

KAPITEL 3 BERUHT AUF EINEM AUSZUG AUS DEM FOLGENDEN VORTRAG:

A tutorial on Database Theory
and a talk on
database query answering under updates

Nicole Schweikardt

Humboldt-Universität zu Berlin

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Computation (WoLLIC 2017)

London, July 19 & 20, 2017

Example database and two queries

<i>Movie</i>	
Name	Actor
Alien	Sigourney Weaver
Blade Runner	Harrison Ford
Blade Runner	Sean Young
Brazil	Jonathan Pryce
Brazil	Kim Greist
Casablanca	Humphrey Bogart
Casablanca	Ingrid Bergmann
Gravity	Sandra Bullock
Gravity	George Clooney
Resident Evil	Milla Jovovich
Terminator	Arnold Schwarzenegger
Terminator	Linda Hamilton
Terminator	Michael Biehn
:	:

<i>Programme</i>		
Cinema	Movietitle	Time
Babylon	Casablanca	17:30
Babylon	Gravity	20:15
Casablanca	Blade Runner	15:30
Casablanca	Alien	18:15
Casablanca	Blade Runner	20:30
Casablanca	Resident Evil	20:30
Kino International	Casablanca	18:00
Kino International	Brazil	20:00
Kino International	Brazil	22:00
Movimiento	Gravity	17:00
Movimiento	Gravity	19:30
Movimiento	Alien	22:00
Urania	Resident Evil	20:00
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a db	:	$D = (M^D, P^D)$, where
M^D	:	a finite subset of \mathbf{dom}^2
P^D	:	a finite subset of \mathbf{dom}^3
dom	:	a fixed, infinite domain of potential db entries

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Query evaluation

Consider a query language L
(e.g., SQL, conjunctive queries CQ, first-order logic FO).

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Special case $k = 0$: **Boolean queries:**

Evaluate $\varphi()$ on D means Decide if $(\text{adom}(D), D) \models \varphi$

Complexity of query evaluation

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CAVEAT: These notions & results cannot handle updates of the db!

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For the static setting: tight characterisation of the tractable CQs:

Boolean: [Grohe, Schwentick, Segoufin 2001], [Grohe 2007], [Marx 2010], [Marx 2013]

counting: [Dalmau, Jonsson 2004], [Chen, Mengel 2015], [Greco, Scarcello 2015]

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Interesting: Sub-linear update time

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 - ▶ Database D arbitrary
 - ▶ query $\varphi(x_1, \dots, x_k)$ CQ
- ▶ **Preprocessing:**
Build a suitable data structure that represents D and $\varphi(D)$
- ▶ **Output:**
For Boolean queries:
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Main result: This is possible $\iff \varphi$ is q-hierarchical.

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Definition: A CQ $\varphi(z_1, \dots, z_k)$ is **q-hierarchical** if for all variables x, y of φ the following is satisfied:

- (i) $\text{atoms}(x) \subseteq \text{atoms}(y)$ or $\text{atoms}(y) \subseteq \text{atoms}(x)$ or $\text{atoms}(x) \cap \text{atoms}(y) = \emptyset$, and
- (ii) if $\text{atoms}(x) \subsetneq \text{atoms}(y)$ and $x \in \text{free}(\varphi)$, then $y \in \text{free}(\varphi)$.

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q-hierarchical CQs are hierarchical CQs where, additionally, the quantifiers respect the query's hierarchical form.

Definition: A CQ $\varphi(z_1, \dots, z_k)$ is **q-hierarchical** if for all variables x, y of φ the following is satisfied:

- (i) $\text{atoms}(x) \subseteq \text{atoms}(y)$ or $\text{atoms}(y) \subseteq \text{atoms}(x)$ or $\text{atoms}(x) \cap \text{atoms}(y) = \emptyset$, and
- (ii) if $\text{atoms}(x) \subsetneq \text{atoms}(y)$ and $x \in \text{free}(\varphi)$, then $y \in \text{free}(\varphi)$.

