Lecture Notes on Propositional and Predicate Logic

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Based on lecture by Petr Gregor

NOVEMBER 15, 2017

Introduction

Generally, logic is a study of arguments and inferences. While it started as philosophical discipline in ancient times, it is now widely studied in mathematics and computer science. As such, logic provides the basic language and tools for most of mathematics. It studies different prove systems and discusses whether they are sound (everything they prove is valid) and complete (everything that is valid can be proven).

While logic provides rather low-level tools for mathematics and computer science, it still has rather wide applications. For example, in some areas of artificial intelligence, logic is used to represent the knowledge of the intelligent agents and reason about it. The agents than use logical reasoning (theorem proving) to decide what to do next, or prove that a certain action is safe in a given environment. Another important area is formal software verification, where logic (and, again, theorem proving) can be used to formally verify, that a program indeed does what it should according to a specification. This is essential while implementing e.g. cryptographic protocols. Formal verification is also used while designing digital circuits.

There are also attempts to formalize the whole mathematics in logic and use computers to check that all the proofs are correct. For example, Mizar[1] is a system that aims to re-create most of the mathematics with formal and verified proofs. The verification starts from the basic mathematical axioms – in the case of the Mizar system, authors of so called Mizar articles can use only axioms of set-theory and theorems from previously verified articles. Therefore, everything published in the Mizar mathematical library is verified to be a correct consequence of the base axioms.

[1] http://www.mizar.org/library/

Logic serves as the formal language of mathematics, and therefore logic also needs to formally specify the syntax of the language. The syntax defines, what is a valid logical formula and what is not, however the meaning and validity of a formula is given by the semantics of the language. Logic itself prescribes the meaning of only a handful of symbols – namely the logical connectives ($\wedge, \vee, \rightarrow, \leftrightarrow, \neg$) and the quantifiers ($\forall, \exists$). Additionally, in languages with equality, the meaning of "$=$" is also given. All other symbols used in logical formulas can have arbitrary meaning, which is given by the semantics. So, for example, if we write a formula $(\forall x)(\forall y)(x + y = y + x)$, we cannot discuss its validity before defining the meaning of "$+$". The formula is valid if we are talking about the real numbers and "$+$" denotes their addition, however, the symbol "$+$" can also represent (quite un-

usually) the multiplication of square matrices, and in such a case, the formula is not valid.

There are different levels of the language of logic. In propositional logic, only propositional variables (those that are either true or false) and the logical connectives can be used. In first order logic, we can additionally use functions, relations and quantifiers for variables that range objects from some universe. In second order logic, there are additionally quantifiers for sets of objects in the universe (and, more specifically, for functions and relations). In higher order logic, we also have variables for sets of sets of objects. For example, a formula in propositional logic

$$(d \wedge c) \rightarrow s$$

can express that if it is dark and clear outside, the stars are visible. In a first order language, we can have a more complex formula

$$(\forall x)(\forall y)(S(x) \wedge E(y) \rightarrow (L(x,y) \rightarrow P(x,y)))$$

that expresses that if x is a student ($S(x)$) and y is an exam ($E(y)$), if x learns for y ($L(x,y)$) then x passes y ($P(x,y)$). As an example of second order language, we can write the axiom of induction:

$$(\forall P)(P(0) \wedge (\forall n)(P(n) \rightarrow P(n+1)) \rightarrow (\forall n)P(n))\,.$$

In the lecture, we will deal mostly with propositional and first-order logic, however there are also other extension of logic. For example, in multi-agent systems, so called modal logic is often used to represent the knowledge. In modal logic, there are special modalities, that can further qualify a statement. For example, there is a modality that says that a statement may be true, or must be true. Other types of modal logic contain modalities that express knowledge of other agents (e.g. "agent A knows that statement S is true" can be written as $\mathrm{K}_A.S$, or even "agent A knows that agent B knows that statement S is true" ($\mathrm{K}_A.\mathrm{K}_B.S$)). Another interesting type of modal logic is temporal logic, which contain modalities about time and can express e.g. "statement S will be true sometimes in the future".

About these lecture notes

In these lecture notes, logic is presented for students of computer science. Therefore, focus is given to areas most needed for computer scientists. For example, we use the more intuitive tableau method instead of the Hilbert-style prove systems. We also explain the resolution method in logic as a background to Prolog and logical programming. The more advanced topics on decidability and incompleteness are explained in a more informal way.

You are currently reading the first version of the lecture notes, which can, and most probably will, contain some errors. If you find an error, or if something is not clear, do not hesitate to contact the author by e-mail[2], or, alternatively, create an issue in the GitHub repository of the book[3].

[2] Martin.Pilat@mff.cuni.cz

[3] https://github.com/martinpilat/logic-book

There are also other resources you may want to check. One of them are the presentations created by Petr Gregor for his version of the lecture[4], that serve as a base for these lecture notes.

[4] http://ktiml.mff.cuni.cz/~gregor/logics/index.html

Informal! The author of these notes sometimes likes to explain things in a more intuitive way with some not-so-formal examples and metaphors. While he believes these can help to get better understanding of the given concept, they sometimes (read "often") are rather informal and have some limitations. Therefore, in these notes, they will be set in boxes like the one you are reading just now with a bold "**Informal!**" warning. The information contained in these boxes is always non-essential to the rest of the text and can be (some would even argue that should be) skipped.

Preliminaries

Like many mathematical texts, these lecture notes also assume that the reader has some basic knowledge. The most important concepts (many of which should sound familiar) are briefly introduced in this short section, both to provide a single place where these can be found and to introduce the notation used in these lecture notes.

We will start with the basic set-theoretic notions. The most basic of these is the *class*. Each property of sets $\varphi(x)$ defines a class $\{x|\varphi(x)\}$. Some classes are also sets, those that are not are called *proper classes*. The distinction between sets and classes is probably new for most of the readers. Why would we need any other collections of objects than sets? How is it possible, that there is a collection of objects, which is not a set? The reason to distinguish between these to is that if everything was considered a set, we could find paradoxes in the set theory. For example, if we had a set of all sets that do not contain themselves, does this set contain itself, or not? Let us assume, it does, but then, by definition, it does not. If we instead assume it does not contain itself, then, again, by definition, it does. This so called Russel's paradox can be avoided by using a notion of classes, that cannot contain other classes.

Informal! A class can be understood as any collection of sets that can be described by the language of set theory. However, some of these collections do not make much sense and can lead to paradoxes. Therefore, any collection, that would lead to some paradox is denoted as a proper class instead of a set and the paradoxes can thus be avoided.

The other set-theoretic notions should be much more familiar. We use $x \in y$ to denote that x is a member of set y, $x \notin y$ and $x \neq y$ are shortcuts for $\neg(x \in y)$ and $\neg(x = y)$. A set containing exactly elements $x_0, x_1, \ldots, x_n$ is denoted as $\{x_0, x_1, \ldots, x_n\}$. A set with only one element $\{x\}$ is called a *singleton* and a set with two elements $\{x_0, x_1\}$ is called an *unordered pair*. We will also use the common notation for set operations: $\emptyset$ denotes an *empty set*, $\cup$ and $\cap$ denote *union* and *intersection* of sets. The $\setminus$ is the *set difference* operator and $\triangle$

is the *symetric set difference* operator

$$x \triangle y = (x \setminus y) \cup (y \setminus x).$$

Two sets are *disjoint*, if their intersection is an empty set, and $x \subseteq y$ denotes that x is a subset of y (all elements of x are also elements of y). The set of all subsets of a set x – the *power set of* x – is denoted as $\mathcal{P}(x)$. The *union of set* x, $\bigcup x$, is the union of all sets contained in x. A *cover of a set* x is a set $y \subseteq \mathcal{P}(x) \setminus \emptyset$, such that $\bigcup y = x$, if all the sets in the cover y are mutually disjoint, than y is a *partition of* x.

The definition of an unordered pair can be used to define the *ordered pair* $(a,b) = \{a,\{a,b\}\}$ and an *ordered n-tuple* $(x_0,\ldots,x_{n-1}) = ((x_0,\ldots,x_{n-2}),x_{n-1})$ for $n > 2$. A Cartesian product of two sets a and b is $a \times b = \{(x,y)|x \in a, y \in b\}$ and the Cartesian power of a set x is $x^0 = \{\emptyset\}$, $x^n = x^{n-1} \times x$. A *binary relation* R is a set of ordered pairs. The *domain of* R is defined as $\mathrm{dom}(R) = \{x|(\exists y)(x,y) \in R\}$, the *range of* R is similarly $\mathrm{rng}(R) = \{y|(\exists x)(x,y) \in R\}$. The *extension of* x *in* R is the set $R[x] = \{y|(x,y) \in R\}$. The symbol R^{-1} denotes the *inverse relation* $R^{-1} = \{(y,x)|(x,y) \in R\}$. The *restriction of* R *to a set* z is defined as $R \restriction z = \{(x,y) \in R|x \in z\}$. Two relations can also be *composed* into one, $R \circ S = \{(x,z)|(\exists y)((x,y) \in R \wedge (y,z) \in S\}$. The *identity relation* on set z, $\mathrm{Id}_z = \{(x,x)|x \in z\}$.

A binary function f is a special type of binary relation where for every $x \in dom(f)$ there is exactly one y such that $(x,y) \in f$, then, y is the value of f in x denoted as $f(x)$. $f : X \to Y$ denotes a function f with $\mathrm{dom}(f) = X$ and $\mathrm{rng}(f) \subseteq Y$. The set of all such functions is ${}^{Y}X$. A function $f : X \to Y$ is a *surjection* (onto) if $\mathrm{rng}(f) = Y$, and it is an *injection* (one-to-one) if for any $x,y \in dom(f)$, $x \neq y \to f(x) \neq f(y)$. A function that is both a surjection and injection is called a *bijection*. Similarly to relation, we can define the inverse function f^{-1}, and the composition of functions $f : X \to Y$ and $g : Y \to Z$ as a function $f \circ g$ with $(f \circ g)(x) = g(f(x))$. The image of a set A, denoted as $f[A]$ is the set of function values for all elements of A, $f[A] = \{y|(x,y) \in f, x \in A\}$.

There are also two special types of relations which will be important later: equivalences and orders. An equivalence on a set X is relation that is reflexive ($R(x,x)$ for all $x \in X$), symmetric ($R(x,y) \to R(y,x)$ for $x,y \in X$) and transitive ($(R(x,y) \wedge R(y,z)) \to R(x,z)$) for all $x,y,z \in X$). The extension of x in R is called the equivalence class of x and is also denoted as $[x]_R$. $X/R = \{R[x]|x \in X\}$ is the *quotient set of* X *by* R. The quotient set is always a partition of X and every partition of X also defines an equivalence on X (two elements are equivalent if they are in the same set in the partition).

The other important types of the relations are the orders, usually an order is denoted as $\leq$. Such a relation is a *partial order of a set* X, if it is reflexive ($x \leq x$ for $x \in X$), antisymmetric ($x \leq y \wedge y \leq x \to x = y$ for $x,y \in X$) and transitive ($x \leq y \wedge y \leq z \to x \leq z$ for $x,y,z \in X$). If, additionally, for every two elements $x,y \in X$ it holds that $x \leq y$ or $y \leq x$ (dichotomy) than $\leq$ is a total (linear) order. It is a well-order if additionally every non-empty subset of Xhas a least element. Finally,

and order of X is dense, if X is not a singleton and for every two elements $x, y \in X$, there is another element $z \in X$ between these two $(x < y \rightarrow (\exists z)(x < z \wedge z < y))$, where $a < b$ means that $a \leq b \wedge a \neq b$.

For example, the common ordering of natural numbers ($\leq$ on $\mathbb{N}$) is a linear well-order (as every two natural numbers are comparable and every subset of natural numbers has a least element under this order), however, it is not a dense order, as for example there is no natural number between 0 and 1. On the other hand, the common ordering of rational numbers is a dense linear order (there is a rational number between any pair of distinct rational numbers), however it is not a well-order, as e.g. the set $\{x \in \mathbb{Q} \,|x \leq 0\}$ has no least element.

The natural numbers can be defined using the empty set in an inductive way – $0 = \emptyset, 1 = \{0\} = \{\emptyset\}, 2 = \{0, 1\} = \{\emptyset, \{\emptyset\}\}, \ldots, n = \{0, \ldots, n-1\}, \ldots$. The set of all natural numbers $\mathbb{N}$ is the smallest set containing $\emptyset$ and closed under the operation of successor $S(x) = x \cup \{x\}$. The other common sets of numbers are the integers, which can be defined as the $\mathbb{Z} = (\mathbb{N} \times \mathbb{N})/\sim$, with $(a, b) \sim (c, d)$ if and only if $a + d = b + c$. Similarly, the set of rational numbers $\mathbb{Q}$ can be defined as $\mathbb{Q} = (\mathbb{Z} \times (\mathbb{Z} \setminus \{0\}))/\sim$, with $(a, b) \sim (c, d)$ if and only if $ad = bc$. The definition of real numbers $\mathbb{R}$ is more complex. These are usually defined as cuts of the rational numbers $\mathbb{Q}$, where a cut is a partition of $\mathbb{Q}$ into two sets A and B, where all numbers in B are greater than all numbers of A, and A has no greatest element. For example, the cut corresponding to the irrational number $\sqrt{2}$ is $A = \{a \in \mathbb{Q} | a^2 < 2 \vee a < 0\}, B = \{b \in \mathbb{Q} | b^2 > 2 \wedge b > 0\}$.

Another important notion for the rest of the lecture deals with the cardinality ("size") of sets. A set X has a cardinality smaller or equal to set Y ($X \preceq Y$) if there is an injective function $f : X \rightarrow Y$. If there is a bijection $f : X \rightarrow Y$ than we say that X and Y have the same cardinality ($X \approx Y$), finally X has strictly smaller cardinality than Y ($X \prec Y$) if ($X \preceq Y \wedge \neg(X \approx Y)$). For each set x, there is a cardinal number $\kappa \approx x$, denoted as $|x| = \kappa$. A set X is *finite* if $|X| = n$ for some $n \in \mathbb{N}$. It is *countable*, if its finite or if $|x| = |\mathbb{N}| = \omega$. Otherwise, it is *uncountable*. The cardinality of $\mathcal{P}(\mathbb{N})$ is called the continuum.

