

# FOUNDATIONS OF DATABASES AND QUERY LANGUAGES

### Lecture 4: Complexity of FO Query Answering

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TU Dresden, 4 May 2015

## Overview

- 1. Introduction | Relational data model
- 2. First-order queries
- 3. Complexity of query answering
- 4. Complexity of FO query answering
- 5. Query optimization
- 6. Conjunctive queries
- 7. Limits of first-order query expressiveness
- 8. Introduction to Datalog
- 9. Implementation techniques for Datalog
- 10. Path queries
- 11. Constraints (1)
- 12. Constraints (2)
- 13. "Buffer time"
- 14. Outlook: database theory in practice

## How to Measure Query Answering Complexity

Query answering as decision problem  $\rightsquigarrow$  consider Boolean queries

Various notions of complexity:

- Combined complexity (complexity w.r.t. size of query and database instance)
- Data complexity (worst case complexity for any fixed query)
- Query complexity (worst case complexity for any fixed database instance)

Various common complexity classes:

### $L \subseteq NL \subseteq P \subseteq NP \subseteq PSPACE \subseteq ExPTIME$

# An Algorithm for Evaluating FO Queries

function  $\mathsf{Eval}(\varphi, I)$ 

- 01 switch  $(\varphi)$  {
- 02 **case**  $p(c_1, \ldots, c_n)$ : return  $\langle c_1, \ldots, c_n \rangle \in p^I$
- 03 **case**  $\neg \psi$ : return  $\neg$ Eval( $\psi$ , I)
- 04 **case**  $\psi_1 \land \psi_2$ : return Eval( $\psi_1, I$ )  $\land$  Eval( $\psi_2, I$ )
- 05 **case**  $\exists x.\psi$ :
- $\mathbf{106} \qquad \mathbf{for} \ c \in \Delta^{\mathcal{I}} \ \{$
- 07 **if**  $Eval(\psi[x \mapsto c], I)$  **then** return **true**
- 08 }
- 09 return **false**

10 }

## FO Algorithm Worst-Case Runtime

Let *m* be the size of  $\varphi$ , and let  $n = |\mathcal{I}|$  (total table sizes)

- How many recursive calls of Eval are there?
  → one per subexpression: at most *m*
- Maximum depth of recursion?
  >> bounded by total number of calls: at most m
- Maximum number of iterations of for loop?
  → |Δ<sup>I</sup>| ≤ n per recursion level
  → at most n<sup>m</sup> iterations
- Checking  $\langle c_1, \ldots, c_n \rangle \in p^I$  can be done in linear time w.r.t. n

Runtime in  $m \cdot n^m \cdot n = m \cdot n^{m+1}$ 

# Time Complexity of FO Algorithm

Let *m* be the size of  $\varphi$ , and let  $n = |\mathcal{I}|$  (total table sizes)

Runtime in  $m \cdot n^{m+1}$ 

Time complexity of FO query evaluation

- Combined complexity: in EXPTIME
- Data complexity (*m* is constant): in P
- Query complexity (*n* is constant): in EXPTIME

FO Algorithm Worst-Case Memory Usage

We can get better complexity bounds by looking at memory

Let *m* be the size of  $\varphi$ , and let  $n = |\mathcal{I}|$  (total table sizes)

- For each (recursive) call, store pointer to current subexpression of φ: log m
- For each variable in φ (at most m), store current constant assignment (as a pointer): m · log n
- Checking  $\langle c_1, \ldots, c_n \rangle \in p^I$  can be done in logarithmic space w.r.t. *n*

Memory in  $m \log m + m \log n + \log n = m \log m + (m + 1) \log n$ 

## Space Complexity of FO Algorithm

Let *m* be the size of  $\varphi$ , and let  $n = |\mathcal{I}|$  (total table sizes)

Memory in  $m \log m + (m + 1) \log n$ 

Space complexity of FO query evaluation

- Combined complexity: in  $\operatorname{PSPACE}$
- Data complexity (*m* is constant): in L
- Query complexity (*n* is constant): in **PSPACE**

## FO Combined Complexity

The algorithm shows that FO query evaluation is in  $\ensuremath{\mathrm{PSPACE}}$  . Is this the best we can get?

Hardness proof: reduce a known  $\operatorname{PSpace}$  hard problem to FO query evaluation

## FO Combined Complexity

The algorithm shows that FO query evaluation is in  $\ensuremath{\mathrm{PSPACE}}$  . Is this the best we can get?

