

Database Theory

VU 181.140, SS 2018

5. Complexity of Query Evaluation

Reinhard Pichler

Institut für Informationssysteme
Arbeitsbereich DBAI
Technische Universität Wien

17 April, 2018



Outline

- 5. Complexity of Query Evaluation
 - 5.1 Measures of Complexity
 - 5.2 Complexity of First-order Queries
 - 5.3 Complexity of Conjunctive Queries
 - 5.4 Complexity of Datalog

Beyond Traktenbrot's Theorem

- By Traktenbrot's Theorem, it is undecidable to check whether a given first-order query Q produces some output over some database.
- What happens if D is actually given as input?

The following are natural (decision) problems in this context:

QUERY-OUTPUT-TUPLE (QOT)

INSTANCE: A database D , a query Q , a tuple \vec{c} of values.

QUESTION: Does $\vec{c} \in Q(D)$ hold?

BOOLEAN-QUERY-EVALUATION (BQE)

INSTANCE: A database D , a Boolean query Q .

QUESTION: Does Q evaluate to **true** in D ?

NOTE: we often view Boolean domain calculus queries $\{\langle \rangle | \phi\}$ simply as closed formulae ϕ .

QUERY-NON-EMPTINESS (QNE)

INSTANCE: A database D , a query Q .

QUESTION: Does query Q yield a non-empty result over the DB D , i.e. $Q(D) \neq \emptyset$?

QOT vs. BQE vs. QNE

We concentrate here mainly on the complexity of **BQE**.

Not a limitation: in our setting **QOT** and **QNE** are essentially the same problems as **BQE**:

From QOT to BQE

Assume a database D , a domain calculus query $Q = \{\vec{x} \mid \phi(\vec{x})\}$, and a value tuple $\vec{c} = \langle c_1, \dots, c_n \rangle$. Then $\vec{c} \in Q(D)$ iff Q' evaluates to **true** in D , where

$$\vec{x} = \langle x_1, \dots, x_n \rangle, \text{ and}$$

$$Q' = \exists \vec{x}. (\phi(\vec{x}) \wedge x_1 = c_1 \wedge \dots \wedge x_n = c_n)$$

From QNE to BQE

Assume a database D and a domain calculus query $Q = \{\vec{x} \mid \phi(\vec{x})\}$. Then $Q(D) \neq \emptyset$ iff $\exists \vec{x}. \phi(\vec{x})$ evaluates to **true** in D .

From **BQE** to **QNE** and **QOT** \rightsquigarrow trivial.

Complexity Measures for **BQE**

Combined complexity

The complexity of **BQE** without any assumptions about the input query Q and database D is called the **combined complexity** of **BQE**.

Further measures are obtained by restricting the input:

Data and query complexity

Data complexity of BQE refers to the following decision problem:

Let Q be some *fixed* Boolean query.

INSTANCE: An input database D .

QUESTION: Does Q evaluate to **true** in D ?

Query complexity of BQE refers to the following decision problem:

Let D be some *fixed* input database.

INSTANCE: A Boolean query Q .

QUESTION: Does Q evaluate to **true** in D ?

Relevant Complexity Classes

We recall the inclusions between some fundamental complexity classes:

$$L \subseteq P \subseteq NP \subseteq PSPACE \subseteq EXPTIME$$

- L is the class of all problems solvable in logarithmic space,
- P —,,— in polynomial time,
- NP —,,— in nondeterministic polynomial time,
- PSPACE —,,— in polynomial space,
- EXPTIME —,,— in exponential time.

Complexity of First-order Queries

Theorem (A)

The query complexity and the combined complexity of domain calculus queries is PSPACE-complete (even if we disallow negation and equality atoms). The data complexity is in L (actually, even in a much lower class).

To prove the theorem, we proceed in steps as follows:

- 1** We provide an algorithm for query evaluation:
 - it shows PSPACE membership for combined complexity (and thus for query complexity as well), and
 - L membership w.r.t. data complexity,
- 2** We show PSPACE-hardness of query complexity (clearly, the lower bound applies to combined complexity as well).

