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Logical Verification in Lean

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lean-forward.github.io/ logical-verification/2019

Contents

Со	itents	iii
Pre	face	vii
ı	Basics	1
1	Definitions and Lemma Statements 1.1 Types and Terms	. 8
	1.3Function Definitions	. 15
2	Tactical Proofs 2.1 Tactic Mode	. 20 . 21 . 24 . 25 . 25 . 26
3	Structured Proofs and Proof Terms 3.1 Structured Proofs	. 31. 32. 33. 34
II	Functional–Logic Programming	37
4	Functional Programming 4.1 Inductive Types	. 39

	4.4 4.5	Example: Colors	45				
	4.6 4.7	New Tactics					
5	Indu	nductive Predicates 47					
	5.1	Introductory Example	47				
	5.2	Logical Symbols					
	5.3	Introduction, Elimination, Induction, and Inversion	48				
	5.4	Example: Full Binary Trees	48				
	5.5	Example: Sorted Lists	50				
	5.6	Example: Well-formed and Ground First-Order Terms	51				
	5.7	Example: Reflexive Transitive Closure	52				
	5.8	Induction Pitfalls	53				
	5.9	New Tactics	54				
	5.10	Summary of New Syntax	54				
6	Mon		55				
	6.1	Motivating Example					
	6.2	Overview					
	6.3	Monadic Laws					
	6.4	A Type Class of Monads					
	6.5	The Option Monad					
	6.6	The State Monad					
	6.7	Example: Generic Iteration over a List					
	6.8	Summary of New Syntax	62				
7	Meta	aprogramming	63				
•	7.1	Overview	_				
	7.2	The Tactic Monad					
	7.3	Names and Expressions	•				
	7.4	Example: A Destructive Tactic					
	7.5	Example: A Solvability Advisor					
	7.6		70				
Ш	Prog	gram Semantics	71				
8	Ope	rational Semantics	73				
	8.1	Motivation	73				
	8.2	A Minimalistic Imperative Language	73				
	8.3	Big-Step Semantics	74				
	8.4	Lemmas about the Big-Step Semantics	76				
	8.5	Small-Step Semantics	•				
	8.6	Lemmas about the Small-Step Semantics					
9	Hoai	re Logic	81				
-	9.1	Hoare Triples	81				
	9.2	Hoare Rules					
	-	A Semantic Approach to Hoare Logic					

	9.4 9.5 9.6 9.7 9.8 9.9	Example: Exchanging Two Variables85Example: Adding Two Numbers85A Verification Condition Generator86Example: Adding Two Numbers, Revisited88Hoare Triples for Total Correctness88Summary of New Syntax89	
10	10.1 10.2 10.3 10.4 10.5	Motivation	
IV	Matl	hematics 97	
11	11.1 11.2 11.3 11.4 11.5	cal Foundations of Mathematics99Type Universes99The Peculiarities of Prop100The Axiom of Choice103Subtypes104Quotient Types106Summary of New Syntax108	
12	12.1 12.2 12.3 12.4 12.5 12.6	Type Classes without Properties	
13	13.1	Rational Numbers	
Bił	oliogr	raphy 125	

Preface

Formal proof assistants are software tools designed to help their users carry out computer-checked proofs in a logical calculus. We usually call them *proof assistants*, or *interactive theorem provers*, but a mischievous student coined the phase "proof-preventing beasts," and dictation software occasionally misunderstands "theorem prover" as "fear improver." Consider yourself warned.

Rigorous and Formal Proofs Interactive theorem proving has its own terminology, already starting with the key notion of "proof." A *formal proof* is a logical argument expressed in a logical formalism. "Formal" is roughly synonymous with "logical" or "logic-based." Logicians—the mathematicians of logics—carried out formal proofs on papers decades before the advent of computers, but nowadays these are almost always carried out using a proof assistant.

In contrast, an *informal proof* is what a mathematician would normally call a proof. These are often carried out in <u>MEX</u> or on the blackboard, and are also called "pen-and-paper proofs." The level of detail can vary a lot, and phrases such as "it is obvious that," "clearly," and "without loss of generality" move some of the proof burden onto the reader. A *rigorous proof* is a very detailed informal proof.

The main strength of proof assistants is that they help develop highly trustworthy, unambiguous proofs of mathematical statements, using a precise logic. They can be used to prove arbitrarily advanced results, not only toy examples or logic puzzles. Formal proofs also helps students understand what constitutes a valid definition or a valid proof. To quote Scott Aaronson:¹

I still remember having to grade hundreds of exams where the students started out by assuming what had to be proved, or filled page after page with gibberish in the hope that, somewhere in the mess, they might accidentally have said something correct.

Formal proofs can help when we develop a new theory, to find out whether it works and exactly which assumptions are needed. They are useful when generalizing, refactoring, or otherwise modifying existing proof, in much the same way as a compiler helps developing correct programs. They give a high level of trustworthiness that makes it easier to review the proof. In addition, formal proofs can form the basis of verified computational tools (e.g., verified computer algebra).

Success Stories There have been a number of success stories in mathematics and computer science. Three landmark results in the formalization of mathematics have been the proof of the four-color theorem by Gonthier et al. [6], the proof

¹https://www.scottaaronson.com/teaching.pdf

of the odd-order theorem by Gonthier et al. [7], and the proof of the Kepler conjecture by Hales et al. [9]. The earliest work in this area was carried out by Nicolaas de Bruijn and his colleagues starting in the 1960s in a system called AUTOMATH.²

Today, few mathematicians use proof assistants, but this is changing. For example, 29 participants of the Lean Together 2019 meeting in Amsterdam,³ about formalization of mathematics, self-identified as mathematicians.

Most users of proof assistants today are computer scientists. A few companies, including AMD [21] and Intel [10], have been using proof assistants to verify their designs. In academia, some of the milestones are the verifications of the operating system kernels seL4 [12] and CertiKOS [8] and the development of the verified compilers CompCert [13] and JinjaThreads [14].

A Selection of Popular Proof Assistants There are dozens of proof assistants in development or use. A list of the main ones follows, classified by logical foundations:

- set theory: Isabelle/ZF, Metamath, Mizar;
- simple type theory (also called higher-order logic): HOL4, HOL Light, Isabelle/HOL;
- · dependent type theory: Agda, Coq, Lean, Matita, PVS.

ACL2 is hard to classify. It is based on a Lisp-like first-order logic.

Lean Lean is a new proof assistant developed primarily by Leonardo de Moura (Microsoft Research) since 2012. Its mathematical library, mathlib, is developed under the leadership of Jeremy Avigad (Carnegie Mellon University).

We will use version 3.4.2, which is likely to remain the last release before Lean 4. We will use both Lean's basic libraries and mathlib.

Lean boasts the following strengths:

- It has a highly expressive logic based on the calculus of inductive constructions, a dependent type theory.
- It is extended with classical axioms and quotient types.
- It includes a convenient metaprogramming framework, which can be used to program custom proof automation.
- It includes a modern user interface via a Visual Studio Code plugin.
- · It has highly readable, fairly complete documentation.
- · It is open source.

Nonetheless, be aware that Lean is a research project, with some rough edges. Lean's core library includes only basic algebraic definitions. More setup and especially tactics are found in Lean's mathematical library, mathlib.⁴ It is more than a mathematical library; it provides a lot of basic automation on top of barebones Lean.

²https://www.win.tue.nl/automath/

³https://lean-forward.github.io/lean-together/2019/index.html

⁴https://github.com/leanprover/mathlib/

This Book This book is a companion to the course Logical Verification (LoVe) taught at the Vrije Universiteit Amsterdam. Despite our attempts to make it self-contained, unlikely to be suitable as a standalone Lean tutorial—we recommend *Theorem Proving in Lean* [1]—nor is it a substitute for attending class or doing the exercises. Ultimately, theorem proving can only be learned by doing.

The Lean files accompanying this book can be found in a public repository. The files' naming scheme follows the book's chapters; thus, love06_monads_demo. lean is the main file associated with Chapter 6 ("Monads"), which is demonstrated in class; love06_monads_exercise_sheet.lean is the exercise sheet; and love06_monads_homework_sheet.lean is the homework sheet.

We have a huge debt to the authors of *Theorem Proving in Lean* [1] and *Concrete Semantics: With Isabelle/HOL* [17], who have taught us Lean and programming language semantics.

Special Symbols Lean lets us enter Unicode symbols such as \mathbb{Z} , \mathbb{Q} , and \to using $\mathbb{E}_{T_E}X$ -style macros. These symbols can be entered in Visual Studio Code by typing \mathbb{Z} , \mathbb{Q} , or \mathbb{Z} , and pressing the tab key or the space bar. We will freely use these notations. For reference, we provide a list of the main non-ASCII symbols that are used in the book and, for each, one of its ASCII-representations. By hovering over a symbol in Visual Studio Code while holding the control or command key pressed, you can see the different ways in which it can be entered.

\neg	\not	\wedge	\and	\vee	\or	\rightarrow	\->
\leftrightarrow	\<->	\forall	\fo	\exists	\ex	\leq	\<=
\geq	\>=	\neq	\neq	λ	\la	Π	\Pi
<	\<	\rangle	\>	\leftarrow	\<-	\mapsto	\ ->
\Rightarrow	\=>	\Longrightarrow	\==>	\mathbb{N}	\N	\mathbb{Z}	\Z
\mathbb{Q}	\ Q	\mathbb{R}	\R	α	\a	β	\ b
γ	\g	0	\0	1	\1	2	\2

⁵https://github.com/blanchette/logical_verification_2019

Part I

Basics

Chapter 1

Definitions and Lemma Statements

We start by the basics of Lean, without carrying out actual proofs yet. We review how to define types and functions and how to state their intended properties as lemmas.

Lean's logical foundation is a rich formalism called the *calculus of inductive* constructions, whose defining feature is its support for dependent types. In this chapter, we restrict our attention to its simply (i.e., nondependently) typed fragment, which is inspired by the λ -calculus and resembles typed functional programming languages such as Haskell, OCaml, and Standard ML. Even if you have not been exposed to these languages, you will recognize many of the concepts from modern programming languages (e.g., Python, C++11, Java 8). In a first approximation:

If your background is in mathematics, you probably already know most of the key concepts underlying functional programming, sometimes under different names. For example, the Haskell program

```
fib 0 = 0
fib 1 = 1
fib n = fib (n - 2) + fib (n - 1)
```

closely corresponds to the mathematical definition

$$fib(n) = \begin{cases} 0 & \text{if } n = 0\\ 1 & \text{if } n = 1\\ fib(n-2) + fib(n-1) & \text{if } n \geq 2 \end{cases}$$

If you are not familiar with typed functional programming, we recommend that you study a tutorial, such as the first six chapters of *Learn You a Haskell for Great Good!*¹

1.1 Types and Terms

Simple type theory (also called higher-order logic) corresponds roughly to the simply typed λ -calculus [4] extended with an equality operator (=). It is an abstract, extremely simplified version of a programming language with a function-calling mechanism that prefigures functional programming.

¹http://learnyouahaskell.com/

Types Types are either basic types such as \mathbb{Z} , \mathbb{Q} , and bool or total functions $\sigma \to \tau$, where σ and τ are themselves types. Types indicate which values an expression may evaluate to. They introduce a discipline that is followed somewhat implicitly in mathematics. In principle, nothing prevents a mathematician from stating $1 \in 2$, but a typing discipline would mark this as the error it likely is.

Semantically, types can be regarded as sets. We would normally define the types \mathbb{Z} , \mathbb{Q} , and bool in such a way that they faithfully capture the mathematicians' \mathbb{Z} and \mathbb{Q} and the computer scientists' Booleans, and similarly for the function arrow (\rightarrow). But despite their similarities, Lean and mathematics are distinct languages; when studying them, it is important to keep them apart. Lean's types may be *interpreted* as mathematical sets, but they *are not* sets.

Higher-order types are types containing left-nested \to arrows. Values of such types are functions that take other functions as arguments. Accordingly, the type $(\mathbb{Z} \to \mathbb{Z}) \to \mathbb{Z}$ is the type of unary functions that take a function of type $\mathbb{Z} \to \mathbb{Z}$ as argument and that return a value of type \mathbb{Z} .

Terms The *terms*, or expressions, of simple type theory consist of constants c, applications t u, λ -expressions λx , t, and variables x, where t and u denote terms. We write t : σ to indicate that term t has type σ .

- A constant $c:\sigma$ is a symbol of type σ whose meaning is fixed in the current logical context. For example, an arithmetic theory might be expected to contain constants such as $0:\mathbb{Z}$, $1:\mathbb{Z}$, $abs:\mathbb{Z}\to\mathbb{N}$, square $:\mathbb{N}\to\mathbb{N}$, and prime $:\mathbb{N}\to bool$. Note that functions (e.g., abs, square) and predicates (e.g., prime) are considered constants.
- An application t u, where t : σ → τ and u : σ, is a term of type τ denoting the result of applying the function t to the argument u—e.g., prime 0.
 No parentheses are necessary around the argument, unless it is a complex term—e.g., prime (abs 0).
- A λ -expression $\lambda x: \sigma$, t, where $t: \tau$, denotes the total function of type $\sigma \to \tau$ that maps each value x of type σ to t, where t is a term that may contain x. For example, $\lambda x: \mathbb{Z}$, square (abs x) denotes the function that maps (the value denoted by) 0 to (the value denoted by) square (abs 0), and likewise 1 to square (abs 1), and so on. A more mathematically oriented syntax would have been $x\mapsto \text{square (abs }x)$, but this is not supported by Lean.
- A $variable \ x : \sigma$ refers back to the input of a λ -expression $\lambda x : \sigma$, t enclosing it. In the expression $\lambda x : \mathbb{Z}$, square (abs x), the second x is a variable that refers back to the λ binder's input x.

Applications and λ -expressions mirror each other: A λ -expression "builds" a function; an application "destructs" a function. Although our functions are unary (i.e., they take one argument), we can build n-ary functions by nesting λ s, using a technique called currying. For example, $\lambda x : \sigma$, $(\lambda y : \tau, x)$ denotes the function of type $\sigma \to (\tau \to \sigma)$ that takes two arguments and that returns the first one. Strictly speaking, $\sigma \to (\tau \to \sigma)$ takes a single argument and returns a function, which in turn takes an argument. Applications work analogously: If K denotes the term $\lambda x : \mathbb{Z}$, $(\lambda y : \mathbb{Z}, x)$, then K 1 = $(\lambda y : \mathbb{Z}, 1)$ and (K 1) 0 = 1. The function K in K 1, which is applied only to one argument, is said to be partially paplied.

Currying is so useful a concept that we will omit most parentheses, writing

```
\begin{split} \sigma \rightarrow \tau \rightarrow v & \text{ for } \sigma \rightarrow (\tau \rightarrow v) \\ & \text{ tu v } \text{ for } (\text{tu}) \text{ v} \\ \lambda \mathbf{x} : \sigma, \lambda \mathbf{y} : \tau, \mathbf{t} & \text{ for } \lambda \mathbf{x} : \sigma, (\lambda \mathbf{y} : \tau, \mathbf{t}) \end{split}
```

and also

```
\lambda(\mathbf{x}:\sigma) (y:\tau), t for \lambda\mathbf{x}:\sigma, \lambda\mathbf{y}:\tau, t \lambda\mathbf{x} y:\sigma, t for \lambda(\mathbf{x}:\sigma) (y:\sigma), t
```

In mathematics, it is customary to write binary operators in infix syntax—e.g., x+y. Such notations are also possible in Lean, as syntactic sugar for (+) x y. The parentheses around the + sign are necessary to disable the parsing of + as an infix operator. Partial application is possible with this syntax. For example, (+) 1 denotes the unary function that adds one to its argument. Other ways to write this function are λx , (+) 1 x and λx , 1 + x.

One way to work with Lean is to declare the types and constants we need using the constant or constants command. Consider the following declarations:

```
constants a b : \mathbb{Z} constant f : \mathbb{Z} \to \mathbb{Z} constant g : \mathbb{Z} \to \mathbb{Z} \to \mathbb{Z} #check \lambda x : \mathbb{Z}, g (f (g a x)) (g x b) #check \lambda x, g (f (g a x)) (g x b)
```

The first three lines introduce four constants (a, b, f, g), which can be used to form terms. The last two lines invoke the #check diagnosis command to type-check some terms and print their types.

The constant(s) command can be used even to declare types. The next example introduces a type of "trooleans" and three values of that type:

```
constant trool : Type
constants ttrue tfalse tmaybe : trool
```

The constants thus introduced have no definition. They are fully unspecified. For example, we cannot tell whether ttrue, tfalse, and tmaybe are equal or not, and we know nothing about trool except that it has at least one value (e.g., ttrue).

Type Checking and Type Inference When Lean parses a term, it checks whether the term is well typed. In the process, Lean will try to infer the types of bound variables if those are omitted—e.g., the type of x in λx , 1 + x. Type inference helps keep notations lighter and saves some keystrokes.

For simple type theory, type checking and type inference are decidable problems. Advanced features such as overloading (the possibility to reuse the same name for several constants—e.g., $0:\mathbb{N}$ and $0:\mathbb{R}$) can lead to undecidability [20]. Lean takes a pragmatic approach, by assuming that numerals 0, 1, 2, etc., are of type \mathbb{N} if several types are possible.

Lean's type system can be expressed as a formal system. A *formal system* consists of *judgments* and of (*derivation*) *rules* for producing judgments. A typing judgment has the form $\Gamma \vdash \mathbf{t} : \sigma$, meaning that term \mathbf{t} has type σ in context Γ . The context gives the types of the variables in \mathbf{t} that are not bound by any λ . There

are four typing rules, corresponding to the four kinds of terms:

$$\begin{array}{c|c} \hline \Gamma \vdash c : \sigma & \text{if c is declared with type σ} \\ \hline \hline \Gamma \vdash c : \sigma \to \tau & \Gamma \vdash u : \sigma \\ \hline \hline \Gamma \vdash t : \sigma \to \tau & \Gamma \vdash u : \sigma \\ \hline \hline \Gamma \vdash t u : \tau & \\ \hline \hline \hline \Gamma \vdash (\lambda x : \sigma, t) : \sigma \to \tau \\ \hline \hline \hline \hline \Gamma \vdash x : \sigma & \text{VAR} \quad \text{if $x : \sigma$ occurs in Γ} \\ \hline \end{array}$$

The first and last rules, labeled CST and VAR, have no premises, but they have side conditions that must be satisfied for the rules to apply. The other two rules take one or two judgments as premises and produce a new judgment.

We can use this rule system to prove that a given term is well typed by working our way backwards (i.e., upwards) and applying the rules, forming a *formal derivation* of a typing judgment. Like natural trees, derivation trees are drawn with their root at the bottom. The derived judgment appears at the root, and each branch ends with the application of a premise-less rule. Rule applications are indicated by a horizontal bar and a label. The following typing derivation establishes that the term $\lambda x: \mathbb{Z}$, abs x has type $\mathbb{Z} \to \mathbb{N}$ in the empty variable context:

$$\begin{tabular}{c|c} \hline \hline $x:\mathbb{Z}\vdash abs:\mathbb{Z}\to\mathbb{N}$ & \hline $x:\mathbb{Z}\vdash x:\mathbb{Z}$ & APP \\ \hline \hline $x:\mathbb{Z}\vdash abs\:x:\mathbb{N}$ & \\ \hline $\vdash(\lambda x:\mathbb{Z},\:abs\:x):\mathbb{Z}\to\mathbb{N}$ & LAM \\ \hline \end{tabular}$$

Reading the proof from the root upwards, notice how LAM moves the variable bound by the λ -expression to the context, making an application of VAR possible further up the tree.

Type inference is a generalization of type checking where the types on the right-hand side of the colon (:) in judgments may be replaced by placeholders. Lean's type inference is based on an algorithm due to J. Roger Hindley [11] and Robin Milner [15], which also forms the basis of Haskell, OCaml, and Standard ML. The algorithm generates type constraints involving placeholders α , β , γ , ..., and attempts to solve them using a type unification procedure. For example, when inferring the type α of λx , abs x, Lean would perform the following schematic type derivation:

In addition, Lean would generate constraints to ensure that all the rule applications are legal:

- 1. For the LAM application, the type of λx , abs x must have the form $\beta \to \gamma$, for some types β and γ . Thus, Lean would generate the constraint $\alpha = \beta \to \gamma$.
- 2. For the CST application, the type of abs must correspond to the declaration as $\mathbb{Z} \to \mathbb{N}$. Thus, Lean would generate the constraint $\beta \to \gamma = \mathbb{Z} \to \mathbb{N}$.

By putting the two constraints together, we obtain the solution $\alpha = \mathbb{Z} \to \mathbb{N}$, which is indeed the type that Lean infers for λx , abs x.

Type Inhabitation Given a type σ , the type inhabitation problem consists of finding an "inhabitant" of that type—i.e., a term of type σ . It may seem like a pointless exercise, but as we will see in Chapter 3, there is a close connection between this problem and the problem of finding a proof of a proposition. In other words, seemingly silly exercises of the form "specify a term of type σ " are good practice towards mastery of theorem proving.

To create a term of a given type, you can recursively apply a combination of the following two steps:

- 1. If the type is of the form $\sigma \to \tau$, a possible inhabitant is an anonymous function, of the form λx , _, where _ is a placeholder for a missing term. Lean will mark _ as an error; if you hover over it in Visual Studio Code, a tooltip will appear, specifying the type of the missing term (here, τ) as well as any constants declared in the context.
- 2. Given a type σ (which may be a function type), you can use any constant $c: \tau_1 \to \cdots \to \tau_n \to \sigma$ to build a term of that type. For each of the n arguments to c, you need to pass a placeholder, yielding $c = \cdots = 0$.

The placeholders can be eliminated recursively using the same procedure. As an example, we will apply the procedure to find a term of type

$$(\alpha \to \beta \to \gamma) \to ((\beta \to \alpha) \to \beta) \to \alpha \to \gamma$$

Initially, only step 1 is possible, with $\sigma:=\alpha\to\beta\to\gamma$ and $\tau:=((\beta\to\alpha)\to\beta)\to\alpha\to\gamma$. This results in the term λf , _, which has the right type but has a placeholder left. Since the argument f has type $\alpha\to\beta\to\gamma$, a function type, it makes sense to use the name f for it. Then we continue recursively with the placeholder, of type $((\beta\to\alpha)\to\beta)\to\alpha\to\gamma$. Again, only step 1 is possible, so we end up with the term λf , λg , _, where the placeholder has type $\alpha\to\gamma$. A third application of step 1 yields the term λf , λg , λa , _, where the placeholder has type γ .

At this point, step 1 is no longer possible. Let us see if step 2 is applicable. The context surrounding the placeholder contains the following local constants:

$$\begin{aligned} \mathbf{f} &: \alpha \to \beta \to \gamma \\ \mathbf{g} &: (\beta \to \alpha) \to \beta \\ \mathbf{a} &: \alpha \end{aligned}$$

Recall that we are trying to build a term of type γ . The only constant we can use to achieve this is f: It takes two arguments and returns a value of type γ . So we replace the placeholder with the term f_- , where the two new placeholders stand for the two missing arguments. Putting everything together, we now have the term λf , λg , λa , f_- .

Following f's type, the placeholders are of type α and β , respectively. The first placeholder is easy to fill, using step 2 again, by simply supplying a, which has type α , with no arguments. For the second placeholder, we apply step 2 with the constant g, which is the only source of β s. Since g takes an argument, we must supply a placeholder. This means our current term is λf , λg , λa , f a (g_{-}) .

We are now close to our goal. The remaining placeholder has type $\beta \to \alpha$, corresponding to g's argument type. Applying step 1, we replace the placeholder with λb , _, where the new placeholder has type α . Like above, we can supply a. Our final term is λf , λg , λa , f a $(g(\lambda b, a))$ —i.e, λf g a, f a $(g(\lambda b, a))$.

The above derivation was tedious but deterministic: At each point, either step 1 or 2 was applicable, but never both. This will not always be the case. We might encounter deadends and need to backtrack. We may also fail altogether, with no possiblity to backtrack; for example, in the empty context, it is impossible to supply a witness for α .

1.2 Type Definitions

A distinguishing feature of Lean's calculus of inductive constructions is its built-in support for inductive types. An *inductive type* is a type whose values by applying special constants called *constructors*. Inductive types are a concise way of representing acyclic data in a program. You may know them under some other, largely equivalent names, including algebraic data types, inductive data types, freely generated data types, or simply "data types" or "datatypes."

Natural Numbers The "Hello, World!" example of inductive types is the type $\mathbb N$ of natural numbers. In Lean, it can be defined as follows:

```
inductive nat : Type | zero : nat | succ : nat \rightarrow nat
```

The first line announces to the world that we are introducing a new type called nat, intended to represent the natural numbers. The second and third line declare two new constructors, nat.zero: nat and nat.succ: nat \rightarrow nat, that can be used to build values of type nat. Following an established convention in computer science and logic, counting starts at zero. The second constructor is what makes this inductive definition interesting—it requires an argument of type nat to produce a value of type nat. The terms

```
nat.zero    nat.succ nat.zero    nat.succ (nat.succ nat.zero)
```

and so on denote the different values of type nat—zero, its successor, its successor's successor, etc. This notation is called *unary*, or *Peano*, after the logician Giuseppe Peano.

The general format of all type declarations presented in this chapter is

For the naturals, it is possible to convince Lean to use the familiar names \mathbb{N} , 0, 1, 2, etc., and indeed the built-in nat type offers such syntactic sugar. But using the lengthier notations sheds more light on the syntax of Lean's type definitions, and anyway syntactic sugar is known to cause cancer of the semicolon [19].

In the Lean file accompanying this chapter, the definition of nat is enclosed by a namespace, delimited by the commands namespace my_nat and end my_nat, to keep its effects contained to a portion of the file. After end my_nat, any occurrence of nat, nat.zero, or nat.succ refers to Lean's predefined type of natural numbers. Similarly, the entire file is enclosed in the LoVe namespace to prevent name clashes with existing Lean libraries.

We can inspect an earlier definition at any point in Lean by using the #print command. For example, #print nat within the my_nat namespaces displays the following information:

```
inductive LoVe.my_nat.nat : Type
constructors:
LoVe.my_nat.nat.zero : nat
LoVe.my_nat.nat.succ : nat -> nat
```

The focus on natural numbers is one of the many features of this book that reveal a bias towards computer science. Number theorists would be more interested in the integers $\mathbb Z$ and the rational numbers $\mathbb Q$; analysts would want to work with the real numbers $\mathbb R$ and the complex numbers $\mathbb C$. But the natural numbers are ubiquitous in computer science and enjoy a very simple definition as an inductive type. They can also be used to build other types, as we will see in Chapter 13.

Arithmetic Expressions If we were to specify a calculator program or a programming language, we would likely need to define a type to represent arithmetic expressions. The next example shows how this could be done in Lean:

```
\begin{array}{l} \text{inductive aexp} : \text{Type} \\ \mid \text{num} : \mathbb{Z} \to \text{aexp} \\ \mid \text{var} : \text{string} \to \text{aexp} \\ \mid \text{add} : \text{aexp} \to \text{aexp} \to \text{aexp} \\ \mid \text{sub} : \text{aexp} \to \text{aexp} \to \text{aexp} \\ \mid \text{mul} : \text{aexp} \to \text{aexp} \to \text{aexp} \\ \mid \text{div} : \text{aexp} \to \text{aexp} \to \text{aexp} \end{array}
```

Mathematically, this definition is equivalent to defining the type aexp inductively by the following formation rules:

- 1. For every integer i, we have that num i is an aexp value.
- 2. For every character string x, we have that var x is an aexp value.
- 3. If e_1 and e_2 are aexp values, then so are add e_1 e_2 , sub e_1 e_2 , mul e_1 e_2 , and div e_1 e_2 .

The above definition is exhaustive. The only possible values for aexp are those built using formation rules 1 to 3. Moreover, aexp values by appealing to different rules are considered distinct. These two properties of inductive types are captured by the motto "No junk, no confusion," due to Joseph Goguen.

Comparison with Java At this point, it may be instructive to compare the concise Lean specification of aexp above with the canonical Java program that achieves the same. The program consists of one interface and six classes that implement it, corresponding to the aexp type and its six constructors:

```
public interface AExp {
}
public class Num implements AExp {
    public int num;
    public Num(int num) { this.num = num; }
}
public class Var implements AExp {
    public String var;
    public Var(String var) { this.var = var; }
}
public class Add implements AExp {
    public AExp left;
    public AExp right;
    public Add(AExp left, AExp right)
    { this.left = left; this.right = right; }
}
public class Sub implements AExp {
    public AExp left;
    public AExp right;
    public Sub(AExp left, AExp right)
    { this.left = left; this.right = right; }
}
public class Mul implements AExp {
    public AExp left;
    public AExp right;
    public Mul(AExp left, AExp right)
    { this.left = left; this.right = right; }
}
public class Div implements AExp {
    public AExp left;
   public AExp right;
    public Div(AExp left, AExp right)
```

```
{ this.left = left; this.right = right; }
}
```

Admittedly, the Java version is more flexible, because users can add new "constructors" simply by providing new classes. But this extensionality comes at a price, especially when we want to reason logically about a program. Extensibility is possible in Lean as well if desired.

Comparison with C In C, the natural counterpart of an inductive type is a tagged union. The type declarations would be as follows:

```
#include <stddef.h>
#include <stdlib.h>
enum AExpKind {
    AET_NUM, AET_VAR, AET_ADD, AET_SUB, AET_MUL, AET_DIV
};
struct aexp;
struct aexp_num {
    int num;
};
struct aexp_var {
    char var[1024];
};
struct aexp_binop {
   struct aexp *left;
    struct aexp *right;
};
struct aexp {
    enum AExpKind kind;
    union {
        struct aexp_num anum;
        struct aexp_var avar;
        struct aexp_binop abinop;
    } data;
};
```

Corresponding to each constructor in Lean, we would need to write a function to allocate an aexp object of the right size in memory. Here is the definition of the function corresponding to the first constructor, num:

```
struct aexp *create_num(int num)
{
    struct aexp *res = malloc(offsetof(struct aexp, data) +
        sizeof(struct aexp_num));
```

```
res->kind = AET_NUM;
res->data.anum.num = num;
return res;
}
```

The subtle pointer arithmetic for the malloc call is needed if we want to allocate exactly the right amount of memory.

Lists The next type we consider is that of finite lists:

```
inductive list (\alpha : Type) : Type | nil : list | cons : \alpha \rightarrow list \rightarrow list
```

The type is *polymorphic*: It is parameterized by a type α , which we can instantiate with concrete types. For example, $\mathtt{list}\ \mathbb{Z}$ is the type of lists over integers, and $\mathtt{list}\ (\mathtt{list}\ \mathbb{R})$ is the type of lists of lists of real numbers. The type constructor \mathtt{list} takes a type as argument and returns a type. Inside the definition, because α is fixed on the left-hand side of the column, it is sufficient to write \mathtt{list} to obtain $\mathtt{list}\ \alpha$.

The following commands allow us to inspect the constructors' types:

```
#check list.nil
#check list.cons
```

The output is

```
list.nil : \Pi(\alpha : \mathsf{Type}), list \alpha list.cons : \mathsf{?M}_1 \to \mathsf{list} \; \mathsf{?M}_1 \to \mathsf{list} \; \mathsf{?M}_1
```

Informally:

- The nil constructor takes a type α as argument and produces a result of type list α . (The Π syntax for the type of arguments will be explained in Chapter 3.)
- The cons constructor takes an element (the *head*) of some arbitrary type ?M_1 as argument and a list over ?M_1 (the *tail*) and produces a result of type list ?M_1. Unlike for nil, there is no need to pass a type argument to cons—the type is inferred from the first argument. If we want to pass the type argument explicitly, we need to write an at sign (@) in front of the constant: @list.cons. We then have the type @list.cons : $\Pi(\alpha: Type)$, $\alpha \to \text{list } \alpha \to \text{list } \alpha$.