It is interesting to know the cardinality of the common sets of numbers. Obviously, the set of natural numbers $\mathbb{N}$ is countable. A less obvious fact is that the sets of integers and rational numbers also have the same cardinality and are therefore also countable. For the integers, we can create an infinite sequence of integers $s = \langle 0, 1, -1, 2, -2, 3, -3, \ldots \rangle$, then a function $f(i) = s_i$ is an injective function $\mathbb{N} \rightarrow \mathbb{Z}$, therefore $\mathbb{Z} \preceq \mathbb{N}$. The other inequality ($\mathbb{N} \preceq \mathbb{Z}$) is obvious (use identity as the injective function). In order to show that the set of rational numbers $\mathbb{Q}$ is also countable, we can create a function $f(\frac{p}{q}) = 2^{|p|} 3^q 5^{\text{sign}(p)}$ (we consider only cases where $p \in \mathbb{Z}, q \in \mathbb{N} \setminus \{0\}$, which clearly covers all the rationals), this is again an injective mapping $\mathbb{Q} \rightarrow \mathbb{N}$ and therefore $\mathbb{Q} \preceq \mathbb{N}$. As before, the other inequality is trivial. Finally, we can show that the set of real numbers $\mathbb{R}$ has bigger cardinality than the set of natural numbers $\mathbb{N}$. Obviously $\mathbb{N} \preceq \mathbb{R}$ as $\mathbb{N} \subseteq \mathbb{R}$. Let us assume both the sets have the

same cardinality, in such a case there is bijection $f : \mathbb{N} \to \mathbb{R}$. We will now define a new real number r in the following way. The integer part of the number is 0, the first digit after the decimal point is different from the first digit after decimal point in $f(0)$, the second digit is different from the second digit in $f(1)$, and so on[5]. This real number is different from all the numbers in $\{f(0), f(1), \dots\}$ as it differs from the number $f(i)$ in the i-th digit after the decimal point. This is a contradiction with the assumption that f is a bijection between $\mathbb{N}$ and $\mathbb{R}$ and therefore $\mathbb{N} \prec \mathbb{R}$.

[5] If we write the decimal value of the number r as $r = 0.r_0r_1r_2\dots$, where r_i is the i-th decimal digit, we can define $r_i = (f(i)_i + 1) \bmod 10$, where $f(i)_i$ is the i-th decimal digit of $f(i)$.

We will conclude the discussion of cardinalities by showing the Cantor's theorem.

Theorem 1 (Cantor)**.** *For every set x, $x \prec \mathcal{P}(x)$.*

Proof. First, $f(x) = \{x\}$ is an injection $X \to \mathcal{P}(x)$ and therefore $x \preceq \mathcal{P}(x)$. Suppose there is also an injective $g : \mathcal{P}(x) \to x$. We can define a set $y = \{g(z) | z \subseteq x \wedge g(z) \notin z\}$. Now, similarly to the Russel's paradox, $g(y) \in y$ if and only if $g(y) \notin y$, which is a contradiction, and therefore there cannot be any such injective g and so $x \prec \mathcal{P}(x)$. □

As the tableau method used in this lecture relies on trees, we will conclude this preliminary section by a brief discussion on trees. Most of the readers are probably familiar with finite trees, however, we will sometimes need to work with infinite trees and therefore we define a *tree* as a set T with a partial order $<_T$ (called the tree order) with a unique least element (*the root*) and in which the set of predecessors of any element is well-ordered by $<_T$. In this definition a branch is a maximal linearly ordered subset of T. Apart from this difference in definition, we will use the common terminology on trees from the graph theory. For simplicity, we will only consider finitely branching trees, where each node except the root has an immediate predecessor[6]. In such trees we can define the *levels of the tree*. The root is on the level 0, the sons of the nodes on the $(n-1)$-th level are on level n. The depth of tree is maximal $n \in \mathbb{N}$ of a non-empty level. In case the tree has an infinite branch it has an infinite depth ω. In an n-ary tree, each node has at most n sons and a tree is finitely branching if each node has a finite number of sons.

[6] This means, we will not deal, for example, with trees where the nodes would be set of rational numbers $\mathbb{Q}$ and the tree order $<_T$ would be the common order on $\mathbb{Q}$.

Lemma 1 (König)**.** *Every infinite, finitely branching tree contains an infinite branch.*

Proof. The root of the tree has only finitely many sons, therefore there is a son of the root that is infinite. We choose this son and continue in the same way with his sons, thus constructing an infinite branch. □

Apart from the tree order $<_T$ we sometimes need to work with *ordered trees* where the sons of each node are additionally ordered from left to right with a *left-to-right order* $<_L$. In a *labeled tree* each node also contains an additional information. For example, the formula

$$(p \wedge q) \to q$$

can be represented as the labeled ordered tree on the left.

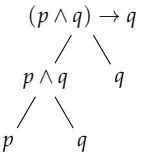

Figure 1: The labeled ordered tree representing the formula $(p \wedge q) \to q$.

Part I

Propositional Logic

Propositional Formulas and Models

In this chapter, we start the discussion of propositional logic. We will define, how propositional formulas look, what is a model in propositional logic and we will also discuss some special forms of formulas.

Propositional logic is the more basic type of logic (and predicate logic is an extension of propositional logic in a sense). Propositional formulas (propositions) are created from so called *propositional variables* that represent an atomic fact which can either be true or false. These propositional variables can only be connected by common logic connectives ($\rightarrow$, $\leftrightarrow$, $\wedge$, $\vee$, $\neg$). Logical formulas can additionally use parentheses to indicate the order of application of connectives. While the propositional formulas are simple compared to formulas in other types of logic, they are still useful. One of the most important problems in propositional logic and in computer science in general is the satisfiability of propositional formulas (SAT). Many other NP-complete problems are often solved by transformation to the SAT problem and using one of the existing SAT solvers.

Syntax of Propositional Logic

The set of propositional variables is often called $\mathbb{P}$ and the variables themselves are usually named p, q, r, s or $p_0, p_1, \ldots, q_0, q_1$, or similarly. Now, we can formally define the propositional formula (over $\mathbb{P}$).

Definition 1. Let $\mathbb{P}$ is the set of propositional variables, than

1. Every propositional variable from $\mathbb{P}$ is a propositional formula.
2. If φ and ψ are propositional formulas, than $(\varphi \rightarrow \psi)$, $(\varphi \wedge \psi)$, $(\varphi \vee \psi)$, $(\varphi \leftrightarrow \psi)$, and $(\neg\varphi)$ are propositional formulas.
3. Every propositional formula is created by finite application of the two rules above.

The last part of the definition ensures that every formula is finite, this also means that each formula can contain only a finite number of distinct variables. The set of propositional variables used in a formula φ will be denoted as $\mathrm{var}(\varphi)$. On the other hand, the set of all propositional formulas using only variables from a set $\mathbb{P}$ will be denoted as $\mathrm{VF}_{\mathbb{P}}$.

Formulas are thus strings created from propositional variables, logical connectives, and parentheses, that fulfill the conditions in the

definition above. A substring of such a string that also fulfills the conditions is called a *sub-formula*.

The formal definition of formula dictates the use of parentheses around every sub-formula, which can be rather cumbersome. Therefore, we define priorities of the logical connectives and can thus omit some of the parentheses. The standard priorities are such, that the negation ($\neg$) has the highest priority (therefore parentheses around $(\neg\varphi)$ can always be omitted), conjunction and disjunction ($\vee, \wedge$) have "middle" priority, and implication and equivalence ($\rightarrow, \leftrightarrow$) have the lowest priority. Therefore, we can write $\varphi \wedge \psi \rightarrow \neg\varphi \vee \xi$ instead of $((\varphi \wedge \psi) \rightarrow ((\neg\varphi) \vee \xi))$.

Each formula can be also represented by a so called *formation tree*, which is a finite ordered tree, whose nodes are labeled with propositions – the leaves are labeled with propositional variables, if a node has label $(\neg\varphi)$, it has a single son labeled with φ, and if a node has label $(\varphi \rightarrow \psi), (\varphi \wedge \psi), (\varphi \vee \psi)$, or $(\varphi \leftrightarrow \psi)$, it has two sons, the left one has label φ, and the right one has label ψ. For example, a formula $p \wedge q \rightarrow \neg(p \vee s)$ is represented by the formation tree on the left.

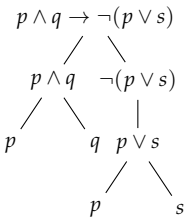

Figure 2: The formation tree representing the formula $p \wedge q \rightarrow \neg(p \vee s)$.

It is simple to show (by the induction on the number of nested parentheses) that each formula is associated with a unique formation tree.

Semantics of Propositional Logic

Once we have the formal definition of the formula (the syntax of propositional logic), we can define its semantics (what the formula means). The propositional variables represent atomic statements, that can have one of two truth values – either 0 (false) or 1 (true). The truth value of the whole proposition in then given by the truth values of the variables and by the semantics of the logical connectives, which is given in Table 1 bellow.

p	q	$\neg p$	$p \vee q$	$p \wedge q$	$p \rightarrow q$	$p \leftrightarrow q$
0	0	1	0	0	1	1
0	1	1	1	0	1	0
1	0	0	1	0	0	0
1	1	0	1	1	1	1

Table 1: The semantics of logical connectives

We can also consider the table above a definition of Boolean functions $\vee_1, \wedge_1, \rightarrow_1, \leftrightarrow_1$, and $-_1$, that implement the logical connectives. We will use these functions in cases where it is needed (e.g. while talking about truth values of propositions). More generally, any propositional formula with n variables defines a Boolean function $f : \{0,1\}^n \rightarrow \{0,1\}$ (later, we will also see that any Boolean function can be expressed using a propositional formula).

We also define two special logical formulas. The formula $\top \equiv p \vee \neg p$, which is always true, and the formula $\bot \equiv p \wedge \neg p$ which is always false.

We can now define the truth assignment and the truth value of

formula more formally.

Definition 2. A *truth assignment* is a function $v : \mathbb{P} \to \{0, 1\}$, that is $v \in {}^{\mathbb{P}}2$.

A *truth value* $\bar{v}(\varphi)$ of a propositional formula φ for a truth assignment v is defined inductively as:

- $\bar{v}(p) = v(p)$ if $p \in \mathbb{P}$
- $\bar{v}(\neg\varphi) = -_1(\bar{v}(\varphi))$
- $\bar{v}(\varphi \vee \psi) = \vee_1(\bar{v}(\varphi), \bar{v}(\psi))$
- $\bar{v}(\varphi \wedge \psi) = \wedge_1(\bar{v}(\varphi), \bar{v}(\psi))$
- $\bar{v}(\varphi \to \psi) = \to_1 (\bar{v}(\varphi), \bar{v}(\psi))$
- $\bar{v}(\varphi \leftrightarrow \psi) = \leftrightarrow_1 (\bar{v}(\varphi), \bar{v}(\psi))$

We can easily show (by the induction on the structure of the formula) that the truth value of a formula φ depends only on the truth assignment of variables from $\mathrm{var}(\varphi)$.

A proposition φ over $\mathbb{P}$ is *true in (satisfied by) an assignment* $v \in {}^{\mathbb{P}}2$, if $\bar{v}(\varphi) = 1$. In such a case, v is called a *satisfying assignment* for φ, we denote this fact $v \vDash \varphi$. If the formula is true for all assignments $v \in {}^{\mathbb{P}}2$, we say that it is *valid (a tautology)* and denote the fact as $\vDash \varphi$. On the other hand, if there is no assignment for which the formula is true, it is called *unsatisfiable (a contradiction)*. A formula φ is *independent (a contingency)* if it is neither a tautology nor a contradiction, i.e. there are two assignments $v_1, v_2 \in {}^{\mathbb{P}}2$, such that $\bar{v}_1(\varphi) = 1$ and $\bar{v}_2(\varphi) = 0$. Finally, a formula is *satisfiable* if there is a truth assignment in which it is true.

A truth assignment of $\mathbb{P}$ is also called a model of the language $\mathbb{P}$. The set of all models of $\mathbb{P}$ is denoted as $M(\mathbb{P})$, and, obviously $M(\mathbb{P}) = {}^{\mathbb{P}}2$. A proposition φ over $\mathbb{P}$ is valid in a model $v \in M(\mathbb{P})$, if $\bar{v}(\varphi) = 1$. Then we also say that v is a model of φ, denoted as $v \vDash \varphi$. $M^{\mathbb{P}}(\varphi) = \{v \in M(\mathbb{P}) | v \vDash \varphi\}$ is the *class of all models* of φ. A formula is valid, if it is true in every model of the language, it is unsatisfiable if it does not have a model, and satisfiable if it has a model. It is independent if it is true in a model of the language and false in another one. Formulas φ and ψ are logically equivalent ($\varphi \sim \psi$), if $M^{\mathbb{P}}(\varphi) = M^{\mathbb{P}}(\psi)$.

The last two paragraphs say basically the same, the difference is that in the latter one, we use the notion of model, which is central to logic. The notion of models, and sets of models will be important later, and "model" is one of the key terms in logic.

In the definition of propositions, we used 5 different logical connectives. However, if we take a look at the table with their semantics, we may notice, that, for example, $p \to q$ is equivalent $\neg p \vee q$. Therefore, even without using the implication ($\to$) we can still express everything we could with them. More formally, for every formula $\varphi \in \mathrm{VF}_{\mathbb{P}}$, there is an equivalent formula φ' that does not use the implication. Moreover, we can notice, that $p \leftrightarrow q$ is equivalent to $(p \to q) \wedge (q \to p)$, therefore we even do not need the equivalence, and every formula can be written using only negation, conjunction, and disjunction ($\neg, \wedge, \vee$). This feature of the set can be defined more formally.

Definition 3. A set of connectives is *adequate* if they can express any Boolean function by some proposition from them.

We have already discussed that the set $\{\neg, \wedge, \vee\}$ is adequate. We can also show, that the set $\{\rightarrow, \neg\}$ is adequate, the easiest way to do that is to realize, that $(p \wedge q) \sim \neg(p \rightarrow \neg q)$ and $(p \vee q) \sim (\neg p \rightarrow q)$.

Generally, we can also define custom connectives, for example, the so called Shaffer stroke (NAND) is defined as $p \uparrow q \sim \neg(p \wedge q)$, or the Pierce arrow (NOR) is defined as $p \downarrow q \sim \neg(p \vee q)$. Interestingly, both $\{\uparrow\}$ and $\{\downarrow\}$ are adequate sets. This is an important fact for the construction of logical circuits as we can use a logical gate of only one kind (either NAND or NOR) to express any Boolean function.