Hardness proof: reduce a known  $\rm PSPACE\text{-}hard$  problem to FO query evaluation  $\sim$  QBF satisfiability

Let  $Q_1X_1.Q_2X_2...Q_nX_n.\varphi[X_1,...,X_n]$  be a QBF (with  $Q_i \in \{\forall, \exists\}$ )

- Database instance I with  $\Delta^{I} = \{0, 1\}$
- One table with one row: true(1)
- Transform input QBF into Boolean FO query

 $Q_1 x_1 . Q_2 x_2 . \cdots Q_n x_n . \varphi[X_1 \mapsto \mathsf{true}(x_1), \ldots, X_n \mapsto \mathsf{true}(x_n)]$ 

## $\operatorname{PSpace}\nolimits$ -hardness for DI Queries

The previous reduction from QBF may lead to a query that is not domain independent

Example: QBF  $\exists p.\neg p$  leads to FO query  $\exists x.\neg true(x)$ 

## $\operatorname{PSpace}\nolimits$ -hardness for DI Queries

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Example: QBF  $\exists p.\neg p$  leads to FO query  $\exists x.\neg$ true(x)

Better approach:

- Consider QBF Q<sub>1</sub>X<sub>1</sub>.Q<sub>2</sub>X<sub>2</sub>...Q<sub>n</sub>X<sub>n</sub>.φ[X<sub>1</sub>,...,X<sub>n</sub>] with φ in negation normal form: negations only occur directly before variables X<sub>i</sub> (still PSPACE-complete: exercise)
- Database instance  $\mathcal{I}$  with  $\Delta^{\mathcal{I}} = \{0, 1\}$
- Two tables with one row each: true(1) and false(0)
- Transform input QBF into Boolean FO query

$$\mathsf{O}_1 x_1 \, \mathsf{O}_2 x_2 \, \cdots \, \mathsf{O}_n x_n \, \varphi'$$

where  $\varphi'$  is obtained by replacing each negated variable  $\neg X_i$ with false( $x_i$ ) and each non-negated variable  $X_i$  with true( $x_i$ ).

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# Combined Complexity of FO Query Answering

### Theorem

The evaluation of FO queries is  $\ensuremath{\operatorname{PSPACE}}$  -complete with respect to combined complexity.

We have actually shown something stronger:

### Theorem

The evaluation of FO queries is  $\ensuremath{\operatorname{PSPACE}}$  -complete with respect to query complexity.

## Data Complexity of FO Query Answering

The algorithm showed that FO query evaluation is in  ${\rm L}$   $\rightsquigarrow$  can we do any better?

What could be better than L?

 $? \subseteq L \subseteq NL \subseteq P \subseteq \ldots$ 

→ we need to define circuit complexities first

## **Boolean Circuits**

### Definition

A Boolean circuit is a finite, directed, acyclic graph where

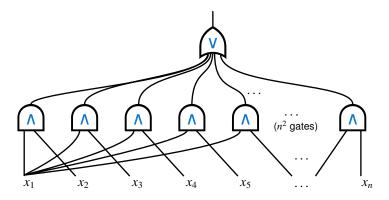
- · each node that has no predecessors is an input node
- each node that is not an input node is one of the following types of logical gate: AND, OR, NOT
- one or more nodes are designated output nodes

ightarrow we will only consider Boolean circuits with exactly one output

 $\rightsquigarrow$  propositional logic formulae are Boolean circuits with one output and gates of fanout  $\leq 1$ 

### Example

A Boolean circuit over an input string  $x_1x_2...x_n$  of length n



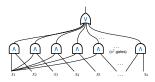
Corresponds to formula  $(x_1 \land x_2) \lor (x_1 \land x_3) \lor \ldots \lor (x_{n-1} \land x_n)$  $\rightsquigarrow$  accepts all strings with at least two 1s

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## Circuits as a Model for Parallel Computation

#### Previous example:



 $\sim n^2$  processors working in parallel  $\sim$  computation finishes in 2 steps

- size: number of gates = total number of computing steps
- depth: longest path of gates = time for parallel computation

ightarrow refinement of polynomial time taking parallelizability into account

# Solving Problems With Circuits

Observation: the input size is "hard-wired" in circuits  $\rightsquigarrow$  each circuit only has a finite number of different inputs  $\rightsquigarrow$  not a computationally interesting problem

How can we solve interesting problems with Boolean circuits?

# Solving Problems With Circuits

Observation: the input size is "hard-wired" in circuits  $\rightsquigarrow$  each circuit only has a finite number of different inputs  $\rightsquigarrow$  not a computationally interesting problem

How can we solve interesting problems with Boolean circuits?

### Definition

A uniform family of Boolean circuits is a set of circuits  $C_n$  ( $n \ge 0$ ) that can be computed from n (usually in logarithmic space or time; we don't discuss the details here).

A language  $\mathcal{L} \subseteq \{0, 1\}^*$  is decided by a uniform family  $(C_n)_{n \ge 0}$  of Boolean circuits if for each word *w* of length |w|:

 $w \in \mathcal{L}$  if and only if  $C_{|w|}(w) = 1$ 

Measuring Complexity with Boolean Circuits

How to measure the computing power of Boolean circuits?