Queries that are not q-hierarchical:

$$\psi_{S-E-T}() := \exists x \exists y (S(x) \wedge E(x, y) \wedge T(y))$$



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$$\begin{aligned}\psi_{S-E-T}() &:= \exists x \exists y (S(x) \wedge E(x, y) \wedge T(y)) \\ \varphi_{S-E-T}(x, y) &:= S(x) \wedge E(x, y) \wedge T(y)\end{aligned}$$



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A q-hierarchical query:

$$\theta_{E-T}(y) := \exists x (E(x, y) \wedge T(y))$$



▶ **Input:**

data complexity

- ▶ Database D arbitrary
- ▶ query $\varphi(x_1, \dots, x_k)$ CQ

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The OMv-problem: [Henzinger et al., STOC'15]

Input: a Boolean $n \times n$ -matrix M and
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Task: output Mv_ℓ before accessing $v_{\ell+1}$

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Let $\epsilon > 0$ and let $\varphi(\bar{x})$ be a self-join free CQ that is not q-hierarchical.

Then, there is no algorithm with arbitrary preprocessing time and $t_u = O(n^{1-\epsilon})$ update time that enumerates $\varphi(D)$ with $t_d = O(n^{1-\epsilon})$ delay, unless the **OMv-conjecture** fails.

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Proof idea for $\varphi_{E-T}(x) := \exists y (E(x, y) \wedge T(y))$

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Given $n \times n$ matrix M , let

$$\blacktriangleright E^{D_0} := \{(i, j) \in [n]^2 : M(i, j) = 1\}, \quad T^{D_0} := \emptyset$$

Create data structure for D_0 in time $n^2 \cdot n^{1-\epsilon}$.

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This solves OMv in total time $O(n^{3-\epsilon})$ \nexists

Intractability result for **Boolean** CQs that are not q-hierarchical

The OuMv-problem: [Henzinger et al., STOC'15]

Input: a Boolean $n \times n$ -matrix M and
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Task: output $(u_\ell)^T M v_\ell$ before accessing $u_{\ell+1}, v_{\ell+1}$

OuMv-Conjecture: For every $\epsilon > 0$, there is no algorithm that solves the OuMv-problem in total time $O(n^{3-\epsilon})$

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Fix an $\epsilon > 0$ and let φ be a Boolean CQ whose homomorphic core is not q-hierarchical.

Then, there is no algorithm with arbitrary preprocessing time and $t_u = O(n^{1-\epsilon})$ update time that answers $\varphi(D)$ in time $t_a = O(n^{2-\epsilon})$, unless the **OuMv-conjecture** fails.

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Intractability result for counting CQs that are not q-hierarchical

The OV-problem:

[cf. R. Williams, 2005]

Input: two sets U and V of n Boolean vectors of dimension $d := \lceil \log^2 n \rceil$

Task: decide if there exist $u \in U$ and $v \in V$ with $u^T v = 0$

OV-Conjecture: For every $\epsilon > 0$, there is no algorithm that solves the OV-problem in time $O(n^{2-\epsilon})$

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Theorem (Counting):

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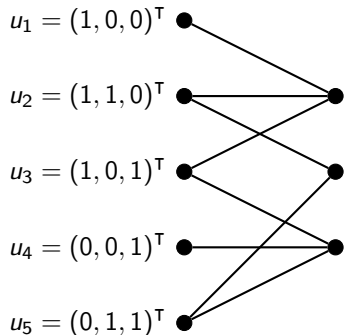
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A lower bound for counting via OV

