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# Description Logics – Reasoning with Data

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# Recap

- For description logic knowledge bases, there are various relevant reasoning problems.
- All can be reduced to knowledge base (in)satisfiability.
- The basic description logic  $\mathcal{ALC}$  can be extended in various ways:
  - Inverse Roles  $\mathcal{I}$
  - (Qualified) Number Restrictions  $(\mathcal{Q})\mathcal{N}$
  - Nominals  $\mathcal{O}$
  - Role Hierarchies  $\mathcal{H}$
  - Transitive Roles  $\mathcal{ALC} \rightsquigarrow \mathcal{S}, \cdot_{R^+}$
- Description Logics have close connections with propositional modal logic ...
- ... and with the two-variable fragments of first-order logic (with counting quantifiers)

# Reasoning with Data

So far we have focused on terminological reasoning

- TBoxes represent general, conceptual domain knowledge
- Terminological reasoning is key to design error-free TBoxes

**New Scenario:** Ontology-based data access (OBDA)

- We have built an (error-free) TBox for our domain
- We want to populate TBox with data (add an ABox)

ABox & TBox should be compatible (no inconsistencies)

- Then, we can query the data

TBox provides vocabulary for queries

Answers reflect both TBox knowledge and ABox data

# Compatibility of Data and Knowledge

The ABox data should be compatible with the TBox knowledge

$$\begin{aligned}\mathcal{T} &= \{\text{GradSt} \sqcap \text{UnderGradSt} \sqsubseteq \perp\} \\ \mathcal{A} &= \{\text{John} : \text{GradSt}, \text{John} : \text{UnderGradSt}\}\end{aligned}$$

Nothing wrong with the TBox

Nothing wrong with the ABox

There is an obvious error when putting them together

To detect these situations we use the following problem:

## Knowledge Base satisfiability:

An instance is knowledge base  $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ .

The answer is **true** iff a model  $\mathcal{I} \models \mathcal{K}$  exists.

In a FOL setting,  $\mathcal{K}$  is satisfiable if and only if  $\pi(\mathcal{K})$  is satisfiable.

# Tableau Algorithm for KB Consistency

Tableau-based knowledge base consistency algorithm:

- Input: Knowledge Base  $\mathcal{K} = (\mathcal{T}, \mathcal{A})$
- Output: **true** iff  $\mathcal{K} = (\mathcal{T}, \mathcal{A})$  is consistent

1. Start with input ABox  $\mathcal{A}$
2. Apply expansion rules until completion or clash
3. Blocking only involves individuals not occurring in  $\mathcal{A}$

Exploit forest-model property: construct **forest-shaped** ABox  
root (ABox) individuals can be arbitrarily connected  
tree individuals (introduced by  $\exists$ -rule) form trees

Typically, we are interested in tableau algorithms that are sound and complete w.r.t. the model theory, whence the terms **satisfiable** (model-theoretic) and **consistent** (proof-theoretic) coincide.

# Tableau Example (Simplified)

$(JRA, John) : Affects$

$JRA : JuvArth$

$(JRA, Mary) : Affects$

$(John, Mary) : hasChild$

$JuvDis \sqsubseteq \exists Affects.\text{Child} \sqcap \forall Affects.\text{Child}$

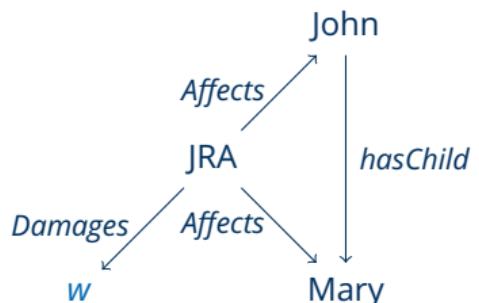
$\exists hasChild.\top \sqsubseteq \text{Adult}$

$\text{Adult} \sqsubseteq \neg \text{Child}$

$\text{Arth} \sqsubseteq \exists Damages.\text{Joint}$

$JuvArth \sqsubseteq \text{Arth} \sqcap \text{JuvDis}$

Tableau expansion (simplified):