An algorithm for evaluating FO queries

- We consider an arbitrary FO formula ψ and a database D .
- W.l.o.g., the formula is of the form

$$\psi = \exists x_1 \forall y_1 \dots \exists x_n \forall y_n \varphi(x_1, y_1, \dots, x_n, y_n).$$

- Let the active domain dom of D be $dom = \{a_1, \dots, a_m\}$.
- For the evaluation of the formula, we design two procedures $evaluate_{\exists}$ and $evaluate_{\forall}$, which call each other recursively.
- The algorithm uses global variables n and $X = \{x_1, y_1, \dots, x_n, y_n\}$.

GLOBAL $x_1, y_1, \dots, x_n, y_n$

Boolean $evaluate_{\exists}(\text{Integer } i)$

for x_i **from** a_1 **to** a_m **do**

if $evaluate_{\forall}(i) = \text{true}$ **then return true**

endfor

return false.

Boolean $evaluate_{\forall}(\text{Integer } i)$

for y_i **from** a_1 **to** a_m **do**

if $i = n$ **then**

if φ evaluates to **false** under the

 current values of $x_1, y_1, \dots, x_n, y_n$ **then return false**

endif

else

if $evaluate_{\exists}(i + 1) = \text{false}$ **then return false**

endif

endfor

return true.

By construction: ψ is **true** in D iff $evaluate_{\exists}(1) = \text{true}$.

Let us analyse the space usage of our algorithm. We have to store:

- 1 The input database D and the formula ψ :
 - do not contribute to the space requirements.
- 2 The global variables $X = \{x_1, y_1, \dots, x_n, y_n\}$.
 - Each variable requires $\mathcal{O}(\log m)$ bits of space. Thus X needs $\mathcal{O}(n \log m)$ bits. Note that X requires logarithmic space if ψ is fixed.
- 3 A call stack $\mathcal{S} = \langle S_1, \dots, S_k \rangle$, where $k \leq 2n$ and each S_j stores a state in which a subroutine is called. Clearly, for both subroutines a state S_j only needs to store the value of i and the return position in the subroutine.
 - Storing a value $i \in \{1, \dots, 2n\}$ requires logarithmic space in the size of ψ (i.e. $\mathcal{O}(\log n)$), but only constant space if ψ is fixed. (The return position requires constant space in both cases.)
 - Hence \mathcal{S} needs $\mathcal{O}(n \log n)$ bits of storage, which is constant if ψ is fixed.
- 4 Space for evaluating φ in an assignment
 - requires a transversal of the parse tree of ψ : space $\mathcal{O}(\log \|\psi\|)$ suffices.

Overall we need $\mathcal{O}(n \log m + n \log n + \log \|\psi\|)$ bits of storage.

$\mathcal{O}(n \log m + n \log n + \log \|\psi\|)$ means that we only need polynomial space in the combined size of D and ψ .

Proposition

BQE \in PSPACE w.r.t. *combined complexity*.

This also implies **BQE** \in PSPACE w.r.t. *query complexity*.

For fixed ψ , the space required is $\mathcal{O}(\log m)$, i.e. logarithmic in the data.

Proposition

BQE \in L w.r.t. *data complexity*.

NOTE: Note that $L \subseteq P$. In fact, one can show completeness of **BQE** w.r.t. data complexity for a much lower circuit class $AC_0 \subseteq L$.

The PSPACE lower bound

To prove the PSPACE-hardness result, we first recall **quantified Boolean formulae**:

QSAT (QBF)

INSTANCE: An expression $\exists x_1 \forall x_2 \exists x_3 \cdots Q x_n \phi$, where Q is either \forall or \exists and ϕ is a Boolean formula in CNF with variables from $\{x_1, x_2, x_3, \dots, x_n\}$.

