# FOUNDATIONS OF DATABASES AND QUERY LANGUAGES 

Lecture 4: Complexity of FO Query Answering

Markus Krötzsch

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## Overview

1. Introduction | Relational data model
2. First-order queries
3. Complexity of query answering
4. Complexity of FO query answering
5. Query optimization
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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
$\leadsto$ 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:

$$
\mathrm{L} \subseteq \mathrm{NL} \subseteq \mathrm{P} \subseteq \mathrm{NP} \subseteq \mathrm{PSPACE} \subseteq \text { ExpTime }
$$

## An Algorithm for Evaluating FO Queries

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

| 01 | switch $(\varphi)\{$ |
| :--- | :---: |
| 02 | case $p\left(c_{1}, \ldots, c_{n}\right): \operatorname{return}\left\langle c_{1}, \ldots, c_{n}\right\rangle \in p^{I}$ |
| 03 | case $\neg \psi:$ return $\neg \operatorname{Eval}(\psi, \mathcal{I})$ |
| 04 | case $\psi_{1} \wedge \psi_{2}:$ return $\operatorname{Eval}\left(\psi_{1}, \mathcal{I}\right) \wedge \operatorname{Eval}\left(\psi_{2}, \mathcal{I}\right)$ |
| 05 | case $\exists x . \psi:$ |
| 06 | for $c \in \Delta^{I}\{$ |
| 07 | $\quad$ if $\operatorname{Eval}(\psi[x \mapsto c], \mathcal{I})$ then return true |
| 08 | $\}$ |
| 09 | return false |
| 10 | $\}$ |

## FO Algorithm Worst-Case Runtime

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

- How many recursive calls of Eval are there?
$\leadsto$ one per subexpression: at most $m$
- Maximum depth of recursion?
$\leadsto$ bounded by total number of calls: at most $m$
- Maximum number of iterations of for loop?
$\leadsto\left|\Delta^{I}\right| \leq n$ per recursion level
$\leadsto$ at most $n^{m}$ iterations
- Checking $\left\langle c_{1}, \ldots, c_{n}\right\rangle \in p^{I}$ 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 $n=|I|$ (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 $\varphi: \log m$
- For each variable in $\varphi$ (at most $m$ ), store current constant assignment (as a pointer): $m \cdot \log n$
- Checking $\left\langle c_{1}, \ldots, c_{n}\right\rangle \in p^{I}$ 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 $n=|I|$ (total table sizes)

Memory in $m \log m+(m+1) \log n$
Space complexity of FO query evaluation

- Combined complexity: in 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 PSpace. Is this the best we can get?

Hardness proof: reduce a known PSpace-hard problem to FO query evaluation

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Hardness proof: reduce a known PSPACE-hard problem to FO query evaluation
$\leadsto$ QBF satisfiability
Let $\bigcirc_{1} X_{1} . \bigcirc_{2} X_{2} \cdots \bigcirc_{n} X_{n} . \varphi\left[X_{1}, \ldots, X_{n}\right]$ be a QBF (with $\bigcirc_{i} \in\{\forall, \exists\}$ )

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

$$
\bigcirc_{1} x_{1} \cdot \varrho_{2} x_{2} \cdots \wp_{n} x_{n} . \varphi\left[X_{1} \mapsto \operatorname{true}\left(x_{1}\right), \ldots, X_{n} \mapsto \operatorname{true}\left(x_{n}\right)\right]
$$

## PSpace-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 $\exists x . \neg \operatorname{true}(x)$

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Example: QBF $\exists p . \neg p$ leads to FO query $\exists x . \neg \operatorname{true}(x)$
Better approach:

- Consider QBF $\bigcirc_{1} X_{1} . \varrho_{2} X_{2} \cdots \bigcirc_{n} X_{n} . \varphi\left[X_{1}, \ldots, X_{n}\right]$ with $\varphi$ in negation normal form: negations only occur directly before variables $X_{i}$ (still PSpace-complete: exercise)
- Database instance $I$ with $\Delta^{I}=\{0,1\}$
- Two tables with one row each: true(1) and false(0)
- Transform input QBF into Boolean FO query

$$
\bigcirc_{1} x_{1} \cdot \bigcirc_{2} x_{2} \cdots \bigcirc_{n} x_{n} \cdot \varphi^{\prime}
$$

where $\varphi^{\prime}$ is obtained by replacing each negated variable $\neg X_{i}$ with false $\left(x_{i}\right)$ and each non-negated variable $X_{i}$ with true $\left(x_{i}\right)$.

## Combined Complexity of FO Query Answering

## Theorem

The evaluation of FO queries is PSPACE-complete with respect to combined complexity.

We have actually shown something stronger:

## Theorem

The evaluation of FO queries is PSPACE-complete with respect to query complexity.