$\text{con}_{\mathcal{A}}(\text{JRA})$	$= \{\text{JuvArth, Arth, JuvDis, } \exists \text{Damages.Joint, } \exists Affects.\text{Child, } \forall Affects.\text{Child}\}$
$\text{con}_{\mathcal{A}}(\text{John})$	$= \{\text{Child, Adult, } \neg \text{Child}\}$
$\text{con}_{\mathcal{A}}(\text{Mary})$	$= \{\text{Child}\}$
$\text{con}_{\mathcal{A}}(w)$	$= \{\text{Joint}\}$

# Querying the Data

It does not make sense to query an inconsistent  $\mathcal{K}$  (previous example).

- An inconsistent ( $\hat{=}$  unsatisfiable)  $\mathcal{K}$  entails all formulas.
- We (typically) fix inconsistencies before we start asking queries.

Once we have determined that  $\mathcal{K}$  is consistent, we want to query the data:

- Which children are affected by a juvenile arthritis?
- Which drugs are used to treat JRA?
- Who is affected by an arthritis and is allergic to steroids?

Similar to the types of queries one would pose to a database.

```
SELECT Child cname
      FROM Child, Affects, JuvArth
     WHERE Child cname = Affects cname AND
           Affects dname = JuvArth dname
```

# Querying the Data: Simple Queries (1)

The basic data queries ask for all the instances of a concept:

$$q_1(x) = \text{Child}(x)$$

Set of children?

$$q_2(x) = (\text{Dis} \sqcap \exists \text{Damages.Joint})(x)$$

Set of diseases affecting a joint?

How to (naively) answer these queries? Try each individual name.

ABox  $\mathcal{A}$

(JRA, John) : *Affects*

JRA : JuvArth

(JRA, Mary) : *Affects*

TBox  $\mathcal{T}$

$(\mathcal{K} = (\mathcal{T}, \mathcal{A}))$

JuvDis  $\sqsubseteq \exists \text{Affects}.\text{Child} \sqcap \forall \text{Affects}.\text{Child}$

Adult  $\sqsubseteq \neg \text{Child}$

Arth  $\sqsubseteq \exists \text{Damages.Joint}$

JuvArth  $\sqsubseteq \text{Arth} \sqcap \text{JuvDis}$

$\mathcal{K} \models \text{JRA} : \text{Child}?$  *No.* JRA is not an answer to  $q_1$

$\mathcal{K} \models \text{John} : \text{Child}?$  *Yes!* John is an answer to  $q_1$

$\mathcal{K} \models \text{Mary} : \text{Child}?$  *Yes!* Mary is an answer to  $q_1$

# Querying the Data: Simple Queries (2)

So, we are interested in the following decision problem:

## Concept Instance Checking:

Given individual name  $a$ , concept  $C$  and KB  $\mathcal{K}$ ,  
an instance is a triple  $\langle a, C, \mathcal{K} \rangle$ .

The answer is **true** iff  $\mathcal{K} \models a : C$

In  $\mathcal{ALC}$  (and extensions) this problem is reducible to KB satisfiability:

$$(\mathcal{T}, \mathcal{A}) \models a : C \quad \text{iff} \quad (\mathcal{T}, \mathcal{A} \cup \{a : \neg C\}) \text{ satisfiable}$$

Note that we can assume, w.l.o.g., that  $C$  is a concept name:

$$(\mathcal{T}, \mathcal{A}) \models a : C \quad \text{iff} \quad (\mathcal{T} \cup \{X \equiv C\}, \mathcal{A}) \models a : X$$

where  $X$  is a concept name that does not occur in  $\mathcal{T}$  or  $\mathcal{A}$ .

# Querying the Data: Simple Queries (3)

What about instances of a role:

$q_2(x, y) = \text{hasChild}(x, y)$  Set of parent-child tuples?

How to (naively) answer these queries? Try each pair of individuals!