QUESTION: Is there a truth value for the variable x_1 such that for both truth values of x_2 there is a truth value for x_3 and so on up to x_n , such that ϕ is satisfied by the overall truth assignment?

Theorem

QSAT is PSPACE-complete.

Remark. A detailed proof is given in the Komplexitätstheorie lecture.

Proof of the PSPACE-Hardness of BQE

The PSPACE-hardness result for Theorem (A) can be shown by a reduction from the QSAT-problem. Let ψ be an arbitrary QBF with

$$\psi = \exists x_1 \forall x_2 \dots Q x_n \alpha(x_1, \dots, x_n)$$

where Q is either \forall or \exists and α is a quantifier-free Boolean formula with variables in $\{x_1, \dots, x_n\}$.

We first define the (fixed) input database D over the predicate symbols $\mathcal{L} = \{\text{istrue}, \text{isequal}, \text{not}, \text{or}, \text{and}\}$ with the obvious meaning:

$$D = \{\text{istrue}(1), \text{isequal}(0, 0), \text{isequal}(1, 1), \text{not}(1, 0), \text{not}(0, 1), \\ \text{or}(1, 1, 1), \text{or}(1, 0, 1), \text{or}(0, 1, 1), \text{or}(0, 0, 0), \\ \text{and}(1, 1, 1), \text{and}(1, 0, 0), \text{and}(0, 1, 0), \text{and}(0, 0, 0)\}$$

Proof of the PSPACE-Hardness (continued)

For each sub-formula β of α , we define a quantifier-free, first-order formula $T_\beta(z_1, \dots, z_n, x)$ with the following intended meaning: if the variables x_i have the truth value z_i , then the formula $\beta(x_1, \dots, x_n)$ evaluates to the truth value x .

The formulae $T_\beta(z_1, \dots, z_n, x)$ can be defined inductively w.r.t. the structure of α as follows:

Case $\beta =$

$$x_i \text{ (with } 1 \leq i \leq n): \quad T_\beta(\bar{z}, x) \equiv \text{isequal}(z_i, x)$$

$$\neg\beta': \quad T_\beta(\bar{z}, x) \equiv \exists t_1 T_{\beta'}(\bar{z}, t_1) \wedge \text{not}(t_1, x)$$

$$\beta_1 \wedge \beta_2: \quad T_\beta(\bar{z}, x) \equiv \exists t_1, t_2 T_{\beta_1}(\bar{z}, t_1) \wedge T_{\beta_2}(\bar{z}, t_2) \wedge \text{and}(t_1, t_2, x)$$

$$\beta_1 \vee \beta_2: \quad T_\beta(\bar{z}, x) \equiv \exists t_1, t_2 T_{\beta_1}(\bar{z}, t_1) \wedge T_{\beta_2}(\bar{z}, t_2) \wedge \text{or}(t_1, t_2, x)$$

Proof of the PSPACE-Hardness (continued)

The first-order query ϕ is then defined as follows:

$$\phi \equiv \exists x \exists z_1 \forall z_2 \dots Q z_n T_\alpha(\bar{z}, x) \wedge \text{istrue}(x)$$

where Q is either \forall or \exists (as in the formula ψ).

We claim that this problem reduction is correct, i.e.:

The QBF $\psi = \exists x_1 \forall x_2 \dots Q x_n \alpha(x_1, \dots, x_n)$ is **true** \Leftrightarrow
the first-order query $\phi \equiv \exists x \exists z_1 \forall z_2 \dots Q z_n T_\alpha(\bar{z}, x) \wedge \text{istrue}(x)$
evaluates to **true** over the database D .

The proof is straightforward. It suffices to show by induction on the structure of α that the formulae $T_\beta(z_1, \dots, z_n, x)$ indeed have the intended meaning.

