

# Knowledge Representation

## Lecture 2: Classical Logics

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# What is a Knowledge Representation?

after [Davis, Shrobe and Szolovits, 1993]:

1. Surrogate
2. Expression of ontological commitment
3. Theory of intelligent reasoning
4. Medium of efficient computation
5. Medium of human expression

Today we look at two examples: **propositional logic** and **first-order logic**

# Propositional Logic

Propositional logic is an example of a simple KR

- ▶ Propositional variables abstract atoms of information

- ▶  $a$ : Tom comes to the VU.
- ▶  $c$ : Tom takes the bike.
- ▶  $b$ : Tom takes the tram.
- ▶  $d$ : It is sunny.

- ▶  $e$ : It is raining.
- ▶  $f$ : The bike is broken.
- ▶  $g$ : The tram company is on strike.

- ▶ Operators allow to build complex formulas

1.  $d \rightarrow \neg e$
2.  $a \leftrightarrow (b \vee c)$
3.  $(e \vee f) \rightarrow \neg b$

4.  $(d \wedge \neg f) \rightarrow b$
5.  $g \rightarrow \neg c$
6.  $e \wedge \neg g$

- ▶ A clear semantics defines what these formulas mean
- ▶ Automated reasoning can be used to infer implicit information
  - ▶ *Does Tom come to the VU?*

# Propositional Logics: Assumptions

This KR formalism makes the following assumptions:

- ▶ **Atomic sentences** as building blocks
  - ▶ Represent facts we want to reason about
  - ▶ No inner structure
- ▶ We can build more **complex sentences** using operators
- ▶ Every sentence is either **true** or **false**

# Propositional Logics: Reasoning Problems

What do we want to do with such sentences?

- ▶ Entailment

- ▶ What does **logically follow** from my knowledge?
- ▶ Example: *Does it follow from my knowledge that Tom comes to the VU?*

- ▶ Consistency

- ▶ Is my knowledge **consistent**?
- ▶ Does it describe a possible situation?
- ▶ Example: "It rains", "It is sunny", "If it rains, it is not sunny"  $\Rightarrow$  **not consistent**
- ▶ In context of propositional logic, usually called **satisfiability**

# Propositional Logic: Vocabulary

*Let's define propositional logic formally!*

Our **vocabulary**  $V$  consists of an infinite set of **propositional variables**:

$$V = \{a, b, c, \dots\}$$

These are our **basic building blocks**:

- ▶ Represent sentences we reason about
- ▶ Using letters makes it easier to write complex formulas
- ▶ But we could also use strings: "It rains", "It is sunny".

# Propositional Logic: Interpretations

Adding **meaning**:

- ▶ We don't know the value of propositional variables unless specified.
- ▶ This **incomplete knowledge** is typical for most KR formalisms.
  - ▶ incomplete knowledge of the world
  - ▶ use **reasoning** to find out more
- ▶ To capture this formally, we use **interpretations**.
- ▶ These interpret the value of the propositional variables.
- ▶ and represent **different possibilities**.

# Interpretations in Propositional Logic

An interpretation in propositional logic is a function  $I : V \rightarrow \{\text{true}, \text{false}\}$

Examples:

- ▶  $I_1(\text{"It rains"}) = \text{true}$ ,  $I_1(\text{"The sun shines."}) = \text{false}$
- ▶  $I_2(\text{"It rains"}) = \text{false}$ ,  $I_2(\text{"The sun shines."}) = \text{true}$
- ▶  $I_3(\text{"It rains"}) = \text{true}$ ,  $I_3(\text{"The sun shines."}) = \text{true}$
- ▶  $I_4(\text{"It rains"}) = \text{false}$ ,  $I_4(\text{"The sun shines."}) = \text{false}$

We do not necessarily know which interpretation is the right one.

Propositional formulas restrict the space of interpretations to those that are models (abstract representations) of possible alternatives of the described situation.



