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# 1 Introduction

This serves as a brief starting point for understanding how mathematical proofs can be formalized in Lean, as well as being an introduction to the language itself. Lean is both a **functional programming language** and a **theorem prover**. We'll focus primarily on its role as a theorem prover. But what does this mean, and how can that be achieved?

A programming language defines a set of rules, semantics, and syntax for writing programs. To achieve a goal, a programmer must write a program that meets given specifications. There are two primary approaches: program derivation and program verification ([NPS90] Section 1.1). In program verification, the programmer first writes a program and then proves it meets the specifications. This approach checks for errors at run-time when the code executes. In program derivation, the programmer writes a proof that a program with certain properties exists, then extracts a program from that proof. This approach enables specification checking at compilation-time, catching errors while typing, thus, before execution. This distinction corresponds to dynamic versus static type systems. Most programming languages combine both approaches; providing basic types for annotation and compile-time checking, while leaving the remaining checks to be performed at runtime.

**Example 1.1** In a dynamically typed language, like JavaScript, variables can change type after they are created. For example, a variable defined as a number can later be reassigned to a string: TypeScript is a statically typed superset of JavaScript. Unlike JavaScript, it performs type checking at compile time. This

means we can prevent the previous behavior, while writing our code, simply by adding a type annotation: Now converting a variable with type annotation number to a string results into a compile time error.

Nonetheless, TypeScript, even though it has a sophisticated type system, cannot fully capture complex mathematical properties. As well as for the most programming languages, program specifications can only be enforced at runtime. Lean, by contrast, uses a much more powerful type system that enables it to express and verify mathematical statements with complete rigor fully during compilation time. This makes it particularly suitable as a **theorem prover** for formalizing mathematics.

Lean's type system is based on **dependent type theory**, specifically the **Calculus of Inductive Constructions** (CIC) with various extensions. It's important to note that **type theory** is not a single, unified theory, but rather a family of related theories with various extensions, ongoing developments, and rich historical ramifications.

Type theory emerged as an alternative foundation for mathematics, addressing paradoxes that arose in naive set theory. Consider Russell's famous paradox: let  $S = \{x \mid x \notin x\}$  be the set of all sets that do not contain themselves. This construction is paradoxical, leading to the contradiction  $S \in S \iff S \notin S$ . Type theory resolves such issues by working with **types** as primary objects rather than sets, and by restricting which constructions are well-formed.

**Dependent type theory**, the framework underlying Lean, extends basic type systems by allowing types to depend on values. For instance, one can define the type of vectors of length n, where n is itself a value. This capability makes dependent type theory particularly expressive for formalizing mathematics.

Various proof assistants have been developed based on different variants of type theory, including Agda, Coq, Idris, and Lean. Each system makes different design choices regarding which rules and features to include. Lean adopts the Calculus of Inductive Constructions, which extends the Calculus of Constructions, introduced in Coq, with inductive types. Inductive types allow for the definition of structures such as natural numbers, lists, and trees.

A fundamental design feature of Lean is its **universe hierarchy** of types, with **Prop** (the proposition type) as a distinguished universe at the bottom. The **Prop** universe exhibits two special properties: **impredicativity** and **proof irrelevance**. Proof irrelevance means that all proofs of the same proposition are considered equivalent. What matters is whether a proposition can be proven, not which specific proof is given. This separation between propositions (**Prop**) and data types (**Type**) was first introduced in **N.G. de Bruijn's AUTOMATH system** (1967) ([Tho99]).

For our purposes, we do not need to delve deeply into the theoretical foundations; instead, we will introduce the relevant concepts as needed while working with Lean. The practical aspects of writing proofs in Lean will be our primary focus, and the theoretical machinery will be explained only insofar as it aids understanding of how to formalize mathematics effectively.

## 1.1 Lean first steps

In the language of type theory, and by extension in Lean, we write  $\mathbf{x}: \mathbf{X}$  to mean that x is a **term** of type X. For example,  $\mathbf{2}: \mathbb{N}$  annotates 2 as a natural number, or more precisely, as a term of the natural number type. Lean has internally defined types such as Nat or  $\mathbb{N}$  (you can type \Nat to get the Unicode symbol). The command #check allows us to inspect the type of any espression, term or variable.

#### Example 1.2

```
#check 2 -- 2 : Nat
#check 2 + 2 -- 2 + 2 : Nat
```

[Try this example in Lean Web Editor] By following the link, you can try out the code in your browser. Lean provides a dedicated **infoview** panel on the right side. Position your cursor after #check 2, and the infoview will display the output 2: Nat. This dynamic interaction, where the infoview responds to your cursor position, is what makes Lean an **interactive theorem prover**. As you move through your code, the infoview continuously updates, showing computations, type information, and proof states at each location.

At first glance, one might be tempted to view the colon notation as analogous to the membership symbol  $\in$  from set theory, treating types as if they were sets. While this intuition can be helpful initially, type theory offers a fundamentally richer perspective. The crucial insight is the **Curry-Howard correspondence**, also known as the **propositions-as-types** principle. This correspondence establishes a deep connection between mathematical proofs and programs: **propositions correspond to types**, and **proofs correspond to terms** inhabiting those types. Under this interpretation, a term x: X can be understood in two more ways:

- As a **computational object**: x is a program or data structure of type X
- As a **logical object**: x is a proof of the proposition X

Lean is a concrete realization of the propositions-as-types principle, proving a theorem, within the language, amounts to constructing a term of the appropriate type. When we write theorem\_name: Proposition, we are declaring that theorem\_name is a proof (term) of Proposition (type). For example, consider proving that 2+2=4:

## Example 1.3

```
theorem \ two\_plus\_two\_eq\_four : 2 + 2 = 4 := rfl
```

Lean's syntax is designed to resemble the language of mathematics. Here, we use the **theorem** keyword to encapsulate our proof, of the statement/proposition 2+2=4, giving it the name two\_plus\_two\_eq\_four. This allows us to reference and reuse this result later in our code. After the semicolon, :, we introduce the

statement; 2 + 2 = 4. The := operator expects the proof term that establishes the theorem's validity. The proof itself consists of a single term: rfl (short for reflexivity). This is a proof term that works by definitional equality, Lean's kernel automatically reduces both sides of the equation to their normal (definitional) form and verifies they are identical. Since 2+2 computes to 4, the proof succeeds immediately. We can now use this theorem in subsequent proofs. For instance:

```
example : 1 + 1 + 1 + 1 = 4 := two_plus_two_eq_four
```

Well, this example is simple enough for Lean to evaluate by itself: 1 + 1 + 1 + 1 = 2 + 2 = 4 and conclude with two\_plus\_two\_eq\_four. Actually, rfl would solve the equation similarly, so this is just applying rfl again (it's a bit of cheating). Here, I used example, which is handy for defining anonymous expressions for demonstration purposes. Before diving into the discussion, here is another keyword, def, used to introduce definitions and functions.

```
def addOne (n : Nat) : Nat := n + 1
```

This definition expects a natural number as its parameter, written (n : Nat) and returns a natural number. [Run in browser]

Let's now turn to how logic is handled in Lean and how the Curry-Howard isomorphism is reflected concretely.

# 2 Logic and Proposition as Types

Logic is the study of reasoning, branching into various systems. We refer to **classical logic** as the one that underpins much of traditional mathematics. It's the logic of the ancient Greeks (not fair) and truthtables, and it remains used nowadays for pedagogical reasons. We first introduce **propositional logic**, which is the simplest form of classical logic. Later we will extend this to **predicate** (or first-order) logic, which includes **predicates** and **quantifiers**. In this setting, a **proposition** is a statement that is either true or false, and a **proof** is a logical argument that establishes the truth of a proposition. Propositions can be combined with logical **connectives** such as "and" ( $\wedge$ ), "or" ( $\vee$ ), "not" ( $\neg$ ), "false" ( $\bot$ ), "true" ( $\top$ ) "implies" ( $\Rightarrow$ ), and "if and only if" ( $\Leftrightarrow$ ). These connectives allow the creation of complex or compound propositions.

Here how connectives are defined in Lean:

## Example 2.1 (LogicaL connectives in Lean)

```
#check And (a b : Prop) : Prop
#check Or (a b : Prop) : Prop
#check True : Prop
#check False : Prop
#check Not (a : Prop) : Prop
#check Iff (a b : Prop) : Prop
```

Prop stands for proposition, and it is an essential component of Lean's type system. For now, we can think of it as a special type whose inhabitants are proofs; somewhat paradoxically, a type of types.

Logic is often formalized through a framework known as the **natural deduction system**, developed by Gentzen in the 1930s ([Wad15]). This approach brings logic closer to a computable, algorithmic system. It specifies rules for deriving **conclusions** from **premises** (assumptions from other propositions), called **inference rules**.

Example 2.2 (Deductive style rule) Here is an hypothetical example of inference rule.

$$\begin{array}{cccc} P_1 & P_2 & \cdots & P_n \\ \hline & C & \end{array}$$

Where the  $P_1, P_2, \ldots, P_n$ , above the line, are hypothetical premises and, the hypothetical conclusion C is below the line.

The inference rules needed are:

- Introduction rules specify how to form compound propositions from simpler ones, and
- Elimination rules specify how to use compound propositions to derive information about their components.

Type theory employs this porocedure too, by referring to deduction rules as judments, and each elements of a with terms and types. here an example:

Example 2.3 
$$\frac{\Gamma, \quad p_1: P_1 \quad p_2: P_2 \quad \cdots \quad P_n}{C}$$

Notation 2.4 We use  $A \vdash B$  (called turnstile) to designate a deduction of B from A. It is employed in Gentzen's sequent calculus ([GTL89]) and mostly used in type theory. The square brackets around a premise [A] mean that the premise A is meant to be discharged at the conclusion. The classical example is the introduction rule for the implication connective. To prove an implication  $A \to B$ , we assume A (shown as [A]), derive B under this assumption, and then discharge the assumption A to conclude that  $A \to B$  holds without the assumption. The turnstile is predominantly used in judgments and type theory with the meaning of "entails that".

Example 2.5 (Type Theory Judgments and Contexts) In type theory, judgments are formal statements about the well-formedness of types and terms. There are typically four fundamental kinds of judgments []:

- 1.  $\Gamma \vdash A \ type "A \ is \ a \ well-formed \ type \ in \ context \ \Gamma"$
- 2.  $\Gamma \vdash t : A$  "t is a term of type A in context  $\Gamma$ "
- 3.  $\Gamma \vdash A \equiv B$  type "types A and B are judgmentally equal in context  $\Gamma$ "
- 4.  $\Gamma \vdash t_1 \equiv t_2 : A$  "terms  $t_1$  and  $t_2$  are judgmentally equal of type A in context  $\Gamma$ "

The context  $\Gamma$  (Greek capital gamma) represents a finite list of type declarations for variables, formally written as  $\Gamma = x_1 : A_1, x_2 : A_2, \ldots, x_n : A_n$  (cite). The context encodes the assumptions or hypotheses under which a judgment is made.