Complexity of Conjunctive Queries

Recall that **conjunctive queries** (CQs) are a special case of first-order queries whose only connective is \wedge and whose only quantifier is \exists (i.e., \vee , \neg and \forall are excluded).

$$\text{E.g.: } Q = \{\langle x \rangle \mid \exists y, z. R(x, y) \wedge R(y, z) \wedge P(z, x)\}$$

Theorem (B)

*The query complexity and the combined complexity of **BQE** for conjunctive queries is NP-complete.*

Proof

NP-Membership (of the combined complexity). For each variable u of the query, we guess a domain element to which u is instantiated. Then we check whether all the resulting ground atoms in the query body exist in D . This check is obviously feasible in polynomial time.

Proof (continued)

Hardness (of the query complexity). We reduce the NP-complete 3-SAT problem to our problem. For this purpose, we consider the following input database (over a ternary relation symbol c and a binary relation symbol v) as fixed:

$$D = \{ c(1, 1, 1), c(1, 1, 0), c(1, 0, 1), c(1, 0, 0), \\ c(0, 1, 1), c(0, 1, 0), c(0, 0, 1), v(1, 0), v(0, 1) \}$$

Now let an arbitrary instance of the **3-SAT problem** be given through the 3-CNF formula $\Phi = \bigwedge_{i=1}^n l_{i1} \vee l_{i2} \vee l_{i3}$ over the propositional variables x_1, \dots, x_k . Then we define a conjunctive query Q as follows:

$$(\exists x_1, \dots, x_k, \bar{x}_1, \dots, \bar{x}_k) \\ c(l_{11}^*, l_{12}^*, l_{13}^*) \wedge \dots \wedge c(l_{n1}^*, l_{n2}^*, l_{n3}^*) \wedge v(x_1, \bar{x}_1) \wedge \dots \wedge v(x_k, \bar{x}_k)$$

where $l^* = x$ if $l = x$, and $l^* = \bar{x}$ if $l = \neg x$. Moreover, $\bar{x}_1, \dots, \bar{x}_k$ are fresh first-order variables. By slight abuse of notation, we thus use x_i to denote either a propositional atom (in Φ) or a first-order variable (in Q).

It is straightforward to verify that the 3-CNF formula Φ is satisfiable $\Leftrightarrow Q$ evaluates to **true** in D .

Complexity of Datalog

Theorem (C)

Query evaluation (i.e., the QOT problem) of Datalog has the following complexity:

- *P-complete w.r.t. data complexity, and*
- *EXPTIME-complete w.r.t. combined and query complexity.*

To prove the theorem, we first concentrate on **ground** Datalog programs:

- A program is ground if it has no variables.
- Such programs are also known as **propositional logic programs**.
- Note that a ground atom $R(\text{tim}, \text{bob})$ can be seen as a propositional variable $R_{\text{tim}, \text{bob}}$.

Ground Datalog

Theorem

Query evaluation for ground Datalog programs is P-complete w.r.t. combined complexity.

Proof: (Membership)

- Recall that the semantics of a given program P can be defined as the least fixed-point of the immediate consequence operator T_P
- This least fixpoint $T_P^\omega(DB)$ can be computed in polynomial time even if the “naive” evaluation algorithm is applied.
- The number of iterations (i.e. applications of T_P) is bounded by the number of rules plus 1.
- Each iteration step is clearly feasible in polynomial time.

P-hardness of Ground Datalog

Proof: (Hardness)

- By encoding of a TM. Assume $M = (K, \Sigma, \delta, q_{start})$, an input string I and a number of steps N , where N is a polynomial of $|I|$.
- We construct in logspace a program $P(M, N)$, a database $DB(I, N)$ and an atom A such that

$$A \in T_{P(M,N)}^\omega(DB(I, N)) \text{ iff } M \text{ accepts } I \text{ in } N \text{ steps.}$$

- Recall that the transition function δ of M with a single tape can be represented by a table whose rows are tuples $t = \langle q_1, \sigma_1, q_2, \sigma_2, d \rangle$. Such a tuple t expresses the following if-then-rule:

if at some time instant τ the machine is in state q_1 , the cursor points to cell number π , and this cell contains symbol σ_1
then at instant $\tau + 1$ the machine is in state q_2 , cell number π contains symbol σ_2 , and the cursor points to cell number $\pi + d$.