Relevant metrics:

- size of the circuit: overall number of gates (as function of input size)
- depth of the circuit: longest path of gates (as function of input size)
- fan in: two inputs per gate or any number of inputs per gate?

Important classes of circuits: small-depth circuits

### Definition

 $(C_n)_{n\geq 0}$  is a family of small-depth circuits if

- the size of  $C_n$  is polynomial in n,
- the depth of  $C_n$  is poly-logarithmic in *n*, that is,  $O(\log^k n)$ .

# The Complexity Classes $\operatorname{NC}$ and $\operatorname{AC}$

Two important types of small-depth circuits

### Definition

 $\operatorname{NC}^k$  is the class of problems that can be solved by uniform families of circuits  $(C_n)_{n\geq 0}$  of fan-in  $\leq 2$ , size polynomial in n, and depth in  $O(\log^k n)$ .

The class NC is defined as NC =  $\bigcup_{k\geq 0}$  NC<sup>k</sup>.

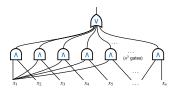
("Nick's Class" named after Nicholas Pippenger by Stephen Cook)

### Definition

 $AC^k$  and AC are defined like  $NC^k$  and NC, respectively, but for circuits with arbitrary fan-in.

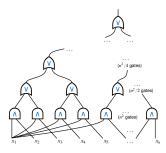
(A is for "Alternating": AND-OR gates alternate in such circuits)

## Example



family of polynomial size, constant depth, arbitrary fan-in circuits  $\sim$  in  $\mathrm{AC}^0$ 

We can eliminate arbitrary fan-ins by using more layers of gates:



family of polynomial size, logarithmic depth, bounded fan-in circuits  $\sim$  in  $\mathrm{NC}^1$ 

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Relationships of Circuit Complexity Classes

The previous sketch can be generalised:

 $\mathbf{NC}^0 \subseteq \mathbf{AC}^0 \subseteq \mathbf{NC}^1 \subseteq \mathbf{AC}^1 \subseteq \ldots \subseteq \mathbf{AC}^k \subseteq \mathbf{NC}^{k+1} \subseteq \ldots$ 

Only few inclusions are known to be proper:  $NC^0 \subset AC^0 \subset NC^1$ Direct consequence of above hierarchy: NC = AC

Interesting relations to other classes:

 $NC^0 \subset AC^0 \subset NC^1 \subseteq L \subseteq NL \subseteq AC^1 \subseteq \ldots \subseteq NC \subseteq P$ 

Intuition:

- $\bullet\,$  Problems in  ${\rm NC}$  are parallelisable
- Problems in  $\mathrm{P} \setminus \mathrm{NC}$  are inherently sequential

### However: it is not known if $NC \neq P$

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### Back to Databases ....

### Theorem

The evaluation of FO queries is complete for (logtime uniform)  ${\rm AC}^0$  with respect to data complexity.

#### Proof:

- Membership: For a fixed Boolean FO query, provide a uniform construction for a small-depth circuit based on the size of a database
- Hardness: Show that circuits can be transformed into Boolean FO queries in logarithmic time (not on a standard TM ... not in this lecture)

# From Query to Circuit

Assumption:

- query and database schema is fixed
- database instance (and thus active domain) are variable

Construct circuit uniformly based on size of active domain

### Sketch of construction:

• one input node for each possible database tuple (over given schema and active domain)

 $\rightsquigarrow$  true or false depending on whether tuple is present or not

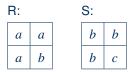
- Recursively, for each subformula, introduce a gate for each possible tuple (instantiation) of this formula
  - $\rightsquigarrow$  true or false depending on whether the subformula holds for this tuple or not
- Logical operators correspond to gate types: basic operators obvious, ∀ as generalised conjunction, ∃ as generalised disjunction
- subformula with n free variables → |adom|<sup>n</sup> gates
  → especially: |adom|<sup>0</sup> = 1 output gate for Boolean query

### Example

### We consider the formula

$$\exists z.(\exists x.\exists y.R(x,y) \land S(y,z)) \land \neg R(a,z)$$

#### Over the database instance:

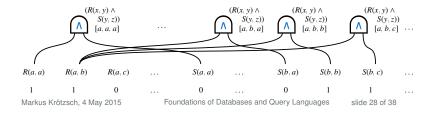


Active domain:  $\{a, b, c\}$ 

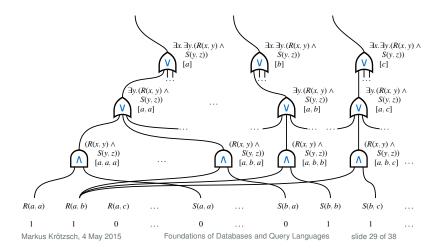
### **Example:** $\exists z.(\exists x.\exists y.R(x,y) \land S(y,z)) \land \neg R(a,z)$

