

# Knowledge Bases and Databases

## Part 1: First-Order Queries

Diego Calvanese

Faculty of Computer Science  
Master of Science in Computer Science

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FREIE UNIVERSITÄT BOZEN  
LIBERA UNIVERSITÀ DI BOLZANO  
FREE UNIVERSITY OF BOZEN - BOLZANO

## Overview of Part 1: First-order queries

- 1 First-order logic
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  - 2 Semantics of first-order logic
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- 2 First-order query evaluation
  - 1 Query evaluation problem
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## Chapter I

### First-Order Logic



## Outline

- 1 Syntax of first-order logic
- 2 Semantics of first-order logic
- 3 First-order logic queries



## Outline

- 1 Syntax of first-order logic
- 2 Semantics of first-order logic
- 3 First-order logic queries



## First-order logic

- First-order logic (FOL) is the logic to speak about **objects**, which are the domain of discourse or universe.
- FOL is concerned about **properties** of these objects and **relations** over objects (resp., unary and  $n$ -ary **predicates**).
- FOL also has **functions** including **constants** that denote objects.



## FOL syntax – Terms

We first introduce:

- A set  $Vars = \{x_1, \dots, x_n\}$  of **individual variables** (i.e., variables that denote single objects).
- A set of **functions symbols**, each of given arity  $\geq 0$ . Functions of arity 0 are called **constants**.

Def.: The set of **Terms** is defined inductively as follows:

- $Vars \subseteq Terms$ ;
- If  $t_1, \dots, t_k \in Terms$  and  $f^k$  is a  $k$ -ary function symbol, then  $f^k(t_1, \dots, t_k) \in Terms$ ;
- Nothing else is in  $Terms$ .



## FOL syntax – Formulas

Def.: The set of **Formulas** is defined inductively as follows:

- If  $t_1, \dots, t_k \in Terms$  and  $P^k$  is a  $k$ -ary predicate, then  $P^k(t_1, \dots, t_k) \in Formulas$  (atomic formulas).
- If  $t_1, t_2 \in Terms$ , then  $t_1 = t_2 \in Formulas$ .
- If  $\varphi \in Formulas$  and  $\psi \in Formulas$  then
  - $\neg\varphi \in Formulas$
  - $\varphi \wedge \psi \in Formulas$
  - $\varphi \vee \psi \in Formulas$
  - $\varphi \rightarrow \psi \in Formulas$
- If  $\varphi \in Formulas$  and  $x \in Vars$  then
  - $\exists x.\varphi \in Formulas$
  - $\forall x.\varphi \in Formulas$
- Nothing else is in  $Formulas$ .

**Note:** a predicate of arity 0 is a proposition of propositional logic.



## Outline

- 1 Syntax of first-order logic
- 2 Semantics of first-order logic
- 3 First-order logic queries



## Interpretations

Given an **alphabet** of predicates  $P_1, P_2, \dots$  and functions  $f_1, f_2, \dots$ , each with an associated arity, a FOL **interpretation** is:

$$\mathcal{I} = (\Delta^{\mathcal{I}}, P_1^{\mathcal{I}}, P_2^{\mathcal{I}}, \dots, f_1^{\mathcal{I}}, f_2^{\mathcal{I}}, \dots)$$

where:

- $\Delta^{\mathcal{I}}$  is the domain (a set of objects)
- if  $P_i$  is a  $k$ -ary predicate, then  $P_i^{\mathcal{I}} \subseteq \Delta^{\mathcal{I}} \times \dots \times \Delta^{\mathcal{I}}$  ( $k$  times)
- if  $f_i$  is a  $k$ -ary function, then  $f_i^{\mathcal{I}} : \Delta^{\mathcal{I}} \times \dots \times \Delta^{\mathcal{I}} \rightarrow \Delta^{\mathcal{I}}$  ( $k$  times)
- if  $f_i$  is a constant (i.e., a 0-ary function), then  $f_i^{\mathcal{I}} : () \rightarrow \Delta^{\mathcal{I}}$  (i.e.,  $f_i$  denotes exactly one object of the domain)



## Assignment

Let  $Vars$  be a set of (individual) variables.

Def.: Given an interpretation  $\mathcal{I}$ , an **assignment** is a function

$$\alpha : Vars \rightarrow \Delta^{\mathcal{I}}$$

that assigns to each variable  $x \in Vars$  an object  $\alpha(x) \in \Delta^{\mathcal{I}}$ .

It is convenient to extend the notion of assignment to terms. We can do so by defining a function  $\hat{\alpha} : Terms \rightarrow \Delta^{\mathcal{I}}$  inductively as follows:

- $\hat{\alpha}(x) = \alpha(x)$ , if  $x \in Vars$
- $\hat{\alpha}(f(t_1, \dots, t_k)) = f^{\mathcal{I}}(\hat{\alpha}(t_1), \dots, \hat{\alpha}(t_k))$

**Note:** for constants  $\hat{\alpha}(c) = c^{\mathcal{I}}$ .



## Truth in an interpretation wrt an assignment

We define when a FOL formula  $\varphi$  is **true** in an interpretation  $\mathcal{I}$  wrt an assignment  $\alpha$ , written  $\mathcal{I}, \alpha \models \varphi$ :

