

Semantic Data

Chapter 2 : Introduction to first order logic

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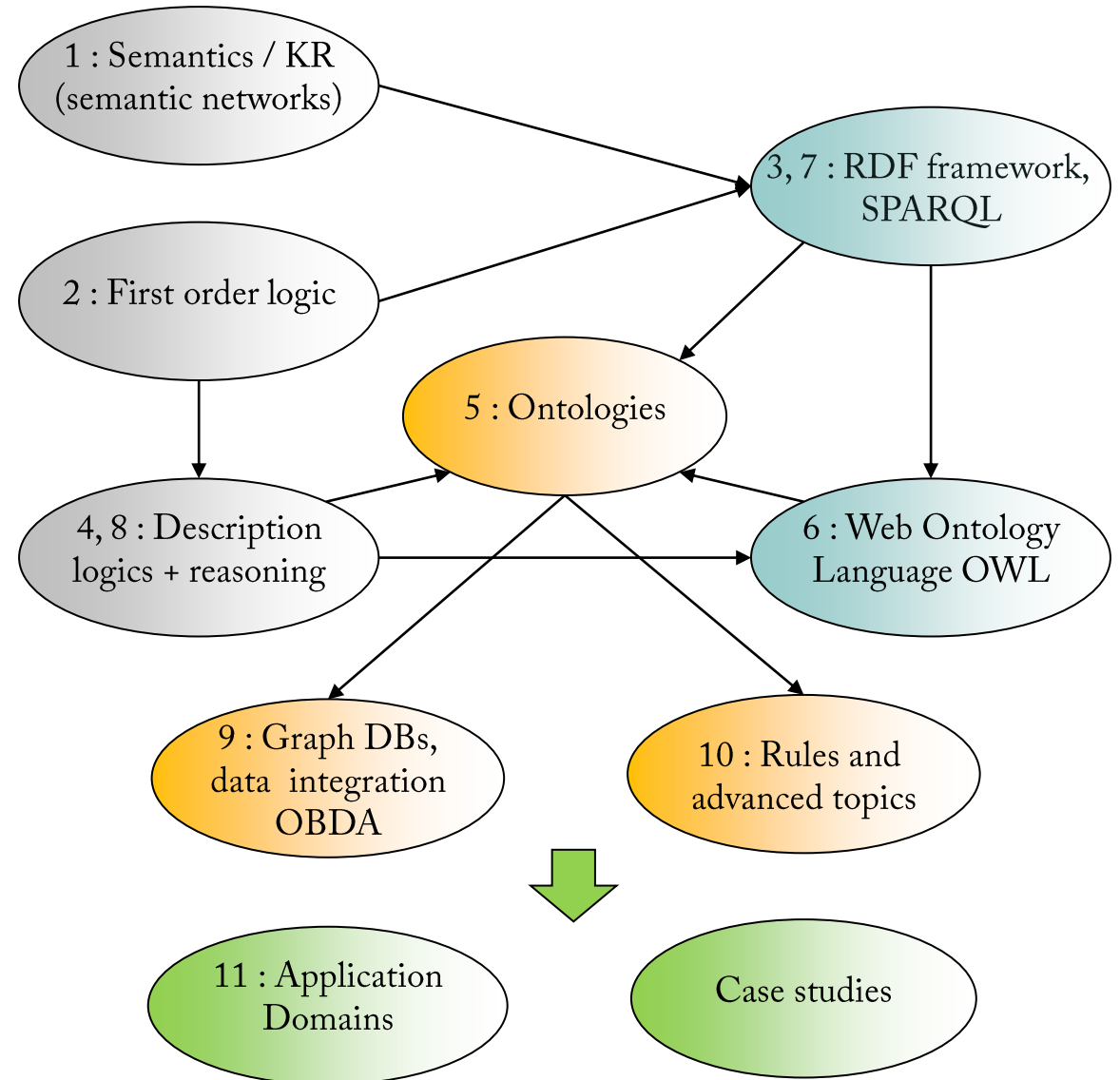
Course content outline

Credits : 5 (theory 25 h, practice 10 h, project 45 h)

Theory (25 h):

1. Semantics and knowledge representation.
2. Introduction to first order logic.
3. The semantic web resource description framework.
4. Description logics.
5. Ontologies and ontology engineering.
6. The Web Ontology Language : OWL.
7. Querying the semantic web : SPARQL.
8. Reasoning with description logics.
9. Data integration and ontology-based data access.
10. Rules and advanced topics.
11. Application domains for semantic data.

Case studies : real cases for genuine business customers;
integrated in the relevant theory sessions.



Sources and recommended readings

- There are no additional required references for this chapter.

- Sources and useful additional readings :
 - *Artificial Intelligence: A Modern Approach (Russel and Norvig 2010).*
 - Chapter 1 from *Handbook of Knowledge Representation (van Harmelen et al. 2007).*

- University courses having partially inspired ideas and examples for this chapter :
 - *Propositional logic and First order logic, M. Frade, university of Minho, Portugal.*
 - *Introduction to first order logic, C. Ghidini and L. Serafini, Fondazione Bruno Kessler, Italy.*

Why study logic?

“Concepts and methods of logic occupy a central place in computer science, insomuch that logic has been called “the calculus of computer science.” (Halpern and Harper 2001).

- ❑ Boolean algebra is widely used in programming languages.
- ❑ SAT Solvers (propositional logic SATisfiability solvers) solve large constraint problems in many domains : planning, diagnostic, design or model checking, PC configuration...
- ❑ First order logic has strong links with the relational database model.
- ❑ Logic is used for knowledge representation, knowledge agents and expert systems.
- ❑ Prolog is a programming language based on logic, used in AI and other areas.
- ❑ The semantic web standards rely on description logics.
- ❑ Program specifications and proofs of program correctness rely on logical formalisms.
- ❑ ...

