Description Logics for Conceptual Data Modeling in UML

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Description Logics
What are Description Logics?

In modeling an application domain we typically need to represent a situation in terms of
- objects
- classes
- relations (or associations)
and to reason about the representation

Description Logics are logics specifically designed to represent and reason on
- objects
- classes – called concepts in DLs
- (binary) relations – called roles in DLs

Origins of Description Logics

Knowledge Representation is a subfield of Artificial Intelligence

Early days KR formalisms (late '70s, early '80s):
- Semantic Networks: graph-based formalism, used to represent the meaning of sentences
- Frame Systems: frames used to represent prototypical situations, antecedents of object-oriented formalisms

Problems: no clear semantics, reasoning not well understood

Description Logics (a.k.a. Concept Languages, Terminological Languages) developed starting in the mid '80s, with the aim of providing semantics and inference techniques to knowledge representation systems
Current applications of DLs

DLs have evolved from being used “just” in KR

Found applications in:

- Databases:
  - schema design, schema evolution
  - query optimization
  - integration of heterogeneous data sources, data warehousing
- Conceptual modeling
- Foundation for the semantic web
- 

Ingredients of a DL

A Description Logic is characterized by:

1. A description language: how to form concepts and roles
   \[ \text{Human} \sqcap \text{Male} \sqcap (\exists \text{child}) \sqcap \forall \text{child}.(\text{Doctor} \sqcup \text{Lawyer}) \]

2. A mechanism to specify knowledge about concepts and roles (i.e., a TBox)
   \[ ggK = \{ \text{Father} \equiv \text{Human} \sqcap \text{Male} \sqcap (\exists \text{child}), \]
   \[ \text{HappyFather} \sqsubseteq \text{Father} \sqcap \forall \text{child}.(\text{Doctor} \sqcup \text{Lawyer}) \} \]

3. A mechanism to specify properties of objects (i.e., an ABox)
   \[ A = \{ \text{HappyFather}(\text{JOHN}), \text{child}(\text{JOHN}, \text{MARY}) \} \]

4. A set of inference services: how to reason on a given knowledge base
   \[ K \models \text{HappyFather} \sqsubseteq \exists \text{child}.(\text{Doctor} \sqcup \text{Lawyer}) \]
   \[ K \cup A \models (\text{Doctor} \sqcup \text{Lawyer})(\text{MARY}) \]

Note: we will consider ABoxes only later, when needed; hence, for now, we consider a knowledge base to be simply a TBox
Architecture of a DL system

Knowledge Base

Terminological knowledge (TBox)

\[
\text{Father} \equiv \text{Human} \sqcap \text{Male} \sqcap (\exists \text{child}) \\
\text{HappyFather} \sqsubseteq \text{Father} \sqcap (\forall \text{child}.(\text{Doctor} \sqcup \text{Lawyer}))
\]

Knowledge about objects (ABox)

\[
\text{HappyFather}(\text{JOHN}) \\
\text{child}(\text{JOHN}, \text{MARY})
\]

Inference Engine

Applications

Description language

A description language is characterized by a set of constructs for building complex concepts and roles starting from atomic ones:

- **concepts** represent classes: interpreted as sets of objects
- **roles** represent relations: interpreted as binary relations on objects

Semantics: in terms of interpretations \( \mathcal{I} = (\Delta^\mathcal{I}, \cdot^\mathcal{I}) \), where

- \( \Delta^\mathcal{I} \) is the interpretation domain
- \( \cdot^\mathcal{I} \) is the interpretation function, which maps
  - each atomic concept \( A \) to a subset \( A^\mathcal{I} \) of \( \Delta^\mathcal{I} \)
  - each atomic role \( P \) to a subset \( P^\mathcal{I} \) of \( \Delta^\mathcal{I} \times \Delta^\mathcal{I} \)

The interpretation function is extended to complex concepts and roles according to their syntactic structure
Syntax and semantics of $\mathcal{AL}$

$\mathcal{AL}$ is the basic language in the family of $\mathcal{AL}$ languages

<table>
<thead>
<tr>
<th>Construct</th>
<th>Syntax</th>
<th>Example</th>
<th>Semantics</th>
</tr>
</thead>
<tbody>
<tr>
<td>atomic concept</td>
<td>$A$</td>
<td>Doctor</td>
<td>$A^I \subseteq \Delta^I$</td>
</tr>
<tr>
<td>atomic role</td>
<td>$P$</td>
<td>child</td>
<td>$P^I \subseteq \Delta^I \times \Delta^I$</td>
</tr>
<tr>
<td>atomic negation</td>
<td>$\neg A$</td>
<td>$\neg$Doctor</td>
<td>$\Delta^I \setminus A^I$</td>
</tr>
<tr>
<td>conjunction</td>
<td>$C \cap D$</td>
<td>Hum $\cap$ Male</td>
<td>$C^I \cap D^I$</td>
</tr>
<tr>
<td>(unqual.) exist. res.</td>
<td>$\exists R$</td>
<td>$\exists$child</td>
<td>${a \mid \exists b. (a, b) \in R^I}$</td>
</tr>
<tr>
<td>value restriction</td>
<td>$\forall R.C$</td>
<td>$\forall$child.$,$Male</td>
<td>${a \mid \forall b. (a, b) \in R^I \supset b \in C^I}$</td>
</tr>
</tbody>
</table>

($C$, $D$ denote arbitrary concepts and $R$ an arbitrary role)

Note: $\mathcal{AL}$ is not propositionally closed (no full negation)

The $\mathcal{AL}$ family

Typically, additional constructs w.r.t. those of $\mathcal{AL}$ are needed:

<table>
<thead>
<tr>
<th>Construct</th>
<th>$\mathcal{AL}$·</th>
<th>Syntax</th>
<th>Semantics</th>
</tr>
</thead>
<tbody>
<tr>
<td>disjunction</td>
<td>$\cup$</td>
<td>$C \cup D$</td>
<td>$C^I \cup D^I$</td>
</tr>
<tr>
<td>qual. exist. res.</td>
<td>$E$</td>
<td>$\exists R.C$</td>
<td>${a \mid \exists b. (a, b) \in R^I \land b \in C^I}$</td>
</tr>
<tr>
<td>(full) negation</td>
<td>$C$</td>
<td>$\neg C$</td>
<td>$\Delta^I \setminus C^I$</td>
</tr>
<tr>
<td>number</td>
<td>$N'$</td>
<td>$(\geq k R)$</td>
<td>${a \mid #{b \mid (a, b) \in R^I} \geq k}$</td>
</tr>
<tr>
<td>restrictions</td>
<td>$N$</td>
<td>$(\leq k R)$</td>
<td>${a \mid #{b \mid (a, b) \in R^I} \leq k}$</td>
</tr>
<tr>
<td>qual. number</td>
<td>$Q$</td>
<td>$(\geq k R.C)$</td>
<td>${a \mid #{b \mid (a, b) \in R^I \land b \in C^I} \geq k}$</td>
</tr>
<tr>
<td>restrictions</td>
<td>$Q$</td>
<td>$(\leq k R.C)$</td>
<td>${a \mid #{b \mid (a, b) \in R^I \land b \in C^I} \leq k}$</td>
</tr>
<tr>
<td>inverse role</td>
<td>$\mathcal{I}$</td>
<td>$P^-$</td>
<td>${(a, b) \mid (b, a) \in P^I}$</td>
</tr>
</tbody>
</table>