# Propositional Logic: Formulas

Propositional formulas are defined as follows:

1. Every **propositional variable** is a formula (stating that this sentence is true)
2. If  $F$  and  $G$  are formulas, then the following are also formulas:
  - ▶  $\neg F$  (“not” / negation)
  - ▶  $F \vee G$  (“or” / disjunction)
  - ▶  $F \wedge G$  (“and” / conjunction)
  - ▶  $F \rightarrow G$  (“implies” / implication)
  - ▶  $F \leftrightarrow G$  (bimplication)
3. Nothing else is a formula.

Examples:

- ▶ *“If it rains, the sun does not shine.”*  
“It rains”  $\rightarrow \neg$  “The sun shines”
- ▶  $(c \wedge d) \leftrightarrow (a \vee \neg b)$

## Exercise: Formalization in Propositional Logic

Assume we have the following propositional variables:

$a$ : "The post brings a parcel."

$d$ : "I take the parcel."

$b$ : "I am at home."

$e$ : "My neighbour takes the parcel."

$c$ : "My neighbour is at home."

$f$ : "The parcel goes back."

Formalize the following facts into propositional logic:

"I am not at home, and the post brings a parcel."	$\neg b \wedge a$
"My neighbour is at home."	$c$
"If I am at home and the post brings a parcel, I take the parcel."	$(b \wedge a) \rightarrow d$
"If I am not at home, I don't take the parcel."	$\neg b \rightarrow \neg d$
"If the post brings a parcel, and neither me nor the neighbour take the parcel, it goes back."	$(a \wedge \neg d \wedge \neg e) \rightarrow f$
"If the post brings a parcel and the neighbour is at home, the neighbour takes the parcel if and only if I am not at home."	$(a \wedge c) \rightarrow (e \leftrightarrow \neg b)$

# Propositional Logic: Semantics

- ▶ What do the complex formulas mean **formally**?
- ▶ Again, **interpretations** capture the meaning
- ▶ Each interpretation  $I : V \rightarrow \{\mathbf{true}, \mathbf{false}\}$  is **extended** to formulas using **truth tables**:

$F$	$\neg F$
<b>false</b>	<b>true</b>
<b>true</b>	<b>false</b>

$F$	$G$	$F \vee G$	$F \wedge G$	$F \rightarrow G$	$F \leftrightarrow G$
<b>false</b>	<b>false</b>	<b>false</b>	<b>false</b>	<b>true</b>	<b>true</b>
<b>false</b>	<b>true</b>	<b>true</b>	<b>false</b>	<b>true</b>	<b>false</b>
<b>true</b>	<b>false</b>	<b>true</b>	<b>false</b>	<b>false</b>	<b>false</b>
<b>true</b>	<b>true</b>	<b>true</b>	<b>true</b>	<b>true</b>	<b>true</b>

Example:  $I(a) = \mathbf{true}$ ,  $I(b) = \mathbf{false}$  and  $I(c) = \mathbf{true}$ :

- ▶  $I(\neg a) = \mathbf{false}$
- ▶  $I(a \wedge b) = \mathbf{false}$
- ▶  $I(a \vee b) = \mathbf{true}$
- ▶  $I((a \wedge b) \rightarrow c) = \mathbf{true}$

## Exercise: Semantics

Exercise:

- ▶ Try to find an interpretation  $I$  that satisfies some of the formulas you have written down in the previous exercise.
- ▶ Of course, you only need to specify the variables that are relevant
- ▶ Example:  $I(a) = \text{true}$  and  $I(b) = \text{false}$  satisfies the first formula

Example solution:

$I(a) = \text{true}$

$I(b) = \text{false}$

$I(c) = \text{true}$

$I(d) = \text{false}$

$I(e) = \text{true}$

$I(f) = \text{false}$

# Propositional Logic: Semantics

$F$	$G$	$F \vee G$	$F \wedge G$
false	false	false	false
false	true	true	false
true	false	true	false
true	true	true	true

We note that conjunction ( $\wedge$ ) and disjunction ( $\vee$ ) are **commutative** and **associative**—the order and how we put brackets does not matter.