Types of logics

- **Proposition logic** (PL) : truth value of propositions.
 - For most students, the only type of logic seen so far; a very brief summary is provided in this chapter.
- **First order predicate logic** (FOL) : predicates, objects and functions.
 - Seen in this chapter (without equality). Reasoning algorithms will be kept for description logics.
- **Description logics** (DL) : structured descriptions of concepts.
 - Main formal basis of this course; seen in chapter 4 (presentation) and 8 (reasoning algorithms).
- Other logics :
 - **Modal logic** : apply modalities (*necessarily, possibly; or believes, knows ...*) to propositions.
 - **Temporal logic** (a type of modal logic) : introduce temporal operators (*always, eventually ...*).
 - **Default logics** : one of the main formal approaches to non-monotonic reasoning.
 - **Fuzzy logic** : degrees of truth (in the range $[0, 1]$) to express vagueness of propositions...

Focus point

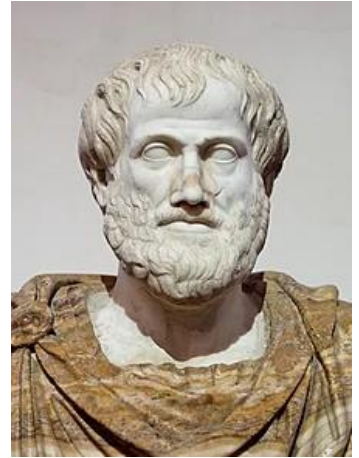
- ❑ Our focus point is not the mathematical logical notation itself (although precise logical notations are used in this chapter).
- ❑ Our focus point is at the knowledge level :
 - What does the representation **mean** ? Can this be defined by **declarative** semantics ?
 - What **reasoning processes / inferences** are supported ?
 - What are their **limitations** ?

Agenda

- 1 First order logic
- 2 Model-theoretic semantics
- 3 Satisfiability, validity, entailment
- 4 Decidability

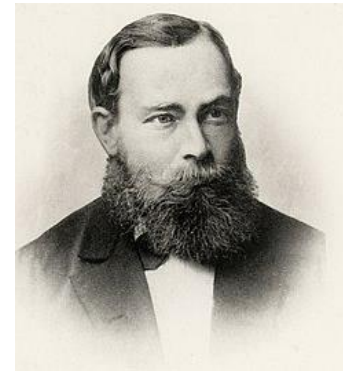
Precursors of first order logic

- The first systematic study of logic goes back to Aristotle (384 – 322 BC).
 - It was part of his body of works called the *Organon*.
 - He built a theory of syllogisms (deductions based on premises and a conclusion).



Among the founding fathers of modern predicate calculus :

- Gottlob Frege (1848, 1925).
- Charles Sanders Peirce (1839, 1914).



A very short reminder of proposition logic

- A logical **proposition** is either *True* or *False*. No other value is possible!
 - *Belgium is a Kingdom.* (a proposition considered as *True* in our real world)
 - *There exists a man who is immortal.* (a proposition considered as *False* in our real world)
 - *Please study !* (not a proposition)

- The **vocabulary** of propositional logic (PL) is made of :
 - **Logical constants** : **true** (also noted \top) and **false** (also noted \perp) (*).
 - **Propositional variables** : a countably infinite set of variables $\Pi = \{p, p_1 \dots p_n, q, r \dots\}$ called atomic formulas.
 - **Logical connectives** : \neg (negation), \wedge (conjunction), \vee (disjunction), \rightarrow (implication), \leftrightarrow (equivalence)

* : **true** and **false**, constants of the logical language, are not the same as truth values (noted *True* and *False*), which characterize states in the real world. However, the distinction between them is often neglected in practice.

A very short reminder of proposition logic ./.

- **Syntax** (Backus-Naur definition) :

$\langle formula \rangle$	$::=$	$p \mid \dots$	(for each $p \in \Pi$)
$\langle formula \rangle$	$::=$	$true \mid false$	(also noted \top and \perp)
$\langle formula \rangle$	$::=$	$\neg \langle formula \rangle$	
$\langle formula \rangle$	$::=$	$(\langle formula \rangle \langle connective \rangle \langle formula \rangle)$	($[], \{\}$ can also be used)
$\langle connective \rangle$	$::=$	$\wedge \mid \vee \mid \rightarrow \mid \leftrightarrow$	

Precedence of operators (higher precedence from left to right) : $\neg, \wedge, \vee, \rightarrow, \leftrightarrow$.

Examples : $a \vee b$ $c \rightarrow d \wedge e$ $p \leftrightarrow \neg(r \vee s)$

- The **semantics** of the logical connectives can be captured by truth tables :

a	b	$\neg(a)$	$\wedge(a, b)$	$\vee(a, b)$	$\rightarrow(a, b)$	$\leftrightarrow(a, b)$
<i>False</i>	<i>False</i>	<i>True</i>	<i>False</i>	<i>False</i>	<i>True</i>	<i>True</i>
<i>False</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>	<i>True</i>	<i>False</i>
<i>True</i>	<i>False</i>	<i>False</i>	<i>False</i>	<i>True</i>	<i>False</i>	<i>False</i>
<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>	<i>True</i>	<i>True</i>	<i>True</i>

Some standard equivalences in propositional logic

- $(a \wedge b) \equiv (b \wedge a)$ commutativity of \wedge
- $(a \vee b) \equiv (b \vee a)$ commutativity of \vee
- $((a \wedge b) \wedge c) \equiv (a \wedge (b \wedge c))$ associativity of \wedge
- $((a \vee b) \vee c) \equiv (a \vee (b \vee c))$ associativity of \vee
- $\neg(\neg a) \equiv a$ double-negation elimination
- $(a \rightarrow b) \equiv (\neg b \rightarrow \neg a)$ contraposition
- $(a \rightarrow b) \equiv (\neg a \vee b)$ implication elimination
- $(a \leftrightarrow b) \equiv ((a \rightarrow b) \wedge (b \rightarrow a))$ biconditional elimination
- $\neg(a \wedge b) \equiv (\neg a \vee \neg b)$ De Morgan's laws
- $\neg(a \vee b) \equiv (\neg a \wedge \neg b)$ De Morgan's laws
- $(a \wedge (b \vee c)) \equiv ((a \wedge b) \vee (a \wedge c))$ distributivity of \wedge over \vee
- $(a \vee (b \wedge c)) \equiv ((a \vee b) \wedge (a \vee c))$ distributivity of \vee over \wedge

These formulas should be known !