We also use: $\bot$ for $A \cap \neg A$ (hence $\bot^I = \emptyset$)

$\top$ for $A \cup \neg A$ (hence $\top^I = \Delta^I$)
The $\mathcal{AL}$ family – Examples

- Disjunction
  $$\forall \text{child}. (\text{Doctor} \sqcup \text{Lawyer})$$

- Qualified existential restriction
  $$\exists \text{child}. \text{Doctor}$$

- Full negation
  $$\neg (\text{Doctor} \sqcup \text{Lawyer})$$

- Number restrictions
  $$\geq 2 \text{ child} \sqcap \leq 1 \text{ sibling}$$

- Qualified number restrictions
  $$\geq 2 \text{ child. Doctor} \sqcap \leq 1 \text{ sibling. Male}$$

- Inverse role
  $$\forall \text{child}^{-}. \text{Doctor}$$

Reasoning on concept expressions

An interpretation $\mathcal{I}$ is a model of a concept $C$ if $C^\mathcal{I} \neq \emptyset$

Basic reasoning tasks:

1. Concept satisfiability: does $C$ admit a model?
2. Concept subsumption: does $C^\mathcal{I} \subseteq D^\mathcal{I}$ hold for all interpretations $\mathcal{I}$? (written $C \subseteq D$)

Subsumption used to build the concept hierarchy:

(1) and (2) are mutually reducible if DL is propositionally closed
Reasoning on concept expressions – Technique

Techniques are based on **tableaux algorithms**: for satisfiability of $C_0$

1. Aims at building a tree representing a model of $C_0$
   - nodes represent objects of $\Delta^I$, labeled with subconcepts of $C_0$
   - edges represent role successorship between objects
2. Concepts are first put in negation normal form (negation is pushed inside)
3. Tree initialized with single root node, labeled with $\{C_0\}$
4. Rules (one for each construct) add new nodes or concepts to the label
   - deterministic rules: for $\cap$, $\forall P.C$, $\exists P.C$, $(\geq kP)$
   - non-deterministic rules: for $\cup$, $(\leq kP)$
5. Stops when:
   - no more rule can be applied, or
   - a clash (obvious contradiction) is detected

Reasoning on concept expressions – Technique (Cont’d)

Properties of tableaux algorithms (must be proved for the various cases):

1. **Termination**: since quantifier depth decreases going down the tree
2. **Soundness**: if there is a way of terminating without a clash, then $C_0$ is satisfiable
   - construct from the tree a model of $C_0$
3. **Completeness**: if $C_0$ is satisfiable, there is a way of applying the rules so that the algorithm terminates without a clash
   - if $I$ is a model of $T$, then there is a rule s.t. $I$ is also a model of the tree obtained by applying the rule to $T$

Tableaux algorithms provide **optimal decision procedures** for concept satisfiability (and subsumption), but not for Knowledge Base reasoning (see later).
Reasoning on concept expressions – Technique (Cont’d)

For gentle introduction on how tableaux for $\mathcal{ALC}$ work without Knowledge base please refer to Ian Horrocks & Uli Sattler’s slides:

http://www.cs.man.ac.uk/~horrocks/Slides/IJCAR-tutorial/Print/p2-algorithmsI.pdf

Reasoning on concept expressions – Complexity

Complexity of concept satisfiability

<table>
<thead>
<tr>
<th>Complexity</th>
<th>Logic</th>
</tr>
</thead>
<tbody>
<tr>
<td>PTIME</td>
<td>$\mathcal{AL}, \mathcal{ALN}$</td>
</tr>
<tr>
<td>NP-complete</td>
<td>$\mathcal{ALU}, \mathcal{ALUN}$</td>
</tr>
<tr>
<td>coNP-complete</td>
<td>$\mathcal{ALE}$</td>
</tr>
<tr>
<td>PSPACE-complete</td>
<td>$\mathcal{ALC}, \mathcal{ALCN}, \mathcal{ALCT}, \mathcal{ALCQT}$</td>
</tr>
</tbody>
</table>

Observations:

- two sources of complexity
  - union ($\cup$) of type NP
  - existential quantification ($\exists$) of type coNP
  When they are combined, the complexity jumps to PSPACE
- number restrictions ($\forall$) do not add to the complexity
Structural properties vs. asserted properties

We have seen how to build complex concept expressions, which allow to denote classes with a complex structure.

However, in order to represent complex domains one needs the ability to assert properties of classes and relationships between them (e.g., as done in UML class diagrams).

The assertion of properties is done in DLs by means of knowledge bases.

### DL knowledge bases

A DL knowledge base consists of a set of inclusion assertions on concepts:

\[ C \sqsubseteq D \]

- when \( C \) is an atomic concept, the assertion is called **primitive**
- \( C \equiv D \) is an abbreviation for \( C \sqsubseteq D, D \sqsubseteq C \)

Example:

\[ \mathcal{K} = \{ \text{Father} \equiv \text{Human} \sqcap \text{Male} \sqcap (\exists \text{child}), \]
\[ \text{HappyFather} \sqsubseteq \text{Father} \sqcap \forall \text{child}.(\text{Doctor} \sqcup \text{Lawyer}) \} \]

Semantics: An interpretation \( \mathcal{I} \) is a model of a knowledge base \( \mathcal{K} \) if

\[ C^\mathcal{I} \sqsubseteq D^\mathcal{I} \quad \text{for every assertion} \ C \sqsubseteq D \text{ in } \mathcal{K} \]
Reasoning on DL knowledge bases

Basic reasoning tasks:

1. Knowledge base satisfiability
   Given $\mathcal{K}$, does it admit a model?

2. Concept satisfiability w.r.t. a KB — denoted $\mathcal{K} \not\models C \equiv \bot$
   Given $C$ and $\mathcal{K}$, do they admit a common model?

3. Logical implication — denoted $\mathcal{K} \models C \sqsubseteq D$
   Given $C$, $D$, and $\mathcal{K}$, does $C^I \subseteq D^I$ hold for all models $I$ of $\mathcal{K}$?

Again, logical implication allows for classifying the concepts in the KB w.r.t. the knowledge expressed by the KB.