ABox  $\mathcal{A}$

JRA:JuvArth  
(JRA, Mary):Affects  
(John, Mary):hasChild

TBox  $\mathcal{T}$   $(\mathcal{K} = (\mathcal{T}, \mathcal{A}))$

JuvDis  $\sqsubseteq \exists \text{Affects}.\text{Child} \sqcap \forall \text{Affects}.\text{Child}$   
Adult  $\sqsubseteq \neg \text{Child}$   
Arth  $\sqsubseteq \exists \text{Damages}.\text{Joint}$   
JuvArth  $\sqsubseteq \text{Arth} \sqcap \text{JuvDis}$

$\mathcal{K} \models (\text{John}, \text{John}): \text{hasChild}$ ? **No.** ( $\text{John}, \text{John}$ ) is not an answer to  $q_2$

$\mathcal{K} \models (\text{John}, \text{Mary}): \text{hasChild}$ ? **Yes!** ( $\text{John}, \text{Mary}$ ) is an answer to  $q_2$

$\mathcal{K} \models (\text{John}, \text{JRA}): \text{hasChild}$ ? **No.** ( $\text{John}, \text{John}$ ) is not an answer to  $q_2$

...

# Querying the Data: Simple Queries (4)

So, we are interested in the following decision problem:

## Role Instance Checking:

Given a pair of individual names  $(a, b)$ , role  $R$  and KB  $\mathcal{K}$ ,  
an instance is a triple  $\langle (a, b), R, \mathcal{K} \rangle$ .

The answer is **true** iff  $\mathcal{K} \models (a, b) : R$

Can this problem be reduced to knowledge base consistency?

$$(\mathcal{T}, \mathcal{A}) \models (a, b) : R \quad \text{iff} \quad (\mathcal{T}, \mathcal{A} \cup \{a : \forall R.X, b : \neg X\}) \text{ is inconsistent}$$

where  $X$  is a concept name that does not occur in  $\mathcal{T}$  or  $\mathcal{A}$ .

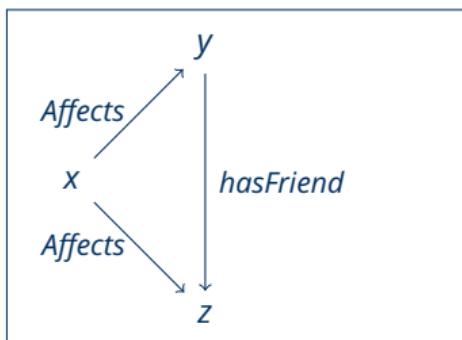
# Limitations of Concept-based Queries

Some natural queries cannot be expressed using a concept:

$$q(y) = \exists x \exists z (Affects(x, y) \wedge Affects(x, z) \wedge hasFriend(y, z))$$

Set of people ( $y$ ) affected by the same disease as a friend?

Query Graph:



We can only represent tree-like queries as concepts

Related to the tree model property of DLs

We need a more expressive query language ...

# Conjunctive Queries

The language of conjunctive queries:

- Generalises concept-based queries in a natural way  
arbitrarily-shaped queries vs. tree-like queries
- Widely used as a query language in databases  
Corresponds to Select-Project-Join fragment of relational algebra  
Fragment of relational calculus using only  $\exists$  and  $\wedge$
- Implemented in most DBMS

We next study the problem of CQ answering over DL knowledge bases

We will not study the problem of answering FOL queries over DL KBs

- ~~ Corresponds to general relational calculus queries.
- ~~ Leads to an undecidable decision problem.

# Conjunctive Queries – Definition

## Conjunctive query

Let  $\mathbf{V}$  be a set of **variables**.

A **term**  $t$  is a variable from  $\mathbf{V}$  or an individual name from  $\mathbf{I}$ .

A **conjunctive query** (CQ)  $q$  has the form  $\exists x_1 \dots \exists x_k (a_1 \wedge \dots \wedge a_n)$  where:

- $k \geq 0, n \geq 1, x_1, \dots, x_k \in \mathbf{V}$ ;
- each  $a_i$  is a **concept atom**  $A(t)$  or a **role atom**  $r(t, t')$ , that is,  $A \in \mathbf{C}$ ,  $r \in \mathbf{R}$ , and  $t, t'$  are **terms**;
- $x_1, \dots, x_k$  are called **quantified variables**; all other variables in  $q$  are called **answer variables**;
- the **arity** of  $q$  is the number of answer variables;
- $q$  is called **Boolean** if it has arity zero.