Lean's built-in lists offer syntactic sugar for writing lists:

```
[] for list.nil h:: t \text{ for list.consht} [x_1, \ldots, x_n] \text{ for } x_1: \ldots : x_n:: []
```

The :: operator, like all other binary operators, binds less tightly than function application. Thus, f x :: reverse ys is parsed as (f x) :: (reverse ys). It is good practice to avoid needless parentheses. They can quickly impair readability. In addition, it is important to put spaces around infix operators, to suggest the right precedences; it is all too easy to misread f x :: reverse ys as f (x :: reverse) ys.

1.3 Function Definitions

If all we want is to declare a function, we can use the constant command described above (Section 1.1). But usually, we want to *define* the function's behavior, and for this we can use the def command.

Going back to the arithmetic expression example (Section 1.2), if we wanted to implement an eval function in Java, we would probably add it as part of AExp's interface and implement it in each subclass. For Add, Sub, Mul, and Div, we would recursively call eval on the left and right objects.

In Lean, the syntax is very compact. We define a single function and use *pattern* matching to distinguish the six cases:

The keyword def introduces the definition. The general format of pattern matching is

```
def name (vars_1 : type_1) \dots (vars_m : type_m) : type
| patterns_1 := result_1
| \vdots
| patterns_n := result_n
```

Patterns may contain variables, whose scope is the corresponding right-hand side, as well as constructors. For example, in the second case of eval's definition, the variable x can be used in the right-hand side env x.

Some definitions do not need pattern matching. For these, the syntax is simply

```
def name (vars_1: type_1) \dots (vars_m: type_m) type := result
```

We can have pattern matching without recursion (e.g., in the num and var cases above), and we can have recursion without pattern matching.

Structural recursion is a form of recursion that allows us to peel off one or more constructors from the value on which we recurse. The eval function above is structurally recursive. Such functions are guaranteed to call themselves only finitely many times before the recursion stops. This is a prerequisite for establish that the function terminates, an important property to ensure logical consistency and termination of evaluation.

For structurally recursive functions, Lean can automatically prove termination. For more general recursive schemes, the termination check may fail. Sometimes it does so for a good reason, as in the following example:

```
-- illegal def evil : \mathbb{N} \to \mathbb{N} | n := nat.succ (evil n)
```

If Lean were to accept this definition, we could exploit it to prove that 0 = 1, by subtracting evil n on each side of the equation evil n = nat.succ (evil n).

The basic arithmetic operations on natural numbers, such as addition, can be defined by structural recursion:

```
\begin{array}{lll} \texttt{def} & \texttt{add} & : \; \mathbb{N} \; \to \; \mathbb{N} \\ \mid \; \texttt{m} \; \texttt{nat.zero} & := \; \texttt{m} \\ \mid \; \texttt{m} \; (\texttt{nat.succ} \; \texttt{n}) \; := \; \texttt{nat.succ} \; (\texttt{add} \; \texttt{m} \; \texttt{n}) \end{array}
```

We pattern-match on two arguments at the same time, distinguishing the case where the second argument is zero and the case where it is nonzero. Each recursive call to add peels off one succ constructor from the second argument. (Given that addition is symmetric, there is no deep reason for choosing to recurse on the second argument.)

We can evaluate the result of applying add to concrete values using #reduce or #eval:

```
#reduce add 2 7
#eval add 2 7
```

Both commands print 9, as expected.

It is generally good practice to provide a few tests each time we define a function, to ensure that it behaves as expected. You can even leave the #reduce or #eval calls in your Lean files as documentation.

The definition of multiplication is similar to that of addition:

```
\begin{array}{lll} \texttt{def mul} & : \; \mathbb{N} \to \mathbb{N} \to \mathbb{N} \\ \mid \; \; \texttt{nat.zero} & := \; \texttt{nat.zero} \\ \mid \; \texttt{m} \; (\texttt{nat.succ} \; \texttt{n}) & := \; \texttt{add} \; \texttt{m} \; (\texttt{mul} \; \texttt{m} \; \texttt{n}) \end{array}
```

This time, #reduce prints 14:

```
#reduce mul 2 7
```

The power operation ("m to the power of n") can be defined in various ways. Our first proposal is analogous to the definition of multiplication:

Since the first argument does not change in the recursive calls, we can factor it out and put it next to the function's name, before the colon introducing the function's type:

```
\begin{array}{lll} \text{def power}_2 \ (\text{m} : \mathbb{N}) : \mathbb{N} \to \mathbb{N} \\ \mid 0 & := 1 \\ \mid \ (\text{nat.succ n}) := \text{m} * \text{power}_2 \ \text{n} \end{array}
```

In fact, we already saw a similar syntax for the type argument α of the list constructor (Section 1.2). Yet another definition is possible by first introducing a general-purpose iterator and then using it with the right arguments:

```
 \begin{tabular}{lll} $\operatorname{def power_3}$ (m n : \mathbb{N}) : \mathbb{N} := \\ $\operatorname{iter}$ \mathbb{N} \ 1 (\lambda 1, \ m * 1) n \\ \end{tabular}
```

Notice that the second definition is not recursive.

Recursive functions on lists can be defined in a similar way:

The append function takes three arguments: a type α and two lists of type list α . By passing the placeholder _, we leave it to Lean to infer the type α from the type of the other two arguments.

To make the type argument implicit by default, we can put it in curly braces {}:

The at sign (@) can be used to make the implicit arguments explicit.

We can use syntactic sugar in the definition, both in the patterns on the lefthand sides of := and in the right-hand sides:

```
def append<sub>3</sub> {\alpha : Type} : list \alpha \to lis
```

In Lean's standard library, ++ is provided as syntactic sugar for the append function. We can use it to define a function that reverses a list:

1.4 Lemma Statements

What makes Lean a proof assistant and not only a programming language is that we can state lemmas about the types and constants we define and prove that they hold. We will use the terms lemma, theorem, corollary, fact, property, and true statement more or less interchangeably. Similarly, logical formulas, propositions, and true/false statements will all mean the same.

In Lean, propositions are simply terms of type Prop. A proposition that can be proved is a theorem (or lemma, corollary, etc.); otherwise it is a nontheorem

or false statement. Mathematicians sometimes use "proposition" as a synonym for theorem (e.g., "Proposition 3.14"), but in formal logic it is customary to let it encompass false statements.

Here are examples of true statements that can be made about the addition, multiplication, and list reversal operations defined in Section 1.3:

```
lemma add_comm (m n : \mathbb{N}):
    add m n = add n m :=
  sorry
  lemma add_assoc (1 m n : \mathbb{N}) :
    add (add l m) n = add l (add m n) :=
  sorry
  lemma mul_comm (m n : \mathbb{N}) :
    mul m n = mul n m :=
  sorry
  lemma mul_assoc (l m n : \mathbb{N}) :
    mul (mul l m) n = mul l (mul m n) :=
  sorry
  lemma mul_add (1 m n : \mathbb{N}) :
    mul l (add m n) = add (mul l m) (mul l n) :=
  sorry
  lemma reverse_reverse \{\alpha : \text{Type}\}\ (\text{xs} : \text{list}\ \alpha) :
    reverse (reverse xs) = xs :=
  sorry
The general format is
  lemma name (vars<sub>1</sub>: type<sub>1</sub>) ... (vars<sub>m</sub>: type<sub>m</sub>) statement := proof
```

The := symbol separates the lemma's statement and its proof. In the examples above, we put the marker sorry as a placeholder for the actual proof. The marker is quite literally an apology to future readers and to Lean for the absence of an actual proof. In Chapters 2 and 3, we will see how to eliminate these sorrys.

1.5 Summary of New Syntax

At the end of this and other chapters, we include a brief summary of the syntax introduced in the chapter. Some commands have multiple meanings, which will be introduced gradually. We refer to *The Lean Reference Manual* [3] and the *Theorem Proving in Lean* [1] and *Programming in Lean* [2] tutorials for details and to Lean's source code for the ultimate truth.

Declarations

constant(s) declares unspecified new constants or types

def defines a new constant

inductive introduces a type and its constructors

lemma states a lemma

namespace ... end puts all declarations in a named scope (to avoid clashes)

Proof Commands

sorry stands for a missing proof or definition

Diagnosis Commands

#check checks and prints type of a term

#eval executes a term using an optimized interpreter prints definition of a constant or function executes a term using Lean's kernel

Chapter 2

Tactical Proofs

In this chapter, we see how to prove Lean lemmas using tactics, and we review the most important Lean tactics. A *tactic* is a procedure that operates on the goal—the formula to prove—and either fully solves it or produces new subgoals. When we state a lemma, the lemma statement is the initial goal. A proof is complete once all (sub)*goals have been eliminated using suitable tactics.

The tactics described here are documented in more detail in Chapter 5 of Theorem Proving in Lean [1] and Chapter 6 of The Lean Reference Manual [3].

2.1 Tactic Mode

In Chapter 1, whenever a proof was needed, we simply put a <code>sorry</code> placeholder. For a tactical proof, we will now write <code>begin</code> and <code>end</code> instead to enter and leave tactic mode. In tactic mode, we can apply a sequence of tactics, separated by commas (,).

Tactics operate on the proof goal, which consists of the conjecture χ that we want to prove and of a context Γ that provides constant declarations of the form $c:\tau$ and hypotheses of the form $h:\varphi$. We write C, $H \vdash \chi$ to denote a goal, where C is a set of (local) constants, H is a set of (local) hypotheses, and χ is the goal's conclusion. The constants and hypotheses form a set; the comma is read as set union. We call C, H the local context of the goal.

This will become clearer with an example:

```
lemma fst_of_two_props : \forall a \ b : Prop, \ a \to b \to a := begin intros a b, intros ha hb, apply ha end
```

Note that the implication arrow \to is right-associative; this means that $a \to b \to a$ is the same as $a \to (b \to a)$. In the example, three tactics are invoked, each on its own line. Let us trace their behavior:

1. Initially, the goal is simply the lemma statement:

```
\vdash \forall a b : Prop, a \rightarrow b \rightarrow a
```

2. The intros a b tactic tells Lean to fix two arbitrary constants, a and b, corresponding to the two bound variables of the same names. It is customary, although not mandatory, to name the constants after the bound variables. The tactic mimics how

mathematicians work on paper: To prove a \forall -quantified formula, it suffices to prove it for some arbitrary value of the bound variable. The goal becomes

```
ab : Prop \vdash a \rightarrow b \rightarrow a
```

3. Next, the intros ha hb tactic tells Lean to move the assumptions a and b to the context and to call these hypotheses ha and hb. Indeed, to prove an implication, we can take its left-hand side as hypothesis and try to prove its right-hand side. The goal becomes

```
(ab: Prop) (ha: a) (hb: b) \vdash a
```

4. The apply ha tactic tells Lean to match ha : a against the goal a. Since a is syntactically equal to a, this finishes the proof.

We can save ourselves the intros invocations by declaring the constants and hypotheses as parameters of the lemma, as follows:

```
lemma fst_of_two_props2 (a b : Prop) (ha : a) (hb : b) :
   a :=
begin
   apply ha
end
```

2.2 Basic Tactics

We already saw the intros and apply tactics. These are the staples of tactical proofs. Other basic tactics include exact, assumption, refl, ac_refl, and use.

These basic tactics can go a long way, if we are patient enough to carry out the reasoning using them, without appealing to stronger proof automation. They can also be used to solve various logic puzzles.

intro(s)

```
intro [name] intros [name_1 \dots name_n]
```

(The symbols $[\]$ encompass optional syntax, here and elsewhere.) The intro tactic moves the leading \forall -quantified variable or the leading assumption $a \to from$ the conclusion to the local context. The tactic takes as an optional argument the name to give to the variable or to the assumption in the context, overriding the default name. The intros variant can be used to move several variables or assumptions at once.

apply

```
apply lemma
```

The apply tactic matches the goal's conclusion with the conclusion of the specified lemma and adds the lemma's hypotheses as new goals.

exact

```
exact lemma
```

The exact tactic matches the goal's conclusion with the specified lemma, closing the goal. We can often use apply in such situations, but exact communicates our intentions better. In the example from Section 2.1, we could have used exact ha instead of apply ha.

assumption

The assumption tactic finds a hypothesis from the local context that matches the goal's conclusion and applies it to solve the goal. In the example from Section 2.1, we could have used assumption instead of apply ha.

refl

The ref1 tactic solves goals of the 1 = r, where the two sides are equal *up* to computation. We also say, equivalently, that the two terms are *definitionally equal*. Computation means unfolding of definitions, reduction of an application of a λ -expression to an argument, and more. These conversions have traditional names. The main conversions are listed below together with examples, in a context with def double $(n : \mathbb{N}) : \mathbb{N} := n + n$:

```
\begin{array}{ll} \alpha\text{-conversion} & (\lambda \texttt{x}, \texttt{f} \texttt{x}) = (\lambda \texttt{y}, \texttt{f} \texttt{y}) \\ \beta\text{-conversion} & (\lambda \texttt{x}, \texttt{f} \texttt{x}) \texttt{a} = \texttt{f} \texttt{a} \\ \delta\text{-conversion} & \text{double } \texttt{m} = \texttt{m} + \texttt{m} \\ \zeta\text{-conversion} & (\text{let } \texttt{n} : \mathbb{N} := 2 \text{ in } \texttt{n} + \texttt{n}) = 4 \\ \eta\text{-conversion} & (\lambda \texttt{x}, \texttt{f} \texttt{x}) = \texttt{f} \\ \iota\text{-conversion} & \text{prod.fst } (\texttt{a}, \texttt{b}) = \texttt{a} \end{array}
```

ac_refl

Proves 1 = r, where the two sides are equal up to associativity and commutativity. This works for binary operations that are registered as associative and commutative (e.g., + and * on arithmetic types). We will see an example in Section 2.6.

use

```
use term
```

The use tactic allows us to supply a witness for an existential quantifier. In the example below, where def double $(n:\mathbb{N}):\mathbb{N}:=n+n$, the witness is 0, and the proof goes through by computation once the witness has been supplied:

```
lemma nat_exists_double_iden : \exists n : \mathbb{N}, double n = n := begin use 0, refl
```

2.3 Proofs about Logical Connectives and Quantifiers

Before we learn to reason about natural numbers, lists, or other data types, we must first learn to reason about the logical connectives and quantifiers of Lean's logic. Let us start with a simple example: commutativity of conjunction (\wedge).

```
lemma and_swap (a b : Prop) : 
 a \wedge b \rightarrow b \wedge a := 
 begin 
 intro hab, 
 apply and.intro, 
 apply and.elim_right, 
 exact hab, 
 apply and.elim_left,
```

```
exact hab
```

At this point, we recommend that you move the cursor over the example in Visual Studio Code, to see the sequence of proof states. By putting the cursor immediately after each comma, you can see the effect of the command on that line.

The proof is a typical combination intro, apply, and exact tactics. The lemmas used are

```
and.intro  \begin{array}{ll} :?a \rightarrow ?b \rightarrow ?a \wedge ?b \\ \text{and.elim\_left} & :?a \wedge ?b \rightarrow ?a \\ \text{and.elim\_right} & :?a \wedge ?b \rightarrow ?b \\ \end{array}
```

where the question marks indicate variables that can be instantiated—for example, by matching the goal's conclusion against one the conclusion of a lemma.

The three lemmas above are the introduction rule and the two elimination rules associated with conjunction. An introduction rule for a symbol (e.g., \land) is a lemma whose conclusion contains that symbol. Dually, an elimination rule has the symbol in an assumption. In the above proof, we apply the introduction rule associated with \land to prove the goal b \land a, and we apply the two elimination rules to extract b and a from the hypothesis a \land b.

The following is an alternative proof of the same lemma:

```
lemma and_swap_2: \forall a b : Prop, a \land b \rightarrow b \land a := begin intros a b hab, apply and.intro, { exact and.elim_right hab }, { exact and.elim_left hab } end
```

The lemma is stated differently, with a and b as \forall -quantified variables instead of parameters of the lemma. Logically, this is equivalent, but in the proof we must then introduce (using intro) a and b in addition to a \land b.

Another difference is the use of curly braces {}. When we face two or more goals to prove, it is generally good style to put each proof in its own block enclosed in curly braces. The {} tactic combinator focuses on the first subgoal; the tactic inside must solve it. In our example, the apply and intro invocation creates two subgoals, b and a.

The third difference between the proofs is that we now apply $\mathtt{and.elim_right}$ and $\mathtt{and.elim_left}$ directly to the hypothesis $\mathtt{a} \land \mathtt{b}$ to obtain \mathtt{b} and \mathtt{a} , respectively, instead of the combination of \mathtt{apply} and \mathtt{exact} used in the first proof. Quantifier instantiation has the same syntax—juxtaposition—as passing an argument to a function. This is not a coincidence, as we will see in Chapter 3.

The introduction and elimination rules for disjunction (\vee) are as follows:

```
\begin{array}{lll} \text{or.intro\_left} & : \forall b : \texttt{Prop}, ?a \rightarrow ?a \lor b \\ \text{or.intro\_right} & : \forall b : \texttt{Prop}, ?a \rightarrow b \lor ?a \\ \text{or.elim} & : ?a \lor ?b \rightarrow (?a \rightarrow ?c) \rightarrow (?b \rightarrow ?c) \rightarrow ?c \end{array}
```

The or.elim rule may seem counterintuitive at a first glance. In essence, it states that if we have $a \lor b$, then to prove an arbitrary c, it suffices to prove c when a holds and when b holds.

The introduction and elimination rules for equivalence (\leftrightarrow) are as follows:

```
\begin{array}{ll} \texttt{iff.intro} & : (?\texttt{a} \to ?\texttt{b}) \to (?\texttt{b} \to ?\texttt{a}) \to (?\texttt{a} \leftrightarrow ?\texttt{b}) \\ \texttt{iff.elim\_left} & : (?\texttt{a} \leftrightarrow ?\texttt{b}) \to ?\texttt{a} \to ?\texttt{b} \\ \texttt{iff.elim\_right} & : (?\texttt{a} \leftrightarrow ?\texttt{b}) \to ?\texttt{b} \to ?\texttt{a} \end{array}
```

The introduction and elimination rules for existential quantification (\exists) are as follows:

```
exists.intro : \forall w, (?p w \rightarrow (\exists x, ?p x))
exists.elim : (\exists x, ?p x) \rightarrow (\forall a, ?p a \rightarrow ?c) \rightarrow ?c
```

The introduction rule for \exists is what the use tactic uses internally. For example:

```
lemma nat_exists_double_iden2 :
    ∃n : N, double n = n :=
begin
    apply exists.intro 0,
    refl
end
```

The elimination rule for \exists has a similar "counterintuitive" flavor to the elimination rule for \lor .

For truth (true), there is only an introduction rule:

```
true.intro : true
```

For falsehood (false), there is only an elimination rule:

```
{\tt false.elim} \qquad \qquad : {\tt false} 
ightarrow {\tt ?a}
```

Negation (not) is defined in terms of implication and falsehood: $\neg \varphi$ abbreviates $\varphi \to \mathtt{false}$. Lean's logic is classical, with support for the law of excluded middle and proofs by contradiction:

```
classical.em : \forall a: \texttt{Prop}, \ a \lor \neg \ a classical.by_contradiction : (\neg ?a \to \texttt{false}) \to ?a
```

Finally, implication (\rightarrow) and universal quantification (\forall) are the dogs that did not bark. For both of them, intro is the introduction rule, and application (juxtaposition) is the elimination rule. For example, if we have hab: $a \rightarrow b$ and ha: a, then hab ha: b is a proof of b.

For proving logic puzzles involving connectives and quantifiers, we advocate a "mindless," "video game" style of reasoning that relies mostly basic tactics such as intro(s) and apply. Here are some strategies that often work:

- If the goal's conclusion is an implication $\varphi \to \psi$, invoke intro $h\varphi$ to move φ into your hypotheses: ... $(h\varphi : \varphi) \vdash \psi$.
- If the goal's conclusion is a universal quantification $\forall x : \sigma$, φ , invoke intro x to move it into the local context: ... $(x : \sigma) \vdash \psi$.
- Otherwise, look for a hypothesis or a lemma whose conclusion has the same shape as the goal's conclusion (possibly containing variables that can be matched against the goal), and apply it. For example, if the goal's conclusion is $\vdash \psi$ and you have a hypothesis $h\varphi\psi:\varphi\to\psi$, try apply $h\varphi\psi$.
- A negated goal $\vdash \neg \varphi$ is definitionally equal to $\vdash \varphi \rightarrow \mathtt{false}$, so you can invoke intro $\mathtt{h}\varphi$ to produce the subgoal $\mathtt{h}\varphi : \varphi \vdash \mathtt{false}$. Expanding negation's definition by invoking dunfold not (described in Section 2.4) is often a good strategy.
- Sometimes you can make progress by replacing the goal by false, by entering apply false.elim. As next step, you would typically apply a hypothesis of the form $\varphi \to \text{false or } \neg \varphi$.
- When you face several choices (e.g., between or.intro_left and or.intro_right), remember which choices you have made, and backtrack when you reach a dead end or have the impression you are not making any progress.

If you have difficulties carrying out a proof, it can be a good idea to check whether
the goal actually is provable under the given assumptions. Be also aware that even
if you started with a provable lemma statement, it is possible that the goal is not
provable (e.g., if you used "unsafe" rules such as or.intro_left).

2.4 Rewriting Tactics

The rewriting tactics replace equals for equals. By default, they apply to the goal's conclusion, but they can also be applied to hypotheses specified using an at clause:

```
at \vdash applies to the conclusion
at h_1 \ldots h_n applies to the specified hypotheses
at * applies to all possible hypotheses and to the conclusion
```

rw

```
rw lemma [at position]
```

The rw tactic applies a single equation as a left-to-right rewrite rule, once. To apply a lemma as a right-to-left rewrite rule, put a short left arrow (\leftarrow) in front of its name, or pass the lemma to eq.symm.

For example, given the lemma 1 : b = a and the goal $\vdash h$ a b b, the tactic invocation rw 1 produces the subgoal $\vdash h$ a a b, whereas $rw \leftarrow 1$ or rw (eq. symm 1) produces the subgoal $\vdash h$ b b b.

```
rw [lemma<sub>1</sub>, ..., lemma<sub>n</sub>] [at position]
```

The above syntax abbreviates $rw lemma_1, ..., rw lemma_n$. (The thick, small brackets [] are part of the syntax.)

simp

```
simp [at position]
```

The simp tactic applies a standard set of rewrite rules, called the simp set, exhaustively. The simp set can be extended by putting the @[simp] attribute on lemmas. The simp tactic is more powerful than rw because it can rewrite terms containing bound variables (e.g., bound by λ , \forall , or \exists).

```
simp [lemma_1, ..., lemma_n] [at position]
```

For the above simp variant, the specified lemmas are temporarily added to the simp set. In the lemma list, an asterisk (*) can be use to represent all local hypotheses. The minus sign (-) in front of a lemma name temporarily removes the lemma from the simp set. To apply a lemma as a right-to-left rewrite rule, prefix its name with a short left arrow (\leftarrow).

For example, given the lemma 1 : b = a and the goal $\vdash h \ a \ b$ b, the tactic invocation simp [1] produces the subgoal $\vdash h \ a \ a$ a, where both occurrences of b have been replaced by a, whereas $simp [\leftarrow 1]$ produces the subgoal $\vdash h \ b \ b$ b.

To find out exactly what simp does, you can enable tracing by adding the command set_option trace.simplify.rewrite true to your Lean file.

dunfold

```
dunfold constant<sub>1</sub>... constant<sub>n</sub> [at position]
```

The dunfold tactic expands the definition of one or more constants that are specified without pattern matching (e.g., not).

2.5 Induction Tactic

induction

```
induction variable
```

The induction tactic performs structural induction on the specified variable. This gives rise to as many subgoals as there are constructors in the definition of the variable's type. Induction hypotheses are available as hypotheses in the subgoals corresponding to recursive constructors (e.g., nat.succ or list.cons).

2.6 Proofs about Natural Numbers

Equipped with a tactic for structural induction, we can now carry out meaningful proofs about the addition and multiplication operations we defined by structural recursion in Section 1.3.

Addition is defined by recursion on its second argument. We will prove two lemmas, add_zero and add_succ, that give us recursive equations on the *first* argument. We start with add_zero:

```
lemma add_zero (n : N) :
  add 0 n = n :=
begin
  induction n,
  { refl },
  { simp [add, n_ih] }
end
```

The name n_{ih} is generated by the induction tactic. It refers to the induction hypothesis for the induction on n. We can supply our own names, and address the cases in an arbitrary order, by using the case keyword in front of each case, together with the name of the case and the desired names for the local constants and hypotheses introduced by the tactic. For example:

```
lemma add_zero2 (n : N) :
  add 0 n = n :=
begin
  induction n,
  case nat.zero {
    refl },
  case nat.succ : m ih {
    simp [add, ih] }
end
```

Instead of n and n_{ih} , we choose the names m and ih.

We can write the proof more concisely if we want:

```
lemma add_zero3 (n : N) :
   add 0 n = n :=
by induction n; simp [add, *]
```

The syntax by tactic abbreviates begin tactic end. It can be used when a single (possibly complex) tactic suffices to solve the goal.

The semicolon (;) is a tactic combinator. The complex tactic $tactic_1$; $tactic_2$ first applies $tactic_1$ and then applies $tactic_2$ to all subgoals emerging from $tactic_1$. Our example works because simp [add, *] solves not only the induction step but also the base case (for which refl would have sufficed).

We can keep on proving lemmas by structural induction:

```
lemma add_succ (m n : \mathbb{N}):
  add (nat.succ m) n = nat.succ (add m n) :=
begin
  induction n,
  case nat.zero {
   refl },
  case nat.succ : m ih {
    simp [add, ih] }
end
lemma add_comm (m n : \mathbb{N}) :
  add m n = add n m :=
  induction n,
  case nat.zero {
    simp [add, add_zero] },
  case nat.succ : m ih {
    simp [add, add_succ, ih] }
end
lemma add_assoc (1 m n : \mathbb{N}) :
  add (add 1 m) n = add 1 (add m n) :=
begin
  induction n,
  case nat.zero {
    refl },
  case nat.succ : m ih {
    simp [add, ih] }
```

Once we have proved that a binary operator is commutative and associative, it is a good idea to let Lean's automation (notably ac_refl) know about this. Here is the syntax to achieve this for add:

```
instance : is_commutative \mathbb N add := \langle add\_comm \rangle instance : is_associative \mathbb N add := \langle add\_assoc \rangle
```

Here is an example that uses ac_refl to reason up to associativity and commutatitivy of add:

```
lemma mul_add (1 m n : N) :
  mul 1 (add m n) = add (mul 1 m) (mul 1 n) :=
begin
  induction n,
  case nat.zero {
    refl },
  case nat.succ : m ih {
    simp [add, mul, ih],
    ac_refl }
end
```

2.7 Goal Management Tactics

The following tactics help manage the goal, by allowing us to give more meaningful names to local constants or hypotheses, remove useless hypotheses, or undo the effect of an intro invocation.

rename

```
rename constant-or-hypothesis new-name
```

The rename tactic renames a local constant or hypothesis.

clear

```
\verb|clear| constant-or-hypothesis|_1 \dots constant-or-hypothesis|_n
```

The clear tactic removes the specified local constants and hypotheses, as long as they are not used anywhere else in the goal.

revert

```
revert constant-or-hypothesis_1 \dots constant-or-hypothesis_n
```

The revert tactic performs the opposite of intros: Moves the specified local constants and hypotheses into the goal's conclusion using universal quantification (\forall) for constants and implication (\rightarrow) for hypotheses.

For example, if the goal is a : Prop, $ha : a \vdash b$, the tactic invocation revert a ha produces the subgoal $\vdash \forall a : Prop$, $a \rightarrow b$.

2.8 Summary of New Syntax

Commands

set_option changes or activates tracing, syntax output, etc.

Proof Commands

begin ... end applies a list of tactics in sequence

by applies a single tactic

Tactic Combinators

; applies the second tactic to all subgoals yielded by the first tactic $\{\,\dots\,\}$ focuses on the first subgoal; needs to solve that goal

Tactics

ac_refl proves 1 = r if 1 equals r up to associativity and commutativity matches the goal's conclusion with the lemma's conclusion apply assumption solves the goal using a local hypothesis removes a local constant or hypothesis from the goal clear unfolds a definition without pattern matching (e.g., not) dunfold solves the goal by the specified term exact performs structural induction on a variable of an inductive type induction moves \forall -quantified variables into the goal's hypotheses intro(s) proves 1 = r where 1 and r are equal up to computation refl

rename renames a local constant or hypothesis

revert undoes the effect of intro

rw applies the given rewrite rules once, in sequence simp applies a set of preregistered rewrite rules exhaustively use provides a witness to an existential quantifier

Chapter 3

Structured Proofs and Proof Terms

Tactics are a backward (or bottom-up) proof mechanism. They start from the goal and break it down. Often it makes sense to work forward: to start with what we already know and proceed step by step towards our goal. Structured proofs are a style that supports this reasoning. They can be combined with the tactical style. Backward proofs tend to be easier to write but harder to read. It is, after all, easier to destruct things than to construct them.

Structured proofs are syntactic sugar sprinkled over Lean's proof terms. All proofs, whether tactical or structured, are reduced internally to proof terms. As the name suggests, these proofs are represented as terms. We have seen some proof terms already, in Chapter 2: If we have hab: $a \rightarrow b$ and ha: a, then hab ha is a proof term for b, and we write hab ha: b.

The concepts covered here are described in more detail in Chapters 2 to 4 of *Theorem Proving in Lean* [1]. Nederpelt and Geuvers's textbook [16] and Van Raamdonk's lecture notes [22] are other useful references.

3.1 Structured Proofs

Structured proofs are built using a number of constructs. We review the main ones below.

Lemma Name

The simplest kind of structured proof is the name of a lemma. If we have

```
lemma some_name : some_proposition :=
```

then some_name can be used as a proof of some_proposition later—for example:

```
lemma some_other_name : some_proposition :=
some_name
```

We can do the same with local hypotheses, and we can pass arguments to lemmas or hypotheses to instantiate \forall quantifiers and discharge assumptions of a lemma.

As an more realistic example, suppose we have a lemma $\mathtt{add_comm}\ (\mathtt{m}\ \mathtt{n}: \mathbb{N}): \mathtt{add}\ \mathtt{m}\ \mathtt{n}=\mathtt{add}\ \mathtt{n}\ \mathtt{m},$ and suppose we want to prove its instance $\mathtt{add}\ \mathtt{0}\ \mathtt{n}=\mathtt{add}\ \mathtt{n}\ \mathtt{0}.$ This can be achieved neatly using the name of the lemma and two arguments:

```
lemma add_comm_zero_left (n : \mathbb{N}):
add 0 n = add n 0 :=
add_comm 0 n
```

This has the same effect as the tactical proof by exact add_comm 0 n, but is more concise.

assume

```
assume name_1 \dots name_n,
```

The assume command moves variables or assumptions from the goal into the local context, as local constants or hypotheses. It can be seen as a structured version of the introstactic. At the proof term level, it amounts to n λ binders.

have

```
have name : proposition := proof,
```

The have command lets us state and prove an intermediate lemma statement, which can refer to the local context. Often we put a _ placeholder for proposition if it can be inferred from the proof.

let

```
let name [: type] := term in
```

The let command introduces a new local definition. It can be used to name a complex object that occurs several times in the proof afterwards. It is similar to have but for computable data, not a proof. Often we omit the type if it can be inferred from the term. Folding or unfolding a let corresponds to ζ -conversion (Section 2.2).

show

```
show proposition, from structured-proof
show proposition, begin tactics end
show proposition, by tactic
```

The show command lets us repeat the goal to prove, which can be useful as documentation. It also allows us to rephrase the goal in an equal form up to computation. Instead of the syntax show proposition, from structured-proof, we can simply write structured-proof if we do not need to rephrase the goal.

The second and third syntaxes listed above remind us that it is possible to mix proof styles. Similarly, the proof part of a have command can be either a structured proof or a tactical proof introduced by begin or by. Generally, we tend to use structured proofs to sketch the main argument and resort to tactical proofs for proving subgoals or straightforward lemmas.