Limits of proposition logic

Proposition logic is not suited to express propositions of a parametric nature.
In **first order logic**, new constructs give us richer expressive possibilities.

- How to express that several propositions concern the same individual ?

- *Michel is a Belgian and Michel is married.*



Belgian(Michel)
Married(Michel)

- **Predicates** and **constants**.

- How to express generalizations ?

- *All humans are animals.*

- *There exists at least one bird which cannot fly.*



$\forall x (Human(x) \rightarrow Animal(x))$
 $\exists x (Bird(x) \wedge Cannotfly(x))$

- **Variables**, universal and existential **quantifiers**.

- How to express functional relations ?

- *The wife of Michel is dead.*



Dead(wife(Michel))

- **Functions**.

Vocabulary of first order logic

□ Non logical symbols : FOL allows us to talk about :

- **Constant** symbols, representing **objects** : Jean, House21, 38 ...
- **Predicate**(arity) symbols, representing **relations** : Red(1), Prime(1), BrotherOf(2), PartOf(2) ...
- **Function**(arity) symbols : fatherOf(1), oneMoreThan(1), plus(2) ...

We will note constant and predicates with an uppercase first letter, functions with a lower case.

□ Logical symbols :

- **Variables** : $x, y, z \dots$
- **Connectives** : $\neg, \rightarrow, \wedge, \vee, \leftrightarrow$
- **Quantifiers** :

\forall : **universal** (intuitively : the quantified formula is *True* for all values of the quantified variable).

\exists : **existential** (intuitively : the quantified formula is *True* for at least 1 value of the quantified variable).

Syntax of first order logic

The language of first order logic includes two main elements :

□ **Terms** represent objects (intuitively correspond to natural language noun phrases).

- Terms can be constants or variables, or can be built by applying a function to other terms :

Jean, Michel, 28, x, father(Michel)

□ **Formulas** have a truth value (intuitively correspond to natural language sentences).

- **Atomic formulas** can be built by applying predicates to terms :

Person(Michel), Dead(father(Michel)), Married(Michel, Sarah)

- **Complex formulas** can be built by using same connectives as in propositional logic :

Dead(Michel) \vee Alive(Michel)

- **Complex formulas** can also be built by applying quantifiers to variables :

$\forall x (Human(x) \rightarrow Animal(x)), \exists x (Bird(x) \wedge Cannotfly(x))$

First order logic: formal syntax

□ **Terms** are defined recursively as follows :

- An individual constant or a variable is an **atomic term**.
- If f is a functional symbol of arity n , and t_1, \dots, t_n are terms, then $f(t_1, \dots, t_n)$ is also a **term**.

□ **Formulas** are defined recursively as follows :

If F and G are formulas, P a predicate, t_1, \dots, t_n are terms, and x a variable then :

- $P(t_1, \dots, t_n)$ is an **atomic formula**.
- **true** (also noted \top) and **false** (also noted \perp) are **formulas**.
- $\neg F$, $(F \wedge G)$, $(F \vee G)$, $(F \rightarrow G)$, $(F \leftrightarrow G)$ are **formulas**.
- $\forall x F$ and $\exists x F$ are **formulas**.

Order of higher precedence from left to right is : \forall and \exists , \neg , \wedge , \vee , \rightarrow , \leftrightarrow .

We accept to omit parentheses where possible, using precedence of operators.

We may also use $[]$ and $\{ \}$ to express parentheses when it makes the notation clearer.

Examples of terms and formulas

□ Terms

an individual named Michel

the number 1

the wife of Michel

anything

the # of victories of McEnroe against Borg

the # of victories of McEnroe's brother against Borg

Michel

1

wife(Michel)

x

victories(McEnroe, Borg)

victories((brother(McEnroe), Borg)

□ Formulas

Paris is a city and France is a country.

Paris is in France.

There is at least one city in France.

City(Paris) \wedge Country(France)

Location(Paris, France)

$\exists x$ (City(x) \wedge Location(x, France))

Intuitive understanding of quantifiers and common mistakes

□ \forall means **for all**.

- The sentence $\forall x F$, where F is a logical formula, says that F is *True for every* object x .
- We can use it to make universal statements such as *every man is mortal*.

$\forall x (\text{Human}(x) \rightarrow \text{Mortal}(x))$

⚠ $\forall x (\text{Human}(x) \wedge \text{Mortal}(x))$ has another meaning : all objects are human and all objects are mortal.

□ \exists means **there exists**.

- The sentence $\exists x F$, where F is a logical formula, says that F is *True for at least one* object x .
- We can use it to make existential statements such as *there is at least one red Ferrari*.

$\exists x (\text{Red}(x) \wedge \text{Ferrari}(x))$

⚠ $\exists x (\text{Red}(x) \rightarrow \text{Ferrari}(x))$ has another meaning : there is one object such as, if it is red, it is a Ferrari.

Universal quantification is typically used with implication, existential with conjunction.

Nested quantifiers

□ Expressing more complex sentences will often require several quantifiers.

□ If they are of the same type, the order of quantifiers is not important :

Brothers are siblings.

$\forall x \forall y (\text{Brother}(x, y) \rightarrow \text{Sibling}(x, y))$

□ If they are of mixed types, the order of quantifiers influences the meaning :

Everybody loves somebody.

$\forall x \exists y \text{Loves}(x, y)$

There is someone who is loved by everyone.

$\exists y \forall x \text{Loves}(x, y)$

De Morgan's rules for quantifiers

- There is a relationship between universal and existential quantifiers.
- \forall is a conjunction over the universe of objects while \exists is a disjunction.

As a consequence, they obey De Morgan's rules :

De Morgan's rules without quantifiers:

$$\neg(P \vee Q) \leftrightarrow \neg P \wedge \neg Q$$

$$\neg(P \wedge Q) \leftrightarrow \neg P \vee \neg Q$$

$$P \wedge Q \leftrightarrow \neg(\neg P \vee \neg Q)$$

$$P \vee Q \leftrightarrow \neg(\neg P \wedge \neg Q)$$

Generalized de Morgan's rules:

$$\forall x \neg F \leftrightarrow \neg \exists x F$$

$$\neg \forall x F \leftrightarrow \exists x \neg F$$

$$\forall x F \leftrightarrow \neg \exists x \neg F$$

$$\exists x F \leftrightarrow \neg \forall x \neg F$$

- We could use only one quantifier, but for readability reasons we will keep both.
- Similarly, we could use only conjunction or disjunction, but we will keep both.