R(a, a)R(a, b)R(a, c)S(a, a)S(b, a)S(b, b)S(b, c)1 1 0 0 0 1 Markus Krötzsch, 4 May 2015 Foundations of Databases and Query Languages slide 27 of 38

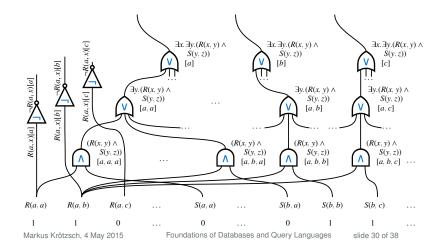
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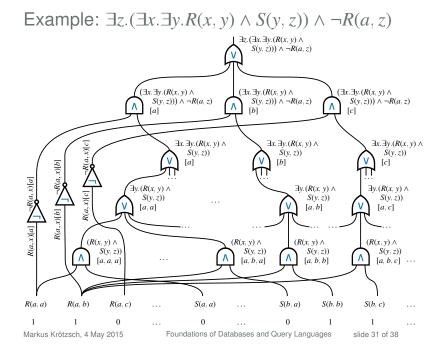


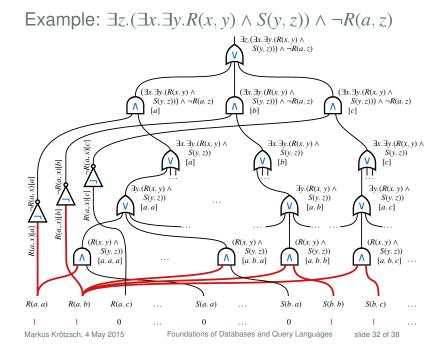
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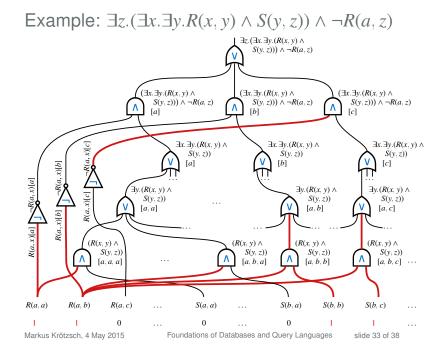


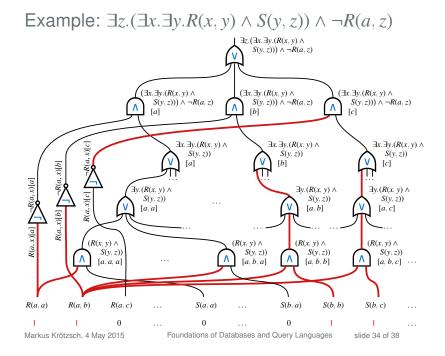
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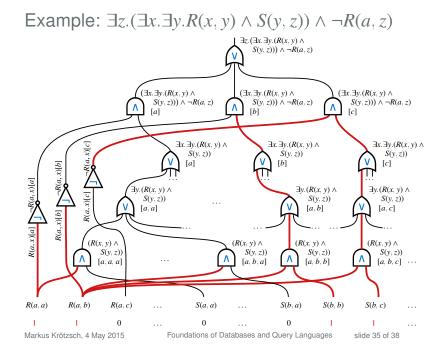


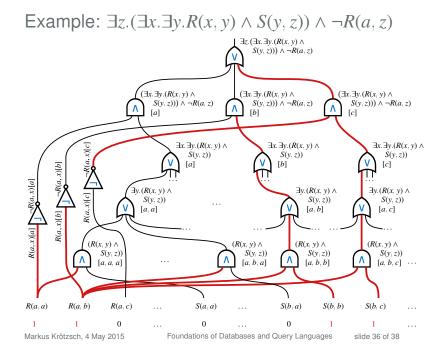


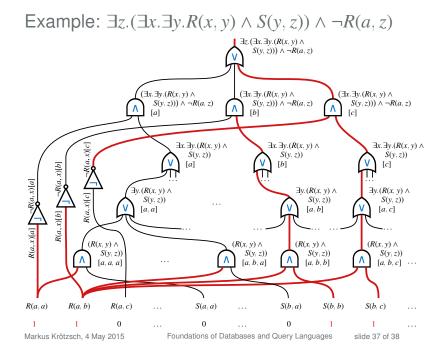












## Summary and Outlook

### The evaluation of FO queries is

- $\operatorname{PSpace}\text{-complete for combined complexity}$
- PSpace-complete for query complexity
- $AC^0$ -complete for data complexity

Circuit complexities help to identify highly parallelisable problems in P

#### Open questions:

- Which other computing problems are interesting? (next lecture)
- Are there query languages with lower complexities?
- How can we study the expressiveness of query languages?