Another kind of mixture arises when we pass arguments to lemma names. For example, given, hab: $a \to b$ and ha: a, the tactic invocation exact hab ha will prove the goal b: hab ha is a (admittedly small) proof term nested in a tactic. Yet another kind of mixture arises when we open a begin... end block after invoking some structured proof commands such as assume and have, as in the following example:

```
lemma and_swap_2 (a b : Prop) : 
 a \wedge b \rightarrow b \wedge a := 
 assume hab : a \wedge b, 
 have ha : a := and.elim_left hab, 
 have hb : b := and.elim_right hab, 
 begin 
 apply and.intro, 
 { exact hb }, 
 { exact ha } 
 end
```

Here is a purer proof:

```
lemma and_swap (a b : Prop) : 
 a \wedge b \rightarrow b \wedge a := 
 assume hab : a \wedge b, 
 have ha : a := and.elim_left hab, 
 have hb : b := and.elim_right hab, 
 show b \wedge a, from and.intro hb ha
```

Forward Tactics

Two of the proof commands presented above—have and let—are also available as tactics. Many users, especially beginners, prefer the tactical mode. Even with tactics, it can be useful to reason in a forward fashion. Observe that the syntax for the tactic let is slightly different than for the structured proof command of the same name, with a comma (,) instead of the keyword in.

```
have [[name :] proposition] := proof,
let name [: type] := term,
```

3.2 Calculational Proofs

In informal mathematics, we often use transitive chains of equalities, inequalities, or equivalences (e.g., a = b = c, $a \ge b \ge c$, or $a \leftrightarrow b \leftrightarrow c$). In Lean, such calculational proofs are supported by the calc command. It provides a lightweight syntax and takes care of applying transitivity lemmas for preregistered relations (including equality and the arithmetic comparison operators).

The general syntax is as follows:

The horizontal dots (\dots) are part of the syntax. Each $proof_i$ establishes the statement $expr_{i-1}$ of i expri. The operators of i need not be identical, but they need to be compatible. For example, =, <, and \leq are compatible with each other, whereas > and < are not.

A simple example follows:

```
lemma two_mul_example (m n : \mathbb{N}) : 
 2 * m + n = m + n + m := 
 calc 2 * m + n = (m + m) + n : by rw two_mul 
 ... = m + n + m : by ac_refl
```

The horizontal dots stand for the term (m + m) + n, which we would have had to repeat had we performed the proof with two haves:

```
lemma two_mul_example_2 (m n : \mathbb{N}) : 
 2 * m + n = m + n + m := 
 have h_1 : 2 * m + n = (m + m) + n := by rw two_mul, 
 have h_2 : (m + m) + n = m + n + m := by ac_refl, 
 show _, from eq.trans h_1 h_2
```

Notice that with haves, we also need to explicitly invoke eq.trans and to give names to the two intermediate steps.

3.3 Induction by Pattern Matching

In Section 2.5, we saw how to use the induction tactic to carry out proofs by induction. An alternative, more structured style relies on pattern matching, as in the example below:

```
lemma add_zero : \forall n : \mathbb{N}, add 0 n = n | 0 := by refl | (nat.succ m) := by simp [add, add_zero m]
```

The patterns on the left, 0 and (nat.succ m), correspond to the two constructors of the type of the \forall -quantified variable n: \mathbb{N} . On the right-hand side of each := symbol is a proof for the corresponding case: the base case or the induction step.

Inside the induction step's proof, the induction hypothesis is available under the same name as the lemma we are proving. We explicitly pass m as argument to the hypothesis: This is useful as documentation, and it also prevents Lean from accidentally invoking the induction hypothesis in a circular or otherwise ill-founded fashion—for example, on $mat.succ\ m$. Lean's termination checker would detect this and raise an error.

Simultaneous pattern matching is also possible, as in the examples below:

```
lemma add_succ :
  \forall m \ n : \mathbb{N}, add (nat.succ m) n = nat.succ (add m n)
                 := by refl
| m (nat.succ n) := by simp [add, add_succ m n]
lemma add_comm :
  \forall m \ n : \mathbb{N}, \ add \ m \ n = add \ n \ m
        := by simp [add, add_zero]
| m (nat.succ n) := by simp [add, add_succ, add_comm m n]
lemma add_comm<sub>2</sub> :
  \forall m \ n : \mathbb{N}, \ add \ m \ n = add \ n \ m
          := by simp [add, add_zero]
| m (nat.succ n) :=
  have ih : \_ := add_comm<sub>2</sub> m n,
  by simp [add, add_succ, ih]
lemma add_assoc :
  \forall 1 \text{ m n} : \mathbb{N}, \text{ add (add 1 m) n = add 1 (add m n)}
1 1 m 0
              := by refl
| 1 m (nat.succ n) := by simp [add, add_assoc 1 m n]
```

Notice the structured proof command have in the proof of add_comm₂.

The general format of lemmas with pattern matching is as follows:

```
lemma name (vars_1 : type_1) \dots (vars_m : type_m) : statement
| patterns_1 := proof_1
| patterns_n := proof_n
```

A few hints on how to carry out proofs by induction follow:

- It is usually beneficial to perform induction following the structure of the definition of one of the functions appearing in the goal.
- If the base case of an induction is difficult, this is often a sign that the wrong variable was chosen or that some lemmas are missing.

3.4 Dependent Types

Dependent types are the defining feature of the *dependent type theory* family of logics. Consider a function pick that takes a natural number x (a value from $\mathbb{N} = \{0, 1, 2, \ldots\}$) and that returns a natural number between 0 and x. Conceptually, pick has a dependent type, namely, $(x : \mathbb{N}) \to \{y : \mathbb{N} \ // \ y \le x\}$. Informally, we read $\{y : \mathbb{N} \ // \ y \le x\}$ as "the type consisting of all y from \mathbb{N} such that $y \le x$." We can think of this type as a \mathbb{N} -indexed family, where each member's type may depend on the index: pick $x : \{y : \mathbb{N} \ // \ y \le x\}$. The name x is immaterial; we could have used any other name except y.

Unless otherwise specified, a *dependent type* means a type depending on a term, as in the above example, with $x:\mathbb{N}$ as the term and $\{y:\mathbb{N} \ //\ y\leq x\}$ as the type that depends on it. This is what we mean when we claim that simple type theory does not support dependent types. But a type may also depend on another type—for example, the type constructor list, its η -expanded variant $\lambda\alpha: \mathrm{Type}$, $\mathrm{list}\ \alpha$, or the polymorphic type λ $\alpha: \mathrm{Type}$, $\alpha \to \alpha$ of functions with the same domain and codomain. A term may depend on a type—for example, the polymorphic identity function $\lambda\alpha: \mathrm{Type}$, $\lambda x:\alpha$, x. And of course, a term may also depend on a term—for example, $\lambda x:\mathbb{N}$, x.

In summary, there are four cases for λx , t in dependent type theory:

Body (t)	Argument (x)	Description
A type A term	depending on depending on depending on depending on	a term a type	Simply typed λ -abstraction Dependent type (stricto senso) Polymorphic term Type constructor

The last three rows correspond to the three axes of Henk Barendregt's λ -cube.

The APP and LAM rules presented in Section 1.1 must be generalized to work with dependent types:

$$\begin{split} \frac{\Gamma \ \vdash \ \mathbf{t} : (\mathbf{x} : \sigma) \to \tau[\mathbf{x}] \qquad \Gamma \vdash \mathbf{u} : \sigma}{\Gamma \vdash \mathbf{t} \ \mathbf{u} : \tau[\mathbf{u}]} \mathsf{APP'} \\ \frac{\Gamma, \ \mathbf{x} : \sigma \vdash \mathbf{t} : \tau[\mathbf{x}]}{\Gamma \vdash (\lambda \mathbf{x} : \sigma, \mathbf{t}) : (\mathbf{x} : \sigma) \to \tau[\mathbf{x}]} \mathsf{LAM'} \end{split}$$

The simply typed case arises when x does not occur in $\tau[x]$. Then, we can simply write $\sigma \to \text{and not } (x:\sigma) \to .$ The notation $\sigma \to \tau$ is equivalent to $(\underline{\ }:\sigma) \to \tau$. It is easy to see that APP' and LAM' are identical to APP and LAM when x does not occur in $\tau[x]$.

The example below demonstrates APP':

$$\frac{ \ \ \, \vdash \ \text{pick}: (\mathtt{x}:\mathbb{N}) \to \{\mathtt{y}:\mathbb{N}\://\:\mathtt{y} \le \mathtt{x}\} \qquad \vdash \ 5:\mathbb{N}}{ \ \ \, \vdash \ \, \text{pick}\:5: \{\mathtt{y}:\mathbb{N}\://\:\mathtt{y} \le 5\}} \mathsf{App'}$$

The next example demonstrates LAM':

$$\frac{\alpha \,:\, \mathsf{Type,}\,\, \mathsf{x} \,:\, \alpha \,\vdash\, \mathsf{x} \,:\, \alpha}{\alpha \,:\, \mathsf{Type}\, \vdash\, (\lambda \mathsf{x} \,:\, \alpha,\,\, \mathsf{x}) \,:\, \alpha \to \alpha} \mathsf{LAM'or}\, \mathsf{LAM'}}{(\lambda \alpha \,:\, \mathsf{Type,}\,\, \lambda \mathsf{x} \,:\, \alpha,\,\, \mathsf{x}) \,:\, (\alpha \,:\, \mathsf{Type}) \to \alpha \to \alpha} \mathsf{LAM'}}$$

We conclude with an important remark on notations. Regrettably, the intuitive syntax $(x:\sigma) \to \tau$ is not available in Lean. Instead, we must write $\Pi x:\sigma$, τ to specify a dependent type. The arrow notation $\sigma \to \tau$ (without the x) is syntactic sugar for $\Pi_-:\sigma$, τ . The Π binder emphasizes that x is a bound variable within Π 's scope.

¹https://en.wikipedia.org/wiki/Lambda_cube

As an alias for Π , we can also write \forall . The two symbols are fully interchangeable. In Section 1.2, we saw the commands

```
#check list.nil
#check list.cons
```

and their output

```
list.nil : \Pi(\alpha : \mathsf{Type}), list \alpha list.cons : \mathsf{?M\_1} \to \mathsf{list} \; \mathsf{?M\_1} \to \mathsf{list} \; \mathsf{?M\_1}
```

We can now make sense of the Π -type. Namely, list.nil is a function that takes a type α as argument and that returns a value of type list α .

3.5 The Curry-Howard Correspondence

You likely have noticed that we use the same symbol \rightarrow both as the implication symbol and as the type constructor of functions. Similarly, \forall is used both as a quantifier and to specify dependent types. When we see $a \rightarrow b$, without further context, we cannot tell whether we have the type of a function with domain a and codomain b or the proposition "a implies b." Similarly, $\forall x$, t[x] can denote a proposition or a dependent type.

It turns out that not only the two pairs of concepts look the same, they *are* the same. This is called the *Curry–Howard correspondence*. It is also called the *PAT principle*, where PAT is a double mnemonic:

```
PAT = propositions as types PAT = proofs as terms
```

Nicolaas Govert de Bruijn, Haskell Curry, William Alvin Howard, and possibly others independently noticed that for some logics, propositions are isomorphic to types, and proofs are isomorphic to terms. Hence, in dependent type theory, we *identify* proofs with terms as well as propositions with types, a considerable economy of concepts.

Let us go through the dramatis personae one by one: types, propositions, terms, and proofs. We use the metavariables σ , τ for types; φ , ψ for propositions; t, u, x for terms; and h for proofs.

For types, we have the following:

- $\sigma \to \tau$ is the type of (total) functions from σ to τ .
- $\Pi x : \sigma$, $\tau[x]$ is the dependent type from $x : \sigma$ to $\tau[x]$.

For propositions, we have the following:

- $\varphi \to \psi$ can be read as " φ implies ψ ", or as the type of functions mapping proofs of φ to proofs of ψ .
- $\forall x : \sigma$, $\psi[x]$ can be read as "for all x, $\psi[x]$ ", or as the type of function mapping values x of type σ to proofs of $\psi[x]$.
- $\forall h : \varphi, \psi[h]$ is the type of function mapping proofs h of φ to proofs of $\psi[h]$.

Notice the similarities between types and propositions. And recall that \forall is an alias for Π .

For terms, we have the following:

- · Declared constants are terms.
- t u is the application of function t to value u.
- λx , t[x] is a function mapping x to t[x].

For proofs (i.e., proof terms), we have the following:

- A lemma name is a proof.
- 1 a, which instantiates the leading parameter or quantifier of lemma 1 with value a, is a proof.
- 11', which discharges the leading assumption of 1 with lemma 1', is a proof. This operation is called *modus ponens*.

- Given $h[n] : \varphi[n]$, we have that $\lambda n : \sigma$, h[n] is a proof of $\forall n : \sigma$, $\varphi[n]$.
- Given $h[h']: \varphi[h']$, we have that $\lambda h': \psi$, h[h'] is a proof of $\forall h': \psi$, $\varphi[h']$.

For the last two cases, in a structured proof we would write assume, and in a tactical proof we would invoke intros.

By the Curry-Howard correspondence, a proof by *induction* is the same as a *recursively* specified proof term. When we invoke the induction hypothesis, we are really just calling a recursive function (the lemma) recursively. This explains why, in a proof by pattern matching (Section 3.3), the induction hypothesis has the same name as the lemma we prove.

In the same way, well-founded induction (also called "complete induction", "course-of-values induction", or "strong induction") is the same as well-founded recursion. Structural induction corresponds to structural recursion. Lean's termination checker is used to prove well-foundedness of the proof by induction.

Finally, you may have noticed that the commands def and lemma have almost the same syntax. They are, in fact, almost the same, with the exception that lemma considers the proof term (or whatever defined term) opaque, whereas def keeps it transparent. Since proofs are irrelevant to Lean's logic, there is no need to store them or unfold them later. A similar difference is visible between let and have.

3.6 Summary of New Syntax

Proof Commands

assume states variables or assumptions calc combines proofs by transitivity states an intermediate lemma let ... in introduces a local definition show states the goal's conclusion

Tactics

have states an intermediate lemma let introduces a local definition

Part II Functional-Logic Programming

Chapter 4

Functional Programming

We take a closer look at the basics of typed functional programming: inductive types, pattern matching, recursive functions, and proofs by induction. The concepts covered here are described in more detail in Chapters 7 to 9 of *Theorem Proving in Lean* [1].

4.1 Inductive Types

Inductive types are modeled after the datatypes of typed functional programming languages (e.g., Haskell, ML, OCaml). As we saw in Chapter 1.2, an inductive type is a type whose members are all the values that can be built by a finite number of applications of its constructors, and only those. Mottos:

- No junk: The type contains no values beyond those expressible using the constructors.
- No confusion: Values built using a different combination of constructors are different

For natural numbers, "no junk" means that there exist no special values such as -1 or ε that cannot be expressed using a finite combination of zero and succ, and "no confusion" ensures that zero \neq succ n for all n. In addition, inductive types are always finite. The infinite term succ (succ (succ ...)) is not a value, nor does there exist a number n such that n = succ n.

Already in Chapter 1, we saw some basic inductive types: the natural numbers, the finite lists, and a type of arithmetic expressions. In this chapter, we revisit the lists and study some new types.

4.2 Example: Lists

We first define the reverse of a list, exploiting the predefined concatenation operator ++:

(In fact, reverse exists in Lean's standard library, as list.reverse, but our definition is more suitable for reasoning.) A nice property to prove is that reverse is its own inverse: reverse (reverse xs) = xs for all lists xs. However, if we try to prove it by induction, we quickly run into an obstacle. For the induction step, the goal to prove is

```
reverse (reverse xs ++ [x]) = x :: xs
```

and the induction hypothesis is

```
\forall xs : list \alpha, reverse (reverse xs) = xs
```

We need a way to "distribute" reverse over ++, to eliminate the occurrence of ++ [x] and obtain a term that matches the induction hypothesis's left-hand side. The trick is to prove and use the following lemma:

```
lemma reverse_append \{\alpha: \text{Type}\}: \forall xs \ ys: \text{list } \alpha, \text{ reverse } (xs ++ \ ys) = \text{reverse } ys ++ \text{ reverse } xs \\ [] ys:= by simp [reverse] \\ [(x:: xs) ys:= by simp [reverse, reverse_append xs]
```

It is important to state the lemma generally, for all xs and ys. If we put [y] instead of ys, we would run into difficulties when proving the lemma's induction step. In general, finding the right inductions and lemmas can be a difficult task, which requires thought and creativity.

With the reverse_append lemma in place, we can prove our initial goal by a straightforward induction:

```
lemma reverse_reverse \{\alpha: \text{Type}\}: \forall xs: \text{list } \alpha, \text{ reverse (reverse xs)} = xs | [] := by refl | (x :: xs) := by simp [reverse, reverse_append, reverse_reverse xs]
```

The next operation we define is a *map function*: a function that applies its argument f—which is itself a function—to all elements stored in a container.

Notice that because f does not change in the recursive call, we put it as a parameter of the entire definition. This means we must write $\mathtt{map}\ \mathtt{xs}$ and $\mathtt{not}\ \mathtt{map}\ \mathtt{f}\ \mathtt{xs}$ in the recursive call. However, when using the function later, we would need to pass an argument for f . The alternative, which is the only option for arguments that change in recursive calls, would be as follows:

Notice the use of _ as a nameless placeholder in the first case. We could also have given the variable a name (e.g., f).

A basic property of map functions is that they have no effect if if their argument is the identity function (λx , x):

```
lemma map_ident \{\alpha: \text{Type}\}: \forall xs: \text{list } \alpha, \text{ map } (\lambda x, x) \text{ xs = xs} \} := by refl | (x :: xs) := by simp [map, map_ident xs]
```

Another basic property is that successive maps can be compressed into a single map, whose argument is the composition of the functions involved:

```
lemma map_comp {\alpha \beta \gamma : Type} (f : \alpha \to \beta) (g : \beta \to \gamma) : \forall xs : list \alpha, map (\lambda x, g (f x)) xs = map g (map f xs) | [] := by refl | (x :: xs) := by simp [map, map_comp xs]
```

When introducing new operations, it is useful to show how these behave when used in combination with other operations. Here are two examples:

```
lemma map_append \{\alpha \ \beta : \text{Type}\}\ (\text{f}: \alpha \to \beta): \forall \text{xs ys}: \text{list } \alpha, \text{ map f (xs ++ ys)} = \text{map f xs ++ map f ys}
```

```
l []
         ys := by refl
| (x :: xs) ys := by simp [map, map_append xs]
lemma map_reverse \{\alpha \ \beta : \text{Type}\}\ (\text{f}: \alpha \rightarrow \beta) :
  \forall xs : list \alpha, map f (reverse xs) = reverse (map f xs)
l []
            := by refl
| (x :: xs) :=
  by simp [map, reverse, map_append, map_reverse xs]
```

The length of a list is defined by recursion:

```
def length \{\alpha : \mathsf{Type}\} : \mathsf{list} \ \alpha \to \mathbb{N}
| [] := 0
\mid (x :: xs) := length xs + 1
```

The next function definition demonstrates the syntax of pattern matching within expressions:

```
def bcount \{\alpha : \mathsf{Type}\}\ (\mathsf{p} : \alpha \to \mathsf{bool}) : \mathsf{list}\ \alpha \to \mathbb{N}
                 := 0
| (x :: xs) :=
  match p x with
   | tt := bcount xs + 1
   | ff := bcount xs
   end
```

The bcount function counts the number of elements in a list that satisfy the given predicate p. The predicate's codomain is bool, not prop. As a general rule, we will use type bool, of Booleans, when specifying programs and use the type prop, of propositions, when stating properties of programs. The two values of type bool are called tt and ff. The connectives are called bor (infix: ||), band (infix: &&), bnot, and bxor.

The match construct has the following general syntax:

```
match term_1, \ldots, term_m with
\mid patterns_{11}, ..., pattern_{1m} := result_1
| \ \textit{patterns}_{\textit{n1}} \text{,..., pattern}_{\textit{nm}} := \textit{result}_{\textit{n}}
```

Notice the differences with the pattern matching syntax in def and lemma: The patterns are separated by commas and not by spaces. This reduces the number of required parentheses. The patterns may contain variables, constructors, and $_{\perp}$ placeholders. The $result_{d}$ expressions may refer to the variables introduced in the corresponding patterns.

match expressions can also be used in structured proofs to eliminate connectives and quantifiers, which are after all defined inductively. For example, the proof

```
example (p q : \mathbb{N} \to \text{Prop}) (h : \exists x, p x \land q x) :
         \exists x, q x \land p x :=
      match h with
      |\langle w, hw \rangle :=
         exists.intro w (and.intro (and.right hw) (and.left hw))
      end
is equivalent to
      example (p q : \mathbb{N} \to \text{Prop}) (h : \exists x, p x \land q x) :
         \exists x, q x \land p x :=
      exists.elim h
         (assume w hw,
         exists.intro w (and.intro (and.right hw) (and.left hw)))
```

The next list operation extracts the first element of a list:

```
def head \{\alpha: {\rm Type}\}\ ({\rm xs}: {\rm list}\ \alpha): {\rm option}\ \alpha:={\rm match}\ {\rm xs}\ {\rm with} | [] := none | x :: _ := some x end
```

Because an empty list contains no elements, there is no meaningful value we can return in that case. For this reason, we use an option. The type option α has two constructors: none: option α and some: $\alpha \to \text{option } \alpha$. We use none when we have no meaningful value to return and some otherwise.

Given two lists of the same length $[x_1, ..., x_n]$ and $[y_1, ..., y_n]$, the zip operation constructs a list of pairs $[(x_1, y_1), ..., (x_n, y_n)]$:

The function is also defined if one list is shorter than the other. For example, zip[a, b, c][x, y] = [(a, x), (b, y)].

Pairs are written using traditional notations. The type constructor \times , which occurs in the result type list $(\alpha \times \beta)$, has an infix syntax. The pair constructor is written using the syntax (t, u), where t and u are the arguments. Given a pair p, the first and second components can be extracted by writing p.fst and p.snd or p.1 and p.2.

The p.1 and p.2 syntaxes are used below to state a distributivity law about map and zip:

```
lemma map_zip {\alpha \alpha' \beta \beta' : Type} (f : \alpha \to \alpha') (g : \beta \to \beta') : \forall xs \ ys, \ map \ (\lambda p : \alpha \times \beta, \ (f \ p.1, \ g \ p.2)) (zip xs \ ys) = zip (map f xs) (map g ys) | (x :: xs) (y :: ys) := begin simp [zip, map], exact (map_zip__) end | [] _ = := by refl | (_ :: _) [] := by refl
```

We can also say something interesting about the length of the result of zip—namely, it is the minimum of the lengths of the two input lists:

```
lemma length_zip {\alpha \beta : Type} : \forall (xs : list \alpha) (ys : list \beta), length (zip xs ys) = min (length xs) (length ys) | (x :: xs) (y :: ys) := by simp [zip, length, length_zip xs ys, min_add_add] | [] _ := by refl | (_ :: _) [] := by refl
```

The proof relies on a lemma about the min function that we need to prove ourselves:

```
lemma min_add_add (1 m n : \mathbb{N}) : min (m + 1) (n + 1) = min m n + 1 := have 1 + m \leq 1 + n \leftrightarrow m \leq n := begin rw add_comm 1, rw add_comm 1, apply nat.add_le_add_iff_le_right
```

```
end, by by_cases m \le n; simp [min, *]
```

The min function is defined such that min a b = (if a \leq b then a else b). To reason about it, we often need to perform a case analysis on the condition a \leq b. This is achieved using by_cases. Given a condition c, it creates two subgoals: one in which c appears as a hypothesis and one where \neg c appears.

In the next example, we show how to exploit injectivity of constructors. The cases tactic can be used to apply injectivity on equations whose both sides have the same constructor applied. In the proof below, the equation on which injectivity is applied is x :: xs = y :: ys. As a result of the case distinction, y is replaced by x and ys is replaced by xs throughout the goal:

```
lemma injection_example \{\alpha: \text{Type}\}\ (\text{x y}: \alpha)\ (\text{xs ys}: \text{list }\alpha)   (h: list.cons x xs = list.cons y ys): x = y \land xs = ys := begin cases h, apply and.intro, { refl }, { refl } end
```

The cases tactic is also useful when the constructors are different, to detect the impossible case:

```
lemma distinctness_example \{\alpha: {\rm Type}\}\ ({\rm x}\ {\rm y}:\alpha)\ ({\rm xs}\ {\rm ys}: {\rm list}\ \alpha) (h : [] = y :: ys) : false := by cases h
```

4.3 Example: Binary Trees

Lists are not the only interesting data type. Inductive types with constructors taking several recursive arguments define tree-like objects. *Binary trees* have nodes with at most two children. A possible definition of binary trees follows:

```
inductive btree ($\alpha$ : Type) : Type | empty {} : btree | node : $\alpha$ \to btree \to btree \to btree
```

(The $\{\}$ annotation is often used with nullary constructors of polymorphic types. Here, it indicates that the type α should be implicitly derived from the context surrounding occurrences of empty.)

By default, the constructor names are prefixed by the name of the type—i.e., tree. empty and tree.node. To lighten notations, we can enter the following command, which exports empty and node from the tree namespace, allowing us to write empty and node with no qualification:

```
export btree (empty node)
```

A peculiarity of binary trees is that structural induction has two induction hypotheses: one for the left subtree of an inner node and one for the right subtree. To prove a goal p t about a tree t by structural induction on t, we must prove (1) p empty and (2) p (node a l r) assuming p l and p r.

The tree counterpart to list reversal is the mirror operation:

```
\begin{array}{lll} \operatorname{def\ mirror\ }\{\alpha\ :\ \operatorname{Type}\}\ :\ \operatorname{btree}\ \alpha\ \to\ \operatorname{btree}\ \alpha\\ \mid\ \operatorname{empty} &:=\ \operatorname{empty}\\ \mid\ (\operatorname{node\ a\ l\ r})\ :=\ \operatorname{node\ a\ (\operatorname{mirror\ r})\ (\operatorname{mirror\ l})} \end{array}
```

Mirroring can be defined directly, without appealing to some append operation. As a result, reasoning about mirror is simpler than reasoning about reverse, as the example below demonstrates:

4.4 Example: Colors

Lean provides a convenient syntax for defining records, or *structures*. These are essentially nonrecursive, single-constructor inductive types, but with some syntactic sugar. The definition below introduces an rgb structure with three fields of type $\mathbb N$ called red, green, and blue:

```
structure rgb := (red green blue : \mathbb{N})
```

We can define a new structure as the extension of an existing structure. The definition below introduces a fourth component, called alpha, to rgb and calls the result rgba:

```
structure rgba extends rgb := (alpha : \mathbb{N})
```

Values can be specified in a variety of syntaxes:

```
def red : rgb := {red := 0xff, green := 0x00, blue := 0x00} def semitransparent_red : rgba := {alpha := 0x7f, ..red} def green : rgb := \langle 0x00, 0xff, 0x00 \rangle
```

The definition of semitransparent_red copies all the values from red (indicated by the ..red syntax) except for the alpha field, which it overrides. The definition of green uses Lean's anonymous constructor syntax $\langle \dots \rangle$. This syntax is available for all single-constructor types, at the condition that the type of the expression can be inferred from the context. Here, it can be inferred because we explicitly specify green's type and the two sides of an equality must have the same type.

Next, we define an operation called shuffle:

```
def shuffle (c : rgb) : rgb :=
{red := c.green, green := c.blue, blue := c.red}
```

The definition uses the automatically generated projection functions color.red, color .green, and color.blue. Instead of color.red c, Lean lets us simply write c.red, and similarly for the other fields.

Applying shuffle three times in a row is the same as not applying it at all:

```
lemma shuffle_shuffle (c : rgb) :
    shuffle (shuffle (shuffle c)) = c :=
by cases c; refl
```

The proof teaches us a new trick. The cases tactic is a relative of induction. It performs a case distinction on its arguments, but it does not generate inductive hypotheses.

4.5 Overlapping Patterns

It is possible to write equations with overlapping patterns. For example:

```
\begin{array}{lll} \operatorname{def} \ \mathbf{f} \ \{\alpha \ : \ \mathsf{Type}\} \ : \ \mathsf{list} \ \alpha \ \to \cdots \\ & | \ [] \ := \ \ldots \\ & | \ \mathsf{xs} \ := \ \ldots \ \mathsf{xs} \ \ldots \end{array}
```

The first matching equation is taken. Internally, this is the same as

```
\begin{array}{lll} \operatorname{def} \ f \ \{\alpha \ : \ \operatorname{Type}\} \ : \ \operatorname{list} \ \alpha \to \cdots \\ & | \ [] & := \ldots \\ & | \ (\mathtt{x} \ :: \ \mathtt{xs}) \ := \ldots \ (\mathtt{x} \ :: \ \mathtt{xs}) \ \ldots \end{array}
```

We generally recommend the latter style, because it is less surprising.

4.6 New Tactics

by_cases

```
by_cases proposition
```

The by_cases tactic performs a case analysis on a proposition. It is useful to reason about the condition in an if condition.

cases

```
cases variable
```

The cases tactic performs a case distinction on the specified variable. This gives rise to as many subgoals as there are constructors in the definition of the variable's type. Its behavior is similar to induction's except that it does not produce induction hypotheses.

4.7 Summary of New Syntax

Declaration

structure introduces a record type and its projection functions

Command

export makes names available outside their namespaces

Term Language

```
if ... then ... else case distinction on a proposition
...
match ... with ... performs nonrecursive pattern matching
end
```

Tactics

by_cases performs a case analysis on an if condition cases performs a case analysis on a variable of an inductive type

Chapter 5

Inductive Predicates

Inductive predicates are reminiscent of the Horn clauses of Prolog-style logic programming languages. But Lean offers a much stronger logic than Prolog, so we need to do some work—interactive proving—to establish theorems instead of just running the Prolog interpreter. A possible view of Lean:

Lean = typed functional programming + logic programming + more logic

5.1 Introductory Example

Inductive predicates, or (more precisely) inductively defined propositions, are familiar from mathematics. Consider the example:

The set *E* of even natural numbers is defined as the smallest set closed under the following rules: (1) $0 \in E$; and (2) for every $n \in \mathbb{N}$, if $n \in E$, then $n + 2 \in E$.

In Lean, we write

```
\begin{array}{lll} \mbox{inductive even} & : & \mathbb{N} & \to \mbox{Prop} \\ \mbox{| zero} & : & \mbox{even 0} \\ \mbox{| add_two} & : & \forall \mbox{n, even n} & \to \mbox{even (n + 2)} \end{array}
```

This should look familiar: By the Curry-Howard correspondence, what we have done is introduce a new unary type constructor, even. Inductive types and inductive predicates are implemented as the same mechanism in Lean.

The even predicate has two constructors: zero and add_two. These can be used to build proof terms. The even predicate can be seen as a tree type, the trees being the corresponding proof terms (or proof trees). Thanks to the "no junk" guarantee of inductive definitions, zero and add_two are the only two ways to construct even. Thus, there is no danger of accidentally proving that 1 is even.

5.2 Logical Symbols

The truth values false and true, the connectives \land and \lor , the \exists quantifier, and the equality predicate = are all defined as inductive propositions or predicates:

```
inductive false : Prop
inductive true : Prop
| intro : true
inductive and (a b : Prop) : Prop
```

```
| intro : a \rightarrow b \rightarrow and inductive or (a b : Prop) : Prop | intro_left : a \rightarrow or | intro_right : b \rightarrow or inductive Exists \{\alpha : \text{Type}\}\ (\text{p} : \alpha \rightarrow \text{Prop}) : \text{Prop} | \text{intro} : \forall \text{a} : \alpha, \text{ p} \text{ a} \rightarrow \text{Exists} inductive eq \{\alpha : \text{Type}\} : \alpha \rightarrow \alpha \rightarrow \text{Prop} | \text{refl} (a : \alpha) : \text{eq a a}
```

The notations $\exists x : \alpha$, p and x = y are syntactic sugar for Exists $(\lambda x : \alpha, p)$ and eq x y, respectively. In contrast, \forall (= Π) and \rightarrow are built directly into the logic and are not defined as inductive predicates. Notice that there is no constructor for false. There are no proofs of false, just like there is no proof that 1 is even.

