# **Structural Description Logics**

(Acknowledgement: Enrico Franconi)

- Description Logics
- The need for a formalism
- \* Ambiguities of object-oriented representations
- A structure upon FOL
  - \* A predicate level language
- Examples from object-oriented programming
- $\mathcal{FL}^-$ : The simplest structural description logic
- Syntax and Semantics
- Reasoning problems and reasoning procedures

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# **Applications**

Description logics based systems are used in many applications:

- Conceptual modeling
- Terminologies and formal ontologies
- Query optimization and view maintenance
- Natural language semantics representation
- Information access and "intelligent" interfaces
- "Intelligent integration of information"
- Configuration
- Formal specifications in engineering
- Software management
- Planning
- . . .

# **Description Logics**

- A logical reconstruction and unifying formalism for (object-oriented) representation tools as
  - Semantic networks
  - Frame-based systems
  - Object-oriented representations
  - Semantic data models
  - Type systems
  - Feature Logics
- A structured fragment of predicate logic
- Provide theories and systems for *expressing* structured information and for *accessing* and *reasoning* with it in a principled way.

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# **Object-Oriented Representations**

- Identifier
- Class (vs. prototype in frame systems)
- Instance
- Attribute ("slot")
  - Value:
  - Identifier; default
  - Value restriction:

Type; concrete domain; cardinality; attached method

# **Ambiguities: Classes and Instances**

**Ambiguities: Classes and Instances** 

Person: AGE: Number

SEX: M, F HEIGHT: Number WIFE: Person

john: AGE: 29

SEX: M HEIGHT: 76 WIFE: mary 29er: AGE: 29

SEX: M

HEIGHT: Number
WIFE: Person

john: AGE: 29

SEX: M

HEIGHT: Number WIFE: Person

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**Ambiguities: IS-A** 

Person: AGE: Number

SEX: M, F HEIGHT: Number WIFE: Person

has subclass

Male: AGE: Number

SEX: M HEIGHT: Number WIFE: Female **Ambiguities: IS-A** 

Male: AGE: Number

SEX: M

HEIGHT: Number WIFE: Female

has instance

john: AGE: 35

SEX: M HEIGHT: 76

WIFE: mary

# **Ambiguities: IS-A**

29er: AGE: 29

SEX: M

HEIGHT: Number WIFE: Person

#### has instance

john: AGE: 29

SEX: M HEIGHT: 76 WIFE: mary

# **Ambiguities: Relations**

## Implicit relation:

john: AGE: 35 SEX: M

HEIGHT: 76

<u>WIFE</u>: mary

mary: AGE: 32

SEX: F

HEIGHT: 59 HUSBAND: john

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# **Ambiguities: Relations**

## **Explicit relation:**

john: AGE: 35

SEX: M HEIGHT: 76

mary: AGE: 32

SEX: F

HEIGHT: 59

m-j-family: <u>WIFE</u>: mary

<u>HUSBAND</u>: john

# **Ambiguities: Relations**

# **Special relation:** (⇒ Mereology)

$$\operatorname{Car} \xrightarrow{\operatorname{HAS-PART}} \operatorname{Engine}$$

$$\texttt{Engine} \xrightarrow{\texttt{HAS-PART}} \texttt{Valve}$$

$$\mathtt{Car} \xrightarrow{\mathtt{HAS}- \mathtt{PART}} \mathtt{Valve}$$

## Normal relation:

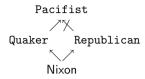
$$\mathsf{John} \xrightarrow{\mathtt{HAS-CHILD}} \mathsf{Ronald}$$

$$\mathsf{Ronald} \xrightarrow{\mathtt{HAS-CHILD}} \mathsf{Bill}$$

$$\begin{array}{c} & \xrightarrow{} \\ \text{John} \xrightarrow{\text{HAS-CHILD}} & \text{Bill} \end{array}$$

# **Ambiguities: Defaults**

The "Nixon diamond"



Quakers are pacifists, Republicans are not.

 $\Rightarrow$  Is Nixon a pacifist?

Remark: Consider depth-first search vs. breadth-first search!

. . . Sceptical inference

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#### **False Friends**

- The meaning of object-oriented representations is logically very ambiguous.
- The appeal of the graphical nature of object-oriented representation tools has led to forms of reasoning that do not fall into standard logical categories, and are not yet well understood.
- Unfortunately, it is much easier to develop some algorithm that appears to reason over structures of a certain kind, than to *justify* its reasoning by explaining what the structures are saying about the domain.

# **Ambiguities: Quantification**

What is the exact meaning of Frog  $\xrightarrow{\text{HAS-COLOR}}$  green ?