- $\mathcal{I}, \alpha \models P(t_1, \dots, t_k)$  if  $(\hat{\alpha}(t_1), \dots, \hat{\alpha}(t_k)) \in P^{\mathcal{I}}$
- $\mathcal{I}, \alpha \models t_1 = t_2$  if  $\hat{\alpha}(t_1) = \hat{\alpha}(t_2)$
- $\mathcal{I}, \alpha \models \neg \varphi$  if  $\mathcal{I}, \alpha \not\models \varphi$
- $\mathcal{I}, \alpha \models \varphi \wedge \psi$  if  $\mathcal{I}, \alpha \models \varphi$  and  $\mathcal{I}, \alpha \models \psi$
- $\mathcal{I}, \alpha \models \varphi \vee \psi$  if  $\mathcal{I}, \alpha \models \varphi$  or  $\mathcal{I}, \alpha \models \psi$
- $\mathcal{I}, \alpha \models \varphi \rightarrow \psi$  if  $\mathcal{I}, \alpha \models \varphi$  implies  $\mathcal{I}, \alpha \models \psi$
- $\mathcal{I}, \alpha \models \exists x. \varphi$  if for some  $a \in \Delta^{\mathcal{I}}$  we have  $\mathcal{I}, \alpha[x \mapsto a] \models \varphi$
- $\mathcal{I}, \alpha \models \forall x. \varphi$  if for every  $a \in \Delta^{\mathcal{I}}$  we have  $\mathcal{I}, \alpha[x \mapsto a] \models \varphi$

Here,  $\alpha[x \mapsto a]$  stands for the new assignment obtained from  $\alpha$  as follows:

$$\begin{aligned} \alpha[x \mapsto a](x) &= a \\ \alpha[x \mapsto a](y) &= \alpha(y) \quad \text{for } y \neq x \end{aligned}$$



## Open vs. closed formulas

## Definitions

- A variable  $x$  in a formula  $\varphi$  is **free** if  $x$  does not occur in the scope of any quantifier, otherwise it is **bounded**.
- An **open formula** is a formula that has some free variable.
- A **closed formula**, also called **sentence**, is a formula that has no free variables.

For **closed formulas** (but not for open formulas) we can define what it means to be **true in an interpretation**, written  $\mathcal{I} \models \varphi$ , without mentioning the assignment, since the assignment  $\alpha$  does not play any role in verifying  $\mathcal{I}, \alpha \models \varphi$ .

Instead, open formulas are strongly related to **queries** — cf. relational databases.



## Outline

- 1 Syntax of first-order logic
- 2 Semantics of first-order logic
- 3 **First-order logic queries**



## FOL queries

Def.: A **FOL query** is an (open) FOL formula.

When  $\varphi$  is a FOL query with free variables  $(x_1, \dots, x_k)$ , then we sometimes write it as  $\varphi(x_1, \dots, x_k)$ , and say that  $\varphi$  has **arity**  $k$ .

Given an interpretation  $\mathcal{I}$ , we are interested in those assignments that map the variables  $x_1, \dots, x_k$  (and only those). We write an assignment  $\alpha$  s.t.  $\alpha(x_i) = a_i$ , for  $i = 1, \dots, k$ , as  $\langle a_1, \dots, a_k \rangle$ .

Def.: Given an interpretation  $\mathcal{I}$ , the **answer to a query**  $\varphi(x_1, \dots, x_k)$  is

$$\varphi(x_1, \dots, x_k)^{\mathcal{I}} = \{ \langle a_1, \dots, a_k \rangle \mid \mathcal{I}, \langle a_1, \dots, a_k \rangle \models \varphi(x_1, \dots, x_k) \}$$

*Note:* We will also use the notation  $\varphi^{\mathcal{I}}$ , which keeps the free variables implicit, and  $\varphi(\mathcal{I})$  making apparent that  $\varphi$  becomes a function from interpretations to set of tuples.



## FOL boolean queries

Def.: A **FOL boolean query** is a FOL query without free variables.

Hence, the answer to a boolean query  $\varphi()$  is defined as follows:

$$\varphi()^{\mathcal{I}} = \{ () \mid \mathcal{I}, \langle \rangle \models \varphi() \}$$

Such an answer is

- $()$ , if  $\mathcal{I} \models \varphi$
- $\emptyset$ , if  $\mathcal{I} \not\models \varphi$ .

As an obvious convention we read  $()$  as “true” and  $\emptyset$  as “false”.



## FOL formulas: logical tasks

## Definitions

- **Validity:**  $\varphi$  is **valid** iff for all  $\mathcal{I}$  and  $\alpha$  we have that  $\mathcal{I}, \alpha \models \varphi$ .
- **Satisfiability:**  $\varphi$  is **satisfiable** iff there exists an  $\mathcal{I}$  and  $\alpha$  such that  $\mathcal{I}, \alpha \models \varphi$ , and **unsatisfiable** otherwise.
- **Logical implication:**  $\varphi$  **logically implies**  $\psi$ , written  $\varphi \models \psi$  iff for all  $\mathcal{I}$  and  $\alpha$ , if  $\mathcal{I}, \alpha \models \varphi$  then  $\mathcal{I}, \alpha \models \psi$ .
- **Logical equivalence:**  $\varphi$  is **logically equivalent** to  $\psi$ , iff for all  $\mathcal{I}$  and  $\alpha$ , we have that  $\mathcal{I}, \alpha \models \varphi$  iff  $\mathcal{I}, \alpha \models \psi$  (i.e.,  $\varphi \models \psi$  and  $\psi \models \varphi$ ).