## Data Complexity of FO Query Answering

The algorithm showed that FO query evaluation is in L
$\leadsto$ can we do any better?
What could be better than L?

$$
? \subseteq \mathrm{~L} \subseteq \mathrm{NL} \subseteq \mathrm{P} \subseteq \ldots
$$

$\leadsto$ 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
$\leadsto$ we will only consider Boolean circuits with exactly one output
$\leadsto$ propositional logic formulae are Boolean circuits with one output and gates of fanout $\leq 1$


## Example

A Boolean circuit over an input string $x_{1} x_{2} \ldots x_{n}$ of length $n$


Corresponds to formula $\left(x_{1} \wedge x_{2}\right) \vee\left(x_{1} \wedge x_{3}\right) \vee \ldots \vee\left(x_{n-1} \wedge x_{n}\right)$
$\leadsto$ accepts all strings with at least two 1 s

## Circuits as a Model for Parallel Computation

Previous example:

$\leadsto n^{2}$ processors working in parallel
$\leadsto$ computation finishes in 2 steps

- size: number of gates = total number of computing steps
- depth: longest path of gates = time for parallel computation
$\leadsto$ refinement of polynomial time taking parallelizability into account


## Solving Problems With Circuits

Observation: the input size is "hard-wired" in circuits
$\leadsto$ each circuit only has a finite number of different inputs
$\leadsto$ 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
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How can we solve interesting problems with Boolean circuits?

## Definition

A uniform family of Boolean circuits is a set of circuits $C_{n}(n \geq 0)$ 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 $\left(C_{n}\right)_{n \geq 0}$ of Boolean circuits if for each word $w$ of length $|w|$ :

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

## 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

$\left(C_{n}\right)_{n \geq 0}$ 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, $O\left(\log ^{k} n\right)$.


## The Complexity Classes NC and AC

Two important types of small-depth circuits

## Definition <br> $\mathrm{NC}^{k}$ is the class of problems that can be solved by uniform families of circuits $\left(C_{n}\right)_{n \geq 0}$ of fan-in $\leq 2$, size polynomial in $n$, and depth in $O\left(\log ^{k} n\right)$.

The class NC is defined as NC $=\bigcup_{k \geq 0} \mathrm{NC}^{k}$.
("Nick's Class" named after Nicholas Pippenger by Stephen Cook)

## Definition

$\mathrm{AC}^{k}$ and AC are defined like $\mathrm{NC}^{k}$ 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 $\leadsto$ 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 $\leadsto$ in $\mathrm{NC}^{1}$

## Relationships of Circuit Complexity Classes

The previous sketch can be generalised:

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

Only few inclusions are known to be proper: $\mathrm{NC}^{0} \subset \mathrm{AC}^{0} \subset \mathrm{NC}^{1}$
Direct consequence of above hierarchy: $\mathrm{NC}=\mathrm{AC}$
Interesting relations to other classes:

$$
\mathrm{NC}^{0} \subset \mathrm{AC}^{0} \subset \mathrm{NC}^{1} \subseteq \mathrm{~L} \subseteq \mathrm{NL} \subseteq \mathrm{AC}^{1} \subseteq \ldots \subseteq \mathrm{NC} \subseteq \mathrm{P}
$$

Intuition:

- Problems in NC are parallelisable
- Problems in P \NC are inherently sequential

However: it is not known if $\mathrm{NC} \neq \mathrm{P}$

## Back to Databases ...

## Theorem

The evaluation of FO queries is complete for (logtime uniform) $\mathrm{AC}^{0}$ 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)
$\leadsto$ 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
$\leadsto$ true or false depending on whether the subformula holds for this tuple or not
- Logical operators correspond to gate types: basic operators obvious, $\forall$ as generalised conjunction, $\exists$ as generalised disjunction
- subformula with $n$ free variables $\sim \mid$ adom $\left.\right|^{n}$ gates $\sim$ especially: |adom| ${ }^{0}=1$ output gate for Boolean query


## Example

We consider the formula

$$
\exists z \cdot(\exists x \cdot \exists y \cdot R(x, y) \wedge S(y, z)) \wedge \neg R(a, z)
$$

Over the database instance:
R:

| $a$ | $a$ |
| :--- | :--- |
| $a$ | $b$ |

S:

| $b$ | $b$ |
| :--- | :--- |
| $b$ | $c$ |

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

## Example: ヨz. $(\exists x . \exists y \cdot R(x, y) \wedge S(y, z)) \wedge \neg R(a, z)$

| $R(a, a)$ | $R(a, b)$ | $R(a, c)$ | $\ldots$ | $S(a, a)$ | $\ldots$ | $S(b, a)$ | $S(b, b)$ | $S(b, c)$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 1 | 0 | $\ldots$ | 0 | $\ldots$ | 0 | 1 | 1 |

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## Summary and Outlook

The evaluation of FO queries is

- PSpace-complete for combined complexity
- PSpace-complete for query complexity
- $\mathrm{AC}^{0}$-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?