For example, consider these concrete judgments:

$$\vdash \mathbb{N} \ type \quad (natural \ numbers \ form \ a \ type)$$
 (1)

$$x: \mathbb{N} \vdash x: \mathbb{N}$$
 (variable x has type  $\mathbb{N}$  when declared) (2)

$$x: \mathbb{N}, y: \mathbb{N} \vdash x + y: \mathbb{N}$$
 (addition of naturals yields a natural) (3)

$$\vdash \lambda x : \mathbb{N}. \ x + 1 : \mathbb{N} \to \mathbb{N} \quad (successor \ function)$$
 (4)

Let's look at how we can define some connectives. The inference rules needed are:

- Introduction rules specify how to form compound propositions from simpler ones, and
- Elimination rules specify how to use compound propositions to derive information about their components.

#### Conjunction $(\land)$

• Introduction

$$\frac{A}{A \wedge B} \wedge -\text{Intro}$$

• Elimination

$$\frac{A \wedge B}{A} \wedge \text{-Elim}_1 \qquad \qquad \frac{A \wedge B}{B} \wedge \text{-Elim}_2$$

## Disjunction $(\vee)$

• Introduction

$$\frac{A}{A \vee B} \vee \text{-Intro}_1 \qquad \qquad \frac{B}{A \vee B} \vee \text{-Intro}_2$$

• Elimination (Proof by cases)

$$\frac{A \vee B \qquad [A] \vdash C \qquad [B] \vdash C}{C} \vee \text{-Elim}$$

# Implication $(\rightarrow)$

• Introduction

$$\frac{[A] \vdash B}{A \to B} \to -Intro$$

• Elimination (Modus Ponens)

$$A \to B$$
  $A \to -\text{Elim}$ 

Let's look at how we can define some connectives.

## Conjunction $(\land)$

• Introduction

$$\frac{A}{A \wedge B} \wedge -Intro$$

• Elimination

$$\frac{A \wedge B}{A} \wedge \text{-Elim}_1$$

$$\frac{A \wedge B}{B} \wedge \text{-Elim}_2$$

# Disjunction $(\vee)$

• Introduction

$$\frac{A}{A \vee B} \vee \text{-Intro}_1$$

$$\frac{B}{A \vee B} \vee -Intro_2$$

• Elimination (Proof by cases)

$$\begin{array}{c|cccc} A \lor B & [A] \vdash C & [B] \vdash C \\ \hline C & \\ \end{array} \lor \text{-Elim}$$

## Implication $(\rightarrow)$

• Introduction

$$\frac{[A] \vdash B}{A \to B} \to -Intro$$

• Elimination (Modus Ponens)

$$A \to B \qquad A \to -\text{Elim}$$

**Notation 2.6** We use  $A \vdash B$  (called turnstile) to designate a deduction of B from A. It is employed in Gentzen's **sequent calculus** ([GTL89]), whereas in natural deduction the corresponding symbol is



There are some minor differences, in fact, which I don't fully understand. The square brackets around a premise [A] mean that the premise A is meant to be **discharged** at the conclusion. The classical example is the introduction rule for the implication connective. To prove an implication  $A \to B$ , we assume A (shown as [A]), derive B under this assumption, and then discharge the assumption A to conclude that  $A \to B$  holds without the assumption. The turnstile is predominantly used in judgments and type theory with the meaning of "entails that".

Lean has its own syntax for connectives and their relative inference rules. For instance  $A \wedge B$  can be presented as And(A, B) or A \land B, . Its untroduction rule is constructed by And.intro\_\_ or shortly  $\langle \_, \_ \rangle$  (underscore are placeholder for assumptions or "propositional functions"). The pair  $A \wedge B$  can be then consumed using elimination rules And.left and And.right.

**Example 2.7** Let's look at our first Lean example

```
example (H_A : A) (H_B : B) : (A \setminus land B) := And.intro H_A H_B
```

Lean aims to resemble the language used in mathematics. For instance, when defining a function or expression, one can use keywords such as theorem or def. Here, I used example, which is handy for defining anonymous expressions for demonstration purposes. After that comes the statement to be proved:

```
(H_A : A) (H_B : B) : (A \setminus land B)
```

Meaning given a proof of A and a proof of B we can form a proof of  $(A \land B)$ . The operator := assigns a value (or return an expression) for the statement which "has to be a proof of it". And intro is implemented as:

```
And. intro: p \setminus to q \setminus to (p \setminus land q).
```

It says: if you give me a proof of p and a proof of q, then i return a proof of  $p \wedge q$ . We therefore conclude the proof by directly giving And.intro H\_A H\_B. Here another way of writing the same statement.

```
example (H_p : p) (H_B : B) : And(A, B) := \langle H_p, H_B \rangle
```

This system of inference rules allows us to construct proofs in an algorithmic and systematical way, organized in what is called a **proof tree**. To reduce complexity, we follow a **top-down**. This methodology forms the basis of **proof** 

assistants like Lean, Coq, and Agda, which help verify the correctness of mathematical proofs by checking each step against these rules. We will see later that Lean, in fact, provides an info view of the proof tree which helps us understand and visualize the proof structure.

Let's examine a concrete example of a proof.

**Example 2.8 (Associativity of Conjunction)** We prove that  $(A \wedge B) \wedge C$  implies  $A \wedge (B \wedge C)$ . First, from the assumption  $(A \wedge B) \wedge C$ , we can derive A:

$$\frac{(A \wedge B) \wedge C}{\underbrace{A \wedge B}_{A} \wedge E_{1}} \wedge E_{1}$$

Second, we can derive  $B \wedge C$ :

$$\frac{(A \wedge B) \wedge C}{\frac{A \wedge B}{B} \wedge E_2} \wedge E_1 \qquad \underbrace{(A \wedge B) \wedge C}_{C \wedge I} \wedge E_2$$

Finally, combining these derivations we obtain  $A \wedge (B \wedge C)$ :

$$\frac{(A \land B) \land C \vdash A \qquad (A \land B) \land C \vdash B \land C}{A \land (B \land C)} \land I$$

Example 2.9 (Lean Implementation) Let us now implement the same proof in Lean.

```
theorem and_associative (a b c : Prop) : (a \land b) \land c \to a \land (b \land c) :=

fun h : (a \land b) \land c =>

-- First, from the assumption (a \land b) \land c, we can derive a:

have hab : a \land b := h.left -- extracts (derive) a proof of (a \land b)

from the assumption

have ha : a := hab.left -- extracts a from (a \land b)

-- Second, we can derive b \land c (here we only extract b and c and combine them in the next step)

have hc : c := h.right

have hb : b := hab.right

-- Finally, combining these derivations we obtain a \land (b \land c)

show a \land (b \land c) from \land ha, \land hb, hc \rand \rand \)
```

We introduce the theorem with the name and\_associative. The type signature  $(a \land b) \land c \rightarrow a \land (b \land c)$  represents our logical implication. Here, we construct the proof term using a function with the fun keyword. Why a function? We have already encountered the Curry-Howard correspondence in Lean previously, though without explicitly stating it. According to this correspondence, a proof of an implication can be understood as a function that takes a hypothesis as input and produces the desired conclusion as output. We will revisit this concept in more detail later. The have keyword introduces local lemmas within our

proof scope, allowing us to break down complex reasoning into manageable intermediate steps, mirroring our natural deduction proof from before. Just before the keyword show, the info view displays the following context and goal:

```
a b c : Prop
h : (a \land b) \land c
hab : a \land b
ha : a
hc : c
hb : b
\vdash a \land b \land c
```

Finally, the show keyword explicitly states what we are proving and verifies that our provided term has the correct type. In this case, show  $a \land (b \land c)$  from  $\langle ha, \langle hb, hc \rangle \rangle$  asserts that we are constructing a proof of  $a \land (b \land c)$  using the term  $\langle ha, \langle hb, hc \rangle \rangle$ . The show keyword serves two purposes: it makes the proof more readable by explicitly documenting what is being proved at this step, and it performs a type check to ensure the provided proof term matches the stated goal up to definitional equality. Two types are definitionally equal in Lean when they are identical after computation and unfolding of definitions—in other words, when Lean's type checker can mechanically verify they are the same without requiring additional proof steps. Here, the goal  $\vdash a \land b \land c$  is definitionally equal to  $a \land (b \land c)$  due to how conjunction associates, so show accepts this statement. If we had tried to use show with a type that was only propositionally equal (requiring a proof to establish equality) but not definitionally equal, Lean would reject it.

## 2.1 Predicate logic and dependency

To capture more complex mathematical ideas, we extend our system from propositional logic to **predicate logic**. A **predicate** is a statement or proposition that depends on a variable. In propositional logic we represent a proposition simply by P. In predicate logic, this is generalized: a predicate is written as P(a), where a is a variable. Notice that a predicate is just a function. This extension allows us to introduce **quantifiers**:  $\forall$  ("for all") and  $\exists$  ("there exists"). These quantifiers express that a given formula holds either for every object or for at least one object, respectively. In Lean if  $\alpha$  is any type, we can represent a predicate P on  $\alpha$  as an object of type  $\alpha \to Prop$ . Thus given an  $\mathbf{x} : \alpha$  (an element with type  $\alpha$ )  $P(\mathbf{x})$ : Prop would be representative of a proposition holding for  $\mathbf{x}$ .

of a variable name can be substituted without changing the meaning of the predicate or statement. This should feel familiar from mathematics, where the meaning of an expression does not depend on the names we assign to variables. We can give an informal reading of the quantifiers as infinite logical operations:

$$\forall x. A(x) \equiv A(a) \land A(b) \land A(c) \land \dots$$
$$\exists x. A(x) \equiv A(a) \lor A(b) \lor A(c) \lor \dots$$

The expression  $\forall x. P(x)$  can be understood as generalized form of implication. If P is any proposition, then  $\forall x. P$  expresses that P holds regardless of the choice of x. When P is a predicate, depending on x, this captures the idea that we can derive P from any assumption about x.

Example 2.10 Lean expresses quantifiers as follow.

```
\forall (x : I), P x
forall (x : I), P x -- another notation

Listing 1: For All

\exists (x : I), P x
exist (x : I), P x -- another notation
```

Listing 2: Exists

Where x is a varible with a type I, and P x is a proposition, or predicate, holding for x.

Example 2.11 (Existential introduction in Lean) When introducing an existential proof, we need a pair consisting of a witness and a proof that this witness satisfies the statement.

```
example (x : Nat) (h : x > 0) : \exists y, y < x := Exists.intro 0 h -- or shortly <math>\langle 0, h \rangle
```

Notice that (0, h) is product type holding a data ((witness) and a proof of it. The **existential elimination rule** (Exists.elim) performs the opposite operation. It allows us to prove a proposition Q from  $\exists x, P(x)$  by showing that Q follows from P(w) for an **arbitrary** value w.