Left: n vertices for the n vectors $u \in U$

Right: $d := \lceil \log^2 n \rceil$ vertices for vector-coordinates

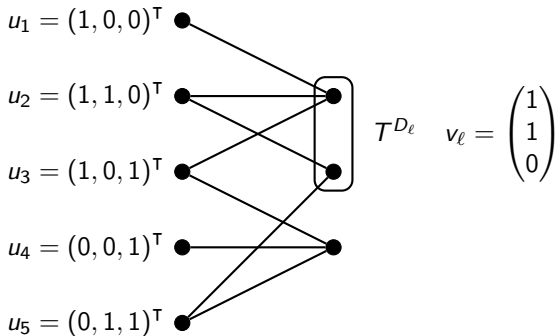


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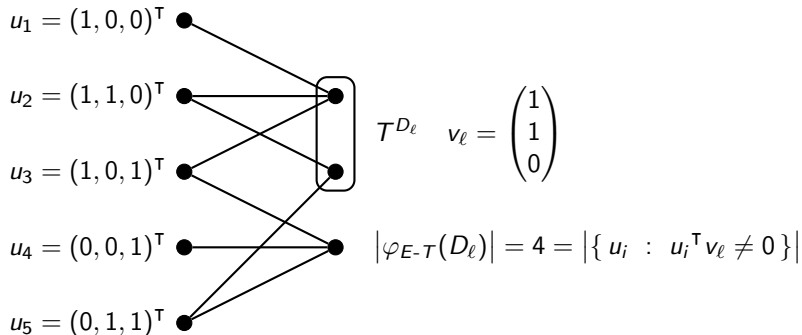
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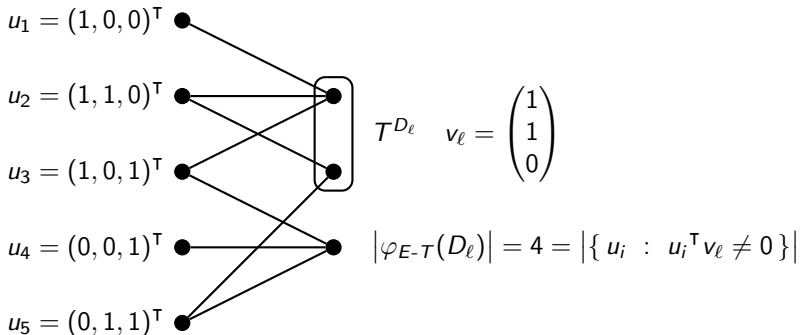
- ▶ for each $v_\ell \in V$: update T^{D_ℓ} in time $d \cdot n^{1-\epsilon} = \lceil \log^2 n \rceil n^{1-\epsilon}$
- ▶ there is $u_i \in U$ with $u_i^T v_\ell = 0 \iff |\varphi_{E-T}(D_\ell)| < n$.

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- ▶ finished for all $v_\ell \in V$ within time $n \cdot \lceil \log^2 n \rceil n^{1-\epsilon} = n^{2-\epsilon'} \neq$

▶ **Input:**

data complexity

- ▶ Database D arbitrary
- ▶ query $\varphi(x_1, \dots, x_k)$ CQ

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For Boolean queries:

- ▶ Decide if $D \models \varphi$

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After every update we want to update the data structure and report the new query result quickly: in time constant or $\text{polylog}(\|D\|)$.

Main result: This is possible $\iff \varphi$ is q-hierarchical.

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After every update we want to update the data structure and report the new query result quickly: in time constant or polylog($\|D\|$).

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Efficient evaluation of a fragment of CQs

Theorem (Upper bound):

For every CQ that is **q-hierarchical**, there is a dynamic data structure that has **constant update time** and allows to

Efficient evaluation of a fragment of CQs

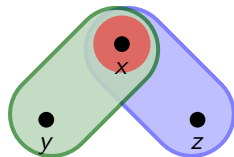
Theorem (Upper bound):

For every CQ that is **q-hierarchical**, there is a dynamic data structure that has **constant update time** and allows to

- ▶ answer a Boolean CQ,
- ▶ count the number of result tuples,
- ▶ enumerate the result relation with constant delay.

q-hierarchical queries

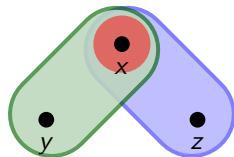
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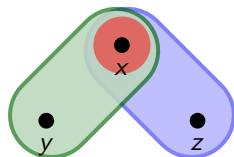
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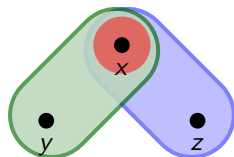


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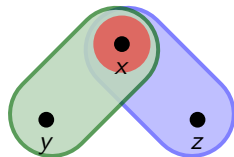