5.3 Introduction, Elimination, Induction, and Inversion

We saw in Section 2.3 that the logical connectives and the \exists quantifier are equipped with introduction and elimination rules. The same is true for arbitrary inductive predicates p.

p's constructors are introduction rules. They typically have the form $\forall\ldots$, $\ldots\to p\ldots$. They can be used to prove goals of the form $p\ldots$.

Elimination works the other way around. It extracts information from a lemma or hypothesis of the form $p ext{...}$ Elimination takes various forms: pattern matching (at the top-level of a definition or lemma, or with match), the cases and induction tactics, or custom elimination rules (e.g., and.elim_left).

Just as we can perform induction on a term, we can perform induction on a proof term using the induction tactic. This is sometimes called *rule induction*, because the induction is on the introduction rules (i.e., the constructors of the proof term).

Often it is convenient to rewrite concrete terms of the form p (c ...), where c is typically a constructor. We can state and prove an *inversion rule* to support such eliminative reasoning. The general format of an inversion rule (assuming a binary constructor) is as follows:

```
\forall x y, p (c x y) \rightarrow (\exists ..., ... \land ...) \lor ... \lor (\exists ..., ... \land ...)
```

It can be useful to combine introduction and elimination in a single lemma, which can be used for rewriting both the hypotheses and the conclusions of goals:

```
\forall x \ y, \ p \ (c \ x \ y) \leftrightarrow (\exists ..., ... \land ...) \lor ... \lor (\exists ..., ... \land ...)
\forall n : \mathbb{N}, \text{ even } n \leftrightarrow n = 0 \lor \exists m : \mathbb{N}, \ n = m + 2 \land \text{ even } m
```

5.4 Example: Full Binary Trees

For example:

We now review in turn four examples of inductive predicates. Our first example is based on the type of binary trees introduced in Section 4.3:

A binary tree is *full* if all its nodes have either zero or two children. This can be encoded as an inductive predicate as follows:

```
inductive is_full {$\alpha$ : Type} : btree $\alpha$ $\rightarrow$ Prop | empty : is_full empty | node (a : $\alpha$) (l r : btree $\alpha$) (hl : is_full l) (hr : is_full r) (empty_iff : l = empty $\leftrightarrow$ r = empty) : is_full (node a l r)
```

The first case states that the empty tree is a full tree. The second case states that a tree is a full tree if has two child trees that are themselves full and that either both are empty or both are nonempty.

Syntactically, we can put the arguments on the left of the colon, as parameters, or on the right. Here is an equivalent definition of the latter kind:

```
inductive is_full_2 {$\alpha$ : Type} : btree $\alpha$ $\rightarrow$ Prop | empty : is_full_2 empty | node : $\forall (a : $\alpha$) (1 r : btree $\alpha$), is_full_2 1 $\rightarrow$ is_full_2 r $\rightarrow$ (1 = empty $\leftrightarrow$ r = empty) $\rightarrow$ is_full_2 (node a 1 r)
```

The tree that consists of a node with empty tree as children is a full tree. Here is a simple proof, using the introduction rule is_full.node:

```
lemma is_full_singleton \{\alpha: \text{Type}\}\ (a:\alpha): \text{is}\_\text{full (node a empty empty)}:= \text{begin} apply is_full.node, repeat \{\text{ apply is}\_\text{full.empty }\}, reflend
```

The proof makes use of the repeat combinator, which applies the tactic specified within curly braces repeatedly, as long as it succeeds. Here it is applied twice, to eliminate the two identical goals is_full empty.

A somewhat more interesting property of full trees is that fullness is preserved by the mirror operation. Our first proof is by structural induction on t:

The key to the proof is the cases tactic invocation on the hypothesis full_t : is_full (node alr). The tactic notices, via a simple syntactic check, that the first introduction rule (is_full.empty) cannot have been used to derive is_full (node alr), so it only produces one case, corresponding to the second introduction rule (is_full.node). As usual, the tactical proof makes more sense if you inspect it in Visual Studio Code, moving the cursor around.

Another way to carry out the proof is to perform pattern matching on $is_{full} t$ at the top level:

```
lemma is_full_mirror_2 {\alpha : Type} : \forall t : btree \alpha, is_full t \rightarrow is_full (mirror t) | _ is_full.empty := begin  
   rw mirror,  
   exact is_full.empty end | _ (is_full.node a l r hl hr empty_iff) :=
```

```
begin
  rw mirror,
  apply is_full.node,
  repeat { apply is_full_mirror2, assumption },
  simp [mirror_eq_empty_iff, *]
```

A third approach relies on an inversion rule. The rule is proved by a combination of introduction and elimination steps:

```
lemma is_full_node_iff \{\alpha: \text{Type}\}\ (a:\alpha)\ (l\ r: \text{btree}\ \alpha): is_full\ (\text{node}\ a\ l\ r) \leftrightarrow is_full\ l\ \land\ is_full\ r\ \land\ (l=\text{empty}\ \leftrightarrow\ r=\text{empty}):= begin apply iff.intro, { intro\ h, cases\ h, cc\ }, { intro\ h, apply is_full.node, repeat\ \{\ cc\ \}\ } end
```

The proof relies on the cc tactic, which implements an algorithm called congruence closure. The tactic studies the equalities present in the problem and attempts to derive a proof from them. It is described in slightly more detail below (Section 5.9).

Equipped with the inversion rule, we can perform a simpler proof that relies on the simp tactic:

5.5 Example: Sorted Lists

Our next example is a predicate that checks whether a list of natural numbers is sorted in increasing order:

```
inductive sorted : list \mathbb{N} \to \operatorname{Prop} | nil : sorted [] | single \{x : \mathbb{N}\} : sorted [x] | two_or_more \{x \ y : \mathbb{N}\} \{xs : \text{list } \mathbb{N}\} (xy : x \le y) (yxs : \text{sorted } (y :: xs)) : sorted (x :: y :: xs)
```

It is always a good idea to test our definitions by trying it on small examples. Is the list [3, 5] sorted? It would appear so:

```
example :
  sorted [3, 5] :=
begin
  apply sorted.two_or_more,
  { linarith },
  { exact sorted.single }
```

end

The example needs two of the introduction rules. Notice the invocation of linarith. This tactic can be invoked to solve linear arithmetic goals. It is described below (Section 5.9). A more compact proof follows, using proof terms for everything except the arithmetic subgoal:

```
example :
    sorted [3, 5] :=
    sorted.two_or_more (by linarith) sorted.single
The same idea can be used to prove that [7, 9, 9, 11] is sorted:
```

```
example :
   sorted [7, 9, 9, 11] :=
sorted.two_or_more (by linarith)
   (sorted.two_or_more (by linarith)
      (sorted.two_or_more (by linarith)
      sorted.single))
```

Conversely, we can show that some lists are not sorted, but this time we need to use elimination, in the form of pattern matching:

```
example : \neg sorted [17, 13] := assume h : sorted [17, 13], have 17 \leq 13 := match h with \mid sorted.two_or_more xy yxs := xy end, have \neg (17 \leq 13) := by linarith, show false, from by cc
```

To prove a negation, we assume the unnegated formula and prove false. From the assumption that [17, 13] is sorted, we derive that $17 \le 13$, which together with the obvious numeric truth that \neg (17 \le 13) leads to a contradiction. The contradiction is easily detected by cc.

5.6 Example: Well-formed and Ground First-Order Terms

Our third example is based on an inductive type of first-order terms:

```
inductive term (\alpha \beta : Type) : Type | var {} : \beta \to \text{term} | fn : \alpha \to \text{list term} \to \text{term} export term (var fn)
```

A first-order term is either a variable x or a function symbol f applied to a list of arguments: $f(t_1, \ldots, t_n)$, where the metavariables t_j stand for the arguments, which are themselves terms. The parameters α and β are the types of function symbols and variables, respectively.

Not all terms are legal. For example, the term g(f(a),f(a,b)) is considered ill formed, because the function f is invoked with different number of arguments (once with one argument and once with two). Along with α and β , we also consider the arity, represented by a function $\mathtt{arity}:\alpha\to \mathbb{N}$ indicating how many arguments each function symbol takes. The $\mathtt{well_formed}$ predicate then checks whether the given term only contains function symbol applications with the specified number of arguments:

```
inductive well_formed \{\alpha \ \beta : \text{Type}\}\ (\text{arity} : \alpha \to \mathbb{N}) : \text{term } \alpha \ \beta \to \text{Prop} | var (x : \beta) : well_formed (var x) | fn (f : \alpha) (ts : list (term \alpha \beta)) (hargs : \formstyle t \in \text{ts}, well_formed t) (hlen : list.length ts = arity f) : well formed (fn f ts)
```

The fn case checks that the arguments ts are recursively well formed. Additionally, it requires that the length of ts equals the specified arity for the function symbol f in question.

Another interesting property of first-order terms is whether they contain variables. This can be checked easily using an inductive predicate:

```
inductive variable_free {\alpha \beta : Type} : term \alpha \beta \rightarrow Prop | fn (f : \alpha) (ts : list (term \alpha \beta)) (hargs : \forallt \in ts, variable_free t) : variable_free (fn f ts)
```

This time, there is no var case, for obvious reasons.

5.7 Example: Reflexive Transitive Closure

Our last example is the reflexive transitive closure of a relation, modeled as a binary predicate rtc.

```
inductive rtc \{\alpha: \mathsf{Type}\}\ (\mathsf{r}: \alpha \to \alpha \to \mathsf{Prop}): \alpha \to \alpha \to \mathsf{Prop} \mid \mathsf{base}\ (\mathsf{a}\;\mathsf{b}: \alpha): \mathsf{r}\;\mathsf{a}\;\mathsf{b} \to \mathsf{rtc}\;\mathsf{a}\;\mathsf{b} \mid \mathsf{refl}\ (\mathsf{a}: \alpha): \mathsf{rtc}\;\mathsf{a}\;\mathsf{a} \mid \mathsf{trans}\ (\mathsf{a}\;\mathsf{b}\;\mathsf{c}: \alpha): \mathsf{rtc}\;\mathsf{a}\;\mathsf{b} \to \mathsf{rtc}\;\mathsf{b}\;\mathsf{c} \to \mathsf{rtc}\;\mathsf{a}\;\mathsf{c}
```

One of the key properties of rtc is that it is idempotent—applying rtc to rtc r has no effect, for any binary predicate r:

```
lemma rtc_rtc_iff_rtc \{\alpha: \text{Type}\}\ (\text{r}: \alpha \to \alpha \to \text{Prop})\ (\text{a b}: \alpha):
  rtc (rtc r) a b \leftrightarrow rtc r a b :=
begin
  apply iff.intro,
  { intro h,
    induction h,
    case rtc.base : x y {
       assumption },
     case rtc.refl : x {
       apply rtc.refl },
     case rtc.trans : x y z {
       apply rtc.trans,
       assumption,
       assumption } },
  { intro h,
     apply rtc.base,
     assumption }
end
```

In the induction proof, we use the case syntax both to increase readability of the proof script and to give intuitive names to the emerging variables. It is easy to get lost in goals containing long, automatically generated names. The goal management tactics discussed in Section 2.7 can also be a great help when facing large goals.

We can state the idempotence property more standardly in terms of equality instead of as an equivalence:

```
lemma rtc_rtc_eq_rtc \{\alpha: \mbox{Type}\}\ (\mbox{r}: \alpha \to \alpha \to \mbox{Prop}): \mbox{rtc (rtc r)} = \mbox{rtc r}:= \mbox{begin} apply funext, intro a, apply funext, intro b, apply propext, apply rtc_rtc_iff_rtc end
```

The proof requires two lemmas that are available because Lean's logic is classical. Functional extensionality (funext) states that two functions that yield equal results for all inputs must be equal. Propositional extensionality (propext) states that equivalence of propositions coincides with equality.

```
funext : (\forall x, ?f \ x = ?g \ x) \rightarrow ?f = ?g
propext : (?a \leftrightarrow ?b) \rightarrow ?a = ?b
```

These properties may seem obvious, but there are proof assistants built on weaker, intuitionistic logics in which the property do not generally hold.

5.8 Induction Pitfalls

Inductive predicates often have parameters that evolve through the induction. Pattern matching and cases handle this gracefully, but some care is necessary when invoking induction. Such details are often glossed over in informal proofs, but proof assistants require us to be precise.

Suppose we have the definition

```
inductive even : \mathbb{N} \to \text{Prop} | zero : even 0 | add_two : \forall n, even n \to \text{even} (n + 2)
```

If the goal has the form $\mathtt{even}\, \mathtt{n} \to \mathtt{p}\, \mathtt{n}$, applying induction will produce the following cases:

```
p0
∀n, pn → p (n + 2)
```

This works as desired.

The problem is that when the argument to even is not a variable, induction destroys the identity of the argument. (With a variable, there is nothing to destroy.) Thus, applying induction on even $(2 * n + 1) \rightarrow false$ produces the subgoals

```
    false
```

• false \rightarrow false

The first subgoal is not provable. To solve this issue, we need to replace 2 * n + 1 by a variable x and add an equation x = 2 * n + 1 as a hypothesis:

```
\forall x, even x \rightarrow x = 2 * n + 1 \rightarrow false
```

This proposition is logically equivalent, but now we obtain very different subgoals:

```
• 2*n+1=0 \rightarrow false
```

```
• \forall x, (2*n+1=x \rightarrow false) \rightarrow 2*n+1=x+2 \rightarrow false
```

Sadly, the second subgoal is not provable. The problem is that we want to instantiate the $\tt n$ in the assumption with $\tt n+1$ but $\tt n$ is not instantiable.

The solution? We need to explicitly quantify over ${\tt n}$ to be able to instantiate it in the induction hypothesis. We state our goal as

$$\forall x$$
, even $x \rightarrow \forall n$, $x = 2 * n + 1 \rightarrow false$

Now we obtain the subgoals

- \forall n, 2 * n + 1 = 0 \rightarrow false;
- $\forall x$, $(\forall n, 2 * n + 1 = x \rightarrow false) \rightarrow \forall n, 2 * n + 1 = x + 2 \rightarrow false$.

Now the variable n in the conclusion is disconnected from the variable n in the assumption, and we can instantiate the assumption's n with the conclusion's n+1.

5.9 New Tactics

CC

The cc tactic applies an algorithm known as congruence closure to derive new equalities from those that are present in the goal. For example, if the goal contains b = a and f $b \neq f$ as hypotheses, the congruence closure procedure will derive f b = f a from b = a and discover a contradiction with f $b \neq f$ a. The tactic is also suitable for more pedestrian work, inferring for example $a \lor b \lor c$ from b.

linarith

The linarith tactic can be used to prove goals involving linear arithmetic equalities or inequalities. "Linear" means that multiplication and division do not occur, of if they do they one of the operand must be a numeric constant. For example, 2 * x < y is a linear constraint—indeed, it can be rewritten as x + x < y—whereas x * y < y is nonlinear.

norm_num

The norm_num tactic can be used to prove goals requiring normalization of arithmetic expressions. For example, it can be used to prove that 1 + (x + -1) = x.

5.10 Summary of New Syntax

Tactic Combinators

repeat repeatedly applies a tactic on all goals until failure

Tactic

cc propagates the equalities in the goal
linarith applies a procedure for linear arithmetic
norm_num normalizes arithmetic expressions

Chapter 6

Monads

We take a look at an important functional programming abstraction: monads. Monads generalize computation with side effects. Haskell has shown that monads can be used very successful to write imperative programs. In Lean, they are used to write imperative programs, but also to reason about them via the so-called monadic laws. Monads are even useful for programming Lean itself—or metaprogramming Lean—as we will see in Chapter 7.

These notes are inspired by Chapter 7 of *Programming in Lean* [2]. We strongly recommend that you read that chapter to get a more complete picture of monads in Lean. We also refer to Chapter 14 of *Real World Haskell* [18] for a general introduction to monads.

6.1 Motivating Example

Consider the following programming task: Implement a function $sum_2_5_7$ ns that sums up the second, fifth, and seventh items of a list ns of natural numbers. Use option $\mathbb N$ for the result so that if the list has fewer than seven elements, you can return none.

A straightforward implementation follows:

(Recall that nth counts elements from 0; thus, the second element is at position 1, etc.) The code is quite ugly, because of all the pattern matching on option. Can we do better? It turns out we can. We can put all the ugliness in one function, which we call bind_option:

That function operates on an option (a value of type option). If the option is none, we leave it as is. This corresponds to an error condition, and errors are preserved. Otherwise, if the option is some a for some a, then we apply the specified operation f on a.

We can now use bind_option to program our sum function:

```
def sum_2_5_7_2 (ns : list \mathbb{N}) : option \mathbb{N} := bind_option (list.nth ns 1) (\lambdan2, bind_option (list.nth ns 4) (\lambdan5, bind_option (list.nth ns 6) (\lambdan7, some (n2 + n5 + n7))))
```

Intuitively, the program reads as follows:

- Extract the second item from the list. If it is none, we are done. Otherwise, we obtain this item as n2 and continue with the next step.
- · Perform the same for the fifth and seventh item, mutatis mutandis.
- Return the sum of n2, n5, and n7 in a some wrapper.

Semantically, our new function $sum_2_5_7_2$ is equal to the original $sum_2_5_7$.

Instead of defining bind_option ourselves, we could have used Lean's predefined general bind operation. It takes the same arguments in the same order. Here is the new code:

```
def sum_2_5_7_3 (ns : list \mathbb{N}) : option \mathbb{N} := bind (list.nth ns 1) (\lambdan2, bind (list.nth ns 4) (\lambdan5, bind (list.nth ns 6) (\lambdan7, some (n2 + n5 + n7))))
```

One of the advantages of using the predefined notion is that it provides syntactic sugar, in the form of an infix >>= operator:

```
def sum_2_5_7_4 (ns : list \mathbb{N}) : option \mathbb{N} := list.nth ns 1 >>= \lambdan2, list.nth ns 4 >>= \lambdan5, list.nth ns 6 >>= \lambdan7, some (n2 + n5 + n7)
```

The syntax ma >>= f expands to bind ma f.

The penultimate version of the sum program uses heavier syntactic sugar:

```
def sum_2_5_7<sub>5</sub> (ns : list \mathbb{N}) : option \mathbb{N} := do n2 \leftarrow list.nth ns 1,
do n5 \leftarrow list.nth ns 4,
do n7 \leftarrow list.nth ns 6,
some (n2 + n5 + n7)
```

The do notation provides an intuitive notation for monads. do a \leftarrow ma, t is syntactic sugar for ma >>= λ a, t. If we do not care about the return result of ma's computation, we can write do ma, t, which expands to ma >>= λ _, t. The notation supports iteration, so there is no need to put the keyword do on every line. This brings us to the final version of the program:

```
\begin{array}{lll} \text{def sum}\_2\_5\_7_6 & (\text{ns}: \text{list } \mathbb{N}) : \text{option } \mathbb{N} := \\ \text{do} \\ & \text{n2} \leftarrow \text{list.nth ns 1,} \\ & \text{n5} \leftarrow \text{list.nth ns 4,} \\ & \text{n7} \leftarrow \text{list.nth ns 6,} \\ & \text{some } (\text{n2} + \text{n5} + \text{n7}) \end{array}
```

Each line with a left arrow (\leftarrow) attempts to read a value. In case of failure, the entire program evaluates to none. Although the notation has an imperative flavor, the function is a pure functional program.

6.2. Overview 57

6.2 Overview

The option type constructor is an example of a monad. In general, a monad is a type constructor m that depends on some type parameter α —i.e., m α —equipped with two distinguished operations:

```
\texttt{pure}: \Pi\{\alpha: \texttt{Type}\}\text{, } \alpha \to \texttt{m} \ \alpha \texttt{bind}: \Pi\{\alpha \ \beta: \texttt{Type}\}\text{, } \texttt{m} \ \alpha \to (\alpha \to \texttt{m} \ \beta) \to \texttt{m} \ \beta
```

Recall that curly braces denote implicit arguments. For options, the pure operation is simply some, whereas bind is what we called bind_option.

Intuitively, we can think of a monad as a "box" containing some data. The operation pure puts the data into the box, whereas bind allows us to access the data in the box and modify it—possibly even changing its type, since the result is an m β monad, not an m α monad. But there is no general way to extract the data from the monad, i.e., to obtain an α from an m α .

To summarize, pure a (also called return a) provides no side effect and simply provides a box containing the the value a, whereas bind ma f (also written ma >>= f) executes ma, then executes f with the boxed result a of ma.

Monads are an abstract concept with many applications. The option monad is only one instance among many. The following table gives an overview of monad instances and the effects they provide.

Туре	Effect
$\overline{\mathtt{option}\;\alpha}$	failure when none; otherwise, some (a : $lpha$) is the result
$\mathtt{set}\ \alpha$	nondeterministic behavior (set of behaviors)
$t \to \alpha$	reading elements of type t (e.g., a configuration)
$\mathbb{N} \times \alpha$	adjoining running time (e.g., to model algorithmic complexity)
$\mathtt{string} \times \alpha$	adjoining text output (e.g., for logging)
$\mathtt{prob}\;\alpha$	probability (e.g., using random number generators)
io $lpha$	interaction with the operating system
$\verb"tactic"\alpha$	interaction with the prover

All of the above are type constructors m are parameterized by a type α . Some effects can be combined (e.g., option $(t \to \alpha)$). Some effects are not executable (e.g., set α , prob α); they are nonetheless useful for modeling programs abstractly in the logic.

Specific monads m may provide further operators. For example, they may provide a way to extract the boxed value stored in the monad without bind's requirement of putting it back in a monad.

Monads have several benefits. We get the convenient and highly readable do notation. They also offer generic operations, such as $\mathtt{mmap}: (\alpha \to \mathtt{M}\ \beta) \to \mathtt{list}\ \alpha \to \mathtt{M}\ (\mathtt{list}\ \beta)$, which work uniformly across all monad instances. To quote $Programming\ in\ Lean\ [2]$, "The power of the abstraction is not only that it provides general functions and notation the can be used in all these various instantiations, but also that it provides a helpful way of thinking about what they all have in common."

6.3 Monadic Laws

The bind and pure operations are normally required to obey three laws, called the monadic laws.

The bind operation combines two programs. If either of these is simply pure data, we can eliminate the bind. This gives us the first two monadic laws. The first law:

The second law:

$$\begin{array}{ccc} (do & & \\ & a \leftarrow x, & = & x \\ & pure \ a) & \end{array}$$

The third law is an associativity rule for bind. It allows us to flatten a nested computation:

6.4 A Type Class of Monads

Monads are a mathematical structure, so we use class to add them as a type class. Type classes will be covered in more detail in Chapter 12. We can think of a type class as a structure or record that is parameterized by a type—or here, by a type constructor $\mathtt{m}: \mathsf{Type} \to \mathsf{Type}$.

A possible Lean definition of monads, together with the three monadic laws, follows:

```
class lawful_monad (m : Type \rightarrow Type) extends has_bind m, has_pure m : Type 1 := (pure_bind \{\alpha \ \beta\} (a : \alpha) (f : \alpha \rightarrow m \beta) : (pure a >>= f) = f a) (bind_pure \{\alpha\} (ma : m \alpha) : ma >>= pure = ma) (bind_assoc \{\alpha \ \beta \ \gamma\} (f : \alpha \rightarrow m \beta) (g : \beta \rightarrow m \gamma) (ma : m \alpha) : ((ma >>= f) >>= g) = (ma >>= (\lambdaa, f a >>= g)))
```

Let us analyze this definition step by step:

- We are creating a structure parameterized by a unary type constructor ${\tt m}:{\tt Type}\to{\tt Type}$.
- The type class inherits the fields, and any syntactic sugar, from type classes called has_bind and has_pure. These provide the bind and pure operations on m, with the expected types and the syntactic sugar.
- The type Type 1 is a technical necessity. Because monads quantify over types, they cannot be standard types themselves. They need to live in a larger type universe, called Type 1. (The familiar Type type of types is also called Type 0.) We will come back to this fine point in Chapter 11.
- Finally, the definition adds three fields to those already provided by has_bind and has_pure. Each field is a proof of one of the three monadic laws.

We call our type class "lawful monad" because the monadic laws are required to hold. To instantiate this definition with a concrete monad (e.g., the option monad), we must supply the monad m, suitable bind and pure operators, and proofs corresponding to the three monadic laws.

Our definition captures the essence of Lean's predefined concept of monads. Lean's actual definition is more complicated. One of the differences is that it distributes the definition over two type classes, called monad and is_lawful_monad.

6.5 The Option Monad

The following code extract shows how to register the type constructor $option: Type \rightarrow Type$ as a lawful monad:

```
namespace option
def pure \{\alpha : \mathsf{Type}\} : \alpha \to \mathsf{option} \ \alpha :=
option.some
def bind \{\alpha \ \beta : \text{Type}\} : option \alpha \to (\alpha \to \text{option } \beta) \to \text{option } \beta
            f := none
| (some a) f := f a
instance lawful_monad_option : lawful_monad option :=
            := @option.pure,
                := @option.bind,
  bind
  pure_bind :=
     begin
       intros \alpha \beta a f,
       refl
     end,
  bind_pure :=
     begin
       intros \alpha m,
       cases m; refl
     end,
  bind_assoc :=
     begin
       intros \alpha \beta \gamma f g m,
       cases m; refl
     end }
end option
```

The registration process requires us to provide five components: the pure and bind operations and the proofs of the three monadic laws. To avoid polluting the name space, we tidily put our code in an option namespace. For the first and second components of the record, we use the @ syntax to make the type arguments explicit, in agreement with the format expected by has_pure and has_bind. The three proofs are straightforward.

6.6 The State Monad

The next monad is called the state monad. It provides an abstraction corresponding to a mutable state. For some programming languages, the compiler can detect the use of the state monad and translate state-monadic programs to more efficient imperative programs.

The type constructor is

```
def state (\sigma : Type) (\alpha : Type) := \sigma \to \alpha \times \sigma
```

For a given σ , we have that state $\sigma: \mathsf{Type} \to \mathsf{Type}$ is a monad. We abstract over the exact memory layout. We could use tuples or lists to represent memory, for example, and instantiate the abstract σ accordingly.

Intuitively, the $\sigma \to \text{part}$ of state's definition provides the current state; the left component of the cartesian product, α , is where the result of a computation goes; and the

right component of the product, σ , can be used to provide a new state, overwriting the received one. Thus, the state is threaded through the monadic program.

We start by defining basic operations: two operations, read and write, to access the memory, and the monadic operations bind and pure:

```
namespace state  \begin{split} & \text{def read } \{\sigma : \text{Type}\} : \text{state } \sigma \ \sigma \\ & | \ \text{s} := (\text{s}, \ \text{s}) \\ & \text{def write } \{\sigma : \text{Type}\} \ (\text{s} : \sigma) : \text{state } \sigma \ \text{unit} \\ & | \ \text{t} := ((), \ \text{s}) \\ & \text{def pure } \{\sigma \ \alpha : \text{Type}\} \ (\text{a} : \alpha) : \text{state } \sigma \ \alpha \\ & | \ \text{s} := (\text{a}, \ \text{s}) \\ & \text{def bind } \{\sigma : \text{Type}\} \ \{\alpha \ \beta : \text{Type}\} \ (\text{f} : \text{state } \sigma \ \alpha) \\ & (\text{g} : \alpha \to \text{state } \sigma \ \beta) : \text{state } \sigma \ \beta \\ & | \ \text{s} := \text{let } (\text{a}, \ \text{t}) := \text{f} \ \text{s in g a t} \end{split}
```

The read operation simply returns the current state s (in the first pair component) and leaves the state unchanged (in the second pair component).

The write operation replaces the state s with t and returns an empty tuple.

The pure operation returns the current state ${\tt s}$ unchanged tupled together with the given value ${\tt a}$.

The bind operation passes the initial state to the f argument, yielding a result a and a new state t. These are passed to g, which returns a new result and a new state.

Registering the state monad as a lawful monad is similar as for option, but we need to reason about functional extensionality (funext). Namely, to prove f = g, it suffices to prove $\forall x$, f = g x.

```
instance \{\sigma : \mathsf{Type}\} : \mathsf{lawful\_monad} \ (\mathsf{state} \ \sigma) :=
{ pure
               := @state.pure \sigma,
              := @state.bind \sigma,
  bind
  pure_bind :=
    begin
       intros \alpha \beta a f,
       apply funext,
       intro s,
       refl
    end,
  bind_pure :=
    begin
       intros \alpha m,
       apply funext,
       intro s,
       simp [bind],
       cases (m s),
       refl
     end,
  bind_assoc :=
    begin
       intros \alpha \beta \gamma f g m,
       apply funext,
       intro s,
       simp [bind],
       cases (m s),
```

```
refl
end }
end state
```

As an example, the following state-monadic program removes all elements that are smaller than a previous element in the list, leaving us with a list of increasing elements. The maximal element is stored as the state in the monad.

Notice how the state is accessed by read and write.

To execute the program, we must provide a start state. We also get back the last state, which is the largest element seen in the list or the start state:

```
#eval diff_list [1, 2, 3, 2] 0
#eval diff_list [1, 2, 3, 2, 4, 5, 2] 0
Output:
    ([1, 2, 3], 3)
    ([1, 2, 3, 4, 5], 5)
```

6.7 Example: Generic Iteration over a List

As a final example, we consider a generic monadic program mmap that iterates over a list and applies a function f to each element. The definition is recursive and monadic:

```
def mmap {m : Type \rightarrow Type} [lawful_monad m] {\alpha \beta : Type} (f : \alpha \rightarrow m \beta) : list \alpha \rightarrow m (list \beta) | [] := pure [] | (a :: as) := do b \leftarrow f a, bs \leftarrow mmap as, pure (b :: bs)
```

Notice that the function returns a single monad containing the resulting list, not a list of monads. Try to work out why the function is well typed and has the expected behavior.

The mmap function distributes over the append operator ++ of lists. The monadic notation is useful not only for defining functions but also for stating their properties:

```
lemma mmap_append {m : Type \rightarrow Type} [lawful_monad m] {\alpha \beta : Type} (f : \alpha \rightarrow m \beta) : \forallas as' : list \alpha, mmap f (as ++ as') = do bs \leftarrow mmap f as, bs' \leftarrow mmap f as',
```

```
pure (bs ++ bs')

| [] _ := by simp [mmap]

| (a :: as) as' := by simp [mmap, mmap_append as as']

Here is an example of how we could use mmap:

def nths \{\alpha : \text{Type}\}\ (\text{xss} : \text{list (list }\alpha))\ (\text{n} : \mathbb{N}) : \text{option (list }\alpha) := \text{mmap }(\lambda \text{xs, list.nth xs n) xss}
```

The function nths xss n selects the (n + 1)st elements of each list in xss. It fails with none if one of the list is not long enough. For example, the call

```
#eval nths
[[11, 12, 13, 14],
[21, 22, 23],
[31, 32, 33]] 2
```

produces the result some [13, 23, 33].

6.8 Summary of New Syntax

Declaration

class declares a structure as a type class

Monadic Notations

do indicates the start of a monadic program >>= combines effectful computations

Chapter 7

Metaprogramming

Like most proof assistants, Lean can be extended with custom functionality, notably tactics. This kind of programming—programming the prover—is called metaprogramming. Lean's metaprogramming framework uses mostly the same notions and syntax as Lean's input language itself, so that we do not need to learn a different programming language.

These notes are partly inspired by an article by Gabriel Ebner et al. [5] and by Chapter 8 of *Programming in Lean* [2]. We strongly recommend that you read the chapter.

7.1 Overview

Lean's metaprogramming framework enables us to program the prover in its own language, using a monadic language. Abstract syntax trees, presented as inductive types in Lean, reflect internal data structures. For example, the type of expressions—i.e., terms—is viewed as an inductive type with ten constructors, but internally in Lean it is a C++ data structure. The prover's C++ internals are exposed through Lean interfaces, which we can use for accessing the current context and goal, unifying expressions, querying and modifying the environment, and setting attributes (e.g., @[simp]).