Example 2.12 (Existential elimination in Lean) The existential rules can be interpreted as an infinite disjunction, so that existential elimination naturally corresponds to a proof by cases (with only one single case). In Lean, this reasoning is carried out using pattern matching, a known mechanism in functional programming for dealing with cases, with let or match, as well as by using cases or reases construct.

```
 \begin{array}{l} \textbf{example} \ (h : \exists \ n : \texttt{Nat}, \ n > 0) : \exists \ n : \texttt{Nat}, \ n > 0 : = \\ \textbf{match} \ h \ \textbf{with} \\ / \ \langle \textbf{witness}, \ proof \rangle \ => \ \langle \textbf{witness}, \ proof \rangle \\ \end{array}
```

**Example 2.13** The universal quantifier may be regarded as a generalized function. Accordingly, In Lean, universal elimination is simply function application.

```
example: \forall n: Nat, n \geq 0:= fun n \Rightarrow Nat.zero_le n
```

# 3 Describing and use properties

Functions are primitive objects in type theory. For example, it is interesting to note that a relation can be expressed as a function:  $\mathbf{R}:\alpha\to\alpha\to\mathsf{Prop}$ . Similarly, when defining a predicate ( $\mathbf{P}:\alpha\to\mathsf{Prop}$ ) we must first declare  $\alpha:$  Type to be some arbitrary type. This is what is called **polymorphism**, more specifically **parametrical polymorphism**. A canonical example is the identity function, written as  $\alpha\to\alpha$ , where  $\alpha$  is a type variable. It has the same type for both its domain and codomain, this means it can be applied to booleans (returning a boolean), numbers (returning a number), functions (returning a function), and so on. In the same spirit, we can define a transitivity property of a relation as follows:

```
def Transitive (\alpha : Type) (R : \alpha \to \alpha \to \text{Prop}) : Prop := \forall x y z, R x y \to R y z \to R x z
```

To use Transitive, we must provide both the type  $\alpha$  and the relation itself. For example, here is a proof of transitivity for the less-than relation on  $\mathbb{N}$  ( in Lean Nat or  $\mathbb{N}$ ):

```
theorem le_trans_proof : Transitive Nat (\cdot \leq \cdot : \text{Nat} \rightarrow \text{Nat} \rightarrow \text{Prop}) :=  fun x y z h1 h2 => Nat.le_trans h1 h2 -- this lemma is provided by Lean
```

Looking at this code, we immediately notice that explicitly passing the type argument Nat is somewhat repetitive. Lean allows us to omit it by letting the type inference mechanism fill it in automatically. This is achieved by using implicit arguments with curly brackets:

```
def Transitive {\alpha : Type} (R : \alpha \to \alpha \to \text{Prop}) : Prop := \forall x y z, R x y \to R y z \to R x z theorem le_trans_proof : Transitive (\cdot \leq \cdot : Nat \to Nat \to Prop) := fun x y z h1 h2 => Nat.le_trans h1 h2
```

Lean's type inference system is quite powerful: in many cases, types can be completely inferred without explicit annotations. For instance, (NEED TO EXPLAIN TYPE INFERENECE). Let us now revisit the transitivity proof, but this time for the less-than-equal relation on the rational numbers (Rat or  $\mathbb{Q}$ ) instead.

Here, Rat denotes the rational numbers in Lean, and Rat.le\_trans is the transitivity lemma for  $\leq$  on rational numbers, provided by Mathlib. We import Mathlib to access Rat and le\_trans. Mathlib is the community-driven mathematical library for Lean, containing a large body of formalized mathematics and ongoing development. It is the defacto standard library for both programming and proving in Lean [Com20], we will dig into it as we go along. Notice that we used a function to discharge the universal quantifiers required by transitivity. The underscores indicate unnamed variables that we do not use later. If we had named them, say x y z, then: h1 would be a proof of  $x \leq y$ , h2 would be a proof of  $y \leq z$ , and Rat.le\_trans h1 h2 produces a proof of  $x \leq z$ . The Transitive definition is imported from Mathlib and similarly defined as before.

**Example 3.1** The code can be made more readable using tactic mode. In this mode, you use tactics—commands provided by Lean or defined by users—to carry out proof steps succinctly, avoid code repetition, and automate common patterns. This often yields shorter, clearer proofs than writing the full term by hand.

This proof performs the same steps but is much easier to read. Using by we enter Lean's tactic mode, which (together with the info view) shows the current goal and context. Move your cursor just before by and observe how the info view changes. The goal is initially displayed as  $\vdash$  Transitive fun x1 x2  $\mapsto$  x1  $\leq$  x2. The tactic intro is mainly used to introduce variables and hypotheses corresponding to universal quantifiers and assumptions into the context (essentially deconstructing universal quantifiers and implications). Now position your cursor just before exact and observe the info view again. The goal is now  $\vdash$  x  $\leq$  z, with the context showing the variables and hypotheses introduced by the previous tactic. The exact tactic closes the goal by supplying the term Rat.le\_trans hxy hyz that exactly matches the goal (the specification of Transitive). You can hover over each tactic to see its definition and documentation.

# 3.1 Exploring Mathlib (The Rat structure)

In these examples we cheated and have used predefined lemmas such as Nat.le\_trans and Rat.le\_trans, just to simplify the presentation. We can now dig into the implementation of these lemmas. Let's look at the source code of Rat.le\_trans. The Mathlib 4 documentation website is at https://leanprover-community.

github.io/mathlib4\_docs, and the documentation for Rat.le\_trans is at https://leanprover-community.github.io/mathlib4\_docs/Mathlib/Algebra/Order/Ring/Unbundled/Rat.html#Rat.le\_trans. Click the "source" link there to jump to the implementation in the Mathlib repository. In editors like VS Code you can also jump directly to the definition (Ctrl+click; Cmd+click on macOS). Another way to check source code is by using #print Rat.le\_trans.

```
variable (a b c : Rat)
protected lemma le_trans (hab : a ≤ b) (hbc : b ≤ c) : a ≤ c := by
  rw [Rat.le_iff_sub_nonneg] at hab hbc
  have := Rat.add_nonneg hab hbc
  simp_rw [sub_eq_add_neg, add_left_comm (b + -a) c (-b), add_comm (b +
        -a) (-b), add_left_comm (-b) b (-a), add_comm (-b) (-a),
      add_neg_cancel_comm_assoc, ← sub_eq_add_neg] at this
  rwa [Rat.le_iff_sub_nonneg]
```

The proof uses several tactics and lemmas from Mathlib. The rw or rewrite tactic is very common and sintactically similar to the mathematical practice of rewriting an expression using an equality. In this case, with at, we use it to rewrite the hypotheses hab and hbc using the another Mathlib's lemma Rat.le\_iff\_sub\_nonneg, which states that for any two rational numbers x and y,  $x \leq y$  is equivalent to  $0 \leq y - x$ . Thus we now have the hypotheses tranformerd to:

The have tactic introduces an intermediate result. If you omit a name, Lean assigns it the default name this. In our situation, from hab:  $a \le b$  and hbc:  $b \le c$  we can derive that b - a and c - b are nonnegative, hence their sum is nonnegative:

```
this : 0 \le b - a + (c - b)
```

The most involved step uses simp\_rw to simplify the expression via a sequence of other existing Mathlib's lemmas. The tactic simp\_rw is a variant of simp: it performs rewriting using the simp set (and any lemmas you provide), applying the rules in order and in the given direction. Lemmas that simp can use are typically marked with the @[simp] attribute. This is particularly useful for simplifying algebraic expressions and equations. After these simplifications we obtain:

```
\textbf{this} \;:\; 0 \,\leq\, c \,-\, a
```

Clearly, the proof relies mostly on Rat.add\_nonneg. Its source code is fairly involved and uses advanced features that are beyond our current scope. Nevertheless, it highlights an important aspect of formal mathematics in Mathlib. Mathlib defines Rat as an instance of a linear ordered field, implemented via a normalized fraction representation: a pair of integers (numerator and denominator) with positive denominator and coprime numerator and denominator [Lea25b]. To achieve this, it uses a **structure**. In Lean, a structure is a dependent record (or product type) type used to group together related fields or properties as a single data type. Unlike ordinary records, the type of later fields may depend on the values of earlier ones. Defining a structure automatically introduces a constructor (usually mk) and projection functions that retrieve (deconstruct) the values of its fields. Structures may also include proofs expressing properties that the fields must satisfy.

#### structure Rat where

```
/-- Constructs a rational number from components.
We rename the constructor to 'mk' to avoid a clash with the smart
constructor. -/
mk' ::
/-- The numerator of the rational number is an integer. -/
num : Int
/-- The denominator of the rational number is a natural number. -/
den : Nat := 1
/-- The denominator is nonzero. -/
den_nz : den ≠ 0 := by decide
/-- The numerator and denominator are coprime: it is in "reduced
form". -/
reduced : num.natAbs.Coprime den := by decide
```

In order to work with rational numbers in Mathlib, we use the Rat.mk' constructor to create a rational number from its numerator and denominator, if omitted the default would be Rat.mk. The fields den\_nz and reduced are proofs that the denominator is nonzero and that the numerator and denominator are coprime, respectively. These proofs are automatically generated by Lean's decide tactic, which can solve certain decidable propositions (to be discussed in the next section).

**Example 3.2** Here is how we can define and manipulate rational numbers in Lean.

```
def half : Rat := Rat.mk' 1 2
def third : Rat := Rat.mk' 1 3
```

When working with rational numbers, or more generally with structures, we must provide the required proofs as arguments to the constructor (or Lean must be able to ensure them). For instance Rat.mk' 1 0 or Rat.mk' 2 6 would be rejected. In the case of rationals, Mathlib unfolds the definition through Rat.numDenCasesOn. This principle states that, to prove a property of an arbitrary rational number, it suffices to consider numbers of the form n /. d in canonical (normalized) form, with d > 0 and gcd n d = 1. This reduction allows mathlib to transform proofs about  $\mathbb Q$  into proofs about  $\mathbb Z$  and  $\mathbb N$ , and then lift the result back to rationals.

**Example 3.3** We present a simplified implementation of addition non-negativity for rationals (Rat.add\_nonneg), maintaining a similar approach: projecting everything to the natural numbers and integers first. To illustrate the proof technique clearly, we avoid using existing lemmas from the Rat module in Mathlib. Mathlib is indeed organized into modules by mathematical domain (e.g., Nat, Int, Rat).

We start by defining helper lemmas needed in the main proof. Given a natural number (which in this case represents the denominator of a rational number) that is not equal to zero, we prove it must be positive. This follows directly by applying the Mathlib lemma Nat.pos\_of\_ne\_zero:

The naming convention follows Mathlib best practices aiming to be descriptive by indicating types and properties involved.