We can therefore leave out brackets for nested conjunctions/disjunctions:

- ▶  $(F \vee G) \vee H = F \vee (G \vee H) = F \vee G \vee H$
- ▶  $(F \wedge G) \wedge H = F \wedge (G \wedge H) = F \wedge G \wedge H$

# Propositional Logic: Reasoning Problems

We can now define different reasoning problems:

- ▶ A formula  $F$  is **satisfiable** if there is an interpretation  $I$  s.t.  $I(F) = \mathbf{true}$ 
  - ▶  $I$  is then called a **model of  $F$**
- ▶ A formula  $G$  is **entailed** by a formula  $F$  if every model of  $F$  is also a model of  $G$ 
  - ▶ We write this as  $F \models G$

Having a method for satisfiability is sufficient:

- ▶  $F \models G$  if and only if  $F \wedge \neg G$  is not satisfiable

# Decision Problems and Decision Procedures

A little bit of theory:

- ▶ Satisfiability and entailment are **decision problems**
  - ▶ Decision problems contain questions that have a yes or a no answer.
  - ▶ Example 1: Given a number  $X$ , is  $X$  a **prime number**?
  - ▶ Example 2: Given a program  $P$ , does  $P$  eventually **stop**?
- ▶ If there is a **decision procedure**, we can use it to **decide** decision problems.

Given a decision problem  $P$ , a **decision procedure** for  $P$  is an algorithm with the following properties:

**Soundness** For each instance of  $P$  for which the method returns **yes**, the answer is also yes.

**Completeness** For each instance of  $P$  for which the method returns **no**, the answer is also no.

**Termination** For each instance of  $P$ , the algorithm stops after a **finite number of steps**.

# Decidability

- ▶ Many problems have corresponding decision problems:
  - ▶ Querying **answers** to some problem  $\Rightarrow$  Deciding whether a given answer is true
- ▶ A decision problem is **decidable** if there is a decision procedure for it.
- ▶ Many decision problems are decidable:
  - ▶ Example: *Is  $X$  a prime number?*
- ▶ But there are also problems that are **undecidable**:
  - ▶ Example *Does program  $P$  eventually stop?*
  - ▶ This corresponds to the famous **halting problem**
  - ▶ It is impossible to devise an algorithm that always answers this correctly.



# Reasoning in Propositional Logic in Practice

How do we reason in propositional logic?

- ▶ Deciding **satisfiability** is usually sufficient
- ▶ Checking whether an interpretation is a model is easy
- ▶ But finding a model can be difficult:
  - ▶ with  $n$  variables,  $2^n$  **interpretations to consider!**
- ▶ Whether there is a tractable method is one of the big open problems in computer science
  - ▶ The famous  **$P = NP$  problem**

In practice:

- ▶ **SAT-solvers** are tools to determine satisfiability of propositional formulas
- ▶ Modern SAT-solvers are **highly optimized**, and can often deal with **very large formulas** in **short time**
- ▶ Examples of SAT solvers: **MiniSAT**, **PicoSAT**, **CaDiCaL**, ...

# Applications of Propositional Logic

- ▶ Symbolic AI applications with **limited/finite set of variables**
- ▶ **Software and hardware verification**
- ▶ **NP-complete problems:** find “easy” to verify solution of fixed size
  - ▶ “Easy”: we can describe it using propositional logic
  - ▶ Many puzzles and games like **Sudoku** have this property
  - ▶ Describing the problem with propositional logic and using a SAT solver can be more efficient than implementing a search procedure from scratch
- ▶ **Explaining classifiers obtained through machine learning**

What we cannot do well is reason using the **inner structure** of sentences:

- ▶ “Socrates is a human.”
- ▶ “All humans are mortal.”
- ▶ Entails: “Socrates is mortal.”