Relationship among reasoning tasks

The reasoning tasks are mutually reducible to each other, provided the description language is propositionally closed:

1. to 3
   $\mathcal{K}$ satisfiable iff not $\mathcal{K} \models T \sqsubseteq \bot$ iff $\mathcal{K} \not\models T \equiv \bot$
   (i.e., $T$ satisfiable w.r.t. $\mathcal{K}$)

2. to 3
   $\mathcal{K} \models C \sqsubseteq D$ iff not $\mathcal{K} \not\models C \cap \neg D \equiv \bot$
   (i.e., $C \cap \neg D$ unsatisfiable w.r.t. $\mathcal{K}$)

3. to 1
   $\mathcal{K} \not\models C \equiv \bot$ iff $\mathcal{K} \cup \{ T \sqsubseteq \exists P_{new} \cap \forall P_{new} \cdot C \}$ satisfiable
   (where $P_{new}$ is a new atomic role)
Tableaux procedures for reasoning with DLs knowledge bases

For details on tableaux based reasoning procedures for $\mathcal{ALC}$ knowledge bases, please refer to Ian Horrocks & Uli Sattler’s slides:

http://www.cs.man.ac.uk/~horrocks/Slides/IJCAR-tutorial/Print/p2-algorithmsII.pdf

For resources on Description Logics, please refer to the Description Logics Site:

- http://dl.kr.org/

Relationship with First Order Logic

Most DLs are well-behaved fragments of First Order Logic

To translate $\mathcal{ALC}$ to FOL:

1. Introduce: a unary predicate $A(x)$ for each atomic concept $A$
   a binary predicate $P(x, y)$ for each atomic role $P$

2. Translate complex concepts as follows, using translation functions $t_x$, for any variable $x$:

   $t_x(A) = A(x)$
   $t_x(C \cap D) = t_x(C) \land t_x(D)$
   $t_x(C \sqcup D) = t_x(C) \lor t_x(D)$
   $t_x(\exists P.C) = \exists y. P(x, y) \land t_y(C)$ with $y$ a new variable
   $t_x(\forall P.C) = \forall y. P(x, y) \supset t_y(C)$ with $y$ a new variable

3. Translate a knowledge base $\mathcal{K} = \bigcup_i \{ C_i \sqsubseteq D_i \}$ as a FOL theory

   $\Gamma_\mathcal{K} = \bigcup_i \{ \forall x. t_x(C_i) \supset t_x(D_i) \}$
Relationship with First Order Logic (Cont’d)

Reasoning services:

- \( C \) is consistent iff its translation \( t_x(C) \) is satisfiable
- \( C \sqsubseteq D \) iff \( t_x(C) \supset t_x(D) \) is valid
- \( C \) is consistent w.r.t. \( K \) iff \( \Gamma_K \cup \{ \exists x. t_x(C) \} \) is satisfiable
- \( K \models C \sqsubseteq D \) iff \( \Gamma_K \models \forall x. (t_x(C) \supset t_x(D)) \)

Relationship with First Order Logic – Exercise

Translate the following \( \mathcal{ALC} \) concepts into FOL formulas:

1. Father \( \sqcap \forall child. (\text{Doctor} \sqcup \text{Manager}) \)
2. \( \exists \text{manages.}(\text{Company} \sqcap \exists \text{employs.}\text{Doctor}) \)
3. Father \( \sqcap \forall child. (\text{Doctor} \sqcup \exists \text{manages.}(\text{Company} \sqcap \exists \text{employs.}\text{Doctor})) \)

Solution:

1. \( \text{Father}(x) \land \forall y. (\text{child}(x, y) \supset (\text{Doctor}(y) \lor \text{Manager}(y))) \)
2. \( \exists y. (\text{manages}(x, y) \land (\text{Company}(y) \land \exists w. (\text{employs}(y, w) \land \text{Doctor}(w)))) \)
3. \( \text{Father}(x) \land \forall y. (\text{child}(x, y) \supset (\text{Doctor}(y) \lor \exists w. (\text{manages}(y, w) \land (\text{Company}(w) \land \exists z. (\text{employs}(w, z) \land \text{Doctor}(z)))))) \)
DLs as fragments of First Order Logic

The above translation shows us that DLs are a fragment of First Order Logic.

In particular, we can translate complex concepts using just two translation functions \( t_x \) and \( t_y \) (thus reusing the same variables):

\[
\begin{align*}
  t_x(A) &= A(x) \\
  t_y(A) &= A(y) \\
  t_x(C \cap D) &= t_x(C) \land t_x(D) \\
  t_y(C \cap D) &= t_y(C) \land t_y(D) \\
  t_x(C \cup D) &= t_x(C) \lor t_x(D) \\
  t_y(C \cup D) &= t_y(C) \lor t_y(D) \\
  t_x(\exists P.C) &= \exists y. P(x, y) \land t_y(C) \\
  t_y(\exists P.C) &= \exists x. P(y, x) \land t_x(C) \\
  t_x(\forall P.C) &= \forall y. P(x, y) \supset t_y(C) \\
  t_y(\forall P.C) &= \forall x. P(y, x) \supset t_x(C)
\end{align*}
\]

\( \sim \mathcal{ALC} \) is a fragment of \( \mathcal{L}_2 \), i.e., FOL with 2 variables, known to be decidable (NEXPTIME-complete).

*Note:* FOL with 2 variables is more expressive than \( \mathcal{ALC} \) (tradeoff expressiveness vs. complexity of reasoning).

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**DLs as fragments of First Order Logic – Exercise**

Translate the following \( \mathcal{ALC} \) concepts into \( \mathcal{L}_2 \) formulas (i.e., into FOL formulas that use only variables \( x \) and \( y \)):

1. Father \( \cap \forall \text{child.}(\text{Doctor} \cup \text{Manager}) \)
2. \( \exists \text{manages.}(\text{Company} \cap \exists \text{employs.}\text{Doctor}) \)
3. Father \( \cap \forall \text{child.}(\text{Doctor} \cup \exists \text{manages.}(\text{Company} \cap \exists \text{employs.}\text{Doctor})) \)

**Solution:**

1. \( \text{Father}(x) \land \forall y. (\text{child}(x, y) \supset (\text{Doctor}(y) \lor \text{Manager}(y))) \)
2. \( \exists y. (\text{manages}(x, y) \land (\text{Company}(y) \land \exists x. (\text{employs}(y, x) \land \text{Doctor}(x)))) \)
3. \( \text{Father}(x) \land \forall y. (\text{child}(x, y) \supset (\text{Doctor}(y) \lor \\
\exists x. (\text{manages}(y, x) \land (\text{Company}(x) \land \exists y. (\text{employs}(x, y) \land \text{Doctor}(y)))))) \)
DLs as fragments of First Order Logic (Cont’d)

Translation can be extended to other constructs:

- For inverse roles, swap the variables in the role predicate, i.e.,

  \[ t_x(\exists P^- C) = \exists y. P(y, x) \land t_y(C) \quad \text{with } y \text{ a new variable} \]
  \[ t_x(\forall P^- C) = \forall y. P(y, x) \supset t_y(C) \quad \text{with } y \text{ a new variable} \]

\( \sim ALCI \) is still a fragment of \( L_2 \)

- For number restrictions, two variables do not suffice; but, \( ALCQT \) is a fragment of \( C_2 \) (i.e, \( L_2+\text{counting quantifiers} \))

Relationship with Modal and Dynamic Logics

In understanding the computational properties of DLs a correspondence with Modal logics and in particular with Propositional Dynamic Logics (PDLs) has been proved essential.