P-hardness of Ground Datalog: the Atoms

The propositional atoms in $P(M, N)$.

(there are many, but only polynomially many...)

$symbol_{\alpha}[\tau, \pi]$ for $0 \leq \tau \leq N$, $0 \leq \pi \leq N$ and $\alpha \in \Sigma$. Intuitive meaning: at instant τ of the computation, cell number π contains symbol α .

$cursor[\tau, \pi]$ for $0 \leq \tau \leq N$ and $0 \leq \pi \leq N$. Intuitive meaning: at instant τ , the cursor points to cell number π .

$state_q[\tau]$ for $0 \leq \tau \leq N$ and $q \in K$. Intuitive meaning: at instant τ , the machine M is in state q .

$accept$ Intuitive meaning: M has reached state q_{yes} .

P-hardness of Ground Datalog: the Database

The construction of the **database** $DB(I, N)$:

$symbol_{\triangleright}[0, 0]$,

$symbol_{\sigma}[0, \pi]$, for $0 < \pi \leq |I|$, where $I_{\pi} = \sigma$

$symbol_{\sqcup}[0, \pi]$, for $|I| < \pi \leq N$

$cursor[0, 0]$,

$state_{q_{start}}[0]$.

P-hardness of Ground Datalog: the Rules

- **transition rules:** for each entry $\langle q_1, \sigma_1, q_2, \sigma_2, d \rangle$, $0 \leq \tau < N$, $0 \leq \pi < N$, and $0 \leq \pi + d$.

$$\begin{aligned} symbol_{\sigma_2}[\tau + 1, \pi] &\leftarrow state_{q_1}[\tau], symbol_{\sigma_1}[\tau, \pi], cursor[\tau, \pi] \\ cursor[\tau + 1, \pi + d] &\leftarrow state_{q_1}[\tau], symbol_{\sigma_1}[\tau, \pi], cursor[\tau, \pi] \\ state_{q_2}[\tau + 1] &\leftarrow state_{q_1}[\tau], symbol_{\sigma_1}[\tau, \pi], cursor[\tau, \pi] \end{aligned}$$

- **inertia rules:** where $0 \leq \tau < N$, $0 \leq \pi < \pi' \leq N$

$$\begin{aligned} symbol_{\sigma_1}[\tau + 1, \pi] &\leftarrow symbol_{\sigma_1}[\tau, \pi], cursor[\tau, \pi'] \\ symbol_{\sigma_1}[\tau + 1, \pi'] &\leftarrow symbol_{\sigma_1}[\tau, \pi'], cursor[\tau, \pi] \end{aligned}$$

- **accept rule:** for $0 \leq \tau \leq N$

$$accept \leftarrow state_{q_{yes}}[\tau]$$

P-hardness of Ground Datalog

- The encoding precisely simulates the behaviour of M on input I up to N steps. (This can be formally shown by induction on the time steps.)
- $accept \in T_{P(M,N)}^\omega(DB(I, N))$ iff M accepts I in N steps.
- The construction is feasible in logarithmic space.
- Note that each rule in $P(M, N)$ has at most 4 atoms. In fact, P-hardness applies already for programs with at most 3 atoms in the rules:
 - Simply replace each $A \leftarrow B, C, D$ in $P(M, N)$ by $A \leftarrow B, E$ and $E \leftarrow C, D$, where E is a fresh atom.

Data Complexity of Datalog

Proposition

Query evaluation in Datalog is P-complete w.r.t. data complexity.