# First Order Logic

First Order Logic (FOL) introduces **structure**:

- ▶ Reason about **objects**, **functions** and **relations**
- ▶ Atomic formulas now have structure
- ▶ Additional logical operators

Examples:

- ▶ Petra is the neighbour of Tom:  $\text{neighbours}(\text{Petra}, \text{Tom})$
- ▶ If  $x$  is a neighbour of  $y$ , then  $y$  is a neighbour of  $x$ :  
 $\forall x, y : (\text{neighbours}(x, y) \rightarrow \text{neighbours}(y, x))$
- ▶ Every parent has a child:  $\forall x : (\text{Parent}(x) \rightarrow \exists y : \text{hasChild}(x, y))$

# First Order Logic: Vocabulary

The vocabulary is now more involved:

- ▶ **constants**  $a, b, c, \dots$  denote **specified objects**
- ▶ **variables**  $x, y, z, \dots$  denote **unspecified objects**
- ▶ **function names**  $f, g, h, \dots$  denote **functions**
  - ▶ Every function name  $f$  has an **arity**  $ar(f) \in \mathbb{N}^+$
  - ▶ The arity determines how many arguments the function takes.
  - ▶ *Examples:* **successor**, **sum**       $ar(\text{successor}) = 1$ ,    $ar(\text{sum}) = 2$
- ▶ **predicate names**  $P, Q, R, \dots$  denote **relations**
  - ▶ Again, every predicate name  $P$  has an **arity**  $ar(P)$ ,
  - ▶ which determines how many arguments  $P$  takes.
  - ▶ *Examples:* **neighbours**, **Parent**       $ar(\text{neighbours}) = 2$ ,    $ar(\text{Parent}) = 1$

**Interpretations** determine what these mean, but first we have to talk about syntax.

# First Order Logic: Syntax

The vocabulary is used as follows to build formulas:

- ▶ First, we define **terms**:
  - ▶ terms refer to **objects**
  - ▶ every **constant** and every **variable** is a term
  - ▶ if  $f$  is a function of arity  $n$ , and  $t_1, \dots, t_n$  are terms, then  $f(t_1, \dots, t_n)$  is also a term
  - ▶ Examples:  $x$ , *tom*, *successor*( $x$ ), *sum*( $x, y$ )
- ▶ Then, we define **atoms**:
  - ▶ if  $P$  is a predicate name of arity  $n$ , and  $t_1, \dots, t_n$  are terms, then  $P(t_1, \dots, t_n)$  is an atom
  - ▶ Example: *EvenNumber*( $x$ ), *Neighbours*(*sister*(*anna*), *peter*)
  - ▶ atoms are like the **propositional variables** in propositional logic

# First Order Logic: Syntax

First order **formulas** are now defined as follows:

- ▶ every **atom** is a formula
- ▶ if  $F$  and  $G$  are **formulas**, and  $x$  is a **variable**, then the following are also formulas:
  - ▶  $\neg F$ ,  $F \wedge G$ ,  $F \vee G$ ,  $F \rightarrow G$ ,  $F \leftrightarrow G$
  - ▶  $\exists x : F$  ("there exists", existential quantification)
  - ▶  $\forall x : F$  ("for all", universal quantification)

# First Order Logic: Quantifiers and Sentences

- ▶ The intuitive idea of the quantifiers is as follows:
  - ▶  $\exists x : F$ : for **some**  $x$ ,  $F$  holds
    - ▶ Example:  $\exists x : \text{hasStudentJob}(\text{peter}, x)$
  - ▶  $\forall x : F$ : for **every**  $x$ ,  $F$  holds
    - ▶ Example:  $\forall x : (\text{Student}(x) \rightarrow \text{Person}(x))$
- ▶ this of course only makes sense if  $F$  contains  $x$
- ▶ we then say that  $x$  is **bound** in  $\exists x : F$  /  $\forall x : F$
- ▶ a variable that is not bound is **free**
- ▶ a formula without free variables is called **sentence**

Some examples:

- ▶  $\forall x : (\text{EvenNumber}(x) \leftrightarrow \text{OddNumber}(\text{sum}(x, 1)))$
- ▶  $\forall x : (\text{Parent}(x) \rightarrow \exists y : \text{HasChild}(x, y))$
- ▶  $\forall x : \forall y : (\exists z : \text{DeliversParcelTo}(z, y, x) \wedge \text{AtHome}(x)) \rightarrow \text{ReceivesParcel}(x, y))$

# First-Order Logic: Example

We can use first-order logic to model the parcel example a bit better:

$\neg \textit{AtHome}(\textit{patrick})$

$\textit{DeliversParcelTo}(\textit{mailMan}, \textit{patrick}, \textit{parcel})$

$\textit{Neighbour}(\textit{patrick}, \textit{lucia})$

$\forall x : \forall y : (\exists z : \textit{DeliversParcelTo}(z, y, x) \wedge \textit{AtHome}(x)) \rightarrow \textit{ReceivesParcel}(x, y)$

$\vdots$



# First-Order Logic: Semantics

While the intuition is maybe easy to understand, the semantics is a bit more involved now. This time, **interpretations** are **structures**  $\langle \Delta, \cdot^I \rangle$ :

- ▶  $\Delta$  is a possibly infinite set called the **domain**
  - ▶ The **elements** occurring in the interpretation
  - ▶ these are called **domain elements**
- ▶  $\cdot^I$  is a **function** that interprets constants, functions and variables
- ▶ We then write their interpretations as  $c^I$ ,  $f^I$ ,  $P^I$ , etc.

# Interpreting Terms

The interpretation function  $\cdot^I$  works as follows:

- ▶ Every constant  $c$  is assigned some element  $c^I \in \Delta$  in the domain
- ▶ Every function name  $f$  with arity  $n$  is assigned a function  $f^I : \Delta^n \rightarrow \Delta$ 
  - ▶ It maps tuples of elements to elements
- ▶ Every predicate name  $P$  with arity  $n$  is assigned a relation  $P \subseteq \Delta^n$ 
  - ▶ This means,  $P$  is a set of **tuples**
  - ▶ For example, if  $P$  has arity 1, it is a subset of  $\Delta$
  - ▶ If  $P$  has arity 3, it contains tuples  $\langle a, b, c \rangle$ , where  $a$ ,  $b$  and  $c$  are from  $\Delta$

## Example: First-Order Interpretation

$$\Delta^I = \{a, b, c, d\}$$

$$\textit{patrick}^I = a \quad \textit{mailMan}^I = b \quad \textit{parcel}^I = c$$

$$\textit{Neighbour}^I = \{\langle a, d \rangle, \langle d, a \rangle\} \quad \textit{AtHome}^I = \{a, d\}$$

$$\textit{DeliversParcelTo}^I = \{\langle b, c, a \rangle\} \quad \textit{ReceivesParcel}^I = \{\langle a, c \rangle\}$$

Which sentences are satisfied in this interpretation?

# Interpreting Variables

Since formulas contain **variables**, we need to also take care of them.

A **variable assignment**  $\sigma_I$  for an interpretation  $I$  is a function that assigns to every variable  $x$  an element  $\sigma_I(x)$  of the domain  $\Delta$ .

We can modify variable assignments:

- ▶  $\sigma_I[x \mapsto d]$  is a new variable assignment which is like  $\sigma_I$ , but assigns  $x$  to  $d$

We **extend**  $\sigma_I$  inductively to work on arbitrary terms:

- ▶ For every **constant**  $c$ ,  $\sigma_I(c) = c^I$ .
- ▶ For every **function**  $f$  of arity  $n$  and terms  $t_1, \dots, t_n$ ,  
$$\sigma_I(f(t_1, \dots, t_n)) = f^I(\sigma_I(t_1), \dots, \sigma_I(t_n))$$

In other words,  $\sigma_I$  replaces variables by the assigned elements, and everything else according to the interpretation function.

Using variable assignments, we can now define what it means for a first-order formula to be satisfied in an interpretation.