The following lemma is slightly more involved. It states that if a rational number (num / den) is non-negative, then its numerator must also be non-negative:

```
lemma rat\_num\_nonneg \{num : \mathbb{Z}\} \{den : \mathbb{N}\} (hden\_pos : 0 < den)
(h : (0 : \mathbb{Q}) \leq num / den) : 0 \leq num := by
contrapose! h
have \ hden\_pos\_to\_rat : (0 : \mathbb{Q}) < den := Nat.cast\_pos.mpr \ hden\_pos
have \ hnum\_neg\_to\_rat : num < (0 : \mathbb{Q}) := Int.cast\_lt.mpr \ h
exact \ div\_neg\_of\_neg\_of\_pos \ hnum\_neg\_to\_rat \ hden\_pos\_to\_rat
```

The lemma requires the denominator to be positive as well as the non-negativity of the rational number, expressed as num / den where the types of num and den are inferred. First, notice the type annotation (0 :  $\mathbb Q$ ). This explicit type annotation on zero forces the entire equation to be casted into rational numbers. Without this annotation, Lean would infer 0 as a natural number by default. However, since the main theorem we are proving concerns rational numbers, we must ensure all comparisons occur in  $\mathbb Q$ . The tactic contrapose! does what you might expect: it proves a statement by contraposition. According to the documentation:

- ullet contrapose turns a goal P ightarrow Q into  $\lnot$  Q ightarrow  $\lnot$  P
- contrapose! turns a goal P o Q into o Q o P and pushes negations inside P and Q using push\_neg
- contrapose h first reverts the local assumption h, then uses contrapose and intro h
- contrapose! h first reverts the local assumption h, then uses contrapose! and intro h

In our case, contrapose! h transforms the goal from proving  $0 \le num$  to assuming num < 0 and proving num / den < 0. We then introduce two local hypotheses. The first, hden\_pos\_to\_rat, proves that the denominator is positive when cast to rationals, using Nat.cast\_pos. The suffix .mpr selects the "modus ponens reverse" direction of the biconditional (the  $\leftarrow$  direction of the  $\leftrightarrow$ ). Next, we introduce hnum\_neg\_to\_rat, which expresses that the numerator is negative when cast to rationals, using Int.cast\_lt with .mpr again. Finally, we apply div\_neg\_of\_neg\_of\_pos, which states that dividing a negative number by a positive number yields a negative result, thus completing the proof by contraposition. Note that we are allowing ourselves to use existing lemmas from Mathlib, such as div\_neg\_of\_neg\_of\_pos from the Field module, but not from the Rat module, to keep the presentation clear and focused on the main proof techniques.

Now we can prove the main result:

```
lemma rat_add_nonneg (a b : Rat) : 0 \le a \to 0 \le b \to 0 \le a + b := by

intro ha hb

cases a with | div a_num a_den a_den_nz a_cop =>
cases b with | div b_num b_den b_den_nz b_cop =>
-- Goal: \vdash 0 \le \uparrow a_num / \uparrow a_den + \uparrow b_num / \uparrow b_den

rw[div_add_div] -- applies the addition formula requiring two new goals
· sorry
· sorry
· sorry
```

We first introduce the two hypotheses ha and hb into the context using intro. As mentioned earlier, a structure can be viewed as a product type or a record type with a single constructor. The tactic cases a with exposes the fields of Rat: the numerator ()a\_num), denominator (a\_den), the proof that the denominator is non-zero (a\_den\_nz), and the coprimality condition (a\_cop). Notice how the goal transforms the rationals a and b into:

```
\vdash 0 \leq \uparrow a\_num / \uparrow a\_den + \uparrow b\_num / \uparrow b\_den
```

variable [Semifield K] {a b d : K}

where  $\uparrow$  denotes type coercion from  $\mathbb{Z}$  or  $\mathbb{N}$  to  $\mathbb{Q}$ . Now we rewrite the goal using rw [div\_add\_div], a theoprem from the Field module, which applies the addition formula for division. Let us briefly examine the source code of this theorem:

```
theorem div_add_div (a : K) (c : K) (hb : b \neq 0) (hd : d \neq 0) : a / b + c / d = (a * d + b * c) / (b * d) := ...
```

The type K here is assumed to be a Semifield. The variable keyword is a way to declare parameters that are potentially used across multiple theorems or definitions. We will explore Lean's powerful algebraic hierarchy and the meaning of the square brackets [] in a later section. Using this rewrite is particularly time-saving, since otherwise one would have to establish the well-definedness of rational addition in terms of the underlying structure (a non-trivial task). This theorem requires proofs (hb: b \neq 0) and (hd: d \neq 0), generating two additional side goals. We handle each goal separately using the focusing bullet. The first bullet addresses the main goal (proving the sum is non-negative), while the subsequent bullets discharge the non-zero denominator conditions. I have omitted the actual proofs, here, using sorry, which we haven't mentioned before. sorry is a useful feature of Lean that tells the system to accept an incomplete proof for the time being, allowing you to continue development without proving every detail immediately. We can now tackle the remaining goals:

```
\cdot -- Goal: \vdash 0 \leq (\uparrow a\_num * \uparrow b\_den + \uparrow a\_den * \uparrow b\_num) / (\uparrow a\_den * \uparrow b\_den)
  have hnum\_nonneg : (0 : \mathbb{Q}) \leq a\_num * b\_den + a\_den * b\_num := by
    have ha_num_nonneg := by
       have ha_den_pos := nat_ne_zero_pos a_den a_den_nz
       exact rat_num_nonneg ha_den_pos ha
    have hb_num_nonneg := by
       have hb_den_pos := nat_ne_zero_pos b_den b_den_nz
       exact rat_num_nonneq hb_den_pos hb
     apply add_nonneg -- works for any OrderedAddCommMonoid
     · apply mul_nonneg -- works for any OrderedSemiring
       \cdot exact Int.cast_nonneg.mpr ha_num_nonneg
       \cdot exact Nat.cast_nonneg b_den
    · apply mul_nonneg
       \cdot exact Nat.cast_nonneg a_den
       · exact Int.cast_nonneg.mpr hb_num_nonneg
  have hden_nonneg : (0 : \mathbb{Q}) \leq a_den * b_den := by
    rw [\leftarrow Nat.cast_mul]
     exact Nat.cast_nonneg (a_den * b_den)
  exact div_nonneg hnum_nonneg hden_nonneg
\cdot exact Nat.cast_ne_zero.mpr a_den_nz -- Goal: \vdash \uparrowa_den \neq 0
\cdot exact Nat.cast_ne_zero.mpr b_den_nz -- Goal: \vdash \uparrow b\_den \neq 0
```

We introduce two key hypotheses, hnum\_nonneg and hden\_nonneg, which will be required by div\_nonneg from the Group With Zero module. This lemma provides us with a term that directly validates our statement. Note that div\_nonneg is a generalized lemma that applies not only to rational numbers but to all ordered groups with zero that are also partially ordered. The hypothesis hnum\_nonneg proves that the numerator is non-negative by working with the coerced expressions in  $\mathbb{Q}$ . It uses add\_nonneg and mul\_nonneg, which are general theorems that work for any ordered additive commutative monoid and ordered semiring, respectively. The actual reasoning is done using integer-related theorems (via Int.cast\_nonneg) for the numerators and natural number theorems (via Nat.cast\_nonneg) for the denominators. The hypothesis hden\_nonneg proves that the denominator is nonnegative by working entirely with natural numbers. We use the rewrite rw [ $\leftarrow$ Nat.cast\_mul], which moves the coercion (in this case from  $\mathbb{N}$  to  $\mathbb{Q}$ ) inside the multiplication:  $\uparrow (m * n) = \uparrow m * \uparrow n$ . The  $\leftarrow$  symbol means that we want the transformation from right to left (i.e., we apply the equality in reverse to move the cast inward). Type casts and coercions require these kinds of rewrite rules, not only for multiplication but also for addition and other operations, and similarly for Z or other numerical types. These lemmas, such as Nat.cast\_mul, Int.cast\_add, etc., ensure that algebraic operations commute with type coercions.

## 3.2 Coercions and Type Casting

We extensively used type casting and coercions in this proof, which requires some explanation [LM20]. Lean's type system lacks subtyping, means that types like  $\mathbb{N}$ ,  $\mathbb{Z}$ , and  $\mathbb{Q}$  are distinct and do not have a subtype relationship. In order to translate between these types, we need to use explicit type casts or rely on automatic coercions. For example, natural numbers ( $\mathbb{N}$ ) can be coerced to integers ( $\mathbb{Z}$ ), and integers can be coerced to rational numbers ( $\mathbb{Q}$ ). Casting and coercion are related but distinct concepts:

- Casting refers to the explicit conversion of a value from one type to another, typically using functions like Int.cast or Nat.cast. These functions have accompanying lemmas that preserve properties across type conversions, such as Int.cast\_lt and Nat.cast\_pos.
- Coercion, on the other hand, is a more general mechanism that allows Lean to automatically convert between types when needed. More generally, in expressions like x + y where x and y are of different types, Lean will automatically coerce them to a common type. For example, if  $x : \mathbb{N}$  and  $y : \mathbb{Z}$ , then x will be coerced to  $\mathbb{Z}$ .

The notation  $\uparrow$  denotes an explicit coercion (in between cast and coercion). To illustrate the expected behavior of coercion simplification, consider the expression  $\uparrow m + \uparrow n < (10 : \mathbb{Z})$ , where m,  $n : \mathbb{N}$  are cast to  $\mathbb{Z}$ . The expected normal form is  $m + n < (10 : \mathbb{N})$ , since +, <, and the numeral 10 are polymorphic (i.e., they can work with any numerical type such as  $\mathbb{Z}$  or  $\mathbb{N}$ ). The simplification should proceed as follows:

- 1. Replace the numeral on the right with the cast of a natural number:  $\uparrow m + \uparrow n < \uparrow (10 : \mathbb{N})$
- 2. Factor  $\uparrow$  to the outside on the left:  $\uparrow$  (m + n) <  $\uparrow$  (10 : N)
- 3. Eliminate both casts to obtain an inequality over  $\mathbb{N}$ :  $m + n < (10 : \mathbb{N})$

Lean provides tactics like norm\_cast to simplify expressions involving such coercions. The norm\_cast tactic normalizes casts by pushing them outward and eliminating redundant coercions, often simplifying proofs significantly by reducing goals to their "native" types.

## 3.3 Type Classes and Algebraic Hierarchy in Lean

In our proof of rat\_add\_nonneg, we used many generalized lemmas from Mathlib, such as add\_nonneg, mul\_nonneg, and div\_nonneg, which apply to a wide range of types beyond just rational numbers. Similarly, in our earlier work with natural numbers, we used Nat.le\_trans, a theorem specifically for natural numbers that is part of Lean's core library (lean/Init/Prelude.lean). Mathlib is built on top of this base library. However, the transitivity property holds not only for naturals but also for integers, reals, and, in fact, for any partially ordered set.

Rather than duplicating this theorem for each type, Mathlib provides a general lemma <code>le\_trans</code> that works for any type  $\alpha$  endowed with a partial ordering. This is achieved through **type classes**, Lean's mechanism for defining and working with abstract algebraic structures in an ad hoc polymorphic manner. Type classes provide a powerful and flexible way to specify properties and operations that can be shared across different types, thereby enabling polymorphism and code reuse. Ad hoc polymorphism arises when a function or operator is defined over several distinct types, with behavior that varies depending on the type. A standard example [WB89] is overloaded multiplication: the same symbol \* denotes multiplication of integers (e.g., 3 \* 3) and of floating-point numbers (e.g., 3.14 \* 3.14). By contrast, parametric polymorphism occurs when a function is defined over a range of types but acts uniformly on each of them. For instance, the List.length function applies in the same way to a list of integers and to a list of floating-point numbers.