Proof: (Membership)

Effective reduction to reasoning over ground Datalog programs is possible. Given a program P , a database DB , and atom A :

- Generate $P' = \text{ground}(P, DB)$, i.e. the set of all ground instances of rules in P :

$$\text{ground}(P, DB) = \bigcup_{r \in P} \text{Ground}(r; P, DB)$$

NB: more details on $\text{Ground}(r; P, DB)$ in Lecture 2.

- Decide whether $A \in T_{P'}^{\omega}(DB)$.

Grounding Complexity

Given a program P and a database DB , the number of rules in $ground(P, DB)$ is bounded by

$$|P| * \#const(P, DB)^{vmax}$$

- $vmax$ is the maximum number of different variables in any rule $r \in P$
- $\#const(P, DB)$ is the number of constants occurring in P and DB .
- $ground(P, DB)$ is polynomial in the size of DB .
- Hence, the complexity of propositional logic programming is an upper bound for the data complexity.
- Note that $ground(P, DB)$ can be exponential in the size of P .

Data Complexity of Datalog: P-hardness

Proof: Hardness

The P-hardness can be shown by writing a simple Datalog *meta-interpreter* for ground programs with at most 3 atoms per rule:

- Represent rules $A_0 \leftarrow A_1, \dots, A_i$ of such a program P , where $0 \leq i \leq 2$, using database facts $\langle A_0, \dots, A_i \rangle$ in an $(i + 1)$ -ary relation R_i on the propositional atoms.
- Then, the program P which is stored this way in a database $DB_{MI}(P)$ can be evaluated by a fixed Datalog program P_{MI} which contains for each relation R_i , $0 \leq i \leq 2$, a rule

$$T(X_0) \leftarrow T(X_1), \dots, T(X_i), R_i(X_0, \dots, X_i).$$

- $T(x)$ intuitively means that atom x is true. Then,

$$A \in T_P^\omega(DB) \text{ iff } T(A) \in T_{P_{MI}}^\omega(DB_{MI}(P))$$
- P-hardness of the data complexity of Datalog is then immediately obtained.

Combined and Query Complexity of Datalog

Proposition

Datalog is EXPTIME-complete w.r.t. query and combined complexity.

Proof

(Membership) Grounding P using DB leads to a propositional program $ground(P, DB)$ whose size is exponential in the size of P and DB . Hence, the query and the combined complexity is in EXPTIME.

(Hardness) We show hardness for query complexity only. Goal: adapt our previous encoding of TM M and input I to obtain a program $P_{dat}(M, I, N)$ and a fixed database DB_{dat} to decide acceptance of M on I within $N = 2^m$ steps, where $m = n^k$ ($n = |I|$) is a polynomial.

Note: We are not allowed to generate an exponentially large program by using exponentially many propositional atoms (the reduction would not be polynomial!).

Query Complexity of Datalog: EXPTIME-hardness

Ideas for lifting $P(M, N)$ and $DB(I, N)$ to $P_{dat}(M, I, N)$ and DB_{dat} :

- use the predicates $symbol_\sigma(\mathbf{X}, \mathbf{Y})$, $cursor(\mathbf{X}, \mathbf{Y})$ and $state_s(\mathbf{X})$ instead of the propositional letters $symbol_\sigma[X, Y]$, $cursor[X, Y]$ and $state_q[X]$ respectively.
- W.l.o.g., let N be of the form $N = 2^m - 1$ for some integer $m \geq 1$. The time points τ and tape positions π from 0 to N are encoded in binary, i.e. by m -ary tuples $t_\tau = \langle c_1, \dots, c_m \rangle$, $c_i \in \{0, 1\}$, $i = 1, \dots, m$, such that $0 = \langle 0, \dots, 0 \rangle$, $1 = \langle 0, \dots, 1 \rangle$, $N = \langle 1, \dots, 1 \rangle$.
- The functions $\tau + 1$ and $\pi + d$ are realized by means of the successor $Succ^m$ from a linear order \leq^m on $\{0, 1\}^m$.