Most of Lean's predefined tactics are implemented as metaprograms, and not in C++. We can find these definitions by clicking on the name of a construct in Visual Studio Code while holding the control or command key.

Here are some example applications of metaprogramming:

- proof goal transformations (e.g., apply all safe introduction rules for connectives, put the goal in negation normal form);
- heuristic proof search (e.g., apply unsafe introduction rules for connectives and hypotheses with backtracking);
- decision procedures (e.g., for propositional logic, linear arithmetic);
- definition generators (e.g., Haskell-style derive for inductive types);
- advisor tools (e.g., lemma finders, counterexample generators);
- exporters (e.g., documentation generators);
- · ad hoc proof automation (to avoid boilerplate or duplication).

A key benefit of Lean's metaprogramming framework is that users, who may not even be computer scientists, do not need to learn another programming language to write metaprograms. They can work with the same constructs and notation used to define ordinary objects in the prover's library. Everything in that library is available for metaprogramming purposes (e.g., \mathbb{Z} , list, algebraic structures). Metaprograms can be written and debugged in the same interactive environment, encouraging a style where formal libraries and supporting automation are developed at the same time.

The framework is articulated around three main types:

- the tactic monad, which contains the proof state, the environment, and more;
- the name type for storing structured names (e.g., option.bind);
- the expr type representing terms, types, proofs, and propositions as abstract syntax trees.

7.2 The Tactic Monad

The tactic monad combines the attributes of several kinds of monads. It is a state monad providing access among others to the list of goals, the environment (including all definitions and inductive types), notations, and attributes (e.g., the list of @[simp] rules).

Goals are represented as metavariables, standing for the missing proofs. Each metavariable has a type and a local context giving the constants and hypotheses. By the Curry–Howard correspondence, constants and hypotheses are not distinguished. Tactics have access to the elaborator, to elaborate expressions and compute their types.

The tactic monad also behaves like an option monad. The program fail msg, where msg is a text string, leaves the monad in an error state.

Moreover, the tactic monad is also an alternative monad. Namely, it provides an operator <|> for specifying alternatives. The program s < > t first executes s; if s is a failure, then the state is reverted to its initial condition and t is executed instead.

Finally, tactic provides some tracing capabilities. We can use trace msg, where msg is a text string, or we can display the current proof state using trace_state.

Let us put the tactic monad to some use:

```
example : true :=
by do
  tactic.trace "Hello, Metacosmos!",
  tactic triv
```

This first example is very modest: It uses the monad's tracing facilities to print the message "Hello, Metacosmos!" before applying the triv tactic, which can be used to prove trivial goals such as true. The do keyword signals entry into the tactic monad. The by keyword is necessary to enter tactic mode.

As is usually the case in computer science, we can introduce a layer of indirection. We can package our combination of tracing and proving as its own tactic, and gloriously apply it to prove true:

```
meta def hello_world : tactic unit :=
do
   tactic.trace "Hello, Metacosmos!",
   tactic.triv

example : true := by hello_world
```

The hello_world tactic's type is tactic unit. In general, tactics will have the type tactic σ , where σ is the type of some value returned by the tactic. Often our tactics will only have side effects on the state but return no value. In such cases, we return the empty tuple, of type unit. This is what the triv tactic returns, and hence this is what we return.

Also notice the presence of the meta keyword in front of the function definition. Any executable (i.e., not defined using the noncomputable keyword) Lean definition can be used as a metaprogram. In addition, we can put meta in front of a definition to indicate that is a metadefinition; these need not terminate but cannot be used in non-meta contexts. Metaprograms (whether defined with meta or not) can communicate with Lean through metaconstants, which are implemented in C++ and have no logical meaning—i.e., they are opaque names. Metaprograms can also call other metaprograms.

Most tactics are designed to work on a proof state, but the Lean concept of a tactic is more general, and they can be used also on the top level, without a proof context, by using the run_cmd command:

7.2. The Tactic Monad 65

```
run_cmd tactic.trace "Hello, Metacosmos!"
```

Tactics usually need to adapt to the situation where they are called. This means they must inspect the current goal. The following example prints the local context, the goal conclusions, and the current goal or "target":

```
open tactic

example (α : Type) (a : α) : true :=
by do
   trace "local context:",
   local_context >>= trace,
   trace "goals:",
   get_goals >>= trace,
   trace "target:",
   target >>= trace,
   triv
```

We start by opening the tactic namespace because we will need much of its functionality. The output is as follows:

```
    local context: [α, a];
    goals: [?m_1];
```

· target: true.

The goals are a list of variables of certain types that must be instantiated. They are normally not displayed in the proof state; instead, their types are shown. In the example, we have $?m_1 : true$.

The next example performs something somewhat more useful. It implements a find_assumption tactic that, like the predefined assumption tactic, looks for a hypothesis of the right type (i.e., right proposition) and applies it to solve the current goal:

The main function extracts all available hypotheses forming the local context, which are stored as expressions of type expr, and iterates through them. For each hypothesis, we first try to apply the predefined exact tactic (which exists also as a monadic program, like many other tactics). If this fails, the <|> operator passes control to the right branch, in which we continue recursively with the remaining hypotheses. If no hypotheses are left, we invoke fail.

A simple invocation follows:

```
example {p : Prop} {\alpha : Type} (a : \alpha) (h : p) : p := by do find_assumption
```

The tracing information informs us that p, α , a, and h have been tried in turn until the matching h was found.

7.3 Names and Expressions

Terms, including proof terms, are represented by an inductive type expr of "expressions" in the metaprogramming framework:

Briefly:

- var represents a bound variable, using a notation known as De Bruijn indices.
- sort is used to represent the type of types Type, Type 1, etc.
- const represents a constant, such as list.reverse or $\mathbb{N}.$
- mvar represents a metavariable, i.e., a variable with a question mark (e.g., ?m_1).
- local_const represents a local constant in the proof state (e.g., h).
- · app represents the application of a function to an argument.
- lam represents a λ binder.
- pi represents a Π binder.

Fortunately, we can often use higher-level interfaces to manipulate expressions. A lightweight way to construct an expression is using Lean's quoting mechanism, based on backticks. Expressions are enclosed in parentheses and preceded by one, two, or three backticks (').

An expression preceded by a single backtick—(e)—must be fully elaborated. It may not contain holes, as in the second example below:

```
run_cmd do let e : expr := '(list.map (\lambdan : \mathbb{N}, n + 1) [1, 2, 3]), trace e run_cmd do let e : expr := '(list.map _ [1, 2, 3]), -- fails (holes are disallowed) skip
```

With two backticks—''(e)—we have *pre-expressions*. These are Lean expressions that may contain some holes to be filled in later, based on some context:

```
run_cmd do let e_1: pexpr := ''(list.map (\lambdan, n + 1) [1, 2, 3]), let e_2: pexpr := ''(list.map _ [1, 2, 3]), trace e_1, trace e_2
```

With three backticks—'''(e)—we have pre-expressions without name checking. We can refer to some_silly_name even if it does not exist at the point when the pre-expression is parsed:

```
run_cmd do
  let e := '''(some_silly_name),
  trace e
```

A similar quoting mechanism exists for names. The syntax consists of one or two backticks, without parentheses:

```
run_cmd trace 'some.silly.name
run_cmd trace 'true
run_cmd trace 'some.silly.name -- fails (not found)
```

The single-backtick variant—'n—does not check whether the name exists; the two-backtick variant—'n—performs this check. Whether the check is desirable depends on the situation.

Sometimes we would like to plug in existing expressions into a larger expression. This can be achieved using *antiquotations*. These are announced by a prefix %% followed by a name from the current context. Antiquotations are available with one, two, and three backticks:

```
run_cmd do
  let x : expr := '(2 : N),
  let e : expr := '(%x + 1),
  trace e

run_cmd do
  let x : expr := '(@id N),
  let e := ''(list.map %xx),
  trace e

run_cmd do
  let x : expr := '(@id N),
  let e := '''(a _ %xx),
  trace e
```

Antiquotations can also be used to perform pattern matching:

```
example : 1 + 2 = 3 := by do  `(\% a + \% b = \% c) \leftarrow \text{target}, \\ \text{trace a,} \\ \text{trace b,} \\ \text{trace c,} \\ `(@eq \% \alpha \% l \% r) \leftarrow \text{target,} \\ \text{trace } \alpha, \\ \text{trace 1,} \\ \text{trace r,} \\ \text{exact } `(\text{refl }\_: 3 = 3)
```

In the example, the two \leftarrow lines, corresponding to monadic bind operations, succeed only if the right-hand side has the right shape (which it has). Given that target is 1 + 2 = 3, executing the first line binds the variables a, b, r to 1, 2, 3, respectively. The second pattern looks at 1 + 2 = 3 through different lenses—namely, without syntactic sugar, as an expression of the form (@eq α 1 r), for some α , 1, r. This also succeeds.

7.4 Example: A Destructive Tactic

Next, we review two examples of metaprograms that accomplish a well-defined task. The first example is a tactic called destruct_and that automates the elimination of conjunctions in premises. Our aim is to automate proofs such as the following:

```
example {a b c d : Prop} (h : a \land (b \land c) \land d) : a := and.elim_left h 
example {a b c d : Prop} (h : a \land (b \land c) \land d) : b := and.elim_left (and.elim_left (and.elim_right h)) 
example {a b c d : Prop} (h : a \land (b \land c) \land d) : b \land c := and.elim_left (and.elim_right h)
```

We would like in each case to simply write destruct_and h.

Our tactic relies on a helper function, which takes as argument the hypothesis ${\tt h}$ to use and its type—i.e., its proposition:

We perform pattern matching with antiquotations to see if the hypothesis h is of the form a \land b. If so, we try three things in turn. First, we try to invoke the exact tactic. If this fails, we try the left elimination rule, producing and elim_left h: a, and continue recursively with that. The to_expr call converts a pre-expression into an expression, elaborating it. If our attempt at left elimination fails, we try right elimination instead.

The main function has very little to do:

```
meta def destruct_and (n : name) : tactic unit :=
do
    h ← get_local n,
    t ← infer_type h,
    destruct_and_helper t h
```

It retrieves the hypothesis h with the specified name h, extracts its type (i.e., its proposition), and calls the helper function.

We can now turn to our motivating examples and prove them again, this time using our new widget:

```
example {a b c d : Prop} (h : a \land b \land c) : a := by destruct_and 'h example {a b c d : Prop} (h : a \land b \land c) : c := by destruct_and 'h example {a b c d : Prop} (h : a \land b \land c) : b \land c := by destruct_and 'h
```

Unfortunately, we need to quote the name of the hypothesis with a backtick. We would need more (not particularly pleasant to write) code to provide proper parsing for our tactic's argument.

7.5 Example: A Solvability Advisor

Our second example enriches Lean with functionality that is available in the Isabelle/HOL proof assistant. It sometimes happens that a user states a lemma, proves it, and later realizes that the lemma already existed in the library. To help prevent this, we implement a solve_direct tactic, which goes through all lemmas in the database and checks whether one of them can be used to fully solve the current goal.

We will review the code in steps. The first step is a function is_theorem that returns true (tt) if a declaration is a theorem or an axiom and false (ff) otherwise:

We will use this function to filter out the declarations that do not interest us. The next function returns the list of all theorem (and axiom) names:

```
meta def get_all_theorems : tactic (list name) := do env \leftarrow get_env, pure (environment.fold env [] (\lambdadecl thms, if is_theorem decl then declaration.to_name decl :: thms else thms))
```

The get_env function (in fact, tactic) returns an object representing the environment. We can traverse it using environment.fold, which is applied on each declaration decl in turn and on an accumulator thms. We extend the accumulator with each declaration's name that satisfies the is_theorem predicate. At the end, fold returns the accumulator.

The fold function is a common idiom in functional programming. If you are not familiar with the concept, we recommend Chapter 6 of Learn You a Haskell for Great Good!

Another component we need is a tactic that attempts to (fully) solve the goal using a theorem called ${\tt n}$.

```
meta def solve_with_name (n : name) : tactic unit :=
do
   cst \( - \text{ mk_const n,} \)
   apply cst { md := transparency.reducible, unify := ff },
   all_goals assumption,
   done
```

 ${\tt mk_const\; n}$ is used to produce an ${\tt expr}$ representing the proof from a name ${\tt n}$ (the theorem name).

The apply tactic applies an expr to the current goal. For performance, we pass an optional configuration, { md := transparency.reducible, unify := ff }, which tells apply to perform less computational unfolding.

The all_goals tactical, in combination with assumption, is used to ensure that the hypothesis from the local context are used to fill in any remaining subgoals.

At the end, done is used to check that no subgoal is left.

Finally, our solve_direct tactic goes through all theorems found by get_all_theorems and succeed with the first theorem that solves the current goal:

```
meta def solve_direct : tactic unit := do  thms \leftarrow \texttt{get\_all\_theorems}, \\ list.mfirst ($\lambda n$, do
```

¹http://learnyouahaskell.com/higher-order-functions

```
solve_with_name n,
  trace ("directly solved by " ++ n.to_string))
thms
```

The list.mfirst function applies a tactic to each element of a list until one application succeeds. We use trace to output the successful theorem to the user, so that they can apply the theorem directly instead of relying on solve_direct.

As a small refinement of the above, we propose a version of $solve_direct$ that also looks for equalities stated in symmetric form—for example, if the goal is l = r but the theorem is r = l:

```
meta def solve_direct_symm : tactic unit :=
solve_direct
<|> (do
   cst ← mk_const 'eq.symm,
   apply cst,
   solve_direct)
```

Our solution is quite simple: We first try solve_direct, and if that fails, we apply symmetry of equality (the property $?1 = ?r \rightarrow ?r = ?1$) to change the goal and try solve_direct again.

7.6 Summary of New Syntax

Declaration

meta prefixes declarations of metaprograms and metaconstants

Command

run_cmd executes a tactic on the top level, without a proof context

Quotations

'n	quotes literal name, with checking
``n	quotes literal name, without checking
'(e)	quotes fully elaborated expressions without holes
''(e)	quotes pre-expressions, with name checking
'''(e)	quotes pre-expressions, without name checking
%%x	embeds an expression in a quotation

Tactic Monad Functions

applies the given tactic to all goals all_goals gives access to the list of all declarations in the environment get_env gives the list of all proof holes to fill in get_goals finds a local constant by name get_local computes the type of an expression infer_type local_context gives the list of all local constants builds a constant from a name mk_const target gives the current goal to prove elaborates a pre-expression to_expr outputs a debug string trace outputs the current proof state trace_state

Part III Program Semantics

Chapter 8

Operational Semantics

In this and the next two chapters, we will see how to use Lean to specify the syntax and semantics of programming languages and to reason about the semantics.

This chapter is inspired by Chapter 7 of Concrete Semantics: With Isabelle/HOL [17], whose reading we recommend.

8.1 Motivation

A formal semantics helps specify and reason about the programming language itself, and about individual programs. It can form the basis of verified compilers, interpreters, verifiers, static analyses, type systems, etc. Without formal proofs, semantics, compilers, interpreters, verifiers, static analyses, type systems, etc., are almost always wrong.

Proof assistants are widely used in the area of programming language research. For example, every year, about 10%–20% of papers presented at POPL (Principles of Programming Languages) are partially or totally formalized. This is possible because comparatively little machinery (background libraries, tactics) is needed to get started, beyond inductive types and predicates and recursive functions, which are provided by the provers. The proofs tend to have lots of cases, which is a good match for computers. Proof assistants are convenient to keep track of what needs to be changed as we extend a programming language with more features.

8.2 A Minimalistic Imperative Language

WHILE¹ is a minimalistic imperative language with the following grammar:

```
S ::= skip (no-op)

\mid x := a (assignment)

\mid S ; S (sequential composition)

\mid if b then S else S (conditional statement)

\mid while b do p (while loop)
```

where S stands for a statement (also called command or program), x for a program variable, a for an arithmetic expression, and b for a Boolean expression.

In our grammar, we deliberately leave the syntax of arithmetic and Boolean expressions unspecified. In Lean, we have the choice:

We can use a type such as aexp from Section 1.2 and similarly for Boolean expressions.

¹Fans of backronyms might enjoy this one: Weak Hypothetical Imperative Language Example.

• We can decide that an arithmetic expression is simply a function from states to natural numbers (e.g., $\mathtt{state} \to \mathbb{N}$) and a Boolean expression is simply a predicate over states (e.g., $\mathtt{state} \to \mathtt{Prop}$ or $\mathtt{state} \to \mathtt{bool}$). A *state* is a mapping from program variables to values. For example, the arithmetic expression $\mathtt{x} + \mathtt{y} + \mathtt{1}$ would be represented by the function $\lambda\mathtt{s}$, \mathtt{s} "x" + \mathtt{s} "y" + 1, and the Boolean expression $\mathtt{a} \neq \mathtt{b}$ by the predicate $\lambda\mathtt{s}$, \mathtt{s} "a" $\neq \mathtt{s}$ "b".

This corresponds to the difference between deep and shallow embeddings: A *deep embedding* of some syntax (expression, formula, program, etc.) consists of an abstract syntax tree specified in the proof assistant (e.g., aexp) with a semantics. In contrast, a *shallow embedding* simply reuses the corresponding mechanisms from the logic (e.g., functions and predicates).

A deep embedding allows us to reason explicitly about a program's syntax. A shallow embedding is more leightweight, because we can use it directly, without having to define a semantics (e.g., an eval function for aexp).

We will use a deep embedding of programs (which we find interesting) and a shallow embeddings of arithmetic and Boolean expressions (which we find boring). Our Lean definition of programs follows:

```
inductive program : Type | skip {} : program | skip {} : program | assign : string \rightarrow (state \rightarrow \mathbb{N}) \rightarrow program | seq : program \rightarrow program \rightarrow program | ite : (state \rightarrow Prop) \rightarrow program \rightarrow program | while : (state \rightarrow Prop) \rightarrow program \rightarrow program infixr '; ':90 := program.seq
```

The infix operation S;; T is provided as alternative to seq ST.

The correspondence between the inductive type's constructors and the grammar rules should be clear. Variables are represented by strings. The type state is defined as string $\to \mathbb{N}$, a mapping from variable names to values. For simplicity, our program variables are all of type natural number, and (somewhat unusually) all possible variable names exist in the state and are assigned a value.

8.3 Big-Step Semantics

An *operational semantics* corresponds to an idealized, abstract interpreter. There are two main variants: big-step semantics and small-step semantics. We will start by giving a big-step semantics to our WHILE language.

The traditional (and also the most readable) way to specify such a semantics is through a formal system of derivation rules, in the style of the typing rules presented in Sections 1.1 and 3.4.

In a big-step (operational) semantics (also called natural semantics), judgments have the form $\langle S, s \rangle \Longrightarrow t$ and the following intuitive interpretation:

Starting in a state s, executing S terminates in the state t.

The angle brackets in $\langle \mathtt{S}, \mathtt{s} \rangle$ can be used instead of parentheses to increase readability when \mathtt{p} or \mathtt{s} is a complex term. In keeping with the definition of WHILE programs, a state \mathtt{s} is a function of type $\mathtt{string} \to \mathbb{N}$. An example of a judgment follows:

```
\langle x := x + y; y := 0, [x \mapsto 3, y \mapsto 5] \rangle \implies [x \mapsto 8, y \mapsto 0]
```

We use the informal notation $[x \mapsto 3, y \mapsto 5]$ to denote the function λvar , if var = "x" then 3 else if var = "y" then 5 else 0 and similarly for $[x \mapsto 8, y \mapsto 0]$. Intuitively, the judgment should hold.

The derivation rules for big-step semantics judgments are given below. The rules can be seen as an abstract interpreter for WHILE program.

$$\frac{}{\langle \text{skip, } s \rangle} \Longrightarrow s \text{SKIP}}{}$$

$$\frac{\langle x := a, s \rangle \Longrightarrow s \{x \mapsto a s\}}{} \text{ASN}}$$

$$\frac{\langle S, s \rangle \Longrightarrow t \qquad \langle T, t \rangle \Longrightarrow u}{\langle S; T, s \rangle \Longrightarrow u} \text{SEQ}$$

$$\frac{\langle S_1, s \rangle \Longrightarrow t}{\langle \text{if } b \text{ then } S_1 \text{ else } S_2, s \rangle \Longrightarrow t} \text{IF-TRUE if } b \text{ s is true}}{}$$

$$\frac{\langle S_2, s \rangle \Longrightarrow t}{\langle \text{if } b \text{ then } S_1 \text{ else } S_2, s \rangle \Longrightarrow t} \text{IF-FALSE if } b \text{ s is false}}{}$$

$$\frac{\langle S, s \rangle \Longrightarrow t \qquad \langle \text{while } b \text{ do } S, t \rangle \Longrightarrow u}{\langle \text{while } b \text{ do } S, s \rangle \Longrightarrow u} \text{WHILE-TRUE if } b \text{ s is true}}$$

$$\frac{\langle S, s \rangle \Longrightarrow t \qquad \langle \text{while } b \text{ do } S, s \rangle \Longrightarrow u}{\langle \text{while } b \text{ do } S, s \rangle \Longrightarrow s} \text{WHILE-FALSE if } b \text{ s is false}}$$

Above, a s denotes the value of arithmetic expression a in state s, and similarly for b s. Moreover, the syntax $s\{x\mapsto n\}$ represents the state that is identical to s except that it maps the variable x to n. Formally: $s\{x\mapsto n\}=(\lambda \text{var}, \text{ if var}=x \text{ then } n \text{ else } s \text{ var}).$ Exercise to the reader: Write the derivation tree for the judgment $\langle x:=x+y; y:=0, [x\mapsto 3, y\mapsto 5]\rangle \implies [x\mapsto 8, y\mapsto 0].$

The derivation rules can be read intuitively. Consider SEQ as an example. The rule can be read as follows:

If (1) executing S in state s leads to state t and (2) executing T in state t leads to state u, then executing the sequential composition S; T in state s leads to state u.

The conditions (1) and (2) correspond to the two premises of SEQ.

The most complicated rule is undoubtedly WHILE-TRUE. Intuitively, it can be understood as follows:

Assume the loop condition b is true in state s. If (1) executing S in state s leads to state t and (2) executing the loop while b do S from state t leads to state u, then executing the loop while b do S in state s leads to state u.

Another way to think about WHILE-TRUE is in terms of sequential composition. If the loop condition is true, while b do S should be equivalent to S; while b do S. The two premises of WHILE-TRUE correspond to the two premises of the instance of the SEQ rule for the statement S; while b do S.

In Lean, a big-step semantics judgment is represented by an inductive predicate whose introduction rules correspond to the derivation rules above:

```
inductive big_step : (program \times state) \rightarrow state \rightarrow Prop | skip {s} : big_step (skip, s) s | assign {x a s} : big_step (assign x a, s) (s{x \mapsto a s}) | seq {S T s t u} (h<sub>1</sub> : big_step (S, s) t)
```

Using an inductive predicate as opposed to a recursive function allows us to cope with nontermination (e.g., a diverging while) and (for languages richer than WHILE) nondeterminism. It also arguably provides a nicer syntax, closer to the judgment rule syntax that is traditionally used in the scientific literature.

We make ample use of implicit arguments, denoted by curly braces, for the variables corresponding to the derivation rule's metavariables.

8.4 Lemmas about the Big-Step Semantics

Equipped with a big-step semantics, we can reason about concrete programs, proving theorems relating final states with initial states. Perhaps more interestingly, we can prove properties of the programming language, such as equivalence proofs between program constructs and determinism.

We will take determinism as an example, stated as follows: If $\langle S, s \rangle \Longrightarrow 1$ and $\langle S, s \rangle \Longrightarrow r$, then 1 = r. Formally:

```
lemma big_step_unique {S s l r} (hl : (S, s) \Longrightarrow 1) (hr : (S, s) \Longrightarrow r) :
```

The proof is by rule induction over $\langle \mathtt{S}, \mathtt{s} \rangle \Longrightarrow \mathtt{1}$, generalizing over r. We give an informal proof sketch:

```
CASE SKIP: The only way to have \langle skip, s \rangle \Longrightarrow 1 or \langle skip, s \rangle \Longrightarrow r is if 1 = r = s.
```

CASE ASN: Similar to SKIP.

CASE SEQ: We have the hypotheses $(S,s)\Longrightarrow t, (T,t)\Longrightarrow 1, (S,s)\Longrightarrow t',$ and $(T,t')\Longrightarrow r$ and the induction hypotheses $\forall r:$ state, $(S,s)\Longrightarrow r\to t=r$ and $\forall r:$ state, $(T,t)\Longrightarrow r\to 1=r$. From the first induction hypothesis together with $(S,s)\Longrightarrow t'$, we derive t=t'. From the second induction hypothesis together with $(T,t')\Longrightarrow r$, we derive 1=r, as desired.

CASE IF-TRUE: The only way to be in this case is if the condition b s is true. As a result, $\langle \text{if b then } S_1 \text{ else } S_2, \text{ s} \rangle \Longrightarrow r$ can only have been derived using IF-TRUE; hence, $\langle S_1, \text{ s} \rangle \Longrightarrow r$. The induction hypothesis is $\forall r : \text{state}$, $\langle S_1, \text{ s} \rangle \Longrightarrow r \to 1 = r$. We can apply $\langle S_1, \text{ s} \rangle \Longrightarrow r$ to it to obtain the desired result 1 = r.

CASE IF-FALSE: Similar to IF-TRUE.

CASE WHILE-TRUE: Similar to SEQ.

CASE WHILE-FALSE: Similar to SKIP.

When reasoning about an inductive predicate, it is often convenient to use inversion rules (Section 5.3). Accordingly, we define the following rules:

```
@[simp] lemma big_step_skip_iff {s t} :
   (skip, s) \Longrightarrow t \leftrightarrow t = s :=
@[simp] lemma big_step_assign_iff {x a s t} :
   (assign x a, s) \Longrightarrow t \leftrightarrow t = s{x \mapsto a s} :=
@[simp] lemma big_step_seq_iff \{S_1 \ S_2 \ s \ t\} :
   (S_1 ;; S_2, s) \Longrightarrow t \leftrightarrow (\exists u, (S_1, s) \Longrightarrow u \land (S_2, u) \Longrightarrow t) :=
@[simp] lemma big_step_ite_iff {b S_1 S_2 s t} :
   (ite b S_1 S_2, s) \Longrightarrow t \leftrightarrow
   (b s \land (S_1, s) \Longrightarrow t) \lor (\neg b s \land (S_2, s) \Longrightarrow t) :=
lemma big_step_while_iff {b S s u} :
   (while b S, s) \Longrightarrow u \leftrightarrow
   (\exists t, b s \land (S, s) \Longrightarrow t \land (while b S, t) \Longrightarrow u)
   \lor (\neg b s \land u = s) :=
@[simp] lemma big_step_while_true_iff \{b : state \rightarrow Prop\}
      {S s u} (hcond : b s) :
   (while b S, s) \Longrightarrow u \leftrightarrow
   (\exists t, (S, s) \Longrightarrow t \land (while b S, t) \Longrightarrow u) :=
@[simp] lemma big_step_while_false_iff \{b : state \rightarrow Prop\}
      {S s t} (hcond : \neg b s) :
   (while b S, s) \Longrightarrow t \leftrightarrow t = s :=
```

We add most of them to the simp set. We leave big_step_while_true_iff out because it could make the simplifier loop due to the presence on the right-hand side of a term that matches the left-hand side.

8.5 Small-Step Semantics

A limitation of big-step semantics is that they do not let us reason about intermediate states. From a judgment $\langle \mathtt{S}, \mathtt{s} \rangle \Longrightarrow \mathtt{t}$, all we see is the initial state and the final state. This is for example too coarse-grained to reason about concurrency with an interleaving semantics. Moreover, for nondeterministic languages, big-step semantics offer no general way to express termination: A judgment indicates a possibility (executing S in state s may result in state t), not a necessity.

Small-step (operational) semantics give us a finer-coarsed view on computation. The transition relation has type $\mathtt{program} \times \mathtt{state} \to \mathtt{program} \times \mathtt{state} \to \mathtt{Prop}$. Notice the presence of $\mathtt{program}$ in the type of the second operand. The intuition is that $\langle \mathtt{S}, \mathtt{s} \rangle \Rightarrow \langle \mathtt{T}, \mathtt{t} \rangle$ indicates that executing one step of program S in state s leaves the program T to be executed, in state t. If there is no program left to be executed, we put \mathtt{skip} .

An execution is a finite or infinite chain $\langle S_0, s_0 \rangle \Rightarrow \langle S_1, s_1 \rangle \Rightarrow \cdots$. A pair $\langle S, s \rangle$ is called a *configuration*; it is *final* if no transition of the form $\langle S, s \rangle \Rightarrow$ _ is possible. Here is an example execution:

```
\langle [x \mapsto 3, y \mapsto 5], x := x + y; y := 0 \rangle

\Rightarrow \langle [x \mapsto 8, y \mapsto 5], y := 0 \rangle

\Rightarrow \langle [x \mapsto 8, y \mapsto 0], skip \rangle
```

The derivation rules for small-step semantics judgments are given below.

Unlike with the big-step semantics, there is no rule for the \mathtt{skip} statement in the small-step semantics. This is because a state of the form $\langle \mathtt{skip}, \mathtt{S} \rangle$ is now considered final; \mathtt{skip} is understood as the program whose execution is vacuous. By inspection of the rules, it is easy to see that a configuration is final if and only if its first component is \mathtt{skip} .

There are two rules for sequential composition S; T. The first rule is applicable if some progress can be made executing S. But if S is skip, no progress can be made and the second rule applies.

The rules for if check the condition and, according to that, put the "then" or the "else" branch as the remaining computation to perform.

Finally, for the while loop, there is a single unconditional rule that unfolds one iteration of the loop, introducing an if statement. It is then the role of the IF-TRUE and IF-FALSE rules to process the if. In the IF-TRUE case, we eventually reach the while loop again. This can continue like this forever for infinite loops.

In Lean, the small-step semantics is specified as follows:

```
inductive small_step : program \times state \rightarrow program \times state \rightarrow Prop | assign \{x \ a \ s\} : small_step (assign x \ a, \ s) (skip, s\{x \mapsto a \ s\}) | seq_step \{S \ T \ s \ s'\} (S') : small_step (S, s) (S', s') \rightarrow small_step (S; T, s) (S';; T, s') | seq_skip \{T \ s\} : small_step (skip;; T, s) (T, s) | ite_true \{b : state \rightarrow Prop\} \ \{S_1 \ S_2 \ s\} : b s \rightarrow small_step (ite b S_1 \ S_2, s) (S_1, s) | ite_false \{b : state \rightarrow Prop\} \ \{S_1 \ S_2 \ s\} : \neg b s \rightarrow small_step (ite b S_1 \ S_2, s) (S_2, s) | while \{b : state \rightarrow Prop\} \ \{S \ s\} : small_step (while b S, s) (ite b (S;; while b S) skip, s)
```

Equipped with a small-step semantics, we can define a big-step semantics as follows:

```
\langle \mathtt{S},\,\mathtt{s}\rangle \Longrightarrow \mathtt{t} \; \mathsf{if} \; \mathsf{and} \; \mathsf{only} \; \mathsf{if} \; \langle \mathtt{S},\,\mathtt{s}\rangle \Longrightarrow \!\!\!\! * \; \langle \mathtt{skip},\,\mathtt{t}\rangle
```

where r* denotes the reflexive transitive closure of a relation r. Alternatively, if we have already defined a big-step semantics, we can *prove* the above equivalence theorem to *validate* our definitions.

We can prove that a configuration $\langle S, s \rangle$ is final if and only if S = skip; this ensures that we have not forgotten a derivation rule.

The main disadvantage of small-step semantics is that we now have two relations, \Rightarrow and \Rightarrow *, and the derivation rules and proofs tend to be more complicated than with big steps.

8.6 Lemmas about the Small-Step Semantics

We can define inversion rules also about the small-step semantics. Here are three examples:

A more important result is the connection between the big-step and the small-step semantics. If $\Rightarrow *$ denotes the reflexive transitive closure of the small-step relation \Rightarrow , we have the following theorem:

```
lemma big_step_iff_refl_trans_small_step {S s t} : (S, s) \implies t \leftrightarrow (S, s) \implies (skip, t) :=
```

Its proof is beyond the scope of this course. We refer to Chapter 7 of Concrete Semantics: With Isabelle/HOL [17] or to the Lean sources accompanying this text.