Free and bound variables, quantifier scope

- The **scope** of a quantification is the formula to which the quantification is applied :

In the expressions $\forall x F$ or $\exists x F$, the scope of x is the formula F .

- Any occurrence of a variable x in the scope of a quantification $\forall x$ or $\exists x$ is **bound**.
- The variable x in the quantification $\forall x$ or $\exists x$ is said to be **quantified**.
- A variable which is not bound nor quantified is said to be **free**.

- If a formula has free variables, its truth value depends on their values.

In $\forall x P(x, y)$, x is **bound** by the universal quantifier \forall , but y is **free**.

- In order to interpret the formula, we need to **assign a value to the free variable**.

If P is the relation \geq and the domain is the set of naturals, the formula is *True* if $y = 0$, but *False* if $y = 2$.

Caution note

- A variable is **bound** by the **innermost** quantifier mentioning it :

$$\forall x (\text{Country}(x) \vee \exists x \text{Location}(\text{Paris}, x))$$


is equivalent to and should be written as :

$$\forall x (\text{Country}(x) \vee \exists y \text{Location}(\text{Paris}, y))$$

- Using different variables with nested quantifiers is less confusing.

Why is it called first order logic ?

- The name indicates that in FOL we may quantify over individual variables.
- In higher logics we would be able to quantify over other types of elements.

- For example over predicates :

If P is a predicate symbol, $\exists P P(b)$ is not a sentence of FOL.

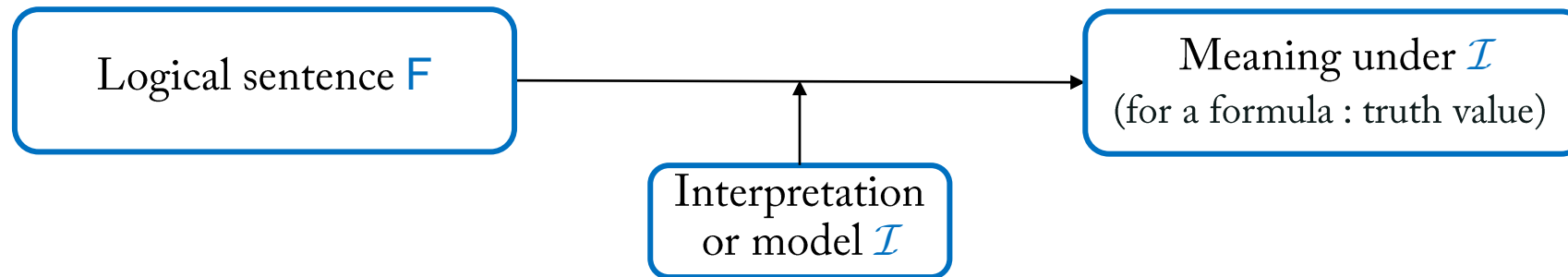
But this is a legitimate sentence of second-order logic.

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Model theoretic semantics

- In the *correspondence theory of truth*, the meaning of logical formulas should be assessed by correspondence to the real world. That is however far too complex.
- Model-theoretic semantics use **models** :
 - A **model** is a description of a small “possible world”, “state of affairs”, or fragment of reality.
 - A model specifies information which allows to **interpret** a sentence and find out its meaning.



- If S is a sentence of the language, $S^{\mathcal{I}}$ represents the **meaning** of S under **interpretation** \mathcal{I} .
- An **interpretation** or **model** \mathcal{I} **satisfies** a sentence F or **is a model of** F if $F^{\mathcal{I}}$ is *True*.
In that case we will note it as $\mathcal{I} \models F$.

Ingredients of model-theoretic semantics

Model-theoretic semantics apply to many logics, with the same ingredients :

1. A definition of **how to build a model**;
2. A definition a priori of the **meaning of atomic syntactic structures** w.r.t. that model;

That definition is provided by the **interpretation function** \mathcal{I} .

3. A formal (recursive) definition of the **meaning of complex syntactic structures** in terms of the atomic ones using the constructors of the language.

The specific details depend on the logic being considered.

Model theoretic semantics of propositional logic (reminder)

- Propositional logic only deals with dependencies between the truth values of statements.
- A PL **model** or **interpretation** \mathcal{I} is an **assignment** from propositional variables into truth values.
 - Example : for the variables p, q and r , $\mathcal{I} = \{p = \text{True}; q = \text{True}; r = \text{False}\}$ is a possible model.
- For any interpretation \mathcal{I} and any formulas F and G , the truth value assigned by \mathcal{I} is determined recursively by the following rules :

- The truth value of any propositional variable $p^{\mathcal{I}}$ is directly specified in \mathcal{I} .
- $\top^{\mathcal{I}} = \text{True}$; $\perp^{\mathcal{I}} = \text{False}$.
- $(\neg F)^{\mathcal{I}}$ is *True* iff $F^{\mathcal{I}}$ is *False*.
- $(F \wedge G)^{\mathcal{I}}$ is *True* iff $F^{\mathcal{I}}$ is *True* and $G^{\mathcal{I}}$ is *True*.
- $(F \vee G)^{\mathcal{I}}$ is *True* iff either $F^{\mathcal{I}}$ is *True* or $G^{\mathcal{I}}$ is *True*.
- $(F \rightarrow G)^{\mathcal{I}}$ is *True* unless $F^{\mathcal{I}}$ is *True* and $G^{\mathcal{I}}$ is *False*.
- $(F \leftrightarrow G)^{\mathcal{I}}$ is *True* iff $F^{\mathcal{I}}$ and $G^{\mathcal{I}}$ are both *True* or both *False*.

Semantics of FOL: intuition

- We will build a definition of FOL semantics by describing how an **interpretation** or **model** maps expressions of the language to the real world.
- FOL formulas can be *True* or *False*; semantics must define their truth value.
- However FOL has additional language elements, including objects.
- The **interpretation domain** is the set of objects referred by an interpretation.