## FOL queries – Logical tasks

- **Validity:** if  $\varphi$  is valid, then  $\varphi^{\mathcal{I}} = \Delta^{\mathcal{I}} \times \dots \times \Delta^{\mathcal{I}}$  for all  $\mathcal{I}$ , i.e., the query always returns all the tuples of  $\mathcal{I}$ .
- **Satisfiability:** if  $\varphi$  is satisfiable, then  $\varphi^{\mathcal{I}} \neq \emptyset$  for some  $\mathcal{I}$ , i.e., the query returns at least one tuple.
- **Logical implication:** if  $\varphi$  logically implies  $\psi$ , then  $\varphi^{\mathcal{I}} \subseteq \psi^{\mathcal{I}}$  for all  $\mathcal{I}$ , written  $\varphi \subseteq \psi$ , i.e., the answer to  $\varphi$  is contained in that of  $\psi$  in every interpretation. This is called **query containment**.
- **Logical equivalence:** if  $\varphi$  is logically equivalent to  $\psi$ , then  $\varphi^{\mathcal{I}} = \psi^{\mathcal{I}}$  for all  $\mathcal{I}$ , written  $\varphi \equiv \psi$ , i.e., the answer to the two queries is the same in every interpretation. This is called **query equivalence** and corresponds to query containment in both directions.

*Note:* These definitions can be extended to the case where we have **axioms**, i.e., **constraints** on the admissible interpretations.



## Chapter II

## First-Order Query Evaluation



## Outline

- 4 Query evaluation problem
- 5 Complexity of query evaluation



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- 4 Query evaluation problem
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## Query evaluation

Let us consider:

- a **finite alphabet**, i.e., we have a finite number of predicates and functions, and
- a **finite interpretation**  $\mathcal{I}$ , i.e., an interpretation (over the finite alphabet) for which  $\Delta^{\mathcal{I}}$  is finite.

Then we can consider query evaluation as an algorithmic problem, and study its computational properties.

*Note:* To study the **computational complexity** of the problem, we need to define a corresponding decision problem.



## Query evaluation problem

## Definitions

- **Query answering problem:** given a finite interpretation  $\mathcal{I}$  and a FOL query  $\varphi(x_1, \dots, x_k)$ , compute

$$\varphi^{\mathcal{I}} = \{(a_1, \dots, a_k) \mid \mathcal{I}, \langle a_1, \dots, a_k \rangle \models \varphi(x_1, \dots, x_k)\}$$

- **Recognition problem (for query answering):** given a finite interpretation  $\mathcal{I}$ , a FOL query  $\varphi(x_1, \dots, x_k)$ , and a tuple  $(a_1, \dots, a_k)$ , with  $a_i \in \Delta^{\mathcal{I}}$ , check whether  $(a_1, \dots, a_k) \in \varphi^{\mathcal{I}}$ , i.e., whether

$$\mathcal{I}, \langle a_1, \dots, a_k \rangle \models \varphi(x_1, \dots, x_k)$$

*Note:* The recognition problem for query answering is the decision problem corresponding to the query answering problem.



## Query evaluation algorithm

We define now an algorithm that computes the function  $\text{Truth}(\mathcal{I}, \alpha, \varphi)$  in such a way that  $\text{Truth}(\mathcal{I}, \alpha, \varphi) = \text{true}$  iff  $\mathcal{I}, \alpha \models \varphi$ .

We make use of an auxiliary function  $\text{TermEval}(\mathcal{I}, \alpha, t)$  that, given an interpretation  $\mathcal{I}$  and an assignment  $\alpha$ , evaluates a term  $t$  returning an object  $o \in \Delta^{\mathcal{I}}$ :

```

 $\Delta^{\mathcal{I}}$  TermEval( $\mathcal{I}, \alpha, t$ ) {
  if ( $t$  is  $x \in \text{Vars}$ )
    return  $\alpha(x)$ ;
  if ( $t$  is  $f(t_1, \dots, t_k)$ )
    return  $f^{\mathcal{I}}(\text{TermEval}(\mathcal{I}, \alpha, t_1), \dots, \text{TermEval}(\mathcal{I}, \alpha, t_k))$ ;
}
```

Then,  $\text{Truth}(\mathcal{I}, \alpha, \varphi)$  can be defined by structural recursion on  $\varphi$ .



## Query evaluation algorithm (cont'd)

```

boolean Truth( $\mathcal{I}, \alpha, \varphi$ ) {
  if ( $\varphi$  is  $t_1 = t_2$ )
    return TermEval( $\mathcal{I}, \alpha, t_1$ ) = TermEval( $\mathcal{I}, \alpha, t_2$ );
  if ( $\varphi$  is  $P(t_1, \dots, t_k)$ )
    return  $P^{\mathcal{I}}$ (TermEval( $\mathcal{I}, \alpha, t_1$ ), ..., TermEval( $\mathcal{I}, \alpha, t_k$ ));
  if ( $\varphi$  is  $\neg\psi$ )
    return  $\neg$ Truth( $\mathcal{I}, \alpha, \psi$ );
  if ( $\varphi$  is  $\psi \circ \psi'$ )
    return Truth( $\mathcal{I}, \alpha, \psi$ )  $\circ$  Truth( $\mathcal{I}, \alpha, \psi'$ );
  if ( $\varphi$  is  $\exists x.\psi$ ) {
    boolean b = false;
    for all ( $a \in \Delta^{\mathcal{I}}$ )
      b = b  $\vee$  Truth( $\mathcal{I}, \alpha[x \mapsto a], \psi$ );
    return b;
  }
  if ( $\varphi$  is  $\forall x.\psi$ ) {
    boolean b = true;
    for all ( $a \in \Delta^{\mathcal{I}}$ )
      b = b  $\wedge$  Truth( $\mathcal{I}, \alpha[x \mapsto a], \psi$ );
    return b;
  }
}

```



## Query evaluation – Results

Theorem (Termination of Truth( $\mathcal{I}, \alpha, \varphi$ ))

The algorithm Truth terminates.

*Proof.* Immediate. □

## Theorem (Correctness)

The algorithm Truth is sound and complete, i.e.,  $\mathcal{I}, \alpha \models \varphi$  if and only if Truth( $\mathcal{I}, \alpha, \varphi$ ) = true.