Chapter 9

Hoare Logic

Hoare logic is due to Tony Hoare. It is also called axiomatic semantics. If operational semantics corresponds to an abstract interpreter, Hoare logic corresponds to an abstract verifier. Hoare logic can be used to specify the semantics of a programming language, but it is particularly convenient to reason about concrete programs and prove them correct.

This chapter is inspired by Chapter 12 of Concrete Semantics: With Isabelle/HOL [17], whose reading we recommend.

9.1 Hoare Triples

Hoare logic is a framework for deriving valid correctness formulas in a mechanical way, using a set of axioms and proof rules. It allows us to reason directly on the program's syntax, without concerning ourselves with the operational semantics. The approach is mechanical in the sense that the applicability of an axiom or a proof rule can easily be checked using a computer or manually.

We start by introducing Hoare logic abstractly, without any connection to Lean; later we will see how we can embed Hoare logic judgments in Lean in a natural, useful way. The basic judgments of Hoare logic are called *Hoare triples*. They have the form $\{P\}$ S $\{Q\}$, where S is a WHILE statement, and P and Q are logical formulas over the state variables. For the moment, we imagine the formulas as syntactic objects built using the familiar connectives, without being more specific. The intended meaning of a Hoare triple is as follows:

If the *precondition* P is true before S is executed and the execution terminates normally, the *postcondition* Q is true at termination.

This is a *partial correctness* statement: The program is correct *if* it terminates normally. For WHILE programs, the only way not to terminate normally is to enter an infinite loop. For other programming languages, infinite recursion and run-time errors such as division by zero may also prevent normal termination.

Based on the intuitive interpretation of Hoare triples, all of the Hoare triples below are valid:

```
 \{ \text{true} \} \ b := 4 \ \{ b = 4 \}   \{ a = 2 \} \ b := 2 * a \ \{ a = 2 \land b = 4 \}   \{ b \ge 0 \} \ b := b + 1 \ \{ b \ge 1 \}   \{ \text{false} \} \ \text{skip} \ \{ b = 10 \}   \{ \text{true} \} \ \text{while} \ i \ne 10 \ \text{do} \ i := i + 1 \ \{ i = 10 \}
```

The first three Hoare triples should be fairly intuitive. The fourth triple is vacuously true, since the precondition false can never be met. The part "If the precondition P is

true" of the definition of Hoare triple is always false; hence the triple is true. In a way, the triple is valid for the same reason that the assertion $\mathtt{false} \to \mathtt{b} = \mathtt{10}$ holds for any value of b. As for the fifth triple, there are no guarantees that control will escape the loop (e.g., if $\mathtt{i} > \mathtt{10}$ initially), but if it does escape, then the loop's condition must be false and hence we have the postcondition $\mathtt{i} = \mathtt{10}$.

9.2 Hoare Rules

The following is a complete set of rules for reasoning about WHILE programs:

$$\begin{array}{c} \hline \{P\} \text{ skip } \{P\} \\ \hline \\ \{\mathbb{Q}[a/x]\} \text{ } x := a \ \{\mathbb{Q}\} \\ \hline \\ \{\mathbb{Q}[a/x]\} \text{ } x := a \ \{\mathbb{Q}\} \\ \hline \\ \{P\} \text{ } S \in \mathbb{R}\} & \{\mathbb{R}\} \text{ } S' \in \mathbb{Q}\} \\ \hline \\ \{P\} \text{ } S \in \mathbb{R}\} & \{\mathbb{R}\} \text{ } S' \in \mathbb{Q}\} \\ \hline \\ \{P\} \text{ } S \in \mathbb{Q}\} & \{\mathbb{P} \land \neg b\} \text{ } S' \in \mathbb{Q}\} \\ \hline \\ \{P\} \text{ } if \text{ } b \text{ } then \text{ } S \text{ } else \text{ } S' \in \mathbb{Q}\} \\ \hline \\ \{\mathbb{Q}\} \text{ } f \text{ } b \text{ } b \text{ } b \text{ } s \text{$$

In the Asn rule, the syntax Q[a/x] denotes the condition Q with x replaced by a. Here is another way to look at the Asn rule:

$$\frac{}{\{Q[a]\} \times := a \{Q[x]\}} ASN$$

where [x] factors out all occurrences of x in Q.

The Asn rule may seem counterintuitive because it works backwards: From the post-condition, it computes a precondition. Nevertheless, it correctly captures the semantics of the assignment statement, as illustrated by the following examples:

$$\{0 = 0\} \ x := 0 \ \{x = 0\}$$
$$\{0 = 0 \land y = 5\} \ x := 0 \ \{x = 0 \land y = 5\}$$
$$\{x + 1 > 5\} \ x := x + 1 \ \{x > 5\}$$

Using elementary arithmetic, we can simplify the preconditions; for example, 0 = 0 is equivalent to true and $x + 1 \ge 5$ is equivalent to $x \ge 4$.

The Seq rule requires us to come up with an intermediate condition R that holds after executing S and before executing S'. Here is an example of Seq in action:

In the While rule, the condition I is called an *invariant*. It is pre- and postcondition of the loop itself but also of its body. The body's precondition is strengthened with the knowledge that b must be true before executing the body. Moreover, the loop's postcondition is strengthened with the knowledge that b must be false when the loop exits.

The Conseq rule is the only rule that has logical formulas among its premises, as opposed to only Hoare triples. These conditions must be discharged somehow—e.g., using pen and paper or a proof assistant. Conseq can be used to strengthen a precondition (i.e., make it more strict), weaken a postcondition (i.e. make it less strict), or both. An example derivation follows:

$$\frac{x > 8 \to x > 4}{\{x > 4\} \ y := x \ \{y > 4\}} ASN \ y > 4 \to y > 0$$

$$\{x > 8\} \ y := x \ \{y > 0\}$$

Reading the tree from top to bottom, we have *strengthened* the triple's precondition from x > 4 to x > 8 and *weakened* the postcondition from y > 4 to y > 0. We can also read the tree from the bottom up: To prove the triple $\{x > 8\}$ $y := x \{y > 0\}$, it suffices to prove $\{x > 4\}$ $y := x \{y > 4\}$, where the precondition is *weakened* and the postcondition is *strengthened*.

Except for Conseq, the rules are syntax-driven; that is, we know which rule to apply in each case, simply by inspecting the program at hand. If the program is, say, an assignment, we apply the Asn rule—or Conseq.

Some of the above rules are nonlinear—i.e., their conclusions are not of the form $\{P\}$... $\{Q\}$ for distinct variables P, Q. This makes them inconvenient to apply. We combine some of the previous rules with Conseq to derive *linear* rules:

$$\frac{P \rightarrow Q}{\{P\} \text{ skip } \{Q\}} \text{SKIP'}$$

$$\frac{P \rightarrow Q[a/x]}{\{P\} \text{ } x \text{ } := \text{ a } \{Q\}} \text{ASN'}$$

$$\frac{\{P \wedge b\} \text{ S } \{P\} \qquad P \wedge \neg b \rightarrow Q}{\{P\} \text{ while } b \text{ do } S \text{ } \{Q\}} \text{WHILE'}$$

As an exercise, you could try to derive each of these rules from Skip, Asn, or While in conjunction with Conseq. This is not difficult.

9.3 A Semantic Approach to Hoare Logic

A natural way to encode Hoare logic in Lean would be to proceed roughly as we have done for the big- and small-step semantics: Define a syntactic notion of Hoare triple and an inductive predicate with one constructor for each of the core Hoare rule. To represent pre- and postconditions, use predicates on states ($\mathtt{state} \to \mathtt{Prop}$). Then we could prove soundness with respect to, say, the big-step semantics, meaning the following. Assume {P} S {Q} is derivable. If P s and $\langle \mathtt{S}, \mathtt{s} \rangle \Longrightarrow \mathtt{t}$, then Q t. This is merely a logical rendition of the informal meaning of a Hoare triple:

If the *precondition* P is true before S is executed and the execution terminates normally, the *postcondition* Q is true at termination.

Instead of pursuing this approach, we propose to define Hoare triples semantically in Lean, in terms of the big-step semantics, so that they are correct by definition. Then we will derive the rules as lemmas, instead of stating them as introduction rules of an inductive predicate. In conjunction with the use of predicates to represent formulas, the approach is resolutely semantic.

Here is the definition of a Hoare triple (for partial correctness):

```
def partial_hoare (P : state \rightarrow Prop) (S : program) (Q : state \rightarrow Prop) : Prop := \foralls t, P s \rightarrow (S, s) \Longrightarrow t \rightarrow Q t
```

Instead of writing partial_hoare P S Q, we introduce some syntactic sugar to allow {* P *} S {* Q *}, which is closer to the informal syntax {P} S {Q}.

The core Hoare rules are stated as follows:

```
lemma skip_intro :
  {* P *} skip {* P *} :=
lemma assign_intro (P : state \rightarrow Prop) :
  \{* \lambda s, P (s\{n \mapsto a s\}) *\} assign n a <math>\{* P *\} :=
lemma seq_intro (h<sub>1</sub> : \{* P_1 *\} S_1 \{* P_2 *\})
     (h_2 : \{* P_2 *\} S_2 \{* P_3 *\}) :
  \{* P_1 *\} S_1 ;; S_2 \{* P_3 *\} :=
lemma ite_intro (h<sub>1</sub> : {* \lambdas, P s \wedge b s *} S<sub>1</sub> {* Q *})
     (h_2 : \{* \lambda s, P s \land \neg b s *\} S_2 \{* Q *\}) :
  {* P *} ite b S_1 S_2 {* Q *} :=
lemma while_intro (P : state \rightarrow Prop)
     (h<sub>1</sub> : {* \lambdas, P s \wedge b s *} S {* P *}) :
  \{* P *\} while b S \{* \lambda s, P s \land \neg b s *\} :=
lemma consequence (h : {* P *} S {* Q *}) (hp : \foralls, P' s \rightarrow P s)
     (hq : \foralls, Q s \rightarrow Q' s) :
  {* P' *} S {* Q' *} :=
```

They all have proofs in terms of the big-step semantics. Some of the triples—for example, the precondition in assign_intro—need to refer to the state explicitly. In such case, we can use a λ to access it. Recall that $P = (\lambda s, P s)$ by η -conversion. Moreover, for premise written informally as $P \to \mathbb{Q}$, in Lean we must write $\forall s, P s \to \mathbb{Q}$ s. This reflects the difference between syntactic formulas and semantic predicates.

The syntax $s\{n \mapsto a s\}$ in the assignment rule denotes the state that is identical to s at all points except n and that maps n to n s. This state update syntax and its automation are provided by the LoVelib library.

You might wonder, when looking at a rule such as skip, why its variable P is not declared. To lighten notation, we have declared most of the variables we need at the beginning of the Lean file:

```
variables  \{ \texttt{b} : \texttt{state} \to \texttt{Prop} \} \ \{ \texttt{a} : \texttt{state} \to \mathbb{N} \} \ \{ \texttt{n} : \texttt{string} \}   \{ \texttt{S} \ \texttt{S}_0 \ \texttt{S}_1 \ \texttt{S}_2 : \texttt{program} \} \ \{ \texttt{s} \ \texttt{s}_0 \ \texttt{s}_1 \ \texttt{s}_2 \ \texttt{t} \ \texttt{u} : \texttt{state} \}   \{ \texttt{P} \ \texttt{P'} \ \texttt{P}_1 \ \texttt{P}_2 \ \texttt{P}_3 \ \texttt{Q} \ \texttt{Q'} : \texttt{state} \to \texttt{Prop} \}
```

Lean then adds them as implicit arguments as necessary in the definitions and lemma statements that follow.

The formalization provide some more convenience rules, which can be derived from the core rules:

```
lemma consequence_left (P' : state \to Prop)
    (h : {* P *} S {* Q *}) (hp : \foralls, P' s \to P s) :
    {* P' *} S {* Q *} :=

lemma consequence_right (Q : state \to Prop)
    (h : {* P *} S {* Q *}) (hq : \foralls, Q s \to Q' s) :
    {* P *} S {* Q' *} :=

lemma skip_intro' (h : \foralls, P s \to Q s):
    {* P *} skip {* Q *} :=
```

```
lemma assign_intro' (h : \forall s, P s \rightarrow Q (s{n \mapsto a s})): {* P *} assign n a {* Q *} := lemma seq_intro' (h<sub>2</sub> : {* P<sub>2</sub> *} S<sub>2</sub> {* P<sub>3</sub> *}) (h<sub>1</sub> : {* P<sub>1</sub> *} S<sub>1</sub> {* P<sub>2</sub> *}) : {* P<sub>1</sub> *} S<sub>1</sub> ;; S<sub>2</sub> {* P<sub>3</sub> *} := lemma while_intro' (I : state \rightarrow Prop) (h<sub>1</sub> : {* \lambda s, I s \wedge b s *} S {* I *}) (hp : \forall s, P s \rightarrow I s) (hq : \forall s, \neg b s \rightarrow I s \rightarrow Q s) : {* P *} while b S {* Q *} :=
```

9.4 Example: Exchanging Two Variables

Let us use the Hoare logic to verify our first program: a program that exchanges the values of its variables a and b, using t for temporary storage. The program is defined as follows:

```
def SWAP : program := assign "t" (\lambdas, s "a") ;; assign "a" (\lambdas, s "b") ;; assign "b" (\lambdas, s "t")

The proof is as follows:

lemma SWAP_correct (x y : \mathbb{N}) :
 {* \lambdas, s "a" = x \wedge s "b" = y *} SWAP {* \lambdas, s "a" = y \wedge s "b" = x *} := begin
 apply seq_intro', apply seq_intro', apply assign_intro, apply assign_intro, apply assign_intro', simp {contextual := tt},
```

end

The applications of seq_intro' and assign_intro are syntax-directed: There are two sequential compositions and three assignments in the program and therefore as many invocations of the corresponding rules. We end up with a very ugly subgoal:

```
 \forall s : state, \\ s "a" = x \land s "b" = y \rightarrow \\ s{"t"} \mapsto s "a"}{"a"} \mapsto s{"t"} \mapsto s "a"} "b"}{"b"}{"b"} \mapsto s{"t"} \mapsto s "a"}{"a} \\ " \mapsto s{"t"} \mapsto s "a"} "b"} "t"} "a" = y \land \\ s{"t"} \mapsto s "a"}{"a"} \mapsto s{"t"} \mapsto s "a"}{"b"}{"b"}{"b"} \mapsto s{"t"} \mapsto s "a"}{"a} \\ " \mapsto s{"t"} \mapsto s "a"} "b"} "t"} "b" = x
```

Fortunately, simp can simplify the subgoal dramatically, and if we enable contextual rewriting—an option that tell simp to use the assumptions in the goal as rewrite rules—simp can even solve the subgoal.

9.5 Example: Adding Two Numbers

Our second example computes m + n, leaving the result in n, using only these primitive operations: k + 1, k - 1, and $k \neq 0$ (for arbitrary k):

```
def ADD : program :=
  while (\lambdas, s "n" \neq 0)
  (assign "n" (\lambdas, s "n" - 1) ;;
   assign "m" (\lambdas, s "m" + 1))
The proof is more involved:
  lemma ADD_correct (n_0 m_0 : \mathbb{N}):
    \{* \lambda s, s "n" = n_0 \land s "m" = m_0 *\} ADD
    \{* \lambda s, s "n" = 0 \land s "m" = n_0 + m_0 *\} :=
  begin
    refine while_intro' (\lambdas, s "n" + s "m" = n_0 + m_0) _ _ _ _,
    apply seq_intro',
    apply assign_intro,
    apply assign_intro',
    { simp,
       -- this looks much better: 'simp' removed all updates
      intros s hnm hn0.
      rw \leftarrow hnm,
       -- subtracting on 'N' is annoying
      cases s "n",
      { apply false.elim, apply hn0, refl },
      { simp [nat.succ_eq_add_one] } },
    { simp {contextual := tt} },
    { simp [not_not_iff] {contextual := tt} }
  end
```

The first step is to apply the derived while rule with a loop invariant. Our invariant is that the sum of the program variables n and m must be equal to the desired mathematical result $n_0 + m_0$, where n_0 and m_0 correspond to the initial values of the program variables, as required by the precondition.

How did we come up with this invariant? Conceptually, most invariants have the form

work already done + work remaining to be done = desired result

where + in general denotes some suitable operator and not necessarily addition. In our example, the "work already done" is m, the "work remaining to be done" is n, and the "desired result" is $n_0 + m_0$. When exiting the loop, we have that n = 0, so the equation simplifies to "work already done = desired result." We can retrieve the desired result from m.

The application of while_intro' is performed using a tactic called refine, which works like exact but lets us specify holes to be filled later.

After applying the obvious rules on the loop body, we are left with three subgoals. The only nontrivial subgoal is the proof that executing the body maintains the invariant. There is some ugly arithmetic on natural numbers necessary due to the use of the minus operator. Our proof simply performs a case distinction to separate the zero case, where -1 has no effect (0 -1 = 0), and the nonzero case, where -1 has an effect. Each case is easy to discharge using the simplifier with contextual rewriting.

9.6 A Verification Condition Generator

Verification condition generators (VCGs) are programs that apply Hoare logic rules automatically, producing verification conditions that must be proved by the user. The user must provide strong enough loop invariants, as annotations in their programs. Hundreds of program verification tools are based on these principles. Often these are based on an

extension called separation logic,¹ which is sometimes called "the Hoare logic of the 21st century."

VCGs typically work backwards from the postcondition, using backward rules—i.e., rules stated to have an arbitrary Q as their postcondition; this works well because the key rule Asn, for assignment, is backward.

We can use Lean's metaprogramming framework to define a simple VCG. First, we introduce a constant called while_inv that carries an invariant I in addition to the loop condition b and body S:

```
def program.while_inv (I : state → Prop) (c : state → Prop)
  (p : program) : program :=
while c p
export program (while_inv)
```

We provide two Hoare rules for the construct: a backward rule and a linear rule. Both are justified in terms of the while_intro' rule we presented earlier:

The rules simply use the invariant annotation as their invariant. When using the framework, we will have to be careful to provide a suitable invariant, because there will not be a second chance to provide one.

The code of the VCG is fairly concise:

```
meta def is_meta : expr \rightarrow bool
| (expr.mvar _ _ _) := tt
                    := ff
meta def vcg : tactic unit := do
 t \leftarrow target,
 match t with
  | '({* \%P *} \%S {* _ *}) :=
    match S with
    | '(program.skip)
      applyc (if is_meta P then ''skip_intro else ''skip_intro')
    | '(program.assign _ _)
      applyc (if is_meta P then ''assign_intro
        else ''assign_intro')
    | '(program.seq _ _)
      do applyc ''seq_intro'; vcg
    | '(program.ite _ _ _)
      do applyc ''ite_intro; vcg
    | '(program.while_inv _ _ _) :=
      do applyc (if is_meta P then ''while_inv_intro
        else ''while_inv_intro');
```

¹https://en.wikipedia.org/wiki/Separation_logic

```
vcg
| _ :=
   fail (to_fmt "cannot analyze " ++ to_fmt S)
   end
| _ := skip
end
```

The is_meta helper function determines whether a Lean term is a metavariable (i.e., a ? variable such as ?P).

The VCG proper extracts the current goal—the target—and inspects it. If it is a Hoare triple, it inspects its precondition P and statement S. If the precondition is a metavariable, then it applies a backward rule if there exists one (e.g., the standard assignment rule assign_intro) because this will usefully instantiate the metavariable. Otherwise, a linear rule is used, with an arbitrary variable as its precondition, which can be matched against the goal's precondition. For while loops, we only consider programs that use while_inv, because we cannot guess the invariant programmatically.

The VCG calls itself recursively on all newly emerging goals (via the ; tactic combinator) for the compound statements seq, ite, and while_inv.

9.7 Example: Adding Two Numbers, Revisited

Using the verification condition generator, we can revisit the proof of ADD presented above:

```
lemma ADD_correct<sub>2</sub> (n_0 m_0 : \mathbb{N}):
  \{* \lambda s, s "n" = n_0 \land s "m" = m_0 *\} ADD
  \{* \lambda s, s "n" = 0 \wedge s "m" = n_0 + m_0 *\} :=
begin
  show \{* \lambda s, s "n" = n_0 \wedge s "m" = m_0 *\}
       while_inv (\lambdas, s "n" + s "m" = n<sub>0</sub> + m<sub>0</sub>) (\lambdas, s "n" \neq 0)
        (assign "n" (\lambdas, s "n" - 1) ;;
         assign "m" (\lambdas, s "m" + 1))
     \{* \lambda s, s "n" = 0 \wedge s "m" = n_0 + m_0 *\},
     simp {contextual := tt},
  intros s hnm hn,
  rw \leftarrow hnm,
  cases s "n",
  { apply false.elim, apply hn, refl },
  { simp [nat.succ_eq_add_one] }
end
```

There are two main points to observe.

First, we use show to annotate the while loop with an invariant. Recall that show allows us to restate the goal in a computationally equivalent way. Here, we use this facility to replace while by while_inv, whose definition is in terms of while. The program and its pre- and postconditions are otherwise the same as in the lemma statement.

Second, we invoke vcg as the first proof step. This will apply all the necessary Hoare rules and leave us with some proof goals. We first simplify them with contextual rewriting enabled; this already takes care of two of the three goals. For the third goal, the proof is as before.

9.8 Hoare Triples for Total Correctness

The focus so far in this chapter has been on partial correctness. When we state the Hoare triple {P} S {Q}, we claim that the final state will satisfy Q if the program S terminates, but

we say nothing when S does not terminate. In particular, we can prove any postcondition for the program while true do skip. This is admittedly too liberal: If we are asked to provide, say, a sorting algorithm, we should not simply give while true do skip as the answer.

Total correctness is a stronger notion asserts that the program terminates normally. One reason to focus on partial correctness is that it is a necessary component of total correctness. It is also simpler, which is a benefit in the classroom.

The Hoare triples for total correctness have the form [P] S [Q] with the following intended meaning:

If the precondition P holds before S is executed, the execution terminates normally and the postcondition Q holds in the final state.

For deterministic programs, the following is an equivalent formulation:

If the precondition P holds before S is executed, there exists a state in which execution terminates normally and the postcondition Q holds in that state.

Here is an example Hoare triple that is valid:

[
$$i \le 10$$
] while $i \ne 10$ do $i := i + 1$ [$i = 10$]

For the WHILE language, the distinction between partial and total correctness only concerns while loops. The Hoare rule for while must now be annotated by a *variant* V—a natural number that decreases with each iteration:

$$\frac{[\text{I} \land \text{b} \land \text{V} = \text{n}] \text{ S } [\text{I} \land \text{V} < \text{n}]}{[\text{I}] \text{ while b do S } [\text{I} \land \neg \text{b}]} \text{WHILE}$$

The variable n is a logical variable whose scope is the entire premise, whereas V is a metavariable standing for the variant. For the example above, we would take 10 - i as the variant. A derivation tree for this example is given on Wikipedia.²

9.9 Summary of New Syntax

Tactics

refine applies a rule that may contain holes (_) left as subgoals simp... {contextual := tt} simplifies using assumptions in the goal as rewrite rules

²https://en.wikipedia.org/wiki/Hoare_logic#While_rule_for_total_correctness

Chapter 10

Denotational Semantics

We review a third way to specify the semantics of a programming language: denotational semantics. Denotational semantics attempts to directly specify the semantics of programs as a mathematical object. If operational semantics corresponds to an abstract interpreter and Hoare logic corresponds to an abstract verifier, then denotational semantics corresponds to an abstract compiler.

This chapter is inspired by Chapter 11 of Concrete Semantics: With Isabelle/HOL [17], whose reading is recommended.

10.1 Motivation

A denotational semantics defines the meaning of each program as a mathematical object and can be seen as a function

```
\llbracket \ \rrbracket : \mathtt{syntax} 	o \mathtt{semantics}
```

A key property of denotational semantics is *compositionality*: The meaning of a compound statement should be defined in terms of the meaning of its components. This compositionality makes it possible to reason equationally about programs. The compositionality requirement disqualifies the trivial definition

$$[S] = \{ st \mid \langle S, st.1 \rangle \Longrightarrow st.2 \}$$

in terms of the big-step semantics \implies because that semantics is not compositional: The semantics of a while loop is given in terms of itself, not of its loop body.

In essence, we want structurally recursive equations of the form

An evaluation function such as $eval: aexp \to ((var \to \mathbb{Z}) \to \mathbb{Z})$ on arithmetic expressions is a denotational semantics. Now we want the same for imperative programs. We need to develop some mathematical notions to reach the point where we can formulate the denotational semantics.

10.2 A Relational Denotational Semantics

Denotational semantics for deterministic languages are normally given as a function from prestate to poststate, but we find relations more convenient to manipulate. Hence we present a relational denotational semantics, following Concrete Semantics [17]. The relational approach was also used for the big-step semantics, which took the form of a

predicate of type state \rightarrow state \rightarrow Prop. In contrast, a denotational semantics of a program will be a mathematical object of type set (state \times state).

To a mathematician working informally, $\mathtt{state} \to \mathtt{state} \to \mathtt{Prop}$ and \mathtt{set} ($\mathtt{state} \times \mathtt{state}$) are "the same," since there is a one-to-one correspondence between predicates on pairs and subsets of the cartesian product. To a type theorist, $\mathtt{state} \to \mathtt{state} \to \mathtt{Prop}$ and \mathtt{set} ($\mathtt{state} \times \mathtt{state}$) are "the same" up to currying, since \mathtt{set} $\alpha = \alpha \to \mathtt{Prop}$ by definition. So why make this distinction? One reason is the different perspective afforded by the different structures. Another is for efficiency of notation or of computation. The final reason is that many systems of formal logic enforce a strict separation between predicates and their arguments, making it formally nonsensical to even ask whether $\mathtt{state} \to \mathtt{state} \to \mathtt{Prop}$ and \mathtt{set} ($\mathtt{state} \times \mathtt{state}$) are "the same."

We call our denotational semantics den, with [] as syntactic sugar. Let us first study the first four equations of the definition, keeping the while case for later:

The skip statement is interpreted as the identity relation over states—i.e., the set of all tuples of the form (s, s). This captures the desired semantics of skip: The poststate is always identical to the prestate.

The semantics of assignment is the set of tuples where the second component reflects the result of the assignment. Recall that if x is a pair, then x.1 and x.2 give access to its first and second components.

The semantics of sequential composition is elegantly expressed as a relational composition \bigcirc , which is defined such that $r_1 \bigcirc r_2 = \{x \mid \exists y, (x.1, y) \in r_1 \land (y, x.2) \in r_2\}.$

The semantics of ite is the union of the behaviors of the two branches, restricted to include only the tuples whose first component makes the Boolean condition true or false, according to the branch. The restriction operator is defined as $r \mid p = \{x \mid p x. 1 \land x \in r\}$.

The difficulties arise when we try to define the while case. We would like to write

```
| (while b S) := ((den S \bigcirc den (while b S)) \downarrow b) \cup (Id state \downarrow \lambdas, \neg b s))
```

but this is not well founded due to the unmodified recursive call to while b S. So we need to write this differently. What we are looking for is a solution for X in the equation

```
X = ((\text{den } S \bigcirc X) \mid b) \cup (\text{Id state } \mid \lambda s, \neg b s))
```

In other words, we are looking for a fixpoint. The rest of this chapter is concerned with building a fixpoint operator lfp that will allow us to define the while case as well:

```
| (while b S) := lfp (\lambdaX, ((den S \bigcirc X) \downarrow b) \cup (Id state \downarrow \lambdas, \neg b s))
```

The union and restriction operators are as for the ite case.

10.3 Fixpoints

A fixpoint (or fixed point) of f is a solution for X in the equation

```
X = f X
```

In general, fixpoints may not exist at all for some f (e.g., f := nat.succ), or there may be many fixpoints (e.g., $f := \lambda x$, x). Under some conditions on f, a unique *least fixpoint* and a unique *greatest fixpoint* are guaranteed to exist.

Consider the following fixpoint equation:

$$X = \lambda n : \mathbb{N}, n = 0 \vee \exists m : \mathbb{N}, n = m + 2 \wedge X m$$

This equation is the β -contracted version of an equation that has more obviously the right format, with $X: \mathbb{N} \to \operatorname{Prop}$ and $f:=(\lambda(p:\mathbb{N} \to \operatorname{Prop})\ (n:\mathbb{N}),\ n=0 \lor \exists m:\mathbb{N},\ n=m+2\land p\ m)$:

$$X = (\lambda(p : \mathbb{N} \to Prop) \ (n : \mathbb{N}), \ n = 0 \lor \exists m : \mathbb{N}, \ n = m + 2 \land p \ m) \ X$$

It turns out that this equation admits only one fixpoint. The fixpoint equation uniquely specifies X as the set of even natural numbers.

In general, the least and greatest fixpoint may be different. Consider the equation

$$X = (\lambda x, x) X$$

where X : $\mathbb{N} \to \operatorname{Prop}$. The least fixpoint of this equation is λ_- , False and the greatest fixpoint is λ_- , True. Conventionally, we have that False < True and thus $(\lambda_-, \operatorname{False}) < (\lambda_-, \operatorname{True})$. Similarly, $\emptyset < \operatorname{@set.univ} \alpha$ if α is inhabited (i.e., if it contains at least one element).

Incidentally, Lean's inductive predicates correspond to least fixpoints, but they are built into Lean's calculus of inductive constructions. We implicitly made use of this observation when we defined the operational semantics as an inductive data type.

For the semantics of programming languages:

- X will have type set (state × state), representing relations between states;
- f will correspond to either taking one extra iteration of the loop (if the condition b is true) or the identity (if b is false).

Which fixpoint should we use for the semantics of while? A program should be prohibited from looping infinitely: Each time a loop is evaluated, it must terminate after some number of iterations n. Let f^n denote the n-fold composition of f, corresponding to performing at most f iterations, then we have $f^n(\emptyset)$ in this case. More generally, we always have

$$X = f^{0}(\emptyset) \cup f^{1}(\emptyset) \cup f^{2}(\emptyset) \cup \cdots$$

Kleene's fixpoint theorem¹ states that the least fixpoint lfp f will produce the value of X satisfying this equation. The greatest fixpoint would also consider cyclic and diverging executions.

10.4 Monotone Functions

Let α and β be arbitrary types, each equipped with a partial order \leq . We say that the function $f: \alpha \to \beta$ is monotone if $a \leq b \to f$ a $\leq f$ b for all a, b.

Many operations on sets (e.g., \cup), relations (e.g., \bigcirc), and functions (e.g., const) are monotone. The set of monotone functions is also well behaved: The identity function is monotone, the composition of monotone functions is again monotone, and so on.

All monotone functions $f: \alpha \to \alpha$, where α is a "complete lattice" (introduced below), admit least and greatest fixpoints.

Here is an example of a nonmonotone function on sets:

$$f s = \begin{cases} s \cup \{a\} & \text{if } a \notin s \\ \emptyset & \text{otherwise} \end{cases}$$

Notice that $\emptyset \subseteq \{a\}$ but $f \emptyset = \{a\} \not\subseteq \emptyset = f \{a\}$.

¹https://en.wikipedia.org/wiki/Kleene_fixed-point_theorem

10.5 Complete Lattices

To define the least fixpoint for sets, we only need one operation: intersection \cap . The same definition works in any instance of the algebraic structure called a complete lattice.

A complete lattice α is an ordered type for which each set α has an infimum, also called *greatest lower bound*. A complete lattice consists of

- a partial order \leq : $\alpha \to \alpha \to \texttt{Prop}$ (i.e., reflexive, transitive, and antisymmetric);
- an operator \square : set $\alpha \to \alpha$, called *infimum*.

The \square s operator satisfies the following two conditions:

- $\sqcap s$ is a lower bound of s: $\sqcap s \leq b$ for all $b \in s$.
- \sqcap s is the greatest lower bound: $b \leq \sqcap$ s for all b such that $\forall x \in s$, $b \leq x$.