Normal Forms

There are also special forms of formulas, which are often used. Among the most common ones are so called conjunctive and disjunctive normal forms. In order to define these two forms, we first need to define a literal. A *literal* is a propositional variable or its negation, for example, if $\mathbb{P} = \{p, q\}$ then all the literals we can construct over $\mathbb{P}$ are $\{p, \neg p, q, \neg q\}$. A formula is in conjunctive normal form (CNF) if it is a conjunction of disjunctions of literals. Disjunctions of literals are also called *clauses*, therefore we can also say, that a CNF formula is a conjunction of clauses. On the other hand, a formula is in disjunctive normal form (DNF) if it is a disjunction of conjunctions of literals. So, for example, $(p \vee \neg q \vee r) \wedge (p \vee q) \wedge (\neg p \vee q \vee r)$ is a formula in CNF and$(\neg p \wedge q \wedge \neg r) \vee (\neg p \wedge \neg q) \vee (p \wedge \neg q \wedge \neg r)$ is a formula in DNF (and, moreover a negation of the previous one in CNF).

Now, we would like to show, that for every formula, there is an equivalent formula in CNF and another equivalent formula in DNF. To this end, we will need the following set of rules, which can be proven by checking the truth table of the propositional connectives:

1. $(\varphi \rightarrow \psi) \sim (\neg\varphi \vee \psi), (\varphi \leftrightarrow \psi) \sim ((\neg\varphi \vee \psi) \wedge (\neg\psi \vee \varphi))$
2. $\neg\neg\varphi \sim \varphi, \neg(\varphi \wedge \psi) \sim (\neg\varphi \vee \neg\psi), \neg(\varphi \vee \psi) \sim (\neg\varphi \wedge \neg\psi)$
3. $(\varphi \vee (\psi \wedge \chi)) \sim ((\psi \wedge \chi) \vee \varphi) \sim ((\varphi \vee \psi) \wedge (\varphi \vee \chi))$
4. $(\varphi \wedge (\psi \vee \chi)) \sim ((\psi \vee \chi) \wedge \varphi) \sim ((\varphi \wedge \psi) \vee (\varphi \wedge \chi))$

We can also easily show (again by induction on the structure of the formula) that if we have a formula φ' which is obtained from φ by replacing some occurrences of its sub-formula ψ with an equivalent sub-formula ψ', then $\varphi \sim \varphi'$.

And finally, we can show the following theorem.

Theorem 2. *For every formula φ over $\mathbb{P}$, there are formulas φ_C and φ_D, such that φ_C is in CNF, φ_D is in DNF and $\varphi \sim \varphi_C$ and $\varphi \sim \varphi_D$.*

Proof. The propositions φ_C and φ_D can be obtained from φ by applying the rules 1 to 4 mentioned above. □

The discussion above shows one of the ways to obtain equivalent formulas in CNF and DNF to a given formula. We can in fact apply

the rules in the order, in which they are presented. First, we remove all the implications and equivalences by using the rules no. 1. Then, we move all negations to the literals (i.e. there are no negations outside of parentheses), using the rule no. 2 and, finally, we repeatedly apply rules no. 3 and 4 to obtain the CNF and DNF.

This syntactic approach is not the only one to obtain CNF/DNF from a given formula. We can also construct the truth table of the formula and then read the CNF/DNF almost directly from the table. We will show a more general approach here, we will construct a CNF and DNF formulas φ_C and φ_D such that $M^{\mathbb{P}}(\varphi_C) = M^{\mathbb{P}}(\varphi_D) = K \subseteq M(\mathbb{P})$, for a given finite set of truth assignments K over a finite $\mathbb{P}$.

Before we show the construction, we will define the notion of p^t for a variable p and a truth value t as

$$p^t = \begin{cases} p & \text{if } t = 1 \\ \neg p & \text{if } t = 0 \end{cases} .$$

Now, we can easily see that for a single assignment $v \in K$, the set of models of the formula $\bigwedge_{p \in \mathbb{P}} p^{v(p)}$ contains only v. For a set of assignments K, we can just make a disjunction over all assignments in K (remember K is a finite set). Therefore,

$$M(\bigvee_{v \in K} \bigwedge_{p \in \mathbb{P}} p^{v(p)}) = K .$$

Thus we constructed a formula in DNF whose models are exactly the set K.

Constructing a formula φ in CNF such that $M(\varphi) = K$ for some given finite K is slightly more complex. However, we can use the fact that the negation of a formula in DNF is a formula in CNF. Negating a formula in CNF/DNF means changing all the conjunctions to disjunctions and vice versa and changing all literals to the complementary ones (i.e. changing p to $\neg p$ and vice versa). So, we start by creating a formula $\neg\varphi$ in DNF for the set ${}^{\mathbb{P}}2 \setminus K$ according to the approach above. Then, we negate the formula, thus obtaining φ in CNF such that $M(\varphi) = K$. Following these two steps we obtain the CNF formula

$$\varphi = \bigwedge_{v \in {}^{\mathbb{P}}2 \setminus K} \bigvee p^{-1 v(p)}$$

such that $M(\varphi) = K$.

If we want to use this approach to create a formula in CNF or DNF equivalent to a formula φ, we simply choose $K = M(\varphi)$. This description also shows that any Boolean function f (i.e. function $f : \{0,1\}^n \to \{0,1\}$) can be expressed as a proposition. We can choose $K = \{v | f(v) = 1\}$.

Both the techniques described above lead to an equivalent formula in CNF/DNF, the table-based method is typically used only for formulas with lower number of variables, as the size of the table for a formula with n variables is 2^n.

Logical theories

In mathematics, we often need to work in theories – we assume that some facts are true (the axioms of the theory) and are interested in which other facts are true. Therefore, in logic, we define a *propositional theory over the language* $\mathbb{P}$ as a set of propositions from $\mathrm{VF}_{\mathbb{P}}$. These propositions are called *axioms*. An assignments $v \in M(\mathbb{P})$ is a *model of theory* T ($v \vDash T$), if all axioms of T are true in v. Similarly to formulas, we define the *class of models of* T as $M(T) = \{v \in M(\mathbb{P}) | v \vDash \varphi \text{for all} \varphi \in T\}$. A finite theory is equivalent to a conjunction of its axioms. We will also write $M(T, \varphi)$ as a shortcut for $M(T \cup \{\varphi\})$.

We can now re-define the semantics concepts with respect to a theory. Let T be a theory over $\mathbb{P}$ and φ a proposition over $\mathbb{P}$. We say that φ *is true (valid) in* T ($T \vDash \varphi$) it if is true in every model of T. In such a case, we also say that φ is a (semantic) consequence of T. A formula φ is *unsatisfiable (contradictory) in* T *(inconsistent with* T*)*, if it is false in every model of T. It is *independent (a contingency) in* T, if it is true is some model of T and false in another model of T and satisfiable in T, if it true in some model of T. Two propositions φ and ψ are *equivalent in* T *(*T*-equivalent)* ($\varphi \sim_T \psi$), if for every model $v \in M(T)$, $v \vDash \varphi$ if and only if $v \vDash \psi$. For an empty theory ($T = \emptyset$), or for a theory where all axioms are tautologies, the re-definitions in this paragraph are equivalent to the definitions mentioned earlier.

The concepts defined above can also be expressed using the sets of models. For example $T \vDash \varphi$ is the same as $M(T) \subseteq M(\varphi)$, and $\varphi \sim_T \psi$ is equivalent to $M(T, \varphi) = M(T, \psi)$.

For each theory, we define its consequence as the set of all propositions that are true in the theory – $\theta^{\mathbb{P}}(T) = \{\varphi \,| \varphi \in \mathrm{VF}_{\mathbb{P}}, T \vDash \varphi\}$. Now, if we have two theories T and T', such that $T \subseteq T'$ over $\mathbb{P}$, we can prove that $T \subseteq \theta^{\mathbb{P}}(T) = \theta^{\mathbb{P}}(\theta^{\mathbb{P}}(T)) \subseteq \theta^{\mathbb{P}}(T')$. The first part says, that each axiom of T is always a consequence of T. This makes sense, as an axiom of T is by definition true on all models of T. The next part says, that the consequences of consequences of T are still the original consequences. However, this is also simple to show. Obviously, $M^{\mathbb{P}}(T) = M^{\mathbb{P}}(\theta(T))$ and therefore also $\theta^{\mathbb{P}}(T) = \theta^{\mathbb{P}}(\theta^{\mathbb{P}}(T))$ by definition of the consequence. Finally, if we have a formula φ which is valid in all models of T then φ is also valid in all models of T' ($T \subseteq T'$) as each model of T' also must be a model of T. Therefore $\theta^{\mathbb{P}}(T) \subseteq \theta^{\mathbb{P}}(T')$.

Similarly, if we have propositions $\varphi, \varphi_1, \varphi_2, \ldots \varphi_n$ over $\mathbb{P}$, we can show that $\varphi \in \theta^{\mathbb{P}}(\{\varphi_1, \ldots, \varphi_n\})$ if and only if $\vDash (\varphi_1 \wedge \cdots \wedge \varphi_n) \rightarrow \varphi$.

A theory T over $\mathbb{P}$ is *inconsistent (unsatisfiable)*, if $T \vDash \bot$, otherwise T is *consistent (satisfiable)*. A theory is consistent if and only if it has a model. A theory is complete, if it is consistent and $T \vDash \varphi$ or $T \vDash \neg\varphi$ for every $\varphi \in \mathrm{VF}_{\mathbb{P}}$, i.e. there are no independent propositions in T. This is also equivalent to the fact that T has exactly one model (if T had two models v_1 and v_2, they there would be a propositional variable p, such that $v_1(p) \neq v_2(p)$ and therefore the formula p is true in one of the models and false in the other one, thus p is independent). In mathematics, we very often create new theories by adding axioms to

other theories. Such new theories are called extensions of the original theories. More formally, a theory T over $\mathbb{P}$ is an *extension* of T' over $\mathbb{P}'$, if $\mathbb{P}' \subseteq \mathbb{P}$ and $\theta^{\mathbb{P}'}(T') \subseteq \theta^{\mathbb{P}}(T)$, an extension is *simple*, if $\mathbb{P} = \mathbb{P}'$, and it is *conservative* if $\theta^{\mathbb{P}'}(T') = \theta^{\mathbb{P}}(T) \cap \mathrm{VF}_{\mathbb{P}'}$. Two theories T and T' are equivalent, if T is an extension of T' and vice versa.

Although we motivated the notion of extension by adding new axioms, it is defined more generally using the sets of consequences of the theory. This abstracts from the particular axioms and considers all equivalent theories the same. The notion of extension can also be expressed with the sets of models, if both theories T and T' are over the same language $\mathbb{P}$. In such a case T is an extension of T', if and only if $M^{\mathbb{P}}(T) \subseteq M^{\mathbb{P}}(T')$ and the two theories are equivalent if $M^{\mathbb{P}}(T) = M^{\mathbb{P}}(T')$.

We will conclude this section by the discussion about the number of nonequivalent propositions and theories over a finite language $\mathbb{P}$. We defined two formulas or theories equivalent, if they have the same sets of models. Therefore, if we want to compute the number of non-equivalent theories/formulas, we can instead compute the number of sets of models. So, if $|\mathbb{P}| = n$, than there are 2^{2^n} non-equivalent formulas (theories) over $\mathbb{P}$, as there are 2^n different assignments, and every set of assignments represents a formula (remember, we know how to write that formula in CNF/DNF) or a theory.

Using a similar reasoning, we can show how many nonequivalent valid (or contradictory – the number is the same) propositions are there in a theory. A valid proposition is true in all models of T, therefore there are $2^n - |M(T)|$ assignments where a valid proposition can be (but does not have to be) true. This means there are $2^{2^n - |M(T)|}$ valid (and contradictory) propositions in T. Every proposition is either valid, contradictory or independent, therefore there are $2^{2^n} - 2 \times 2^{2^n - |M(T)|}$ nonequivalent independent propositions in T. A theory has $2^{|M(T)|}$ simple extensions, one of these is contradictory (the set of models of an extension are a subset of the models of the original theory), and the same theory has $|M(T)|$ simple complete extensions (those correspond to single-element subsets of $M(T)$).

Instead of talking about nonequivalent propositions, we can also discuss T-nonequivalent propositions. There are $2^{|M(T)|}$ T-nonequivalent propositions (we know consider only subsets of $M(T)$ as the possible sets of models for the proposition), one of them is valid and one is contradictory in T, thus the number of T-nonequivalent propositions in T is $2^{|M(T)|} - 2$.

The fact, that we can use the number of subsets of models while computing the number of nonequivalent theories or formulas is more formally explained by so called Lindenbaum-Tarski algebra. For a consistent theory T over $\mathbb{P}$, we can define operations $\neg, \wedge, \vee, \bot, \top$ on the quotient set $\mathrm{VF}_{\mathbb{P}} \,/\, \sim_T$ by use of representatives, e.g. $[\varphi]_{\sim_T} \wedge [\psi]_{\sim_T} = [\varphi \wedge \psi]_{\sim_T}$. Then $AV^{\mathbb{P}}(T) = \langle \mathrm{VF}_{\mathbb{P}} \,/\, \sim_T, \neg, \wedge, \vee, \bot, \top \rangle$ is *Lindenbaum-Tarski algebra for* T. Since $\varphi \sim_T \psi$ if and only if $M(T, \varphi) = M(T, \psi)$ then $h([h]_{\sim_T}) = M(T, \varphi)$ is an injective function $h : \mathrm{VF}_{\mathbb{P}} \to \mathcal{P}(M(T))$. If $M(T)$ is finite then, h is additionally surjective, and

therefore AV is isomorphic to the algebra of sets $\mathcal{P}(M(T))$.

Satisfiability of Propositional Formulas

The problem of satisfiability of logical formulas is one of the central problems in computer science. The general question posed by the problem is, whether a given formula in CNF is satisfiable. In general, this problem is NP-complete[7], which means that we do not know any polynomial-time algorithm to solve it. However, there are some special types of formulas, for which SAT can be solved in polynomial time. In this section, we will discuss such formulas and show the algorithms that solve SAT for these, we will also briefly discuss local search algorithms for SAT, and describe the complete (but generally exponential) DPLL procedure.

[7] A problem c is NP-complete, if it is NP and if any other NP problem is reducible to c in polynomial time. A problem is NP if given a candidate solution, we can check in polynomial time that it is indeed a solution. A problem p is reducible to c in polynomial time, if we can transform each instance of c into an instance of p in polynomial time.

The first class of formulas are so called 2-CNF. A formula is in *k-CNF* if it is a disjunction of clauses and each of the clauses contain at most k literals. The SAT problem for k-CNF formulas is called the k-SAT. The k-SAT problem is NP-complete for $k > 2$, however for $k = 2$ it can be solved in polynomial time using the so called implication graph of the formula.

