min }(s \models P) \geqslant 1-\varepsilon .
\]

Similarly, if \(P\) describes the bad behaviors, then a reasonable requirement is to verify whether \(P\) holds with some sufficiently small probability at most \(\varepsilon\) under all schedulers. That is:
\[
\sum_{s \in S} \iota_{\mathrm{init}}(s) \cdot \operatorname{Pr}^{\max }(s \models P) \leqslant \varepsilon
\]

The previous sections covered special cases of this setting such as reachability properties. For limit LT property \(P\) given by a Rabin condition, the values \(\operatorname{Pr}^{\max }(s \models P)\) can be obtained by computing the values \(\operatorname{Pr}^{\max }\left(s \vDash \diamond U_{P}\right)\). The set \(U_{P}\) is determined using graph algorithms, and the reachability probabilities \(\operatorname{Pr}^{\max }\left(s \models \diamond U_{P}\right)\) are obtained by (again) graph analysis followed by solving linear programs. This technique for limit LT properties can be generalized toward arbitrary \(\omega\)-regular properties in the following way.

Let \(P\) be an arbitrary \(\omega\)-regular property. As a first step, a deterministic Rabin automaton \(\mathcal{A}\) for \(P\) is constructed. Likewise as for Markov chains, the idea is to reduce the problem of computing \(\operatorname{Pr}^{\min }(s \models P\) ) (or its dual) to determining reachability probabilities in a product MDP. To that end, the product \(\operatorname{MDP} \mathcal{M} \otimes \mathcal{A}\) is considered (defined just below) and the maximal probabilities of reaching the success set \(U_{\mathcal{A}}\) in \(\mathcal{M} \otimes \mathcal{A}\) are computed. The success set \(U_{\mathcal{A}}\) depends on the Rabin acceptance condition in \(\mathcal{A}\).

\section*{Notation 10.128. Product MDP}

For finite \(\operatorname{MDP} \mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) and DRA \(\mathcal{A}=\left(Q, 2^{A P}, \delta, q_{0}, A c c\right)\) with \(A c c=\) \(\left\{\left(L_{1}, K_{1}\right), \ldots,\left(L_{k}, K_{k}\right)\right\}\), the product \(\mathcal{M} \otimes \mathcal{A}\) is an MDP
\[
\mathcal{M}^{\prime}=\left(S \times Q, A c t, \mathbf{P}^{\prime}, \iota_{\mathrm{init}}^{\prime}, Q, L^{\prime}\right)
\]
with
- \(\mathbf{P}^{\prime}\left(\langle s, q\rangle, \alpha,\left\langle s^{\prime}, q^{\prime}\right\rangle\right)= \begin{cases}\mathbf{P}\left(s, \alpha, s^{\prime}\right) & \text { if } q^{\prime}=\delta\left(q, L\left(s^{\prime}\right)\right) \\ 0 & \text { otherwise. }\end{cases}\)
- \(\iota_{\text {init }}^{\prime}(\langle s, q\rangle)= \begin{cases}\iota_{\text {init }}(s) & \text { if } q=\delta\left(q_{0}, L(s)\right) \\ 0 & \text { otherwise. }\end{cases}\)
- \(L^{\prime}(\langle s, q\rangle)=\{q\}\).

As for the product construction of a Markov chain and a DRA, there is a one-to-one correspondence between the path
\[
\pi=s_{0} \xrightarrow{\alpha_{1}} s_{1} \xrightarrow{\alpha_{2}} s_{2} \xrightarrow{\alpha_{3}} \ldots
\]
in the MDP \(\mathcal{M}\) and the path
\[
\pi^{+}=\left\langle s_{0}, q_{1}\right\rangle \xrightarrow{\alpha_{1}}\left\langle s_{1}, q_{2}\right\rangle \xrightarrow{\alpha_{2}}\left\langle s_{2}, q_{3}\right\rangle \xrightarrow{\alpha_{3}} \ldots
\]
in \(\mathcal{M} \otimes \mathcal{A}\) that starts in state \(\left\langle s_{0}, q_{1}\right\rangle\) where \(q_{1}=\delta\left(q_{0}, L\left(s_{0}\right)\right)\). Given a path \(\pi^{+}\)in \(\mathcal{M} \otimes \mathcal{A}\) the corresponding path in \(\mathcal{M}\) is simply obtained by omitting all automata states \(q_{i}\). Vice versa, given a path \(\pi\) as above, the corresponding path \(\pi^{+}\)is obtained by adding the automata states \(q_{i+1}=\delta\left(q_{i}, L\left(s_{i}\right)\right)\) to \(\pi\). Thus, \(\pi^{+}\)arises by combining \(\pi\) with the unique run for \(\operatorname{trace}(\pi)\) in the DRA \(\mathcal{A}\). In particular:
\[
\operatorname{trace}(\pi) \in P=\mathcal{L}_{\omega}(\mathcal{A})
\]
iff the run \(q_{0} q_{1} q_{2} q_{3} \ldots\) for \(\operatorname{trace}(\pi)\) in \(\mathcal{A}\) is accepting
\[
\text { iff } \quad \pi^{+} \models \underset{0<i \leqslant k}{\bigvee}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)
\]

In fact, this one-to-one correspondence on the path level induces a one-to-one correspondence for schedulers on \(\mathcal{M}\) and \(\mathcal{M} \otimes \mathcal{A}\). That is, any scheduler \(\mathfrak{S}\) for \(\mathcal{M}\) induces a scheduler \(\mathfrak{S}^{\prime}\) for \(\mathcal{M} \otimes \mathcal{A}\) such that for any \(\mathfrak{S}\)-path \(\pi\) in \(\mathcal{M}\) the corresponding path \(\pi^{+}\)in \(\mathcal{M} \otimes \mathcal{A}\) is a \(\mathfrak{S}^{\prime}\)-path, and vice versa. The scheduler \(\mathfrak{S}^{\prime}\) is obtained by simply ignoring the automata states, i.e.:
\[
\mathfrak{S}^{\prime}\left(\left\langle s_{0}, q_{1}\right\rangle\left\langle s_{1}, q_{2}\right\rangle \ldots\left\langle s_{n}, q_{n+1}\right\rangle\right)=\mathfrak{S}\left(s_{0} s_{1} \ldots s_{n}\right)
\]

We then have
\[
\operatorname{Pr}^{\mathfrak{S}}(s \models P)=\operatorname{Pr}^{\mathfrak{S}^{\prime}}\left(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle \models \bigvee_{0<i \leqslant k}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)\right)
\]

Vice versa, for a given scheduler \(\mathfrak{S}^{\prime}\) for \(\mathcal{M} \otimes \mathcal{A}\), the corresponding scheduler \(\mathfrak{S}\) for \(\mathcal{M}\) is obtained as follows. For history \(s_{0} s_{1} \ldots s_{n}\) in \(\mathcal{M}\), scheduler \(\mathfrak{S}\) chooses the same action as \(\mathfrak{S}^{\prime}\) selects for the history \(\left\langle s_{0}, q_{1}\right\rangle \ldots\left\langle s_{n}, q_{n+1}\right\rangle\) in \(\mathcal{M} \otimes \mathcal{A}\) (where \(q_{i+1}=\delta\left(q_{i}, L\left(s_{i}\right)\right.\) ) for \(0 \leqslant i \leqslant n)\).

Thus, there is a one-to-one-correspondence between the schedulers for \(\mathcal{M}\) and \(\mathcal{M} \otimes \mathcal{A}\). This correspondence preserves the finite-memory property-if \(\mathfrak{S}\) is finite memory, then \(\mathfrak{S}^{\prime}\) is also. In addition, the probabilities for \(P(\) in \(\mathcal{M})\) under \(\mathfrak{S}\) agree with the probabilities for \(\mathcal{A}\) 's acceptance condition \(\bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)\) under \(\mathfrak{S}^{\prime}\). But then, for all states \(s\) in \(\mathcal{M}\) :
\[
\begin{aligned}
\operatorname{Pr}_{\mathcal{M}}^{\max }(s \models P) & =\operatorname{Pr}_{\mathcal{M} \otimes \mathcal{A}}^{\max }\left(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle \models \underset{1 \leqslant i \leqslant k}{\bigvee}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)\right) \\
& =\operatorname{Pr}_{\mathcal{M} \otimes \mathcal{A}}^{\max }\left(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle \models \diamond U_{\mathcal{A}}\right)
\end{aligned}
\]
where \(U_{\mathcal{A}}\) denotes the success set of \(\mathcal{A}\) 's acceptance condition \(\bigvee_{1 \leqslant i \leqslant k}\left(\diamond \square \neg L_{i} \wedge \square \diamond K_{i}\right)\). Hence, the techniques presented in the previous section can be applied to compute \(\operatorname{Pr}^{\max }(s \models\) \(P)\). For the special case where the aim is to check whether \(\operatorname{Pr}^{\max }(s \models P)>0\), a graph analysis in \(\mathcal{M} \otimes \mathcal{A}\) suffices that checks whether \(\left\langle s, \delta\left(q_{0}, L(s)\right)\right\rangle\) can reach the success set \(U_{\mathcal{A}}\) in \(\mathcal{M} \otimes \mathcal{A}\).

The same techniques can be used to determine \(\operatorname{Pr}^{\min }(s \models P)\) by constructing a DRA for the complement \(\bar{P}\) of \(P\), i.e., \(\bar{P}=\left(2^{A P}\right)^{\omega} \backslash P\). As the class of \(\omega\)-regular properties is closed under complementation, \(\bar{P}\) is an \(\omega\)-regular property. Then:
\[
\operatorname{Pr}^{\min }(s \models P)=1-\operatorname{Pr}^{\max }(s \models \bar{P}) .
\]

In case the \(\omega\)-regular property \(P\) is given as LTL formula \(\varphi\), the worst-case time complexity of this technique to compute \(\operatorname{Pr}^{\max }(s \models \varphi)\) or \(\operatorname{Pr}^{\min }(s \models \varphi)\) is polynomial in the size of \(\mathcal{M}\), but double exponential in the length of \(\varphi\). (The double-exponential blowup is caused by the transformation from LTL to DRA.) From the complexity-theoretic point of view this algorithm is optimal, since the qualitative model-checking problem for MDPs-given a finite MDP \(\mathcal{M}\) and an LTL formula \(\varphi\) is \(\operatorname{Pr}^{\max }(s \models \varphi)=1\)-is in 2EXPTIME. This result is due to Courcoubetis and Yannakakis and stated here without proof [104].

\section*{Theorem 10.129.}

The qualitative model-checking problem for finite MDPs is in 2EXPTIME.

Recall that in the setting of Markov chains, this problem is PSPACE-complete.
As for Markov chains, the PCTL model-checking algorithm can be extended to treat the logic PCTL* by using the above techniques for computing extreme probabilities for LTL formulae. For PCTL* state formulae \(\mathbb{P}_{\leqslant p}(\varphi)\), one first recursively computes the satisfaction sets of the maximal state subformulae of \(\varphi\). These maximal state subformulae are replaced by new atomic propositions. This yields an LTL formula \(\varphi^{\prime}\), say. Subsequently, construct a DRA for \(\varphi^{\prime}\) and apply a quantitative analysis to compute \(\operatorname{Pr}^{\max }\left(s \models \varphi^{\prime}\right)\) for all states \(s\) in the MDP \(\mathcal{M}\). Then, \(\operatorname{Sat}\left(\mathbb{P}_{\leqslant p}(\varphi)\right)\) is the set of all states \(s\) in \(\mathcal{M}\) where \(\operatorname{Pr}^{\max }\left(s \models \varphi^{\prime}\right) \leqslant p\). The treatment of strict upper bounds \(<p\) is similar. For lower bounds \(\geqslant p\) or \(>p\), we have to compute \(\operatorname{Pr}^{\min }\left(s \models \varphi^{\prime}\right)\) for all states \(s\) in \(\mathcal{M}\). This yields a PCTL* model checking procedure that runs in time polynomial in the size of the MDP and double exponential in the length of the input PCTL* state formula.

\subsection*{10.6.5 Fairness}

This chapter is completed by discussing fairness assumptions in MDPs. Let us first remark that for each scheduler \(\mathfrak{S}\), almost all paths that visit state \(s\) infinitely often and take action \(\alpha\) in \(s\) infinitely often will visit each \(\alpha\)-successor of \(s\) infinitely often. This is similar to the setting of Markov chains, cf. Theorem 10.25 (page 772). (This fact has been already used in the proof of Theorem 10.120.) Thus, probabilistic choices are almost surely strongly fair. However, this does not address the resolution of the nondeterministic choices. As for
transition systems, fairness assumptions for the resolution of the nondeterministic choices are often necessary to establish liveness properties. This typically applies to distributed systems modeled by an MDP that relies on an interleaving semantics, and where (process) fairness simply serves to rule out unrealistic behaviors where certain processes eventually stop their execution without having reached a terminal state.

Consider, e.g., the simple mutual exclusion protocol with a randomized arbiter (cf. Example 10.83, page 835). In the absence of any fairness assumption, this protocol cannot guarantee that each process eventually enters its critical section almost surely. For instance, a scheduler which only selects the actions of the second process while completely ignoring the first one, is not excluded. Similarly, for the randomized dining philosophers algorithm (Example 10.87 on page 839), one cannot guarantee that each philosopher eats infinitely often almost surely, as there exist schedulers which only select the actions of one of the philosophers and never an action from one of the other philosophers.

For MDPs, fairness assumptions on the resolution of the nondeterministic choices are constraints on the schedulers. Instead of ranging over all schedulers, only the schedulers that generate fair paths are considered and are taken into account for the analysis. The underlying notion of fairness for paths is as for transition systems. In the sequel, it is assumed that fairness constraints are given as LTL fairness assumptions, i.e., conjunctions of unconditional fairness assumptions \(\square \diamond \Psi\), strong fairness assumptions \(\square \diamond \Phi \rightarrow \square \diamond \Psi\), and weak fairness assumptions \(\diamond \square \Phi \rightarrow \square \diamond \Psi\). Here, \(\Phi\) and \(\Psi\) are propositional logic formulae. A scheduler is fair if it almost surely generates fair paths.

\section*{Definition 10.130. Fair Scheduler}

Let \(\mathcal{M}\) be a Markov decision process and fair an LTL fairness assumption. A scheduler \(\mathfrak{F}\) for \(\mathcal{M}\) is fair (with respect to fair) if for each state \(s\) of \(\mathcal{M}\)
\[
\operatorname{Pr}_{s}^{\mathfrak{F}}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \text { fair }\}=1 .
\]

The fairness assumption fair is realizable in \(\mathcal{M}\) if there exists some fair scheduler for \(\mathcal{M}\).

Without any additional assumptions, fair schedulers need not exist. This is evident when there are no paths satisfying fair. But even if each finite path fragment can be extended to a path satisfying fair, the existence of fair schedulers is not guaranteed. For example, consider the following Markov chain:


Let the strong fairness assumption fair be defined as \(\square \diamond a \rightarrow \square \diamond b\). Each path fragment can be extended to a fair path, since there is always the possibility of eventually entering state \(t\) and staying there forever (and violating \(\square \diamond a\) ). On the other hand, when \(\mathcal{M}\) is considered as an MDP (labeling each transition with \(\alpha\), say), then \(\mathcal{M}\) has just one (memoryless) scheduler \(\mathfrak{S}\), viz. one that in each state always selects \(\alpha\). But \(\mathfrak{S}\) is not fair since almost surely both states \(s\) and \(t\) will be visited infinitely often. Hence, \(\square \diamond a \wedge \square \neg b\) holds almost surely.

In the sequel, we require the realizability of fair in the MDP \(\mathcal{M}\) under consideration. As LTL fairness assumptions constitute limit LT properties, for finite MDPs realizability is equivalent to the existence of a finite-memory fair scheduler (see Exercise 10.28, page 907).

The first observation is that realizable fairness assumptions are irrelevant for maximal reachability probabilities:

Lemma 10.131. Fairness and Max Reachability Probabilities
Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S, B, C \subseteq S\) and fair a realizable fairness assumption for \(\mathcal{M}\). For each state s of \(\mathcal{M}\) :
\[
\sup _{\substack{\mathcal{F} \text { far } \\ \text { eduler for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{F}}(s \models C \cup B)=\operatorname{Pr}^{\max }(s \models C \cup B) \text {. }
\]

Furthermore, there exists a finite-memory fair scheduler that maximizes the probabilities for \(C \cup B\).

Proof: Let \(\mathcal{M}\) be a finite MDP with state space \(S, B, C \subseteq S\) and fair a realizable fairness assumption for \(\mathcal{M}\). For \(s \in S\), it holds that
\[
\sup _{\substack{\mathcal{F} \text { far } \\ \text { schedurer for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{F}}(s \models C \cup B) \leqslant \sup _{\substack{\mathcal{S} \text { arbitrary } \\ \text { scheduler for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{G}}(s \models C \cup B)=\operatorname{Pr}^{\max }(s \models C \cup B)
\]
since fair schedulers are more restrictive than arbitrary ones.

The basic idea is now to construct a fair finite-memory scheduler \(\mathfrak{G}\) that maximizes the probabilities for the event \(C \cup B\) on the basis of a memoryless (possibly unfair) scheduler \(\mathfrak{S}\) that maximizes the probabilities for that event (cf. Lemma 10.102 on page 852).

Let \(\Pi\) be the set of all finite path fragments \(s_{0} s_{1} \ldots s_{n}\) in the Markov chain \(\mathcal{M}_{\mathfrak{S}}\) such that \(s_{n} \in B\) and \(s_{i} \in C \backslash B\) for \(0 \leqslant i<n\). Furthermore, since fair is realizable and \(\mathcal{M}\) is finite, there exists a fair finite-memory scheduler \(\mathfrak{F}\) for \(\mathcal{M}\).

The fair scheduler \(\mathfrak{G}\) that maximizes the probabilities for \(C \cup B\) is now defined as follows on the basis of \(\mathfrak{S}\) and \(\mathfrak{F}\). \(\mathfrak{G}\) behaves as the memoryless scheduler \(\mathfrak{S}\) for all histories \(s_{0} s_{1} \ldots s_{n}\) that are proper prefixes of some \(\widehat{\pi} \in \Pi\). As soon as \(\mathfrak{G}\) has generated the path fragment \(s_{0} s_{1} \ldots s_{n} \in \Pi\) then \(\mathfrak{G}\) continues in a fair way by mimicking the finite-memory scheduler \(\mathfrak{F}\) for \(\mathcal{M}\). Similarly, once a path fragment \(s_{0} \ldots s_{n}\) is generated that is not a prefix of some \(\widehat{\pi} \in \Pi, \mathfrak{G}\) behaves as \(\mathfrak{F}\).

Since the \(\mathfrak{S}\)-path fragments in \(\Pi\) are path fragments that are also generated by \(\mathfrak{G}\), we have
\[
\operatorname{Pr}^{\mathfrak{G}}(s \models C \cup B)=\operatorname{Pr}^{\mathfrak{S}}(s \models C \cup B)=\operatorname{Pr}^{\max }(s \models C \cup B) .
\]

Furthermore,
\[
\operatorname{Pr}_{s}^{\mathscr{G}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{pref}(\pi) \subseteq \operatorname{pref}(\Pi)\}=0
\]
where \(\operatorname{pref}(\pi)\) denotes the set of all finite prefixes of \(\pi\) and \(\operatorname{pref}(\Pi)\) the set of all finite prefixes of the path fragments in \(\Pi\). The event " \(\operatorname{pref}(\pi) \subseteq \operatorname{pref}(\Pi)\) " means that \(\mathfrak{G}\) never stops to mimic \(\mathfrak{S}\), and thus might behave unfairly along those paths.