To indicate that the answer variables in a CQ  $q$  are  $\vec{x}$ , we often write  $q(\vec{x})$  instead of just  $q$ .

# Example Conjunctive Queries

1. Return all pairs of individual names  $(a, b)$  such that  $a$  is a professor who supervises student  $b$ :

$$q_1(x_1, x_2) = \text{Professor}(x_1) \wedge \text{supervises}(x_1, x_2) \wedge \text{Student}(x_2).$$

2. Return all individual names  $a$  such that  $a$  is a student supervised by some professor:

$$q_2(x) = \exists y (\text{Professor}(y) \wedge \text{supervises}(y, x) \wedge \text{Student}(x)).$$

3. Return all pairs of students supervised by the same professor:

$$q_3(x_1, x_2) = \exists y (\text{Professor}(y) \wedge \text{supervises}(y, x_1) \wedge \text{supervises}(y, x_2) \wedge \text{Student}(x_1) \wedge \text{Student}(x_2)).$$

4. Return all students supervised by professor smith (an individual name):

$$q_4(x) = \text{supervises}(\text{smith}, x) \wedge \text{Student}(x).$$

# Answers on an Interpretation

We first define query answers on a given interpretation  $\mathcal{I}$ .

## Definition

Let  $q$  be a conjunctive query and  $\mathcal{I}$  an interpretation.

We use  $\text{term}(q)$  to denote the terms in  $q$ .

A **match of  $q$  in  $\mathcal{I}$**  is a mapping  $\pi: \text{term}(q) \rightarrow \Delta^{\mathcal{I}}$  such that

- $\pi(a) = a^{\mathcal{I}}$  for all  $a \in \text{term}(q) \cap \mathbf{I}$ ,
- $\pi(t) \in A^{\mathcal{I}}$  for all concept atoms  $A(t)$  in  $q$ , and
- $(\pi(t_1), \pi(t_2)) \in r^{\mathcal{I}}$  for all role atoms  $r(t_1, t_2)$  in  $q$ .

Let  $\vec{x} = x_1, \dots, x_k$  be the answer variables in  $q$  and  $\vec{a} = a_1, \dots, a_k$  be individual names from  $\mathbf{I}$ .

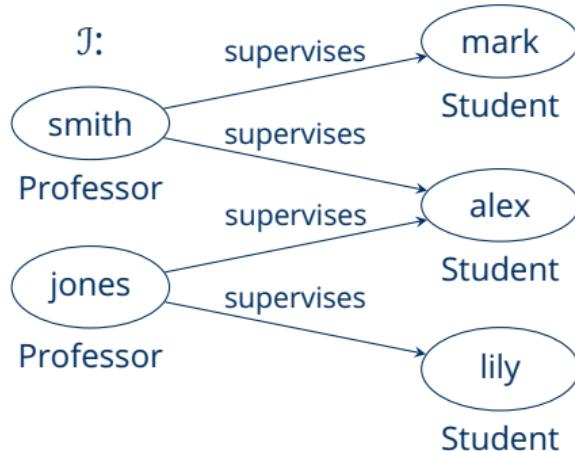
We call the match  $\pi$  of  $q$  in  $\mathcal{I}$  an  **$\vec{a}$ -match** if  $\pi(x_i) = a_i^{\mathcal{I}}$  for  $1 \leq i \leq k$ .

We say that  $\vec{a}$  is an **answer to  $q$  on  $\mathcal{I}$**  if there is an  $\vec{a}$ -match  $\pi$  of  $q$  in  $\mathcal{I}$ .

We use  $\text{ans}(q, \mathcal{I})$  to denote the set of all answers to  $q$  on  $\mathcal{I}$ .