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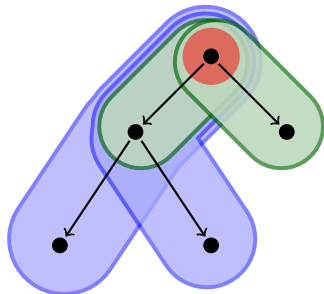
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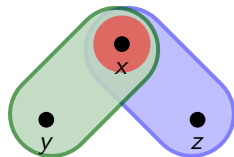
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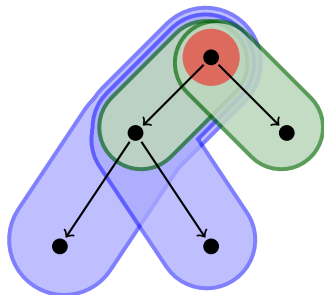
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Definition (q-tree):

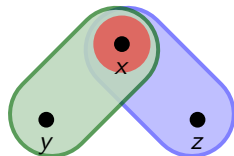
A q-tree T for a CQ $\varphi(x_1, \dots, x_\ell)$ is a rooted tree with $V(T) = \text{vars}(\varphi)$ and



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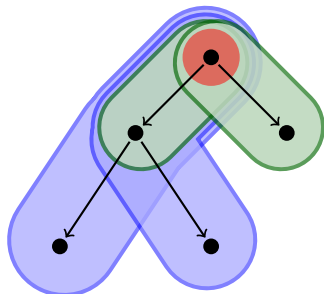


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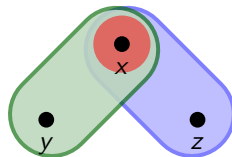
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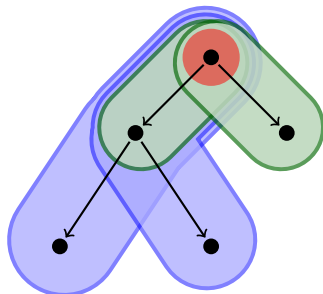


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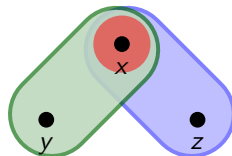
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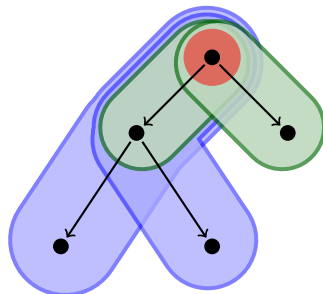


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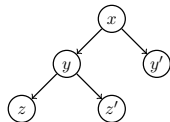
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Lemma: A CQ $\varphi(\bar{x})$ is q-hierarchical \iff every connected component of $\varphi(\bar{x})$ has a q-tree.

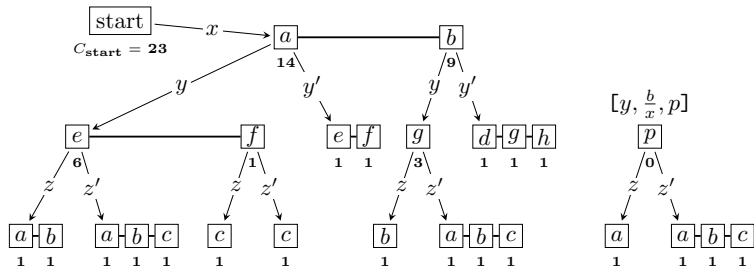
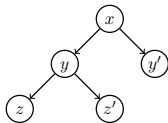
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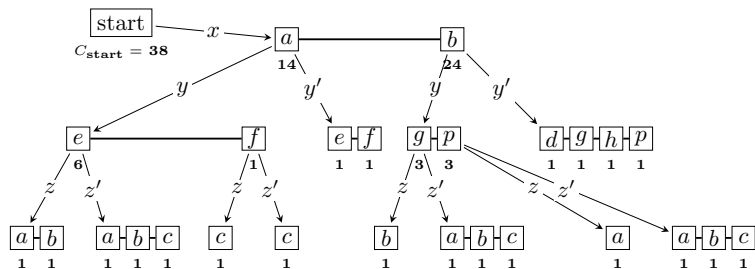
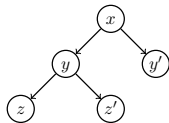
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[Berkholz, Keppeler, S., PODS'17]

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Ongoing work: Similar results for UCQs & FDs.