The latter statement follows from the following reasoning. Since the Markov chain \(\mathcal{M}_{\mathfrak{S}}\) is finite (as \(\mathfrak{S}\) is memoryless), almost surely one of its BSCCs is reached and all its states visited infinitely often. But there is no \(\mathfrak{S}\)-path \(\pi\) that reaches a BSCC \(T\) of \(\mathcal{M}_{\mathfrak{S}}\) and visits all states of \(T\) infinitely often while fulfilling \(\operatorname{pref}(\pi) \subseteq \operatorname{pref}(\Pi)\). This can be seen as follows. Let \(\pi=s_{0} s_{1} s_{2} \ldots\) be a \(\mathfrak{S}\)-path such that \(\inf (\pi)=T\) and \(\operatorname{pref}(\pi) \subseteq \operatorname{pref}(\Pi)\). We show that \(B \cap T=\varnothing\). This goes by contraposition. Assume \(B \cap T \neq \varnothing\). Select a finite prefix \(\widehat{\pi}=s_{0} s_{1} \ldots s_{n}\) of \(\pi\) that ends in a state \(s_{n} \in B \cap T\). Then, by definition of \(\Pi\) and since \(\widehat{\pi} \in \operatorname{pref}(\pi) \subseteq \operatorname{pref}(\Pi)\), it follows that \(s_{0} s_{1} \ldots s_{n} \in \Pi\). But then none of the prefixes \(\widehat{\pi}_{m}=s_{0} s_{1} \ldots s_{n} s_{n+1} \ldots s_{m}\) of \(\pi\) of length \(m>n\) can be extended to a path fragment in \(\Pi\). That is, \(\widehat{\pi}_{m} \in \operatorname{pref}(\pi) \backslash \operatorname{pref}(\Pi)\) for all \(m>n\). This contradicts the assumption that \(\operatorname{pref}(\pi) \subseteq \operatorname{pref}(\Pi)\).

This ensures that almost surely the scheduler \(\mathfrak{G}\) will stop to mimic the memoryless scheduler \(\mathfrak{S}\) and will behave in a fairly manner from some moment on by simulating \(\mathfrak{F}\). This yields that the finite-memory scheduler \(\mathfrak{G}\) is fair.

As Lemma 10.131 asserts the existence of a fair scheduler that maximizes the probabilities for \(C \cup B\), the supremum in Lemma 10.131 can be replaced with a maximum.

The above result can be understood as the probabilistic counterpart to the fact that
realizable fairness assumptions are irrelevant for verifying safety properties; see Theorem 3.55 (page 140). Recall that a typical usage of \(\operatorname{Pr}^{\max }(s \models C \cup B)\) is to show that a safety property \(a_{1} \mathrm{~W} a_{2}\) holds with some sufficiently large probability \(1-\varepsilon\) under all schedulers. (Then \(C\) characterizes the states where \(a_{1} \wedge \neg a_{2}\) holds, while \(B\) represents the states satisfying \(\neg a_{1} \wedge \neg a_{2}\).) In this sense, computing \(\operatorname{Pr}^{\max }(s \models C \cup B)\) can be understood as quantitative reasoning about safety properties.

Due to the previous result, maximal probabilities can be computed without taking fairness assumptions into account. This, however, does not apply to minimal probabilities. Fairness assumptions may be essential when considering the minimal probabilities of reaching a certain set of states \(B\). This is a typical task to show that the liveness property \(\diamond b\) holds with probability \(\geqslant 1-\varepsilon\) under all schedulers, for some small \(\varepsilon\). For instance, consider the following MDP:


Consider the strong fairness assumption:
\[
\text { fair }=\square \diamond a \rightarrow \square \diamond b \text {, }
\]
which can be read as fair \(=\square \diamond u \rightarrow \square \diamond t\). All fair schedulers have to take action \(\alpha\) in state \(s\) infinitely often, and thus, \(\diamond t\) almost surely holds for all fair schedulers. On the other hand, the memoryless scheduler \(\mathfrak{S}\) that selects action \(\beta\) for state \(s\) never visits \(t\). Hence:
\[
\operatorname{Pr}^{\min }(s \models \diamond b)=0<1=\inf _{\substack{\mathfrak{F} \text { far } \\ \text { scheduler for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{F}}(s \models \diamond b) .
\]

We now show that the problem of computing the minimal probabilities of reaching \(B\) under fair schedulers is reducible to the problem of computing maximal reachability probabilities. Let \(\mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite \(\mathrm{MDP}, B \subseteq S\) and fair a fairness assumption that is realizable in \(\mathcal{M}\), i.e., \(\mathcal{M}\) is fair with respect to fair. Let \(F_{=0}^{\min }\) be the set of all states \(t \in S\) such that \(B\) will never be reached from \(t\) for some fair scheduler \(\mathfrak{F}\) :
\[
F_{=0}^{\min }=\left\{t \in S \mid \operatorname{Pr}^{\mathfrak{F}}(t \models \diamond B)=0 \text { for some fair scheduler } \mathfrak{F}\right\} .
\]

By the following result, the set \(F_{=0}^{\min }\) can be characterized by the end components \((T, A)\) of \(\mathcal{M}\) such that \(T \models\) fair. Notice that the fairness assumption fair is a limit LT property (see Notation 10.121 on page 872). Recall that \(T \models\) fair means that all paths \(\pi\) with \(\inf (\pi)=T\) fulfill fair.

\section*{Lemma 10.132. Characterization of the Set \(F_{=0}^{\min }\)}

Let \(\mathcal{M}\) be a finite \(M D P, B \subseteq S\), fair and
\[
F_{=0}^{\min }=\left\{t \in S \mid \operatorname{Pr}^{\mathfrak{F}}(t \models \Delta B)=0 \text { for some fair scheduler } \mathfrak{F}\right\} .
\]

For any state \(s \in S\), the following statements are equivalent:
(a) \(s \in F_{=0}^{\min }\), i.e., \(\operatorname{Pr}^{\mathfrak{F}}(s \models \diamond B)=0\) for some fair scheduler \(\mathfrak{F}\).
(b) \(\operatorname{Pr}^{\mathfrak{F}}(s=\diamond B)=0\) for some fair, finite-memory scheduler \(\mathfrak{F}\).
(c) \(\operatorname{Pr}^{\max }(s \models(\neg B) \cup V)=1\) where \(V\) is the union of the state sets \(T\) of all end components \((T, A)\) of \(\mathcal{M}\) such that \(T \cap B=\varnothing\) and \(T \models\) fair.

Proof: \((\mathrm{a}) \Longrightarrow(\mathrm{c}):\) Suppose \(s \in F_{=0}^{\min }\) and consider a fair scheduler \(\mathfrak{F}\) with \(\operatorname{Pr}^{\mathfrak{F}}(s \models \diamond B)=\) 0 . The limit of almost all \(\mathfrak{F}\)-paths is an end component. Let \((T, A)\) be an end component of \(\mathcal{M}\) such that \(\operatorname{Pr}_{s}^{\mathfrak{\xi}}\left(\Pi_{(T, A)}\right)>0\) where \(\Pi_{(T, A)}=\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi)=(T, A)\}\). Then, \(T \cap B=\varnothing\) and \(T \models\) fair, and therefore \(T \subseteq V\). Moreover, all paths \(\pi \in \Pi_{(T, A)}\) fulfill \(\square \neg B\), and hence also \((\neg B) \cup V\). Hence:
\[
\operatorname{Pr}^{\mathfrak{F}}(s \models(\neg B) \cup V)=1 .
\]

In particular, \(\operatorname{Pr}^{\max }(s \models(\neg B) \cup V)=1\).
(c) \(\Longrightarrow(\mathrm{b})\) : Let \(\operatorname{Pr}^{\max }(s \vDash(\neg B) \cup V)=1\). Consider a finite-memory scheduler \(\mathfrak{S}\) with \(\operatorname{Pr}^{\mathfrak{G}}(s \models(\neg B) \cup V)=1\). By definition of \(V\), there exists a fair finite-memoryless scheduler \(\mathfrak{H}\), say, such that \(\operatorname{Pr}^{\mathfrak{H}}(t \mid \diamond B)=0\) for any state \(t \in V\). (The scheduler \(\mathfrak{H}\) can be constructed using the same technique as used in the proof of Theorem 10.122, page 873.) We now compose \(\mathfrak{G}\) and \(\mathfrak{H}\) to obtain a fair finite-memory scheduler \(\mathfrak{F}\) which ensures that \(B\) will not be reached from \(s\). For starting state \(s\), scheduler \(\mathfrak{F}\) first acts as \(\mathfrak{G}\). As soon as \(V\) has been reached (which happens almost surely), \(\mathfrak{F}\) behaves as \(\mathfrak{H}\). It is now clear that \(\mathfrak{F}\) is fair (almost surely, eventually \(\mathfrak{F}\) simulates the fair scheduler \(\mathfrak{H}\) ) and has the finite-memory property (the union of the modes of \(\mathfrak{G}\) and \(\mathfrak{H}\) suffices). Moreover:
\[
\operatorname{Pr}^{\mathfrak{F}}(s \models \diamond B)=1-\operatorname{Pr}^{\mathfrak{G}}(s \models(\neg B) \cup V)=0 .
\]
\((\mathrm{b}) \Longrightarrow(\mathrm{a})\) : obvious.

The characterization of the set \(F_{=0}^{\min }\) enables characterizing the minimal probabilities for the event \(\diamond B\) under fair schedulers by means of maximal probabilities for a constrained reachability property:

\section*{Theorem 10.133. Fair Min Reachability Probabilities}

Let \(\mathcal{M}\) be a finite \(M D P\) with state space \(S, B \subseteq S\), fair a strong fairness constraint, and \(F_{=0}^{\min }\) as above. Then, for any state \(s \in S\) :
\[
\inf _{\substack{\mathfrak{F} \text { fair } \\ \text { scheduler for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{F}}(s \models \diamond B)=1-\operatorname{Pr}^{\max }\left(s \models(\neg B) \cup F_{=0}^{\min }\right) \text {. }
\]

Furthermore, there exists a fair finite-memory scheduler \(\mathfrak{F}\) with
\[
\operatorname{Pr}^{\widetilde{F}}(s \models \diamond B)=1-\operatorname{Pr}^{\max }\left(s \models(\neg B) \cup F_{=0}^{\min }\right) .
\]

Proof: We first prove the second statement. The proof is by constructing a finite-memory fair scheduler \(\mathfrak{G}\) such that for any state \(s\)
\[
\operatorname{Pr}^{\mathfrak{G}}(s \models \diamond B) \leqslant 1-\operatorname{Pr}^{\max }\left(s \models(\neg B) \cup F_{=0}^{\min }\right) .
\]

Let \(\mathfrak{S}\) be a memoryless scheduler maximizing the probabilities for \((\neg B) \cup F_{=0}^{\min }\); see Lemma 10.102 (page 852). From the proof of Lemma 10.132, it follows that there exists a fair finite-memory scheduler \(\mathfrak{H}\) such that \(\operatorname{Pr}^{\mathfrak{H}}(t \mid=\Delta B)=0\) for each state \(t \in F_{=0}^{\min }\). We now combine \(\mathfrak{S}\) and \(\mathfrak{H}\) to obtain a finite-memory scheduler \(\mathfrak{G}\) with the desired properties. In its starting mode, \(\mathfrak{G}\) simulates \(\mathfrak{S}\) until a path fragment \(s_{0} s_{1} \ldots s_{n}\) has been generated such that \(s_{i} \in S \backslash\left(B \cup F_{=0}^{\min }\right)\) for \(0 \leqslant i<n\) and either \(s_{n} \in F_{=0}^{\min }\) or \(s_{n} \in B\). In the former case, \(\mathfrak{G}\) switches mode and simulates \(\mathfrak{H}\) from now on. In the latter case, \(\mathfrak{G}\) behaves from now on as an arbitrary finite-memory fair scheduler. Obviously, \(\mathfrak{G}\) is fair and is finite memory, and
\[
\operatorname{Pr}^{\mathfrak{G}}\left(s \models(\neg B) \cup F_{=0}^{\min }\right)=\operatorname{Pr}^{\max }\left(s \models(\neg B) \cup F_{=0}^{\min }\right) .
\]

All \(\mathfrak{G}\)-paths \(\pi\) with \(\pi \models(\neg B) \cup F_{=0}^{\min }\) fulfill \(\pi \models \square(\neg B)\). Hence:
\[
\begin{aligned}
\operatorname{Pr}^{\mathfrak{G}}(s \models \diamond B) & =1-\operatorname{Pr}^{\mathfrak{G}}(s \models \square(\neg B)) \\
& \leqslant 1-\operatorname{Pr}^{\mathfrak{G}}\left(s \models(\neg B) \cup F_{=0}^{\min }\right) \\
& =1-\operatorname{Pr}^{\max }\left(s \models(\neg B) \cup F_{=0}^{\min }\right) .
\end{aligned}
\]

This shows the second claim. It remains to show that for each state \(s\) of \(\mathcal{M}\) :
\[
\inf _{\substack{\mathfrak{F} \text { fair } \\ \text { scheduler for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{F}}(s \models \diamond B) \geqslant 1-\operatorname{Pr}^{\max }\left(s \models(\neg B) \cup F_{=0}^{\min }\right) .
\]

As for each fair scheduler \(\mathfrak{F}\) we have \(\operatorname{Pr}^{\mathfrak{F}}(s \models \diamond B)=1-\operatorname{Pr}^{\mathfrak{F}}(s \models \square(\neg B))\), it suffices to show that
\[
\operatorname{Pr}^{\widetilde{F}}(s \models \square(\neg B)) \leqslant \operatorname{Pr}^{\widetilde{F}}\left(s \models(\neg B) \cup F_{=0}^{\min }\right) .
\]

By Theorem 10.120 on page 871 , we have
\[
\begin{aligned}
& \operatorname{Pr} r^{\mathfrak{F}}(s \models \square(\neg B)) \\
& =\sum_{\substack{(T, A) \text { end comp. } \\
T \cap B=\varnothing}} \operatorname{Pr}_{s}^{\mathfrak{F}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi)=(T, A) \wedge \pi \models \square(\neg B)\} \\
& \leqslant \sum_{\substack{(T, A) \text { end comp. } \\
T \cap B=\varnothing}} \operatorname{Pr}_{s}^{\mathfrak{F}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi)=(T, A) \wedge \pi \models(\neg B) \cup T\} .
\end{aligned}
\]

Furthermore, since \(\mathfrak{F}\) is fair, for each end component \((T, A)\) such that \(\operatorname{Pr}_{s}^{\mathfrak{F}}\{\pi \in \operatorname{Paths}(s) \mid\) \(\operatorname{Limit}(\pi)=(T, A)\}>0\), we have \(T \models\) fair. Lemma 10.132 asserts that the end components \((T, A)\) with \(T \models\) fair and \(T \cap B=\varnothing\) belong to \(F_{=0}^{\min }\). Hence:
\[
\begin{aligned}
& \operatorname{Pr}^{\widetilde{F}}(s \models \square(\neg B)) \\
& \leqslant \sum_{\substack{(T, A) \text { end comp. } \\
T \cap B=\varnothing, T \vDash \text { fair }}} \operatorname{Pr}_{s}^{\mathfrak{\lessgtr}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi)=(T, A) \wedge \pi \models(\neg B) \cup T\} \\
& \leqslant \sum_{\substack{(T, A \text { end comp. } \\
T \cap B=\varnothing, T \models \text { fair }}} \operatorname{Pr}_{s}^{\mathfrak{\Psi}}\left\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi)=(T, A) \wedge \pi \models(\neg B) \cup F_{=0}^{\min }\right\} \\
& \leqslant \quad \sum \quad \operatorname{Pr}_{s}^{\mathfrak{F}}\left\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi)=(T, A) \wedge \pi \vDash(\neg B) \cup F_{=0}^{\min }\right\} \\
& (T, A) \text { end comp. } \\
& =\operatorname{Pr}^{\mathfrak{F}}\left(s \models(\neg B) \cup F_{=0}^{\min }\right) \text {. }
\end{aligned}
\]

Since \(\operatorname{Pr}^{\mathfrak{F}}\left(s \models(\neg B) \cup F_{=0}^{\min }\right)\) is bounded above by \(\operatorname{Pr}^{\max }\left(s \models(\neg B) \cup F_{=0}^{\min }\right)\), this yields the first claim.

Theorem 10.133 suggests the following recipe to determine the minimal probabilities for reachability properties \(\forall B\) (for all fair schedulers) of a finite MDP: (1) determine the set \(F_{=0}^{\min }\) and (2) solve the linear program for the maximal probabilities for \((\neg B) \cup F_{=0}^{\min }\). The computation of \(F_{=0}^{\min }\) relies on a graph analysis of the end components of the MDP.

In case fair consists of weak fairness constraints \(\diamond \square \Phi_{j} \rightarrow \square \diamond \Psi_{j}, j=1, \ldots, k\), the maximal end component may be exploited. In this case, \(V\) is the union of the sets \(T\) of all maximal end components \((T, A)\) such that, for each \(1 \leqslant j \leqslant k, T \cap \operatorname{Sat}\left(\Psi_{j}\right) \neq \varnothing\) or \(T \backslash \Phi_{j} \neq \varnothing\). For strong fairness constraints, the analysis of maximal end components might not be sufficient. However, the algorithm for computing maximal end components (Algorithm 47) can be reformulated to compute all end components \((T, A)\) such that \(T \models\) fair and \((T, A)\) is not contained in another end component \(\left(T^{\prime}, A^{\prime}\right)\) where \(T^{\prime} \models\) fair. See Exercise 10.29 (page 907).

Constrained reachability properties \(C \cup B\) can be treated as simple reachability properties by making all states \(s \in S \backslash(C \cup B)\) in the MDP absorbing. Let \(\mathcal{M}^{\prime}\) be the thus obtained MDP. The paths in \(\mathcal{M}\) satisfying \(C \cup B\) agree with the paths in \(\mathcal{M}^{\prime}\) satisfying \(\diamond B\). Thus:
\[
\inf _{\substack{\mathcal{F} \text { fair } \\ \text { scheduler for } \mathcal{M}}} \operatorname{Pr}_{\mathcal{M}}^{\mathcal{F}}(s \models C \cup B)=\inf _{\substack{\mathcal{F}^{\prime} \text { fair } \\ \text { scheduler for } \mathcal{M}^{\prime}}} \operatorname{Pr}_{\mathcal{M}^{\prime}}^{\mathfrak{F}^{\prime}}(s \models \diamond B) .
\]

The ingredients to treat (constrained) reachability properties under fairness assumptions can be combined with the standard approach to compute satisfaction sets of CTL-like logics to obtain a model-checking algorithm for PCTL under fairness assumptions and finite MDPs. Let \(\mathcal{M}\) be a finite MDP and fair an LTL fairness assumption that is realizable in \(\mathcal{M}\). The satisfaction relation \(\models_{\text {fair }}\) for PCTL state- and path formulae is defined as the standard satisfaction relation \(\models\), except that the probabilistic operator ranges over all fair schedulers (instead of all schedulers). That is:
\[
s \models_{\text {fair }} \mathbb{P}_{J}(\varphi) \quad \text { iff } \quad \operatorname{Pr} r^{\mathfrak{F}}(s \models \varphi) \in J \text { for all fair schedulers } \mathfrak{F} \text {. }
\]

The satisfaction set
\[
\operatorname{Sat}_{\text {fair }}\left(\mathbb{P}_{J}(\varphi)\right)=\left\{s \in S \mid s \models_{\text {fair }} \mathbb{P}_{J}(\varphi)\right\}
\]
is obtained as follows. For path formulae with the next-step operator, the fairness constraints are irrelevant due to the realizability of fair. Thus,
\[
\operatorname{Sat}_{f a i r}\left(\mathbb{P}_{J}(\bigcirc a)\right)=\operatorname{Sat}\left(\mathbb{P}_{J}(\bigcirc a)\right)
\]
for \(a \in A P\). For path formulae with the until operator we apply the techniques explained above.