Query Complexity of Datalog: EXPTIME-hardness

The predicates $Succ^m$, $First^m$, and $Last^m$ are provided.

- The database facts $symbol_{\sigma}[0, \pi]$ are readily translated into the Datalog rules

$$symbol_{\sigma}(\mathbf{X}, \mathbf{t}) \leftarrow First^m(\mathbf{X}),$$

where \mathbf{t} represents the position π ,

- Similarly for the facts $cursor[0, 0]$ and $state_{s_0}[0]$.
- Database facts $symbol_{\sqcup}[0, \pi]$, where $|I| \leq \pi \leq N$, are translated to the rule

$$symbol_{\sqcup}(\mathbf{X}, \mathbf{Y}) \leftarrow First^m(\mathbf{X}), \leq^m(\mathbf{t}, \mathbf{Y})$$

where \mathbf{t} represents the number $|I| + 1$.

Query Complexity of Datalog: EXPTIME-hardness

- **Transition** and **inertia rules**: for realizing $\tau + 1$ and $\pi + d$, use in the body atoms $Succ^m(\mathbf{X}, \mathbf{X}')$. For example, the clause

$$symbol_{\sigma_2}[\tau + 1, \pi] \leftarrow state_{q_1}[\tau], symbol_{\sigma_1}[\tau, \pi], cursor[\tau, \pi]$$

is translated into

$$symbol_{\sigma_2}(\mathbf{X}', \mathbf{Y}) \leftarrow state_{q_1}(\mathbf{X}), symbol_{\sigma_1}(\mathbf{X}, \mathbf{Y}), cursor(\mathbf{X}, \mathbf{Y}), Succ^m(\mathbf{X}, \mathbf{X}').$$

- The translation of the **accept rules** is straightforward:

$$accept \leftarrow state_{q_{yes}}(\mathbf{X}).$$

Defining $Succ^m(\mathbf{X}, \mathbf{X}')$ and \leq^m

- The ground facts $Succ^1(0, 1)$, $First^1(0)$, and $Last^1(1)$ are provided in DB_{dat} .
- For an inductive definition, suppose $Succ^i(\mathbf{X}, \mathbf{Y})$, $First^i(\mathbf{X})$, and $Last^i(\mathbf{X})$ tell the successor, the first, and the last element from a linear order \leq^i on $\{0, 1\}^i$, where \mathbf{X} and \mathbf{Y} have arity i .

Then, use rules

$$\begin{aligned}
 Succ^{i+1}(Z, \mathbf{X}, Z, \mathbf{Y}) &\leftarrow Succ^i(\mathbf{X}, \mathbf{Y}) \\
 Succ^{i+1}(Z, \mathbf{X}, Z', \mathbf{Y}) &\leftarrow Succ^1(Z, Z'), Last^i(\mathbf{X}), First^i(\mathbf{Y}) \\
 First^{i+1}(Z, \mathbf{X}) &\leftarrow First^1(Z), First^i(\mathbf{X}) \\
 Last^{i+1}(Z, \mathbf{X}) &\leftarrow Last^1(Z), Last^i(\mathbf{X})
 \end{aligned}$$

Defining $Succ^m(\mathbf{X}, \mathbf{X}')$ and \leq^m

- The ground facts $Succ^1(0, 1)$, $First^1(0)$, and $Last^1(1)$ are provided in DB_{dat} .
- For an inductive definition, suppose $Succ^i(\mathbf{X}, \mathbf{Y})$, $First^i(\mathbf{X})$, and $Last^i(\mathbf{X})$ tell the successor, the first, and the last element from a linear order \leq^i on $\{0, 1\}^i$, where \mathbf{X} and \mathbf{Y} have arity i .