*Proof.* Easy, since the algorithm is very close to the semantic definition of  $\mathcal{I}, \alpha \models \varphi$ . □



## Outline

- 4 Query evaluation problem
- 5 Complexity of query evaluation



## Query evaluation – Time complexity I

Theorem (Time complexity of Truth( $\mathcal{I}, \alpha, \varphi$ ))

The time complexity of Truth( $\mathcal{I}, \alpha, \varphi$ ) is  $(|\mathcal{I}| + |\alpha| + |\varphi|)^{|\varphi|}$ , i.e., polynomial in the size of  $\mathcal{I}$  and exponential in the size of  $\varphi$ .

*Proof.*

- $f^{\mathcal{I}}$  (of arity  $k$ ) can be represented as  $k$ -dimensional array, hence accessing the required element can be done in time linear in  $|\mathcal{I}|$ .
- TermEval(...) visits the term, so it generates a polynomial number of recursive calls, hence is time polynomial in  $(|\mathcal{I}| + |\alpha| + |\varphi|)$ .



## Query evaluation – Time complexity II

- $P^{\mathcal{I}}$  (of arity  $k$ ) can be represented as  $k$ -dimensional boolean array, hence accessing the required element can be done in time linear in  $|\mathcal{I}|$ .
- $\text{Truth}(\dots)$  for the boolean cases simply visits the formula, so generates either one or two recursive calls.
- $\text{Truth}(\dots)$  for the quantified cases  $\exists x.\varphi$  and  $\forall x.\psi$  involves looping for all elements in  $\Delta^{\mathcal{I}}$  and testing the resulting assignments.
- The total number of such testings is  $O(|\mathcal{I}|^{\#Vars})$ .

Hence the claim holds.  $\square$



## Query evaluation – Space complexity I

Theorem (Space complexity of  $\text{Truth}(\mathcal{I}, \alpha, \varphi)$ )

The space complexity of  $\text{Truth}(\mathcal{I}, \alpha, \varphi)$  is  $|\varphi| \cdot (|\varphi| \cdot \log |\mathcal{I}|)$ , i.e., logarithmic in the size of  $\mathcal{I}$  and polynomial in the size of  $\varphi$ .

## Proof.

- $f^{\mathcal{I}}(\dots)$  can be represented as  $k$ -dimensional array, hence accessing the required element requires  $O(\log |\mathcal{I}|)$ ;
- $\text{TermEval}(\dots)$  simply visits the term, so it generates a polynomial number of recursive calls. Each activation record has a constant size, and we need  $O(|\varphi|)$  activation records;
- $P^{\mathcal{I}}(\dots)$  can be represented as  $k$ -dimensional boolean array, hence accessing the required element requires  $O(\log |\mathcal{I}|)$ ;



## Query evaluation – Space complexity II

- $\text{Truth}(\dots)$  for the boolean cases simply visits the formula, so generates either one or two recursive calls, each requiring constant size;
- $\text{Truth}(\dots)$  for the quantified cases  $\exists x.\varphi$  and  $\forall x.\psi$  involves looping for all elements in  $\Delta^{\mathcal{I}}$  and testing the resulting assignments;
- The total number of activation records that need to be at the same time on the stack is  $O(\#Vars) \leq O(|\varphi|)$ .

Hence the claim holds.  $\square$

*Note:* the worst case form for the formula is

$$\forall x_1. \exists x_2. \dots \forall x_{n-1}. \exists x_n. P(x_1, x_2, \dots, x_{n-1}, x_n).$$



## Query evaluation – Complexity measures [Var82]

## Definition (Combined complexity)

The **combined complexity** is the complexity of  $\{\langle \mathcal{I}, \alpha, \varphi \rangle \mid \mathcal{I}, \alpha \models \varphi\}$ , i.e., interpretation, tuple, and query are all considered part of the input.

## Definition (Data complexity)

The **data complexity** is the complexity of  $\{\langle \mathcal{I}, \alpha \rangle \mid \mathcal{I}, \alpha \models \varphi\}$ , i.e., the query  $\varphi$  is fixed (and hence not considered part of the input).

## Definition (Query complexity)

The **query complexity** is the complexity of  $\{\langle \alpha, \varphi \rangle \mid \mathcal{I}, \alpha \models \varphi\}$ , i.e., the interpretation  $\mathcal{I}$  is fixed (and hence not considered part of the input).





## Query evaluation – Combined, data, query complexity

### Theorem (Combined complexity of query evaluation)

The complexity of  $\{\langle \mathcal{I}, \alpha, \varphi \rangle \mid \mathcal{I}, \alpha \models \varphi\}$  is:

- time: exponential
- space: PSPACE-complete — see [Var82] for hardness

### Theorem (Data complexity of query evaluation)

The complexity of  $\{\langle \mathcal{I}, \alpha \rangle \mid \mathcal{I}, \alpha \models \varphi\}$  is:

- time: polynomial
- space: in LOGSPACE

### Theorem (Query complexity of query evaluation)

The complexity of  $\{\langle \alpha, \varphi \rangle \mid \mathcal{I}, \alpha \models \varphi\}$  is:

- time: exponential
- space: PSPACE-complete — see [Var82] for hardness

## Chapter III

## Conjunctive Queries

## Outline

- 6 Evaluation of conjunctive queries
- 7 Containment of conjunctive queries
- 8 Unions of conjunctive queries

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## Conjunctive queries (CQs)

Def.: A **conjunctive query (CQ)** is a FOL query of the form

$$\exists \vec{y}. conj(\vec{x}, \vec{y})$$

where  $conj(\vec{x}, \vec{y})$  is a conjunction (i.e., an “and”) of atoms and equalities, over the free variables  $\vec{x}$ , the existentially quantified variables  $\vec{y}$ , and possibly constants.