As a next verification problem, we consider the quantitative analysis of finite MDPs against an \(\omega\)-regular property \(P\) in the presence of fairness assumptions. As for the case without fairness assumptions, we adopt the automata-based approach and first construct a deterministic Rabin automaton \(\mathcal{A}\) for \(P\). Subsequently, the product-MDP \(\mathcal{M} \otimes \mathcal{A}\) is considered; see Notation 10.128 on page 881 . Let us first assume that we are interested in the maximal probabilities for \(P\), when ranging over all fair schedulers. By a graph analysis (similar to the technique sketched above), we determine the union \(V\) of all end components \((T, A)\) in \(\mathcal{M} \otimes \mathcal{A}\) that satisfy the fairness assumption of \(\mathcal{M}\) and the acceptance condition of \(\mathcal{A}\). Then:
\[
\sup _{\substack{\mathcal{F} \text { far } \\ \text { eduler for } \mathcal{M}}} \operatorname{Pr}_{\mathcal{M}}^{\mathfrak{\mathcal { S }}}(s \models P)=\operatorname{Pr}_{\mathcal{M} \otimes \mathcal{A}}^{\max }\left(\left\langle s, q_{s}\right\rangle \models \diamond V\right)
\]
where \(q_{s}=\delta\left(q_{0}, L(s)\right)\). In fact, there is a finite-memory fair scheduler for \(\mathcal{M}\) that maximizes the probabilities for \(P\). Hence, the supremum can be replaced by a maximum. Such a finite-memory fair scheduler can be derived from (i) a memoryless scheduler \(\mathfrak{S}\) for
\(\mathcal{M} \otimes \mathcal{A}\) that maximizes the probabilities of reaching \(V\), and (ii) a finite-memory scheduler \(\mathfrak{F}\) for \(\mathcal{M} \otimes \mathcal{A}\) which ensures that whenever an end component \((T, A)\) of \(\mathcal{M} \otimes \mathcal{A}\) is reached such that \(T \models\) fair and \(T\) satisfies the acceptance condition of \(\mathcal{A}\), then \(T\) will never be left and all actions in \(T\) are visited infinitely often. These two finite-memory schedulers \(\mathfrak{S}\) and \(\mathfrak{F}\) for \(\mathcal{M} \otimes \mathcal{A}\) can be combined to obtain a finite-memory fair scheduler \(\mathfrak{G}\) for \(\mathcal{M} \otimes \mathcal{A}\) that maximizes the probabilities for \(P\). A corresponding finite-memory scheduler \(\mathfrak{G}\) is obtained by encoding the states in \(\mathcal{A}\) in the modes of \(\mathfrak{G}\).

To compute minimal probabilities for \(P\) under all fair schedulers, we consider the complement property \(\bar{P}\) and compute the maximal probabilities for \(\bar{P}\) under all fair schedulers. This is sufficient, since
\[
\min _{\substack{\mathfrak{F} \text { fair } \\ \text { scheduler for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{F}}(s \models P)=1-\underset{\substack{\mathcal{F} \text { fax } \\ \text { scheduler for } \mathcal{M}}}{ } \operatorname{Pr}^{\mathfrak{F}}(s \models \bar{P}) .
\]

Combining the techniques for model checking PCTL and those for checking \(\omega\)-regular properties yields a model-checking procedure for PCTL* and finite MDPs under fairness assumptions. The worst-case time complexity is roughly the same for PCTL* model checking of MDPs without fairness. The main difference is that the graph analysis for determining the end components is more advanced and can cause an additional factor \(\mid\) fair \(\mid\) in the cost function.

In LTL and CTL*, fairness assumptions can be encoded syntactically into the formula to be checked. This allows one to reduce the fair satisfaction relation \(\models_{\text {fair }}\) to the standard satisfaction relation \(\models\). In CTL*, e.g., it holds that
\[
s \models_{\text {fair }} \exists \varphi \quad \text { if and only if } \quad s \models \exists(\text { fair } \wedge \varphi)
\]
and
\[
s \models_{\text {fair }} \forall \varphi \quad \text { if and only if } \quad s \models \text { fair } \rightarrow \varphi .
\]

To conclude this section, we show that an analogous result can also be established for the logic PCTL*. Clearly,
\[
s \models \mathbb{P}_{\geqslant p}(\text { fair } \rightarrow \varphi) \quad \text { implies } \quad s \models_{\text {fair }} \mathbb{P}_{\geqslant p}(\varphi)
\]
and
\[
s \models \mathbb{P}_{\leqslant p}(\text { fair } \wedge \varphi) \quad \text { implies } \quad s \models_{\text {fair }} \mathbb{P}_{\leqslant p}(\varphi) \text {. }
\]

The question is whether the reverse implications also hold. At first glance this does not seem to be the case since, e.g., \(s \models_{\text {fair }} \mathbb{P}_{\geqslant p}(\varphi)\) only concerns the fair schedulers, while \(s \neq \mathbb{P}_{\geqslant p}(\) fair \(\rightarrow \varphi)\) considers all schedulers, including the unfair ones. While schedulers \(\mathfrak{S}\) with \(0<\operatorname{Pr}^{\mathfrak{S}}(s \models\) fair \()<1\) are ignored by the fair satisfaction relation \(\models_{\text {fair }}\), satisfaction \(s \models \mathbb{P}_{\geqslant p}(\) fair \(\rightarrow \varphi)\) under the standard relation \(\models\) without fairness requires
\[
\operatorname{Pr}^{\mathfrak{S}}(s \not \vDash \text { fair })+\operatorname{Pr}^{\mathfrak{S}}(s \models \text { fair } \wedge \varphi) \geqslant p
\]
for all schedulers. In fact, when dropping the realizability assumption, then \(s \models_{\text {fair }} \mathbb{P}_{\geqslant p}(\varphi)\) and \(s \not \vDash \mathbb{P}_{\geqslant p}(\) fair \(\rightarrow \varphi)\) is possible. A simple example is a Markov chain \(\mathcal{M}\) (viewed as an MDP) with \(\operatorname{Pr}(s \models\) fair \()=\frac{1}{2}\) and \(\operatorname{Pr}(s \models \varphi)=0\). Since there is no fair scheduler, \(s \neq_{\text {fair }} \mathbb{P}_{=1}(\varphi)\), but \(s \not \vDash \mathbb{P}_{=1}(\) fair \(\rightarrow \varphi)\). However, the realizability assumption permits an encoding of the fair satisfaction relation by means of the standard satisfaction relation for finite MDPs. This is stated by the following theorem:

Theorem 10.134. Reduction of \(\models_{\text {fair }}\) to \(\models\)
Let \(\mathcal{M}\) be a finite \(M D P\) and fair an LTL fairness assumption that is realizable for \(\mathcal{M}\). Then, for each LTL formula \(\varphi\) and state \(s\) of \(\mathcal{M}\) :
\[
\begin{aligned}
& \min _{\substack{\mathfrak{F} \text { fair } \\
\text { ssheduler for } \mathcal{M}}} \operatorname{Pr}^{\mathfrak{F}}(s \models \varphi)=\operatorname{Pr}_{\substack { \min \\
\begin{subarray}{c}{\mathfrak{F} \text { fair } \\
\text { scheduler for } \mathcal{M}{ \operatorname { m i n } \\
\begin{subarray} { c } { \mathfrak { F } \text { fair } \\
\text { scheduler for } \mathcal { M } } }\end{subarray}}^{\operatorname{Pr}^{\mathfrak{F}}(s \models \text { fair } \rightarrow \varphi)}=\operatorname{Pr}^{\max }(s \models \text { fair } \wedge \varphi)
\end{aligned}
\]

In particular, \(s \models_{\text {fair }} \mathbb{P}_{\geqslant p}(\varphi)\) iff \(s \models \mathbb{P}_{\geqslant p}(\) fair \(\rightarrow \varphi)\) and \(s \models_{\text {fair }} \mathbb{P}_{\leqslant p}(\varphi)\) iff \(s \models \mathbb{P}_{\leqslant p}(\) fair \(\wedge\) \(\varphi)\). The same holds for strict probability bounds \(<p\) and \(>p\).

Proof: We first prove the statement about minimal probabilities. For any fair scheduler \(\mathfrak{F}\), it holds that \(\operatorname{Pr}^{\widetilde{\mathfrak{F}}}(s \models \varphi)=\operatorname{Pr}^{\mathfrak{F}}(s \models\) fair \(\rightarrow \varphi)\). Hence, the minimal probability for \(\varphi\) under all fair schedulers is at least the minimal probability for fair \(\rightarrow \varphi\) under all schedulers.

We now show that there exists a fair (finite-memory) scheduler such that the probability for \(\varphi\) equals the minimum for fair \(\rightarrow \varphi\) under all schedulers. For the sake of simplicity, assume that \(\varphi\) describes a limit LT property. (This is not a restriction, as \(\varphi\) can be replaced by the acceptance condition of the DRA for \(\varphi\).) Let \(\mathfrak{S}\) be a finite-memory (possibly unfair) scheduler that minimizes the probabilities for fair \(\rightarrow \varphi\). Thus, \(\mathfrak{S}\) maximizes the probabilities for fair \(\wedge \neg \varphi\) and:
\[
\begin{aligned}
\operatorname{Pr}^{\max }(s \models \operatorname{fair} \wedge \neg \varphi) & =\operatorname{Pr}^{\mathfrak{G}}(s \models \text { fair } \wedge \neg \varphi) \\
& =\sum_{(T, A)} \operatorname{Pr}_{s}^{\mathfrak{S}}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{Limit}(\pi)=(T, A)\}
\end{aligned}
\]
where \((T, A)\) ranges over all end components of \(\mathcal{M}\) such that \(T \models\) fair \(\wedge \neg(T \models \varphi)\). Since \(\mathfrak{S}\) has the finite-memory property, the induced Markov chain \(\mathcal{M}_{\mathfrak{S}}\) is finite, and the above sum can be rewritten as
\[
\sum_{T} \operatorname{Pr}_{s}^{\mathfrak{S}}\{\pi \in \operatorname{Paths}(s) \mid \inf (\pi)=T\}
\]
where \(T\) ranges over all BSCCs of \(\mathcal{M}_{\mathfrak{S}}\) such that \(T \models\) fair \(\wedge \neg(T \models \varphi)\). We now consider the finite-memory scheduler \(\mathfrak{F}\) that first mimics \(\mathfrak{S}\) until \(\mathfrak{S}\) reaches a BSCC \(T\). If \(T \models\) fair \(\wedge \neg(T \models \varphi)\), then \(\mathfrak{F}\) behaves forever as \(\mathfrak{S}\). If \(\neg(T \models\) fair \()\) or \(T \models \varphi\), then \(\mathfrak{F}\) behaves from now on in an arbitrary but fair way (by acting like some fair finite-memory scheduler). \(\mathfrak{F}\) is, in fact, a fair scheduler (because \(\mathfrak{S}\) reaches almost surely some BSCC). Furthermore:
\[
\begin{aligned}
\operatorname{Pr}^{\mathfrak{F}}(s \models \neg \varphi) & \geqslant \sum_{T} \operatorname{Pr}_{s}^{\mathfrak{S}}\{\pi \in \operatorname{Paths}(s) \mid \inf (\pi)=T\} \\
& =\operatorname{Pr}^{\max }(s \models \text { fair } \wedge \neg \varphi)
\end{aligned}
\]
where \(T\) ranges over all BSCCs with \(T \models\) fair \(\wedge \neg(T \models \varphi)\). As \(\operatorname{Pr}^{\mathfrak{F}}(s \models \neg \varphi)=\operatorname{Pr}^{\mathfrak{F}}(s \models\) fair \(\wedge \neg \varphi\) ), the inequality \(\geqslant\) in the above formula can be replaced with an equality \(=\). This yields
\[
\begin{aligned}
\operatorname{Pr}^{\mathfrak{F}}(s \models \varphi) & =1-\operatorname{Pr}^{\widetilde{\mathscr{}}}(s \models \neg \varphi) \\
& =1-\operatorname{Pr}^{\mathfrak{F}}(s \models \text { fair } \wedge \neg \varphi) \\
& =\operatorname{Pr}^{\min }(s \models \text { fair } \rightarrow \varphi) .
\end{aligned}
\]

The statement for maximal probabilities follows by duality. Let \(\mathfrak{F}\) range over all fair schedulers and \(\mathfrak{S}\) over all schedulers. Then:
\[
\begin{aligned}
\max _{\mathfrak{F}} \operatorname{Pr}^{\mathfrak{F}}(s \models \varphi) & =1-\min _{\mathfrak{F}} \operatorname{Pr}^{\mathfrak{F}}(s \models \neg \varphi) \\
& =1-\operatorname{Pr}^{\min }(s \models \text { fair } \wedge \neg \varphi) \\
& =\operatorname{Pr}^{\max }(s \models \text { fair } \rightarrow \varphi) .
\end{aligned}
\]

As a consequence:
\[
s \models_{\text {fair }} \mathbb{P}_{=1}(\varphi) \quad \text { if and only if } \quad s \models \mathbb{P}_{=1}(\text { fair } \rightarrow \varphi)
\]
and
\[
s \models_{\text {fair }} \mathbb{P}_{=0}(\varphi) \quad \text { if and only if } \quad s \models \mathbb{P}_{=0}(\text { fair } \wedge \varphi) \text {. }
\]

That is to say, there exists a fair scheduler \(\mathfrak{F}\) with \(\operatorname{Pr}^{\mathfrak{F}}(\varphi)>0\) if and only if there exists a scheduler \(\mathfrak{S}\) with \(\operatorname{Pr}^{\mathfrak{S}}(\) fair \(\wedge \varphi)>0\). These reductions emphasize the expressiveness of PCTL* as a logic for reasoning about finite MDPs. However, for efficiency reasons it is recommended to perform a quantitative analysis of finite MDPs with fairness assumptions using the techniques described before.

\subsection*{10.7 Summary}
- Markov chains are transition systems with fixed probability distributions over the successors of each state.
- A qualitative property is an event that either holds with probability one or zero. Checking a qualitative property in finite Markov chains can be done by graph algorithms. This does not hold for infinite Markov chains.
- (Constrained) Reachability probabilities can be computed by a graph analysis and solving a linear equation system.
- Almost surely, the long-run behavior of a finite Markov chain ends in a bottom strongly connected component (BSCC). Quantitative (and qualitative) properties about the long-run behavior-such as repeated reachability, persistence, or Boolean combinations thereof - of a finite Markov chain \(\mathcal{M}\) can be checked by computing the probability of reaching an accepting BSCC in \(\mathcal{M}\).
- Probabilistic Computation Tree Logic (PCTL) is a quantitative variant of CTL where the path quantifiers \(\exists\) and \(\forall\) are replaced by a probabilistic operator \(\mathbb{P}_{J}(\varphi)\) that specifies lower and/or upper probability bounds (given by \(J\) ) for the event \(\varphi\).
- PCTL model checking for finite Markov chains relies on the standard CTL modelchecking procedure in combination with methods for computing constrained reachability probabilities.
- The qualitative fragment of PCTL is obtained by only allowing bounds \(>0\) and \(=1\). For finite Markov chains, CTL is at least as expressive as the qualitative fragment of PCTL. For infinite Markov chains, the expressivity of the qualitative fragment of PCTL and CTL is incomparable. As opposed to CTL, persistence properties can be expressed in PCTL (both qualitative and quantitative).
- Computing the probability for an LT property \(P\) on a finite Markov chain \(\mathcal{M}\) can be reduced to computing the acceptance probability in the product of \(\mathcal{M}\) and a deterministic Rabin automaton (DRA) for the complement of \(P\).
- A probabilistic bisimulation equivalence only relates states that are equally labeled and whose cumulative probability of moving to the equivalence classes coincides.
- Probabilistic bisimulation on Markov chains coincides with PCTL and PCTL* equivalence. (The logic PCTL* is obtained by combining PCTL and LTL.) This holds for any arbitrary Markov chain - the restriction to finitely-branching models (as for the logical characterization of bisimulation on transition systems) is not required.
- Long-run averages and expected cost-bounded reachability in Markov reward chains can be determined using graph algorithms and solving linear equation systems.
- Markov decision processes (MDPs) are transition systems in which in any state a nondeterministic choice between probability distributions exists. A Markov chain is an MDP in which for any state the set of probability distributions is a singleton. MDPs are adequate for modeling randomized distributed algorithms.
- Reasoning about probabilities in MDPs requires the concept of schedulers. Schedulers resolve the nondeterminism and yield a stochastic process. Computing extreme (i.e., minimal or maximal) probabilities for constrained reachability properties relies on graph algorithms and linear programs. The latter can be solved by means of an iterative approximation algorithm (called value iteration).
- When interpreting PCTL on MDPs, the formula \(\mathbb{P}_{J}(\varphi)\) ranges over all schedulers. The PCTL model-checking problem for MDPs is reducible to the reachability problem.
- Almost surely, the long-run behavior of a finite MDP ends in one of its end components. Quantitative (and qualitative) properties about the long-run behavior-such as repeated reachability, persistence, or Boolean combinations thereof-of a finite MDP \(\mathcal{M}\) can be checked by computing the extreme probability of reaching an "accepting" end component in \(\mathcal{M}\).
- Model-checking a finite MDP against an \(\omega\)-regular property can be solved by an automata-based approach, analogous to finite Markov chains.
- A scheduler is fair if it almost surely generates only fair paths. For maximal reachability properties, fairness is irrelevant. Minimal reachability probabilities under all fair schedulers can be computed by graph-based techniques and determining maximal probabilities for a constrained reachability property. Model checking a finite MDP with fairness assumptions against an \(\omega\)-regular property can be performed by an automata-based approach similar to that for MDPs without fairness assumptions.

\subsection*{10.8 Bibliographic Notes}

Markov chains and Markov decision processes. The main principles of Markov chains as a mathematical model for stochastic processes goes back to Markov in 1906. Since then a variety of different aspects of Markov chains has been studied in the literature such as queuing theory [55], numerical algorithms [63, 378], reliability and performance analysis [196], and lumping [72, 237]. Markov chains have been applied in various areas ranging from systems biology, social sciences, and psychology, to electrical engineering and operations research. Important textbooks on Markov chains are, e.g., [237, 238, 248]. The reader should bear in mind that this monograph treats Markov chains from the statebased view as a graph (transition system) with probabilities, rather than - the more usual interpretation - as a sequence of random variables. Markov reward models are extensively described in the monograph by Howard [216]. The zeroconf protocol example has been adopted from Bohnenkamp et al. [54].

Markov decision processes (MDPs) have their roots in operations research and stochastic control theory. They are useful for studying a wide range of optimization problems. MDPs were known at least as early as in the fifties; see the seminal work by Bellman [38, 39]. MDPs are used in a variety of application areas, including robotics, automated control, economics and in manufacturing. Vardi [407] proposed using MDPs as models for concurrent probabilistic systems. Major textbooks on MDPs are by Puterman [346], Howard [215], and Bertsekas [49]. The randomized philosophers example originates from Lehmann and Rabin [268] and the randomized leader election algorithm of Itai and Rodeh [223]. Other randomized algorithms can be found in, e.g., Rabin [349] or the textbooks by Motwani and Raghavan [306], and Lynch [280].

Verifying qualitative properties. Verification techniques for probabilistic models date back to the early eighties and were focused on qualitative LT properties. Hart, Sharir, and Pnueli [191] observed that graph-based algorithms are sufficient for proving almost sure termination for finite-state concurrent probabilistic programs. The verification of qualitative \(\omega\)-regular properties for finite MCs and finite MDPs has first been addressed by Vardi and Wolper [407, 409, 411, 412]. They represent \(\omega\)-regular properties by means of NBAs that are deterministic in the limit. Vardi and Wolper also showed that the qualitative LTL model-checking problem for finite Markov chains is PSPACE-complete. Courcoubetis and Yannakakis [104] extended these results by techniques for the quantitative analysis of MCs against LTL and NBA specifications (see below). They also established a doubleexponential lower bound for the problem of verifying whether an LTL formula holds almost surely for a finite MDP. Pnueli and Zuck have developed a tableau-based verification technique for MDPs and qualitative LTL formulae [340] as well as proof methods for MDPs that rely on the connection between probabilistic choice and fairness; see e.g., [341]. A generalization of some of these concepts has been discussed by Baier and Kwiatkowska [32]. Fairness assumptions that impose restrictions on the resolution of nondeterminism in MDPs have been first addressed by Hart, Sharir, and Pnueli [191] and Vardi [407]. The role of fairness in the context of PCTL (and PCTL*) model checking has been discussed by Baier and Kwiatkowska [31].