# Answers on Interpretation $\mathcal{I}$ (1)

$q_2(x)$ :



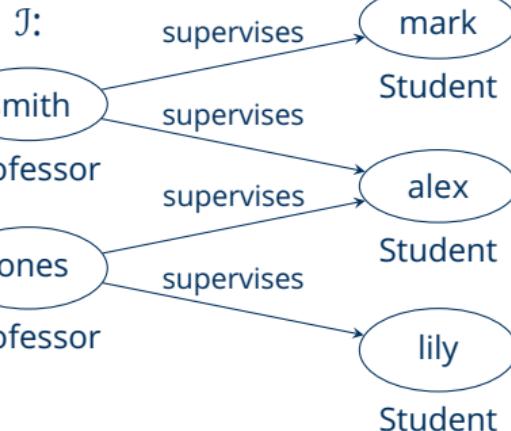
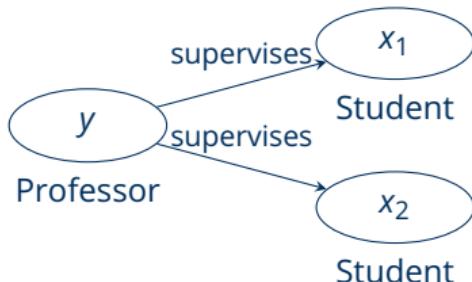
$$q_2(x) = \exists y (\text{Professor}(y) \wedge \text{supervises}(y, x) \wedge \text{Student}(x))$$

There are 3 answers to  $q_2(x)$  on  $\mathcal{I}$ : mark, alex, and lily.

Note that a match is a **homomorphism** from the query to the interpretation (both viewed as a graphs).

# Answers on Interpretation $\mathcal{I}$ (2)

$q_3(x_1, x_2)$ :



$$q_3(x_1, x_2) = \exists y (\text{Professor}(y) \wedge \text{supervises}(y, \underline{x_1}) \wedge \text{supervises}(y, \underline{x_2}) \wedge \text{Student}(\underline{x_1}) \wedge \text{Student}(\underline{x_2})).$$

There are 7 answers to  $q_3(x_1, x_2)$  on  $\mathcal{I}$ , including (mark, alex), (alex, lily), (lily, alex) and (mark, mark). Note that a match need not be injective.

# Certain Answers

Usually we are interested in answers on a KB, which may have many models. In this case, so-called **certain answers** provide a natural semantics.

## Definition

Let  $q$  be a CQ and  $\mathcal{K} = (\mathcal{T}, \mathcal{A})$  be a KB.

We say that  $\vec{a}$  is a **certain answer to  $q$  on  $\mathcal{K}$**  if

- all individual names from  $\vec{a}$  occur in  $\mathcal{A}$ , and
- $\vec{a} \in \text{ans}(q, \mathcal{I})$  for every model  $\mathcal{I}$  of  $\mathcal{K}$ .

We use  $\text{cert}(q, \mathcal{K})$  to denote the set of all certain answers to  $q$  on  $\mathcal{K}$ :

$$\text{cert}(q, \mathcal{K}) = \bigcap_{\mathcal{I} \models \mathcal{K}} \text{ans}(q, \mathcal{I})$$

# Certain Answers: Examples

Consider the  $\mathcal{ALCJ}$  KB  $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ :

$$\mathcal{T} = \{\text{Student} \sqsubseteq \exists \text{supervises}^-. \text{Professor}\},$$

$$\mathcal{A} = \{\text{smith} : \text{Professor}, \text{mark} : \text{Student}, \text{alex} : \text{Student}, \text{lily} : \text{Student},$$
$$(\text{smith}, \text{mark}) : \text{supervises}, (\text{smith}, \text{alex}) : \text{supervises}\}.$$