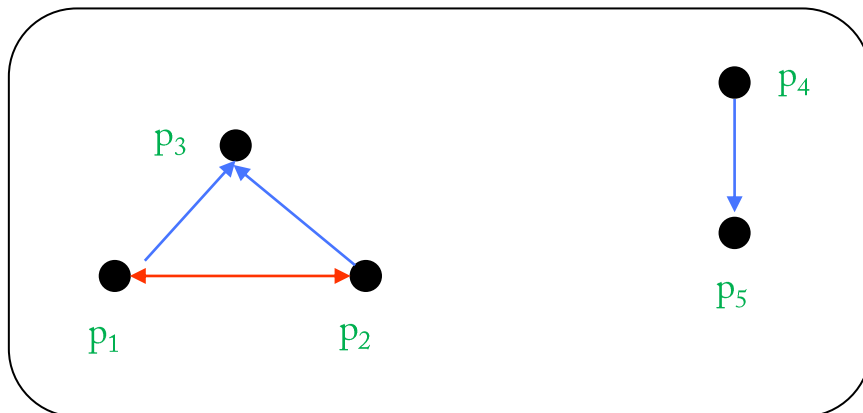
Using that interpretation domain, we will build correspondences between :

- *constant symbols* and **objects** of the interpretation domain;
- *predicate symbols* and **relations** (sets of n-tuples) on the interpretation domain;
- *function symbols* and **functions** on the interpretation domain.

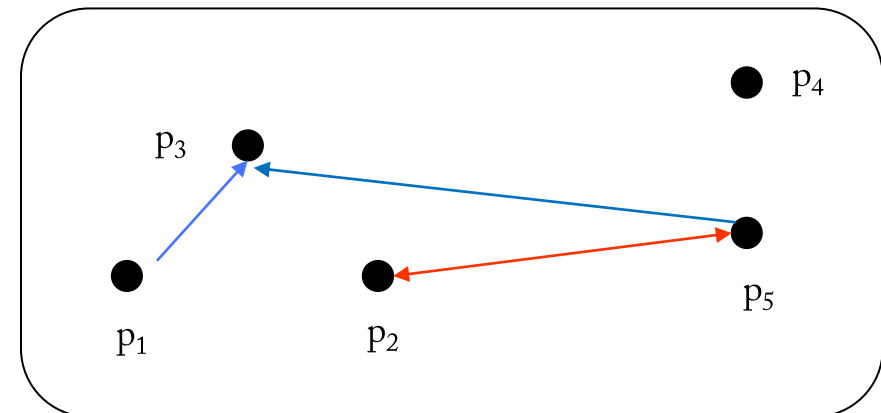
Interpretation domain : example

- KB : constants *Jean*, *René*, *Alice*, *Michel*, *Sarah*, a function *manager* and a predicate *Work_with*.
- A possible interpretation \mathcal{I}_1 is:
 - $Jean^{\mathcal{I}_1} = p_1$; $René^{\mathcal{I}_1} = p_2$; $Alice^{\mathcal{I}_1} = p_3$;
 $Michel^{\mathcal{I}_1} = p_4$; $Sarah^{\mathcal{I}_1} = p_5$.
 - $manager^{\mathcal{I}_1} : \{p_1 \rightarrow p_3; p_2 \rightarrow p_3; p_4 \rightarrow p_5\}$
 - $Works_with^{\mathcal{I}_1} : True$ for $\{<p_1, p_2>\}$
- Another model with the same domain.
Works_with(*Jean*, *René*) is *True* under \mathcal{I}_1 ,
but not under \mathcal{I}_2 .

Interpretation domain and model \mathcal{I}_1



Interpretation domain and model \mathcal{I}_2



First order logic semantics

A **first order interpretation** \mathcal{I} is a triple $\{ \Delta^{\mathcal{I}}, \cdot^{\mathcal{I}}, \alpha \}$, where

- $\Delta^{\mathcal{I}}$ is a **non-empty** set called the **interpretation domain**;
- $\cdot^{\mathcal{I}}$ is an **interpretation function** associating :
 - to every **constant** symbol c_i an element of the domain $c_i^{\mathcal{I}} \in \Delta^{\mathcal{I}}$,
 - to every **predicate** symbol P_i of arity n a relation $P_i^{\mathcal{I}} : \Delta^{\mathcal{I} \ n} \rightarrow \{True, False\}^{(*)}$,
 - to every **function** symbol f_i of arity m a function $f_i^{\mathcal{I}} : \Delta^{\mathcal{I} \ m} \rightarrow \Delta^{\mathcal{I}}$.
- α is an **assignment** associating to each free variable x_i an element of the domain : $\alpha(x_i) \in \Delta^{\mathcal{I}}$.
- We will note $t^{\mathcal{I}}$ and $F^{\mathcal{I}}$ the meaning assigned to any term t or formula F by \mathcal{I} .

* : the notation $\Delta^{\mathcal{I} \ n}$ denotes the n -ary cartesian product of $\Delta^{\mathcal{I}}$.

First order logic semantics : terms

The interpretation function $\cdot^{\mathcal{I}}$ is defined recursively.

□ For **terms** :

- For a **constant** c , its interpretation is given directly by the interpretation function $c^{\mathcal{I}}$.
- For a **free variable** x , its interpretation is given by the variable assignment $\alpha : x^{\mathcal{I}} = \alpha(x)$.
- For a **functional symbol** f of arity m , if $t_1, t_2 \dots t_m$ are terms : $f(t_1, t_2 \dots t_m)^{\mathcal{I}} = f^{\mathcal{I}}(t_1^{\mathcal{I}}, t_2^{\mathcal{I}} \dots t_m^{\mathcal{I}})$.

First order logic semantics : formulas

□ For atomic formulas :

- If P is a predicate symbol of arity n , and if $t_1, t_2 \dots t_n$ are terms : $P(t_1, t_2 \dots t_n)^{\mathcal{I}} = P^{\mathcal{I}}(t_1^{\mathcal{I}}, t_2^{\mathcal{I}} \dots t_n^{\mathcal{I}})$.

□ For logical constants and formulas with connectives ($\neg, \rightarrow, \wedge, \vee, \leftrightarrow$) :

- Their interpretation follows the same rules as in propositional logic.