Under the hood, a type class is a structure. An important aspect of structures, and hence type classes, is that they support hierarchy and composition through inheritance. For example, mathematically, a monoid is a semigroup with an identity element, and a group is a monoid with inverses. In Lean, we can express this by defining a Monoid structure that extends the Semigroup structure, and a Group structure that extends the Monoid structure using the extends keyword:

```
-- A semigroup has an associative binary operation structure Semigroup (\alpha: Type*) where mul: \alpha \to \alpha \to \alpha mul_assoc: \forall a b c: \alpha, mul (mul a b) c = mul a (mul b c) -- A monoid extends semigroup with an identity element structure Monoid (\alpha: Type*) extends Semigroup \alpha where one: \alpha one_mul: \forall a: \alpha, mul one a = a mul_one: \forall a: \alpha, mul a one = a -- A group extends monoid with inverses structure Group (\alpha: Type*) extends Monoid \alpha where inv: \alpha \to \alpha mul_left_inv: \forall a: \alpha, mul (inv a) a = one
```

The symbol \* in ( $\alpha$ : Type\*) indicates a universe variable (we will discuss universes later). Sometimes, to avoid inconsistencies between types (such as Girard's paradox), universes must be specified explicitly. This is an example of universe polymorphism. Thus we have now seen all the polymorphism flavors in Lean: parametric, ad hoc, and universe polymorphism.

Type classes are defined using the class keyword, which is syntactic sugar for defining a structure. Thus, the previous example can be rewritten using type classes:

```
-- A semigroup has an associative binary operation class Semigroup (\alpha: Type*) where mul: \alpha \to \alpha \to \alpha mul_assoc: \forall a b c: \alpha, mul (mul a b) c = mul a (mul b c) -- A monoid extends semigroup with an identity element class Monoid (\alpha: Type*) extends Semigroup \alpha where one: \alpha one_mul: \forall a: \alpha, mul one a = a mul_one: \forall a: \alpha, mul a one = a -- A group extends monoid with inverses class Group (\alpha: Type*) extends Monoid \alpha where inv: \alpha \to \alpha mul_left_inv: \forall a: \alpha, mul (inv a) a = one
```

The main difference is that type classes support **instance resolution**. We use the keyword **instance** to declare that a particular type is an instance of a type class, which inherits the properties and operations defined in the type class. Instances can be automatically inferred by Lean's type inference system, allowing for concise and expressive code. For example, we can declare that  $\mathbb Z$  is a group under addition:

```
instance : Group Z where
  mul := Int.add
  one := 0
  inv := Int.neg
  mul_assoc := Int.add_assoc
  one_mul := Int.zero_add
  mul_one := Int.add_zero
  mul_left_inv := Int.neg_add_cancel
```

Now, any theorem proven for an arbitrary Group  $\alpha$  automatically applies to  $\mathbb Z$  without any additional work. This mechanism is particularly useful for defining and working with order structures like preorders and partial orders. Mathematically, a preorder consists of a set P and a binary relation  $\leq$  on P that is reflexive and transitive [Lea25a].

```
-- A preorder is a reflexive, transitive relation '≤' with '<' defined in terms of '≤'

class Preorder (α: Type*) extends LE α, LT α where

le_refl: ∀ a: α, a ≤ a

le_trans: ∀ a b c: α, a ≤ b → b ≤ c → a ≤ c

lt:= fun a b ⇒ a ≤ b ∧ ¬ b ≤ a

lt_iff_le_not_ge: ∀ a b: α, a < b ↔ a ≤ b ∧ ¬ b ≤ a:= by intros;

rfl

instance [Preorder α]: Lean.Grind.Preorder α where

le_refl:= Preorder.le_refl

le_trans:= Preorder.le_trans____

lt_iff_le_not_le:= Preorder.lt_iff_le_not_ge___

Listing 3: Preorder Type Class in Lean
```

Listing 3: Preorder Type Class in Lean

The class Preorder declares a type class over a type  $\alpha$ , bundling the  $\leq$  and < relations (inherited via extends LE  $\alpha$ , LT  $\alpha$ ) with the preorder axioms: reflexivity (le\_refl) and transitivity (le\_trans). The field 1t provides a default definition of strict inequality in terms of  $\leq$ , and the theorem lt\_iff\_le\_not\_ge characterizes this relationship, proved automatically via reflexivity (by intros; rfl). The instance declaration connects the Preorder class to Lean's Grind tactic automation, which allows automatic reasoning with preorder properties during proof search. Returning to our rational number proof, this explains why lemmas like add\_nonneg and mul\_nonneg work seamlessly:  $\mathbb Q$  is an instance of OrderedSemiring, which extends Preorder and other algebraic structures, automatically providing all their theorems. (Cocnlude the discussion of less or equal is trnaitive)

An important aspect of type classes is that they can be parameterized by other type classes, enabling the construction of complex hierarchies of algebraic structures. Topological spaces in Mathlib are represented using type classes and they can be derived metric spaces and normed spaces. For example, the TopologicalSpace type class defines the structure of a topological space, and it can be extended to define metric spaces and normed spaces, which are topological spaces with additional structure. This hierarchical organization allows for the reuse of definitions and theorems across different mathematical domains, facilitating a modular and scalable approach to formalization. (Little Intro on THe TOPOLOGY CLass, Continuos CLosure ecc.)

Topology in Lean is built upon the concept of open sets, defined using the IsOpen predicate. A topological space is a type equipped with a collection of open sets satisfying the standard axioms. In Lean, this is represented using the TopologicalSpace type class:

```
class TopologicalSpace (\alpha : Type*) where
IsOpen : Set \alpha \to \text{Prop}
isOpen_univ : IsOpen univ
isOpen_inter : \forall s t, IsOpen s \to IsOpen t \to IsOpen (s \cap t)
isOpen_sUnion : \forall s, (\forall t \in s, IsOpen t) \to IsOpen (sUnion s)
```

Key topological concepts used in our formalization include:

**Connectedness:** A space X is connected if it cannot be written as the union of two disjoint non-empty open sets. Equivalently, every continuous function from X to a discrete space is constant. In Lean, this is formalized as IsConnected: Set  $\alpha \to \text{Prop}$ .

**Path-connectedness:** A space X is path-connected if for any two points  $x,y \in X$ , there exists a continuous path  $\gamma:[0,1]\to X$  with  $\gamma(0)=x$  and  $\gamma(1)=y$ . In Lean, the unit interval [0,1] is represented as the type unitInterval, and path-connectedness is formalized as IsPathConnected: Set  $\alpha\to \text{Prop}$ .

Continuous functions: A function  $f: X \to Y$  between topological spaces is continuous if the preimage of every open set is open. Lean provides Continuous f: Prop, as well as ContinuousAt and ContinuousOn for local versions.

Closure: The closure of a set S is the smallest closed set containing S, equivalently the set of all limit points of S. In Lean, this is closure: Set  $\alpha \to \text{Set } \alpha$ .

(Maybe include part of the REal construcion / classical noncomputable ecc ecc  $^{\prime}$ )

# 4 Example in Topology: A Connected but not Path-Connected Space

As part of my thesis work, with the help and revision from Prof David Loeffler, I have formalized a well-known counterexample in topology: the **topologist's sine curve**. This classic example illustrates a space that is **connected** but not **path-connected**. It demonstrates the use of type classes, structures, and several tactics in Lean that were discussed earlier. In my presentation, I will provide a high-level overview along with some specific examples, of my original code, wich is way longer than the final result. My original proof follows Conrad's paper ([Con]), with a few modifications and some differences from the final formalization **Counterexamples** – **Topologist's Sine Curve**. The topologist's sine curve is defined as the graph of  $y = \sin(1/x)$  for  $x \in (0, \infty)$ , together with the origin (0,0). We define three sets in  $\mathbb{R}^2$ :

• S: the oscillating curve  $\{(x, \sin(1/x)) : x > 0\}$ 

- Z: the singleton set  $\{(0,0)\}$
- T: their union  $S \cup Z$

In Lean, this is expressed as follows:

```
open Real Set def pos_real := Ioi (0 : \mathbb{R})
noncomputable def sine_curve := fun x \mapsto (x, sin (x<sup>-1</sup>))
def S : Set (\mathbb{R} \times \mathbb{R}) := sine_curve '' pos_real
def Z : Set (\mathbb{R} \times \mathbb{R}) := { (0, 0) }
def T : Set (\mathbb{R} \times \mathbb{R}) := S \cup Z
```

We open the Real and Set namespaces to avoid prefixing real number and set operations with Real. and Set., respectively. We define the interval  $(0,\infty)$  as pos\_real, using the predefined notation Ioi 0, from Set. The function sine\_curve maps a positive real number to a point on the topologist's sine curve in  $\mathbb{R}^2$ . Here, '' denotes the image of a set under a function. It's noncomputable because it involves the sine function, which is not computable in Lean's core logic. The sets S, Z, and T are defined using set operations. , and { (0, 0) } denotes the singleton set containing the point (0,0). Set is the type of sets, defined as predicates (i.e., functions from a type to Prop). The sets are subsets of the product space  $\mathbb{R}^2$ , represented as  $\mathbb{R} \times \mathbb{R}$ . The sin function sin is defined in the Real.

The goal is to prove that T is connected but not path-connected. Let's start with connectedness.

## 4.1 T is connected

First of all one can directly see that S is connected, since it is the image of the set  $((0,\infty))$  under the continuous map  $x\mapsto (x,\sin(1/x))$  and a interval in  $\mathbb R$  is connected. Moreover, the closure of S is connected, and every set in between a connected set and its closure are connected. Since T is contained in the closure of S, T is connected. This is how a mathematician would argue informally, using known facts. However, in a formal proof, one must justify each step. For instance, justifying that S is connected requires proving that the map  $x\mapsto (x,\sin(1/x))$  is continuous on  $(0,\infty)$  and that  $(0,\infty)$  is connected.

As we have seen, even showing that a rational number is non-negative requires several steps and the use of various lemmas from Mathlib. Similarly, proving that a set is connected can involve multiple steps for the newer programmer.

We can use the structure IsConnected to set up the statement and see if we can argue similarly in Lean.

```
lemma S_is_conn : IsConnected S := by sorry
```

In the file where IsConnected is defined, Topology/Connected/Basic.lean, we see that it requires S to be nonempty and preconnected. You can verify this by unfolding IsConnected in the goal.

Following the definition of IsPreconnected, we see that it captures the usual definition of preconnectedness: that S cannot be partitioned into two nonempty disjoint open sets. This trivially requires nonemptiness to make sense. The unfold tactic helps to expand definitions; one can use it to expand the definition of S or pos\_real defined before, as well as other Mathlib expressions. Reflecting our argument, we can check if Mathlib includes the fact that every interval is connected and that connectedness is preserved under continuous maps. Indeed, in Topology/Connected/Interval.lean, we find the theorem isConnected\_Ioi.image, stating that the image of an interval of the form  $(a, \infty)$  under a continuous map is connected.