**Note:**

- CQs contain no disjunction, no negation, no universal quantification, and no function symbols besides constants.
- Hence, they correspond to relational algebra **select-project-join (SPJ) queries**.
- CQs are the most frequently asked queries.



## Conjunctive queries and SQL – Example

Relational alphabet:

Person(name, age), Lives(person, city), Manages(boss, employee)

Query: return name and age of all persons that live in the same city as their boss.

Expressed in SQL:

```
SELECT P.name, P.age
FROM Person P, Manages M, Lives L1, Lives L2
WHERE P.name = L1.person AND P.name = M.employee AND
      M.boss = L2.person AND L1.city = L2.city
```

Expressed as a CQ:

$$\exists b, e, p_1, c_1, p_2, c_2. Person(n, a) \wedge Manages(b, e) \wedge Lives(p_1, c_1) \wedge Lives(p_2, c_2) \wedge n = p_1 \wedge n = e \wedge b = p_2 \wedge c_1 = c_2$$

Or simpler:  $\exists b, c. Person(n, a) \wedge Manages(b, n) \wedge Lives(n, c) \wedge Lives(b, c)$



## Datalog notation for CQs

A CQ  $q = \exists \vec{y}. conj(\vec{x}, \vec{y})$  can also be written using **datalog notation** as

$$q(\vec{x}_1) \leftarrow conj'(\vec{x}_1, \vec{y}_1)$$

where  $conj'(\vec{x}_1, \vec{y}_1)$  is the list of atoms in  $conj(\vec{x}, \vec{y})$  obtained by equating the variables  $\vec{x}, \vec{y}$  according to the equalities in  $conj(\vec{x}, \vec{y})$ .

As a result of such an equality elimination, we have that  $\vec{x}_1$  and  $\vec{y}_1$  can contain constants and multiple occurrences of the same variable.

Def.: In the above query  $q$ , we call:

- $q(\vec{x}_1)$  the **head**;
- $conj'(\vec{x}_1, \vec{y}_1)$  the **body**;
- the variables in  $\vec{x}_1$  the **distinguished variables**;
- the variables in  $\vec{y}_1$  the **non-distinguished variables**.

z.it

## Conjunctive queries – Example

- Consider an **interpretation**  $\mathcal{I} = (\Delta^{\mathcal{I}}, E^{\mathcal{I}})$ , where  $E^{\mathcal{I}}$  is a binary relation – note that such interpretation is a (directed) graph.
- The following **CQ**  $q$  returns all nodes that participate to a triangle in the graph:

$$\exists y, z. E(x, y) \wedge E(y, z) \wedge E(z, x)$$

- The query  $q$  in **datalog notation** becomes:

$$q(x) \leftarrow E(x, y), E(y, z), E(z, x)$$

- The query  $q$  in **SQL** is (we use Edge(f, s) for  $E(x, y)$ ):

```
SELECT E1.f
FROM Edge E1, Edge E2, Edge E3
WHERE E1.s = E2.f AND E2.s = E3.f AND E3.s = E1.f
```



## Nondeterministic evaluation of CQs

Since a CQ contains only existential quantifications, we can evaluate it by:

- 1 guessing a truth assignment for the non-distinguished variables;
- 2 evaluating the resulting formula (that has no quantifications).

```
boolean ConjTruth( $\mathcal{I}, \alpha, \exists \vec{y}. conj(\vec{x}, \vec{y})$ ) {
  GUESS assignment  $\alpha[\vec{y} \mapsto \vec{a}]$  {
    return Truth( $\mathcal{I}, \alpha[\vec{y} \mapsto \vec{a}], conj(\vec{x}, \vec{y})$ );
  }
}
```

where  $Truth(\mathcal{I}, \alpha, \varphi)$  is defined as for FOL queries, considering only the required cases.



## Nondeterministic CQ evaluation algorithm

```
boolean Truth( $\mathcal{I}, \alpha, \varphi$ ) {
  if ( $\varphi$  is  $t_1 = t_2$ )
    return TermEval( $\mathcal{I}, \alpha, t_1$ ) = TermEval( $\mathcal{I}, \alpha, t_2$ );
  if ( $\varphi$  is  $P(t_1, \dots, t_k)$ )
    return  $P^{\mathcal{I}}$ (TermEval( $\mathcal{I}, \alpha, t_1$ ), ..., TermEval( $\mathcal{I}, \alpha, t_k$ ));
  if ( $\varphi$  is  $\psi \wedge \psi'$ )
    return Truth( $\mathcal{I}, \alpha, \psi$ )  $\wedge$  Truth( $\mathcal{I}, \alpha, \psi'$ );
}

 $\Delta^{\mathcal{I}}$  TermEval( $\mathcal{I}, \alpha, t$ ) {
  if ( $t$  is a variable  $x$ ) return  $\alpha(x)$ ;
  if ( $t$  is a constant  $c$ ) return  $c^{\mathcal{I}}$ ;
}
```



## CQ evaluation – Combined, data, and query complexity

### Theorem (Combined complexity of CQ evaluation)

$\{\langle \mathcal{I}, \alpha, q \rangle \mid \mathcal{I}, \alpha \models q\}$  is **NP-complete** — see below for hardness.

- time: exponential
- space: polynomial

### Theorem (Data complexity of CQ evaluation)

$\{\langle \mathcal{I}, \alpha \rangle \mid \mathcal{I}, \alpha \models q\}$  is in **LOGSPACE**

- time: polynomial
- space: logarithmic

### Theorem (Query complexity of CQ evaluation)

$\{\langle \alpha, q \rangle \mid \mathcal{I}, \alpha \models q\}$  is **NP-complete** — see below for hardness.