# Satisfaction of First-Order Formulas

- ▶ Satisfaction of formulas  $F$  is first defined relative to a variable assignment  $\sigma_I$  for the interpretation in question.
- ▶ We write this as  $\sigma_I \models F$  ( $F$  is satisfied under  $\sigma_I$ ).

We define satisfaction under a variable assignment *inductively*:

- ▶ For any atom  $P(t_1, \dots, t_n)$ :  $\sigma_I \models P(t_1, \dots, t_n)$  if  $\langle \sigma_I(t_1), \dots, \sigma_I(t_n) \rangle \in P^I$
- ▶ For any formula  $F$ :  $\sigma_I \models \neg F$  if  $\sigma_I \not\models F$
- ▶ For any formulas  $F$  and  $G$ :
  - ▶  $\sigma_I \models F \vee G$  if  $\sigma_I \models F$  or  $\sigma_I \models G$
  - ▶  $\sigma_I \models F \wedge G$  if  $\sigma_I \models F$  and  $\sigma_I \models G$
  - ▶  $\sigma_I \models F \rightarrow G$  if  $\sigma_I \not\models F$  or  $\sigma_I \models G$
  - ▶  $\sigma_I \models F \leftrightarrow G$  if  $\sigma_I \models F \rightarrow G$  and  $\sigma_I \models G \rightarrow F$
- ▶ For a variable  $x$  and a formula  $F$ :
  - ▶  $\sigma_I \models \exists x : F$  if there is **some** domain element  $d \in \Delta$  s.t.  $\sigma_I[x \mapsto d] \models F$ .
  - ▶  $\sigma_I \models \forall x : F$  if for **every** domain element  $d \in \Delta$ ,  $\sigma_I[x \mapsto d] \models F$ .

# Satisfaction of First-Order Formulas

We have now everything to define when a sentence is **satisfied** by an interpretation, and when an interpretation is a **model** of a formula.

- ▶ Recall: in a sentence, all variables are bound (occur in the scope of a quantifier).
- ▶ If all variables are **bound**, the variable assignment is not relevant anymore.
- ▶ We can then write  $I \models F$  instead of  $\sigma_I \models F$
- ▶ This is how *satisfaction of sentences* is defined:
  - ▶  $I \models F$ :  $I$  **satisfies**  $F$  /  $I$  is a **model** of  $F$ .

## Example: First-Order Interpretation

$$\Delta^I = \{a, b, c, d\}$$

$$\textit{patrick}^I = a \quad \textit{mailMan}^I = b \quad \textit{parcel}^I = c$$

$$\textit{Neighbour}^I = \{\langle a, d \rangle, \langle d, a \rangle\} \quad \textit{AtHome}^I = \{a, d\}$$

$$\textit{DeliversParcelTo}^I = \{\langle b, c, a \rangle\} \quad \textit{ReceivesParcel}^I = \{\langle a, c \rangle\}$$

Some examples of sentences that are satisfied:

- ▶  $\textit{AtHome}(\textit{patrick})$ ,  $\textit{DeliversParcelTo}(\textit{mailMan}, \textit{parcel}, \textit{patrick})$
- ▶  $\exists x : \textit{Neighbour}(\textit{patrick}, x)$
- ▶  $\forall x : \forall y : (\textit{Neighbour}(x, y) \leftrightarrow \textit{Neighbour}(y, x))$
- ▶  $\forall x : \forall y : (\exists z : \textit{DeliversParcelTo}(z, y, x) \wedge \textit{AtHome}(x)) \rightarrow \textit{ReceivesParcel}(x, y))$

# First-Order Logic: Reasoning

- ▶ Satisfiability and entailment are defined the same way as for propositional logic
  - ▶ **Satisfiable**: has some model
  - ▶ **Entailed**:  $F \models G$  if every model of  $F$  is a model of  $G$
- ▶ **Theorem provers** are systems that can be used to determine whether a formula is satisfiable
  - ▶ Examples: **Vampire**, **E**, **SPASS**, **Otter**, etc.
- ▶ However, reasoning in first-order logic is **much much harder** than in propositional logic
- ▶ In particular, it is only **semi-decidable**