Verifying quantitative properties. The presented algorithm for the quantitative analysis of MCs against \(\omega\)-regular specifications exploits deterministic Rabin automata (DRAs). This approach is conceptually simple as the DRA does not affect the transition probabilities in the product Markov chain. Although for many \(\omega\)-regular properties DRA exist whose size is of the same order as NBA [241], in the worst case the smallest DRA can be exponentially larger than the smallest equivalent NBA. Several alternative algorithms have been presented in the literature that have a time complexity which is polynomial in the size of the Markov chain and exponential in the length of the LTL formula \(\varphi\). Such algorithms have been proposed in the seminal paper by Courcoubetis and Yannakakis [104], and more recently (using different techniques) by Couvreur, Saheb, and Sutre [108],
and Bustan, Rubin, and Vardi [77]. The observation that the quantitative analysis of MDPs can be solved by linear programs goes back to an earlier paper by Courcoubetis and Yannakakis [103].

Branching-time properties. A branching-time logic for reasoning about probabilistic systems has been originally proposed by Hart and Sharir [190]. Their focus was on qualitative properties and deductive proof rules. Hansson and Jonsson introduced Probabilistic Computation Tree Logic (PCTL) and the PCTL model-checking procedure for finite Markov chains [187]. Variants of PCTL have been proposed for MDPs (or similar models). Hansson [186] and Segala and Lynch [370] presented action-based variants of PCTL. The (state-based) variant of PCTL and PCTL* for MDPs is due to Bianco and de Alfaro [51]. The concept of end components was introduced by Courcoubetis and Yannakakis [104] and has been studied in more detail by de Alfaro [115, 116]. Efficient counterexample generation algorithms for PCTL have recently been proposed by Han and Katoen [185]. A detailed account of (among others) PCTL model checking has been given by Kwiatkowska, Norman, and Parker [254]. Extensions of PCTL-like logics for MDPs (e.g., long-run properties) have been studied by de Alfaro; see e.g., [114, 115, 117]. The logic PCRTL (PCTL with rewards) which, amongst others, supports long-run averages and expected cumulative rewards has been proposed by Andova, Hermanns, and Katoen [14]. They also provide model-checking algorithms for this logic.

Probabilistic bisimulation. Bisimulation for Markov chains and an action-based model á la MDPs has been introduced in the seminal paper by Larsen and Skou [264]. Aziz et al. [23] have shown that bisimulation equivalence on Markov chains preserves the validity of PCTL* formulae. The observation that PCTL \(^{-}\)-equivalence agrees with bisimulation equivalence goes back to Desharnais et al. [121]. These authors study this for labeled Markov processes, a probabilistic model with continuous state spaces. Jonsson and Larsen [227] defined simulation relations for probabilistic models. Simulation and bisimulation relations, as well as the use of MDP-like models as semantical model for process algebras, can be found in the survey paper by Jonsson, Yi, and Larsen [228]. Further results on the connection between PCTL/PCTL* and bisimulation and simulation relations for Markov chains have been recently established by Baier et al. [30]. For MDPs (and the like) we refer to the works by Hansson [186], Segala and Lynch [370], and Desharnais and her colleagues \([122,123,124]\). Bisimulation minimization algorithms for MCs have been considered by Huynh and Tian [220], Baier, Engelen, and Majster-Cederbaum [27], and Derisavi, Hermanns, and Sanders [120]. Recently, Katoen et al. [232] have shown experimentally that substantial reductions in both memory and time can be obtained by exploiting bisimulation minimization for PCTL model checking and reward properties.

Probabilistic model checkers. One of the first prototypical PCTL model checkers has been reported by Fredlund [156]. More recent PCTL model checkers are PRISM [255]
and ETMCC [201] (and its successor, MRMC [233]). The former supports both MCs and MDPs, the latter only MCs. PRISM uses a variant of BDDs [26, 181] to enable the compact representation of transition matrices [254, 323]. MRMC is based on a sparse matrix representation and supports minimization techniques for probabilistic bisimulation. Both MRMC/ETMCC and PRISM also support CSL model checking, a continuous-time variant of PCTL proposed by Aziz et al. [22] and Baier et al. [29], and expected measures for Markov reward chains. Other model checkers for Markov chains are ProbVerus [192] and FMur \(\varphi\) [331]. Alternative model checkers for Markov decision processes are LiQuor [82] and Rapture [112, 226]. The latter tool focuses on reachability properties and exploits abstraction-refinement techniques. LiQuor supports the verification of quantitative and qualitative \(\omega\)-regular properties of MDPs that are modeled using a probabilistic variant of Promela, the input language of the LTL model checker SPIN. This variant [25] is based on the probabilistic guarded command language by Morgan and McIver [304]. LiQuor uses partial order reduction techniques to combat the state-space explosion problem [28, 113].

\subsection*{10.9 Exercises}

Exercise 10.1. Consider the Markov chain \(\mathcal{M}\) shown in Figure 10.22. Let \(C=\left\{s_{0}, s_{1}, s_{4}, s_{6}\right\}\)


Figure 10.22: Markov chain \(\mathcal{M}\) for Exercise 10.1.
and \(B=\left\{s_{2}, s_{3}\right\}\).
(a) Compute the probability measure of the union of the following cylinder sets:
\[
\operatorname{Cyl}\left(s_{0} s_{1}\right), \operatorname{Cyl}\left(s_{0} s_{5} s_{6}\right), \operatorname{Cyl}\left(s_{0} s_{5} s_{4} s_{3}\right), \operatorname{Cyl}\left(s_{0} s_{1} s_{6}\right)
\]
given that the initial distribution is given by \(i_{\text {init }}\left(s_{0}\right)=1\).
(b) Compute \(\operatorname{Pr}\left(s_{0} \models \diamond B\right)\) using the least fixed point characterization.
(c) Compute \(\operatorname{Pr}\left(s_{0} \models C \mathrm{U}^{\leqslant 5} B\right)\) using:
(i) the least fixed point characterization;
(ii) transient state probabilities.
(d) Determine \(\operatorname{Pr}\left(s_{0} \models \diamond \square D\right)\) with \(D=\left\{s_{3}, s_{4}\right\}\).

Exercise 10.2. Consider the Markov chain of Figure 10.23. Let \(B_{1}=\left\{s_{1}, s_{7}\right\}, B_{2}=\left\{s_{6}, s_{7}, s_{8}\right\}\),


Figure 10.23: Markov chain \(\mathcal{M}\) for Exercise 10.2.
\(B_{3}=\left\{s_{1}, s_{3}, s_{7}, s_{9}\right\}\) and \(B_{4}=\left\{s_{2}, s_{3}, s_{4}, s_{5}, s_{6}, s_{7}, s_{8}, s_{9}\right\}\). Determine whether:
(a) \(\operatorname{Pr}\left(s_{0} \models \diamond B_{1}\right)=1\),
(b) \(\operatorname{Pr}\left(s_{7}, \diamond B_{2}\right)=1\),
(c) \(\operatorname{Pr}\left(s_{0}, \square \diamond B_{i}\right)=1\), for \(i \in\{1,2,3\}\),
(d) \(\operatorname{Pr}\left(s_{0}, \diamond \square B_{2}\right)=1\), and
(e) \(\operatorname{Pr}\left(s_{0}, \diamond \square B_{4}\right)=1\).

Exercise 10.3. Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite Markov chain, \(s \in S\) and \(C, B \subseteq S\), \(n \in \mathbb{N}, n \geqslant 1\). Let \(C \mathrm{U}^{=n} B\) denote the event that a \(B\)-state will be entered after exactly \(n\) steps and all states that are visited before belong to \(C\). That is, \(s_{0} s_{1} s_{2} \ldots \vDash C \mathrm{U}^{=n} B\) if and only if
\(s_{n} \in B\) and \(s_{i} \in C\) for \(0 \leqslant i<n\). The event \(C \mathrm{U}^{\geqslant n} B\) denotes the union of the events \(C \mathrm{U}^{=k} B\) where \(k\) ranges over all natural numbers \(\geqslant n\).

Question: Provide an algorithm to compute
(a) the values \(\operatorname{Pr}\left(s \models C \mathrm{U}^{=n} B\right)\), and
(b) an algorithm for computing \(\operatorname{Pr}\left(s \models C \mathrm{U}^{\geqslant n} B\right)\).

Exercise 10.4. Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite Markov chain and \(s \in S, a, b \in A P\). Prove or disprove the following statements:
(a) \(\operatorname{Pr}(s \models \square a)=1\) iff \(s \models \forall \square a\)
(b) \(\operatorname{Pr}(s \models \diamond a)<1\) iff \(s \not \models \forall \diamond a\)
(c) \(\operatorname{Pr}(s \models \square a)>0\) iff \(s \models \exists \square a\)
(d) \(\operatorname{Pr}(s \models \diamond \square a)=1\) iff \(\operatorname{Pr}(s \models \diamond B)=1\) where \(B=\{t \in S \mid t \models \forall \square a\}\)
(e) \(\operatorname{Pr}(s \models a \cup b)=1\) iff \(s \models \forall(a \cup b)\)
(f) \(\operatorname{Pr}(s \models a \cup b)=0\) iff \(s \not \models \exists(a \cup b)\)

EXERCISE 10.5. Complete the proof of Corollary 10.33 on page 778 and provide the proof for the following statement:
\[
\operatorname{Pr}(s \models \square \diamond B)=1 \quad \text { iff } \quad s \models \forall \square \exists \diamond B
\]
provided that \(s\) is a state of a finite Markov chain \(\mathcal{M}\) and \(B\) a set of states in \(\mathcal{M}\).

Exercise 10.6. Consider the following MC:


Question: Determine the set of states for which the PCTL formula \(\mathbb{P}_{\geqslant p}\left(b \cup_{c}\right)\) holds for \(p=\frac{17}{19}\).

Exercise 10.7. Prove or disprove the following PCTL equivalences:
(a) \(\mathbb{P}_{=1}\left(\bigcirc \mathbb{P}_{=1}(\square a)\right) \equiv \mathbb{P}_{=1}\left(\square \mathbb{P}_{=1}(\bigcirc a)\right)\)
(b) \(\mathbb{P}_{>0.5}\left(\bigcirc \mathbb{P}_{>0.5}(\diamond a)\right) \equiv \mathbb{P}_{>0.5}\left(\diamond \mathbb{P}_{>0.5}(\bigcirc a)\right)\)
(c) \(\mathbb{P}_{=1}\left(\bigcirc \mathbb{P}_{=1}(\diamond a)\right) \equiv \mathbb{P}_{=1}\left(\diamond \mathbb{P}_{=1}(\bigcirc a)\right)\)

Exercise 10.8. Let \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a finite Markov chain such that \(S \subseteq A P\) and \(L(s) \cap A P=\{s\}\) for each state \(s \in S\). Let sfair be the following CTL fairness assumption:
\[
\text { sfair }=\bigwedge_{s \in S} \bigwedge_{t \in \operatorname{Post}(s)}(\square \diamond s \rightarrow \square \diamond t) .
\]

Show that for \(a, b \in A P\) :
(a) \(s \models \mathbb{P}_{=1}(a \cup b) \quad\) iff \(s \models_{\text {sfair }} \forall(a \cup b)\).
(b) \(s \models \mathbb{P}_{>0}(\square a) \quad\) iff \(\quad s \models_{\text {sfair }} \exists \square a\).

ExErcise 10.9. Provide definitions of the weak until operator and the release operator in PCTL. That is, define PCTL formulae \(\mathbb{P}_{J}(\Phi \mathrm{~W} \Psi)\) and \(\mathbb{P}_{J}(\Phi \mathrm{R} \Psi)\) such that for each Markov chain \(\mathcal{M}\) and each state \(s\) in \(\mathcal{M}\) :
\[
\begin{array}{lll}
s \models \mathbb{P}_{J}(\Phi \mathrm{~W} \Psi) & \text { iff } & \operatorname{Pr}(s \models \Phi \mathrm{~W} \Psi) \in J, \\
s \models \mathbb{P}_{J}(\Phi \mathrm{R} \Psi) & \text { iff } & \operatorname{Pr}(s \models \Phi \mathrm{R} \Psi) \in J,
\end{array}
\]
where
\[
\begin{aligned}
\operatorname{Pr}(s \models \Phi \mathrm{~W} \Psi) & =\operatorname{Pr}_{s}^{\mathcal{M}}\{\pi \in \operatorname{Paths}(s) \mid \pi \models \Phi \cup \Psi \vee \pi \models \square \Phi\}, \\
\operatorname{Pr}(s \models \Phi \mathrm{R} \Psi) & =\operatorname{Pr}_{s}^{\mathcal{M}}\{\pi \in \operatorname{Paths}(s) \mid \pi \not \models \neg \Phi \cup \neg \Psi\} .
\end{aligned}
\]

Exercise 10.10. As for CTL and LTL, PCTL in positive normal form (PNF) can be defined as a fragment of PCTL where negation is only allowed adjacent to atomic propositions. To avoid a decrease in expressiveness, the syntax of PCTL formulae in PNF contains for each operator of the base syntax of PCTL a dual operator (e.g., disjunction as the dual for conjunction, release as the dual for until, etc.).

Questions:
(a) Provide the precise definition of the syntax of PCTL formulae, and
(b) Prove that for any PCTL formula there exists an equivalent PCTL formula in PNF.

Exercise 10.11. Let \(\mathcal{M}\) be a finite Markov chain over \(A P, s\) a state of \(\mathcal{M}\) and \(a, b \in A P\) : Show that
\[
s \models \mathbb{P}_{=1}(a \cup b) \quad \text { iff } \quad s \models \forall((\exists(a \cup b)) \mathrm{W} b) .
\]

Exercise 10.12. Provide deterministic Rabin automata for the following LTL formulae: \(\square(a \rightarrow\) \(\diamond b), \neg \square(a \rightarrow \diamond b)\), and \(a \cup(\square b)\).

ExERCISE 10.13. Let \(\mathcal{A}_{1}\) and \(\mathcal{A}_{2}\) be DRA over the same alphabet. Define the DRA \(\mathcal{A}=\mathcal{A}_{1} \cup \mathcal{A}_{2}\) such that
\[
\mathcal{L}_{\omega}(\mathcal{A})=\mathcal{L}_{\omega}\left(\mathcal{A}_{1}\right) \cup \mathcal{L}_{\omega}\left(\mathcal{A}_{2}\right)
\]
and \(\operatorname{size}(\mathcal{A})=\mathcal{O}\left(\operatorname{poly}\left(\operatorname{size}\left(\mathcal{A}_{1}\right), \operatorname{size}\left(\mathcal{A}_{2}\right)\right)\right)\).

Exercise 10.14. Consider the Markov chain \(\mathcal{M}\) in Figure 10.22 (page 899), and let the labeling be given by \(L\left(s_{2}\right)=L\left(s_{3}\right)=L\left(s_{4}\right)=\{a\}\) and \(L(s)=\varnothing\) for the remaining states. Question: Compute the probability \(\operatorname{Pr}^{\mathcal{M}}(\varphi)\) for the LTL formula \(\varphi=\square \diamond a\). (Hint: Construct a DRA \(\mathcal{A}\) for \(\varphi\) and perform a quantitative analysis in \(\mathcal{M} \otimes \mathcal{A}\).)

Exercise 10.15. Prove the following statement. For finite Markov chain \(\mathcal{M}=\left(S, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) and \(\omega\)-regular property \(P\) over \(A P\) such that \(\operatorname{Pr}^{\mathcal{M}}(P)>0\), there exists a finite path fragment \(\widehat{\pi}=s_{0} s_{1} \ldots s_{n}\) in \(\mathcal{M}\) with \(\iota_{\text {init }}\left(s_{0}\right)>0\) such that almost all paths in the cylinder set \(C y l(\widehat{\pi})\) fulfill \(P\), i.e.,
\[
\operatorname{Pr}^{\mathcal{M}}\{\pi \in \operatorname{Cyl}(\widehat{\pi}) \mid \operatorname{trace}(\pi) \in P\}=\mathbf{P}(\widehat{\pi}) .
\]

Exercise 10.16. Show that there is no PCTL formula that is equivalent to the PCTL* formula \(\mathbb{P}_{\geqslant 0.5}(\bigcirc \bigcirc a)\) where \(a\) is an atomic proposition.

Exercise 10.17. Consider the Markov chain \(\mathcal{M}\), which is given by


Questions:
(a) Determine the bisimulation quotient \(\mathcal{M} / \sim_{\mathcal{M}}\)
(b) For each pair of equivalence classes \(C\) and \(D\) under \(\sim_{\mathcal{M}}\), give a PCTL formula \(\Phi\) such that \(C \models \Phi\) and \(D \not \models \Phi\).

Exercise 10.18. Let ( \(\mathcal{M}\), rew) be a finite Markov reward model and \(s\) a state in \(\mathcal{M}, T\) a set of states in \(\mathcal{M}\) such that \(\operatorname{Pr}(s \models \diamond T)=1\). Show that the infinite series
\[
\operatorname{ExpRew}(s \models \diamond T)=\sum_{r=0}^{\infty} r \cdot \operatorname{Pr}_{s}\{\pi \in \operatorname{Paths}(s) \mid \operatorname{rew}(\pi, \diamond B)=r\}
\]
converges.

Exercise 10.19. Consider the Markov reward model for the simulation of a die by coin-tossing actions as in Example 10.75 on page 825. Show that for each \(r \in \mathbb{N}, r \geqslant 0\) :
\[
\operatorname{Pr}\left(s_{0} \models \diamond\{1,2,3,4,5,6\}\right)=1-\frac{1}{4^{r}}
\]

Exercise 10.20. In Section 10.5, we only considered rewards for the states. Let us now study reward models that rely on a finite Markov chain where the edges are augmented by rewards. I.e., we deal with reward functions rew : \(S \times S \rightarrow \mathbb{N}\) such that \(\operatorname{rew}\left(s, s^{\prime}\right)=0\) if \(\mathbf{P}\left(s, s^{\prime}\right)=0\). The cumulative reward of a finite path fragment \(s_{0} s_{1} \ldots s_{n}\) is now defined as the sum of the reward earned by traversing the edges \(\left(s_{0}, s_{1}\right),\left(s_{1}, s_{2}\right), \ldots,\left(s_{n-1}, s_{n}\right)\). That is,
\[
\operatorname{rew}\left(s_{0} s_{1} \ldots s_{n}\right)=\sum_{1 \leqslant i \leqslant n} \operatorname{rew}\left(s_{i-1}, s_{i}\right)
\]

Expected rewards \(\operatorname{Exp} \operatorname{Rew}(s \models \diamond T)\) and reward-bounded reachability events \(\diamond_{\leqslant r} T\) are defined as in Section 10.5. Provide algorithms to compute expected rewards ExpRew \((s \models \diamond T)\), and
probabilities for reward-bounded reachability properties \(\left.\operatorname{Pr}(s \models\rangle_{\leqslant r} T\right)\) for finite Markov chains with reward structures that attach rewards to the edges.

Exercise 10.21. Provide a definition of bisimulation equivalence on Markov reward models such that bisimulation equivalence agrees with PRCTL-equivalence. Prove the correctness of your notion of bisimulation equivalence.

Exercise 10.22. Consider the MDP \(\mathcal{M}\) shown in Figure 10.24. Let \(B=\left\{s_{6}\right\}\). Compute the values \(x_{s}=\operatorname{Pr}^{\max }(s \vDash \Delta B)\) where \(s\) is a state in the MDP \(\mathcal{M}\). Take the following approach: first determine the states for which \(\operatorname{Pr}^{\max }(s \models \diamond B) \in\{0,1\}\); then solve the corresponding linear program for the remaining states.