- time: exponential
- space: polynomial



## 3-colorability

A graph is  **$k$ -colorable** if it is possible to assign to each node one of  $k$  colors in such a way that every two nodes connected by an edge have different colors.

Def.: **3-colorability** is the following decision problem

Given a graph  $G = (V, E)$ , is it 3-colorable?

### Theorem

3-colorability is NP-complete.

We exploit 3-colorability to show NP-hardness of conjunctive query evaluation.



## Reduction from 3-colorability to CQ evaluation

Let  $G = (V, E)$  be a graph. We define:

- An **Interpretation**:  $\mathcal{I} = (\Delta^{\mathcal{I}}, E^{\mathcal{I}})$  where:
  - $\Delta^{\mathcal{I}} = \{r, g, b\}$
  - $E^{\mathcal{I}} = \{(r, g), (g, r), (r, b), (b, r), (g, b), (b, g)\}$
- A **conjunctive query**: Let  $V = \{x_1, \dots, x_n\}$ , then consider the boolean conjunctive query defined as:

$$q_G = \exists x_1, \dots, x_n. \bigwedge_{(x_i, x_j) \in E} E(x_i, x_j) \wedge E(x_j, x_i)$$

## Theorem

$G$  is 3-colorable iff  $\mathcal{I} \models q_G$ .



## NP-hardness of CQ evaluation

The previous reduction immediately gives us the hardness for combined complexity.

## Theorem

**CQ evaluation is NP-hard** in combined complexity.

*Note:* in the previous reduction, the interpretation does not depend on the actual graph. Hence, the reduction provides also the lower-bound for query complexity.

## Theorem

**CQ evaluation is NP-hard** in query (and combined) complexity.



## Outline

- 6 Evaluation of conjunctive queries
- 7 **Containment of conjunctive queries**
- 8 Unions of conjunctive queries



## Homomorphism

Let  $\mathcal{I} = (\Delta^{\mathcal{I}}, P^{\mathcal{I}}, \dots, c^{\mathcal{I}}, \dots)$  and  $\mathcal{J} = (\Delta^{\mathcal{J}}, P^{\mathcal{J}}, \dots, c^{\mathcal{J}}, \dots)$  be two interpretations over the same alphabet (for simplicity, we consider only constants as functions).

**Def.:** A **homomorphism** from  $\mathcal{I}$  to  $\mathcal{J}$

is a mapping  $h : \Delta^{\mathcal{I}} \rightarrow \Delta^{\mathcal{J}}$  such that:

- $h(c^{\mathcal{I}}) = c^{\mathcal{J}}$
- $h(P^{\mathcal{I}}(a_1, \dots, a_k)) = P^{\mathcal{J}}(h(a_1), \dots, h(a_k))$

*Note:* An **isomorphism** is a homomorphism that is one-to-one and onto.

## Theorem

FOL is unable to distinguish between interpretations that are isomorphic.

*Proof.* See any standard book on logic.  $\square$



## Recognition problem and boolean query evaluation

Consider the recognition problem associated to the evaluation of a query  $q$  of arity  $k$ . Then

$$\mathcal{I}, \alpha \models q(x_1, \dots, x_k) \quad \text{iff} \quad \mathcal{I}_{\alpha, \vec{c}} \models q(c_1, \dots, c_k)$$

where  $\mathcal{I}_{\alpha, \vec{c}}$  is identical to  $\mathcal{I}$  but includes new constants  $c_1, \dots, c_k$  that are interpreted as  $c_i^{\mathcal{I}_{\alpha, \vec{c}}} = \alpha(x_i)$ .

That is, we can **reduce the recognition problem to the evaluation of a boolean query**.



## Canonical interpretation of a (boolean) CQ

Let  $q$  be a conjunctive query  $\exists x_1, \dots, x_n \cdot \text{conj}$

Def.: The **canonical interpretation**  $\mathcal{I}_q$  associated with  $q$

is the interpretation  $\mathcal{I}_q = (\Delta^{\mathcal{I}_q}, P^{\mathcal{I}_q}, \dots, c^{\mathcal{I}_q}, \dots)$ , where

- $\Delta^{\mathcal{I}_q} = \{x_1, \dots, x_n\} \cup \{c \mid c \text{ constant occurring in } q\}$ ,  
i.e., all the variables and constants in  $q$ ;
- $c^{\mathcal{I}_q} = c$ , for each constant  $c$  in  $q$ ;
- $(t_1, \dots, t_k) \in P^{\mathcal{I}_q}$  iff the atom  $P(t_1, \dots, t_k)$  occurs in  $q$ .

Sometimes the procedure for obtaining the canonical interpretation is called **freezing** of  $q$ .



## Canonical interpretation of a (boolean) CQ – Example

Consider the boolean query  $q$

$$q(c) \leftarrow E(c, y), E(y, z), E(z, c)$$

Then, the canonical interpretation  $\mathcal{I}_q$  is defined as

$$\mathcal{I}_q = (\Delta^{\mathcal{I}_q}, E^{\mathcal{I}_q}, c^{\mathcal{I}_q})$$

where

- $\Delta^{\mathcal{I}_q} = \{y, z, c\}$
- $E^{\mathcal{I}_q} = \{(c, y), (y, z), (z, c)\}$
- $c^{\mathcal{I}_q} = c$



## Canonical interpretation and (boolean) CQ evaluation

**Theorem ([CM77])**

For boolean CQs,  $\mathcal{I} \models q$  iff there exists a homomorphism from  $\mathcal{I}_q$  to  $\mathcal{I}$ .