- $q_4(x) = \text{supervises}(\text{smith}, \underline{x}) \wedge \text{Student}(\underline{x})$ ;  $\text{cert}(q_4, \mathcal{K}) = \{\text{mark}, \text{alex}\}$ : there are models of  $\mathcal{K}$  in which smith supervises other students, but only mark and alex are supervised by smith in *all* models.
- $q_2(x) = \exists y (\text{Professor}(y) \wedge \text{supervises}(y, \underline{x}) \wedge \text{Student}(\underline{x}))$ ;  
 $\text{cert}(q_2, \mathcal{K}) = \{\text{mark}, \text{alex}, \text{lily}\}$ : note that lily is included because she is a student and thus the TBox enforces that in every model of  $\mathcal{K}$  she has a supervisor who is a professor.
- $q_1(x_1, x_2) = \text{Professor}(\underline{x_1}) \wedge \text{supervises}(\underline{x_1}, \underline{x_2}) \wedge \text{Student}(\underline{x_2})$ ;  
 $\text{cert}(q_1, \mathcal{K}) = \{(\text{smith}, \text{mark}), (\text{smith}, \text{alex})\}$ : lily always has a supervisor, but there is no supervisor (known by name) on which all models agree.

# Boolean Conjunctive Query Answering

(Arbitrary) CQ answering reduces to Boolean CQ answering.

Given query  $q$  of arity  $n$  and  $\mathcal{K} = (\mathcal{T}, \mathcal{A})$  in which  $m$  individual names occur:

- Iterate through  $m^n$  tuples of arity  $n$ .
- For each tuple  $\vec{a} = (a_1, \dots, a_n)$  create a Boolean query  $q_{\vec{a}}$  by replacing the  $i$ th answer variable with  $a_i$ , for all  $1 \leq i \leq n$ .
- $\vec{a} \in \text{cert}(q, \mathcal{K})$  iff  $\mathcal{K} \models q_{\vec{a}}$ .

## Boolean Conjunctive Query Entailment:

An instance is a pair  $\langle \mathcal{K}, q \rangle$   
with  $\mathcal{K}$  a KB and  $q$  a Boolean CQ.

The answer is **true** iff  $\mathcal{I} \models q$  for each  $\mathcal{I} \models \mathcal{K}$ .

This problem is **not** trivially reducible to knowledge base satisfiability.

It is ExpTime-complete for  $\mathcal{ALC}$ , the same as satisfiability.  
(A proof is beyond the scope of this course.)

# Boolean Conjunctive Query Answering

Many types of query **can** be reduced to KB satisfiability:

- Concept and role instance queries, e.g.,  $q() = C(a)$  and  $q() = r(a, b)$ .
- Fully ground queries, e.g.,  $q() = C(a) \wedge D(b) \wedge r(a, b)$   
(idea: check each atom independently).
- Forest shaped queries, e.g.,  $q() = \exists x(C(a) \wedge D(x) \wedge r(a, x))$   
(idea: roll up the tree parts of the query).

Reduction may or may not be possible in general (possible for  $\mathcal{SHIQ}$ ; open problem for  $\mathcal{SHOIQ}$ ).

# Conjunctive Query Answering (1)

How to interpret the answer to a Boolean Query?

$(\mathcal{K} = (\mathcal{T}, \mathcal{A}))$

ABox  $\mathcal{A}$ :

$(\text{JRA, John}) : \text{Affects}$   
 $\text{JRA} : \text{JuvArth}$   
 $(\text{JRA, Mary}) : \text{Affects}$

TBox  $\mathcal{T}$ :

$\text{JuvDis} \sqsubseteq \exists \text{Affects}.\text{Child} \sqcap \forall \text{Affects}.\text{Child}$   
 $\text{Adult} \sqsubseteq \neg \text{Child}$   
 $\text{Arth} \sqsubseteq \exists \text{Damages}.\text{Joint}$   
 $\text{JuvArth} \sqsubseteq \text{Arth} \sqcap \text{JuvDis}$

$q_1 = \text{Affects}(\text{JRA, Mary})$

$q_2 = \text{Child}(\text{Mary})$

$q_3 = \text{Adult}(\text{Mary})$

$q_4 = \exists y(\text{Damages}(\text{JRA}, y) \wedge \text{Organ}(y))$

$\mathcal{A} \models q_1$	Yes
$\mathcal{A} \not\models q_2, \mathcal{A} \not\models \neg q_2$	???
$\mathcal{K} \models q_2$	Yes
$\mathcal{A} \not\models q_3, \mathcal{A} \not\models \neg q_3$	???
$\mathcal{K} \models \neg q_3$	No
$\mathcal{A} \not\models q_4, \mathcal{A} \not\models \neg q_4$	???
$\mathcal{K} \not\models q_4, \mathcal{K} \not\models \neg q_4$	???