Figure 10.24: Markov decision process \(\mathcal{M}\) for Exercise 10.22.

Exercise 10.23. Given a finite Markov decision process \(\mathcal{M}\) with state space \(S\) and a subset \(B\) of \(S\), show that there exists a memoryless scheduler \(\mathfrak{S}\) such that
\[
\operatorname{Pr}^{\max }(s \models \square \diamond B)=\operatorname{Pr}^{\mathfrak{G}}(s \models \square \diamond B) .
\]

For which other types of limit LT properties \(P\) exist memoryless schedulers that maximize the probabilities for \(P\) ? Consider
(a) persistence properties \(\diamond \square B\),
1. Rabin conditions \(\underset{1 \leqslant i \leqslant k}{\bigvee}\left(\diamond \square B_{i} \wedge \square \diamond C_{i}\right)\),
(b) strong fairness assumptions \(\bigwedge_{1 \leqslant i \leqslant k}\left(\square \diamond B_{i} \rightarrow \square \diamond C_{i}\right)\).

Exercise 10.24. Show or disprove that the class of limit LT properties is closed under intersection. Do the same for union and complementation.

Exercise 10.25. Determine the maximal end components of the following MDP:


Exercise 10.26. Let \(\mathcal{M}\) be a finite MDP over \(A P\) and \(\mathcal{A}\) a DRA over the alphabet \(2^{A P}\). On page 882 we described a transformation "scheduler \(\mathfrak{S}\) for \(\mathcal{M} \rightsquigarrow\) scheduler \(\mathfrak{S}^{\prime}\) for \(\mathcal{M} \otimes \mathcal{A}\) ", and vice versa. Questions:
(a) Is it true that if \(\mathfrak{S}\) is memoryless, then so is \(\mathfrak{S}^{\prime}\) ?
(b) Is it true that if \(\mathfrak{S}^{\prime}\) is memoryless, then so is \(\mathfrak{S}\) ?
(c) Assume that \(\mathfrak{S}^{\prime}\) is a finite-memory scheduler for \(\mathcal{M} \otimes \mathcal{A}\). Describe the corresponding finitememoryless scheduler \(\mathfrak{S}\) by its modes.

Exercise 10.27. Let \(\mathcal{M}=\left(S, A c t, \mathbf{P}, \iota_{\text {init }}, A P, L\right)\) be a Markov decision process. A bisimulation on \(\mathcal{M}\) is an equivalence \(\mathcal{R}\) on \(S\) such that for all \(\left(s_{1}, s_{2}\right) \in \mathcal{R}\) :
(a) \(L\left(s_{1}\right)=L\left(s_{2}\right)\)
(b) for all \(\alpha \in \operatorname{Act}\left(s_{1}\right)\) there exists \(\beta \in \operatorname{Act}\left(s_{2}\right)\) such that for all \(\mathcal{R}\)-equivalence classes \(T \in S / \mathcal{R}\) :
\[
\mathbf{P}\left(s_{1}, \alpha, T\right)=\mathbf{P}\left(s_{2}, \beta, T\right) .
\]

States \(s_{1}, s_{2}\) of \(\mathcal{M}\) are called bisimulation equivalent iff there exists a bisimulation \(\mathcal{R}\) with \(\left(s_{1}, s_{2}\right) \in\) \(\mathcal{R}\).
(a) Show that bisimulation-equivalent states satisfy the same PCTL* formulae over \(A P\).
(b) Suppose that \(\mathcal{M}\) is finite. Show that PCTL-equivalent states are bisimulation equivalent.

In (b), PCTL-equivalence of two states \(s_{1}, s_{2}\) of \(\mathcal{M}\) means that \(s_{1}\) and \(s_{2}\) fulfill the same PCTL formulae over \(A P\).

Exercise 10.28. Let \(\mathcal{M}\) be a finite Markov decision process and fair an LTL fairness assumption. Prove the equivalence of the following statements:
(a) There exists a fair scheduler for \(\mathcal{M}\).
(b) There exists a fair finite-memory scheduler for \(\mathcal{M}\).
(c) There exists a scheduler \(\mathfrak{S}\) for \(\mathcal{M}\) such that, for all states \(s\) of \(\mathcal{M}, \operatorname{Pr}^{\max }\left(s \models \diamond U_{f a i r}\right)=1\), where \(U_{\text {fair }}\) denotes the success set of fair, i.e., the union of the sets \(T\) of all end components \((T, A)\) of \(\mathcal{M}\) such that \(T \models\) fair.

Here, fairness of a scheduler \(\mathfrak{F}\) for \(\mathcal{M}\) means that for all states \(s\) of \(\mathcal{M}\) it holds that
\[
\operatorname{Pr}_{s}^{\mathfrak{F}}\{\pi \in \operatorname{Paths}(s)|\pi|=\text { fair }\}=1
\]

Exercise 10.29. Let \(\mathcal{M}\) be a finite Markov decision process and
\[
\text { sfair }=\bigwedge_{1 \leqslant i \leqslant k}\left(\square \diamond a_{i} \rightarrow \square \diamond b_{i}\right)
\]
a strong fairness assumption. Questions:
(a) Design an algorithm that runs in time \(\mathcal{O}(\operatorname{poly}(\operatorname{size}(\mathcal{M})) \cdot k)\) and computes all end components \((T, A)\) of \(\mathcal{M}\) such that \(T \models\) sfair and such that \((T, A)\) is not contained in another end component \(\left(T^{\prime}, A^{\prime}\right)\) with \(T^{\prime} \models\) sfair.
(b) Design an algorithm that takes as input a finite MDP \(\mathcal{M}\) and an LTL fairness assumption sfair and checks in time \(\mathcal{O}(\operatorname{poly}(\operatorname{size}(\mathcal{M})) \cdot|\varphi|)\) whether \(\varphi\) is realizable for \(\mathcal{M}\).
(Hint for part (a): combine the ideas of the algorithm to compute the maximal end components (Algorithm 47) with the ideas of model checking CTL with fairness.)

\section*{Appendix A}

\section*{Appendix: Preliminaries}

\section*{A. 1 Frequently Used Symbols and Notations}

The symbol \(\square\) only serves readability. It indicates the end of a definition, of a remark, or of other numbered text fragments. The abbreviation "iff" stands for "if and only if". We also often use logical symbols, such as \(\wedge\) for "and", \(\vee\) for "or" and the following quantifiers:
\(\exists \quad\) "there exists ..."
\(\forall \quad\) "for all"
\(\stackrel{\infty}{\exists}\) "there exist infinitely many"
\(\stackrel{\infty}{\forall}\) "for almost all, i.e., for all except for finitely many".
The Greek Alphabet At various places, Latin and Greek letters serve as symbols for certain mathematical objects. Although not all Greek letters will be used in this monograph, Figure A. 1 shows the complete Greek alphabet. The symbols \(\Gamma, \Delta, \Theta, \Lambda, \Xi, \Sigma, \Upsilon\), \(\Phi, \Psi\), and \(\Omega\) are capital letters. All other symbols in Figure A. 1 are lowercase letters.

Natural and Real Numbers The symbol \(\mathbb{N}\) denotes the set \(\{0,1,2, \ldots\}\) of natural numbers. The set of real numbers is denoted \(\mathbb{R}\). Subsets of \(\mathbb{N}\) and \(\mathbb{R}\) are often denoted by subscripts. For instance, \(\mathbb{R}_{\geqslant 0}\) denotes the set of non-negative reals, while \(\mathbb{R}_{>5}\) denotes the interval \(] 5, \infty[\).
\begin{tabular}{llllll}
\(\alpha\) & alpha & \(\iota\) & iota & \(\rho, \varrho\) & rho \\
\(\beta\) & beta & \(\kappa\) & kappa & \(\sigma, \varsigma, \Sigma\) & sigma \\
\(\gamma, \Gamma\) & gamma & \(\lambda, \Lambda\) & lambda & \(\tau\) & tau \\
\(\delta, \Delta\) & delta & \(\mu\) & mu & \(v, \Upsilon\) & upsilon \\
\(\epsilon, \varepsilon\) & epsilon & \(\nu\) & nu & \(\phi, \varphi, \Phi\) & phi \\
\(\zeta\) & zeta & \(\xi, \Xi\) & xi & \(\chi\) & chi \\
\(\eta\) & eta & 0 & omicron & \(\psi, \Psi\) & psi \\
\(\theta, \vartheta, \Theta\) & theta & \(\pi, \varpi, \Pi\) & pi & \(\omega, \Omega\) & omega
\end{tabular}

Figure A.1: The Greek alphabet.

Asymptotic operators \(\mathcal{O}, \Omega\), and \(\Theta\) To specify the asymptotic growth of (cost) functions, the Landau symbols \(\mathcal{O}, \Omega\), and \(\Theta\) are used (see, e.g., [100]):
\(\mathcal{O}\) : asymptotic upper bound
\(\Omega\) : asymptotic lower bound
\(\Theta\) : asymptotic upper and lower bound
The precise meaning is as follows. If \(f: \mathbb{N} \rightarrow \mathbb{N}\) is a function, then \(\mathcal{O}(f)\) denotes the set of all functions \(g: \mathbb{N} \rightarrow \mathbb{N}\) such that there exists a constant \(C>0\) and natural number \(N\) with \(g(n) \leqslant C \cdot f(n)\) for all \(n \in \mathbb{N}\) where \(n \geqslant N\). The function class \(\Omega(f)\) consists of all functions \(g: \mathbb{N} \rightarrow \mathbb{N}\) such that there exists a constant \(C>0\) and a natural number \(N\) with \(g(n) \geqslant C \cdot f(n)\) for all \(n \geqslant N\). The class \(\Theta(f)\) is given by \(\Theta(f)=\mathcal{O}(f) \cap \Omega(f)\). It is common to use the equality symbol instead of the element symbol " \(\in\) " and to write, e.g., \(g=\mathcal{O}(f)\) or \(g(n)=\mathcal{O}(f(n))\) rather than \(g \in \mathcal{O}(f)\). Thus, "equations" with the asymptotic operators have to be read from the left to the right.

We write \(\mathcal{O}(\operatorname{poly}(n))\) to denote the class of all functions \(g: \mathbb{N} \rightarrow \mathbb{N}\) that are polynomially bounded, i.e., \(g(n)=\mathcal{O}\left(n^{k}\right)\) for some natural number \(k\). Similarly, \(\mathcal{O}(\exp (n))\) denotes the class of all functions \(g: \mathbb{N} \rightarrow \mathbb{N}\) that are exponentially bounded, i.e., \(g(n)=\mathcal{O}\left(a^{n^{k}}\right)\) where \(a>1\) is a real constant and \(k\) a natural number.

Notation for Sets Let \(X\) be a set. The symbol \(2^{X}\) stands for the powerset of \(X\), i.e., the set consisting of all subsets of \(X\). We write \(|X|\) to denote the cardinality of \(X\), i.e., the number of elements of \(X\). (If \(X\) is infinite, then \(|X|=\omega\).) Besides the standard symbols \(\cup\) for union, \(\cap\) for intersection, and \(\backslash\) for "set-minus", i.e., \(X \backslash Y=\{x \in X \mid x \notin Y\}\), the symbol \(\uplus\) will be used which denotes disjoint union. Formally, for sets \(X, Y, X \uplus Y\) is defined by \(\{(x, 1) \mid x \in X\} \cup\{(y, 2) \mid y \in Y\}\). Additionally, we mention special cases for union and intersection. For \(x \in X\), let \(Y_{x}\) be subset of a set \(Y\). Then,
\[
\bigcup_{x \in \varnothing} Y_{x}=\varnothing, \quad \bigcap_{x \in \varnothing} Y_{x}=Y
\]

Thus, \(\bigcap_{x \in \varnothing} \ldots\) depends on the chosen universal set \(Y\).

Relations A relation denotes any set of tuples of fixed length. More precisely, let \(X_{1}, \ldots, X_{k}\) be sets where \(k \in \mathbb{N}\) with \(k \geqslant 1\). Subsets of the Cartesian products \(X_{1} \times \ldots \times X_{k}\) are also called relations or predicates. The number \(k\) denotes the arity. If \(X_{1}=X_{2}=\) \(\ldots=X_{k}=X\), then the subsets of
\[
X^{k}=X_{1} \times \ldots \times X_{k}
\]
are called \(k\)-ary relations on \(X\). Very often we have to deal with binary (2-ary) relations \(\mathcal{R}\) over a set \(X\). For these relations the infix notation \(x \mathcal{R} y\) is often used instead of \((x, y) \in \mathcal{R}\). If \(X\) is a set and \(\mathcal{R}\) a binary relation on \(X\), then \(\mathcal{R}\) is called
- transitive if for all \(x, y, z \in X, x \mathcal{R} y\), and \(y \mathcal{R} z\) implies \(x \mathcal{R} z\);
- reflexive if for all \(x \in X, x \mathcal{R} x\);
- symmetric if for all \(x, y \in X, x \mathcal{R} y\) implies \(y \mathcal{R} x\);
- antisymmetric if for all \(x, y \in X, x \mathcal{R} y\), and \(y \mathcal{R} x\) implies \(x=y\).

Equivalences An equivalence relation (or briefly equivalence) on \(X\) is a transitive, reflexive, and symmetric binary relation on \(X\). Symmetric symbols (like \(\sim\), \(\equiv\), or similar symbols) are often used to denote an equivalence relation. If \(x \in X\), then \([x]_{\mathcal{R}}=\{y \in\) \(X \mid x \mathcal{R} y\}\) is called the equivalence class of \(x\) with respect to \(\mathcal{R}\). If \(\mathcal{R}\) follows from the context, \([x]\) is often written for short instead of \([x]_{\mathcal{R}}\). The quotient space of \(X\) with respect to \(\mathcal{R}\) is the set of all equivalence classes with respect to \(\mathcal{R}\) and is denoted \(X / \mathcal{R}\). Then, \(X / \mathcal{R}=\left\{[x]_{\mathcal{R}} \mid x \in X\right\}\). If \(\mathcal{R}\) is an equivalence relation on \(X\), then for all \(x, y \in X\) :
\[
x \mathcal{R} y \text { iff }[x]_{\mathcal{R}}=[y]_{\mathcal{R}} \text { iff }[x]_{\mathcal{R}} \cap[y]_{\mathcal{R}} \neq \varnothing
\]

So the quotient space consists of pairwise disjoint nonempty subsets of \(X\) whose union just results in \(X\). For each element \(x \in X\), there is always exactly one element \(A\) of the quotient space with \(x \in A\) (i.e. \(A=[x]_{\mathcal{R}}\) ). The index of an equivalence relation \(\mathcal{R}\) denotes the number of equivalence classes, i.e., the cardinality of \(X / \mathcal{R}\). It is usual to use spellings like " \(\mathcal{R}\) is of finite index" if the index of \(\mathcal{R}\) is finite (a natural number).

Let \(\mathcal{R}\) and \(\mathcal{R}^{\prime}\) be two equivalence relations on \(X\). Relation \(\mathcal{R}\) is a refinement of \(\mathcal{R}^{\prime}\) if \(\mathcal{R} \subseteq \mathcal{R}^{\prime}\), i.e., \(\mathcal{R}\) "distinguishes" more elements than \(\mathcal{R}^{\prime}\). In this case, it is also said that \(\mathcal{R}\) is finer than \(\mathcal{R}^{\prime}\) and that \(\mathcal{R}^{\prime}\) is coarser than \(\mathcal{R}\). It is referred to as a proper refinement if \(\mathcal{R}\) is a refinement of \(\mathcal{R}^{\prime}\) with \(\mathcal{R} \neq \mathcal{R}^{\prime}\). If \(\mathcal{R}\) is a refinement of \(\mathcal{R}^{\prime}\), then every equivalence class with respect to \(\mathcal{R}\) is contained in exactly one equivalence class with respect to \(\mathcal{R}^{\prime}\), because we have \([x]_{\mathcal{R}} \subseteq[x]_{\mathcal{R}^{\prime}}\). This observation can be enforced as follows. Each equivalence class with respect to \(\mathcal{R}^{\prime}\) can be written as a disjoint union of equivalence classes with respect
to \(\mathcal{R}\). Thus, if \(\mathcal{R}\) is a refinement of \(\mathcal{R}^{\prime}\), then \(\left|X / \mathcal{R}^{\prime}\right| \leqslant|X / \mathcal{R}|\). Therefore, the index of \(\mathcal{R}^{\prime}\) is at most the index of \(\mathcal{R}\).

Transitive and Reflexive Closure Let \(\mathcal{R}\) be a binary relation over \(X\). The transitive and reflexive closure of \(\mathcal{R}\) is the smallest transitive, reflexive binary relation on \(X\) containing \(\mathcal{R}\). Usually, this is denoted by \(\mathcal{R}^{*}\). Thus,
\[
\mathcal{R}^{*}=\bigcup_{n \geqslant 0} \mathcal{R}^{n}
\]
where \(\mathcal{R}^{0}=\{(x, x) \mid x \in X\}\) and \(\mathcal{R}^{n+1}=\{(x, y) \in X \times X \mid \exists z \in X .(x, z) \in \mathcal{R} \wedge(z, y) \in\) \(\left.\mathcal{R}^{n}\right\}\). If \(\mathcal{R}\) is symmetric, then \(\mathcal{R}^{*}\) is an equivalence relation.

Partitions A partition (or partitioning) of a set \(X\) is a set \(\mathcal{B} \subseteq 2^{X}\) consisting of pairwise disjoint, nonempty subsets of \(X\) such that \(\bigcup_{B \in \mathcal{B}} B=X\). Elements of \(\mathcal{B}\) are called blocks, as well. In particular, each element \(x \in X\) is included in exactly one block \(B \in \mathcal{B}\). There is a close connection between equivalence relations on \(X\) and partitions of \(X\). If \(\mathcal{R}\) is an equivalence relation on \(X\), then the quotient space \(X / \mathcal{R}\) is a partition of \(X\). Vice versa, if \(\mathcal{B}\) is a partition of \(X\), then
\[
\{(x, y) \mid x \text { and } y \text { are in the same block } B \in \mathcal{B}\}
\]
is an equivalence relation with quotient space \(\mathcal{B}\).
Preorder A preorder \(\mathcal{R}\) on \(X\) denotes a reflexive, transitive relation on \(X\). Any preorder induces an equivalence, the so-called kernel of \(\mathcal{R}\), which is given by \(\mathcal{R} \cap \mathcal{R}^{-1}\), i.e., the relation \(\{(x, y) \mid x \mathcal{R} y\) and \(y \mathcal{R} x\}\).

\section*{A. 2 Formal Languages}

This section summarizes the main concepts of regular languages. Further details and the proofs of the results mentioned here can be found in any text book on formal language theory, see e.g., [214, 272, 363, 383].

Words over an Alphabet An alphabet is an arbitrary nonempty and finite set \(\Sigma\). The elements of \(\Sigma\) are typically called symbols or letters. A word over \(\Sigma\) denotes a finite or infinite sequence of symbols in \(\Sigma\), i.e., words have the form \(w=A_{1} A_{2} \ldots A_{n}\) where \(n \in \mathbb{N}\) or \(\sigma=A_{1} A_{2} A_{3} \ldots\) and where the \(A_{i}\) 's are symbols in \(\Sigma\). \({ }^{1}\) The special case \(n=0\) is

\footnotetext{
\({ }^{1}\) Formally, infinite words can be defined as functions \(\sigma: \mathbb{N} \rightarrow \Sigma\) and the notation \(\sigma=A_{1} A_{2} A_{3} \ldots\) means that \(\sigma(i)=A_{i}\) for all \(i \in \mathbb{N}\). Similarly, finite words are obtained by functions \(w:\{1, \ldots, n\} \rightarrow \Sigma\).
}
allowed, in which case the so-called empty word, denoted \(\varepsilon\), is obtained. The length of a word is the number of symbols in the given word. Thus, for \(w=A_{1} A_{2} \ldots A_{n}\), the length is \(n\), while each infinite word has the length \(\omega\). (The Greek letter \(\omega\) (omega) is typically used to denote "infinity".) \(\Sigma^{*}\) denotes the set consisting of all finite words over \(\Sigma\), while \(\Sigma^{\omega}\) denotes the set of all infinite words over \(\Sigma\). Note that \(\varepsilon \in \Sigma^{*}\). Thus, the set of all nonempty finite words, denoted \(\Sigma^{+}\), is \(\Sigma^{*} \backslash\{\varepsilon\}\). A set of finite words over the alphabet \(\Sigma\) is called a language, and is ranged over by \(\mathcal{L}\) (and primed and subscripted versions thereof).