*Proof.*

“ $\Rightarrow$ ” Let  $\mathcal{I} \models q$ , let  $\alpha$  be an assignment to the existential variables that makes  $q$  true in  $\mathcal{I}$ , and let  $\hat{\alpha}$  be its extension to constants. Then  $\hat{\alpha}$  is a homomorphism from  $\mathcal{I}_q$  to  $\mathcal{I}$ .

“ $\Leftarrow$ ” Let  $h$  be a homomorphism from  $\mathcal{I}_q$  to  $\mathcal{I}$ . Then restricting  $h$  to the variables only we obtain an assignment to the existential variables that makes  $q$  true in  $\mathcal{I}$ .  $\square$



## Canonical interpretation and (boolean) CQ evaluation

The previous result can be rephrased as follows:

(The recognition problem associated to) **query evaluation can be reduced to finding a homomorphism.**

Finding a homomorphism between two interpretations (aka relational structures) is also known as solving a **Constraint Satisfaction Problem (CSP)**, a problem well-studied in AI – see also [KV98].



## Query containment

### Def.: Query containment

Given two FOL queries  $\varphi$  and  $\psi$  of the same arity,  $\varphi$  is **contained in**  $\psi$ , denoted  $\varphi \subseteq \psi$ , if for all interpretations  $\mathcal{I}$  and all assignments  $\alpha$  we have that

$$\mathcal{I}, \alpha \models \varphi \text{ implies } \mathcal{I}, \alpha \models \psi$$

(In logical terms:  $\varphi \models \psi$ .)

*Note:* Query containment is of special interest in query optimization.

### Theorem

For FOL queries, query containment is undecidable.

*Proof.:* Reduction from FOL logical implication.  $\square$



## Query containment for CQs

For CQs, query containment  $q_1(\vec{x}) \subseteq q_2(\vec{x})$  can be reduced to query evaluation.

- 1 **Freeze the free variables**, i.e., consider them as constants. This is possible, since  $q_1(\vec{x}) \subseteq q_2(\vec{x})$  iff
  - $\mathcal{I}, \alpha \models q_1(\vec{x})$  implies  $\mathcal{I}, \alpha \models q_2(\vec{x})$ , for all  $\mathcal{I}$  and  $\alpha$ ; or equivalently
  - $\mathcal{I}_{\alpha, \vec{c}} \models q_1(\vec{c})$  implies  $\mathcal{I}_{\alpha, \vec{c}} \models q_2(\vec{c})$ , for all  $\mathcal{I}_{\alpha, \vec{c}}$ , where  $\vec{c}$  are new constants, and  $\mathcal{I}_{\alpha, \vec{c}}$  extends  $\mathcal{I}$  to the new constants with  $c^{\mathcal{I}_{\alpha, \vec{c}}} = \alpha(x)$ .
- 2 **Construct the canonical interpretation  $\mathcal{I}_{q_1(\vec{c})}$  of the CQ  $q_1(\vec{c})$  on the left hand side** ...
- 3 ... and **evaluate on  $\mathcal{I}_{q_1(\vec{c})}$  the CQ  $q_2(\vec{c})$  on the right hand side**, i.e., check whether  $\mathcal{I}_{q_1(\vec{c})} \models q_2(\vec{c})$ .



## Reducing containment of CQs to CQ evaluation

### Theorem ([CM77])

For CQs,  $q_1(\vec{x}) \subseteq q_2(\vec{x})$  iff  $\mathcal{I}_{q_1(\vec{c})} \models q_2(\vec{c})$ , where  $\vec{c}$  are new constants.

### Proof.

" $\Rightarrow$ " Assume that  $q_1(\vec{x}) \subseteq q_2(\vec{x})$ .

- Since  $\mathcal{I}_{q_1(\vec{c})} \models q_1(\vec{c})$  it follows that  $\mathcal{I}_{q_1(\vec{c})} \models q_2(\vec{c})$ .

" $\Leftarrow$ " Assume that  $\mathcal{I}_{q_1(\vec{c})} \models q_2(\vec{c})$ .

- By [CM77] on hom., for every  $\mathcal{I}$  such that  $\mathcal{I} \models q_1(\vec{c})$  there exists a homomorphism  $h$  from  $\mathcal{I}_{q_1(\vec{c})}$  to  $\mathcal{I}$ .
- On the other hand, since  $\mathcal{I}_{q_1(\vec{c})} \models q_2(\vec{c})$ , again by [CM77] on hom., there exists a homomorphism  $h'$  from  $\mathcal{I}_{q_2(\vec{c})}$  to  $\mathcal{I}_{q_1(\vec{c})}$ .
- The mapping  $h \circ h'$  (obtained by composing  $h$  and  $h'$ ) is a homomorphism from  $\mathcal{I}_{q_2(\vec{c})}$  to  $\mathcal{I}$ . Hence, once again by [CM77] on hom.,  $\mathcal{I} \models q_2(\vec{c})$ .

So we can conclude that  $q_1(\vec{c}) \subseteq q_2(\vec{c})$ , and hence  $q_1(\vec{x}) \subseteq q_2(\vec{x})$ .  $\square$



## Query containment for CQs

For CQs, we also have that (boolean) query evaluation  $\mathcal{I} \models q$  can be reduced to query containment.

Let  $\mathcal{I} = (\Delta^{\mathcal{I}}, P^{\mathcal{I}}, \dots, c^{\mathcal{I}}, \dots)$ .