□ For formulas with quantifiers :

Let us note $\mathcal{I}_{x=d}$ the variant of interpretation \mathcal{I} where free variable x has value d ($d \in \Delta^{\mathcal{I}}$). Then :

- $(\forall x F)^{\mathcal{I}} = \text{True}$ iff for all elements $d \in \Delta^{\mathcal{I}}$, $F^{\mathcal{I}_{x=d}} = \text{True}$.
- $(\exists x F)^{\mathcal{I}} = \text{True}$ iff there is exists one element $d \in \Delta^{\mathcal{I}}$ such as $F^{\mathcal{I}_{x=d}} = \text{True}$.

Agenda

1

First order logic

2

Model-theoretic semantics

3

Satisfiability, validity, entailment

4

Decidability

Fundamental semantic concepts

We will use model-theoretic semantics to define key concepts usable in any logic :

- *Satisfiability.*
- *Validity.*
- *Logical equivalence.*
- *Entailment, or logical consequence.*

Satisfiability and validity

- An interpretation or model \mathcal{I} **satisfies** a formula F , noted $\mathcal{I} \models F$, if $F^{\mathcal{I}}$ is *True*.
 - In that case we say that \mathcal{I} is a **model of** F .
 - If F is *False* under interpretation \mathcal{I} , we say that \mathcal{I} does not satisfy F , noted $\mathcal{I} \not\models F$.
- A formula F is **satisfiable** if it has at least one model.
- A formula F is **valid**, noted $\models F$, if it is *True* for all models.
- A formula F is **unsatisfiable**, noted $\not\models F$, if it is *False* under all models.
 - A formula is valid iff its negation is unsatisfiable : $\models F$ iff $\not\models \neg F$.

Examples :

$p \vee q$ is **satisfiable**.

$p \vee \neg p$ is **valid**.

$p \wedge \neg p$ is **unsatisfiable**.

Satisfiability for set of formulas

- An interpretation or model \mathcal{I} **satisfies** a set of formulas E if it satisfies each formula in E .
- A set of formulas E is **satisfiable** if E has at least one model :
there is at least one interpretation \mathcal{I} satisfying all formulas in E .
- A set of formulas E is **unsatisfiable** if it is not possible to satisfy it :
for every interpretation \mathcal{I} , we have at least one formula $F \in E$ where $\mathcal{I} \not\models F$.

Consequence :

- The models of a finite set of formulas $\{F_1, \dots, F_n\}$ are the same as the models of the unique formula being the conjunction $F_1 \wedge \dots \wedge F_n$.

Entailment and logical equivalence

- A formula G **entails** a formula F , noted $G \models F$, if every model of G is a model of F (*).
 - We also say then that F is a **logical consequence** of G .

- A set of formulas E **entails** a formula F , or F is a logical consequence of E , noted $E \models F$, if every model of E is a model of F .
 - If E is empty, E can only entail F if F is valid : $\emptyset \models F$ is equivalent to $\models F$.

- Two formulas F and G are **logically equivalent**, noted $F \equiv G$, if they have the same models :
 $F^{\mathcal{I}} = G^{\mathcal{I}}$ for any interpretation \mathcal{I} .

* : The symbol \models is used to represent both entailment and satisfiability, which are tightly connected.

Some useful results

For any sentences F and G :

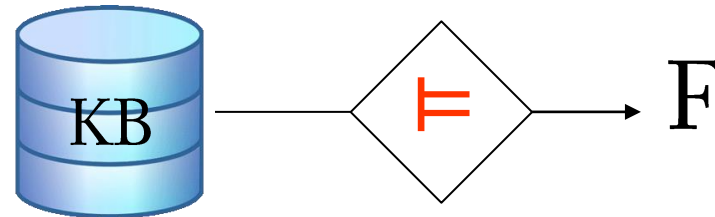
- $G \models F$ iff $\models (G \rightarrow F)$: G entails F iff $G \rightarrow F$ is valid (**deduction theorem**).
- $G \models F$ iff $\not\models (G \wedge \neg F)$: G entails F iff $G \wedge \neg F$ is unsatisfiable.

These results will be very useful when we discuss logical reasoners and decision procedures.

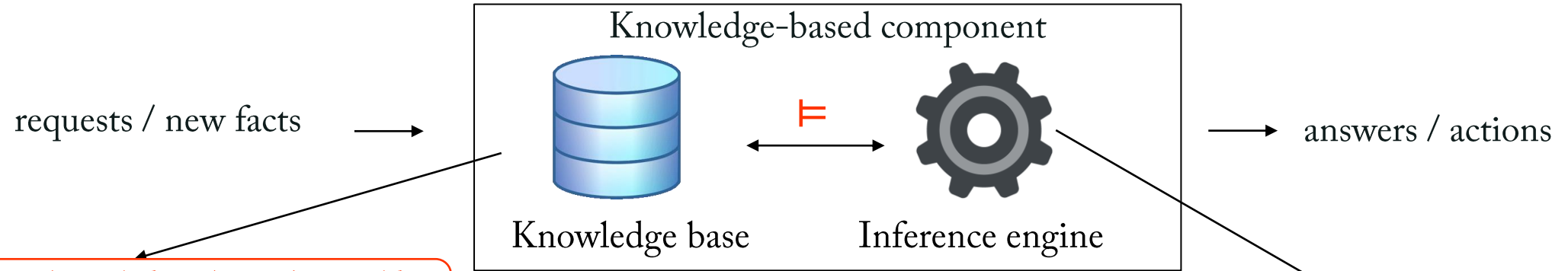
Logical consequence and knowledge-based inferences

- A knowledge base **KB** is a set of sentences expressed in a knowledge representation language.
 - A knowledge base using logic is a set of logical formulas.
 - It can be considered as a single formula being the conjunction of all formulas included in it.
- In a KB using logic, **inference is related to entailment** :

A knowledge base **KB** allows to infer a formula **F** iff **F is entailed by KB**, or
iff **F** is *True* in all models where **KB** is *True*.



Updated framework of reference



How to express knowledge about the world?
Use a logical formalism

In what kind of databases shall we store this type of data / knowledge ?

Which formalism(s) to represent and access the meaning of web resources?

How to perform deductions?
Use entailment.