Definition 4. Let φ is a formula over $\mathbb{P}$ in 2-CNF, $\varphi \equiv \bigwedge_{i=1}^{j}(l_{i1} \vee l_{i2}) \wedge \bigwedge_{i=j+1}^{k} l_i$. The *implication graph* of φ is a oriented graph $G_\varphi = (V, E)$, where the set of vertices is

$$V = \{p | p \in \mathbb{P}\} \cup \{\neg p | p \in \mathbb{P}\}$$

and the set of edges is

$$E = \{(\bar{l}_{i1}, l_{i2}) | 1 \leq i \leq j\} \cup \{(\bar{l}_{i2}, l_{i1}) | 1 \leq i \leq j\} \cup \{(\bar{l}_i, l_i) | j+1 \leq i \leq k\}\,.$$

In the implication graph, the set of vertices corresponds to all literals from variables in $\mathrm{var}(\varphi)$ and each clause in the formula is represented as one or two edges. For a clause $l_1 \vee l_2$, we include two edges (implications) $\bar{l}_1 \to l_2$ and $\bar{l}_2 \to l_1$. These two implications are logically equivalent to the clause. For a *unit clause* l (a clause with only a single literal), we include a single edge $\bar{l} \to l$, this is also equivalent to l. The implication graph thus contains the 2-CNF formula written as implications between its literals. The implication graph of a formula can be constructed in a time linear in the length of the formula.

Let us now assume that a truth assignment $v \in {}^{\mathbb{P}}2$ satisfies a formula φ. In such a case, in every strongly connected component[8] in G_φ, all the literals have the same truth value. Otherwise there would be an implication which is not true in the assignment which is a contradiction with the fact that the whole formula is true in the assignment[9]. This also means that if we have a satisfying assignment for φ, none of the strongly connected components contain both a literal and its negation.

[8] In a strongly connected component, there is an oriented path between every pair of vertices.

[9] Assume that literals l_1 and l_n are in the same strongly connected component and that $v(l_1) = 1$ and $v(l_n) = 0$. There is a chain of implications $l_1 \to l_2, l_2 \to l_3, \ldots, l_{n-1} \to l_n$, at least one of these must be $1 \to 0$ and therefore false.

Can we use the implication graph of a formula to obtain a satisfying truth assignment? We indeed can, but only if none of the strongly connected components contain a pair of complementary literals. In

such a case, we can contract each of the strongly connected components into a single vertex (thus obtaining a graph G^*_φ). Such a graph would be acyclic and therefore has a topological ordering $<$. We create an assignment v in a few steps: for every unassigned component in increasing order of $<$, we assign 0 to all its literals and 1 to all the complementary literals in the graph (they would in fact also form an strongly connected component). Such an assignment is satisfying for φ. If not, G^*_φ would contain edges $p \to q$ and $\bar{p} \to \bar{q}$ with $v(p) = 1$ and $v(q) = 0$, but that contradicts the order of assignments as $p < q$ and $\bar{q} < \bar{p}$.

The discussion above can be summarized in the theorem bellow.

Theorem 3. *Proposition φ in 2-CNF is satisfiable if and only if no strongly connected component of its implication graph G_φ contains a pair of complementary literals.*

As the implication graph can be constructed in linear time and the strongly connected components can also be found in linear time. This also shows that the 2-SAT problem can be solved in linear time.

Another class of formulas where SAT can be solved in polynomial time are conjunctions of clauses with at most one positive literal. Such clauses are called Horn clauses, and such formulas are called Horn formulas. The problem of satisfiability of Horn formulas is called Horn-SAT.

The Horn clauses can also be interpreted as implications. The Horn clause $(\neg p_1 \vee \neg p_2 \vee \cdots \vee \neg p_n \vee q)$ is equivalent to the implication $(p_1 \wedge p_2 \wedge \cdots \wedge p_n) \to q$.

Deciding whether a Horn formula φ is satisfiable or not is simple, and can be done using the following algorithm:

1. If φ contains a pair of unit clauses l and $\bar{l}$ (a pair of complementary literals) it is not satisfiable.

2. If φ contains a unit clause l, assign 1 to l, remove all clauses containing l, remove $\bar{l}$ from all clauses, and continue from the start.

3. If φ does not contain a unit clause, it is satisfied by assigning 0 to all remaining propositional variables.

The first step of the algorithm is obviously correct, as $p \wedge \neg p$ is a contradiction, and the last step follows from the form of Horn formulas, as each of the remaining clauses contains at least one negative literal. It remains to show that the second step (also called *unit propagation*) is also correct. The formula φ can be satisfied only if each of its clauses is true, and therefore the unit clause l must also be true. Once we assign 1 to l, we can remove all the clauses that contain l (these are already satisfied) and we can also remove $\bar{l}$ from all the remaining clauses as $\bar{l}$ is 0, and therefore the clauses need to be satisfied by other literals.

This shows, that Horn-SAT can be solved in polynomial time. The direct implementation of the algorithm described above is quadratic, however there are even linear implementations.

We already mentioned that there are no polynomial algorithms for the SAT problem in general, but we can use some local search algorithms to attempt to solve the problem. For example, the GSAT algorithm starts with a random truth assignment. If this assignment satisfies the formula, the algorithm ends. Otherwise it flips the truth value for one of the variables – it chooses the variable whose change leads to the smallest number of unsatisfied clauses in the new assignment. There is a small chance to change a random variable (this allows the algorithm to escape local minima in the number of unsatisfied clauses). The WalkSAT algorithm works in a similar way, but instead of picking a variable from the whole formula, it first selects a random clause and picks a variable from it which minimizes the number of previously satisfied clauses that become unsatisfied by the change. It also has a small chance to pick a variable at random.

While none of these algorithms can guarantee that they find the satisfying assignment if it exists, they are very fast, and very often can indeed find the satisfying assignment.

A complete algorithm (such that always finds the assignment, if it exists) can be implemented using backtracking and testing all possible assignments, however, such an algorithm is generally exponential.

The DPLL procedure implements such a backtracking with some improvements. It first removes all clauses that are tautologies, than if a clause becomes empty during the run of the algorithm, it indicates that the current partial assignment cannot satisfy the formula and the DPLL procedure fails. After these simple steps, the DPPL procedure simplifies the formula using unit propagation and so called *pure literal elimination*[10]. If none of the previous step can be applied, the DPLL procedure uses a splitting rule – it selects a literal and tries to call the DPLL procedure twice. Once for each possible truth assignment of that literal. If at least one of these calls succeeds, the formula is satisfiable.

[10] if a literal l is only positive or only negative in the formula, it can be assigned such value $v(l) = 1$ and all the clauses containing it can be removed

Formal Proof Systems

Up to now, we mostly discussed the semantics of the propositional logic, and also defined many different terms semantically using the notion of truth assignments and models. We have defined a consequence of a theory as a formula that is true in all models of the theory. However, in mathematics, we usually do not check all possible models of a theory in order to tell whether a given formula is a consequence or not. Instead, we prove the formula from the axioms of the theory.

In this chapter, we will formalize the notion of proof as a syntactical method that can be used to prove formulas in propositional logic. The formalization will be called a proof system, and, informally, a proof system is a collection of syntactical rules that provide a proof of a given formula in a given theory. The proof is than a finite object that can be built from the axioms of the theory, and if a formula has a proof, it can be found algorithmically.

There are many different proof systems, among them is the tableau method we will describe in detail, and also the Hilbert systems and Gentzen systems. However, a proof system can only be useful, if any formula proved by the system is also valid, and vice versa, if every valid formula can be proven. These two features of a proof system are called soundness and completeness.

Tableau Method

The tableau method is a proof system, where the proof (tableau) of a formula φ in a theory T is a binary labeled tree representing search for a model of T, where φ is not true (a counterexample). If the search fails, the formula is proved and in such a case the tableau is finite. In case there is a counterexample of φ, the tableau can be infinite and there is a branch in the tree that provides the counterexample.

In tableau methods, we assume a fixed and countable language $\mathbb{P}$, in this case, also every theory over $\mathbb{P}$ is countable.

We already mentioned, that every tableau is a labeled binary tree. The nodes in the tree are labeled by *entries*, which are formulas with a *sign* T/F that represent the assumption the formula is true (T) or false (F). The tree will be constructed using the *atomic tableaux* and a set of rules. For a propositional variable p and propositions φ, ψ, the atomic tableaux are given in the figure bellow.

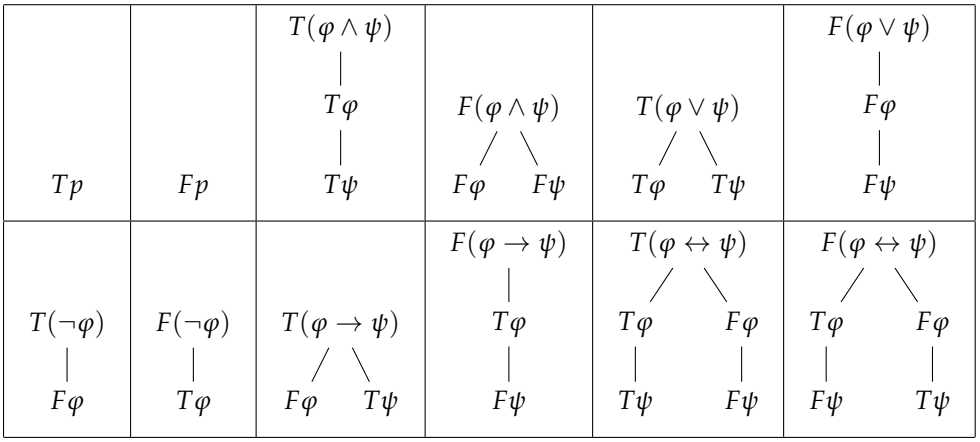

Figure 3: The atomic tableaux

Informal! The labels in each of the tableaux show, whether a formula φ should be true ($T\varphi$) or false ($F\varphi$). The tableaux themselves then "rewrite" the requirement in their root into more simple requirements. For example, the atomic tableau for $T(\varphi \to \psi)$ expresses that a formula $(\varphi \to \psi)$ is true (T), if φ is false ($F\varphi$) or ψ is true $T\psi$. The "or" is expressed by the two sons. On the other hand, the atomic tableau for $F(\varphi \to \psi)$ says, that $\varphi \to \psi$ is false, if φ is true and ψ is false. The "and" is presented by the fact, that both these facts are on a single branch.
While the atomic tableaux are based on the semantics of propositional logic, the tableau method itself is purely syntactic – it only says, how to manipulate tableaux in order to obtain the proof of a formula.

Using the atomic tableaux, we can define the tableau in general.

Definition 5. A *finite tableau* is a binary tree labeled with entries defined inductively as

1. every atomic tableau is a finite tableau,
2. if P is an entry on a branch V in a finite tableau τ and τ' is obtained by adjoining the atomic tableau for P at the end of branch V, then τ' is also a finite tableau,
3. every finite tableau is formed by finite number of steps above.

A *tableau* τ is a (potentially infinite) sequence $\tau_0, \tau_1, \ldots, \tau_n, \ldots$ of finite tableaux such that τ_{n+1} is formed from τ_n by an application of step 2 above, formally, $\tau = \bigcup \tau_n$.

An example of a tableau is shown bellow. In propositional logic, we do not need to repeat the entries that we expand, therefore, we will generally use only the version on the right, where these repeated entries are removed.[11]

The definition above does not specify, how to choose the entry P on branch V for expansion. Later, we define the systematic tableau, where this is specified.

Before we can define the formal notion of proof using the tableau method, we first need to discuss some of the terms related to tableaux.

[11] In predicate logic, some of the repeated entries need to be included in the tableau again.

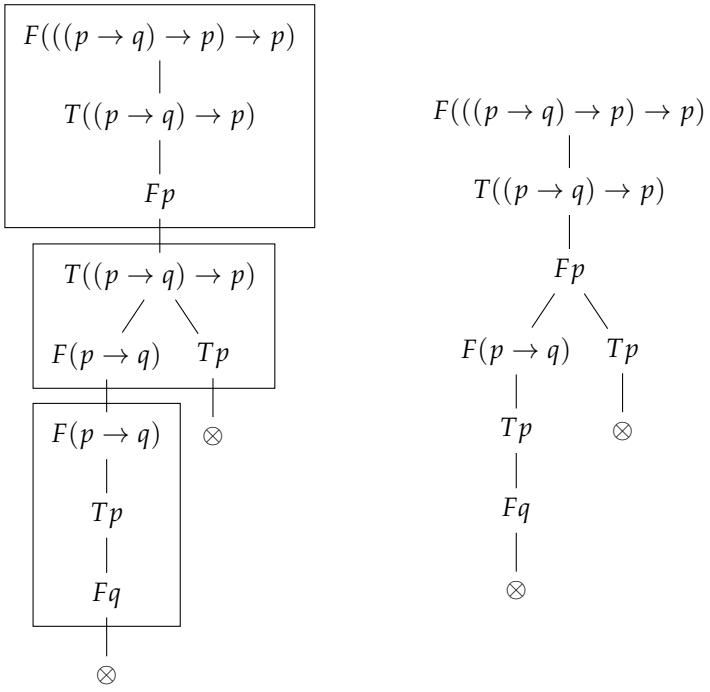

Figure 4: Example tableau. The rectangles on the left show the atomic tableaux used. The version on the right removes the repeated entries. The symbol $\otimes$ denotes a contradictory branch.

For an entry P on a branch V in a tableau τ, we say that *P is reduced on V* if it occurs on V as a root of an atomic tableau. A *branch V is contradictory* if it contains entries $T\varphi$ and $F\varphi$ for some proposition φ, otherwise it is noncontradictory. A *branch V is finished*, if it is contradictory, or every entry on V is reduced on V, and finally, a *tableau τ is finished* if every branch in τ is finished and τ is contradictory, if every branch in τ is contradictory.

A *tableau proof of φ* is a contradictory tableau with the root entry $F\varphi$. A formula *φ is tableau provable* ($\vdash \varphi$) if it has tableau proof. On the other hand, a *refutation of φ by tableau* is a contradictory tableau with the root entry $T\varphi$, and *φ is tableau refutable* it it has a tableau refutation, in this case we write $\vdash \neg\varphi$.

Informal! Why does a tableau proof of φ start with $F\varphi$? Tableaux in fact represent systematic searches for assignments that fulfill the condition expressed by the entry in the root. Therefore, if we cannot find a truth assignment in which φ is false (the tableau for $F\varphi$ is contradictory), then φ must be true in all assignments, and therefore valid. The formal proof of the soundness and completeness of the tableau methods will be discussed shortly.