# Conjunctive Query Answering (2)

$\mathcal{A}$  is seen as a FOL knowledge base, but  $\mathcal{D}$  is seen as a FOL model:

ABox  $\mathcal{A}$

$(JRA, John) : Affects$   
 $JRA : JuvArth$   
 $(JRA, Mary) : Affects$

Database  $\mathcal{D}$

<i>Affects</i>		<i>JuvArthritis</i>	
JRA	John	JRA	
JRA	Mary		

$q_1 = Affects(JRA, Mary)$

$\mathcal{A} \models q_1$  Yes

$q_2 = \text{Child}(Mary)$

$\mathcal{D} \models q_1$  Yes

$q_3 = \text{Adult}(Mary)$

$\mathcal{A} \not\models q_2, \mathcal{A} \not\models \neg q_2$  ???

$q_4 = \exists y(Damages(JRA, y) \wedge \text{Organ}(y))$

$\mathcal{D} \not\models q_2$  No

$\mathcal{A} \not\models q_3, \mathcal{A} \not\models \neg q_3$  ???

$\mathcal{D} \not\models q_3$  No

$\mathcal{A} \not\models q_4, \mathcal{A} \not\models \neg q_4$  ???

$\mathcal{D} \not\models q_4$  No

# Ontologies vs. Database Systems

## Conceptual DB-Schema:

- Typically formulated as an ER or UML diagram (used in DB design)
- Schema leads to a set of FOL-based constraints
- Constraints are used to check conformance of the data
- Constraints are disregarded for query answering
  - ~ In databases, query answering is a FOL **model checking** problem.

## Description Logic TBoxes:

- Formulated in a Description Logic (fragment of FOL)
- TBox axioms are used to check conformance of the data
  - The way this is done differs from DBs
- TBox axioms participate in query answering
  - ~ In description logics, query answering is a FOL **entailment** problem.

# KB Consistency: Practicality Issues

- Addition of ABox may greatly exacerbate practicality problems
  - No obvious limit to size of data – could be millions or even billions of individuals
  - Tableau algorithm applied to whole ABox
- Optimisations can ameliorate but not eliminate the problem
- Can exploit decomposition of an ABox:
  - $\mathcal{A}$  can be decomposed into a set of disjoint connected components  $\{\mathcal{A}_1, \dots, \mathcal{A}_n\}$  such that:

$$\mathcal{A} = \mathcal{A}_1 \cup \dots \cup \mathcal{A}_n$$
$$\forall 1 \leq i < j \leq n \text{ ind}(\mathcal{A}_i) \cap \text{ind}(\mathcal{A}_j) = \emptyset$$

where  $\text{ind}(\mathcal{A}_i)$  is the set of individuals (constants) occurring in  $\mathcal{A}_i$

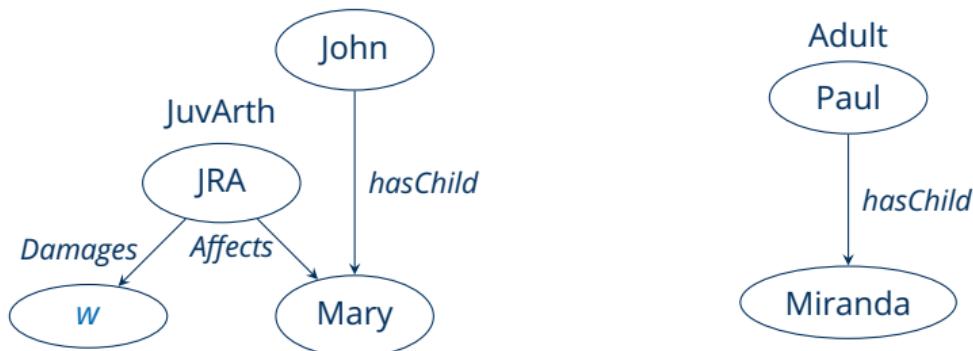
- An  $\mathcal{ALC}$  KB  $(\mathcal{T}, \mathcal{A})$  is consistent iff  $(\mathcal{T}, \mathcal{A}_i)$  is consistent for each  $\mathcal{A}_i$  in a decomposition  $\{\mathcal{A}_1, \dots, \mathcal{A}_n\}$  of  $\mathcal{A}$