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Related work: [Idris, Ugarte, Vansummeren, SIGMOD'17]:

q-hierarchical queries are also efficient in practice!

Overview

Introduction

Conjunctive Queries on Arbitrary Databases

First-Order Queries on Bounded Degree Databases

FO+MOD queries and FOC(\mathbb{P}) queries

<i>Movie</i>	
Name	Actor
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Blade Runner	Harrison Ford
Blade Runner	Sean Young
Brazil	Jonathan Pryce
Brazil	Kim Greist
Casablanca	Humphrey Bogart
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Gravity	Sandra Bullock
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Resident Evil	Milla Jovovich
Terminator	Arnold Schwarzenegger
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⋮	⋮

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In FO+MOD:

$\exists^{0 \bmod 2} y \text{ Movie}(y, \text{"Sigourney Weaver"})$

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FO+MOD = extension of first-order logic with modulo-counting quantifiers $\exists^{i \bmod m} y \psi(y, \bar{z})$

FO+MOD queries and FOC(\mathbb{P}) queries

Movie	
Name	Actor
Alien	Sigourney Weaver
Blade Runner	Harrison Ford
Blade Runner	Sean Young
Brazil	Jonathan Pryce
Brazil	Kim Greist
Casablanca	Humphrey Bogart
Casablanca	Ingrid Bergmann
Gravity	Sandra Bullock
Gravity	George Clooney
Resident Evil	Milla Jovovich
Terminator	Arnold Schwarzenegger
Terminator	Linda Hamilton
Terminator	Michael Biehn
⋮	⋮

Is the number of movies with Sigourney Weaver even?

In FO+MOD:

$$\exists^{0 \bmod 2} y \text{ Movie}(y, \text{"Sigourney Weaver"})$$

In FOC(\mathbb{P}):

$$P_{\text{even}}(\#(y).\text{Movie}(y, \text{"Sigourney Weaver"}))$$

FO+MOD = extension of first-order logic with modulo-counting quantifiers $\exists^{i \bmod m} y \psi(y, \bar{z})$

Let \mathbb{P} be a collection of numerical predicates. E.g., \mathbb{P} may contain the predicates $\llbracket P_{\text{even}} \rrbracket = \{i \in \mathbb{Z} : i \text{ is even}\}$ and $\llbracket P_{\leq} \rrbracket = \{(i, j) \in \mathbb{Z}^2 : i \leq j\}$.

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FOC(\mathbb{P}) = extension of first-order logic with formulas of the form $P(t_1, \dots, t_r)$ for $P \in \mathbb{P}$ of arity r , and where each t_i is a **counting term** built using integers, $+$, \cdot , and basic counting terms $t(\bar{x})$ of the form $\#\bar{y}.\psi(\bar{x}, \bar{y})$

Bounded degree databases

Graph $G = (V, E)$:

degree of a node v : the number of neighbours of v in G
degree of G : $\max \{\text{degree}(v) : v \in V\}$

Database D :

degree of D : degree of the Gaifman graph of D

Gaifman graph of D :

the graph $G = (V, E)$ with $V = \text{adom}(D)$ and an edge between two distinct nodes $a, b \in V$ iff some tuple in some relation of D contains a and b

Known results for the static setting (i.e., without updates)

FO query evaluation on dbs of degree $\leq d$

Boolean queries:

- ▶ evaluation in linear time

(Seese 1996)

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- ▶ evaluation in time $f(\varphi, d)\|D\|$, for (Frick, Grohe 2004)

$$f(\varphi, d) = 2^{d^{2^{O(\|\varphi\|)}}} = 3\text{-exp}(\|\varphi\| + \lg \lg d)$$

and the 3-fold exponential blow-up is unavoidable assuming $\text{FPT} \neq \text{AW}[*]$.