- Every frog is just green
- Every frog is also green
- Every frog is of some green
- There is a frog, which is just green
- . . .
- Frogs are typically green, but there may be exceptions

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# A Structured Logic

- Any (basic) Description Logic is a fragment of FOL.
- The representation is done at the *predicate level*: there are no variables in the formalism.
- A Description Logic theory consists of two parts:
  - definitions of predicates ("TBox")
  - assertions over constants ("ABox")
- Any (basic) Description Logic is a subset of  $\mathcal{L}_3$ , i.e. the function-free FOL using only at most three variable names.

# Why not FOL?

If FOL is directly used without any restrictions then

- the structure of the knowledge is destroyed, and it can not be exploited for driving the inference;
- the expressive power is too high for obtaining decidable and efficient inference problems:
- the inference power may be too low for expressing interesting, but still decidable theories.

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# Types of the TBox Language

```
    Concepts — denote entities
(unary predicates, classes)
```

Example: Student, Married  $\{x \mid \mathtt{Student}(x)\}, \{x \mid \mathtt{Married}(x)\}$ 

 Roles — denote properties (binary predicates, relations)

Example: FRIEND, LOVES  $\{\langle x,y \rangle \mid \text{FRIEND}(x,y)\}, \{\langle x,y \rangle \mid \text{LOVES}(x,y)\};$ 

## Structured Inheritance Networks: KL-ONE

- Structured Descriptions
  - corresponding to the complex relational structure of objects,
  - built using a restricted set of epistemologically adequate constructs;
- Distinction between conceptual (terminological, T-Box) and instance (assertional, A-Box) knowledge;
- Strict reasoning, no defaults.

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# **Concept Expressions**

Description Logics organize the information in classes — *concepts* — gathering homogeneous data, according to the relevant common properties among a collection of instances.

Example:

Student □ ∃FRIEND.Married

 $\{x \mid \mathtt{Student}(x) \land \bigvee_{y} \mathtt{.FRIEND}(x,y) \land \mathtt{Married}(y).\}$ 

## $\lambda$ to Define Predicates

Predicates are special case of functions: truth functions.

 $\lambda x.P(x)$  denotes a function from domain individuals to truth values.

E.g., we can write for the *unary* predicate Person:

$$Person \doteq \lambda x.Person(x)$$

which is equivalent to say that Person denotes the set of persons:

$$\begin{array}{c} \mathtt{Person} \Longrightarrow \{x \mid \mathtt{Person}(x)\} \\ \mathtt{Person}^{\mathcal{I}} = \{x \mid \mathtt{Person}(x)\} \\ \mathtt{Person}(\mathsf{john}) \quad \mathsf{IFF} \quad \mathsf{john}^{\mathcal{I}} \in \mathtt{Person}^{\mathcal{I}} \end{array}$$

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# **Concept Expressions**

```
\begin{split} &(\mathtt{Student} \sqcap \exists \mathtt{FRIEND.Married})^{\mathcal{I}} \\ &= \\ &(\mathtt{Student})^{\mathcal{I}} \cap (\exists \mathtt{FRIEND.Married})^{\mathcal{I}} \\ &= \\ &\{x \mid \mathtt{Student}(x)\} \cap \{x \mid \bigvee_{y} (\mathtt{FRIEND}(x,y) \wedge \mathtt{Married}(y))\} \\ &= \\ &\{x \mid \mathtt{Student}(x) \wedge \bigvee_{y} (\mathtt{FRIEND}(x,y) \wedge \mathtt{Married}(y))\} \end{split}
```

In the same way for the binary predicate FRIEND:

$$\begin{aligned} & \mathtt{FRIEND} \doteq \lambda x, y.\mathtt{FRIEND}(x,y) \\ & \mathtt{FRIEND}^{\mathcal{I}} = \{ \langle x,y \rangle \mid \mathtt{FRIEND}(x,y) \} \end{aligned}$$

The functions we are defining with the  $\lambda$  operator may be parametric:

$$\texttt{Student} \sqcap \texttt{Worker} = \lambda x. (\texttt{Student}(x) \land \texttt{Worker}(x))$$
 
$$(\texttt{Student} \sqcap \texttt{Worker})^{\mathcal{I}} = \{x \mid (\texttt{Student}(x) \land \texttt{Worker}(x))\}$$
 
$$(\texttt{Student} \sqcap \texttt{Worker})^{\mathcal{I}} = \texttt{Student}^{\mathcal{I}} \cap \texttt{Worker}^{\mathcal{I}}$$
 (Verify as an exercise)