# ABox Decomposition: Example

JRA : JuvArth  
(JRA, Mary) : Affects  
(John, Mary) : hasChild  
(Paul, Miranda) : hasChild  
Paul : Adult

JuvDis  $\sqsubseteq \exists Affects.\text{Child} \sqcap \forall Affects.\text{Child}$   
 $\exists hasChild.\top \sqsubseteq \text{Adult}$   
 $\text{Adult} \sqsubseteq \neg \text{Child}$   
Arth  $\sqsubseteq \exists Damages.\text{Joint}$   
JuvArth  $\sqsubseteq \text{Arth} \sqcap \text{JuvDis}$

Perform separate consistency tests on the disjoint connected components:



# Query Answering: Practicality Issues

- Recall our example query:

$$q(y) = \exists x \exists z (Affects(x, y) \wedge Affects(x, z) \wedge hasFriend(y, z))$$

- To answer this query we have to:

- check for each individual  $a$  occurring in  $\mathcal{A}$  if  $(\mathcal{T}, \mathcal{A}) \models q_{[y/a]}$ , where  $q_{[y/a]}$  is the Boolean CQ

$$q() = \exists x \exists z (Affects(x, a) \wedge Affects(x, z) \wedge hasFriend(a, z))$$

- checking  $(\mathcal{T}, \mathcal{A}) \models q_{[y/a]}$  involves performing (possibly many) consistency tests;
- each test could be very costly.

- And what if we change the query to

$$q(x, y, z) = Affects(x, y) \wedge Affects(x, z) \wedge hasFriend(y, z)?$$

- In general, there are  $m^n$  “candidate” answer tuples, where  $m$  is the number of individuals occurring in  $\mathcal{A}$  and  $n$  the arity of the query.

# Optimised Query Answering

Many optimisations are possible, for example:

- Exploit the fact that we cannot entail ABox roles in  $\mathcal{ALC}$ , that is:

$$(\mathcal{T}, \mathcal{A}) \models R(a, b) \text{ iff } R(a, b) \in \mathcal{A}$$

- Only check candidate tuples with relevant relational structure
- So for

$$q(y, z) = \exists x ( \text{JuvArth}(x) \wedge \text{Affects}(x, y) \wedge \text{hasFriend}(y, z) )$$

only check tuples  $(a, b)$  such that

$$\text{hasFriend}(a, b) \in \mathcal{A}$$

and for these we only need to check the Boolean CQ:

$$\exists x ( \text{JuvArth}(x) \wedge \text{Affects}(x, a) \wedge \text{Affects}(x, b) )$$

# Conflicting Requirements

Ontology-based data access applications require:

1. Very expressive ontology languages
  - As large fragment of FOL as possible
2. Powerful query languages
  - As large fragment of SQL as possible
3. Efficient query answering algorithms
  - Low complexity, easy to optimise

**The requirements are in conflict!**

~~ We need to make compromises.

# Conclusion

- DL KB consistency can be decided using tableau algorithms  
~~ Idea: Make implicit inconsistencies explicit/construct model
- Query answering for DL KBs is understood as FOL *entailment*
- Conjunctive Queries (CQs) constitute natural query language
- CQs induce answers on a single interpretation, and *certain answers* on a KB
- Boolean CQ Entailment is not trivially reducible to KB consistency
- In contrast, CQ Entailment in databases is understood as FOL *model checking*