Figure 5 shows a tableau with the root entry $F(((\neg p \land \neg q) \lor p) \to (\neg p \land \neg q))$. The tableau has three branches, the leftmost one is contradictory, as it contains both $F(\neg p \land \neg q)$ and $T(\neg p \land \neg q)$, the middle one is finished and noncontradictory, as every entry on that branch is expanded on it, and the rightmost one is unfinished, as the entry $F(\neg q)$ is not expanded on that branch.

On the other hand, the tableau in Figure 4 is a tableau proof of the proposition $(((p \to q) \to p) \to p)$, as it starts with the entry $F(((p \to q) \to p) \to p)$ and all its branches are contradictory.

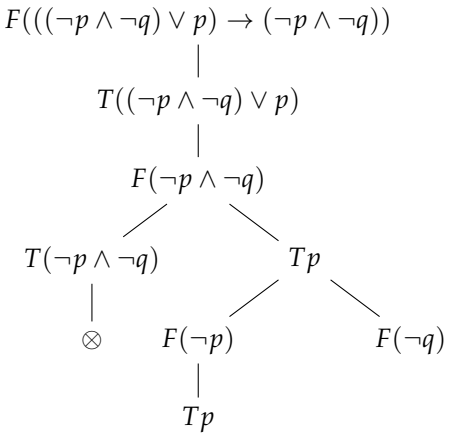

Figure 5: Example tableau. Both left and middle branches are finished. The left one is also contradictory, while the middle one is noncontradictory. The right branch is not finished.

We often need to work with theories, and also prove propositions in a theory. Therefore, the notion of tableau needs to be generalized to the notion of tableau from a theory. Theories provide axioms, these are assumed to be true, and therefore the tableau from a theory T can additionally contain entries of the from $T\varphi$ for an axiom $\varphi \in T$. More formally, a *finite tableau from a theory* T is a generalized tableau with an additional rule – if V is a branch of finite tableau (from T) and $\varphi \in T$, then by adjoining $T\varphi$ at the end of V we obtain a finite tableau from T. The rest of the definitions related to tableaux can be generalized in the same way. *A tableau from* T is a sequence $\tau_0, \tau_1, \ldots, \tau_n, \ldots$ of finite tableaux from T such that τ_{n+1} is formed from τ_n applying the rule 2 (from the definition of tableaux), or the additional rule above, formally $\tau = \bigcup \tau_n$. A *tableau proof of* φ *from* T is a contradictory tableau from T with $F\varphi$ in the root. $T \vdash \varphi$ denotes that φ is tableau provable from T. A *refutation of* φ *by a tableau from* T is a contradictory tableau from T with $T\varphi$ in the root. A branch V of a tableau from T is finished, if it is contradictory, or every entry on V is already reduced on V and, additionally, V contains $T\varphi$ for each $\varphi \in T$.

While the current definition of tableaux is enough for proving propositions in theories. Here, we provide a stricter definition of so called systematic tableau. We will see later, that a systematic tableau is always finished and, in case the tableau is a proof of a proposition, it is also finite. The definition prescribes the precise order of steps to use while constructing tableaux from theories – it specifies which entry in the tableau should be expanded next and also which axiom from the theory should be added next.

Definition 6. Let R be an entry and $T = \{\varphi_0, \varphi_1, \ldots\}$ a theory. Then the *systematic tableau* τ from T for the entry R is the result of the following construction, i.e. $\tau = \bigcup \tau_n$

1. t_0 is the atomic tableau for R, then proceed with the following steps until possible

2. Let P be the leftmost entry in the smallest possible level of the tableau τ_n, such that P is not reduced on some noncontradictory branch through P.

3. Let τ'_n be the tableau obtained from τ_n by adjoining the atomic tableau for P to every noncontradictory branch through P ($\tau'_n = \tau_n$ if no such P exists).

4. Let τ_{n+1} be the tableau obtained from τ'_n by adjoining $T\varphi_n$ to every noncontradictory branch that does not contain $T\varphi_n$ (if φ_n does not exist, $\tau_{n+1} = \tau'_n$).

The first thing to notice is that every systematic tableau is finished. Assume we have a tableau $\tau = \bigcup \tau_n$. If there is a noncontradictory branch in τ, the prefix of this branch is noncontradictory in each τ_n. Therefore, the branch must contain $T\varphi_n$ for each φ_n in T. Let us now assume, there is an entry R, such that R is not reduced on a branch. However, there are only finitely many levels above R in τ and therefore only finitely many entries above R, thus R will be eventually selected in step 2 and reduced in step 3, which is a contradiction with R not being reduced. So, every noncontradictory branch in the tableau is finished (it contains $T\varphi_n$ for each $\varphi_n \in T$, and every entry on the branch is reduced).

Interestingly, if tableau is used as a proof, it is not only finished, it is also finite. More specifically – for every contradictory tableau $\tau = \bigcup t_n$, there is some n such that τ_n is contradictory finite tableau. Why? Let S be the set of nodes in τ that have no pair of contradictory entries $T\varphi$, $F\varphi$ amongst their predecessors. We can imagine this set as a "top part" of the tableau – the root is definitely in this set. If there is a node in the set, all of its predecessors are also there. Such a set S must be finite, because otherwise, by König's lemma, the subtree of τ induced by the set S would have an infinite branch (it is a finitely branching infinite tree), and therefore the tableau τ would not be contradictory. Now, since S is finite, all of the nodes in S belong to levels up to m for some m. Thus every node in level $m+1$ has a pair of contradictory entries among its predecessors. We can now choose n such that the top $m+1$ levels of τ are a subtree of τ_n. Every branch in τ_n now contains a pair of contradictory entries and is thus contradictory.

In the construction of systematic tableaux, we extend only noncontradictory branches, therefore if a systematic tableau (from a theory) is a proof, it is finite (remember that a proof is a contradictory tableau with $F\varphi$ in its root). This is an important results, it shows that if a formula has a proof, we have an algorithm (the construction of systematic tableau) that can find the proof in finite amount of time. It also shows that any proof from a theory depends only on a finite number of axioms from the theory.

Soundness and Completeness

Now, we want to show the soundness and completeness of the tableau method. We start with the soundness and show, that if a formula has a tableau proof from a theory, the formula is also valid in the theory. However, before we get to the proof, we need a definition and a lemma. We say that an *entry P agrees with an assignment v*, if P is $T\varphi$

and $\bar{v}(\varphi) = 1$, or if P is $F\varphi$ and $\bar{v}(\varphi) = 0$. A *branch V agrees with v* if every entry on V agrees with v.

Lemma 2. *Let v be a model of a theory T that agrees with the root entry of a tableau $\tau = \bigcup \tau_n$. Than τ contains a branch that agrees with v.*

Proof. We will find a sequence $V_0, V_1, \ldots$ for every n, such that V_n is a branch in τ_n, $V_n \subseteq V_{n+1}$ and V_n agrees with n. We start by verifying the lemma for all atomic tableaux, thus verifying the base of the induction. For example, if we have $v(p) = 1, v(q) = 0$ and the atomic tableau with root entry $T(p \vee q)$, then v agrees with the root entry, and the branch of the tableau containing Tp also agrees with v. We can check the other atomic tableaux similarly. Now, if τ_{n+1} is obtained from τ_n without extending V_n, we take $V_{n+1} = V_n$. If τ_{n+1} is obtained from τ_n by adjoining $T\varphi$ to V_n for some $\varphi \in T$, let V_{n+1} be this branch, v agrees with V_{n+1} as v is a model of T (and therefore all axioms of T are true in v). Finally, if τ_{n+1} is obtained from τ_n by adjoining the atomic tableau for some entry P on V_n to the end of V_n, we can extend V_n to V_{n+1} as required as P agrees with v and all atomic tableaux are verified (for example, if $P = T(p \vee q)$, and v is as in example on atomic tableaux above, we obtain V_{n+1} by adding $T(p \vee q)$ and Tp to the end of V_n). □

Using the lemma, we can now easily proof the soundness of the tableau method in propositional logic.

Theorem 4 (Soundness of tableau method in propositional logic)**.** *For every theory T and proposition φ, if φ is tableau provable from T, then φ is valid in T, i.e. $T \vdash \varphi \Rightarrow T \vDash \varphi$.*

Proof. If φ is tableau provable from T, there is a contradictory tableau τ from T with the root entry $F\varphi$. Suppose φ is not valid in T. In such a case, there is a model v of the theory T in which φ is false. Therefore, the root entry of the proof ($F\varphi$) agrees with v and by the previous lemma, there is a branch in τ that agrees with v. However, that leads to contradiction as τ is the proof of φ from T, and therefore every branch of τ is contradictory and cannot agree with v. □

The soundness theorem says that whenever we have a tableau proof of a formula in a theory, the formula is valid. However, can we also prove any valid formula using the tableau method? We indeed can, as the completeness theorem states. Again, before we get to the proof of the completeness theorem, we prove a helper lemma, that formally shows that a noncontradictory branch in a finished tableau provides a counterexample.

Lemma 3. *Let V be a noncontradictory branch of a finished tableau τ. Then V agrees with the following assignment v:*

$$v(p) = \begin{cases} 1 & \text{if } Tp \text{ occurs on } V \\ 0 & \text{otherwise} \end{cases}$$

Proof. We prove the lemma by induction on the structure of formulas in entries on V.

- For entry Tp on V, where p is a propositional variable, we have $\bar{v}(p) = 1$ by definition.
- For entry Fp on V, the entry Tp is not on V as V is noncontradictory, and thus we have $\bar{v}(p) = 0$ by definition.
- For entry $T(\varphi \wedge \psi)$, we have both $T\varphi$ and $T\psi$ on V as τ is finished, and by induction, we know $\bar{v}(\varphi) = \bar{v}(\psi) = 1$, therefore $\bar{v}(\varphi \wedge \psi) = 1$ and v agrees with $T(\varphi \wedge \psi)$.
- For entry $F(\varphi \wedge \psi)$, we have $F\varphi$ or $F\psi$ on V as τ is finished, therefore we have $\bar{v}(\varphi) = 0$, or $\bar{v}(\psi) = 0$, which leads to $\bar{v}(\varphi \wedge \psi) = 0$ and thus v agrees with $F(\varphi \wedge \psi)$.

The lemma can be proven for the other possible types of entries (with $\vee, \rightarrow, \leftrightarrow, \neg$) similarly to the last two steps for entries with $\wedge$. □

Using this lemma, it is simple to prove the completeness theorem.

Theorem 5. *For every theory T and proposition φ, if φ is valid in T, then φ is tableau provable from T, i.e. $T \vDash \varphi \Rightarrow T \vdash \varphi$.*

Proof. We will show that an arbitrary finished tableau τ from theory T with root entry $F\varphi$ is contradictory, if φ is valid in T.

Assume (for contradiction), there is noncontradictory branch V in τ. The previous lemma provides an assignment v, such that V agrees with v, therefore also the root entry $F\varphi$ agrees with v and thus $\bar{v}(\varphi) = 0$. Since V is finished, it contains $T\psi$ for every $\psi \in T$, but that means that v is a model of T (V agrees with v, therefore $\bar{v}(\psi) = 1$ for all $\psi \in T$). However, this is contradiction with the assumption that φ is valid in T, therefore every branch in τ is contradictory and τ is a proof of φ from T. □

We can now introduce syntactic definition of the semantic terms defined earlier and discuss the relation between the syntactic and semantic notions. First of all, we define the *set of proposition provable from T*

$$\mathrm{Thm}^{\mathbb{P}}(T) = \{\varphi | \varphi \in \mathrm{VF}_{\mathbb{P}}, T \vdash \varphi\}.$$

We say that a theory T *is inconsistent*, if $T \vdash \bot$, otherwise T *is consistent*. A theory T *is complete*, if it is consistent and every proposition is provable or refutable from T, i.e. if $T \vdash \neg\varphi$ or $T \vdash \varphi$ for every $\varphi \in \mathrm{VF}_{\mathbb{P}}$. A theory T over $\mathbb{P}$ is *an extension* of T' over $\mathbb{P}'$, if $\mathbb{P}' \subseteq \mathbb{P}$ and $\mathrm{Thm}^{\mathbb{P}'}(T') \subseteq \mathrm{Thm}^{\mathbb{P}}(T)$, the extension is *simple*, if $\mathbb{P} = \mathbb{P}'$, and it is *conservative* if $\mathrm{Thm}^{\mathbb{P}'}(T') = \mathrm{Thm}^{\mathbb{P}}(T) \cap \mathrm{VF}_{\mathbb{P}'}$. Two *theories T and T' are equivalent*, if T is an extension of T' and vice versa.

There are strong relations between the syntactic terms introduced above and the semantic terms introduced in the previous chapter. Most of these are corollaries of the soundness and completeness of tableau method. For each theory T and propositions φ, ψ over $\mathbb{P}$

1. $T \vdash \varphi$ if and only if $T \vDash \varphi$,
2. $\mathrm{Thm}^{\mathbb{P}}(T) = \theta^{\mathbb{P}}(T)$,

3. T is inconsistent if and only if T is unsatisfiable, i.e. it has no model,
4. T is complete if and only if T is semantically complete, i.e. it has a single model,
5. (deduction theorem) $T \cup \{\varphi\} \vdash \psi$ if and only if $T \vdash \varphi \rightarrow \psi$.

Another important corollary of the theorems is the compactness theorem.

Theorem 6. *A theory T has a model if and only if every finite subset of T has a model.*

Proof. The implication to right (if a theory has a model, every finite subset has a model) is trivial. In order to prove the other implication, we first realize that if T has no model, it is inconsistent, thus $T \vdash \bot$ and $\bot$ is provable by a systematic tableau τ from T. The tableau is finite, therefore τ is also provable from a finite subset of $T' \subseteq T$ (T' contains the axioms from T that were used in the proof), T' is inconsistent and, therefore, has no model. □

While the compactness theorem is interesting itself, it is also a very strong theorem that can be used to prove other theorems in different parts of mathematics. Consider for example the theorem on infinite k-colorable graphs[12]: a countably infinite graph $G = (V, E)$ is k-colorable if and only if each finite subgraph of G is k-colorable. Again, if the infinite graph is colorable, every finite subgraph is obviously also colorable. The other implication is more interesting. Consider a set of propositional variables $\mathbb{P} = \{p_{u,i} | u \in V, i \in k\}$, where $p_{u,i}$ means that vertex u has color i. We can create a theory T with axioms $p_{u,0} \vee p_{u,1} \vee \cdots \vee p_{u,k-1}$ for each $u \in V$ (every vertex has a color), $\neg(p_{u,i} \wedge p_{u,j})$ for every $u \in V, i < j < k$ (every vertex has only one color), and $\neg(p_{u,i} \wedge p_{v,i})$ for each $\{u, v\} \in E, i < k$ (two vertices connected with an edge do not have the same color). Obviously, G is colorable if and only if T has a model. We only need to show, that every finite $T' \subseteq T$ has a model (and use the compactness theorem). Let G' be a subgraph of G induced by vertices u such that $p_{u,i}$ appears in T' for some i. By assumption, G' is k-colorable, and therefore T' has a model.