A prefix of a finite word \(w=A_{1} A_{2} \ldots A_{n}\), is a word \(v\) of the form \(A_{1} A_{2} \ldots A_{i}\) for some \(i\) where \(0 \leqslant i \leqslant n\). A suffix of \(w\) is a word \(v\) of the form \(A_{i} A_{i+1} \ldots A_{n}\) where \(1 \leqslant i \leqslant\) \(n+1\). (In particular, \(\varepsilon\) is a prefix and suffix of any finite word.) The words of the form \(A_{i} A_{i+1} \ldots A_{j}\) with \(1 \leqslant i \leqslant j \leqslant n\) are called subwords of \(w\). The definition of prefixes, suffixes, and subwords of infinite words are similar. For infinite word \(\sigma=A_{0} A_{1} A_{2} \ldots\), the suffix \(A_{j} A_{j+1} A_{j+2} \ldots\) is denoted \(\sigma[j .\).\(] . For technical reasons, we start with the index\) 0 . Thus, \(\sigma=\sigma[0 .\).\(] .\)

Operations on Words Important operations on words are concatenation and finite repetition. Concatenation takes two words and "glues them together" to construct a new word. It is denoted by juxtaposition. For instance, the concatenation of the words \(B A\) and \(A A B\) yields the word \(B A . A A B=B A A A B\). The concatenation of a word with itself is denoted by squaring, e.g., \((A B)^{2}\) equals \(A B A B\); this is generalized in a straightforward manner for arbitrary \(n\). In the special cases \(n=0\) and \(n=1\), we have \(w^{0}=\varepsilon\) (the empty word) and \(w^{1}=w\).

Finite repetition, also called Kleene star and denoted by *, of a finite word w yields the language consisting of all finite words that arise by zero or more (but finitely many) repetitions of \(w\). Formally, for \(w \in \Sigma^{*}\) we have \(w^{*}=\left\{w^{i} \mid i \in \mathbb{N}\right\}\). For instance, \((A B)^{*}=\{\varepsilon, A B, A B A B, A B A B A B, \ldots\}\). Note that the empty word \(\varepsilon\) is included in \(w^{*}\) for each finite word \(w\). This is precisely the difference with the slight variant of the Kleene star, denoted \({ }^{+}\), defined by \(W^{+}=\left\{w^{i} \mid i \in \mathbb{N}, i \geqslant 1\right\}\), or, equivalently, \(W^{+}=w^{*} \backslash\{\varepsilon\}\). For instance, \((A B)^{+}\)denotes the set \(\{A B, A B A B, A B A B A B, \ldots\}\).

Operations on Languages Concatenation is lifted to languages in a natural way as a pointwise extension of concatenation on words. The same applies to repetition. For languages \(\mathcal{L}, \mathcal{L}_{1}, \mathcal{L}_{2} \subseteq \Sigma^{*}\) we have
\[
\mathcal{L}_{1} \cdot \mathcal{L}_{2}=\left\{w_{1} \cdot w_{2} \mid w_{1} \in \mathcal{L}_{1}, w_{2} \in \mathcal{L}_{2}\right\}
\]
and
\[
\mathcal{L}^{*}=\bigcup_{i=0}^{\infty} \mathcal{L}^{i}, \quad \text { and } \quad \mathcal{L}^{+}=\bigcup_{i=1}^{\infty} \mathcal{L}^{i}
\]
where \(\mathcal{L}^{i}\) denotes the concatenation of \(i\) times \(\mathcal{L}\). Note that the star and plus notation for
languages over finite words is consistent with the standard notations \(\Sigma^{*}\) and \(\Sigma^{+}\)for the set of all finite words over \(\Sigma\) with or without the empty word, respectively. For instance, for \(\mathcal{L}_{1}=\{A, A B\}\) and \(\mathcal{L}_{2}=\{\varepsilon, B B B\}\) we have that
\[
\begin{aligned}
\mathcal{L}_{1} \cdot \mathcal{L}_{2} & =\{A, A B, A B B B, A B B B B\}, \text { and } \\
\mathcal{L}_{1}^{2} & =\{A A, A A B, A B A B, A B A\}
\end{aligned}
\]

There are several equivalent formalisms to describe regular languages. In this monograph we only use regular expressions and finite automata. We start with the former and introduce the automata approach later.

Regular Expressions Regular expressions (denoted E or F) are built from the symbols \(\underline{\varnothing}\) (to denote the empty language), \(\underline{\varepsilon}\) (to denote the language \(\{\varepsilon\}\) consisting of the empty word), the symbols \(\underline{A}\) for \(A \in \Sigma\) (for the singleton sets \(\{A\}\) ), and the language operators "+" (union), "*" (Kleene star, finite repetition), and "." (concatenation). Formally, regular expressions can be defined inductively:
1. \(\underline{\varnothing}\) and \(\underline{\varepsilon}\) are regular expressions over \(\Sigma\).
2. If \(A \in \Sigma\), then \(\underline{A}\) is a regular expression over \(\Sigma\).
3. If \(E, E_{1}\) and \(E_{2}\) are regular expressions over \(\Sigma\), then so are \(E_{1}+E_{2}, E_{1} \cdot E_{2}\), and \(E^{*}\).
4. Nothing else is a regular expression over \(\Sigma\).
\(\mathrm{E}^{+}\)is an abbreviation for the regular expression E.E*. The semantics of a regular expression \(E\) is a language \(\mathcal{L}(E) \subseteq \Sigma^{*}\) that is defined as follows:
\[
\mathcal{L}(\underline{\varnothing})=\varnothing, \quad \mathcal{L}(\underline{\varepsilon})=\{\varepsilon\}, \quad \mathcal{L}(\underline{A})=\{A\}
\]
and
\[
\mathcal{L}\left(\mathrm{E}_{1}+\mathrm{E}_{2}\right)=\mathcal{L}\left(\mathrm{E}_{1}\right) \cup \mathcal{L}\left(\mathrm{E}_{2}\right), \quad \mathcal{L}\left(\mathrm{E}_{1} \cdot \mathrm{E}_{2}\right)=\mathcal{L}\left(\mathrm{E}_{1}\right) \cdot \mathcal{L}\left(\mathrm{E}_{2}\right), \quad \mathcal{L}\left(\mathrm{E}^{*}\right)=\mathcal{L}(\mathrm{E})^{*}
\]

From this definition, we derive \(\mathcal{L}\left(\mathrm{E}^{+}\right)=\mathcal{L}(\mathrm{E})^{+}\).
A language \(\mathcal{L} \subseteq \Sigma^{*}\) is called regular if there is some regular expression \(E\) over \(\Sigma\) such that \(\mathcal{L}(\mathrm{E})=\mathcal{L}\). For instance, \(\mathrm{E}=(\underline{A}+\underline{B})^{*} \cdot \underline{B} \cdot \underline{B} \cdot(\underline{A}+\underline{B})\) is a regular expression over \(\Sigma=\{A, B\}\) representing the language
\[
\mathcal{L}(\mathrm{E})=\left\{w B B A \mid w \in \Sigma^{*}\right\} \cup\left\{w B^{3} \mid w \in \Sigma^{*}\right\}
\]
consisting of all finite words that end with three \(B\) 's or with the word \(B B A\). The regular expression \(\mathrm{E}^{\prime}=(\underline{A}+\underline{B})^{*} \cdot \underline{B} \cdot \underline{B} \cdot(\underline{A}+\underline{B})^{*}\) represents the regular language consisting of all finite words over \(\Sigma=\{A, B\}\) that contain the subword \(B B\).

It is standard to use a simplified syntax for regular expressions that does not distinguish between the atomic expression \(\underline{x}\) and the symbol \(x\) for \(x \in\{\varnothing, \varepsilon\} \cup \Sigma\) and skips the operator symbol ". " for concatenation. For example,
\[
(A+B)^{*} B B(A+B)
\]
stands for the regular expression \((\underline{A}+\underline{B})^{*} \cdot \underline{B} \cdot \underline{B} \cdot(\underline{A}+\underline{B})\).

\section*{A. 3 Propositional Logic}

This section summarizes the basic principles of propositional logic. For more elaborate treatments we refer to the textbook [342].

Given is a finite set AP of atomic propositions, sometimes also called propositional symbols. In the following, Latin letters like \(a, b\), and \(c\) (with or without subscripts) are used to denote elements of \(A P\). The set of propositional logic formulae over AP, formulae for short, is inductively defined by the following four rules:
1. true is a formula.
2. Any atomic proposition \(a \in A P\) is a formula.
3. If \(\Phi_{1}, \Phi_{2}\) and \(\Phi\) are formulae, then so are \((\neg \Phi)\) and \(\left(\Phi_{1} \wedge \Phi_{2}\right)\).
4. Nothing else is a formula.

Any formula stands for a proposition that might hold or not, depending on which of the atomic propositions are assumed to hold. Intuitively, the formula \(a\) stands for the propositions stating that \(a\) holds. The intuitive meaning of the symbol \(\wedge\) is conjunction ("and"), i.e., \(\Phi_{1} \wedge \Phi_{2}\) holds if and only if both propositions \(\Phi_{1}\) and \(\Phi_{2}\) hold. The symbol \(\neg\) denotes negation, i.e., \(\neg \Phi\) holds if and only if \(\Phi\) does not hold. Thus, e.g., \(a \wedge \neg b\) holds if and only if \(a\) holds and \(b\) does not hold. The constant true stands for a proposition which holds in any context, independent of the interpretation of the atomic propositions \(a\).

It is standard to use simplified notations for formulae with the help of derived operators and by declaring a precedence order on the basic and derived operators, which often allows skipping brackets. The standard precedence order assigns a higher priority to the unary negation operator \(\neg\) than the binary conjunction operator \(\wedge\). Thus, \(\neg a \wedge b\) is short for
\(((\neg a) \wedge b)\). Moreover, conjunction \(\wedge\) binds stronger than the derived binary operators, such as
\[
\begin{array}{lll}
\Phi_{1} \vee \Phi_{2} & \stackrel{\text { def }}{=} \neg\left(\neg \Phi_{1} \wedge \neg \Phi_{2}\right) & \text { disjunction ("or") } \\
\Phi_{1} \rightarrow \Phi_{2} \stackrel{\text { def }}{=} \neg \Phi_{1} \vee \Phi_{2} & \text { implication } \\
\Phi_{1} \leftrightarrow \Phi_{2} \stackrel{\text { def }}{=}\left(\neg \Phi_{1} \wedge \neg \Phi_{2}\right) \vee\left(\Phi_{1} \wedge \Phi_{2}\right) & \text { equivalence } \\
\Phi_{1} \oplus \Phi_{2} \stackrel{\text { def }}{=}\left(\neg \Phi_{1} \wedge \Phi_{2}\right) \vee\left(\Phi_{1} \wedge \neg \Phi_{2}\right) & \text { parity (xor) }
\end{array}
\]

For example, \(\neg a \vee \neg b \wedge c\) is a short-form notation for \((\neg a) \vee((\neg b) \wedge c)\) which - by the definition of \(\vee-\) stands for the formula \(\Phi=(\neg(\neg a) \wedge \neg((\neg b)) \wedge c))\).

Abstract Syntax Throughout this monograph, we provide the definition of the syntax of logics in a more relaxed way. Skipping the syntactic rules for brackets (which can be derived from the precedence order of the operators that will be declared in words), the above inductive definition of propositional formulae over AP can be rewritten as
\[
\Phi \quad::=\text { true }|a| \Phi_{1} \wedge \Phi_{2} \mid \neg \Phi
\]
where \(a \in A P\). The above can be understood as a casual notation for the Backus-Naur form of a context-free grammar over the alphabet \(\Sigma=\{\) true \(\} \cup A P \cup\{\neg, \wedge\}\). In this short-form notation, the symbol \(\Phi\) serves simultaneously for (1) a nonterminal symbol (variable) of the grammar and (2) its derived words over \(\Sigma^{*}\) (i.e., propositional formulae). The latter explains the indices in the term \(\Phi_{1} \wedge \Phi_{2}\), which is correct on the formula level, although the correct notation would be \(\Phi \wedge \Phi\) (without indices) in the grammar.

Length of a Formula The length of a formula \(\Phi\) is defined by the number of operators in \(\Phi\) and is denoted by \(|\Phi|\). For instance, the formula \(\Phi=(\neg b) \wedge c\) has the length 2 . Since in most cases we are only interested in the asymptotic length of formulae in formula sequences \(\left(\Phi_{n}\right)\), we may also assign one cost unit to the derived operators \(V\) and \(\rightarrow\). In fact, the asymptotic formula length does not depend on whether \(\vee\) and \(\rightarrow\) are treated as a basic operator (with one cost unit per occurrence in a formula) or a derived one (using conjunction and negation).

Semantics of Propositional Logic To formalize the intuitive meaning of propositional formulae, we first need a precise definition of the "context" that declares which atomic propositions hold and which do not hold. This is done by means of an evaluation which assigns a truth value 0 ("false") or 1 ("true") to each atomic proposition. Formally, an evaluation for \(A P\) is a function \(\mu: A P \rightarrow\{0,1\}\). \(\operatorname{Eval}(A P)\) denotes the set of all evaluations for \(A P\). The semantics of propositional logic is specified by a satisfaction relation \(\models\) indicating the evaluations \(\mu\) for which a formula \(\Phi\) is true. Formally, \(\models\) is a set of pairs \((\mu, \Phi)\) where \(\mu\) is an evaluation and \(\Phi\) is a formula. It is written
\[
\begin{array}{llll}
\mu & =\text { true } & & \\
\mu \models a & \text { iff } & & \mu(a)=1 \\
\mu \models \neg \Phi & \text { iff } & & \mu \not \models \Phi \\
\mu & \models \Phi \wedge \Psi & \text { iff } &
\end{array} \quad \mu \models \Phi \text { and } \mu \models \Psi . ~ \$
\]

Figure A.2: The satisfaction relation \(\models\) of propositional logic.
\[
\mu \models \Phi \text { instead of }(\mu, \Phi) \in \models \text {. }
\]

Accordingly, \(\mu \not \models \Phi\) stands for \((\mu, \Phi) \notin \models\). Intuitively, \(\mu \models \Phi\) stands for the fact that \(\Phi\) is true under evaluation \(\mu\). The satisfaction relation \(\models\) is inductively defined by the conditions indicated in Figure A.2. If \(\mu \models \Phi\), then \(\mu\) is called a satisfied condition for \(\Phi\). In literature, the notation \(\mu(\Phi)=1\) if \(\mu \models \Phi\), and \(\mu(\Phi)=0\), if \(\mu \not \models \Phi\), is used, too. The value \(\mu(\Phi) \in\{0,1\}\) is called the truth-value of \(\Phi\) under \(\mu\).

Formulae with derived operators like disjunction \(\vee\) or implication \(\rightarrow\) have the expected semantics. Thus,
\[
\begin{array}{lll}
\mu \models \Phi \vee \Psi & \text { iff } & \mu \models \Phi \text { or } \mu \models \Psi \\
\mu \models \Phi \rightarrow \Psi & \text { iff } & \mu \not \models \Phi \text { or } \mu \models \Psi \\
& \text { iff } & \mu \models \Phi \text { implies } \mu \models \Psi .
\end{array}
\]

Set Notation for Evaluations An alternative representation of evaluations for \(A P\) is based upon representation of sets. Each evaluation \(\mu\) can be represented by the set \(A_{\mu}=\{a \in A P \mid \mu(a)=1\}\). And conversely, an evaluation \(\mu=\mu_{A}\) with \(A=A_{\mu}\) can be assigned to each subset \(A\) of \(A P\). Evaluation \(\mu_{A}\) is the characteristic function of \(A_{\mu}\), that is, \(\mu_{A}(a)=1\) if \(a \in A\) and \(\mu_{A}(a)=0\) if \(a \notin A\). This observation suggests extending the satisfaction relation \(\models\) to subsets of \(A P\) by
\[
A \models \Phi \quad \text { iff } \quad \mu_{A} \models \Phi
\]

As an example, we look at \(\Phi=(a \wedge \neg b) \vee c\). Given an evaluation \(\mu\) with \(\mu(a)=0\) and \(\mu(b)=\mu(c)=1\), then \(\mu \not \models a \wedge \neg b\) and \(\mu \models c\), and thus, \(\mu \models \Phi\). The accompanying set \(A_{\mu}\) is \(A_{\mu}=\{b, c\}\). Hence, \(\{b, c\} \models \Phi\). The empty set induces an evaluation \(\mu_{\varnothing}\) with \(\mu_{\varnothing}(a)=\mu_{\varnothing}(b)=\mu_{\varnothing}(c)=0\). Due to \(\mu_{\varnothing} \not \vDash \Phi\) (where \(\Phi=(a \wedge \neg b) \vee c\) as above), we get \(\varnothing \not \vDash \Phi\). However, we have \(\varnothing \models \neg a \wedge \neg b\) since \(\neg a\) and \(\neg b\) hold for the associated evaluation \(\mu_{\varnothing}\).