PDLs are logics specifically designed for reasoning about programs.

PDLs have been widely studied in computer science, especially from the point of view of computational properties:

- tree model property
- small model property
- automata based reasoning techniques
Relationship with Modal Logics

\( \mathcal{ALC} \) is a syntactic variant of \( \mathcal{K}_m \) (i.e., multi-modal \( \mathcal{K} \)):

\[
\begin{align*}
C \sqcap D & \iff C \land D & \exists P.C & \iff \Diamond_P C \\
C \sqcup D & \iff C \lor D & \forall P.C & \iff \Box_P C \\
\neg C & \iff \neg C
\end{align*}
\]

- no correspondence for inverse roles
- no correspondence for number restrictions

\( \sim \) Concept consistency, subsumption in \( \mathcal{ALC} \) \iff \text{satisfiability, validity in } \mathcal{K}_m

To encode inclusion assertions, \textit{axioms} are used

\( \sim \) Logical implication in DLs corresponds to “global logical implication” in Modal Logics

\[\text{Relationship with Propositional Dynamic Logics}\]

\( \mathcal{ALC} \) and \( \mathcal{ALCT} \) can be encoded in Propositional Dynamic Logics (PDLs)

\[
\begin{align*}
C \sqcap D & \iff C \land D & \exists R.C & \iff \langle R \rangle C \\
C \sqcup D & \iff C \lor D & \forall R.C & \iff [R]C \\
\neg C & \iff \neg C
\end{align*}
\]

Universal modality (or better “master modality”) can be expressed in PDLs using reflexive-transitive closure:

- for \( \mathcal{ALC} / \text{PDL} \):
  \[u = (P_1 \cup \cdots \cup P_m)^*\]
- for \( \mathcal{ALCT} / \text{conversePDL} \):
  \[u = (P_1 \cup \cdots \cup P_m \cup P_1^- \cup \cdots \cup P_m^-)^*\]

Universal modality allows for \textit{internalizing assertions}:

\[C \sqsubseteq D \iff [u](C \supset D)\]
Relationship with Propositional Dynamic Logics (Cont’d)

Concept satisfiability w.r.t. a KB (resp., logical implication) reduce to PDL (un)satisfiability:

$$\bigcup_i \{ C_i \sqsubseteq D_i \} \not\models C \equiv \bot \iff C \land \bigwedge_i \lbrack u \rbrack (C_i \supset D_i) \text{ satisfiable}$$

$$\bigcup_i \{ C_i \sqsubseteq D_i \} \models C \subseteq D \iff C \land \neg D \land \bigwedge_i \lbrack u \rbrack (C_i \supset D_i) \text{ unsatisfiable}$$

Correspondence also extended to other constructs, e.g., number restrictions:
- polynomial encoding when numbers are represented in unary
- technique more involved when numbers are represented in binary

*Note*: there are DLs with non first-order constructs, such as various forms of fixpoint constructs. Such DLs still have a correspondence with variants of PDLs

Consequences of correspondence with PDLs

- PDL, conversePDL, DPDL, converseDPDL are EXPTIME-complete
- Logical implication in $\mathcal{ALCQI}$ is in EXPTIME

- PDLs enjoy the tree-model property: every satisfiable formula admits a model that has the structure of a (in general infinite) tree of linearly bounded width
- A satisfiable $\mathcal{ALCQI}$ knowledge base has a tree model

- PDLs admit optimal reasoning algorithms based on (two-way alternating) automata on infinite trees
- Automata-based algorithms are optimal for $\mathcal{ALCQI}$ logical implication
DL reasoning systems

Systems are available for reasoning on DL knowledge bases:

- FaCT/Fact++ [University of Manchester]
- Pellet [University of Maryland, Clark&Parsia]
- Racer/RacerPro [University of Hamburg, Racer Systems]

Some remarks on these systems:

- the state-of-the-art DL reasoning systems are based on tableaux techniques and not on automata techniques
  + easier to implement
  - not computationally optimal (NEXPTIME, 2NEXPTIME)

- the systems are highly optimized

- despite the high computational complexity, the performance is surprisingly good in real world applications:
  - knowledge bases with thousands of concepts and hundreds of axioms
  - outperform specialized modal logics reasoners

Identification constraints

Identification constraints (aka keys) are well-studied in
- relational databases
- conceptual data models (Entity-Relationship model, UML class diagrams)

Examples of keys:

- a student is identified by its id,
  i.e., no two students have the same id

- a company is identified by its telephone number,
  i.e., given a telephone number there is a unique company which owns it
  (although a company may own more than one telephone number)

- a person is identified by its name and surname,
  i.e., no two persons have the same name and surname
Keys in $\textit{ALCQI}$

Limited forms of keys can be expressed in $\textit{ALCQI}$ using number restrictions

Examples:

- a student is identified by its id

  $$\text{StudentId} \sqsubseteq \forall \text{hasId}.\text{Student} \sqcap (\leq 1 \text{hasId})$$

  and has a unique id, i.e., the student identifies the id

  $$\text{Student} \sqsubseteq \forall \text{hasId}.\text{StudentId} \sqcap (= 1 \text{hasId})$$

- a company is identified by its telephone number

  $$\top \sqsubseteq (\leq 1 \text{telephone}.\text{Company})$$

In $\textit{ALCQI}$ only unary keys can be expressed

Keys in $\textit{ALCQI}_{id}$

$\textit{ALCQI}_{id}$ KBs extend $\textit{ALCQI}$ KBs by key assertions:

$$(\text{id } C R_1, \ldots, R_h)$$

A key assertion acts as a constraint, rather than denoting a set of objects

Semantics of a key assertion:

no two instances of $C$ agree on the participation to $R_1, \ldots, R_h$

$$a = b$$

Example: a person is identified by its name and surname

$$(\text{id } \text{Person name}, \text{surname})$$
Reasoning in $\text{ALCQI}_{id}$