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## **Objects: Classes**

Student

Person	
name:	[String]
address:	[String]
enrolled:	[Course]

```
 \begin{cases} x \mid \mathsf{Student}(x) \rbrace = \\ \{x \mid \mathsf{Person}(x) \land \\ & \bigvee_{y} (\mathsf{NAME}(x,y) \land \mathsf{String}(y)) \land \\ & \bigvee_{z} (\mathsf{ADDRESS}(x,z) \land \mathsf{String}(z)) \land \\ & \bigvee_{w} (\mathsf{ENROLLED}(x,w) \land \mathsf{Course}(w)) \} \end{cases}   \mathsf{Student} \doteq \mathsf{Person} \ \sqcap \\ \exists \mathsf{NAME.String} \ \sqcap \\ \exists \mathsf{ADDRESS.String} \ \sqcap \\ \exists \mathsf{ENROLLED.Course}
```

# **Objects: Instances**

# s1: Student name: "John" address: "Abbey Road" enrolled: cs415

$$\begin{split} \text{Student}(\texttt{s1}) &= \text{Person}(\texttt{s1}) \land \\ &\quad \text{NAME}(\texttt{s1}, \texttt{"john"}) \land \\ &\quad \text{String}(\texttt{"john"}) \land \\ &\quad \text{ADDRESS}(\texttt{s1}, \texttt{"AbbeyRoad"}) \land \\ &\quad \text{String}(\texttt{"AbbeyRoad"}) \land \\ &\quad \text{ENROLLED}(\texttt{s1}, \texttt{cs415}) \land \\ &\quad \text{Course}(\texttt{cs415}) \end{split}$$

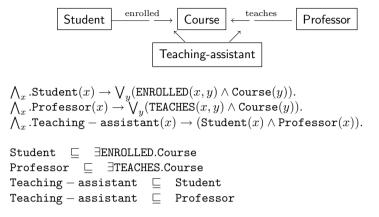
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# Quantification

 $\texttt{Frog} \xrightarrow{\texttt{HAS}-\texttt{COLOR}} \texttt{Green}$ 

- Frog ☐ ∃HAS COLOR.Green: Every frog is also green
- Frog  $\sqsubseteq$   $\forall$ HAS COLOR.Green Frog(x), HAS COLOR(x,y): There is a frog, which is just green

## **Semantic Networks**



Frog 
$$\xrightarrow{\text{HAS}-\text{COLOR}}$$
 Green

Every frog is also green

Frog  $\sqsubseteq$   $\exists \text{HAS}-\text{COLOR}.\text{Green}$ 
 $\bigwedge_x.\text{Frog}(x) \to \bigvee_y(\text{HAS}-\text{COLOR}(x,y) \land \text{Green}(y)).$ 

Exercise: Is this a model?

Frog(oscar), Green(green), HAS  $-\text{COLOR}(\text{oscar},\text{green}),$  Red(red), HAS  $-\text{COLOR}(\text{oscar},\text{green}).$ 

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```
Frog \xrightarrow{\text{HAS}-\text{COLOR}} Green

Every frog is only green

Frog \sqsubseteq \forall \text{HAS}-\text{COLOR}.\text{Green}
\bigwedge_x.\text{Frog}(x) \to \bigwedge_y(\text{HAS}-\text{COLOR}(x,y) \to \text{Green}(y)).

Exercise: Is this a model?

Frog(oscar), Green(green),
HAS - COLOR(oscar,green),
Red(red),
HAS - COLOR(oscar,red).

... and this?

Frog(sing), AGENT(sing,oscar).
```

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## $\mathcal{FL}^-$

- For the rest of this chapter, we will focus to the simplest conceivable structural description logic:  $\mathcal{FL}^-$
- We will regard  $\mathcal{FL}^-$  as a logical language:
  - Syntax
  - Semantics
  - Reasoning problems: decidability and complexity
- Reasoning procedures soundness, completeness, and asymptotic complexity

# **Analytic reasoning**

## The Grammar

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```
\begin{array}{lll} C,D & \to A \mid C \sqcap D \mid \forall R.C \mid \exists R \\ \\ A \in \textit{atomic-concept} \\ R \in \textit{atomic-role} \\ \\ C,D \in \textit{concept} \\ \\ & \textit{concept} \mid \neg \left\langle \textit{concept} \right\rangle \mid \\ & \left\langle \textit{concept} \right\rangle \sqcap \left\langle \textit{concept} \right\rangle \\ & \exists \left\langle \textit{atomic-role} \right\rangle . \left\langle \textit{concept} \right\rangle \\ & \forall \left\langle \textit{atomic-role} \right\rangle . \left\langle \textit{concept} \right\rangle \end{array}
```

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## Intuitive Model-theoretic Semantics

Intuitively (as we already know):

- Concepts represent classes, i.e., sets of individuals.
- Roles represent relations between pairs of individuals.
- Atomic concepts are the names of primitive (undefined) concepts.
- □-constructions represent conjoined concepts, so, e.g.,
   Adult □ Male □ Person would represent the concept of something that is at the same time an adult, a male, and a person.
- This allows us to combine several properties (i.e. super-concepts or attribute restrictions) in the definition of a concept.