# Semi-Decidable?

First order logic is **semi-decidable**:

- ▶ **The good news**: for every **unsatisfiable** formula, an algorithm can in theory show in finite time that it is unsatisfiable.
- ▶ **The bad news**: there are **satisfiable formulas** for which **no algorithm can ever show that they are satisfiable**.
  - ▶ Recall: the **domain** of an interpretation can be arbitrarily large, even **infinite**.
- ▶ This means: even if we want to know whether the formula is **unsatisfiable**, we might have to wait forever.
- ▶ Future technological advancements will not help us here (this is a result from theoretical computer science).

**This limits the usefulness of FOL as KR drastically!**

# And now?

This is very bad news! What now?

There are **restrictions** that make FOL **fully decidable**.

- ▶ Restrict the **number of variables** to 2
- ▶ Restrict the use of **quantifiers**
- ▶ Or **more involved restrictions** on formulas

# From First-Order to Propositional Logic

- ▶ In some cases, we can use SAT solvers also in first-order contexts
- ▶ The basic idea is grounding
- ▶ A formula is ground if it contains no variables
  - ▶  $hasLocation(robot, livingRoom) \vee hasLocation(robot, kitchen)$
- ▶ Ground formulas are like propositional formulas:
  - ▶ every ground atom can be treated as a propositional variable
- ▶ We can thus use SAT-solvers to reason with them.

# From First-Order to Proposition Logic: Grounding

Under certain circumstances, we can focus on ground formulas

- ▶ For instance, if we have the following restrictions:
  - ▶ No functions
  - ▶ All quantifiers are universal quantifiers
  - ▶ No quantifier nested under a negation symbol or (bi-)implication
- ▶ Examples:
  - ▶  $\forall x : (\textit{Student}(x) \rightarrow \textit{Person}(x))$
  - ▶  $\forall x : \forall y : (\textit{Neighbour}(x, y) \leftrightarrow \textit{Neighbour}(y, x))$
  - ▶  $\forall x : \forall y : ((\textit{Person}(x) \wedge \textit{hasChild}(x, y)) \rightarrow \textit{Parent}(x))$
- ▶ Idea: use all groundings of quantified variables:
  - ▶  $\forall x : (\textit{Student}(x) \rightarrow \textit{Person}(x))$   
 $\implies \textit{Student}(\textit{tom}) \rightarrow \textit{Person}(\textit{tom}) \wedge \textit{Student}(\textit{peter}) \rightarrow \textit{Person}(\textit{peter}) \wedge \dots$
  - ▶ We use only the constants that occur in our formulas.
  - ▶ If we have no constant, we use an arbitrary one (e.g. *a*)
  - ▶ The resulting formula is satisfiable if the original one is

# From First-Order to Proposition Logic: Grounding

- ▶ This trick fails once we have functions or existential quantifiers
  - ▶ refer to new objects
  - ▶ there may be many:  $\forall x : A(x) \rightarrow A(f(x)) \Rightarrow A(a), A(f(a)), A(f(f(a))), \dots$
- ▶ negating universal quantifiers has the same effect:  
 $\neg \forall x : A(x) \equiv \exists x. \neg A(x)$

# Conclusion

## Propositional Logic

- ▶ Limited expressivity
- ▶ Reasoning may take exponential time
- ▶ Efficient SAT-solvers

## First-Order Logic

- ▶ High expressivity
- ▶ Semantics a bit involved
- ▶ Only semi-decidable
- ▶ But there are syntactical restrictions that makes it decidable

# Outlook

Next: A family of fragments of first-order logic that

- ▶ can deal with existential and universal quantification
- ▶ is decidable
- ▶ has a more readable syntax (optimized for its use case)
- ▶ has easier semantics
- ▶ allows for efficient reasoning in practical applications

*Meet the **Description Logics!***