Important observations:
- $\text{ALCQI}$ knowledge bases have the tree-model property
- On tree-models, non-unary keys are trivially satisfied

**Theorem:** let $\mathcal{K}$ be a set of inclusion assertions, and $\mathcal{F}$ be a set of non-unary key assertions

$\mathcal{K} \cup \mathcal{F}$ satisfiable iff $\mathcal{K}$ satisfiable

Since logical implication of inclusion assertions and concept satisfiability w.r.t. a KB can be reduced to KB satisfiability, we also have:

$\mathcal{C}$ satisfiable w.r.t. $\mathcal{K} \cup \mathcal{F}$ iff $\mathcal{C}$ satisfiable w.r.t. $\mathcal{K}$

$\mathcal{K} \cup \mathcal{F} \models \mathcal{C} \subseteq \mathcal{D}$ iff $\mathcal{K} \models \mathcal{C} \subseteq \mathcal{D}$

$\leadsto$ Key assertions do not interact with reasoning on inclusion assertions

**Logical implication of keys in $\text{ALCQI}_{id}$**

To check $\mathcal{K} \cup \mathcal{F} \models (\text{id} \; C \; R_1, \ldots, R_h)$, reduce it to unsatisfiability of $\mathcal{K} \cup \mathcal{F} \cup \mathcal{A}$, where $\mathcal{A}$ is an ABox violating the key assertion:

To check satisfiability of $\mathcal{K} \cup \mathcal{F} \cup \mathcal{A}$, it is sufficient to check the key assertions in $\mathcal{F}$ on the objects of the ABox:

1. guess a saturation $\mathcal{A}_s$ of $\mathcal{A}$, i.e., a way of completing the knowledge about objects in $\mathcal{A}$ regarding concepts and roles in $\mathcal{F}$ ($\mathcal{A}_s$ is polynomial)
2. check that $\mathcal{A}_s$ satisfies $\mathcal{F}$ (polynomial)
3. check that $\mathcal{K} \cup \mathcal{A} \cup \mathcal{A}_s$ is satisfiable (exponential)

$\leadsto$ Logical implication in $\text{ALCQI}_{id}$ is EXPTIME-complete
Reasoning on DL knowledge bases – Lower bounds

We have seen that reasoning on DL knowledge bases can be done in EXPTIME (e.g., by exploiting automata based techniques)

Are such techniques optimal for DL reasoning?
What is the intrinsic complexity of reasoning on DL knowledge bases?

Theorem: Logical implication in $\mathcal{ALC}$ (and hence concept satisfiability w.r.t. an $\mathcal{ALC}$ KB) is EXPTIME-hard

Reasoning on DL knowledge bases – Lower bounds (Cont’d)

The lower bound for logical implication in DLs can be strengthened

Theorem: concept satisfiability w.r.t. an $\mathcal{AL}$ KB and logical implication in $\mathcal{AL}$ are EXPTIME-hard

Proof: by reducing concept satisfiability w.r.t. an $\mathcal{ALC}$ KB in various steps to concept satisfiability w.r.t. an $\mathcal{AL}$ KB:

1. Reduce to satisfiability of an atomic concept w.r.t. a KB with primitive inclusion assertions only
2. Eliminate nesting of constructs in right hand side by introducing new assertions
3. Encode away qualified existential quantification
4. Encode away disjunction
1. Simplify assertions and concept

Reduce to satisfiability of an atomic concept w.r.t. a KB $\mathcal{K}$ with primitive inclusion assertions only:

$C$ satisfiable w.r.t. $\bigcup_i\{C_i \subseteq D_i\}$

iff

$A_T \cap C$ satisfiable w.r.t. $\{A_T \subseteq \bigcap_i(\neg C_i \sqcup D_i) \cap \bigcap_P \forall P.A_T\}$

iff

$A_C$ satisfiable w.r.t. $\begin{cases} A_T \subseteq \bigcap_i(\neg C_i \sqcup D_i) \cap \bigcap_P \forall P.A_T, \\ A_C \subseteq A_T \cap C \end{cases}$

with $A_T$ and $A_C$ new atomic concepts

2. Eliminate nesting of constructs in right hand side

Proceed as follows:

1. Push negation inside

2. Replace assertions as follows:

$A \sqsubseteq C_1 \cap C_2$  $\Rightarrow$  $A \sqsubseteq C_1$,  $A \sqsubseteq C_2$

$A \sqsubseteq C_1 \sqcup C_2$  $\Rightarrow$  $A \sqsubseteq A_1 \sqcup A_2$,  $A_1 \sqsubseteq C_1$,  $A_2 \sqsubseteq C_2$

$A \sqsubseteq \forall P.C$  $\Rightarrow$  $A \subseteq \forall P.A_1$,  $A_1 \sqsubseteq C$

$A \sqsubseteq \exists P.C$  $\Rightarrow$  $A \subseteq \exists P.A_1$,  $A_1 \sqsubseteq C$

with $A_1$, $A_2$ new atomic concepts for each replacement

Let $\mathcal{K}'$ be obtained from $\mathcal{K}$ by (1) and (2) above. We have $A_C$ satisfiable w.r.t. $\mathcal{K}$ iff $A_C$ satisfiable w.r.t. $\mathcal{K}'$
3. Encode away qualified existential quantification

Proceed as follows:

1. For each $\exists P.A$ appearing in $K$, introduce a new atomic role $P_A$

2. Replace assertions as follows:

$$A \subseteq \exists P.A' \Rightarrow A \subseteq \exists P_A \cap \forall P_A \cdot A'$$

$$A \subseteq \forall P.A' \Rightarrow A \subseteq \forall P.A' \cap \bigcap_{P_A} \forall P_A \cdot A'$$

Let $K''$ be obtained from $K'$ by (1) and (2) above. We have

$$A_C \text{ satisfiable w.r.t. } K' \iff A_C \text{ satisfiable w.r.t. } K''$$

$\Rightarrow$ Concept satisfiability w.r.t. a (primitive) $\mathcal{ALU}$ KB is EXPTIME-hard

4. Encode away disjunction

Replace assertions as follows:

$$A_1 \subseteq A_2 \sqcup A_3 \Rightarrow \neg A_2 \cap \neg A_3 \subseteq \neg A_1$$

The two assertions are logically equivalent

$\Rightarrow$ Concept satisfiability w.r.t. an $\mathcal{AL}$ KB is EXPTIME-hard

Concept satisfiability w.r.t. an $\mathcal{AL}$ KB can be reduced to logical implication in $\mathcal{AL}$:

$$C \text{ satisfiable w.r.t. } K \iff \text{not } K \models C \sqsubseteq \bot$$

$\Rightarrow$ Logical implication in $\mathcal{AL}$ is EXPTIME-hard
Summary on Description Logics