Monotonicity, closed-world and open-world assumptions

- **Monotonicity** : the set of entailed sentences can only **increase** as information is added to the KB.
 - Facts known as *True* remain *True* regardless of what is else is in the KB or what is added to it :
For any sentences **F** and **G**, if $KB \models F$ then $KB \wedge G \models F$.
 - Default inheritance (cf. chapter 1) leads to non-monotonicity.
- **Closed-world assumption (CWA)** : sentences not known to be *True* are assumed *False*.
 - Database systems usually make the closed-world assumption.
 - CWA is associated with non-monotonicity : if a formula **F** is not in **KB** then $KB \models \neg F$, but $KB \wedge F \models F$.
- **Open-world assumption (OWA)** : sentences not known to be *True* or *False* are *unknown*.
 - Description logics and the semantic web make the open-world assumption, as the web can always provide more information.
 - OWA is associated with monotonicity.

Agenda

- 1 First order logic
- 2 Model-theoretic semantics
- 3 Satisfiability, validity, entailment
- 4 Decidability

Questions regarding the use of FOL for reasoning

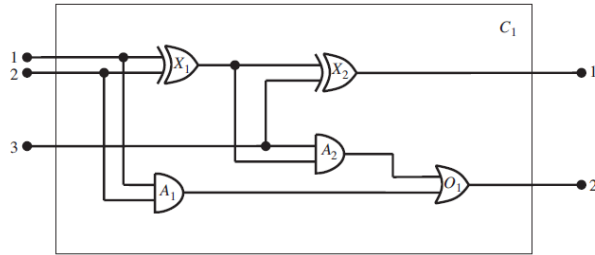
- Entailment is our inference mechanism of choice.
- The model-theoretic semantics on which it relies apply to many logics.
- Is **first order** logic our logical formalism of choice ?

We will look at this question from two points of view :

1. Suitability of the formalism for representing knowledge.
2. Capability to define efficient reasoning algorithms.

1. Knowledge representation in FOL

Excerpt from an example KB about circuits and gates (*source Russel and Norwig 2010*)



1. If two terminals are connected, they have the same signal.
2. The signal at every terminal is either 1 or 0.
3. Connected is commutative.
4. There are four types of gates.
5. An AND gate's output is 0 if and only if any of its inputs is 0.
6. An OR gate's output is 1 if and only if any of its inputs is 1.
7. An XOR gate's output is 1 if and only if its inputs are different.
8. Gates are circuits.
- ...

1. $\forall t1, \forall t2 (Terminal(t1) \wedge Terminal(t2) \wedge Connected(t1, t2) \rightarrow Signal(t1)=Signal(t2))$
2. $\forall t (Terminal(t) \rightarrow Signal(t)=1 \vee Signal(t)=0)$
3. $\forall t1, \forall t2 (Connected(t1, t2) \leftrightarrow Connected(t2, t1))$
4. $\forall g (Gate(g) \wedge k = Type(g) \rightarrow k = AND \vee k = OR \vee k = XOR \vee k = NOT)$
5. $\forall g (Gate(g) \wedge Type(g)=AND \rightarrow Signal(Out(1, g))=0 \leftrightarrow \exists n Signal(In(n, g))=0)$
6. $\forall g (Gate(g) \wedge Type(g)=OR \rightarrow Signal(Out(1, g))=1 \leftrightarrow \exists n Signal(In(n, g))=1)$
7. $\forall g Gate(g) \wedge Type(g)=XOR \Rightarrow Signal(Out(1, g))=1 \Leftrightarrow Signal(In(1, g)) \neq Signal(In(2, g))$
8. $\forall g (Gate(g) \rightarrow Circuit(g))$

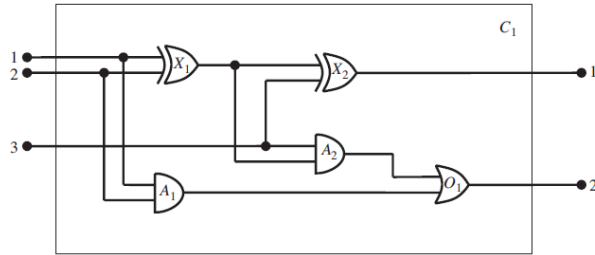
...

Encoding a particular circuit :

- a. $Circuit(C1) \wedge Arity(C1, 3, 2)$
- b. $Gate(X1) \wedge Type(X1)=XOR$
- c. ...

1. Knowledge representation in FOL

Excerpt from an example KB about circuits and gates (source Russel and Norwig 2010)



- ✓ It can be done.
- ✓ It supports precise axiomatization.
- ✗ Knowledge is not easy to structure and to manage in FOL:
 - Where are the concept definitions ?
 - Where is the hierarchy ?

1. $\forall t1, \forall t2 (Terminal(t1) \wedge Terminal(t2) \wedge Connected(t1, t2) \rightarrow Signal(t1)=Signal(t2))$
2. $\forall t (Terminal(t) \rightarrow Signal(t)=1 \vee Signal(t)=0)$
3. $\forall t1, \forall t2 (Connected(t1, t2) \leftrightarrow Connected(t2, t1))$
4. $\forall g (Gate(g) \wedge k = Type(g) \rightarrow k = AND \vee k = OR \vee k = XOR \vee k = NOT)$
5. $\forall g (Gate(g) \wedge Type(g)=AND \rightarrow Signal(Out(1, g))=0 \leftrightarrow \exists n Signal(In(n, g))=0)$
6. $\forall g (Gate(g) \wedge Type(g)=OR \rightarrow Signal(Out(1, g))=1 \leftrightarrow \exists n Signal(In(n, g))=1)$
7. $\forall g Gate(g) \wedge Type(g)=XOR \Rightarrow Signal(Out(1, g))=1 \Leftrightarrow Signal(In(1, g)) \neq Signal(In(2, g))$
8. 12. $\forall g (Gate(g) \rightarrow Circuit(g))$

...

Encoding a particular circuit :

- a. $Circuit(C1) \wedge Arity(C1, 3, 2)$
- b. $Gate(X1) \wedge Type(X1)=XOR$
- c. ...