We are justified in writing "the greatest lower bound": If α is a partial order, for any $s:set \alpha$ there is at most one $a:\alpha$ satisfying both conditions on the infimum.

Warning: $\prod s$ is not necessarily an element of s. For example, an open interval a, b[has infimum a, which is not in a, b[.

Here are some examples of complete lattices:

```
• enat := \mathbb{N} \cup \{\infty\};
• ereal := \mathbb{R} \cup \{-\infty, \infty\};
```

- set α is a complete lattice with respect to \subseteq and \bigcap for all types α ;
- Prop is a complete lattice with respect to \rightarrow and \forall , i.e., $\Box s := \forall p \in s$, p.

If α is a complete lattice, then $\beta \to \alpha$ is also a complete lattice. If α and β are complete lattices, then $\alpha \times \beta$ is also a complete lattice. In both cases, \leq and \square are defined componentwise.

Here are some *non*examples of complete lattices: \mathbb{N} , \mathbb{Z} , \mathbb{Q} , \mathbb{R} . The issue is that there is no largest element, which the infimum for \emptyset must be. Another nonexample: erat := \mathbb{Q} \cup $\{-\infty, \infty\}$. The issue is that $\prod \{q \mid 2 < q * q\} = \operatorname{sgrt} 2$ is not in \mathbb{Q} .

Finally, we define

$$lfp f := \prod \{x \mid f x < x\}$$

The Knaster–Tarski theorem² gives us the following properties for any monotone function f on a complete lattice:

- lfp f is a fixpoint: lfp f = f (lfp f);
- 1fp f is smaller than any other fixpoint: $X = f X \rightarrow 1fp f \leq X$.

With lfp, we can complete the definition of the denotational semantics, as announced:

```
| (while b S) := lfp (\lambdaX, ((den S \bigcirc X) \downarrow b) \cup (Id state \downarrow \lambdas, \neg b s))
```

10.6 A Relational Denotational Semantics, Continued

The proof of the pudding is in the eating. We defined the semantics of while in terms of the lfp operator, but perhaps monotonicity—which guarantees the existence of a least fixpoint—is not met. To quell such doubts, we prove the lemma

```
lemma den_while (S : program) (b : state → Prop) :
    [while b S] = [ite b (S ;; while b S) skip] :=
begin
    apply lfp_eq,
    apply monotone.union,
    { apply sorry_lemmas.monotone_restrict,
        apply sorry_lemmas.monotone_comp,
```

²https://en.wikipedia.org/wiki/Knaster-Tarski_theorem

The lemma gives us a convenient way to unfold one iteration of a loop. The first proof step expands the least fixpoint, exploiting definitional equality to recognize the semantics of an ite hiding in the semantics of a while. The rest of the proof is all about proving monotonicity of the body of the fixpoint operator.

Another important result is the connection between the denotational and the big-step semantics:

```
lemma den_eq_bigstep (S : program) (s t : state) : (s, t) \in [S] \leftrightarrow (S, s) \Longrightarrow t :=
```

Its proof is beyond the scope of this course. We refer to Chapter 11 of *Concrete Semantics:* With Isabelle/HOL [17] or to the Lean files accompanying this book.

Part IV Mathematics

Chapter 11

Logical Foundations of Mathematics

In this chapter, we dive deeper into the logical foundations of Lean.

11.1 Type Universes

In type theory—and in particular in Lean—not only all terms have a type, but also all types have a type themselves. We have already seen one occurrence of this principle. The Curry–Howard correspondence tells us to view proofs as terms and propositions as types. For example:

```
@and.intro : \forall a b, a \rightarrow b \rightarrow a \wedge b
```

Thus, a proposition such as $\forall a \ b$, $a \to b \to a \land b$ is a type, and this type has in turn a type, namely Prop:

$$\forall a b, a \rightarrow b \rightarrow a \wedge b : Prop$$

What is the type of Prop then? Prop has the same type as virtually all other types we have constructed so far, namely, Type:

```
\mathbb{N}: Type Prop: Type
```

What is the type of Type? It may seem like the simplest solution to assign Type: Type, but this choice leads to Girard's paradox, the type theory equivalent of Russel's paradox. To avoid inconsistencies, we need a fresh type to contain Type, which we call Type 1. Type 1 again cannot be assigned any of the other types introduced so far without causing contradictions, leading to the introduction of Type 2, and so on:

```
Type: Type 1
Type 1: Type 2
Type 2: Type 3
:
```

In fact, Type is an abbreviation for Type 0. If we want to incorporate Prop in this hierarchy, we can also use the syntax Sort u, where Sort 0 is the same as Prop and Sort (u+1) is the same as Type u.

All of these types containing other types are called (type) universes, and the u in the expression Sort u is a universe level. Although universe levels look like terms of type $\mathbb N$, they are in fact neither of type $\mathbb N$ nor even terms.

To formalize most of computer science and mathematics, Type (i.e., Type 0) suffices. Instead of using Type everywhere though, you can make your theorems slightly more general without having to think about universe levels by writing Type*, which abbreviates

Type _. It leaves it up to Lean figure out the right universe level or to create a fresh universe variable.

11.2 The Peculiarities of Prop

Although Prop seems to fit nicely into the hierarchy of type universes, it is different from the other universes in many respects.

Impredicativity

When constructing a new type from other types (e.g., $\alpha \to \beta$ from α : Type u and β : Type v), the newly constructed type is more complex than each of its components, and it is natural to put it into the largest type universe involved (e.g., $\alpha \to \beta$: Type (max u v)). This is exactly what Lean does. The following typing rule expresses this idea generally for dependent types:

$$\frac{\vdash \alpha : \text{Type u} \qquad (\texttt{a} : \alpha) \vdash \beta \texttt{ a} : \text{Type v}}{\vdash (\Pi \texttt{a} : \alpha, \beta \texttt{ a}) : \text{Type (max u v)}} \mathsf{P}_{\mathsf{I}}$$

This behavior of the universes Type v is called *predicativity*.

However, it is convenient to have Prop behave differently. We would like $\forall (a: Prop)$, $a \to a$ to be of type Prop. Unfolding the syntactic sugar \forall and \to , this expression is the same as $\Pi(a: Prop)$ (x:a), a. If we had Type u instead of Prop, the above typing rule would yield

$$(\Pi(a: Type u) (x:a), a): Type (u+1)$$

because Type u: Type (u+1) and max (u+1) $(max\ u\ u)=u+1$. Thus the universe level is increased by one when typing this expression. To force expressions such as $\forall (a: \texttt{Prop})$, $a \to a$ (which equals $\Pi(a: \texttt{Prop})$ (x:a), a) to be of type Prop anyway, we have a special typing rule for Π -expressions involving Prop:

$$\frac{\vdash \alpha : \mathtt{Sort} \ \mathtt{u} \qquad (\mathtt{a} : \alpha) \vdash \beta \ \mathtt{a} : \mathtt{Prop}}{\vdash (\Pi \mathtt{a} : \alpha, \beta \ \mathtt{a}) : \mathtt{Prop}} \ \mathsf{P}_{\mathsf{I}}$$

This is called *predicativity* of Prop. The rule yields

$$(\Pi(a: Prop) (x:a), a): Prop$$

as desired. The two typing rules for Π -expressions above can be summarized in one rule

$$\frac{\Gamma \vdash \alpha : \mathtt{Sort} \ \mathsf{u} \qquad \Gamma, \ \mathsf{a} : \alpha \vdash \beta \ \mathsf{a} : \mathtt{Sort} \ \mathsf{v}}{\Gamma \vdash (\Pi \mathsf{a} : \alpha, \beta \ \mathsf{a}) : \mathtt{Sort} \ (\mathtt{imax} \ \mathsf{u} \ \mathsf{v})} \mathsf{P}_{\mathsf{I}}$$

where

```
imax u 0 = 0

imax u v = max u v if v \neq 0
```

In other systems, such as Agda, all type universes are predicative.

Proof Irrelevance

A second difference between Prop and Type is proof irrelevance. It means that any two proofs of the same proposition are equal:

```
lemma proof_irrel {a : Prop} (h_1 \ h_2 : a) : h_1 = h_2 := by refl
```

In Lean, this equality is even a definitional equality, allowing us to prove that any two proofs of the same proposition are equal using the refl tactic. When viewing a proposition as a type and a proof as an element of that type, proof irrelevance means that a proposition is either an empty type or has exactly one inhabitant. If it is empty, it is false. If it has exactly one inhabitant, it is true. Proof irrelevance is very helpful when reasoning about dependent types.

Other systems and logics make different choices in this matter. For example, Agda and Coq are proof relevant by default, but are compatible with proof irrelevance. Homotopy type theory and other *constructive* or *intuitionistic* type theories build on data in equality proofs and are therefore incompatible with proof irrelevance.

Small Elimination

A further difference between Prop and Type is that Prop does not allow *large elimination*, meaning that it is impossible to extract data from an element of Prop. Consider the following inductive type:

```
inductive square_roots (sq : \mathbb{Z}) : Type u | mk (z : \mathbb{Z}) (h : sq = z * z) : square_roots
```

The type $square_roots\ sq$ contains all square roots of a given integer sq. It contains no elements if sq is not square, one element if sq = 0, and two elements otherwise. For example:

```
def two_as_root : square_roots 4 :=
square_roots.mk 2 (by refl)

def neg_two_as_root : square_roots 4 :=
square_roots.mk (-2) (by refl)
```

A large eliminator can be defined as follows:

```
def square_roots.large_elim {sq : \mathbb{Z}} {\alpha : Sort v} : square_roots sq \rightarrow (\forall(z : \mathbb{Z}) (h : sq = z * z), \alpha) \rightarrow \alpha | (square_roots.mk z h) f := f z h
```

This eliminator's first explicit argument an element of square_roots sq. The second one is a function that converts a number and a proof that sq is the square of that number into an element of an arbitrary type α . Given those two things, it can produce the element of α corresponding to the given element of square_roots sq. For example, it can be used to extract the number that was used to construct the given element of square_roots:

Therefore, the large eliminator can be used to prove that two elements of square_roots are different if they were constructed using different roots:

```
example : two_as_root \neq neg_two_as_root := begin assume h : two_as_root = neg_two_as_root, have h : (2 : \mathbb{Z}) = -2 := congr_arg (\lambdahsq, square_roots.large_elim hsq (\lambdaz _, z)) h, show false, by linarith
```

(This proof could be written more elegantly using the injection tactic.)

Moreover, the large eliminator can be used to show that all numbers with a square root are nonnegative:

```
example (sq : \mathbb{Z}) (z : square_roots sq) : sq \geq 0 := begin apply z.large_elim, intros z hz, rw hz, apply mul_self_nonneg end
```

(This proof could be written more elegantly using the cases tactic.)

Now we make a second definition that strongly resembles square_roots. The only difference is that we define it as an inductive predicate, i.e., we put into the Prop universe instead of Type u.

```
inductive square_number (sq : Z) : Prop
| mk (z : Z) (h : sq = z * z) : square_number
Again, we can show that 4 is square using either 2 or -2:
lemma four_is_square : square_number 4 :=
square_number.mk 2 (by ref1)

lemma four_is_square' : square_number 4 :=
square_number.mk (-2) (by ref1)
```

The following lemma is one that we can only prove for square_number, even though we cannot prove the corresponding lemma for square_roots. We can show that the two definitions above are equal, as a consequence of proof irrelevance:

```
example : four_is_square = four_is_square' :=
by apply proof_irrel
```

Proof irrelevance forces the type square_number 4 to contain only one element. Therefore, this element cannot contain the information as to whether it was constructed using 2 or -2. This is why large elimination on square_number would lead to an inconsistency and hence Lean will not allow us to define a large eliminator:

```
def square_number.large_elim {sq : \mathbb{Z}} {\alpha : Sort v} : square_number sq \rightarrow (\forall(z : \mathbb{Z}) (h : sq = z * z), \alpha) \rightarrow \alpha | (square_number.mk z h) f := f z h -- induction tactic failed, recursor -- 'LoVe.square_number.dcases_on' can only eliminate into Prop
```

But without being able to eliminate the square_number predicate at all, it would be quite useless. As a compromise, Lean allows us to define a small eliminator:

```
def square_number.small_elim {sq : \mathbb{Z}} {\alpha : Prop} : square_number sq \rightarrow (\forall(z : \mathbb{Z}) (h : sq = z * z), \alpha) \rightarrow \alpha | (square_number.mk z h) f := f z h
```

It is called a small eliminator because it eliminates only into Prop, whereas a large eliminator can eliminate into an arbitrary large type universe Sort v. This eliminator cannot be used to prove that four_is_square and four_is_square, are different, but it can be used to prove that all square numbers are nonnegative, just like we could with the large eliminator:

```
example (sq : \mathbb{Z}) (h : square_number sq) : sq \geq 0 := begin apply h.small_elim, intros z hz, rw hz, apply mul_self_nonneg end
```

11.3 The Axiom of Choice

To understand the axiom of choice, it will help to first take a look at the following inductive predicate:

```
inductive nonempty (\alpha : Sort u) : Prop | intro (val : \alpha) : nonempty
```

The predicate states that the given type α has at least one element. To prove nonempty α , we must provide an element of α to the intro rule:

```
example : nonempty \mathbb{N} := nonempty.intro 0
```

Since nonempty lives in Prop, large elimination is not available, and thus we cannot extract the element that was used from a proof of nonempty α .

The axiom of choice allows us to obtain some element of type α if we can show nonempty α :

```
axiom classical.choice \{\alpha: \text{Sort u}\}: \text{nonempty } \alpha \to \alpha
```

We have no way of knowing whether this is the element that was used to prove nonempty α . It will just give us an arbitrary element of α , thereby avoiding inconsistencies.

The constant classical.choice is noncomputable. If we ask Lean for its value using #reduce or #eval, it will refuse to compute it. This is one of the reasons why some logicians prefer to work without this axiom. Unlike proof irrelevance and the eliminators, this principle is not built into the Lean kernel; it is only an axiom in Lean's core library, giving users the freedom to work with or without it. Lean requires us to mark definitions as noncomputable if they use choice to define constants outside of Prop.

The following tools rely on choice:

• The function classical.some can help us find a witness of $\exists \mathtt{a}:\alpha$, p a, if we do not care which one.

```
classical.some : (\existsa : \alpha, p a) \rightarrow \alpha classical.some_spec (h : \existsa : \alpha, p a) : p (some h)
```

• We can also derive the traditional axiom of choice:

```
classical.axiom_of_choice : \forall \alpha \ \beta \ \{r : \alpha \rightarrow \beta \rightarrow Prop\}, (\forall x : \alpha, \exists y : \beta, r x y) \rightarrow (\exists f, \forall x, r x (f x))
```

• The choose tactic provides an elegant way to use the set theoretic version of the axiom of choice in proofs:

```
example (x : list \mathbb{N})

(h : \forall y : list \mathbb{N}, \exists z : list \mathbb{N}, [0] ++ z = x ++ y) : x.head = 0 := begin choose z h using h, have hx : x = [0] ++ z [], by simp [h []], simp [hx] end
```

• Using choice and two other principles called propositional extensionality and functional extensionality, the law of excluded middle can be derived:

```
classical.em : \forall p : Prop, p \lor \neg p
```

With the law of excluded middle, every proposition is decidable. This means that we can construct proofs based on a case distinction on whether a certain proposition is true. The by_cases tactic offers a convenient interface for such proofs. To be able to use it with any proposition, add the line

local attribute [instance, priority 0] classical.prop_decidable at the top of your file. Then you can use the by_cases tactic as follows:

```
example (p : Prop) : p V ¬ p :=
begin
  by_cases h : p,
  { exact or.intro_left (¬ p) h, },
  { exact or.intro_right p h, },
end
```

In this example, the line by_cases h:p introduces a case distinction on whether p is true. It splits the goal into two subgoals, one with the additional assumption h:p, the other with the additional assumption h:p.

11.4 Subtypes

Subtyping is a mechanism to create new types from existing ones. Given a predicate on the elements of the original type, the new type contains only those elements of the original type that fulfill this property.

To illustrate this, we will define a type of full binary trees, building on the example from Section 4.3. In that section, we defined an inductive type btree, representing binary trees. In Section 5.4, we introduced a predicate is_full that is true if each node of a tree has either zero or two child nodes. Based on this type and this predicate, we can construct a subtype full_btree, containing only full binary trees, as follows:

```
def full_btree (\alpha : Type) := { t : btree \alpha // is_full t }
```

This is syntactic sugar for

```
\begin{tabular}{ll} $\operatorname{def full\_btree} & (\alpha : \operatorname{Type}) := \\ & (\operatorname{Subtype} & (\operatorname{btree} & \alpha) \end{tabular} is full \\ \end{tabular}
```

where subtype is defined as an inductive predicate:

```
inductive subtype \{\alpha: {\tt Type}\}\ ({\tt p}:\alpha\to{\tt Prop}): {\tt Type}\ |\ {\tt mk}:\Pi{\tt x}:\alpha , p x \to subtype
```

Thus, every element of full_btree consists of a binary tree t and a proof that t is full:

The elements can also be written using the angle bracket notation:

```
def some_full_btree': full_btree \mathbb{N} := \langle empty, is\_full.empty \rangle def another_full_btree': full_btree \mathbb{N} := \langle node \ 6 \ empty \ empty, is_full.node 6 empty empty is_full.empty is_full.empty (by ref1)\rangle
```

11.4. Subtypes 105

From a given element of full_btree, we can retrieve its two components via subtype.val and subtype.property. For example, another_full_btree.val gives us the tree node 6 empty empty and another_full_btree.property gives us a proof that this binary tree is full (i.e., is_full (another_full_btree.val)).

We can transfer functions from the base type btree to the subtype full_btree if they preserve the property is_full. As we have shown in the lemma is_full_mirror in Section 5.4, mirror does preserve the property is_full. We can use that lemma to define a function full_btree.mirror operating on our new subtype as follows:

```
def full_btree.mirror \{\alpha: \text{Type}\}\ (\text{t}: \text{full_btree}\ \alpha): full_btree}\ \alpha:= \text{subtype.mk}\ (\text{mirror t.val})\ (\text{is_full_mirror t.val}\ \text{t.property})
```

The input of the function is an element t of our subtype full_btree. We decompose it into the binary tree t.val and the proof of being full t.property. We use the mirror function to mirror the tree component and the lemma is_full_mirror to transform the proof component into a proof that the mirrored tree is full. Finally, we recompose the elements using subytpe.mk into an element of our subtype.

For proofs about subtypes, the lemma subtype.eq is useful. It states that two elements of a subtype are equal if their subtype.val components are equal. Here is how it can be used to prove that the mirror image of a mirror image of a full_btree is the full_btree itself:

The lemma subtype.eq allows us to transform the goal into mirror (mirror (t.val)) = t.val, which matches the statement of the lemma mirror_mirror we proved earlier. The two unfold calls help understand the proof but are not necessary.

As a second example, consider the following definition of vectors from mathlib:

```
def vector (\alpha : Type u) (n : \mathbb{N}) := { 1 : list \alpha // 1.length = n }
```

Here, a vector is defined as a list of a given length. With lists, there is only one type for a list of all lengths. For vectors, we have one dedicated type for every length of a vector. The advantage of this is that some operations such as addition or scalar product of vectors rely on the two involved vectors having the same length.

Vectors can be built from lists using subtype.mk or the angle bracket notation:

```
def some_vector : vector \mathbb{Z} 3 := \langle [1, 2, 3], \text{ by refl} \rangle
```

Basic operations on vectors can be defined by decomposing them with subtype.val and subtype.property, manipulating the underlying lists, and then recomposing them with subtype.mk. For example, we can define the negation of an integer vector as follows:

```
def vector.neg \{n : \mathbb{N}\}\ (v : vector \mathbb{Z} n) : vector \mathbb{Z} n := \langle \text{list.map int.neg v.val},  begin rw list.length_map, exact v.property end\rangle
```

We use list.map to negate every entry of the underlying list and list.length_map to show that this operation does not change the length of the list. Using subtype.eq, we can prove the following lemma about this new operator:

```
lemma neg_neg (n : N) (v : vector Z n) :
   v.neg.neg = v :=
begin
   apply subtype.eq,
   unfold vector.neg,
   unfold subtype.val,
   apply list.neg_neg
end
```

The application of subtype.eq reduces the goal to showing the corresponding property on the underlying lists. We can then use a lemma list.neg_neg, which states (1.map int.neg).map int.neg = 1 for all lists 1.

11.5 Quotient Types

Quotients are a powerful construction in mathematics used to define \mathbb{Z} , \mathbb{R} , and many other sets. Lean supports *quotient types*, an analogous mechanism on types. Just like subtypes, quotient types construct a new type from an existing type. unlike a subtype, a quotient type contains all of the elements of the underlying type—but some elements that were different in the underlying type are considered equal in the quotient type.

The prerequisites to construct a quotient type are a type α and a relation ${\tt r}:\alpha\to\alpha\to{\tt Prop}$ specifying which elements of α will be considered equal in the quotient. Given that,

```
quot r
```

represents our new quotient type. Every element a in α belongs to some element in quot ${\bf r}$. We can find that element using

```
\mathtt{quot.mk}\;\mathtt{r}:\alpha\to\mathtt{quot}\;\mathtt{r}
```

As an an abbreviation for quot.mk r a, we can also write [a]. The axiom quot.sound guarantees that elements for which the relation r holds are indeed equal in the quotient type:

```
\mathtt{Qquot.sound}\ \alpha\ \mathtt{r}:\forall\mathtt{a}\ \mathtt{b}:\alpha\text{, ra}\ \mathtt{b}\to [\![\mathtt{a}]\!]=[\![\mathtt{b}]\!]
```

Finally, we can *lift* functions from $\alpha \to \beta$ into quot $r \to \beta$ using quot.lift, which has the following computation rule. Given some $f: \alpha \to \beta$ such that $h: \forall a \ b, \ r \ a \ b \to f \ a = f \ b$, we have

We can think of quot.lift as an eliminator.

In most cases, the relation \mathbf{r} we want to use is an equivalence relation, which implies $[a] = [b] \leftrightarrow \mathbf{r}$ a b. For equivalence relations, quotient provides a convenient interface on top of quot.

As an example for a quotient type using the quotient interface, we will now construct the integers. A construction that turns out to be very convenient is to construct a quotient over pairs of natural numbers (i.e., $\mathbb{N} \times \mathbb{N}$). The idea is that a pair (m, n) of natural numbers represents the integer m-n. In this way, we can represent all nonnegative integers m by (m, 0) and all negative integers -n by (0, n). In addition, we get many more representations of the same integers, e.g., (2, 1), (3, 2), and (4, 3) all represent the integer 1.

To use the quotient interface, we first need to register the equivalence relation that we want to use. We want two pairs (k, 1) and (m, n) to be equal when k-1 and m-n yield the same integer. However, the condition k-1=m-n does not work because the negation on $\mathbb N$ does not behave like the negation on integers, e.g., 0-1=0. Instead, we use the condition k+n=m+1, which contains only addition.

To register the equivalence relation, we use the type class \mathtt{setoid} , which equips a type with an equivalence relation \approx . We provide the definition of our equivalence relation, followed by a proof that it is indeed a equivalence relation (i.e., that it is reflexive, symmetric, and transitive):

```
instance int.rel : setoid (\mathbb{N} \times \mathbb{N}) := { r := \lambdaa b, a.1 + b.2 = b.1 + a.2, iseqv := \langle (assume a, by refl), (assume a b eq, eq.symm), (assume a b c eq_ab eq_bc, have (a.1 + c.2) + b.2 = (c.1 + a.2) + b.2 := calc (a.1 + c.2) + b.2 = (a.1 + b.2) + c.2 : by ac_refl ... = a.2 + (b.1 + c.2) : by rw [eq_ab]; ac_refl ... = (c.1 + a.2) + b.2 : by rw [eq_bc]; ac_refl, calc a.1 + c.2 = c.1 + a.2 : eq_of_add_eq_add_right this) \rangle }
```

We can now write \approx for our equivalence relation:

```
lemma rel_iff (a b : \mathbb{N} \times \mathbb{N}) : a \approx b \leftrightarrow a.1 + b.2 = b.1 + a.2 := by refl
```

Using the name int.rel of our setoid instance that we registered above, we can then define the integers as

```
def int : Type := quotient int.rel
```

We can define the integer zero as

```
def zero : int := [(0, 0)]
```

In fact, any term of the form [(n, n)] represents the same integer:

```
example (n : N) : zero = [(n, n)] :=
begin
  unfold zero,
  apply quotient.sound,
  rw [rel_iff],
  simp
end
```

In this proof, we use the lemma quotient. sound, which tells us that two elements $[\![a]\!]$ and $[\![b]\!]$ are equal, if $a\approx b$.

Next, we define addition on our newly constructed integers. To define functions on a quotient type, one can define the function on the original type first, and then lift the definition to the quotient type. To do that, we must prove that the definition of the function f does not depend on the chosen representative of an equivalence class, i.e., $a \approx b \rightarrow f$ a = f b. The functions quotient.lift_on (for unary functions) and quotient.lift_on₂ (for binary functions) can be used to lift functions in this way.

Addition can be defined as simply adding the pairs of natural numbers componentwise. We then need provide a proof that this can be lifted to a function on the quotient by showing that $a_1 \approx b_1$ and $a_2 \approx b_2$ imply $[(a_1.1 + a_2.1, a_1.2 + a_2.2)] = [(b_1.1 + b_2.1, b_1.2 + b_2.2)]$:

```
def add (a b : int) : int := quotient.lift_on<sub>2</sub> a b (\lambdaa b, [(a.1 + b.1, a.2 + b.2)]) begin intros a<sub>1</sub> b<sub>1</sub> a<sub>2</sub> b<sub>2</sub> ha hb, apply quotient.sound,
```

```
simp only [rel_iff] at ha hb \vdash,
calc (a_1.1 + b_1.1) + (a_2.2 + b_2.2) =
(a_1.1 + a_2.2) + (b_1.1 + b_2.2) : by ac_refl
... = (a_2.1 + a_1.2) + (b_2.1 + b_1.2) : by rw [ha, hb]
... = (a_2.1 + b_2.1) + (a_1.2 + b_1.2) : by ac_refl
```

This yields a function add with the intended behavior:

```
lemma add_mk (ap an bp bn : \mathbb{N}) : add [(ap, an)] [(bp, bn)] = [(ap + bp, an + bn)] := by refl
```

We can use this to prove other lemmas about add, such as

```
lemma add_zero (i : int) : add zero i = i :=
begin
   apply quotient.induction_on i,
   intro p,
   apply quotient.sound,
   simp
end
```

11.6 Summary of New Syntax

Tactics

by_cases makes a case distinction on whether a given statement is true applies the axiom of choice to a hypothesis of the form $\forall \dots \exists \dots$

Constants

classical.choice function that returns an arbitrary element of a nonempty type function that returns a witness given a proof of an existential classical.some abbreviation for Sort 0 Prop function that creates a quotient type given an arbitrary relation quot function that creates a quotient type of a given setoid instance quotient type class for a type with an equivalence relation on it setoid type universe at level u Sort u abbreviation for Sort (u + 1) Type u

Notation

 $\{x : \alpha // p x\}$ subtype of all x in α fulfilling the predicate p equivalence relation on a setoid

Chapter 12

Basic Mathematical Structures

In this chapter, we introduce definitions and proofs about basic mathematical structures.

12.1 Type Classes without Properties

A type class is a collection of abstract constants and their properties. A type can be declared an instance of a type class by providing concrete definitions for the constants and proving that the properties hold.

A simple example for a type class is the class inhabited, which requires only a constant default_value and no properties:

```
class inhabited (\alpha : Type) := (default_value : \alpha)

This is syntactic sugar for 

@[class]

structure inhabited (\alpha : Type) : Type := (default_value : \alpha)
```

The type class inhabited is very similar to the inductive predicate nonempty that we discussed in Section 11.3. But since inhabited lives in Type, there is a large eliminator and we can extract the default_value that was used to construct an instance of the type class inhabited α .

Any type that has at least one element can be declared an instance of this class. For example, we can declare the natural numbers $\mathbb N$ as an instance by choosing an arbitrary natural number to be the default value:

```
instance : inhabited \mathbb{N} := { default_value := 0 }
```

Using this type class, we can define a variant of the head operation on lists that we defined in Section 4.2. The issue with head is that there is no meaningful value that we can assign to head []. Given a type of type class inhabited, we can simply return the default value instead of using the option type.

```
def ihead \{\alpha: {\tt Type}\} [inhabited \alpha] (xs : list \alpha) : \alpha:={\tt match}\ {\tt xs}\ {\tt with} | [] := inhabited.default_value \alpha | x :: _ := x end
```

We require that the type α must be of type class inhabited by writing [inhabited α]. This allows us to access inhabited.default_value in the definition. The syntax [inhabited α] adds an implicit argument to the constant ihead. But unlike for other implicit argument

ments, Lean performs not only type inference but also a type class search through all declared instances to determine the value of this argument. Thus, when we write

```
#reduce ihead ([] : list \mathbb{N})
```

Lean will search for an instance $inhabited \ \mathbb{N}$ and will find the instance we declared above. In that instance declaration, we defined $default_value$ to be 0 and hence this is what the #reduce command returns.

We can prove abstract lemmas about the type class inhabited such as

```
lemma ihead_ihead \{\alpha: {\tt Type}\} [inhabited \alpha] (xs : list \alpha) : ihead [ihead xs] = ihead xs := sorry
```

For such lemmas, it is important to add the assumption [inhabited α] to be allowed to use the operator ihead on lists of type list α . If we omit this assumption, Lean will raise an error telling us that type class synthesis failed.

There are a few more type classes only requiring a constant but no properties—for example:

```
class has_zero (\alpha : Type) := (zero : \alpha) class has_neg (\alpha : Type) := (neg : \alpha \to \alpha) class has_add (\alpha : Type) := (add : \alpha \to \alpha \to \alpha) class has_one (\alpha : Type) := (one : \alpha) class has_inv (\alpha : Type) := (inv : \alpha \to \alpha) class has_mul (\alpha : Type) := (mul : \alpha \to \alpha \to \alpha)
```

These syntactic type classes introduce constants that are used in many different contexts with different semantics. For example, one can stand for the natural number 1, the integer 1, the real 1, the identity matrix, and many other concepts of 1. The main purpose of these type classes is to form the foundation for rich hierarchy of algebraic type classes and to allow overloading of common symbols such as *, 1, and $^{-1}$.

The syntactic type classes do not impose serious restrictions on the types that can be declared instances, except that a type of type class has_zero or has_one must be inhabited. In contrast, the *semantic* type classes, which we will discuss in the following sections, contain properties that restrict how the given constants behave.

12.2 Type Classes over a Single Binary Operator

The type classes group and add_group are examples of semantic type classes. In mathematics, a group is a set G with a a binary operator $\bullet: G \times G \to G$ fulfilling the following properties, called group axioms:

- Associativity: For all $a, b, c \in G$, we have $(a \bullet b) \bullet c = a \bullet (b \bullet c)$.
- Identity element: There exists an element $e \in G$ such that for all $a \in G$, we have $e \bullet a = a$.
- Inverse element: For each $a \in G$, there exists an inverse element denoted a^{-1} such that $a^{-1} \bullet a = e$.

Commonly, group operations are written multiplicatively (i.e., with operator *, identity element 1, and inverse element a^{-1}) or additively (i.e., with operator +, identity element 0, and inverse element -a). This is why Lean offers two type classes for groups: the multiplicative group and the additive add_group. They are essentially the same but use different names for their constants and properties.