- Description Logics are logics for class-based modeling:
  - can be seen as a fragment of FOL with nice computational properties
  - tight relationship with Modal Logics and Propositional Dynamic Logics

- For reasoning over concept expressions, tableaux algorithms are optimal

- For most (decidable) DLs, reasoning over KBs is EXPTIME-complete:
  - tight upper bounds by automata based techniques
  - implemented systems exploit tableaux techniques, are suboptimal, but perform well in practice

- Techniques can be extended to deal also with key constraints

EXPTIME-Hardness
of Reasoning on UML Class Diagrams
We are now ready to answer our initial questions

1. **Can we develop sound, complete, and **terminating** reasoning procedures for reasoning on UML Class Diagrams?**
   
   To answer this question we polynomially encode UML Class Diagrams in DLs
   
   $\sim$ reasoning on UML Class Diagrams can be done in EXPTIME

2. **How hard is it to reason on UML Class Diagrams in general?**
   
   To answer this question we polynomially reduce reasoning in EXPTIME-complete DLs to reasoning on UML class diagrams
   
   $\sim$ reasoning on UML Class Diagrams is in fact EXPTIME-hard

We start with point (2)

---

**Reasoning tasks on UML class diagrams**

1. Consistency of the whole class diagram
2. Class consistency
3. Class subsumption
4. Class equivalence
5. · · ·

Obviously:

- Consistency of the class diagram can be reduced to class consistency
- Class equivalence can be reduced to class subsumption

We show that also class consistency and class subsumption are mutually reducible

This allows us to concentrate on class consistency only
Reducing class subsumption to class consistency

To check whether a class $C_1$ subsumes a class $C_2$ in a class diagram $D$:

1. Add to $D$ the following part, with $O$, $C$, and $C_1$ new classes

2. Check whether $C$ is inconsistent in the resulting diagram

Reducing class consistency to class subsumption

To check whether a class $C$ is inconsistent in a class diagram $D$:

1. Add to $D$ the following part, with $O$, $C_1$, $\overline{C_1}$, and $C_0$ new classes

2. Check whether $C_0$ subsumes $C$ in the resulting diagram
Lower bound for reasoning on UML class diagrams

EXPTIME lower bound established by encoding satisfiability of a concept w.r.t. an $\mathcal{ALC}$ KBs into consistency of a class in an UML class diagram

We exploit the reductions in the hardness proof of reasoning over $\mathcal{ALC}$ KBs:

- By step (1) it suffices to consider satisfiability of an atomic concept w.r.t. an $\mathcal{ALC}$ knowledge base with primitive inclusion assertions only, i.e., of the form

  \[ A \sqsubseteq C \]

- By step (2) it suffices to consider concepts on the right hand side that contain only a single construct, i.e., assertions of the form

  \[ A \sqsubseteq B \quad A \sqsubseteq \neg B \quad A \sqsubseteq B_1 \sqcup B_2 \quad A \sqsubseteq \forall P. B \quad A \sqsubseteq \exists P. B \]

Note: by step (3) it would suffice to encode $A \sqsubseteq \exists P$ instead of $A \sqsubseteq \exists P. B$

UML class diagram corresponding to an $\mathcal{ALC}$ KB - Optional

Given an $\mathcal{ALC}$ knowledge base $\mathcal{K}$ of the simplified form above, we construct an UML class diagram $\mathcal{D}_\mathcal{K}$:

- we introduce in $\mathcal{D}_\mathcal{K}$ a class $O$, intended to represent the whole domain
- for each atomic concept $A$ in $\mathcal{K}$, we introduce in $\mathcal{D}_\mathcal{K}$ a class $A$

\[ O \]

\[ A \]

- for each atomic role $P$ in $\mathcal{K}$, we introduce in $\mathcal{D}_\mathcal{K}$ a binary association $P$ with related association class

\[ O \]

\[ P \]
Encoding of $\mathcal{ALC}$ assertions - Optional

$A \subseteq B$

$A \subseteq \neg B$

$A \subseteq B_1 \sqcup B_2$

Encoding of $\mathcal{ALC}$ assertions (Cont’d) - Optional

$A \subseteq \exists P.B$
Encoding of \textit{ALC} assertions (Cont’d) - Optional

\[
A \subseteq \forall P.B
\]

Correctness of the encoding - Optional

The encoding of an \textit{ALC} knowledge base (of the simplified form) into an UML class diagram is \textit{correct}, in the sense that it \textit{preserves concept satisfiability}.

\textbf{Theorem:}

An atomic concept \(A\) is satisfiable w.r.t. an \textit{ALC} knowledge base \(\mathcal{K}\) if and only if the class \(A\) is consistent in the UML class diagram \(\mathcal{D}_\mathcal{K}\) encoding \(\mathcal{K}\).

\textbf{Proof idea:} by showing a correspondence between the models of \(\mathcal{K}\) and the models of (the FOL formalization of) \(\mathcal{D}_\mathcal{K}\).
Lower bound for reasoning on UML class diagrams

The UML class diagram $D_K$ constructed from an $\textit{ALC}$ knowledge base $K$ is of polynomial size in $K$

From

- EXPTIME-hardness of concept satisfiability w.r.t. an $\textit{ALC}$ knowledge base
- the fact that the encoding is polynomial

we obtain:

Reasoning on UML class diagrams is EXPTIME-hard
Reasoning on UML class diagrams using DLs

We can use DLs to polynomially encode UML class diagrams: this gives us EXPTIME upper bound on reasoning with UML class diagrams.

More precisely from such encoding we get

- DLs admit decidable inference
  \( \sim \) decision procedure for reasoning in UML

- (most) DLs are decidable in EXPTIME
  \( \sim \) EXPTIME method for reasoning in UML (provided the encoding in polynomial)

- exploit DL-based reasoning systems for reasoning in UML

Encoding of UML class diagrams in DLs

We encode an UML class diagram \( \mathcal{D} \) into an \( \mathcal{ALCQI}_{id} \) knowledge base \( \mathcal{K}_D \):

- classes are represented by concepts
- attributes and association roles are represented by roles
- each part of the diagram is encoded by suitable inclusion assertions
- the properties of association classes are encoded through suitable key assertions

\( \sim \) Consistency of a class in \( \mathcal{D} \) is reduced to consistency of the corresponding concept w.r.t. \( \mathcal{K}_D \)
Encoding of classes and attributes

- An UML class $C$ is represented by an atomic concept $C$.
- Each attribute $a$ of type $T$ for $C$ is represented by an atomic role $a_C$.
  - To encode the typing of $a$ for $C$:
    \[
    \exists a_C \sqsubseteq C \quad \exists a_C^- \sqsubseteq T
    \]
    We use $a_C$ as name of the role to take into account that the attribute $a$ is local to the class $C$. Sometimes, for simplicity, we directly use $a$ instead of $a_C$.
  