Alternatively, use rules

$$\begin{aligned}
 Succ^{i+1}(0, \mathbf{X}, 0, \mathbf{Y}) &\leftarrow Succ^i(\mathbf{X}, \mathbf{Y}) \\
 Succ^{i+1}(1, \mathbf{X}, 1, \mathbf{Y}) &\leftarrow Succ^i(\mathbf{X}, \mathbf{Y}) \\
 Succ^{i+1}(0, \mathbf{X}, 1, \mathbf{Y}) &\leftarrow Last^i(\mathbf{X}), First^i(\mathbf{Y}) \\
 First^{i+1}(0, \mathbf{X}) &\leftarrow First^i(\mathbf{X}) \\
 Last^{i+1}(1, \mathbf{X}) &\leftarrow Last^i(\mathbf{X})
 \end{aligned}$$

Defining $Succ^m(\mathbf{X}, \mathbf{X}')$ and \leq^m

- The ground facts $Succ^1(0, 1)$, $First^1(0)$, and $Last^1(1)$ are provided in DB_{dat} .
- For an inductive definition, suppose $Succ^i(\mathbf{X}, \mathbf{Y})$, $First^i(\mathbf{X})$, and $Last^i(\mathbf{X})$ tell the successor, the first, and the last element from a linear order \leq^i on $\{0, 1\}^i$, where \mathbf{X} and \mathbf{Y} have arity i .

Alternatively, use rules

$$\begin{aligned}
 Succ^{i+1}(0, \mathbf{X}, 0, \mathbf{Y}) &\leftarrow Succ^i(\mathbf{X}, \mathbf{Y}) \\
 Succ^{i+1}(1, \mathbf{X}, 1, \mathbf{Y}) &\leftarrow Succ^i(\mathbf{X}, \mathbf{Y}) \\
 Succ^{i+1}(0, \mathbf{X}, 1, \mathbf{Y}) &\leftarrow Last^i(\mathbf{X}), First^i(\mathbf{Y}) \\
 First^{i+1}(0, \mathbf{X}) &\leftarrow First^i(\mathbf{X}) \\
 Last^{i+1}(1, \mathbf{X}) &\leftarrow Last^i(\mathbf{X})
 \end{aligned}$$

- The order \leq^m is easily defined from $Succ^m$ by two clauses

$$\begin{aligned}
 \leq^m(\mathbf{X}, \mathbf{X}) &\leftarrow \\
 \leq^m(\mathbf{X}, \mathbf{Y}) &\leftarrow Succ^m(\mathbf{X}, \mathbf{Z}), \leq^m(\mathbf{Z}, \mathbf{Y})
 \end{aligned}$$

Combined and Query Complexity of Datalog: Conclusion

- Let L be an arbitrary language in EXPTIME, i.e., there exists a Turing machine M deciding L in exponential time. Then there is a constant k such that M accepts/rejects every input I within $2^{|I|^k}$ steps.
- The program $P_{dat}(M, I, |I|^k)$ is constructible from M and I in polynomial time (in fact, careful analysis shows feasibility in logarithmic space).
- $accept$ is in the answer of $P_{dat}(M, I, |I|^k)$ evaluated over $DB_{dat} \Leftrightarrow M$ accepts input I within N steps.
- Thus the EXPTIME-hardness follows.

Complexity of Datalog with Stratified Negation

Theorem

Reasoning in stratified ground Datalog programs with negation is P-complete. Stratified Datalog with negation is

- *P-complete w.r.t. data complexity, and*
 - *EXPTIME-complete w.r.t. combined and query complexity.*
-
- A ground stratified program P can be partitioned into disjoint sets S_1, \dots, S_n s.t. the semantics of P is computed by successively computing in polynomial time the fixed-points of the immediate consequence operators T_{S_1}, \dots, T_{S_n} .
 - As with plain Datalog, for programs with variables, the grounding step causes an exponential blow-up.

Learning Objectives

- The **BQE**, **QOT** and **QNE** problems
- The notions of combined, data and query complexity
- The complexity of first-order queries
- The complexity of conjunctive queries
- The complexity of plain and stratified Datalog