Any type that fulfills the group axioms can be instantiated as group or add_group. To illustrate this, we will define \mathbb{Z}_2 , also known as $\mathbb{Z}/2\mathbb{Z}$, and instantiate it as an add_group. \mathbb{Z}_2 has two elements, which we will call zero and one:

```
inductive \mathbb{Z}_2 : Type \mid zero \mid one
```

Addition is defined as follows:

```
\begin{array}{lll} \text{def } \mathbb{Z}_2.\text{add} : \mathbb{Z}_2 \to \mathbb{Z}_2 \to \mathbb{Z}_2 \\ \mid \mathbb{Z}_2.\text{zero } \mathbf{x} & := \mathbf{x} \\ \mid \mathbf{x} & \mathbb{Z}_2.\text{zero} := \mathbf{x} \\ \mid \mathbb{Z}_2.\text{one} & \mathbb{Z}_2.\text{one} := \mathbb{Z}_2.\text{zero} \end{array}
```

Invoking #print add_group tells us all the constants and properties we need to provide:

```
structure add_group : Type u \rightarrow Type u fields: add_group.add : \Pi\{\alpha\} [add_group \alpha], \alpha \rightarrow \alpha \rightarrow \alpha add_group.add_assoc : \forall \{\alpha\} [add_group \alpha] (a b c : \alpha), a + b + c = a + (b + c) add_group.zero : \Pi(\alpha) [add_group \alpha], \alpha add_group.zero_add : \forall \{\alpha\} [add_group \alpha] (a : \alpha), 0 + a = a add_group.add_zero : \forall \{\alpha\} [add_group \alpha] (a : \alpha), a + 0 = a add_group.neg : \Pi\{\alpha\} [c : add_group \alpha], \alpha \rightarrow \alpha add_group.add_left_neg : \forall \{\alpha\} [add_group \alpha] (a : \alpha), -a + a = 0
```

The constants add_group.add, add_group.zero, and add_group.neg respectively correspond to the binary operator •, the identity element e, and the inverse $^{-1}$ in the definition of a group above. The properties add_group.add_assoc, add_group.zero_add, and add_group.add_left_neg correspond to the three group axioms. Due to the way the add_group type class is constructed, we also need to provide the redundant property add_group.add_zero.

 \mathbb{Z}_2 can be instantiated as a group as follows:

```
instance \mathbb{Z}_2.add_group : add_group \mathbb{Z}_2 := { add := \mathbb{Z}_2.add, add_assoc := begin intros a b c, cases a; cases b; cases c; refl end, zero := \mathbb{Z}_2.zero, zero_add := begin intro a, cases a; refl end, add_zero := begin intro a, cases a; refl end, neg := id, add_left_neg := begin intro a, cases a; refl end }
```

With this instance declaration, we can now use the notations 0, +, and -:

```
#reduce \mathbb{Z}_2.one + 0 - 0 - \mathbb{Z}_2.one
```

Moreover, we get any lemma that has already been proved in general for the type class add_group for free:

```
example : \forall a : \mathbb{Z}_2, a + - a = 0 := add\_right\_neg
```

The algebraic hierarchy contains some more type classes with one binary operator. Here is an overview of the most important ones:

Structure	Properties
semigroup	associativity
monoid	semigroup with unit (1)
left_cancel_semigroup	semigroup with $x * a = x * b \rightarrow a = b$
right_cancel_semigroup	semigroup with $a * x = b * x \rightarrow a = b$
group	monoid with inverse $(^{-1})$

Most of these structures have commutative versions (where $a \bullet b = b \bullet a$ for all elements a, b), prefixed with comm_:

```
comm_semigroup, comm_monoid, comm_group
```

All of these type classes have additive counterparts with the prefix add_:

```
add_semigroup, add_monoid, add_group
add_comm_semigroup, add_comm_monoid, add_comm_group
add_left_cancel_semigroup, add_right_cancel_semigroup
```

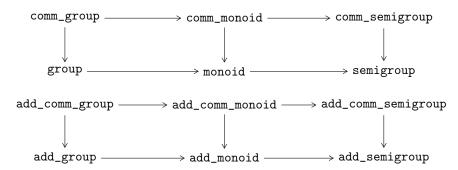
Although these additive type classes are isomorphic to their multiplicative counterparts, they are crucial when constructing algebraic structures with more than one binary operator such as rings and fields. To avoid duplicating all lemmas and definitions based on the multiplicative type classes, the copying process is automated by a tactic (transport_multiplicative_to_additive) and an attribute (@[to_additive]).

An example for an instance of add_monoid is the type list with the empty list [] as zero and the append operator ++ as addition:

```
instance \{\alpha : \text{Type}\} : \text{add\_monoid (list } \alpha) := \{ \text{zero} := [], \\ \text{add} := (++), \\ \text{add\_assoc} := \text{list.append\_assoc}, \\ \text{zero\_add} := \text{list.nil\_append}, \\ \text{add\_zero} := \text{list.append\_nil } \}
```

We could go further an register this combination as a add_left_cancel_semigroup and an add_right_cancel_semigroup.

The graph below illustrates the relationships between some of these type classes. The arrows mean "inherits all properties from."



12.3 Type Classes over Two Binary Operators

The additive and multiplicative structures are amalgamated to form more complex type classes over two binary operators. One of these is field. A field F is defined by the following properties:

- F forms a commutative group under an operator +, called addition, with identity element o.
- $F\setminus\{o\}$ forms a commutative group under an operator *, called multiplication.
- Multiplication distributes over addition—i.e., a * (b + c) = a * b + a * c for all $a, b, c \in E$

By invoking #print field, we can display all constants and properties required by the field type class. Again, the class includes some redundant properties such as left_distrib and right_distrib, due to the way in which it is constructed.

We will now show that \mathbb{Z}_2 is a field by instantiating the field type class with it. First, we define multiplication on \mathbb{Z}_2 as follows:

```
\begin{array}{llll} \text{def } \mathbb{Z}_2.\text{mul} & : \mathbb{Z}_2 \to \mathbb{Z}_2 \to \mathbb{Z}_2 \\ \mid \mathbb{Z}_2.\text{one} & x & := x \\ \mid x & \mathbb{Z}_2.\text{one} & := x \\ \mid \mathbb{Z}_2.\text{zero} & \mathbb{Z}_2.\text{zero} & := \mathbb{Z}_2.\text{zero} \end{array}
```

To declare \mathbb{Z}_2 a field, we can reuse the instance \mathbb{Z}_2 .add_group that we declared above using the syntax .. \mathbb{Z}_2 .add_group. We can prove the remaining properties as follows:

```
instance : field \mathbb{Z}_2 :=
{ one
                  := \mathbb{Z}_2.one,
                  := \mathbb{Z}_2.mul,
  mul
  inv
                  := id,
                := begin intros a b, cases a; cases b; refl end,
  add_comm
  zero_ne_one := begin intro h, cases h end,
 one_mul := begin intros a, cases a; refl end, mul_one := begin intros a, cases a; refl end,
  mul_inv_cancel :=
    begin intros a h, cases a, contradiction, refl end,
  inv_mul_cancel :=
    begin intros a h, cases a, contradiction, refl end,
    begin intros a b c, cases a; cases b; cases c; refl end,
  mul_comm := begin intros a b, cases a; cases b; refl end,
  left_distrib
    begin intros a b c, cases a; cases b; cases c; refl end,
  right_distrib :=
    begin intros a b c, cases a; cases b; cases c; refl end,
  ..\mathbb{Z}_2.add\_group }
```

With this declaration in place, we can now use the notations 1, *, /, and more:

```
#reduce (1 : \mathbb{Z}_2) * 0 / (0 - 1) -- result: \mathbb{Z}_2.zero
```

The type annotation : \mathbb{Z}_2 is necessary here to tell Lean that we want to calculate in \mathbb{Z}_2 and not in \mathbb{N} , the default. We can even use arbitrary numerals in \mathbb{Z}_2 . For example, the numeral 3 is interpreted as 1 + 1 + 1, which is the same as 1 in \mathbb{Z}_2 :

```
#reduce (3 : \mathbb{Z}_2) -- result: \mathbb{Z}_2.one
```

Moreover, we can use the tactic ring to normalize terms. For example:

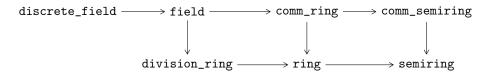
```
example (a b : \mathbb{Z}_2) : (a + b) ^ 3 = a ^ 3 + 3 * a^2 * b + 3 * a * b ^ 2 + b ^ 3 := by ring
```

This tactic is available for any type that is declared as a field or more generally as a ring. For commutative monoids and groups, there is a similar normalization tactic called abel.

Besides field, there are many more type classes for structures with two binary operators. These are the most important ones:

Structure	Properties	Examples
semiring	monoid and add_comm_monoid with distributivity	\mathbb{R} , \mathbb{Q} , \mathbb{Z} , \mathbb{N}
comm_semiring	semiring with commutativity of *	\mathbb{R} , \mathbb{Q} , \mathbb{Z} , \mathbb{N}
ring	monoid and add_comm_group with distributivity	\mathbb{R} , \mathbb{Q} , \mathbb{Z}
comm_ring	ring with commutativity of *	\mathbb{R} , \mathbb{Q} , \mathbb{Z}
division_ring	ring with multiplicative inverse	\mathbb{R} , \mathbb{Q}
field	division_ring with commutativity of *	\mathbb{R} , \mathbb{Q}
discrete_field	field with decidable equality and $\forall n, n / 0 = 0$	\mathbb{R} , \mathbb{Q}

The property $\forall n$, $n \neq 0$ = 0 required by the discrete_field type class is simply a convention to make division a total function. Mathematicians would view division as a partial function.



The hierarchy between ring and field is more complex than shown in this overview and includes type classes for domains, integral domains, Euclidean rings, and more.

12.4 Coercions

When dealing with different kinds of numbers from \mathbb{N} , \mathbb{Z} , \mathbb{Q} , and \mathbb{R} at the same time, we may want to cast from one type into another. For example, given a natural number n, we may need to convert it to an integer:

```
lemma neg_mul_neg_nat (n : \mathbb{N}) (z : \mathbb{Z}) : (- z) * (- n) = z * n := neg_mul_neg z n
```

Surprisingly, this statement does not lead to an error, although negation – n is not defined on $n:\mathbb{N}$, multiplication of $z:\mathbb{Z}$ with $n:\mathbb{N}$ is not defined, and the lemma neg_mul_neg does not apply to natural numbers. By invoking $print neg_mul_neg_nat$, we can see what happened:

```
theorem neg_mul_neg_nat : \forall (z: \mathbb{Z}) (n : \mathbb{N}), \neg z * \neg \uparrow n = z * \uparrow n := <math>\lambda(z: \mathbb{Z}) (n : \mathbb{N}), neg_mul_neg z \uparrow n
```

Lean has a mechanism to introduce coercions, represented by \uparrow or coe, whenever necessary. This coercion operator can be set up to provide implicit conversions between arbitrary types. Many coercions are already in place, including the following:

• coe : $\mathbb{N} \to \alpha$ casts \mathbb{N} into another semiring α ; • coe : $\mathbb{Z} \to \alpha$ casts \mathbb{Z} into another ring α ; • coe : $\mathbb{Q} \to \alpha$ casts \mathbb{Q} into another division ring α .

In some cases, Lean is unable to figure out where to place the coercions. We can then provide some type annotations to guide Lean's inference algorithm, as in the following example:

```
lemma neg_mul_neg_nat_2 (n : \mathbb{N}) (z : \mathbb{Z}) : (- n : \mathbb{Z}) * (- z) = n * z := neg_mul_neg n z
```

In proofs involving coercions, the tactic $norm_cast$ can be convenient. It can help with proof goals such as

```
m n : \mathbb{N},

h : \uparrow m = \uparrow n

\vdash m = n
```

in the proof

```
example (m n : \mathbb{N}) (h : (m : \mathbb{Z}) = (n : \mathbb{Z})) : m = n := begin norm_cast at h, exact h end
```

```
or
```

```
\begin{array}{c} \text{m n}: \ \mathbb{N} \\ \vdash \ \uparrow \text{m} + \uparrow \text{n} = \uparrow (\text{m} + \text{n}) \end{array} in the proof \begin{array}{c} \text{example (m n}: \ \mathbb{N}): \\ \text{(m}: \ \mathbb{Z}) + (\text{n}: \ \mathbb{Z}) = ((\text{m} + \text{n}: \ \mathbb{N}): \ \mathbb{Z}):= \\ \text{by norm\_cast} \end{array}
```

12.5 Lists, Multisets, and Finite Sets

We have seen many examples of how lists can be used in previous chapters. But when making a new definition or stating a new lemma, we should also reflect on alternatives such as multisets and finite sets.

Consider this definition based on the binary trees we introduced in Section 4.3:

This function returns a list of all the elements occurring in the tree. It traverses the tree depth first, from left to right. But for some applications, we might not care in which order the elements occur in the tree. This is where multisets can help us. For multisets, we have $\{3, 2, 1, 2\} = \{1, 2, 2, 3\}$, whereas the lists [3, 2, 1, 2] and [1, 2, 2, 3] are different. Multisets are defined as the quotient type over lists up to reordering. We can redo the above definition using multisets as follows:

```
\begin{array}{lll} \texttt{def nodes\_multiset} \ : \ \texttt{btree} \ \alpha \ \to \ \texttt{multiset} \ \alpha \\ | \ \texttt{empty} & := \ \emptyset \\ | \ (\texttt{node a } \ \texttt{t}_1 \ \texttt{t}_2) \ := \\ & \texttt{insert a } \ (\texttt{nodes\_multiset} \ \texttt{t}_1 \ \cup \ \texttt{nodes\_multiset} \ \texttt{t}_2) \end{array}
```

Using this definition, we can prove lemmas such as nodes_multiset t = nodes_multiset (mirror t), whereas nodes_list t = nodes_list (mirror t) does not hold.

For some applications, we might want to go a step further and ignore not only the order but also how often each element occurs in the tree, distinguishing only between occurrence and nonoccurrence. This is where finite sets—finsets—come into play. On finsets, we have $\{3, 2, 1, 2\} = \{1, 2, 3\}$. Finsets are defined as the subtype of multisets that do not contain any repeated elements. We can redo the definition above using finsets as follows:

For all three of these structures, Lean offers sum and product operators to add or multiply all of the values of a list, multiset, or finset:

The operators finset.sum and finset.prod take a second argument, a function that maps the values of the finset to the values that should be added or multiplied. If we put the identity function id for that second argument, they behave like the corresponding operators for lists and multisets.

These operators require the type of the elements to be declared as an instance of add_monoid for addition or of monoid for multiplication. For multiset and finset, we additionally require an instance declaration for add_comm_monoid or comm_monoid because we cannot have the result depend on the order of adding or multiplying the elements.

mathlib provides a collections of lemmas describing how the big operators behave under reorderings, basic operators on list, multiset, and finset, homomorphisms, order relations, and more.

12.6 Order Type Classes

Many of the structures introduced above can be ordered. For example, the well-known order on the natural numbers can be defined as

```
inductive nat.le : \mathbb{N} \to \mathbb{N} \to \mathsf{Prop} | refl : \forall a : \mathbb{N}, nat.le a a | step : \forall a b : \mathbb{N}, nat.le a b \to nat.le a (b + 1)
```

This is an example of a *linear order*. A linear (or total) order is a binary relation \leq such that for all a, b, and c, the following properties hold:

- Reflexivity: a < a.
- Transitivity: If $a \le b$ and $b \le c$, then $a \le c$.
- Antisymmetry: If $a \le b$ and $b \le a$, then a = b.
- Totality: $a \le b$ or $b \le a$.

If a relation fulfills the first three of these properties (reflexivity, transitivity, and antisymmetry), it is a *partial order*. An example of a partial order is the subset relation \subseteq on sets, finite sets, or multisets. If a relation fulfills the first two of the above properties (reflexivity and transitivity), it is a *preorder*. An example of a preorder is comparing lists by their length.

In Lean, there are type classes for these different kinds of orders: linear_order, partial_order, and preorder. The preorder type class has the fields

```
le : \alpha \to \alpha \to \mathsf{Prop}
le_refl : \forall \mathsf{a} : \alpha, le a a
le_trans : \forall \mathsf{a} b c : \alpha, le a b \to le b c \to le a c
```

The partial_order type class has the additional field

```
le_antisymm : \forall \mathtt{a} \ \mathtt{b} \ : \ \alpha, le \mathtt{a} \ \mathtt{b} \ \to \ \mathtt{le} \ \mathtt{b} \ \mathtt{a} \ \to \ \mathtt{a} = \mathtt{b}
```

and linear_order has the additional field

```
le_total : \forall \mathbf{a} \ \mathbf{b} \ : \ \alpha, le \mathbf{a} \ \mathbf{b} \ \lor \ \mathbf{le} \ \mathbf{b} \ \mathbf{a}
```

We can declare the preorder on lists that compares lists by their length as follows:

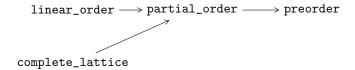
```
instance \{\alpha: \text{Type}\}: \text{preorder (list } \alpha):= \{ \text{le} := \lambda x \text{ y, x.length} \leq \text{y.length,} \\ \text{le_refl} := \lambda x, \text{nat.le_refl x.length,} \\ \text{le_trans} := \lambda x \text{ y z, nat.le_trans } \}
```

This type class instance gives us access to the infix syntax \leq and to the corresponding relations \geq , \prec , and \succ :

```
example \{\alpha : \text{Type}\}\ (c : \alpha) : [c] > [] := dec_trivial
```

The proof uses the dec_trivial tactic, which attempts to decide a goal using an inferred decision procedure, relying on definitional equalities to evaluate the goal. This tactic tends to work well for expressions consisting only of closed terms.

Complete lattices, which we discussed in Chapter 10, are formalized as another type class complete_lattice, which is derived form partial_order. The arrows in the following graphical overview mean "inherits all properties from."



Type classes combining orders and algebraic structures are also available:

All these mathematical structures relate \leq and < with the constants 0, 1, +, and * by monotonicity rules (e.g., $a \leq b \rightarrow c \leq d \rightarrow a + c \leq b + d$) and cancellation rules (e.g., $x + b \leq x + c \rightarrow b \leq c$).

12.7 Summary of New Syntax

Tactics

abel normalizes terms of commutative monoids and groups dec_trivial decides a goal (e.g., a true closed expression)
norm_cast normalizes coercions
ring normalizes ring terms

Chapter 13

Rational and Real Numbers

In the previous chapters, we have seen how the natural numbers $\mathbb N$ can be defined as an inductive type and how the integers $\mathbb Z$ can be defined as a quotient over $\mathbb N \times \mathbb N$. In this chapter, we present the construction of the rational numbers $\mathbb Q$ and the real numbers $\mathbb R$. The tools used for these constructions are the same as for $\mathbb N$ and $\mathbb Z$: inductive types and quotients.

A general recipe to construct new types with specific properties is as follows:

- 1. Create a new type that can represent all elements, but not necessarily in a unique manner.
- 2. Quotient this representation, equating elements that should be considered equal.
- 3. Define operators on the quotient type by lifting functions from the base type, and prove that they respect the quotient relation.

We have used this approach before to construct $\mathbb Z.$ It can be used to construct $\mathbb Q$ and $\mathbb R$ as well.

13.1 Rational Numbers

A rational number is a number that can be expressed as a fraction n/d of integers n and $d \neq 0$:

```
\begin{array}{lll} \text{structure fraction} := & \\ (\text{num} & : & \mathbb{Z}) \\ (\text{denom} & : & \mathbb{Z}) \\ (\text{denom\_ne\_zero} : & \text{denom} \neq 0) \end{array}
```

The number n is called the numerator, and the number d is called the denominator. The representation of a rational number as a fraction is not unique. For example, the rationals 1/2, 2/4, and -1/-2 are all equal. This representation as a fraction will serve as the base type from which we will derive a quotient.

Two fractions n_1/d_1 and n_2/d_2 represent the same rational number if the ratio between numerator and denominator are the same, meaning that $n_1*d_2=n_2*d_1$. To construct the quotient of the type fraction with respect to this relation, we show that the relation is an equivalence relation. This is achieved by declaring fraction an instance of the setoid type class:

```
instance : setoid fraction := 
{ r := \lambdaa b, a.num * b.denom = b.num * a.denom, iseqv := 
 \langleassume \langlen, d, hd\rangle, by simp, assume a b h, h.symm,
```

Then we can define $\mathbb Q$ as the quotient over this setoid. To avoid a name clash, we call it $my\mathbb Q$:

```
def my \mathbb{Q} : Type := quotient fraction.setoid
```

To define zero, one, addition, multiplication, and other operations, we first define them on the fraction type. To add two fractions, we convert the fractions to a common denominator and add the numerators. The easiest common denominator to use is simply the product of the two denominators:

We register these operations directly as instances of the syntactic type classes such as has_add to be able to use convenient notation such as + on fraction. Similarly, we define zero as 0 := 0 / 1, one as 1 := 1 / 1, and multiplication as the pairwise multiplication of numerators and denominators.

To lift these operations to $my\mathbb{Q}$, we first need to show that they respect the relation \approx :

```
@[simp] lemma add_num (a b : fraction) :
  (a + b).num = a.num * b.denom + b.num * a.denom :=
by refl
@[simp] lemma add_denom (a b : fraction) :
  (a + b).denom = a.denom * b.denom :=
by refl
lemma add_equiv_add {a b c d : fraction} (hac : a \approx c)
    (hbd : b \approx d) :
  a + b \approx c + d :=
begin
  simp at hac hbd,
  have h_1: a.num * b.denom * (c.denom * d.denom)
      = (c.num * a.denom) * b.denom * d.denom :=
    by rw \leftarrow hac; ac\_refl,
  have h2 : b.num * a.denom * (c.denom * d.denom)
      = (d.num * b.denom) * a.denom * c.denom :=
    by rw \leftarrowhbd; ac_refl,
  simp [add_mul, h1, h2],
  ac_refl
end
```

Then we can use $quotient.lift_on$ and $quotient.lift_on_2$ to define the operations on $my\mathbb{Q}$, and we can instantiate the relevant the syntactic type classes:

```
instance : has_add myQ :=
{ add := λa b, quotient.lift_on2 a b (λa b, [a + b])
  begin
    assume a b c d hac hbd,
    apply quotient.sound,
    exact fraction.add_equiv_add hac hbd
end }
```

From here, we can proceed and prove all the properties needed to make myQ an instance of field or, if we are more ambitious, discrete_linear_ordered_field.

Alternative Definitions of the Rational Numbers

We will discuss two more approaches to construct the rational numbers.

The first approach is the one used in mathlib. It is similar to the approach described above, but adds an additional property cop: coprime num denom to the structure, which states that that numerator and denominator do not have a common divisor (except 1 and -1):

```
structure rat := (\text{num} : \mathbb{Z}) (\text{denom} : \mathbb{N}) (\text{pos} : 0 < \text{denom}) (\text{cop} : \text{coprime num denom})
```

On the positive side, no quotient construction is required, computation may be more efficient, and more theorems are definitional equalities. A disadvantage is that function definitions become more complicated.

A second alternative approach is to define all elements syntactically, including the desired operations:

```
inductive pre_rat : Type
| zero : pre_rat
| one : pre_rat
| add : pre_rat \rightarrow pre_rat \rightarrow pre_rat
| sub : pre_rat \rightarrow pre_rat \rightarrow pre_rat
| mul : pre_rat \rightarrow pre_rat \rightarrow pre_rat
| div : pre_rat \rightarrow pre_rat \rightarrow pre_rat
```

Then we define an equivalence relation and take its quotient to enforce congruence rules and the field axioms.

```
inductive equiv : pre_rat → pre_rat → Prop
| add_congr {a b c d : pre_rat} :
        equiv a b → equiv c d → equiv (add a c) (add b d)
| add_assoc {a b c : pre_rat} :
        equiv (add a (add b c)) (add (add a b) c)
| zero_add {a : pre_rat} : equiv (add zero a) a
| add_comm {a b : pre_rat} : equiv (add a b) (add b a)
| etc : equiv sorry sorry
def rat : Type :=
quot equiv
```

The advantages are that this construction does not require \mathbb{Z} and that it is easy to declare as an instance of field. Most importantly, though, this approach can be used for other

algebraic constructions, such as free monoids and free groups. A disadvantage is that the definition of orders and lemmas about them become more complicated.

13.2 Real Numbers

There are sequences of rational numbers that seem to converge because the numbers in the sequence get closer and closer to each other, and yet do not converge to a rational number. The sequence

```
a_0 = 1
a_1 = 1.4
a_2 = 1.41
a_3 = 1.414
a_4 = 1.4142
a_5 = 1.41421
a_6 = 1.414213
a_7 = 1.4142135
a_8 = 1.41421356
\vdots
```

where a_n is the largest number with n decimals such that $a_n^2 < 2$, is such a sequence. It seems to converge because each a_n is at most 10^{-n} away from any of the following numbers, but the limit is $\sqrt{2}$, which is not a rational number.

In that sense, the rational numbers are incomplete, and the reals are their *completion*. To construct the reals, we need to fill in the gaps that are revealed by these sequences that seem to converge, but do not.

Cauchy sequences capture the notion of a sequence that seems to converge. A sequence a_0, a_1, \ldots is Cauchy if for any $\varepsilon > 0$, there exists an $N \in \mathbb{N}$ such that for all $m \ge N$, we have $|a_N - a_m| < \varepsilon$. In other words, no matter how small we choose ε , we can always find a point in the sequence from which all following numbers do not deviate more than ε .

We formalize sequences of rational numbers as functions $f: \mathbb{N} \to \mathbb{Q}$ and denote the absolute value $|\cdot|$ by abs. This yields the following Lean definition of Cauchy sequences:

```
def is_cau_seq (f : \mathbb{N} \to \mathbb{Q}) : Prop := \forall \varepsilon > 0, \exists N, \forall m > N, abs (f N - f m) < \varepsilon
```

We define a type of Cauchy sequences as a subtype:

```
\label{eq:def_cau_seq} \begin{array}{l} \texttt{def} \ \texttt{cau\_seq} \ : \ \texttt{Type} \ := \\ \{\texttt{f} \ : \ \mathbb{N} \ \to \ \mathbb{Q} \ \ / \ \ \texttt{is\_cau\_seq} \ \ \texttt{f} \} \end{array}
```

The idea of this construction is to let the real numbers be represented by Cauchy sequences. Each Cauchy sequence represents the real number that is its limit; for example, the sequence $a_n=1/n$ represents the real number o, and the sequence 1, 1.4, 1.41, ... represents the real number $\sqrt{2}$. Two different Cauchy sequences can represent the same real number; for example, the sequence $a_n=1/n$ and the constant sequence $b_n=0$ both represent o. Therefore, we need to take the quotient over sequences representing the same real number. Formally, two sequences represent the same real number when their difference converges to zero:

```
instance equiv : setoid cau_seq := { r := \lambdaf g, \forall \varepsilon > 0, \existsN, \forallm \geq N, abs (f.val m - g.val m) < \varepsilon, iseqv := sorry }
```

We omitted the proof that this is an equivalence relation using sorry for brevity. Using this setoid instance, we can now define the real numbers:

```
def my_real : Type :=
quotient cau_seq.equiv
```

Like for the rational numbers, we need to define zero, one, addition, multiplication, and other operators. We define them on cau_seq first and lift them to my_real afterwards. For the constants 0 and 1, we can define them simply as a constant sequence. Any constant sequence is a Cauchy sequence:

```
def const (x : \mathbb{Q}) : cau_seq := \langle \lambda_-, x, \text{ begin intros } \varepsilon \text{ h}\varepsilon, \text{ use 0, simpa using h}\varepsilon \text{ end} \rangle
```

We can declare my_real instances of the syntactic type classes has_zero and has_one as follows:

```
instance : has_zero my_real := \langle [cau\_seq.const \ 0] \rangle instance : has_one my_real := \langle [cau\_seq.const \ 1] \rangle
```

Defining addition of real numbers requires a little more effort. We define addition on Cauchy sequences by adding the elements of the sequence pairwise:

```
instance : has_add cau_seq := { add := \lambda f g, \langle \lambda n, f.val n + g.val n, sorry\rangle}
```

This definition requires a proof that the result is a Cauchy sequence, given that ${\tt f}$ and ${\tt g}$ are Cauchy sequences. We omit that proof here. Next, we need to show that this addition is compatible with our equivalence relation:

```
lemma add_equiv_add {f_1 f_2 g_1 g_2 : cau_seq} (hf : f_1 \approx f_2)
      (hg : g_1 \approx g_2) :
   f_1 + g_1 \approx f_2 + g_2 :=
begin
   intros \varepsilon_0 h\varepsilon_0,
   obtain \langle \mathtt{Nf},\ \mathtt{hNf} 
angle : \exists \mathtt{Nf},\ orall (\mathtt{m}\ :\ \mathbb{N})\,,\ \mathtt{m}\ \geq\ \mathtt{Nf}\ 	o
         abs (f<sub>1</sub>.val m - f<sub>2</sub>.val m) < \varepsilon_0 / 2,
      from hf (\varepsilon_0 / 2) (half_pos h\varepsilon_0),
   obtain \langle \text{Ng, hNg} \rangle : \exists \text{Ng, } \forall (\text{m} : \mathbb{N}), \text{ m} \geq \text{Ng} \rightarrow
         abs (g_1.val m - g_2.val m) < \varepsilon_0 / 2,
      from hg (\varepsilon_0 / 2) (half_pos h\varepsilon_0),
   use max Nf Ng,
   intros m hm,
   calc
      abs ((f_1 + g_1).val m - (f_2 + g_2).val m)
      = abs ((f_1.val m + g_1.val m) - (f_2.val m + g_2.val m))
         : by refl
      \dots = abs ((f<sub>1</sub>.val m - f<sub>2</sub>.val m) + (g<sub>1</sub>.val m - g<sub>2</sub>.val m))
      \dots \leq abs (f_1.val m - f_2.val m) + abs (g_1.val m - g_2.val m)
         : by apply abs_add
      ... < \varepsilon_0 / 2 + \varepsilon_0 / 2
         : add_lt_add (hNf m (le_of_max_le_left hm))
                (hNg m (le_of_max_le_right hm))
      \dots = \varepsilon_0
         : by simp
```

We will discuss the proof in detail. It has a similar flavor to the two proofs we omitted above. To show that $f_1 + g_1 \approx f_2 + g_2$, we are given an $\varepsilon_0 > 0$ and must prove that there

exists an $\mathbb{N}:\mathbb{N}$ such that

```
\forall \mathtt{m}: \mathbb{N}, \mathtt{m} \geq \mathtt{N} 	o \mathtt{abs} \; ((\mathtt{f}_1 + \mathtt{g}_1).\mathtt{val} \; \mathtt{m} - (\mathtt{f}_2 + \mathtt{g}_2).\mathtt{val} \; \mathtt{m}) < \varepsilon_0
```

To obtain such an N, we use the facts $f_1 \approx f_2$ and $g_1 \approx g_2$. The fact $f_1 \approx f_2$ gives us for any $\varepsilon > 0$ a number Nf such that abs $(f_1.val\ m - f_2.val\ m) < \varepsilon$ for all $m \geq Nf$. The fact $g_1 \approx g_2$ gives us a number Ng with a similar property. For the calculations to work out in the end, we take the numbers Nf and Ng for $\varepsilon := \varepsilon_0 / 2$. Then we choose N to be the maximum of Nf and Ng, so that we get the inequalities for any $m \geq N$. The calc block at the end of the proof establishes that abs $((f_1 + g_1).val\ m - (f_2 + g_2).val\ m) < \varepsilon_0$ for all $m \geq N$.

Using the lemma, we can then define addition on my_real:

```
instance : has_add my_real :=
{ add := λa b, quotient.lift_on2 a b (λa b, [a + b])
   begin
    assume a b c d hac hbd,
   apply quotient.sound,
   exact cau_seq.add_equiv_add hac hbd,
   end }
```

Multiplication can be defined similarly.

Alternative Definitions of the Real Numbers

In mathlib, the construction of the real numbers is essentially as described above. Only some definitions are stated in a more general fashion to allow construction of other algebraic structures, such as the *p*-adic numbers.

Alternatively, the real numbers can be defined using Dedekind cuts. Essentially, a number $r : \mathbb{R}$ is then represented as $\{x : \mathbb{Q} \mid x < r\}$.

Another alternative definition of the reals is to define them using binary sequences $\mathbb{N} \to \mathtt{bool}$. The elements of the sequence represent the digits of the real number. This works particularly well if we only need the real numbers in the interval [0,1]. This construction does not require the rational numbers.

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126 Bibliography

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