  - To encode the multiplicity $[i..j]$ of $a$:
    \[
    C \sqsubseteq (\geq i a_C) \cap (\leq j a_C)
    \]
    * when $j$ is $\ast$, we omit the second conjunct
    * when the multiplicity is $[0..\ast]$ we omit the whole assertion

  * when the multiplicity is missing (i.e., $[1..1]$), the assertion becomes:
    \[
    C \sqsubseteq \exists a_C \cap (\leq 1 a_C)
    \]
Encoding of classes and attributes – Example

<table>
<thead>
<tr>
<th>Phone</th>
</tr>
</thead>
<tbody>
<tr>
<td>number[1..*]: String</td>
</tr>
<tr>
<td>brand: String</td>
</tr>
<tr>
<td>lastDialed(): String</td>
</tr>
<tr>
<td>callLength(String): Integer</td>
</tr>
</tbody>
</table>

- To encode the class **Phone**, we introduce a concept **Phone**
- Encoding of the attributes: **number** and **brand**
  
  \[ \exists \text{number} \sqsubseteq \text{Phone} \quad \exists \text{brand} \sqsubseteq \text{Phone} \]
  
  \[ \exists \text{number} \sqsubseteq \text{String} \quad \exists \text{brand} \sqsubseteq \text{String} \]
  
  \[ \text{Phone} \sqsubseteq \exists \text{number} \quad \text{Phone} \sqsubseteq \exists \text{brand} \sqcap (\leq 1 \text{brand}) \]

- Encoding of the operations: **lastDialed()** and **callLength(String)**
  
  see later

Encoding of associations

The encoding depends on:
- the presence/absence of an association class
- the arity of the association

<table>
<thead>
<tr>
<th></th>
<th>without association class</th>
<th>with association class</th>
</tr>
</thead>
<tbody>
<tr>
<td>binary</td>
<td>via ALC QT role</td>
<td>via reification</td>
</tr>
<tr>
<td>non-binary</td>
<td>via reification</td>
<td>via reification</td>
</tr>
</tbody>
</table>
Encoding of binary associations without association class

- $A$ is represented by an $\textit{ALCQI}$ role $A$, with:
  \[ \exists A \sqsubseteq C_1 \quad \exists A^- \sqsubseteq C_2 \]

- To encode the multiplicities of $A$:
  - each instance of $C_1$ is connected through $A$ to at least $\min_1$ and at most $\max_1$ instances of $C_2$:
    \[ C_1 \sqsubseteq (\geq \min_1 A) \sqcap (\leq \max_1 A) \]
  - each instance of $C_2$ is connected through $A^-$ to at least $\min_2$ and at most $\max_2$ instances of $C_1$:
    \[ C_2 \sqsubseteq (\geq \min_2 A^-) \sqcap (\leq \max_2 A^-) \]

Binary associations without association class – Example

- $\exists \text{reference} \sqsubseteq \text{PhoneBill}$
- $\exists \text{reference}^- \sqsubseteq \text{PhoneCall}$
- $\text{PhoneBill} \sqsubseteq \exists \text{reference}$
- $\text{PhoneCall} \sqsubseteq \exists \text{reference}^- \sqcap (\leq 1 \text{ reference}^-)$
Encoding of associations via reification

- Association $A$ is represented by a concept $A$.
- Each instance of the concept represents a tuple of the relation.
- $n$ (binary) roles $r_{A,1}, \ldots, r_{A,n}$ are used to connect the object representing a tuple to the objects representing the components of the tuple.
- To ensure that the instances of $A$ correctly represent tuples:

$$
\exists r_{A,i} \sqsubseteq A \quad \exists r_{A,i}^{-} \sqsubseteq C_i \quad i = 1, \ldots, n
$$
$$
A \sqsubseteq \exists r_{A,1} \sqcap \cdots \sqcap \exists r_{A,n} \sqcap (\leq 1 r_{A,1}) \sqcap \cdots \sqcap (\leq 1 r_{A,n})
$$

We have not ruled out the existence of two instances of $A$ representing the same tuple of association $A$:

To rule out such a situation we could add a key assertion:

$$(\text{id } A r_1, \ldots, r_n)$$

Note: In a tree-model the above situation cannot occur.

$\leadsto$ Since in reasoning on an $\textit{ALCQI}$ KB we can restrict the attention to tree-models, we can ignore the key assertions.
Multiplicities of binary associations with association class

To encode the multiplicities of $A$ we need qualified number restrictions:

- each instance of $C_1$ is connected through $A$ to at least $min_1$ and at most $max_1$ instances of $C_2$:
  
  \[
  C_1 \sqsubseteq (\geq min_1 r_{A,1}) \cap (\leq max_1 r_{A,1})
  \]

- each instance of $C_2$ is connected through $A^-$ to at least $min_2$ and at most $max_2$ instances of $C_1$:
  
  \[
  C_2 \sqsubseteq (\geq min_2 r_{A,2}) \cap (\leq max_2 r_{A,2})
  \]

Associations with association class – Example

\[
\exists \text{call} \sqsubseteq \text{Origin} \hspace{1cm} \exists \text{call}^- \sqsubseteq \text{PhoneCall} \\
\exists \text{from} \sqsubseteq \text{Origin} \hspace{1cm} \exists \text{from}^- \sqsubseteq \text{Phone} \\
\text{Origin} \sqsubseteq \exists \text{call} \cap (\leq 1 \text{call}) \cap \exists \text{from} \cap (\leq 1 \text{from}) \\
\text{PhoneCall} \sqsubseteq (\geq 1 \text{call}^-) \cap (\leq 1 \text{call}^-)
\]
Encoding of ISA and generalization

- When the generalization is **disjoint**
  \[ C_i \subseteq \neg C_j \quad \text{for } 1 \leq i < j \leq k \]

- When the generalization is **complete**
  \[ C \subseteq C_1 \sqcup \cdots \sqcup C_k \]

ISA and generalization – Example

ETACSphone $\subseteq$ CellPhone 
GSMSphone $\subseteq$ CellPhone 
UMTSSphone $\subseteq$ CellPhone 

CellPhone $\sqsubseteq$ ETACSphone $\sqcup$ GSMphone $\sqcup$ UMTSphone

ETACSphone $\subseteq$ $\neg$GSMPhone 
ETACSphone $\subseteq$ $\neg$UMTSSphone 

GSMSphone $\subseteq$ $\neg$UMTSSphone 
UMTSSphone $\subseteq$ $\neg$UMTSSphone
Translate the above UML class diagram into an ALCQI knowledge base.
Encoding of UML in DLs – Solution of Exercise 1 (Cont’d)

Encoding of hierarchies

- Internal ⊑ Scene
- External ⊑ Scene
- Scene ⊑ Internal ⊔ External
- Internal ⊑ ¬ External

Encoding of associations

- \( \exists \text{stp}\_\text{for scn} \) ⊑ Scene
- \( \exists \text{stp}\_\text{for scn}^- \) ⊑ Setup
- Scene ⊑ \( \exists \text{stp}\_\text{for scn}^- \) \( \sqcap (\leq 1 \text{stp}\_\text{for scn}^-) \)
- \( \exists \text{tk}\_\text{of stp} \) ⊑ Setup
- \( \exists \text{tk}\_\text{of stp}^- \) ⊑ Take
- Setup ⊑ \( \exists \text{tk}\_\text{of stp}^- \) \( \sqcap (\leq 1 \text{tk}\_\text{of stp}^-) \)
- \( \exists \text{located} \) ⊑ External
- \( \exists \text{located}^- \) ⊑ Location
- External ⊑ \( (\geq 1 \text{located}) \sqcap (\leq 1 \text{located}) \)
How does the translation change w.r.t. the one for Exercise 1?