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#### Formal Semantics

An interpretation  $\mathcal{I} = (\Delta, \cdot^{\mathcal{I}})$  consists of:

- a nonempty set  $\Delta$  (the *domain*)
- $\bullet$  a function  $\cdot^{\mathcal{I}}$  (the interpretation function) that maps
  - every *concept* to a subset of  $\Delta$
  - every *role* to a subset of  $\Delta \times \Delta$
  - every  $\mathit{individual}$  to an element of  $\Delta$

## Quantifiers

- The  $\forall$ -construct provides a concept restriction on the values of an attribute (x is an  $\forall$ R.C if and only if each R of x is a C). Thus,  $\forall$ CHILD.Doctor corresponds to the concept of something all of whose children are doctors. It is a way to restrict the value of a slot at a frame.
- The  $\exists$ -operator guarantees that there will be at least one value for the attribute named. x is a  $\exists R$  if and only if x has at least one R. E.g., Person  $\sqcap \exists CHILD$  would represent the concept of a parent. This is a way to *introduce* a slot at a frame.

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# **Extension of Concepts**

An interpretation function  $\cdot^{\mathcal{I}}$  is an extension function iff:

$$\begin{array}{rcl} (C \sqcap D)^{\mathcal{I}} & = & C^{\mathcal{I}} \cap D^{\mathcal{I}} \\ (\forall R.C)^{\mathcal{I}} & = & \{x \in \Delta \mid \bigwedge_y . (x,y) \in R^{\mathcal{I}} \rightarrow y \in C^{\mathcal{I}}\}. \\ (\exists R)^{\mathcal{I}} & = & \{x \in \Delta \mid \bigvee_y (x,y) \in R^{\mathcal{I}}\} \end{array}$$

Recall that  $C^{\mathcal{I}}$  is the set of all individuals in the extension of C: so,  $x \in C^{\mathcal{I}}$  has the same truth value as C(x).

Analogously,  $(x,y) \in R^{\mathcal{I}}$  is the same as R(x,y).

## The Subsumption Problem

 $\mathtt{C} \sqsubseteq \mathtt{D}, \ \mathtt{C} \ \textit{is subsumed} \ \ \mathtt{by} \ \mathtt{D} \ \ \mathsf{iff}$  for any domain  $\Delta$  and any extension function  $\cdot^{\mathcal{I}}$  over  $\Delta$ :  $\mathtt{C}^{\mathcal{I}} \subseteq \mathtt{D}^{\mathcal{I}}$ , i.e.,  $\bigwedge_x . C(x) \to D(x)$ .

## Simple examples:

- Adult  $\sqcap$  Male  $\sqsubseteq$  Adult
- Adult  $\sqcap$  Male  $\sqcap$  Rich  $\sqsubseteq$  Adult  $\sqcap$  Male
- $\forall$ CHILD.Adult  $\sqcap$  Male  $\sqsubseteq$   $\forall$ CHILD.Adult
- $\forall$ CHILD.Adult  $\sqcap \exists$ CHILD  $\sqsubseteq \forall$ CHILD.Adult
- $\forall$ CHILD.Adult  $\not\sqsubseteq \exists$ CHILD
- ∃CHILD ⊄ ∀CHILD.Adult

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#### **Normal Form**

The algorithm works in two phases: first, concepts are rewritten into a normal form, then their structures are compared.

#### Normal Form:

- 1. All nested conjunctions are flattened, i.e.  $A \sqcap (B \sqcap C) \Longrightarrow A \sqcap B \sqcap C$ .
- 2. All conjunctions of universal quantifications are factorized, i.e.  $\forall R.C \sqcap \forall R.D \Longrightarrow \forall R.(C \sqcap D).$

The rewritten concepts are logically equivalent to the previous ones, hence subsumption is preserved by this transformation.

(Proof: exercise)

# **Computational Properties of Subsumption**

Subsumption for  $\mathcal{FL}^-$  has the following computational properties (constructive proofs sketched):

#### decidable and in P

#### The structural subsumption algorithm

- The algorithm for computing subsumption is based on structural comparisons between concept expressions.
- The central idea of structural comparison is that if the two concept expressions to be compared are made of subexpressions, one can compare separately one subexpression of a concept with all subexpressions of the other concept.