Semantic Equivalence Two propositional logic formulae \(\Phi, \Psi\) are called (semantically) equivalent if they have the same truth-value for each evaluation. That is, for all evaluations \(\mu\) :
\begin{tabular}{|c|c|}
\hline rule for double negation
\[
\neg \neg \Phi \equiv \Phi
\] & idempotency law
\[
\begin{aligned}
\Phi \vee \Phi & \equiv \Phi \\
\Phi \wedge \Phi & \equiv \Phi
\end{aligned}
\] \\
\hline absorption law
\[
\begin{aligned}
& \Phi \wedge(\Psi \vee \Phi) \\
& \equiv \Phi \\
& \Phi \vee(\Psi \wedge \Phi)
\end{aligned}
\] & commutativity law
\[
\begin{aligned}
& \Phi \wedge \Psi \equiv \Psi \wedge \Phi \\
& \Phi \vee \Psi \equiv \Psi \vee \Phi
\end{aligned}
\] \\
\hline associativity law
\[
\begin{aligned}
\Phi \wedge(\Psi \wedge \Xi) & \equiv(\Phi \wedge \Psi) \wedge \Xi \\
\Phi \vee(\Psi \vee \Xi) & \equiv(\Phi \vee \Psi) \vee \Xi
\end{aligned}
\] & de Morgan's law
\[
\begin{aligned}
\neg(\Phi \wedge \Psi) & \equiv \neg \Phi \vee \neg \Psi \\
\neg(\Phi \vee \Psi) & \equiv \neg \Phi \wedge \neg \Psi
\end{aligned}
\] \\
\hline \multicolumn{2}{|l|}{distributivity law
\[
\begin{aligned}
& \Phi \vee\left(\Psi_{1} \wedge \Psi_{2}\right) \equiv\left(\Phi \vee \Psi_{1}\right) \wedge\left(\Phi \vee \Psi_{2}\right) \\
& \Phi \wedge\left(\Psi_{1} \vee \Psi_{2}\right) \equiv\left(\Phi \wedge \Psi_{1}\right) \vee\left(\Phi \wedge \Psi_{2}\right)
\end{aligned}
\]} \\
\hline
\end{tabular}

Figure A.3: Some equivalence rules for propositional logic.
\[
\mu \models \Phi \text { iff } \mu \models \Psi .
\]

In this case we write \(\Phi \equiv \Psi\). For example, the formulae \(a \wedge \neg \neg b\) and \(a \wedge b\) are semantically equivalent. A few of the most important equivalence rules for propositional logic and the operators \(\neg, \wedge\) and \(\vee\) are shown in Figure A.3. Here, the Greek capital letters \(\Phi, \Psi\), \(\Xi\) (with or without subscripts) serve as metasymbols for formulae of propositional logic. The associativity and commutativity law for disjunction \(\vee\) and conjunction \(\wedge\) justify the omission of brackets and notations like
\[
\bigwedge_{1 \leqslant i \leqslant n} \Phi_{i} \text { or } \Phi_{1} \wedge \ldots \wedge \Phi_{n}
\]

Note that the length of a formula of type \(\bigwedge_{1 \leqslant i \leqslant n} \Phi_{i}\) is equal to \(n-1\) (and not to 1 ). Furthermore, notations like \(\bigwedge_{i \in I} \Phi_{i}\) or \(\bigwedge\left\{\Phi_{i} \mid i \in I\right\}\) are often used where \(I\) is an arbitrary finite index set. If \(I\) is nonempty, then \(\Phi\) stands for one of the formulae \(\Phi_{i_{1}} \wedge \ldots \wedge \Phi_{i_{k}}\) where \(I=\left\{i_{1}, \ldots, i_{k}\right\}\) and \(i_{1}, \ldots, i_{k}\) are pairwise different. For \(I=\varnothing\), the convention is
\[
\bigwedge_{i \in \varnothing} \Phi_{i} \stackrel{\text { def }}{=} \text { true, while } \bigvee_{i \in \varnothing} \Phi_{i} \stackrel{\text { def }}{=} \text { false. }
\]

Satisfiability and Validity Propositional formula \(\Phi\) is called satisfiable if there is an evaluation \(\mu\) with \(\mu \models \Phi\). \(\Phi\) is called valid (or a tautology) if \(\mu \models \Phi\) for each evaluation \(\mu\). \(\Phi\) is unsatisfiable if \(\Phi\) is not satisfiable. For example, \(a \wedge \neg a\) is unsatisfiable, while
\(a \vee \neg(a \wedge b)\) is a tautology. The formulae \(a \vee \neg b\) and \(a \wedge \neg b\) are satisfiable, but not tautologies. Obviously:
\[
\begin{array}{ll} 
& \Phi \text { is unsatisfiable } \\
\text { iff } & \mu \not \models \Phi \text { for all evaluations } \mu \\
\text { iff } & \mu \models \neg \Phi \text { for all evaluations } \mu \\
\text { iff } & \neg \Phi \text { is valid. }
\end{array}
\]

Thus, \(\Phi\) is unsatisfiable if and only if \(\neg \Phi\) is a tautology.
Literals and Positive Normal Form (PNF) A literal means a formula of the form \(a\) or \(\neg a\) where \(a \in A P\) is an atomic proposition. Propositional formulae in positive normal form (PNF for short, also sometimes called negation normal form) use the negation operator only on the level of literals. To ensure that the class of PNF formulae is as expressive as full propositional logic, both the conjunction and the disjunction operator serve as basic operators. Thus, the abstract syntax of PNF formulae is
\[
\Phi \quad::=\text { true } \mid \text { false }|a| \neg a\left|\Phi_{1} \wedge \Phi_{2}\right| \Phi_{1} \vee \Phi_{2}
\]
where \(a \in A P\). Given a (non-PNF) formula \(\Phi\), de Morgan's laws and the rule for double negation allow for "pushing the negation inside" until an equivalent formula in PNF arises. That is, successive application of the transformations
\[
\begin{array}{ll}
\neg\left(\Phi_{1} \wedge \Phi_{2}\right) & \rightsquigarrow \neg \Phi_{1} \vee \neg \Phi_{2} \\
\neg\left(\Phi_{1} \vee \Phi_{2}\right) & \rightsquigarrow \neg \Phi_{1} \wedge \neg \Phi_{2} \\
\neg \neg \Psi & \rightsquigarrow \Psi
\end{array}
\]
to \(\Phi\) 's subformulae yields an equivalent formula in PNF of the same asymptotic length.
Conjunctive and Disjunctive Normal Form Special cases of PNF are the conjunctive normal form (CNF) and disjunctive normal form (DNF). A CNF formula has the form
\[
\bigwedge_{i \in I} \bigvee_{j \in J_{i}} l i t_{i, j}
\]
where \(I\) and \(J_{i}\) are arbitrary finite index sets and lit \(t_{i, j}\) for \(i \in I\) and \(j \in J_{i}\) are literals. The subformulae \(\bigvee_{j \in J_{i}}\) lit \(i_{i, j}\) are called clauses. For instance, \(\left(a_{1} \vee \neg a_{3} \vee a_{4}\right) \wedge\left(\neg a_{2} \vee \neg a_{3} \vee a_{4}\right) \wedge a_{4}\) is a CNF formula with three clauses. Note that, e.g., true and false are also representable by CNF formulae: true is obtained by \(I=\varnothing\), while false is equivalent to \(a \wedge \neg a\), a CNF with two clauses consisting of one literal each. Given a PNF formula \(\Phi\), an equivalent CNF formula is obtained (on the basis of the distributivity laws) by applying the transformation rules:
\[
\begin{aligned}
& \Phi_{0} \vee\left(\Psi_{1} \wedge \Psi_{2}\right) \rightsquigarrow\left(\Phi_{0} \vee \Psi_{1}\right) \wedge\left(\Phi_{0} \vee \Psi_{2}\right) \\
& \left(\Psi_{1} \wedge \Psi_{2}\right) \vee \Phi_{0} \rightsquigarrow\left(\Psi_{1} \vee \Phi_{0}\right) \wedge\left(\Psi_{2} \vee \Phi_{0}\right)
\end{aligned}
\]
to \(\Phi\) 's subformulae as long as possible. Thus, for each propositional formula \(\Phi\) there exists an equivalent CNF formula \(\Phi^{\prime}\). Similarly, DNF formulae are formulae of the form
\[
\bigvee_{i \in I} \bigwedge_{j \in J_{i}} l i t_{i, j}
\]
where \(I\) and \(J_{i}\) are arbitrary finite index sets and \(l_{i t_{i, j}}\) for \(i \in I\) and \(j \in J_{i}\) are literals. E.g., \(\left(a_{1} \wedge \neg a_{2}\right) \vee\left(\neg a_{2} \wedge \neg a_{3} \wedge a_{4}\right) \vee\left(\neg a_{1} \wedge \neg a_{3}\right)\) is a DNF formula. A transformation similar to the one for CNF can be applied to prove that any propositional formula \(\Phi\) has an equivalent DNF formula \(\Phi^{\prime}\).

\section*{A. 4 Graphs}

In most chapters of this monograph, the reader is supposed to be familiar with the basic principles of graph theory and elementary graph algorithms. These are supposed to be known from basic courses. This section is a brief summary of the terms and concepts that are central to this monograph. The details can be found in any elementary textbook on algorithms and data structures; see, e.g. [24, 100], or textbooks that provide an introduction to graph theory, see, e.g. [396].

Digraphs (Graphs, for short) A digraph (or directed graph, in the following called graph \()\) is a pair \(G=(V, E)\) consisting of a set \(V\) of vertices and an edge relation \(E \subseteq V \times V\). The elements of \(E\) are called edges. \(G\) is called finite if the set \(V\) (and hence also \(E\) ) is finite. The basic models considered in this monograph will rely on graphs where the vertices and edges are augmented with certain information. E.g., instead of defining \(E\) as a set of vertex pairs, we may also deal with labeled edges, in which case \(E\) is a subset of \(V \times \Sigma \times V\) for a certain alphabet \(\Sigma\). If this additional information is irrelevant, the edge labels may be omitted and rather than considering \(E \subseteq V \times \Sigma \times V\) we use \(E^{\prime}=\{(v, w) \mid\) there exists an edge \((v, \sigma, w) \in E\}\) to obtain a digraph in the above sense.

The following explanations refer to the case of a digraph \(G=(V, E)\) where \(E\) is a binary relation over the vertices. For \(v \in V\), let \(\operatorname{Post}_{G}(v)\), or briefly \(\operatorname{Post}(v)\), denote the set of direct successors of \(v\); that is, \(\operatorname{Post}(v)=\{w \in V \mid(v, w) \in E\}\). Similarly, \(\operatorname{Pre}_{G}(v)\), or briefly \(\operatorname{Pre}(v)\), stands for the set of direct predecessors of \(v\), i.e., \(\operatorname{Pre}(v)=\{w \in V \mid\) \((w, v) \in E\}\). A vertex \(v\) is called terminal if \(\operatorname{Post}(v)=\varnothing\). Edges of the form \((v, v)\) are called self-loops.

Paths in Graphs A path in \(G\) denotes a (finite or infinite) non-empty sequence of vertices \(\hat{\pi}=v_{0} v_{1} \ldots v_{r}\) with \(r \geqslant 0\) or \(\pi=v_{0} v_{1} v_{2} \ldots\) such that \(v_{i+1} \in \operatorname{Post}\left(v_{i}\right), i=0,1,2, \ldots\). (In
the context of transition systems, the term path is used in the sense of a maximal path, that is, a path is either infinite or ending in a terminal state.) A path is called simple if its vertices are pairwise distinct, e.g., the finite path \(v_{0} v_{1} \ldots v_{n}\) is simple if \(v_{i} \neq v_{j}\) for all \(0 \leqslant i<j \leqslant n\). The length of a path defines the number of traversed edges along the path and is denoted by \(|\cdot|\). If \(\hat{\pi}\) and \(\pi\) are as above, then \(|\hat{\pi}|=r\) and \(|\pi|=\omega\). The set \(\operatorname{Post}^{*}(v)\) or Reach \((v)\) denotes the set of all vertices reachable from \(v\); that is, the set of all vertices \(w \in V\) with a finite path \(v_{0} v_{1} \ldots v_{r}\) with \(v=v_{0}\) and \(w=v_{r}\). A finite path \(v_{0}, v_{1}, \ldots, v_{r}\) is called a cycle if \(v_{0}=v_{r}\) and \(r>0\). Graph \(G\) is called acyclic (or cycle-free) if \(G\) does not contain any cycle; otherwise \(G\) is called cyclic.

A popular representation of finite digraphs is by adjacency lists. This entails a list representation for the sets \(\operatorname{Post}(v)\) for each vertex \(v\). To support direct access from vertex \(v\) to its direct successors, an array for the vertices (in arbitrary order) can be used which contains pointers to the heads of the adjacency lists.

Depth- and Breadth-First Search (DFS, BFS) Depth-first search (DFS) and breadth-first search (BFS) are important graph traversal strategies. They are both based on the skeleton given in Algorithm 48 which deterimines each vertex \(v\) that is reachable from the vertex \(v_{0}\). \(R\) is the set of vertices that have been already visited, while \(U\) keeps track of all vertices that still have to be explored (provided they are not contained in \(R\) ). The procedure terminates since each edge is tagged as taken at most once. Any vertex \(u\) can thus be added to \(U\) (and taken from \(U\) ) at most \(|\operatorname{Pre}(u)|\) times. Hence, Algorithm 48 terminates after at most \(\mathcal{O}(M)\) iterations where \(M=|E|\) is the number of edges. (Note that \(M=\sum_{v \in V}|\operatorname{Post}(v)|=\sum_{u \in V}|\operatorname{Pre}(u)|\).) On termination, \(R\) contains all vertices that are reachable from \(v_{0}\).

In case all vertices \(v \in V\) have to be visited, the algorithm can be started with some arbitrary vertex \(v_{0}\) and restarted with a new vertex \(v_{0}^{\prime}\), say, that has not yet been visited, i.e., \(v_{0}^{\prime} \notin R\). This can be repeated until all vertices belong to \(R\). Given an adjacency list representation of the sets \(\operatorname{Post}(v)\), the time complexity is \(\Theta(N+M)\) where \(N=|V|\) is the number of vertices and \(M=\sum_{v \in V}|\operatorname{Post}(v)|=|E|\) the number of edges. Here, it is assumed that \(R\) and \(U\) are represented such that access to the elements of \(R\) and \(U\), as well as the insertion and deletion into \(U\), takes constant time. For moderately sized graphs, the set \(R\) can be represented by a bit-vector. In the context of model checking, \(R\) is typically represented by a hash table. Using this data structure, the expected time to insert an element into \(R\) and to check membership in \(R\) is constant.

Depth-first and breadth-first search differ in the realization of the multiset \(U\). The DFS approach organizes \(U\) as a stack, while BFS relies on a queue implementation for \(U\). Accordingly, a stack obeys the LIFO principle (last-in, first-out), while the insertion and deletion of elements from a queue relies on the FIFO principle (first-in, first-out).
```

Algorithm 48 Reachability analysis
Input: finite digraph $G=(V, E)$, vertex $v_{0} \in V$
Output: Reach $\left(v_{0}\right)$
set of vertex $R:=\varnothing$;
multiset of vertex $U:=\left\{v_{0}\right\}$;
(* the set of explored vertices *)
(* vertices still to be explored *)
(* initially none of the edges is tagged as taken *)
while $U \neq \varnothing$ do
let $v \in U ; \quad$ (* choose some vertex still to be explored *)
if $\exists u \in \operatorname{Post}(v)$ such that $(v, u)$ is not tagged as taken then
let $u \in \operatorname{Post}(v)$ be such a vertex;
tag the edge $(v, u)$ as taken;
if $u \notin R$ then
$U:=U \cup\{u\} ; \quad$ (* ensure that all successors *)
$R:=R \cup\{u\}$
(* of $v$ will be explored ${ }^{*}$ )
fi
else
$U:=U \backslash\{v\}$
fi
od
return $R$.
(* $\left.R=\operatorname{Reach}\left(v_{0}\right)^{*}\right)$

```

Let us briefly summarize some details of the DFS approach. As mentioned above, DFSbased graph traversal organizes the multiset of vertices that are still to be explored by a stack. Stacks support the operations \(\operatorname{top}(U)\) (which returns the first element of \(U\) ), \(\operatorname{pop}(U)\) (which deletes the first element of \(U\) ), and \(\operatorname{push}(U, v)\) (which inserts \(v\) into \(U\) as the first element). The empty stack will be denoted with \(\varepsilon\). Using an appropriate array or list implementation of \(U\), each of these operations, as well as checking emptiness, can be executed in constant time.

DFSs provide the basis for many algorithms that analyze the topological structure of a graph. An important algorithm in this monograph is a DFS-based cycle check, i.e., the problem of deciding whether a given finite digraph contains a cycle. Such an algorithm exploits the concept of backward edges. The edge \((v, u)\) is a backward edge whenever vertex \(u\) is on the DFS stack \(U\) on tagging \((v, u)\) as taken; see Algorithm 49.
```

Algorithm 49 Depth-first search
Input: finite digraph $G=(V, E)$, vertex $v_{0} \in V$
Output: Reach (vo

```

```

(* any vertex $v$ in $G$ has been visited ${ }^{*}$ )

```

Any vertex \(w\) which is visited (i.e., inserted into \(R\) ) when \(u\) is on the stack \(U\) is reachable from \(u\). Thus, any backward edge "closes" a cycle. Vice versa, if \(v_{0} v_{1} \ldots v_{n}\) is a cycle in \(G\) and \(v_{0}\) is visited during the DFS after visiting \(v_{1}, \ldots, v_{n-1}\), then-on removing \(v_{0}\) as top element from \(U\) tagging the edge \(\left(v_{0}, v_{1}\right)\) as taken - the vertices \(v_{2}\) through \(v_{n-1}\) are on the stack \(U\) (since \(v_{0}\) is reachable from each of these vertices). As \(v_{1}\) has been visited before \(v_{0}\), the edge \(\left(v_{0}, v_{1}\right)\) is a backward edge. Hence, for any cycle \(v_{0} v_{1} \ldots v_{n}\) in \(G\), the DFS classifies at least one edge \(\left(v_{i-1}, v_{i}\right)\) as a backward edge. This yields that \(G\) has a cycle if and only if the DFS detects a backward edge. Moreover, if we consider a fixed starting vertex \(v_{0}\) and apply the DFS approach to visit (exactly) the vertices that are reachable from \(v_{0}\), then \(v_{0}\) belongs to a cycle in \(G\) if and only if the DFS finds a backward edge of the form \(\left(w, v_{0}\right)\) for some vertex \(w\).

Strongly Connected Components (SCCs) Let \(G=(V, E)\) be a finite digraph and \(C \subseteq V . C\) is strongly connected if for every pair of vertices \(v, w \in C\), the vertices \(v\) and \(w\) are mutually reachable, i.e., \(v \in \operatorname{Post}^{*}(w)\) and \(w \in \operatorname{Post}^{*}(v)\). A strongly connected component (SCC, for short) of \(G\) is a maximally strongly connected set of vertices. That is, \(C\) is an SCC if \(C\) is strongly connected and \(C\) is not contained in another strongly connected vertex set \(D \subseteq V\) with \(C \neq D\). SCC \(C\) is called trivial if \(C=\{v\}\) and \((v, v) \notin E\). SCC \(C\) is called terminal if there is no SCC \(D \neq C\) such that \((v, w) \in E\) for some \(v \in C\) and \(w \in D\). Finite digraph \(G\) is cyclic if and only if \(G\) contains a nontrivial SCC. The time complexity of determining the SCCs in finite digraph \(G\) is \(\Theta(N+M)\) where \(N=|V|\) and \(M=|E|\). This complexity is obtained using a variant of DFS.

Trees The digraph \(T=(V, E)\) is a (directed) tree if there exists a vertex \(v_{0}\) with \(\operatorname{Pre}\left(v_{0}\right)=\) \(\varnothing\) such that each vertex \(v \in V\) is reachable from \(v_{0}\) via a unique path. The vertex \(v_{0}\) is the unique vertex with this property and is called the root of \(T\). Any vertex \(v\) with \(\operatorname{Post}(v)=\varnothing\) is called a leaf. Any vertex \(v \in V \backslash\left\{v_{0}\right\}\) has exactly one direct predecessor, i.e., \(|\operatorname{Pre}(v)|=1\). This unique direct predecessor \(w\) of \(v\) is the father of \(v ; v\) is called a son of \(w\). Tree \(T\) is called finitely branching if the maximal outdegree of all its vertices is finite, i.e., \(\sup _{v \in V}|\operatorname{Post}(v)| \in \mathbb{N}\). A binary tree is a tree with \(|\operatorname{Post}(v)| \leqslant 2\) for each vertex \(v\).

Hamiltonian Paths A Hamiltonian path in a finite digraph \(G=(V, E)\) is a path \(v_{1} v_{2} \ldots v_{n}\) where each vertex \(v \in V\) appears exactly once, i.e., \(V=\left\{v_{1}, \ldots, v_{n}\right\}\) and \(i \neq j\) implies \(v_{i} \neq v_{j}\). The (directed) Hamiltonian path problem questions whether a given finite digraph \(G\) has a Hamiltonian path.

Undirected Graphs An undirected graph is a digraph in which the orientation of the edges is ignored. Formally, an undirected graph is a pair \((V, E)\) where \(V\) is a set of vertices and \(E\) a set consisting of subsets of \(V\) with exactly two elements. Then, \(\{v, w\} \in E\) means that there is an edge connecting \(v\) and \(w\). Paths in undirected graphs are defined as in
the directed case. For cycles it is required that they are composed of simple paths, i.e., paths \(v_{0} v_{1} \ldots v_{n}\) where \(n>1\) and \(v_{0}, \ldots, v_{n}\) are pairwise distinct. (This constraint avoids that a path of the form \(v w v\) with \(\{v, w\} \in E\) is classified as cycle.)