2. Logical reasoning

- A **logical reasoner** is an algorithm checking entailment.
- Let us note $M(F)$ the set of all possible models of a formula F . Then :
$$KB \models F \text{ iff } M(KB) \subseteq M(F) \quad \text{by definition of entailment (note the sense of the inclusion).}$$
- Entailment can be tested by a systematic model checking algorithm :
 - The algorithm will enumerate all possible models and checks if F is true whenever KB is true.
- This algorithm will only terminate if the number of possible models is finite.
- Is that the case in propositional logic?
 - Yes, why ?
 - If n is the number of propositional variables, there are 2^n of combinations of truth values.

A truth table enumeration algorithm for PL entailment

```
function TT-ENTAILS?(KB,  $\alpha$ ) returns true or false
  inputs: KB, the knowledge base, a sentence in propositional logic
          $\alpha$ , the query, a sentence in propositional logic

  symbols  $\leftarrow$  a list of the proposition symbols in KB and  $\alpha$ 
  return TT-CHECK-ALL(KB,  $\alpha$ , symbols, { })

function TT-CHECK-ALL(KB,  $\alpha$ , symbols, model) returns true or false
  if EMPTY?(symbols) then
    if PL-TRUE?(KB, model) then return PL-TRUE?( $\alpha$ , model)
    else return true // when KB is false, always return true
  else do
    P  $\leftarrow$  FIRST(symbols)
    rest  $\leftarrow$  REST(symbols)
    return (TT-CHECK-ALL(KB,  $\alpha$ , rest, model  $\cup$  {P = true})
           and
           TT-CHECK-ALL(KB,  $\alpha$ , rest, model  $\cup$  {P = false}))
```

Figure 7.10 A truth-table enumeration algorithm for deciding propositional entailment. (TT stands for truth table.) PL-TRUE? returns *true* if a sentence holds within a model. The variable *model* represents a partial model—an assignment to some of the symbols. The keyword “and” is used here as a logical operation on its two arguments, returning *true* or *false*.

(Source Russel and Norvig 2010)

- ❑ This algorithm performs a recursive enumeration on a finite space of assignments to propositional symbols.
- ❑ It is **sound**: it only derives formulas entailed by the KB.
- ❑ It is **complete**: it can derive any formula entailed by KB and always terminate.
- ❑ Its **time complexity** is $O(2^n)$, where n is the number of propositional symbols, and space complexity is $O(n)$.
- ❑ More efficient algorithms exist, but propositional satisfiability is **NP-complete** (Cook’s theorem).

Every known inference algorithm for propositional logic has a worst-case complexity probably exponential in the size of the input.

Quid of FOL ?

Decision problems

- A **decision problem** is any problem that, given a certain input, asks a question to be answered with “yes” or “no”.
 - Entailment checking is an example of a decision problem.
- A problem is **decidable** if it can always receive a correct “yes” or “no” answer for every instance of the input.
- In formal logic we have several important decision problems :
 - Validity problem: *Is F valid ?*
 - Satisfiability problem: *Is F satisfiable ?*
 - Entailment or logical consequence problem: *Is G a consequence of F ?*
 - Equivalence problem: *Are F and G equivalent ?*
- Are these questions decidable ?

Decidability of propositional logic

□ We have seen that:

- | | | |
|----------------------|-----|---|
| ■ $F \models G$ | iff | $\models (G \wedge \neg F) : F$ entails G iff $G \wedge \neg F$ is not satisfiable. |
| ■ $\models F$ | iff | $\models \neg F : F$ is valid iff $\neg F$ is unsatisfiable. |
| ■ F is satisfiable | iff | $\neg F$ is not valid. |
| ■ $F \equiv G$ | iff | $F \models G$ and $G \models F$. |

Any algorithm that works for one of our decision problems will work for all !

□ Are the above problems in propositional logic decidable ?

□ Yes ! For PL, truth table enumeration is guaranteed to terminate in finite time.

=> **Propositional logic is decidable**, and its worst-case time complexity is $O(2^n)$.

- PL decision algorithms, called **SAT solvers**, have focused on the SATisfiability problem.
- In practice, truth-table enumeration is not used. The best algorithms use backtracking search.

Fundamental complexity is unchanged, but optimizations allow to handle tens of millions of variables.

Undecidability of first order logic

- We consider the same decision problems :
 - Validity problem: is F valid ?
 - Satisfiability problem: is F satisfiable ?
 - Entailment or logical consequence problem: is G a consequence of F ?
 - Equivalence problem: are F and G equivalent ?

- Alan Turing (1936) and Alonzo Church (1936) both proved independently that the questions of **validity and entailment for first-order logic are semi-decidable**.
 - Given the question $F \models G$?, algorithms exist that say yes to every entailed sentence, but they are not guaranteed to terminate in the case of non-entailment. No algorithm will always answer by yes or no.

- The question of **satisfiability** is not even semi-decidable : it is **undecidable**.
 - If it was semi-decidable, we could answer on the satisfiability of $\neg F$ and hence on the non validity of F .

Practical approaches for reasoning with first order logic

□ Two important approaches have focused on specific subsets of first order logic :

1. Definite clauses => Prolog.

- Definite clauses : disjunctions of literals of which exactly one is positive.

$\neg P(X, Y) \vee \neg Q(Y, Z) \vee R(X, Z)$, which is equivalent to $P(X, Y) \wedge Q(Y, Z) \rightarrow R(X, Z)$

Definite clauses in FOL are still undecidable but lend themselves to fast implementations.

2. Description Logics => Semantic web, OWL.

- Description logics work with decidable fragments of first order logic.
- They are the theoretical basis for this course.

Summary

- ❑ Proposition logic is decidable and has very useful and efficient reasoning algorithms for specialized tasks but is not expressive enough for most knowledge-based systems.
- ❑ First order logic is a powerful knowledge representation language, the standard to which other languages are compared. It presents however two drawbacks :
 - FOL is **too expressive**, to the extent that FOL is not decidable.
 - FOL is not **structured** for knowledge representation : it lacks the syntactic tools to organize knowledge, identify concept definitions, make visible the taxonomic hierarchy...
- ❑ Some restricted fragments of first order logic are decidable and support efficient reasoners; they will offer a solution for our purpose.

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THANK YOU