The apply tactic applies the theorem similar to exact, the latter tries to close the goal with rfl. The theorem isConnected\_Ioi.image requires proving the continuity of the map on the interval  $(0,\infty)$ , which is expressed as ContinuousOn sine\_curve (Ioi O). The predicate ContinuousOn f S expresses that a function f is continuous on a set S, which is what we need to prove now. The function  $x\mapsto (x,\sin(1/x))$  is continuous on  $(0,\infty)$  as the product of two functions continuous on the given domain: the identity map  $x\mapsto x$  and the map  $x\mapsto\sin(1/x)$ .

Here is the full proof in Lean:

Starting from the bottom lemma, ContinuousOn.prodMk states that the product of two functions continuous on a set is continuous on that set, requiring a proof of the continuity of each component. The first component is the identity map, which is continuous on any set. Mathlib provides continuous\_id.continuousOn for this purpose. The second component is the composition of the sine function with the inverse function. The sine function is continuous everywhere, and for this we can use continuous\_sin. The method comp\_continuousOn is accessible from the fact that continuous\_sin gives an instance of a continuous map and is generalized in the ContinuousOn module. The theorem Continuous.comp\_continuousOn states that the composition of a continuous function with a function that is continuous on a set is continuous on that set, and requires proof of the continuity on the set of the inner function. We separate the proof that the inverse function is continuous on the positive reals into the auxiliary lemma inv\_is\_continuous\_on\_pos\_real. The theorem continuousOn\_invo states that if a function is continuous and non-zero on a set, then its inverse is continuous on that set. The continuity of the identity map is proved as before. The second argument requires proving that  $x \neq 0$  for all x in  $(0, \infty)$ .

```
· intro x hx
exact ne_of_gt hx
```

The hypothesis hx states that x is in  $(0, \infty)$ , which implies that x > 0. The theorem ne\_of\_gt states that if a real number is greater than zero, then it is non-zero, which completes the proof. Thus the final proof goes as follows:

```
lemma S_is_conn : IsConnected S := by
apply isConnected_Ioi.image
    exact sin_comp_inv_is_continuous_on_pos_real
```

When writing a proof, one starts by working out the informal argument on paper. Then one tries to translate it into Lean, step by step, looking for theorems in Mathlib. Afterwards, one can try to optimize the proof by removing unnecessary steps or refactoring it. Proving properties like continuity and connectedness is very common, and there are obviously ways to achieve this with less work. Let's showcase a refactoring of the entire proof. First, the auxiliary lemmas can be reduced to one-liners.

We removed the by keyword since we can provide a **term** that directly proves the statement. In inv\_is\_continuous\_on\_pos\_real, we directly apply ContinuousOn. inv\_0 with the two required arguments. Notice that we can use a lambda function fun\_hx => ne\_of\_gt hx to prove that  $x \neq 0$  for all x in  $(0, \infty)$  (recall the propositions-as-types correspondence). In the next lemma, we use the <\| reverse application operator, which allows us to avoid parentheses by changing the order of application. This means that f < g < |h| is equivalent to f(gh). We can inline these two lemmas into the main proof to get a final one-liner:

Notice again the use of the **pipe** operator. Reading from left to right, we are building up the proof by successive applications:

- We start with isConnected\_Ioi.image sine\_curve, which states that the image of  $(0, \infty)$  under sine\_curve is connected if we can prove the function is continuous.
- We then apply continuous\_id.continuousOn.prodMk, which constructs the product of two continuous functions.
- Next, continuous\_sin.comp\_continuousOn provides the continuity of the sine composition.
- Finally, ContinuousOn.inv<sub>0</sub> continuous\_id.continuousOn (fun \_ hx => ne\_of\_gt hx) proves the continuity of the inverse function on positive reals.

The entire chain can be read as building the continuity proof from the innermost function (the inverse) outward to the complete sine curve function, which is then used to prove that S is connected.

Since the intersection of Z and S is empty, we cannot directly conclude that T is connected from the connectedness of its components alone. However, we can use the fact that every subset between a connected set and its closure is connected.

**Theorem 4.1** Let C be a connected topological space, and denote  $\overline{C}$  as its closure. It follows that every subset  $C \subseteq S \subseteq \overline{C}$  is connected.

In Mathlib, this theorem is available as IsConnected.subset\_closure. We can set up the statement and progress from there.

```
theorem T_is_conn : IsConnected T := by apply IsConnected.subset_closure  
• exact S_is_conn -- \vdash IsConnected ?s  
• tauto_set -- \vdash S \subseteq T  
• sorry -- \vdash T \subseteq closure S
```

The theorem requires three goals:

- 1. That S is connected, which was already proved in  $S_{is\_conn.}$
- 2. That  $S \subseteq T$ , which is a trivial set operation. The tactic tauto\_set handles this kind of set tautologies.
- 3. That  $T \subseteq \overline{S}$  (the closure of S), which requires proof.

Let's continue with the final point.

```
lemma T_sub_cls_S : T ⊆ closure S := by
intro x hx
cases hx with
| inl hxS => exact subset_closure hxS
| inr hxZ =>
sorry
```

Proving that one set is contained in another can be done naively in a pointwise manner. We introduce an element  $x \in \mathbb{R}^2$  together with the proof that  $x \in T$ . Since T is a union, we use cases to separate the two cases. When  $x \in S$ , the goal is trivially solved by exact subset\_closure hxS. The case where  $x \in Z$ , requires more work.

Now a trick. One of the most painful issue in formalizing math in Lean is the use of existing theorems. One can use several ways to look for the exact theorem. Let's try using the apply? tactic to see what the infoview suggests:

```
lemma T_sub_cls_S : T ⊆ closure S := by
intro x hx
cases hx with
| inl hxS => exact subset_closure hxS
| inr hxZ =>
apply?
sorry
```

Depending on the previous work in the file, Lean can already unify the goal with available theorems and suggest the next step. Similar tactics include:

- exact? for finding an exact match to close the goal
- rw? for suggesting rewrites and definitionally equal replacements

• simp? for suggesting simplifications

Another useful tool is Loogle (similar to Haskell's Hoogle), which helps you find theorems by their type signature or name patterns. You can access it at https://loogle.lean-lang.org/ or use it directly in VS Code. The best approach is, anyway, to think first about how you would tackle the problem on a piece of paper, as mentioned earlier. Since we are working with a topology on  $\mathbb{R}$ , we know that this is a metrizable topology, therefore it is induced by the metric space structure on  $\mathbb{R}$ . Thus, we can expand our toolkit by working with Lean's MetricSpace module, which provides specialized tools for reasoning about metric spaces, such as balls, distances, and metric-specific characterizations of continuity and convergence. We know that the closure of a set contains all its limit points. To show that the point (0,0) is contained in the closure of S, we need to show that it is a limit point of S. Thus, one can define a sequence in S tending to (0,0), and we are done. At this point, apply? suggests several ways to proceed, involving new symbols such as  $\mathcal{N}$  (neighborhoods) and f (eventually/frequently). For example:

```
Try this: refine Frequently.mem_closure ?_ Remaining subgoals: \vdash \exists^f \ (\mathtt{x} : \mathbb{R} \times \mathbb{R}) \ \text{in} \ \mathcal{N} \ \mathtt{x}, \ \mathtt{x} \in \mathtt{S} or the more familiar metric space approach:  \text{Try this: refine Metric.mem\_closure\_iff.mpr ?\_}  Remaining subgoals: \vdash \forall \ \varepsilon > \mathtt{0}, \ \exists \ \mathtt{b} \in \mathtt{S}, \ \mathtt{dist} \ \mathtt{x} \ \mathtt{b} < \varepsilon
```

While we could work with metric space properties directly using the familiar  $\varepsilon$ - $\delta$  formulation, we instead introduce and explain filters. This approach may seem more abstract initially, but it provides a more general and powerful framework that works uniformly across all topological spaces, not just metric spaces. Moreover, once understood, filters often make proofs more concise and elegant by allowing us to reason at a higher level of abstraction.

### 4.1.1 Limits and Convergence with Filters

As seen before, continuity is defined in the Topology type class. One can create their own epsilon-delta definition given a notion of distance. Similarly, defining the convergence of a sequence in  $\mathbb{R}$  can be done manually as follows:

Where s is a sequence and a is the limit point. However, there are many other types of limits to consider, like lmits of a function at a point, limits at infinity (from above or below), one-sided limits (from the left or right) and so on. Defining each of these separately would require a huge amount of work to include in Mathlib, with significant duplication of theorems and proofs. Moreover, many fundamental theorems (like the characterization of continuity via limits)

would need to be reproved for each type of limit. Fortunately, Bourbaki solved this issue by introducing the notion of **filters** to unify all concepts of limits, convergence, neighborhoods and terms like eventually or frequently often into a single framework. Mathlib adopts this notion to achieve an elegant solution that fully covers the entire landscape of limit-related concepts. Intuitively, a filter represents a notion of "sufficiently large" subsets. More formally, a filter F on a type X is a collection of subsets of X satisfying three axioms:

- 1. Non-emptiness:  $X \in F$  (the whole space is in the filter)
- 2. **Upward closure:** If  $U \in F$  and  $U \subseteq V$ , then  $V \in F$  (supersets of "large" sets are "large")
- 3. Intersection closure: If  $U, V \in F$ , then  $U \cap V \in F$  (finite intersections of "large" sets are "large")

We are going to use some of the following concetps:

- At top filter atTop: Filter  $\mathbb{N}$ : Contains sets that include all sufficiently large natural numbers. Formally,  $U \in \mathtt{atTop}$  if and only if there exists N such that  $\{n \mid n \geq N\} \subseteq U$ . This captures the idea of " $n \to \infty$ ."
- Neighborhood filter  $\mathcal{N}$  x: In a topological space, this filter contains all neighborhoods of the point x. A set is in  $\mathcal{N}$  x if it contains an open set containing x. This captures the idea of "near x."
- $\forall^f$  x in f, p x (f.Eventually p): "Eventually in filter f, property p holds." This means there exists some set  $U \in f$  such that p holds for all  $x \in U$ . For example,  $\forall^f$  n in atTop, n > 100 means "for all sufficiently large n, we have n > 100."
- $\exists^f$  x in f, p x (f.Frequently p): "Frequently in filter f, property p holds." This means for every set  $U \in f$ , there exists some  $x \in U$  where p holds. This captures the idea that p holds "infinitely often" or "arbitrarily close." For example,  $\exists^f$  n in atTop, Even n means "there are arbitrarily large even numbers."
- Tendsto f  $1_1$   $1_2$ : "Function f tends from filter  $l_1$  to filter  $l_2$ ." This is used for convergence.

Example 4.2 Here some examples of the use of filter in Lean.