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# The Core Algorithm: SUBS? [C, D]

Let  $C = C_1 \sqcap \ldots \sqcap C_n$  and  $D = D_1 \sqcap \ldots \sqcap D_n$  (in normal form).

Then SUBS?[C, D] returns TRUE if and only if for all  $C_i$ :

- 1. if  $C_i$  is either an atomic concept, or is a concept of the form  $\exists R$ , then there exists a  $D_i$  such that  $C_i = D_i$ ;
- 2. if  $C_i$  is a concept of the form  $\forall R.C'$ , then there exists a  $D_j$  of the form  $\forall R.D'$  (same atomic role R) such that SUBS?[C',D'].

**Exercise**: Check the examples given above with the structural algorithm.

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# **Asymptotic Complexity**

- By induction on the nesting of  $\forall$ -quantifiers, one can prove that the complexity of the structural algorithm is  $O(|C| \times |D|)$  (i.e., quadratic in the length of the longest argument).
- If the subexpressions of each concept are ordered (e.g. lexicographically), it can be shown that the complexity is only linear, so the dominant factor becomes the complexity of ordering concepts.

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#### Informal Proof

- Suppose that SUBS?[C, D] returns TRUE and consider one of the conjuncts of C call it  $C_i$ .
- Either  $C_i$  is among the  $D_i$ , or it is of the form  $\forall R.C'$
- In the latter case, there is a  $\forall R.D'$  among the  $D_j$  , where SUBS?[C',D'].
- Then, by induction, any extension of D' must be a subset of C', and so any extension of  $D_j$  must be a subset of  $C_i$ 's.
- So, no matter what  $C_i$  is, the extension of D which is the conjunction of all the  $D_i$ 's must be a subset of  $C_i$ .

### Soundness

Remember: Whenever a **sound** reasoning procedure claims to have found a solution for a given instance of the problem, then this is actually a solution.

The structural subsumption algorithm is sound, since, when it says that a concept C subsumes a concept D — i.e., SUBS? [C,D] returns TRUE — then it holds that  $\mathbf{D}^{\mathcal{I}} \subseteq \mathbf{C}^{\mathcal{I}}$  for all interpretations.

#### Observation:

The part of the algorithm computing the normal form does not change the extension of the concepts for any interpretation; thus it does not affect the soundness (and the completeness) of the algorithm.

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- Since this is true for every  $C_i$ , the extension of D must also be a subset of the extension of C which is the intersection of all the extensions of  $C_i$ .
- $\bullet$  So, whenever SUBS? [C,D] returns TRUE, C subsumes D, i.e.,  $\mathbf{D}^{\mathcal{I}}\subseteq\mathbf{C}^{\mathcal{I}}$

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## **Completeness**

Remember: Whenever an instance of the problem has a solution, a **complete** reasoning procedure computes the solution for that instance.

The structural subsumption algorithm is complete, since, whenever it holds that  $D^{\mathcal{I}} \subseteq C^{\mathcal{I}}$  for all interpretations, then the algorithm says that C subsumes D.

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# Structural Algorithms or Normalize-and-Compare

What if we enrich the expressivity?

- CLASSIC
- BACK
- LOOM

These systems have incomplete algorithms, due to the interactions between constructors, which cannot be taken into account by structural algorithms which are based on syntactical comparisons between subexpressions of the concepts.

Example:  $A \sqcup \neg A$  subsumes every concept, even if such a concept does not mention at all the atomic concepts A in its definition.

**Proof Idea** 

- ullet The proof is done by showing that anytime SUBS?[C,D] returns FALSE, there exists an interpretation assigning an element to D but not to C, i.e., in that interpretation the extension of C is not a superset of the extension of D.
- This shows a counter-example, i.e., anytime SUBS?[C,D] returns FALSE it is not true that  $\mathbf{D}^{\mathcal{I}} \subseteq \mathbf{C}^{\mathcal{I}}$  for that particular interpretation, and so  $D \not \sqsubseteq C$ .
- The proof relies on the fact that anytime SUBS?[C, D] returns FALSE
  it is possible to find a conjunct C<sub>i</sub> of C which has no corresp. conjunct
  in D.
- It is shown that in this case there exists an interpretation which assigns an object to any primitive concept, but not to the factorized one  $C_i$ .
- Thus, it is not possible that  $D^{\mathcal{I}} \subset C^{\mathcal{I}}$ .

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