[12] A graph is k-colorable if there is a function $c : V \rightarrow k$, such that $c(u) \neq c(v)$ for every edge $\{u, v\} \in E$.

Hilbert systems

A (more traditional) alternative to the tableau method is the Hilbert calculus. In this proof system, formulas are defined using only implication ($\rightarrow$) and negation ($\neg$), and all other logical connectives are defined using these two (we already know, that the set $\{\rightarrow, \neg\}$ is adequate, so this can be done). The Hilbert proof system then defines the following set of schemas of axioms (for two proposition $\varphi, \psi \in \mathrm{VF}_{\mathbb{P}}$):

1. $\varphi \rightarrow (\psi \rightarrow \varphi)$
2. $(\varphi \rightarrow (\psi \rightarrow \chi)) \rightarrow ((\varphi \rightarrow \psi) \rightarrow (\varphi \rightarrow \chi))$
3. $(\neg\varphi \rightarrow \neg\psi) \rightarrow (\psi \rightarrow \varphi)$

Apart from the axioms, there is also a single inference rule: *modus ponens*, which can be expressed as

$$\frac{\varphi, \varphi \to \psi}{\psi}.$$

That means that if φ and $\varphi \to \psi$ are true we can infer that also ψ is true.

A proof of formula φ from a theory T in the Hilbert-style is defined as a finite sequence of formulas $\varphi_0, \varphi_1, \ldots \varphi_n = \varphi$ such that for every $i \leq n$, φ_i is a logical axiom, or an axiom from the theory ($\varphi_n \in T$), or φ_i is inferred from φ_j and φ_k ($j, k < i$) using the modus ponens rule. As with tableau method, a formula φ is provable from T ($T \vdash_H \varphi$), if it has a proof.

For example, we can show, that $\varphi \to \psi$ is provable from $T = \{\neg\varphi\}$ for every ψ.

1. $\neg\varphi$
2. $\neg\varphi \to (\neg\psi \to \neg\varphi)$
3. $\neg\psi \to \neg\varphi$
4. $(\neg\psi \to \neg\varphi) \to (\varphi \to \psi)$
5. $\varphi \to \psi$

The first two steps are axiom of a theory and logical axiom (the schema number 2). The third formula is obtained from the previous two by modus ponens, the fourth one is again an axiom (by schema number 3), and the last one is obtained from formulas number 3 and 4 using modus ponens.

It is easy to prove the soundness of the Hilbert calculus ($T \vdash_H \varphi \Rightarrow T \vDash \varphi$). Logical axioms are tautologies, and axioms from T hold in all models of T, therefore the soundness holds for axioms of any kind, and the modus ponens rule is sound (as can be easily checked using the truth tables of φ, $\varphi \to \psi$, and ψ). Thus, the soundness is proved. The Hilbert calculus is also complete, but we will not show the proof here.

Resolution method

The resolution method is the base of many automated systems – SAT solvers, automated deduction or verification systems and Prolog[13] interpreters. The method assumes the input formulas are given in CNF and it works with a set representation of the formulas (a CNF formula is represented as a set of sets of literals). The method has no explicit axioms, but some of the axioms are implicitly included. It uses a single inferences rule (the resolution rule). Similarly to the tableau method, the resolution method is also a refutation procedure, i.e. it tries to show that a given formula or theory is unsatisfiable. There are several variants of the resolution method that gives more specific rules

[13] Prolog is a programming language based on the specification of programs as sets of Horn formulas.

on when the resolution rule can be applied (e.g. the LI resolution, or the SLD resolution).

Before we describe the resolution method formally, we must define the set representation of CNF formulas. Similarly to our discussion on CNF formulas, a literal is either a propositional variable or its negation. The complementary literal to l is still denoted as $\bar{l}$. A *clause* C is a finite set of literals, and an *empty clause*, denoted as $\square$, is never satisfied. A *formula* S is then a (possibly infinite) set of clauses. An empty formula $\emptyset$ is always satisfied. Infinite formulas represent infinite theories. A *(partial) assignment* $\mathcal{V}$ is a consistent set of literals (i.e. the set does not contain a complementary pair of literals). An assignment is *total*, if it contains a positive or negative literal for each propositional variable. An assignment $\mathcal{V}$ satisfies a formula S (denoted as $\mathcal{V} \vDash S$), if $C \cap \mathcal{V} \neq \emptyset$ for each clause $C \in S$.

For example, the CNF-formula $((\neg p \vee q) \wedge (\neg p \vee \neg q \vee r) \wedge (\neg r \vee \neg s) \wedge s)$ is represented as $S = \{\{\neg p, q\}, \{\neg p, \neg q, r\}, \{\neg r, \neg s\}, \{s\}\}$ and $\mathcal{V} = \{s, \neg r, \neg p\}$ is a satisfying assignment for S.

Informal! While the definitions above are different from those we used previously, they are in fact equivalent. The only reason why they are worded differently is the set representation of the CNF formulas. We know that a formula in CNF is a conjunction of clauses, therefore, we can only represent them as a set of clauses. A clause is a disjunction of literals, and therefore it is again natural to represent each clause as a set of literals. The definition of assignment may seem strange, but the set of literals only says which literals are true and which are false.

There is only one inference rule in resolution – the resolution rule: let C_1 and C_2 are clauses such that $l \in C_1$ and $\bar{l} \in C_2$, then infer a clause C (called a resolvent) such that $C = (C_1 \setminus \{l\}) \cup (C_2 \setminus \{\bar{l}\})$. The resolution rule is a special case of the cut rule:

$$\frac{\varphi \vee \psi \qquad \neg\varphi \vee \chi}{\psi \vee \chi},$$

for any formulas φ, ψ, χ.

It is easy to realize that the resolution rule is sound, i.e. if $\mathcal{V} \vDash C_1$ and $\mathcal{V} \vDash C_2$, then $\mathcal{V} \vDash C$ – the assignment $\mathcal{V}$ cannot contain a pair of complementary literals (by definition), therefore at least one of the intersections $\mathcal{V} \cap (C_1 \setminus \{l\})$ or $\mathcal{V} \cap (C_2 \setminus \{\bar{l}\})$ must be non-empty, therefore $\mathcal{V} \cap C$ is also non-empty.

A *resolution proof (deduction) of a clause C from formula S* is a finite sequence of clauses $C_0, C_1, \ldots, C_n = C$ such that for each $i \leq n$, $C_i \in S$, or C_i is a resolvent of some previous clauses. As usual, a *clause C is provable from formula S* ($S \vdash_R C$), if it has a resolution proof from S. We already mentioned that resolution is used as a refutation procedure. A *resolution refutation of formula S* is a resolution proof $S \vdash_R \square$, and a *formula is resolution refutable*, if there is such a proof.

Let us now show, that resolution is also a sound and complete method. The soundness is simple, and follows from the soundness of the resolution rule.

Theorem 7 (Soundness of resolution)**.** *If a formula S is resolution refutable, it is unsatisfiable.*

Proof. Let $S \vdash_R \square$ and assume (for contradiction) there is an assignment $\mathcal{V}$ such that $\mathcal{V} \models S$. Because the resolution rule is sound, also $\mathcal{V} \models \square$, but that is not possible ($\square$ is never satisfied). $\square$

The proof of completeness is a bit more involved. To this end, we first define resolution trees, which in fact show, how we obtained a proof of a clause. A *resolution tree* of clause C from formula S is a finite binary tree with nodes labeled by clauses such that the root is labeled by C, the leaves are labeled by clauses from S, and every inner node is labeled by the resolvent of its sons. Obviously, there is a resolution tree for C from S if and only if $S \vdash_R C$.

Another important notion is the *resolution closure* of a formula S, denoted as $\mathcal{R}(S)$ and defined as the smallest set containing all clauses of S and closed under the resolution rule, i.e. if $C_1, C_2 \in \mathcal{R}(S)$ and C is the resolvent of C_1 and C_2, then also $C \in \mathcal{R}(S)$. Obviously, $C \in \mathcal{R}(S)$ if and only if $S \vdash_R C$, and all the notions on resolution proofs can be also defined using the resolution trees and closures.

As a simple example of the resolution method, we can show that formula $S = \{\{p,r\},\{q,\neg r\},\{\neg q\},\{\neg p,t\},\{\neg s\},\{s,\neg t\}\}$ is unsatisfiable as $S \vdash_R \square$.

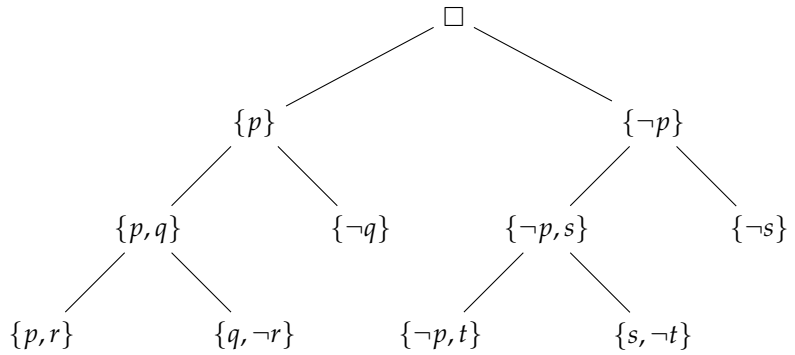

Figure 6: The resolution proof of $S \vdash_R \square$.

We can also compute the resolution closure

$$\begin{aligned}\mathcal{R}(S) = \{&\{p,r\},\{q,\neg r\},\{\neg q\},\{\neg p,t\},\{\neg s\},\{s,\neg t\},\{p,q\},\\ &\{\neg r\},\{r,t\},\{q,t\},\{\neg t\},\{\neg p,s\},\{r,s\},\{t\},\{q\},\\ &\{q,s\},\square,\{\neg p\},\{p\},\{r\},\{s\}\}\,,\end{aligned}$$

and as $\square \in \mathcal{R}(S)$, we also know that S in unsatisfiable.

In the proof of completeness, we will use the notion of reduction by substitution. Let S be a formula and l a literal, we define

$$S^l = \{C \setminus \{\bar{l}\} | l \notin C \in S\}\,.$$

The new formula S^l is in fact equivalent to a formula, where the literal l was assigned a true value ($\top$) and $\bar{l}$ was assigned false value ($\bot$). In such a case, any clause containing l can be removed (as it is satisfied), and $\bar{l}$ is removed from all other clauses. The formula S^l does not

contain any of the literals l and $\bar{l}$, and if S contained a clause $\{\bar{l}\}$, then S^l contains $\square$.

In the proof of completeness of the resolution method, we will need the following lemma.

Lemma 4. *A formula S is satisfiable if and only if S^l or $S^{\bar{l}}$ is satisfiable.*

Proof. Let $\mathcal{V} \vDash S$ and (without loss of generality) $\bar{l} \in \mathcal{V}$. Then, $\mathcal{V} \vDash S^l$, as for clauses C such that $l \notin C \in S$, $\mathcal{V} \vDash C \setminus \{\bar{l}\}$, as $\mathcal{V}$ does not contain $\{\bar{l}\}$ and it is satisfying for each clause $C \in S$.

On the other hand, assume (without loss of generality) $\mathcal{V} \vDash S^l$ for some $\mathcal{V}$. Since neither l nor $\bar{l}$ occur in S^l, $\mathcal{V}' = (\mathcal{V} \setminus \{\bar{l}) \cup \{l\}\} \vDash S^l$. Then, $\mathcal{V}' \vDash S$, as for $C \in S$, such that $l \in C$, also $l \in \mathcal{V}$ and for $C \in S$ not containing l, we have $\mathcal{V}' \vDash (C \setminus \{\bar{l}\}) \in S^l$. $\square$

The reductions of literals can be represented in a binary tree – so called reduction tree. The root of the tree is the formula S and each node N has two sons – N^l and $N^{\bar{l}}$. With the reduction tree, formula S is unsatisfiable if and only if every branch contains $\square$.

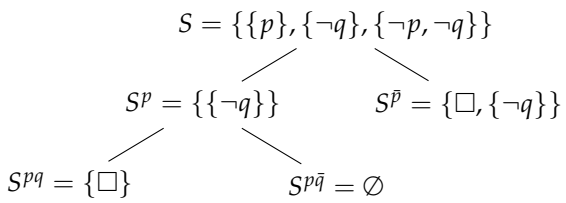

Figure 7: An example of a reduction tree.

Interestingly, since S can be infinite over countable language, the tree can also be infinite. However, if S is unsatisfiable, according to the compactness theorem, there is a finite $S' \subseteq S$ such that S' is unsatisfiable. Therefore, after the reduction of all literals from S', there will be $\square$ on every branch after finitely many steps.

Finally, we can prove the completeness of the resolution. The theorem bellow shows the completeness for finite formulas, the general version is obtained from that theorem by using the compactness, similarly to the discussion on the reduction trees above.

Theorem 8 (completeness of resolution)**.** *If a finite S is unsatisfiable, it is resolution refutable, i.e. $S \vdash_R \square$.*

Proof. We will prove the theorem by induction on the number of variables in S. There is only one unsatisfiable S without variables – $\{\square\}$ and therefore $S \vdash_R \square$ (the proof is the single step $\square$).

Let us now assume, that S is unsatisfiable and contains a literal l. Then, by the previous lemma, S^l and $S^{\bar{l}}$ are unsatisfiable. These contain less literals than S and therefore by induction there are resolution trees T^l and $T^{\bar{l}}$ for derivation of $\square$ from S^l and $S^{\bar{l}}$ respectively. Now, if every leaf of T^l is in S, then T^l is a resolution tree of $\square$ from S and therefore $S \vdash_R \square$. Otherwise, we can append the literal l to each leaf of T^l which is not in S and to all of its predecessors, thus obtaining the resolution tree for $\{l\}$ from S (if the original leaf was not in S, the one with added l will be, as the only difference between S^l and S is the removal of l).