- Sequence convergence: Tends to s at Top (N a) means  $s_n \to a$  as  $n \to \infty$
- Function limits: Tendsto f (N x) (N y) means  $f(x') \to y$  as  $x' \to x$
- Continuity: A function is continuous at x iff Tendsto f ( $\mathcal{N}$  x) ( $\mathcal{N}$  (f x)) (does this remember the continuity in terms of open sets?)

Using filters, we can prove that  $T \subseteq \overline{S}$  by showing that the origin is a limit point of S. We construct a sequence  $f: \mathbb{N} \to \mathbb{R}^2$  in S converging to (0,0) using the Tendsto framework:

```
lemma T_sub_cls_S : T ⊆ closure S := by
  intro x hx
  cases hx with
  | inl hxS => exact subset_closure hxS
  inr hxZ =>
      rw [hxZ]
       -- Define sequence: f(n) = (1/(n\pi), 0)
      let f : \mathbb{N} \to \mathbb{R} \times \mathbb{R} := \text{fun n} \Rightarrow ((n * \text{Real.pi})^{-1}, 0)
      -- Show f converges to (0, 0)
      have hf: Tendsto f atTop (\mathcal{N} (0, 0)) := by
         refine Tendsto.prodMk_nhds ?_ tendsto_const_nhds
         exact tendsto_inv_atTop_zero.comp
           (Tendsto.atTop_mul_const' Real.pi_pos
    tendsto_natCast_atTop_atTop)
       -- Show f eventually takes values in S
      have hf' : \forall^f n in atTop, f n \in S := by
         filter_upwards [eventually_gt_atTop 0] with n hn
         exact ((n * Real.pi)^{-1},
           inv_pos.mpr (mul_pos (Nat.cast_pos.mpr hn) Real.pi_pos),
           by simp [f, sine_curve, inv_inv, Real.sin_nat_mul_pi]
       -- Apply sequential characterization of closure
      exact mem_closure_of_tendsto hf hf'
```

The proof is already reduced as much as possible. Let's break down what's happening in without getting into details. Using let, we define  $f(n) = (\frac{1}{n\pi}, 0)$ , which we will show converges to (0,0) and stays in S.

- 1. Convergence proof (hf): We show Tendsto f atTop ( $\mathcal{N}$  (0, 0)).
  - We use Tendsto.prodMk\_nhds to split the product: we need to show the first coordinate tends to 0 and the second is constantly 0.
  - For the first coordinate, we compose tendsto\_inv\_atTop\_zero (which states  $\frac{1}{x} \to 0$  as  $x \to \infty$ ) with the fact that  $n\pi \to \infty$ .
  - The second constant coordinate is handled by tendsto\_const\_nhds.
- 2. Membership proof (hf'): We show  $\forall^f$  n in atTop, f n  $\in$  S, meaning  $f(n) \in S$  for all sufficiently large n.
  - We use filter\_upwards, which allows us to combine hypotheses about properties that hold eventually to prove another property holds eventually. Here, we combine it with eventually\_gt\_atTop 0, which states that eventually n > 0.
  - For such n, we show  $f(n) = \left(\frac{1}{n\pi}, 0\right)$  is in S by noting that the second term is:

$$\sin\left(\frac{1}{\left(\frac{1}{n\pi}\right)}\right) = \sin(n\pi) = 0.$$

Finally, mem\_closure\_of\_tendsto combines these facts: if a sequence eventually stays in S and converges to x, then x is in the closure of S.

## 4.1.2 Finalising the first part of the proof

If you are a one-liner enthusiast like me, you don't mind trying to combine bits and pieces to get a clean final result. We can simplify the final theorem as follows initially:

```
theorem T_is_conn : IsConnected T :=
   IsConnected.subset_closure S_is_conn (by tauto_set) T_sub_cls_S
```

The second argument is still in tactic mode with by tauto\_set, but it looks clean and we can keep it as is. With a bit of courage, we can also inline the proof of S\_is\_conn (while T\_sub\_cls\_S is way too long to inline) to get a more self-contained one-liner:

```
theorem T_is_conn : IsConnected T :=
   IsConnected.subset_closure (isConnected_Ioi.image sine_curve <|
      continuous_id.continuousOn.prodMk <|
      Real.continuous_sin.comp_continuousOn <|
      ContinuousOn.invo continuous_id.continuousOn
      (fun _ hx => ne_of_gt hx)) (by tauto_set) T_sub_cls_S
```

Making these amendments is not only for the sake of shortening the proof. Lean will, obviously, compile the proof faster by not entering tactic mode or using multiple tactics. Tactics internally hide many operations they automatically perform to close the goal. Moreover, if we directly provide a term for the proof, Lean will infer and unify everything by definitional equality; remember the very first example we did. By providing explicit proof terms, we give Lean less work to do, making the proof more transparent and efficient. This practice of "golfing" is essential in a huge library such as Mathlib community that needs to balance performance and maintainability. From now on the rest of th code will be presented in it's reduced form. Here is the link to the entire first part of the proof: [link to Lean live]

Note 4.3 The proof merged into the Mathlib library takes Z as  $\{0\} \times [-1,1]$  instead of the singleton  $\{(0,0)\}$ . This, together with the fact that T equals the closure of S, yields a stronger and more general result. The Mathlib version demonstrates that the entire vertical segment at x=0 lies in the closure of the oscillating curve, providing a more complete characterization of the topologist's sine curve. This stronger version shows that a closed set (specifically, the closure of S) can be connected but not path-connected. Showing that forming closure can destroy the property of path connectedness for subsets of a topological space.

## 4.2 T is not path-connected

The main and most substantial part is showing that T is not path-connected. Showing this informally already requires constructing and pointing out various steps in order to convince an ideal reader. One can argue informally by contradiction: suppose a path exists in the topologist's sine curve T connecting a point in S to a point in S. As the path approaches the y-axis (where  $x \to 0$ ),

the y-coordinate must oscillate infinitely between -1 and 1 due to the behavior of  $\sin(1/x)$  as  $x \to 0^+$ . This infinite oscillation contradicts the continuity of the path, which is a fundamental requirement for path-connectedness. To be more precise, we need to construct a sequence that it eventually oscillates, establishing the contradiction. We start by setting up the theorem:

```
theorem T_is_not_path_conn : ¬ (IsPathConnected T) :=
by sorry
```

In mathematics, we normally define a path-connected space as follows.

**Definition 4.4** A topological space X is said to be path-connected if for every two points  $a, b \in X$ , there exists a path, i.e., a continuous map  $p : [0,1] \to X$  such that p(0) = a and p(1) = b.

The interval [0,1] is the standard choice for the domain of paths. In Mathlib, PathConnectedSpace X is a type class that asserts the entire topological space X is path-connected, while IsPathConnected S is a predicate used to infer that a subset S of a topological space is path-connected.

The auxiliary predicate JoinedIn is defined as:

where Path x y denotes a continuous map  $\gamma:[0,1]\to X$  with  $\gamma(0)=x$  and  $\gamma(1)=y$ . Mathlib uses unitInterval as the standard definition for [0,1] in constructions such as the definition of a path.

Now let's start with the first part of the proof:

```
theorem T_is_not_path_conn : ¬ (IsPathConnected T) := by
   -- Assume we have a path from z = (0, 0) to w = (1, sin(1))
have hz : z ∈ T := Or.inr rfl
have hw : w ∈ T := Or.inl ⟨1, ⟨zero_lt_one' ℝ, rfl⟩⟩
intro p_conn
apply IsPathConnected.joinedIn at p_conn
specialize p_conn z hz w hw
let p := JoinedIn.somePath p_conn
```

We introduce two points: z = (0,0) and  $w = (1,\sin(1))$ , and prove they are both in T. Using intro p\_conn, we assume that T is path-connected. Notice that the goal is now False, meaning we must find a contradiction. The last three lines extract an explicit path p connecting z and w:

- apply IsPathConnected.joinedIn at p\_conn transforms the path-connectedness assumption into the statement that any two points in T are joined.
- specialize  $p_{conn}$  z hz w hw specializes this to our specific points z and w.
- let p := JoinedIn.somePath p\_conn extracts a concrete path from the existential statement.

Conrad's paper ([Con]) defines a time  $t_0 \in [0,1]$  as the first time the path p jumps from (0,0) to the graph of  $\sin(1/x)$ , where the x-coordinate map  $(x : \mathbb{R}^2 \to \mathbb{R})$  of p is positive.

$$t_0 = \inf\{t \in [0,1] : x(p(t)) > 0\}$$

The argument then uses the continuity of the x-coordinate map composed with the path p. By continuity at  $t_0$ , we can find a neighborhood around  $t_0$  where the path stays close to (0,0). Specifically, with  $\varepsilon=1/2$ , there exists  $\delta>0$  such that for all t with  $|t-t_0|<\delta$ , we have  $||p(t)-p(t_0)||<1/2$ . We want to show the oscillating behavior around (0,0) indeed. To simplify some steps, we instead define

$$t_0 = \sup\{t \in [0,1] : x(p(t)) = 0\}$$

to be the last time the path remains at (0,0). The same continuity argument applies with this definition.

```
-- Consider the composition of the x-coordinate map with p, which is
     continuous
have xcoord_pathcont : Continuous fun t \mapsto (p t).1 :=
     continuous_fst.comp p.continuous
-- Let t_0 be the last time the path is on the y-axis
let t_0: unitInterval := sSup {t | (p t).1 = 0}
let xcoord_path := fun t => (p t).1
-- The x-coordinate of the path at t_0 is 0
have hpt_0_x : (p t_0).1 = 0 :=
  (isClosed_singleton.preimage xcoord_pathcont).sSup_mem \langle \tt 0, \ by \ aesop \rangle
-- By continuity of the path, we can find a \delta > 0 such that
-- for all t in [t_0 - \delta, t_0 + \delta], //p(t) - p(t_0)//<1/2
-- Hence the path stays in a ball of radius 1/2 around (0, 0)
obtain \langle \delta, h\delta, ht \rangle : \exists \delta > 0, \forall t, dist t t_0 < \delta \rightarrow
  dist (p t) (p t_0) < 1/2 :=
  Metric.eventually_nhds_iff.mp 
Metric.tendsto_nhds.mp
     (p.continuousAt t<sub>0</sub>) _ one_half_pos
```

The final statement uses the obtain tactic to extract witnesses from an existential statement. This tactic destructures the existential quantifier  $\exists \delta > 0, \ldots$  into concrete values:  $\delta$  (the distance),  $h\delta$  (the proof that  $\delta > 0$ ), and ht (the proof that the distance condition holds). Since  $\mathbb{R}^2$  is a metric space, we can work with the distance function dist:  $\mathbb{R} \times \mathbb{R} \to \mathbb{R} \times \mathbb{R} \to \mathbb{R}$ , which computes the Euclidean distance between two points. The statement dist  $t t_0 < \delta$  expresses  $|t - t_0| < \delta$  in the unit interval, while dist (p t) (p t<sub>0</sub>) < 1/2 expresses  $|p(t) - p(t_0)| < 1/2$  in  $\mathbb{R}^2$ . The proof itself leverages Metric module.