Encoding of UML in DLs – Solution of Exercise 2

The change is in the encoding of the association \(\text{located}\), which now must be reified into a concept \(\text{Located}\), i.e.,

replace

\[
\exists \text{located} \sqsubseteq \text{External} \\
\exists \text{located}^- \sqsubseteq \text{Location} \\
\text{External} \sqsubseteq (\geq 1 \text{located}) \cap (\leq 1 \text{located})
\]

with

\[
\exists r_i \sqsubseteq \text{Located} \quad i = 1, 2 \\
\exists r_1^- \sqsubseteq \text{External} \\
\exists r_2^- \sqsubseteq \text{Location} \\
\text{Located} \sqsubseteq \exists r_1 \cap (\leq 1 r_1) \cap \exists r_2 \cap (\leq 1 r_2) \\
\text{External} \sqsubseteq \exists r_1^- \cap (\leq 1 r_1^-) \\
\exists \text{nbr\_days} \sqsubseteq \text{Located} \\
\exists \text{nbr\_days}^- \sqsubseteq \text{Integer} \\
\text{Located} \sqsubseteq \exists \text{nbr\_days} \cap (\leq 1 \text{nbr\_days})
\]
Encoding of UML in DLs – Exercise 3

Encoding of operations

Operation $f(P_1, \ldots, P_m) : R$ for class $C$ corresponds to an $(m+2)$-ary relation that is functional on the last component

- Operation $f() : R$ without parameters directly represented by an atomic role $P_{f()}$, with:

  \[
  \exists P_{f()} \sqsubseteq C \quad \exists P_{f()}^- \sqsubseteq R \quad C \sqsubseteq (\leq 1 P_{f()})
  \]

- Operation $f(P_1, \ldots, P_m) : R$ with one or more parameters can be cannot expressed in $\mathcal{ALCQI}_{id}$ through reification:
  - relation is reified by using a concept $A_f(P_1, \ldots, P_m)$
  - each instance of the concept represents a tuple of the relation
  - (binary) roles $\tau_0, \ldots, \tau_{m+1}$ connect the object representing a tuple to the objects representing the components of the tuple
Reification of operations

To represent operation $f(P_1, \ldots, P_m) : R$ for class $C$:

1. $\forall r_0 \ldots \forall r_{m+1} \subseteq (\leq r_0) \cap \ldots \cap (\leq r_{m+1})$
2. $\exists r_i \subseteq A_f(P_1, \ldots, P_m) \quad \exists r_i^- \subseteq P_i \quad i = 1, \ldots, m$
3. $\exists r_o \subseteq A_f(P_1, \ldots, P_m) \quad \exists r_o^- \subseteq C$
4. $\exists r_{m+1} \subseteq A_f(P_1, \ldots, P_m) \quad \exists r_{m+1}^- \subseteq R$

(1) ensures that the instances of $A_f(P_1, \ldots, P_m)$ represent tuples
(2) ensures that the parameters of the operation have the correct types
(3) ensures that, when the operation is applied to an instance of $C$, then the result is an instance of $R$

Reification of operations (Cont’d)

Again, we have not ruled out two instances of $A_f(P_1, \ldots, P_m)$ representing two applications of the operation with identical parameters but different result:

To rule out such a situation we could add a key assertion:

$(id A_f(P_1, \ldots, P_m) \; r_0, r_1, \ldots, r_m)$

Again, by the tree-model property of $\mathcal{ALCQI}$, we can ignore the key assertion for reasoning
Encoding of operations – Example

Encoding of the operations: \texttt{lastDialed()} and \texttt{callLength(String)}

\[
\exists P_{\text{lastDialed}()} \sqsubseteq \text{Phone} \\
\exists P_{\text{lastDialed}()} \sqsubseteq \text{String} \\
\text{Phone} \sqsubseteq (\leq 1 P_{\text{lastDialed}()} ) \\
P_{\text{callLength(String)}} \sqsubseteq \exists r_0 \cap (\leq 1 r_0) \cap \exists r_1 \cap (\leq 1 r_1) \cap \exists r_2 \cap (\leq 1 r_2) \\
\exists r_i \sqsubseteq P_{\text{callLength(String)}} \quad i = 0, 1, 2 \\
\exists r c_0^- \sqsubseteq \text{Phone} \\
\exists r c_1^- \sqsubseteq \text{String} \\
\exists r c_2^- \sqsubseteq \text{Integer}
\]

Correctness of the encoding

The encoding of an UML class diagram into an \texttt{ALCQT} knowledge base is correct, in the sense that it preserves the reasoning services over UML class diagrams.

Theorem:

A class \( C \) is consistent in an UML class diagram \( D \) if and only if the concept \( C \) is satisfiable in the \texttt{ALCQT} knowledge base \( \mathcal{K}_D \) encoding \( D \)

Proof idea: by showing a correspondence between the models of (the FOL formalization of) \( D \) and the models of \( \mathcal{K}_D \).
Complexity of reasoning on UML class diagrams

All reasoning tasks on UML class diagrams can be reduced to reasoning tasks on $\mathcal{ALCQI}$ knowledge bases

From
- EXPTIME-completeness of reasoning on $\mathcal{ALCQI}$ knowledge bases
- the fact that the encoding in polynomial

we obtain:

Reasoning on UML class diagrams can be done in EXPTIME

Conclusions

- We have formalized UML class diagrams in logics, which gives us the ability to reason on them so as to detect and deduce relevant properties
- We have provided an encoding in the DL $\mathcal{ALCQI}$ thus showing that:
  1. Reasoning on UML class diagrams is decidable, and in fact EXPTIME-complete, and thus can be automatized
  2. We can perform such automated reasoning using state-of-the-art DL reasoning systems

The above results lay the foundation for advanced CASE tools with integrated automated reasoning support