Similarly, by appending $\{\bar{l}\}$ to leaves in $T^{\bar{l}}$ we obtain resolution tree for $\{\bar{l}\}$ from S. Resolving the roots of these trees yields the resolution tree of $\square$ from S. $\square$

We already mentioned that resolution is widely used in different automated systems – SAT solvers, formal verification systems etc. One of the important examples is the Prolog interpreter. Prolog is a programming language, where programs are sets of Horn clauses. The program can then answer queries (goals). As the Prolog programs are limited to Horn formulas, the resolution method can be improved. The Prolog interpreter uses so called SLD resolution which is based on LD resolution and that is in turn based on LI resolution, which is a special case of linear resolution. Therefore, we will now define the linear resolution, show that LI resolution is complete for Horn formulas and finally define the LD and SLD resolution as simple improvements of the LI resolution.

Linear resolution

The general resolution procedure can be further simplified without losing the completeness. We define a linear proof of a clause C from a formula S as a finite sequence of pairs $(C_0, B_0), \ldots, (C_n, B_n)$, such that $C_0 \in S$ and for every $i \leq n$, $B_i \in S$ or $B_i = C_j$ for some $j < i$, and C_{i+1} is a resolvent of C_i and B_i, where $C_{n+1} = C$. In the linear proof C_0 is called the *starting clause*, C_i a *central clause*, and B_i a *side clause*. Again, we say that C is *linearly provable* from S ($S \vdash_L C$), if it has a linear proof from S. A linear proof of $\square$ from S is a *linear refutation* of S and S is *linearly refutable* if $S \vdash \square$.

Obviously, every linear proof can be transformed into a general resolution proof and therefore the linear resolution is also sound. Moreover, is is also complete (we omit the proof of completeness here).

LI-resolution

If we deal with Horn formulas, we can use an even more refined resolution procedure called linear input (LI) resolution. A LI-resolution from a formula S is a linear resolution from S where each side clause B_i is from the input formula S (i.e. B_i cannot be a previously resolved central clause). We write $S \vdash_{LI} C$ to denote that C is provable by LI-resolution from S.

We already defined Horn formulas while discussing the satisfiability problem. The definition from the resolution point of view is similar, the only difference is that we again use the set representation instead of the general one (and also formulas in this case can be infinite, as there is no distinction between theories and formulas in resolution). So, a *Horn clause* is a finite set of literals containing at most one positive literal. A Horn formula is then a (potentially infinite) set of Horn clauses. A clause {p}, where p is a positive literal is called a *fact*, and a clause with exactly one positive literal is called a *rule*. Rules and facts

are also called *program clauses*. A non-empty Horn clause without any positive literal is called a *goal*.

We can easily see that if a Horn formula S is unsatisfiable and it does not contain $\square$, it must contain some fact and some goal. Why? If S does not contain any fact, it is satisfied by setting all the propositional variables to 0. if it does not contain any goal, it is satisfied by setting all variables to 1.

The LI-resolution is complete for Horn formulas, as the following theorem says. The proof of the theorem is similar to the proof of completeness of general resolution.

Theorem 9 (completeness of LI-resolution)**.** *If T is satisfiable Horn formula but $T \cup \{G\}$ is unsatisfiable for some goal G, then $\square$ a LI-resolution from $T \cup \{G\}$ with starting clause G.*

Proof. As in the proof of completeness of general resolution, we use induction on the number of variables, this time in T. By the observation above, T must contain a fact $\{p\}$ for some variable p ($T \cup \{G\}$) is unsatisfiable, therefore it must contain a goal and a fact, G is a goal, so the fact must be in T). By the lemma we used in the proof of completeness of general resolution, $T' = (T \cup \{G\})^p = T^p \cup \{G^p\}$ is unsatisfiable and $G^p = G \setminus \{\neg p\}$. Now, if $G^p = \square$, we have $G = \{\neg p\}$ and thus $\square$ is a resolvent of G and $\{p\} \in T$. Otherwise, since T^p is satisfiable (by the satisfying assignment for T) and has less variables, by induction assumption, there is an LI-resolution of $\square$ form T' starting with G^p. If we now append the literal $\neg p$ to all leaves that are not in $T \cup \{G\}$ (and their predecessors), we have an LI-resolution proof of $\{\neg p\}$ from T, we can resolve it with $\{p\}$ from T to obtain the LI-resolution proof of $\square$ from T. $\square$

Resolution in Prolog

A program in Prolog is a set of program clauses (i.e. rules or facts), an example program is shown bellow (the program contains seven clauses, the numbers indicate line numbers and are not a part of the program):

```
1: p:-q,r.
2: p:-s.
3: q.
4: q:-s.
5: r.
6: s:-t.
7: s.
```

The formulas on lines 3, 5, and 7 are facts, the rest are rules. The symbol `:-` can be interpreted as an implication from right to left ($\leftarrow$). So, the meaning of the clauses is as given bellow:

1. $q \wedge r \rightarrow p$
2. $s \rightarrow p$
3. q
4. $s \rightarrow q$
5. r
6. $t \rightarrow s$
7. s

In Prolog, we want to know whether a query is a consequence of a given program. A query is a conjuction of goals (positive literals), e.g. $p \wedge q$. That means, the question is, whether for a program P and query $(q_1 \wedge \cdots \wedge q_n)$, it holds that $P \vDash (q_1 \wedge \cdots \wedge q_n)$. However, such a question is equivalent to the fact that $P \cup \{\neg q_1, \ldots, \neg q_n\}$ is unsatisfiable, which is equivalent to $\square$ having LI-resolution from $P \cup \{G\}$ starting with goal $G = \{\neg q_1, \ldots, \neg q_n\}$.

In the Prolog interpreter, the clauses are represented as sequences of literals (as opposed to sets of literals as in LI resolution). Therefore, Prolog uses a slightly different version of LI-resolution called the LD-resolution (linear definite). In LD resolution, the resolvent of a goal $(\neg p_1, \ldots, \neg p_{i-1}, \neg p_i, \neg p_{i+1}, \ldots, \neg p_n)$ and a rule $(p_i, \neg q_1, \ldots, \neg q_m)$ is a new goal $(\neg p_1, \ldots, \neg p_{i-1}, \neg q_1, \ldots, \neg q_m, \neg p_{i+1}, \ldots, \neg p_n)$, i.e. one of the negative literals in the current goal is replaced by the negative literals from the rule.

The LD resolution does not specify, which of the negative literals in the goal should be resolved next. This would make programming in Prolog hard, therefore it extends the LD resolution with a selection rule $\mathcal{R}$. Typically, the rule is "select the first literal". More formally, an SLD-resolution via $\mathcal{R}$ is an LD-resolution in which each step (C_i, B_i) we resolve through $\mathcal{R}(C_i)$.

Obviously, any LI-resolution can be expressed as LD-resolution (just use the sequences of literals instead of the sets of literals), and any LD-resolution can be expressed as an SLD-resolution with the correct selection rule (select the literal that was selected in the LD resolution). Therefore, we can see that SLD resolution is complete for Prolog programs.

Further discussion on Prolog is out of the scope of this lecture, so we will omit it here. The important message was to show an application of logic in computer science. Prolog is quite often used in certain areas of artificial intelligence.

This concludes the part of the lecture dedicated to propositional logic. We started with the discussion about the syntax of propositions, explained their semantics and showed some formal proof systems. In the next part of the lecture, we will build on the understanding of concepts from propositional logic and extend them to predicate logic (more precisely to first order logic).

Part II

First-Order Logic

Basic Syntax and Semantics

We are now ready to discuss the predicate logic, mostly the first-order logic. The language of predicate logic is more expressive than the one of propositional logic and allows us to express complex formulas in a much more concise way. In propositional logic, we often needed to use many variables and created long formulas. In predicate logic, some of these can be written more elegantly, as we can now use functions, relations, and logical quantifiers.

This whole part will follow the structure of the previous one – we will again start by the basic syntax and semantics of predicate logic, then we will discuss the logical theories and their models, the tableau method in predicate logic and also the resolution method. While the basic ideas remain the same, there are also important differences. For example, a model of a theory will now be defined as a mathematical structure in which all the axioms of the theory are true, instead of the simpler definition as a truth assignment.

First-order formulas and theories

The symbols used in first-order language can be divided into two groups – symbols of logic and non-logical symbols. The *symbols of logic* consist of *variables* ($x, y, z, \ldots, x_1, x_2, \cdots \in \mathrm{Var}$), logical connectives ($\rightarrow, \wedge, \vee, \leftrightarrow, \neg$), the quantifiers $(\forall x), (\exists x)$ for each variable $x \in \mathrm{Var}$, and parenthesis.

The *non-logical symbols* consist of function symbols $(f, g, \ldots)$, including constant symbols $(c, d, \ldots)$, which are nullary function symbols, and relation (predicate) symbols (P, Q, R). Each function and relation symbol S, has an associated arity $\mathrm{ar}(S) \in \mathbb{N}$ that expresses the number of arguments the symbol takes.

The equality $(=)$ is a special relation symbol that is often considered separately, as it is central to many parts of mathematics and there are even special axioms regarding the equality. Equality is also not considered a non-logical symbol.

The language in first-order logic is determined by the sets of function and relation symbols – these are coupled in the so called *signature*, which is a pair $\langle \mathcal{R}, \mathcal{F} \rangle$ of relation and function symbols with their arities. None of the symbols is the equality symbol. The *language* is then given by a signature $L = \langle \mathcal{R}, \mathcal{F} \rangle$ and by specifying whether the language is with equality or not. A language must always contain at least one relation symbol (either equality or a non-logical one), otherwise, it

would not be possible to write formulas in the language.

The meaning of the symbols in the language is not given by logic, i.e. even the common symbols like $+$ or $\leq$ do not need to represent addition or ordering.

There are many languages that are commonly used in mathematics, for example (all the languages are with equality):

1. $L = \langle\rangle$ is the language of pure equality,
2. $L = \langle c_i \rangle_{i \in \mathbb{N}}$ is the language of countably many constants,
3. $L = \langle \leq \rangle$ is the language of orderings,
4. $L = \langle E \rangle$ is the language of graph theory,
5. $L = \langle +, -, 0 \rangle$ it the language of group theory,
6. $L = \langle +, -, \cdot, 0, 1 \rangle$ it the language of field theory,
7. $L = \langle -, \wedge, \vee, 0, 1 \rangle$ is the language of Boolean algebras, and
8. $L = \langle S, +, \cdot, 0, \leq \rangle$ is the language of arithmetic.

In the examples, 0, 1, and c_i are constant symbols, $-$ and S are unary function symbols, $+, \cdot, \wedge, \vee$ are binary function symbols, and E and $\leq$ are binary relation symbols.

The structure of formulas in first-order language is more complex that the structure of propositional formulas. Before we formally define the formula, we first need to define terms and atomic formulas. Informally, terms are expressions created from variables and functions, while atomic formulas are relations applied to terms.

More formally, a *term of a language* L is defined inductively as

1. Every variable $x \in \text{Var}$ or a constant symbol in L is a term.
2. If f is a function symbol in L with arity $n > 0$ and $t_1, \ldots, t_n$ are terms, then $f(t_1, \ldots, t_n)$ is a term.
3. Every term is obtained by finite amount of applications of steps 1 and 2 above.

A term without variables is called a *ground term*, the set of all terms of a language L is denoted as Term_L. A term that is a part of another term t is called a *subterm* of t. The terms can also be expressed using formation trees. For binary functions, we often use the infix notation, so we write $x + y$ instead of $+(x, y)$.

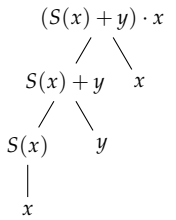

Figure 8: A formation tree of the term $(S(x) + y) \cdot x$.

The simplest type of formulas are the atomic formulas. These are only relations applied to terms. More formally, and *atomic formula of a language* L is an expression $R(t_1, \ldots, t_n)$, where R is a relation symbol in L and $t_1, \ldots, t_n$ are terms of L. The set of all atomic formulas of language L is denoted as AFm_L. Similarly to terms, atomic formulas can also be represented using formation trees from the formation trees of its terms and for binary relations, we use the infix notation, e.g. $\leq (x, y)$ can be written as $(x \leq y)$. For example $(x + y) = 0$, or

$R(f(x), g(y,z), x)$ are atomic formulas (f is a unary function, g and $+$ are binary functions, and R is a ternary relation).

We can finally define formulas in first-order language. The definition is similar to the one in propositional language, but this time the propositional variables are represented by atomic formulas and we additionally have the quantifiers. Formally, a *formula of a language L* is defined inductively by

1. Every atomic formula is a formula
2. If φ and ψ are formulas, $(\varphi \to \psi), (\varphi \wedge \psi), (\varphi \vee \psi), (\varphi \leftrightarrow \psi), (\neg\varphi)$ are also formulas.
3. If φ is a formula and $x \in \mathrm{Var}$ is a variable, then $((\forall x)\varphi)$ and $((\exists x)\varphi)$ are formulas.
4. Every formulas is obtained by a finite application of the steps above.

The set of all formulas of a language L is denoted by Fm_L. A formula that is a part of another formula φ is a subformula of φ. Of course, formulas can also be expressed as their formation tree. An example is on the right.

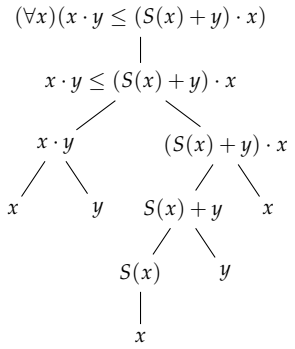

Figure 9: A formation tree of the formula $(\forall x)(x \cdot y \leq (S(x) + y) \cdot x)$. Moreover, $x \cdot y$ and $(S(x) + y) \cdot x$ are roots of formation trees of the terms included in the formula.

As before, we can define some conventions to simplify writing the formulas. After introducing priorities of binary function symbols $(+, \cdot, \dots)$ we can omit parenthesis in the infix notation of terms that are around a subterm formed by a symbol of higher priority. We also introduce the priority of logical connectives similar to the priorities in the propositional logic. The negation and quantifiers $(\neg, (\forall x), (\exists x))$ have the highest priorities, then we have conjunction and disjunction $(\wedge, \vee)$ and, finally, the implication and equivalence $(\to, \leftrightarrow)$ have the lowest priority. Now, we can omit some of the parenthesis in the formulas.

In predicate logic, there is an important difference between so called free and bound (occurrences of) variables, as we deal with each type differently in the semantics. For a formula φ and variable x, an *occurrence of x in φ* is a leaf labeled by x in the formation tree of φ. The occurrence of x in φ is *bound*, if it is in some subformula ψ that starts with $(\forall x)$ or $(\exists x)$. Otherwise the occurrence is *free*. A *variable is free* in a formula, if it has at least one free occurrence in the formula and it is *bound* if it has at least one bound occurrence. A variable can be both free and bound at the same time. For example, in formula $x > 0 \vee (\forall x)(\exists y)(x > y)$ the variable x is both free and bound, as its first occurrence is free and the second one is bound. We will use the notation $\varphi(x_1, \dots, x_n)$ to denote that $x_1, \dots, x_n$ are all the free variables in φ.