We construct the (boolean) CQ  $q_{\mathcal{I}}$  as follows:

- $q_{\mathcal{I}}$  has no existential variables (hence no variables at all);
- the constants in  $q_{\mathcal{I}}$  are the elements of  $\Delta^{\mathcal{I}}$ ;
- for each relation  $P$  interpreted in  $\mathcal{I}$  and for each fact  $(a_1, \dots, a_k) \in P^{\mathcal{I}}$ ,  $q_{\mathcal{I}}$  contains one atom  $P(a_1, \dots, a_k)$  (note that each  $a_i \in \Delta^{\mathcal{I}}$  is a constant in  $q_{\mathcal{I}}$ ).

### Theorem

For CQs,  $\mathcal{I} \models q$  iff  $q_{\mathcal{I}} \subseteq q$ .



## Query containment for CQs – Complexity

From the previous results and NP-completeness of combined complexity of CQ evaluation, we immediately get:

### Theorem

Containment of CQs is NP-complete.

Since CQ evaluation is NP-complete even in query complexity, the above result can be strengthened:

### Theorem

Containment  $q_1(\vec{x}) \subseteq q_2(\vec{x})$  of CQs is NP-complete, even when  $q_1$  is considered fixed.



## Outline

- 6 Evaluation of conjunctive queries
- 7 Containment of conjunctive queries
- 8 Unions of conjunctive queries



## Union of conjunctive queries (UCQs)

Def.: A **union of conjunctive queries (UCQ)** is a FOL query of the form

$$\bigvee_{i=1, \dots, n} \exists \vec{y}_i. \text{conj}_i(\vec{x}, \vec{y}_i)$$

where each  $\text{conj}_i(\vec{x}, \vec{y}_i)$  is a conjunction of atoms and equalities with free variables  $\vec{x}$  and  $\vec{y}_i$ , and possibly constants.

*Note:* Obviously, each conjunctive query is also a union of conjunctive queries.



## Datalog notation for UCQs

A union of conjunctive queries

$$q = \bigvee_{i=1, \dots, n} \exists \vec{y}_i. \text{conj}_i(\vec{x}, \vec{y}_i)$$

is written in **datalog notation** as

$$\left\{ \begin{array}{l} q(\vec{x}) \leftarrow \text{conj}'_1(\vec{x}, \vec{y}'_1) \\ \vdots \\ q(\vec{x}) \leftarrow \text{conj}'_n(\vec{x}, \vec{y}'_n) \end{array} \right\}$$

where each element of the set is the datalog expression corresponding to the conjunctive query  $q_i = \exists \vec{y}_i. \text{conj}_i(\vec{x}, \vec{y}_i)$ .

*Note:* in general, we omit the set brackets.



## Evaluation of UCQs

From the definition of FOL query we have that:

$$\mathcal{I}, \alpha \models \bigvee_{i=1, \dots, n} \exists \vec{y}_i. \text{conj}_i(\vec{x}, \vec{y}_i)$$

if and only if

$$\mathcal{I}, \alpha \models \exists \vec{y}_i. \text{conj}_i(\vec{x}, \vec{y}_i) \quad \text{for some } i \in \{1, \dots, n\}.$$

Hence to evaluate a UCQ  $q$ , we simply evaluate a number (linear in the size of  $q$ ) of conjunctive queries in isolation.

Hence, **evaluating UCQs has the same complexity as evaluating CQs.**



## UCQ evaluation – Combined, data, and query complexity

**Theorem (Combined complexity of UCQ evaluation)**

$\{\langle \mathcal{I}, \alpha, q \rangle \mid \mathcal{I}, \alpha \models q\}$  is **NP-complete**.

- time: exponential
- space: polynomial

**Theorem (Data complexity of UCQ evaluation)**

$\{\langle \mathcal{I}, q \rangle \mid \mathcal{I}, \alpha \models q\}$  is in **LOGSPACE** (query  $q$  fixed).

- time: polynomial
- space: logarithmic

**Theorem (Query complexity of UCQ evaluation)**

$\{\langle \alpha, q \rangle \mid \mathcal{I}, \alpha \models q\}$  is **NP-complete** (interpretation  $\mathcal{I}$  fixed).

- time: exponential
- space: polynomial

## Query containment for UCQs

**Theorem**

For UCQs,  $\{q_1, \dots, q_k\} \subseteq \{q'_1, \dots, q'_n\}$  iff for each  $q_i$  there is a  $q'_j$  such that  $q_i \subseteq q'_j$ .

*Proof.*

“ $\Leftarrow$ ” Obvious.

“ $\Rightarrow$ ” If the containment holds, then we have

$\{q_1(\vec{c}), \dots, q_k(\vec{c})\} \subseteq \{q'_1(\vec{c}), \dots, q'_n(\vec{c})\}$ , where  $\vec{c}$  are new constants:

- Now consider  $\mathcal{I}_{q_i(\vec{c})}$ . We have  $\mathcal{I}_{q_i(\vec{c})} \models q_i(\vec{c})$ , and hence  $\mathcal{I}_{q_i(\vec{c})} \models \{q_1(\vec{c}), \dots, q_k(\vec{c})\}$ .
- By the containment, we have that  $\mathcal{I}_{q_i(\vec{c})} \models \{q'_1(\vec{c}), \dots, q'_n(\vec{c})\}$ . I.e., there exists a  $q'_j(\vec{c})$  such that  $\mathcal{I}_{q_i(\vec{c})} \models q'_j(\vec{c})$ .
- Hence, by [CM77] on containment of CQs, we have that  $q_i \subseteq q'_j$ .

□





## Query containment for UCQs – Complexity

From the previous result, we have that we can check  $\{q_1, \dots, q_k\} \subseteq \{q'_1, \dots, q'_n\}$  by at most  $k \cdot n$  CQ containment checks.

We immediately get:

### Theorem

Containment of UCQs is NP-complete.



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