- p.continuousAt  $t_0$  asserts that the path p is continuous at  $t_0$
- Metric.tendsto\_nhds.mp converts this to the metric space characterization: for any  $\varepsilon > 0$ , there exists  $\delta > 0$  such that points within  $\delta$  of  $t_0$  map to points within  $\varepsilon$  of  $p(t_0)$
- Metric.eventually\_nhds\_iff.mp further unpacks this into the  $\forall t, dist \ t \ t_0 < \delta \rightarrow dist \ (p \ t) \ (p \ t_0) < \varepsilon$  form
- We instantiate with  $\varepsilon = 1/2$  using one\_half\_pos

We can find a time  $t_1$  greater than  $t_0$  that remains in the neighborhood of  $t_0$ , and obtain a point  $a = x(p(t_1)) > 0$  which is positive.

```
-- Let t_1 be a time when the path is not on the y-axis
-- t_1 is in (t_0, t_0 + \delta], hence t_1 > t_0
obtain \langle \mathsf{t}_1, \; \mathsf{h} \mathsf{t}_1 \rangle : \exists \; \mathsf{t}_1, \; \mathsf{t}_1 > \mathsf{t}_0 \; \wedge \; \mathsf{dist} \; \mathsf{t}_0 \; \mathsf{t}_1 < \delta := \mathsf{by}
  let \mathbf{s}_0 := (\mathbf{t}_0 : \mathbb{R}) -- cast t_0 from unitInterval to \mathbb{R} for manipulation
  let s_1 := \min (s_0 + \delta/2) 1
   have hs_0_delta_pos : 0 \le s_0 + \delta/2 := add_nonneg t_0.2.1 (by positivity)
   have hs_1 : 0 \le s_1 := le_min hs_0_delta_pos zero_le_one
   have hs_1': s_1 \leq 1 := min_le_right ...
   sorry
-- Let a = xcoord_path t_1 > 0
-- This follows from the definition of t_0 and t_0 < t_1
-- so t_1 must be in S, which has positive x-coordinate
let a := (p t_1).1
have ha: a > 0 := by
   \verb"obtain" \langle \texttt{x}\texttt{,} \; \texttt{hxI}\texttt{,} \; \texttt{hx\_eq} \rangle \; : \; \texttt{p} \; \texttt{t}_1 \; \in \; \texttt{S} \; := \; \texttt{by}
      cases p_conn.somePath_mem t1 with
      | inl hS => exact hS
      | inr hZ =>
        -- If p t_1 \in Z, then (p t_1).1 = 0
        have : (p t_1).1 = 0 := by rw [hZ]
        -- So t_1 \leq t_0, contradicting t_1 > t_0
        have hle : t_1 \le t_0 := le_sSup this
        have hle_real : (t_1 : \mathbb{R}) \leq (t_0 : \mathbb{R}) := Subtype.coe_le_coe.mpr hle
        have hgt_real : (t_1 : \mathbb{R}) > (t_0 : \mathbb{R}) := Subtype.coe_lt_coe.mpr
      ht_1.1
        linarith
   simpa only [a, \leftarrow hx_eq] using hxI
```

The code is quite convoluted in Lean, and i will omit a detailed explanation as well as some part of it. However, it's worth mentioning a few key technical points. The type unit Interval is a subtype of  $\mathbb{R}$ , defined as  $\{x : \mathbb{R} \mid 0 \le x \le 1\}$ . In Lean, a subtype  $\{x : \alpha // P x\}$  bundles a value x of type  $\alpha$  together with a proof that x satisfies the predicate P. Manipulating terms of unitInterval directly is challenging because this type lacks many algebraic operations such as addition, minimum, etc. Therefore, we cast to  $\mathbb{R}$  (with let  $s_0 := (t_0 : \mathbb{R})$ ) to perform arithmetic operations, then cast back to unitInterval by providing proofs that the bounds [0,1] are satisfied  $(hs_1, hs_1)$ . In the second case of the inner statuent of have ha : a > 0, if  $p(t_1) \in Z$ , then  $(p \ t_1).1 = 0$  by definition of  $Z = \{(0,0)\}$ . This implies  $t_1 \leq t_0$  by the definition of  $t_0$  as the supremum. However, we also have  $t_1 > t_0$  from our construction of  $t_1$  (ht<sub>1</sub>.1). The tactic linarith, an automated solver for linear arithmetic, recognizes this contradiction by observing both hle\_real : (t\_1 :  $\mathbb{R}$ )  $\leq$  (t\_0 :  $\mathbb{R}$ ) and hgt\_real : (t\_1  $\mathbb{R}$  > (t<sub>0</sub> :  $\mathbb{R}$ ). Since these statements are contradictory, linarith proves False. Lemmas like Subtype.coe\_lt\_coe allow us to transfer inequalities between the subtype and its underlying type, needed for linarith.

Finally, simpa only [a,  $\leftarrow$  hx\_eq] using hxI completes the proof. The tactic simpa combines simplification (simp) with assumption matching. The directive only [a,  $\leftarrow$  hx\_eq] unfolds the definition of  $a=(p\ t_1).1$  and rewrites

using  $hx_eq$  in the reverse direction, transforming the goal from  $(p t_1).1 > 0$  to  $(sine_curve x).1 > 0$ . Since  $sine_curve x = (x, sin(1/x))$ , this simplifies to x > 0, which is exactly the hypothesis hxI. The using hxI clause applies this hypothesis to close the goal.

Next, the image  $x(p([t_0, t_1]))$  is connected (as the continuous image of a connected set), and it contains  $0 = x(p(t_0))$  and  $a = x(p(t_1))$ . Since every connected subset of  $\mathbb{R}$  is an interval, we have

$$[0, a] \subseteq x(p([t_0, t_1]))$$

This will be crucial for the next step, where we show that the path must oscillate.

```
-- The image x(p([t_0, t_1])) is connected and contains 0 and a -- Therefore [0, a] \subseteq x(p([t_0, t_1])) have Icc\_of\_a\_b\_sub\_Icc\_t_0\_t_1: Set.Icc 0 a \subseteq xcoord\_path '' Set.Icc t_0 t_1 := IsConnected.Icc\_subset ((isConnected_Icc (le_of_lt ht_1.1)).image \_ xcoord\_pathcont.continuousOn) (\langle t_0, left_mem_Icc.mpr (le_of_lt ht_1.1), hpt_0_x\rangle) (\langle t_1, right_mem_Icc.mpr (le_of_lt ht_1.1), rf1\rangle)
```

Now we construct a sequence that demonstrates the contradiction. Recall that  $\sin(\theta) = 1$  if and only if  $\theta = \frac{(4k+1)\pi}{2}$  for some  $k \in \mathbb{Z}$ . Therefore,  $(x, \sin(1/x)) = (x, 1)$  when

$$x = \frac{2}{(4k+1)\pi}$$

for  $k \in \mathbb{N}$ . As  $k \to \infty$ , these x-values approach 0, so infinitely many of them lie in any interval [0, a]. We define this sequence and establish its key properties:

```
noncomputable def xs_pos_peak := fun (k : \mathbb{N}) => 2/((4 * k + 1) * Real.pi)

lemma xs_pos_peak_tendsto_zero : Tendsto xs_pos_peak atTop (\mathcal{N} 0) := sorry

lemma xs_pos_peak_nonneg : \forall k : \mathbb{N}, 0 \leq xs_pos_peak k := sorry

lemma sin_xs_pos_peak_eq_one (k : \mathbb{N}) : Real.sin ((xs_pos_peak k)^{-1}) = 1 := sorry
```

The crucial property is that this sequence eventually enters [0, a]:

```
-- For any k \in \mathbb{N}, sin(1/xs\_pos\_peak(k)) = 1
-- Since xs\_pos\_peak converges to 0 as k \to \infty,
-- there exist indices i \ge 1 for which xs\_pos\_peak i \in [0, a]
have xpos\_has\_terms\_in\_Icc\_of\_a\_b: \exists i : \mathbb{N}, i \ge 1 \land xs\_pos\_peak i \in Set.Icc 0 a := sorry
```

This gives us points on the topologist's sine curve with y-coordinate equal to 1, lying arbitrarily close to the y-axis.

Now we can establish the final contradiction. Since  $[0, a] \subseteq x(p([t_0, t_1]))$  by the previous argument, and  $xs_pos_peak(i) \in [0, a]$  for some i, there must exist some  $t' \in [t_0, t_1]$  such that  $x(p(t')) = xs_pos_peak(i)$ . This means

 $p(t') = (xs_pos_peak(i), \sin(1/xs_pos_peak(i))) = (xs_pos_peak(i), 1)$ , so the y-coordinate of p(t') equals 1. However, since  $t' \in [t_0, t_1] \subseteq [t_0, t_0 + \delta)$ , we have  $dist(t', t_0) < \delta$ , which by our earlier continuity argument implies  $||p(t') - p(t_0)|| < 1/2$ . But  $||p(t') - (0, 0)|| \ge |(p(t')).2| = |1| = 1 > 1/2$ , yielding a contradiction.

```
-- Show there exists time t' in [t_0, t_1] \subseteq [t_0, t_0 + \delta) such that p(t') =
      (*, 1)
obtain \langle t', ht', hpath_t' \rangle : \exists t' \in Set.Icc t_0 t_1, (p t').2 = 1 := sorry
-- Derive the final contradiction using t', ht', hpath_t'
-- First show that p t_0 = (0, 0)
have hpt_0 : p t_0 = (0, 0) := sorry
-- t ' is within \delta of t_0 (since t ' \in [t_0, t_1] and dist t_0 t_1 < \delta)
have t'_close : dist t' t_0 < \delta := by
  calc dist t, t_0

    dist t<sub>1</sub> t<sub>0</sub> := dist_right_le_of_mem_uIcc (Icc_subset_uIcc' ht')

    \_ = dist t<sub>0</sub> t<sub>1</sub> := dist_comm \_ \_
    _{-} < \delta := ht<sub>1</sub>.2
-- By continuity, p(t') should be close to p(t_0)
have close : dist (p t') (p t_0) < 1/2 := ht t' t'_close
-- But p(t') has y-coordinate 1, so it's actually far from p(t_0) = (0, 0)
have far : 1 \le dist (p t') (p t_0) := by
  calc 1 = |(p t').2 - (p t_0).2| := by simp [hpath_t', hpt_0]
       _ \leq \|p t' - p t_0\| := norm_ge_abs_snd
       \_ = dist (p t') (p t<sub>0</sub>) := by rw [dist_eq_norm]
-- This is a contradiction: 1 \leq dist (p t') (p t_0) < 1/2
linarith
```

## 4.3 T is connected not path-connected

Finally, we combine the two parts in the following concise and pleasant theorem:

And now, since this code compiles successfully, these two lines stand as verified witnesses to the correctness of our entire proof. This showcases the power of proof assistants and formal reasoning: mathematics becomes not only more rigorous but also automatically verifiable. Furthermore, the formalization becomes a learning tool in its own right. Future readers can inspect each part of the code. Here the full proof: [link to Lean live]

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[] Placeholder for web:1.

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