A formula is *open*, if it contains no quantifiers. The set of all open formulas in a language L will be denoted as OFm_L. Obviously, $\mathrm{AFm}_L \subsetneq \mathrm{OFm}_L \subsetneq \mathrm{Fm}_L$. On the other hand, a formula is *closed (a sentence)* if it has no free variables. A formula can be both closed and open at the same time, all terms of such formulas are ground terms.

In mathematics, we very often have general theorems and we later use them with a more specific substitutions. This can more formally be

expressed by substituting terms for free variables in formulas, however, we need to be careful, as in some cases the substitution can change the meaning of the formula. For example, if we substituted the term y for x in $(\exists y)(x + y = 1)$, we would change the original meaning of the formula "there is a y such that $y = 1 - x$" to a new meaning that says "y is divisible by 2". We want to avoid such situation while performing the substitution. Therefore, we define a term t is *substitutable* for a variable x in formula φ, if after the substitution of t for all free occurrences of x, none of the variables of t become bound in φ. The new formula is denoted as $\varphi(x/t)$ and we call it an *instance of the formula* φ after a substitution of term t for variable x. Alternatively, we can also define that t is not substitutable for x in φ if x has a free occurrence is a subformula of form $(\exists y)\psi$ or $(\forall y)\psi$ for some variable y in t.

We can also rename the quantified variables, but we again need to be careful. In this case, we would like to obtain an equivalent formula. Let $(Qx)\psi$ us a subformula of φ where Q is either $\forall$ or $\exists$ and y is a variable. Then, if y is substitutable for x in ψ and y is not free in ψ, we can replace the subformula $(Qx)\psi$ with $(Qy)\psi(x/y)$ to obtain a *variant* of φ in subformula $(Qx)\psi$. A variant of φ is obtained by variation of one of more subformulas of φ.

Informal! Creating variants in predicate logic serves some important purposes. One of them is, that we can easily transform any formula with a variable that is both free and bound into its variant, where each variable is "pure", i.e. only free or only bound. Moreover, we will very often create variants from formulas in order to fulfill assumptions such as "variable x is not free in φ".

In propositional logic, models were defined as truth assignments. The truth assignment was enough to tell whether a proposition is true or false. In predicate logic, the situation is a bit more complex. First of all, the values of variables can be taken from a larger set than only $\{0, 1\}$. Moreover, we need to define, what all the functions a relations mean. A natural representation of models in first-order logic is a mathematical structure. A structure is a set and a definition of functions and relations on this set.

More formally, if we have a signature of a language $L = \langle \mathcal{R}, \mathcal{F} \rangle$ and a non-empty set A. A *realization (interpretation) of a relation symbol* $R \in \mathcal{R}$ on set A is any relation $R^A \subseteq A^{\mathrm{ar}(R)}$. A realization of $=$ is the identity relation on A, i.e. $\mathrm{Id}_A = \{(x, x) | x \in A\}$. *A realization (interpretation) of a function symbol* $f \in \mathcal{F}$ is any function $f^A : A^{\mathrm{ar}(f)} \to A$. Specifically, a realization of a constant symbol is some element of A. A *structure for the language* L (L-structure) is a triple $\mathcal{A} = \langle A, \mathcal{R}^A, \mathcal{F}^A \rangle$, where A is a non-empty set called the domain of the structure $\mathcal{A}$, $\mathcal{R}^A$ is a collection of realizations of the relation symbols on A, and $\mathcal{F}^A$ is a collection of realizations of function symbols on A. A structure of the language is also called a model of the language, and the class of all models of a language L will be denoted as $M(L)$.

You probably already know different mathematical structures from

other parts of mathematics, for example:

1. $S = \langle S, \leq \rangle$ is an ordered set, where $\leq$ is reflexive, antisymmetric, and transitive binary relation,
2. $G = \langle V, E \rangle$ is a graph,
3. $\mathbb{Z}_p = \langle \mathbb{Z}_p, +, -, 0 \rangle$ is the additive group of integers modulo p,
4. $\mathbb{Q} = \langle \mathbb{Q}, +, -, 0, 1 \rangle$ is the field of rational numbers,
5. $\mathcal{P}(X) = \langle \mathcal{P}(X), \setminus, \cap, \cup, \emptyset, X \rangle$ is the set algebra over X, and
6. $\mathbb{N} = \langle \mathbb{N}, S, +, \cdot, 0, \leq \rangle$ is the standard model of arithmetic.

But also many other objects can be defined as structures, e.g. the finite automata or even databases.

We now aim to define the truth value of formulas in first-order logic. We already know, that a formula is constructed from atomic formulas, which are in turn constructed from terms. Therefore, in order to define the truth value of a formula, we need to start with the definition of the value of a term. Let t be a term of $L = \langle \mathcal{R}, \mathcal{F} \rangle$ and $\mathcal{A} = \langle A, \mathcal{R}^A, \mathcal{F}^A \rangle$ an L-structure. A *variable assignment* over the domain A is a function $e : \mathrm{Var} \to A$. The *value* $t^A[e]$ *of term* t in structure $\mathcal{A}$ with respect to the assignment e is defined inductively by

1. $x^A[e] = e(x)$, for $x \in \mathrm{Var}$,
2. $(f(t_1, \ldots, t_n))^A[e] = f^A(t_1^A[e], \ldots, t_n^A[e])$ for $f \in \mathcal{F}$.

For a constant symbol $c^A[e] = c^A$, i.e. the values of constants do not depend on the assignment e, and therefore also the value of ground terms does not depend on the assignment. Obviously, the value of a term t depends only on the assignment of variables in t.

We now know, how to compute the values of individual terms, therefore, we can define the value of atomic formulas. Contrary to the values of terms, values of formulas are always from the set $\{0, 1\}$. Let φ be an atomic formula of $L = \langle \mathcal{R}, \mathcal{F} \rangle$ in the from $R(t_1, \ldots, t_n)$, $\mathcal{A} = \langle A, \mathcal{R}^A, \mathcal{F}^A \rangle$ is an L-structure and e a variable assignment over A. The *value* $H_{at}^A(\varphi)[e]$ *of the atomic formula* φ in the structure A with respect to e is

$$H_{at}^A(\varphi)[e] = \begin{cases} 1 & \text{if } (t_1^A[e], \ldots t_n^A[e]) \in R^A \\ 0 & \text{otherwise} \end{cases}$$

Specifically, for the equality relation $=$, the only possible realization is Id_A and therefore $H_{at}^A(t_1 = t_2)[e] = 1$, if $t_1^A[e] = t_2^A[e]$ and 0 otherwise. We can again see that the value of a formula depends only on the assignment of variables in the formula and that the value of a ground formula does not depend on the assignment at all.

We can finally define the value of a general formula. The definition is quite long, but also very similar to the one in propositional logic. In fact, the only difference is in the last two cases. The atomic formulas in this case play the role of the propositional variables.

The value $H^A(\varphi)[e]$ of formula φ in the structure A with respect to e is

1. $H^{\mathcal{A}}(\varphi)[e] = H^{\mathcal{A}}_{at}(\varphi)[e]$ if φ is atomic
2. $H^{\mathcal{A}}(\neg\varphi)[e] = -_1(H^{\mathcal{A}}(\varphi)[e])$
3. $H^{\mathcal{A}}(\varphi \wedge \psi)[e] = \wedge_1(H^{\mathcal{A}}(\varphi)[e], H^{\mathcal{A}}(\psi)[e])$
4. $H^{\mathcal{A}}(\varphi \vee \psi)[e] = \vee_1(H^{\mathcal{A}}(\varphi)[e], H^{\mathcal{A}}(\psi)[e])$
5. $H^{\mathcal{A}}(\varphi \to \psi)[e] = \to_1 (H^{\mathcal{A}}(\varphi)[e], H^{\mathcal{A}}(\psi)[e])$
6. $H^{\mathcal{A}}(\varphi \leftrightarrow \psi)[e] = \leftrightarrow_1 (H^{\mathcal{A}}(\varphi)[e], H^{\mathcal{A}}(\psi)[e])$
7. $H^{\mathcal{A}}((\forall x)\varphi)[e] = \min_{a\in A}(H^{\mathcal{A}}(\varphi)[e(x/a)])$
8. $H^{\mathcal{A}}((\exists x)\varphi)[e] = \max_{a\in A}(H^{\mathcal{A}}(\varphi)[e(x/a)])$

where $-_1, \wedge_1, \vee_1, \to_1, \leftrightarrow_1$ are the function given by the truth tables in the part on propositional logic and $e(x/a)$ is an assignment assigning value a to variable x and otherwise identical to e. We can see that the value of a formula depends only on the assignment of free variables in the formula (we check all possible assignments for the bound variables in steps 7 and 8).

The structure $\mathcal{A}$ satistifies the formula φ if $H^{\mathcal{A}}(\varphi) = 1$, we denote the fact as $\mathcal{A} \vDash \varphi[e]$, otherwise we write $\mathcal{A} \nvDash \varphi[e]$. We can easily check that all the following hold

$$\begin{aligned}
\mathcal{A} \vDash \neg\varphi[e] &\Leftrightarrow \mathcal{A} \nvDash \varphi[e] \\
\mathcal{A} \vDash (\varphi \wedge \psi)[e] &\Leftrightarrow \mathcal{A} \vDash \varphi[e] \text{ and } \mathcal{A} \vDash \psi[e] \\
\mathcal{A} \vDash (\varphi \vee \psi)[e] &\Leftrightarrow \mathcal{A} \vDash \varphi[e] \text{ or } \mathcal{A} \vDash \psi[e] \\
\mathcal{A} \vDash (\varphi \to \psi)[e] &\Leftrightarrow \mathcal{A} \vDash \varphi[e] \text{ implies } \mathcal{A} \vDash \psi[e] \\
\mathcal{A} \vDash (\varphi \leftrightarrow \psi)[e] &\Leftrightarrow \mathcal{A} \vDash \varphi[e] \text{ if and only if } \mathcal{A} \vDash \psi[e] \\
\mathcal{A} \vDash (\forall x)(\varphi)[e] &\Leftrightarrow \mathcal{A} \vDash \varphi[e(x/a)] \text{ for every } a \in A \\
\mathcal{A} \vDash (\exists x)(\varphi)[e] &\Leftrightarrow \mathcal{A} \vDash \varphi[e(x/a)] \text{ for some } a \in A
\end{aligned}$$

Furthermore, if t is substitutable for x in φ, then for every structure $\mathcal{A}$ and assignment e, $\mathcal{A} \vDash \varphi(x/t)[e]$ if and only if $\mathcal{A} \vDash \varphi[e(x/a)]$, where $a = t^{\mathcal{A}}[e]$. If ψ is a variant of φ then $\mathcal{A} \vDash \varphi[e]$ if and only if $\mathcal{A} \vDash \psi[e]$.

As in propositional logic, we can generalize the notion above to validity in structure and in theory. Let φ be a formula of a language L, and $\mathcal{A}$ an L-structure. We say, that φ *is valid in* $\mathcal{A}$, denoted as $\mathcal{A} \vDash \varphi$, if $\mathcal{A} \vDash \varphi[e]$ for every $e : \text{Var} \to A$. We also say that $\mathcal{A}$ *satisfies* φ. Otherwise, we write $\mathcal{A} \nvDash \varphi$. The formula φ *is contradictory in* $\mathcal{A}$ if $\mathcal{A} \vDash \neg\varphi$, i.e. if $\mathcal{A} \nvDash \varphi[e]$ for every e.

We can easily check, that for any structure $\mathcal{A}$ and formulas φ, ψ the following holds:

$$\mathcal{A} \vDash \varphi \Rightarrow \mathcal{A} \nvDash \neg\varphi \tag{1}$$

$$\mathcal{A} \vDash \varphi \wedge \psi \Leftrightarrow \mathcal{A} \vDash \varphi \text{ and } \mathcal{A} \vDash \psi \tag{2}$$

$$\mathcal{A} \vDash \varphi \vee \psi \Leftarrow \mathcal{A} \vDash \varphi \text{ or } \mathcal{A} \vDash \psi \tag{3}$$

$$\mathcal{A} \vDash \varphi \Leftrightarrow \mathcal{A} \vDash (\forall x)\varphi \tag{4}$$

Moreover, if φ and ψ are sentences, the implications in (1) and (3) are in fact equivalences. The last equivalence (4) also shows, that $\mathcal{A} \vDash \varphi$ if

and only if $\mathcal{A} \vDash \psi$, where ψ is the *universal closure* of φ, i.e. the formula $(\forall x_1)(\forall x_1)\dots(\forall x_n)\varphi$, where $x_1, \dots, x_n$ are all the free variables of φ.

A *theory* of a language L is any set T of formulas of L (the *axioms* of the theory). A *model of a theory* T is an L-structure $\mathcal{A}$ such that $\mathcal{A} \vDash \varphi$ for every $\varphi \in T$. We also write $\mathcal{A} \vDash T$ and say that $\mathcal{A}$ satisfies T. The *class of all models* of theory T is $M(T) = \{\mathcal{A} \in M(L) | \mathcal{A} \vDash T\}$. A formula is *valid in* T (true in T) ($T \vDash \varphi$) if $\mathcal{A} \vDash \varphi$ for every model $\mathcal{A}$ of T. Otherwise we write $T \nvDash \varphi$. A formula φ *is contradictory in* T if $T \vDash \neg\varphi$ and φ *is independent in* T if it is neither valid nor contradictory in T. For empty theory T, we can omit T in the notation and $M(T) = M(L)$. In this case, ($\vDash T$) means that the formula φ is *logically valid* (*a tautology*). A *consequence of* T is the set $\theta^L(T)$ of all sentences of L valid in T, i.e.

$$\theta^L(T) = \{\varphi \in \mathrm{Fm}_L \mid T \vDash \varphi \text{ and } \varphi \text{ is a sentence}\}.$$

Informal! The definitions above should closely resemble those we saw in propositional logic, the main difference is in the definition of the model. In propositional logic, we could use truth assignments, while in predicate logic, the model is a structure. The structure, and its definitions of functions and relations in fact give the truth values to the atomic formulas. The atomic formulas than play the role of the propositional variables. Of course, in predicate logic, we also need to take care of the quantifiers, which brings another complexity to the definitions.

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