\section*{A. 5 Computational Complexity}

In this monograph, some lower bounds on the time complexity of verification problems are presented. These lower bounds are based on showing the hardness for a certain complexity class. This section summarizes the main basic concepts at a rather intuitive level. For the precise definitions and technical details we refer to textbooks on complexity theory, such as [160, 320].

The complexity class we consider here refers to decision problems, i.e., problems that take a certain (finite) input and are required to return either "yes" or "no". Example decision problems are:

\section*{SAT (satisfiability problem for propositional logic)}

Given a propositional formula \(\Phi\). Is \(\Phi\) satisfiable?

\section*{(Directed) Hamiltonion path problem}

Given a finite digraph \(G\). Does \(G\) have a Hamiltonion path?

\section*{Cycle problem}

Given a finite digraph \(G\). Is \(G\) cyclic?

\section*{Pattern matching}

Given finite words \(w=A_{1} \ldots A_{n}\) (a text), \(v=B_{1} \ldots B_{m}\) (a pattern) over some alphabet. Is \(v\) a subword of \(w\) ?

Complexity classes are classes of decision problems, classified by the resources (i.e., time and space) that are required to solve them algorithmically.

Deterministic and Nondeterministic Algorithms The precise definition of the complexity classes uses Turing machines as a formalization of algorithms. The input is encoded by a finite word of the input alphabet of the Turing machine. The answers "yes" and "no" correspond to the requirement that the Turing machine halts in an accept state ("yes") or in a nonaccept state ("no"). These details are omitted here. We describe informally the complexity classes by means of an intuitive notion of what an algorithm is. For the purpose of this monograph we need both deterministic and nondeterministic algorithms. Although one typically aims at a deterministic algorithm to solve a certain problem (since
real programs do not work nondeterministically) the concept of nondeterministism plays a crucial role in various theoretical considerations.

A deterministic algorithm is an algorithm where for any configuration it is uniquely determined whether the algorithm halts or not, which output (if any) is provided, or what the next step (and the successor configuration) is. A deterministic algorithm solves a given decision problem \(P\) if for each input \(w(1)\) the unique computation of the algorithm started with \(w\) terminates, and (2) returns the correct answer. An input \(w\) (for \(P\) ), of course, needs to satisfy all criteria as given by the decision problem \(P\), e.g., for the SAT problem, the input \(w\) is a propositional formula \(\Phi\). A deterministic algorithm solves the SAT problem if it provides the answer "yes" if \(\Phi\) is satisfiable and "no" if \(\Phi\) is not satisfiable.

The time complexity of a deterministic algorithm \(A\) is typically measured by a function \(T_{A}: \mathbb{N} \rightarrow \mathbb{N}\) where \(T_{A}(n)\) is the maximal number of "steps" that \(A\) performs on an input of size \(n\). As the maximal number of steps is considered, \(T_{A}\) reflects the worst-case running time. Similarly, the space complexity is formalized by a function \(S_{A}: \mathbb{N} \rightarrow \mathbb{N}\) where \(S_{A}(n)\) is the maximal number of atomic memory units that are used for inputs of length \(n .{ }^{2}\)

Nondeterministic algorithms differ from deterministic algorithms in the sense that they may provide finitely many alternatives for the next step of a configuration. Thus, there might be several possible computations of a nondeterministic algorithm on a given input \(w\). A nondeterministic algorithm is said to solve a decision problem \(P\) if for each input \(w\) :
(1) all computations for \(w\) terminate,
(2a) if the correct answer for \(w\) is "yes", then there is at least one computation for \(w\) that returns the answer "yes", and
(2b) if the correct answer for \(w\) is "no", then all computations for \(w\) return the answer "no".

An example of a nondeterministic algorithm for SAT is shown in Algorithm 50. Here, \(\operatorname{sat}(\mu, \Phi)\) is a subprocedure which computes deterministically the truth value of \(\Phi\) under the evaluation \(\mu\) (in time \(\Theta(|\Phi|)\) ). It returns the Boolean value true if \(\mu \models \Phi\) and false otherwise. The details of this algorithm sat are not of importance here. Algorithm 50 relies on the so-called "guess and check" principle which is the basis for many nondeterministic algorithms. After nondeterministically guessing an evaluation \(\mu\) it is checked whether or not \(\mu \models \Phi\). If the guess is correct, the evaluation is a witness for " \(\mu \models \Phi\) ", and the answer

\footnotetext{
\({ }^{2}\) For some complexity classes, \(S_{A}(n)\) only takes into account the memory requirement that is needed in addition to the memory units for storing the input. As these complexity classes are not used in this monograph, this detail is ignored here.
}
```

Algorithm 50 Nondeterministic algorithm for SAT
Input: propositional logic formula $\Phi$ over $A P=\left\{a_{1}, \ldots, a_{n}\right\}$
Output: "yes" or "no"
(* Guess an evalution $\mu$ for $A P{ }^{*}$ )
for $i=1, \ldots, n$ do
choose nondeterministically a truth value $t \in\{0,1\}$;
$\mu\left(a_{i}\right):=t$
od
(* Check whether $\mu \models \Phi^{*}$ )
if $\operatorname{sat}(\mu, \Phi)$ then
return"yes" (* $\Phi$ is satisfiable $\left.{ }^{*}\right)$
else
return "no" $\quad(* \Phi$ might or might not be satisfiable *)
fi

```
"yes" is correct. If, however, \(\mu \not \models \Phi\), then the given answer "no" might be wrong, as there might be another evaluation \(\mu^{\prime}\) with \(\mu^{\prime} \models \Phi\). However, Algorithm 50 solves SAT since (1) it terminates for all input formulae \(\Phi\). The fact that it also fulfills conditions (2a) and (2b) can be seen as follows. Condition (2a) holds since whenever \(\Phi\) is satisfiable, then there exists a computation which chooses an evalution under which \(\Phi\) holds, and hence, the algorithm returns "yes". Condition (2b) holds since whenever \(\Phi\) is not satisfiable, then for each choice of \(\mu\) in the guess phase we have \(\mu \not \vDash \Phi\) and the returned answer is "no".

The time complexity of a nondeterministic algorithm \(A\) that solves a decision problem (i.e., all computations terminate) is given by the function \(T_{A}: \mathbb{N} \rightarrow \mathbb{N}\) where \(T_{A}(n)\) is the maximum number of steps that \(A\) may perform for inputs of length \(n\). Similarly, the space complexity of \(A\) is given by the function \(S_{A}: \mathbb{N} \rightarrow \mathbb{N}\) where \(S_{A}(n)\) is the maximum number of memory units that \(A\) may access on an input of length \(n\).

Algorithm \(A\) (deterministic or nondeterministic) is called polynomially time-bounded (or briefly, a polytime algorithm) if \(T_{A}(n)=\mathcal{O}(\operatorname{poly}(n))\). Similarly, algorithm \(A\) is a polyspace algorithm if \(S_{A}(n)=\mathcal{O}(\operatorname{poly}(n))\).

Complexity Classes Using the above intuitive explanations of deterministic and nondeterministic algorithms, we are in a position to introduce the complexity classes used in this monograph:

PTIME (or briefly P) denotes the class of all decision problems that can be solved by a deterministic polytime algorithm.

NP denotes the class of all decision problems that can be solved by a nondeterministic polytime algorithm.

PSPACE denotes the class of all decision problems that can be solved by a deterministic polyspace algorithm.

The complexity class NPSPACE is the class of all decision problems that can be solved by a nondeterministic polyspace algorithm. It is known that PSPACE \(=\) NPSPACE. Since any deterministic algorithm can be viewed as a nondeterministic algorithm (where the number of alternative steps per configuration is at most one) and since in \(N\) steps at most \(N\) memory units can be used, it follows that:
\[
\text { PTIME } \subseteq \mathrm{NP} \subseteq \text { PSPACE }=\text { NPSPACE }
\]

One of the most important open questions in theoretical computer science is whether PTIME agrees with NP. It is also unknown so far whether NP agrees with PSPACE.

The next complexity class considered is coNP which is somehow "complementary" to NP. Let \(P\) be a decision problem. Then, the complementary problem of \(P\), denoted \(\bar{P}\), is the problem which takes the same inputs as \(P\) but requires the answer "no" if \(P\) requires the answer "yes", and vice versa. For example, the complementary problem of SAT takes as input a propositional logic formula \(\Phi\) and asks whether \(\Phi\) is not satisfiable. The complementary problem of the (directed) Hamiltonian path problem asks whether a given finite digraph \(G\) does not have a Hamiltonian path.
coNP denotes the class of all decision problems \(P\) where the complementary problem \(\bar{P}\) belongs to NP.
coNP subsumes PTIME and is contained in PSPACE, while it is unknown whether these inclusions are strict. The complexity classes coPTIME and coPSPACE can be defined as the classes of the complementary problems of PTIME and PSPACE, respectively. However, the symmetry of the treatment of "yes" and "no" by deterministic algorithms allows swapping the outputs of a deterministic polytime (or polyspace) algorithm that solves problem \(P\) to obtain a deterministic polytime (polyspace) algorithm for \(\bar{P}\). Hence, PTIME \(=\) coPTIME and PSPACE \(=\) coPSPACE. Note that this argument does not apply to nondeterministic complexity classes due to the asymmetry of "yes" and "no" answers according to (2a) and (2b). The question whether NP and coNP agree is an open problem.

Further complexity classes that will be needed in this monograph are EXPTIME and 2EXPTIME. EXPTIME denotes the class of decision problems \(P\) that are solvable by a deterministic algorithm where the worst-case time complexity is exponentially bounded, i.e., bounded above by a function \(n \mapsto 2^{p(n)}\) where \(p\) is a polynomial. The complexity
class 2EXPTIME consists of all decision problems that can be solved by a deterministic algorithm for which the worst-case time complexity is bounded by some function \(n \mapsto 2^{2^{p(n)}}\) where \(p\) is a polynomial. It holds that
\[
\text { PSPACE } \subseteq \text { EXPTIME } \subseteq \text { 2EXPTIME. }
\]

Completeness To specify the "hardest" problem in NP, Cook [99] introduced in the early seventies the notion of NP-completeness. Decision problem \(P \in\) NP is NP-complete if the existence of a deterministic polytime algorithm \(A\) for \(P\), implies that any other problem \(Q \in\) NP can be solved deterministically with the help of \(A\) plus a polytime transformation from \(Q\) 's inputs to P's inputs. (The transformation of \(Q\) 's inputs into \(P\) 's inputs relies on the reduction principle.) Thus, if one NP-complete problem is shown to be in PTIME, then NP = PTIME.

Let us briefly consider these concepts in more detail. Let \(P\) and \(Q\) be decision problems. Problem \(Q\) is called polynomially reducible to \(P\) if there exists a polytime deterministic algorithm which transforms a given input \(w_{Q}\) for \(Q\) into an input \(w_{P}\) for \(P\) such that the correct answer for \(Q\) on \(w_{Q}\) is "yes" if and only if the correct answer for \(P\) on \(w_{P}\) is "yes". Decision problem \(P\) is said to be NP-hard if each problem \(Q \in\) NP is polynomially reducible to \(P\). \(P\) is called NP-complete if (1) \(P\) belongs to NP and (2) \(P\) is NP-hard. There is a wide range of NP-complete problems. For many cases, the proof of membership of \(P\) in NP is quite simple - one provides a polytime nondeterministic algorithm for \(P\) on the basis of the guess-and-check paradigm. The NP-hardness is often harder to prove. The first NP-completeness result was by Cook [99] who showed that the SAT problem is NPcomplete. His proof for the NP-hardness of SAT is generic: it starts with a formalization of a polytime nondeterministic algorithm for some NP-problem \(Q\) by a nondeterministic Turing machine \(\mathcal{M}\) and constructs for a given input word \(w\) for \(\mathcal{M}\) a propositional formula \(\Phi_{\mathcal{M}, w}\) such that \(\Phi_{\mathcal{M}, w}\) is satisfiable if and only if \(\mathcal{M}\) has an accepting computation for \(w\). For proving the NP-hardness of a given problem \(P\), it is often simpler to use the reduction principle:
\[
\text { if } Q \text { is NP-hard and } Q \text { is polynomially reducible to } P \text {, then } P \text { is NP-hard. }
\]

For instance, a polynomial reduction from SAT to 3SAT (a variant of SAT which takes as input a CNF formula \(\Phi\) with at most three literals per clause and asks for the satisfiability of \(\Phi)\) is not very involved and yields the NP-hardness of 3SAT. The NP-hardness of the Hamiltonian path problem can be proven by a polynomial reduction from 3SAT to the Hamiltonian path problem. The latter problem is again polynomially reducible to the three-coloring problem, a decision problem on finite undirected graphs that asks whether the vertices can be colored with three colors such that for no edge \(\{v, w\}\) the vertices \(v\) and \(w\) are equally colored. Hence, the three-coloring problem is NP-hard too. In this way, a wide range of NP-hardness results have been obtained.

Hardness or completeness can be defined in an analogous way for other complexity classes such as PSPACE and coNP. \(P\) is called PSPACE-complete if (1) \(P\) belongs to PSPACE and (2) \(P\) is PSPACE-hard, i.e., all problems in PSPACE are polynomially reducible to \(P\). An example of a PSPACE-complete problem is the universality problem: does the language described by a regular expression over the alphabet \(\Sigma\) equal \(\Sigma^{*}\) ? Similarly, \(P\) is called coNP-complete if (1) \(P\) belongs to coNP and (2) \(P\) is coNP-hard, i.e., all problems in coNP are polynomially reducible to \(P\). The symmetry of NP and coNP yields that a problem \(P\) is NP-hard if and only if its complementary problem \(\bar{P}\) is coNP-hard. Thus, \(P\) is NP-complete if and only if \(\bar{P}\) is coNP-complete. For instance, the problem whether a given propositional logic formula \(\Phi\) is valid is coNP-complete, since \(\Phi \mapsto \neg \Phi\) yields a polynomial reduction from SAT to the validity problem-recall that \(\Phi\) is not satisfiable if and only if \(\neg \Phi\) is valid (see page 919) -and \(\overline{\text { SAT }}\) is coNP-complete (as SAT is NPcomplete). Since coNP and NP are contained in PSPACE, any problem that is hard for PSPACE is also coNP- and NP-hard.

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[^0]:    ${ }^{1}$ For some products this is much higher, though. Microsoft has acknowledged that Windows 95 contained at least 5000 defects. Despite the fact that users were daily confronted with anomalous behavior, Windows 95 was very successful.

[^1]:    ${ }^{2}$ Parts of the more advanced model-checking procedures have been formally proven correct using theorem provers to circumvent this.

[^2]:    ${ }^{1}$ Recall that $2{ }^{A P}$ denotes the power set of $A P$.

[^3]:    ${ }^{2}$ An evaluation $s \in E \operatorname{val}(\cdot)$ is a mapping which assigns a value $s\left(x_{i}\right) \in\{0,1\}$ to any input bit $x_{i}$. Similarly, every register $r_{j}$ is mapped onto a value $s\left(r_{j}\right) \in\{0,1\}$. To simplify matters, we assume every element $s \in S$ to be a bit-tuple of length $n+k$. The $i$ th bit is set if and only if $x_{i}$ is evaluated to 1 $(0<i \leqslant n)$. Accordingly, the $n+j$ th bit indicates the evaluation of $r_{j}(0<j \leqslant k)$.

[^4]:    ${ }^{3}$ For simplicity, the operators are supposed to be total. For operators that require special arguments (e.g., division requires a nonzero second argument) we assume that the corresponding domain contains a special element with the meaning of "undefined".

[^5]:    ${ }^{4}$ Operator $*$ is typically assumed to be commutative and associative.

[^6]:    ${ }^{1}$ Formally, we add a loop to the transition system of a defective philosopher at state think ${ }_{i}$.

[^7]:    ${ }^{2}$ It is important to realize the difference between the notion of a path in a transition system and the notion of a path in a digraph. A path in a transition system is maximal, whereas a path in a digraph in the graph-theoretical sense is not always maximal. Besides, paths in a digraph are usually required to be finite whereas paths in transition systems may be infinite.

[^8]:    ${ }^{3}$ A further alternative is to adapt the linear-time framework for transition systems with terminal states. The main concepts of this chapter are still applicable, but require some adaptions to distinguish nonmax-

[^9]:    imal and maximal finite paths.

[^10]:    ${ }^{4}$ Here, we assume two transition systems with sets of propositions that include $A P$.

[^11]:    ${ }^{5}$ The basic principles of propositional logic are treated in Appendix A.3.

[^12]:    ${ }^{6}$ To cover the special case where $\Phi$ is an atomic proposition, in which case $|\Phi|=0$, we deal with $1+|\Phi|$ for the cost to check whether $\Phi$ holds for a given state $s$.

[^13]:    ${ }^{7}$ The main concepts of a finite automaton as acceptors for languages over finite words are summarized in Section 4.1.

[^14]:    ${ }^{8}$ If $T S$ is a transition system with terminal states, then $\operatorname{Traces}(T S)$ is defined as the set of all words $\operatorname{trace}(\pi)$ where $\pi$ is an initial, maximal path fragment in $T S$.

[^15]:    ${ }^{1}$ The symbol $\left\}\right.$ denotes the empty subset of $A P$ which serves as symbol in the alphabet $\Sigma=2^{A P}$. It must be distinguished from the regular expression $\varnothing$ representing the empty language.

[^16]:    ${ }^{2}$ The symbol $\omega$ denotes the first infinite ordinal. It already appeared in the notation $\Sigma^{\omega}$ for the set of infinite words over the alphabet $\Sigma$.

[^17]:    ${ }^{3}$ A strongly connected component is nontrivial if it contains at least one edge. In fact, any cycle is contained in a nontrivial strongly connected component, and vice versa, any nontrivial strongly connected component contains a cycle that goes through all its states.

[^18]:    ${ }^{4}$ In Chapter 10, deterministic Rabin automata will be used for representing $\omega$-regular properties.

[^19]:    ${ }^{5}$ Recall that $G(T S)$ denotes the underlying directed graph of $T S$.

[^20]:    ${ }^{6}$ For a recursive formulation of the outer depth-first search, this is the moment where the depth-first search call for $s$ terminates.

[^21]:    ${ }^{7}$ exactly once, if Algorithm 8 does not abort with the answer "no".

[^22]:    ${ }^{1}$ The Backus Naur form (BNF) is used in a somewhat liberal way. More concretely, nonterminals are identified with derived words (formulae) and indices in the rules. Moreover, brackets will be used, e.g. in $a \wedge(b \cup c)$, which are not shown in the grammar. Such simplified notations for grammars to determine the syntax of formulae of some logic (or terms of other calculi) are often called abstract syntax.

[^23]:    ${ }^{2}$ Only the identity of a process is unique.

[^24]:    ${ }^{1}$ The semantics of the CTL path formulae is formulated more simply than for LTL, since in CTL each

[^25]:    ${ }^{2}$ Recall that the notion CTL formula is used for a CTL state formula.

[^26]:    For each CTL formula there exists an equivalent CTL formula in PNF.

[^27]:    ${ }^{3}$ Although the rewrite rules for $\neg \forall U$ and $\neg \exists U$ could be simplified by $\neg \forall(\Phi \cup \Psi) \equiv$ $\exists((\neg \Psi) \mathrm{W}(\neg \Phi \wedge \neg \Psi))$ and $\neg \exists(\Phi \cup \Psi) \equiv \forall((\neg \Psi) \mathrm{W}(\neg \Phi \wedge \neg \Psi))$, there is still a duplication of formula $\Psi$.

[^28]:    ${ }^{4}$ A strongly connected component (SCC) of a digraph $G$ is a maximal, connected subgraph of $G$. Stated differently, the SCCs of a graph are the equivalence classes of vertices under the "are mutually reachable" relation. A nontrivial SCC is an SCC that contains at least one transition.

[^29]:    ${ }^{5}$ This observation is mainly of theoretical interest since it is more efficient to design special LTL model-

[^30]:    checking algorithms when dealing with fairness assumptions than to apply the reduction to the standard semantics $\models$.