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Similar results for other classes of databases

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New: Generalisation to the **dynamic setting** and **FO+MOD**
[Berkholz, Keppeler, S., ICDT'17] and **FOC(\mathbb{P})** [Kuske, S., LICS'17]

▶ Input:

- ▶ Database D of degree $\leq d$
- ▶ query $\varphi(x_1, \dots, x_k)$ in $\text{FOC}(\mathbb{P})[\sigma]$

▶ Preprocessing:

Build a suitable data structure that represents D and $\varphi(D)$

▶ Output:

For Boolean queries:

- ▶ Decide if $D \models \varphi$

For k -ary queries:

- ▶ Compute the number of tuples in $\varphi(D)$
- ▶ Test for a given tuple \bar{a} whether $\bar{a} \in \varphi(D)$
- ▶ Enumerate the tuples in $\varphi(D)$

▶ Dynamic setting:

Tuples may be inserted into or deleted from D

- ▶ **Input:** data complexity
 - ▶ Database D of degree $\leq d$
 - ▶ query $\varphi(x_1, \dots, x_k)$ in $\text{FOC}(\mathbb{P})[\sigma]$
- ▶ **Preprocessing:** in time $O(\|D\|)$
 Build a suitable data structure that represents D and $\varphi(D)$
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 - ▶ Decide if $D \models \varphi$
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 - For Boolean queries:
 - ▶ Decide if $D \models \varphi$ in constant time
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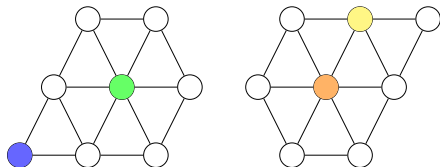
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Proof method: use Hanf normal form for FO+MOD

Hanf normal form for FO+MOD

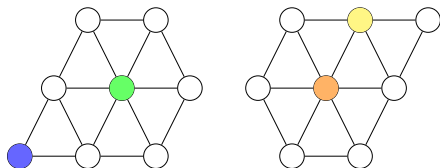
- ▶ A type τ with k centres and radius r :



Example type with
 $k = 4$ centres
and radius $r = 1$

Hanf normal form for FO+MOD

- ▶ A type τ with k centres and radius r :



Example type with
 $k = 4$ centres
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- ▶ $\mathcal{N}_r^D(\bar{b})$ is the induced substructure of D on

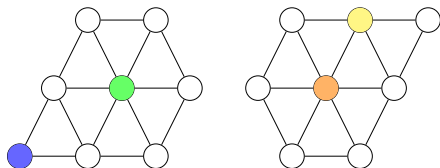
$$N_r^D(\bar{b}) = N_r^D(b_1) \cup \dots \cup N_r^D(b_k)$$

where

$$N_r^D(b_i) = \{a \in \text{adom}(D) : \text{dist}^D(b_i, a) \leq r\}$$

Hanf normal form for FO+MOD

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where

$$\mathcal{N}_r^D(b_i) = \{a \in \text{adom}(D) : \text{dist}^D(b_i, a) \leq r\}$$

- ▶ Sphere-formula $\text{sph}_\tau(\bar{x})$:

$$(D, \bar{a}) \models \text{sph}_\tau(\bar{x}) \iff (\mathcal{N}_r^D(\bar{a}), \bar{a}) \cong \tau$$

Hanf normal form for FO+MOD

A **Hanf normal form** $\psi(\bar{x})$ is a Boolean combination of

- ▶ sphere-formulas $\text{sph}_\rho(\bar{x})$ and
- ▶ Hanf-sentences $\exists^{\geq m} y \text{sph}_\tau(y)$ and $\exists^{i \bmod m} y \text{sph}_\tau(y)$
where τ is a type with 1 centre and radius r .

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Two queries $\varphi(\bar{x})$ and $\psi(\bar{x})$ are **d -equivalent** iff

$$(D, \bar{a}) \models \varphi \iff (D, \bar{a}) \models \psi$$

for all dbs D of degree $\leq d$.

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Theorem (Heimberg, Kuske, S., LICS'16)

There is an algorithm which receives as input a degree bound $d \geq 2$ and a FO+MOD[σ]-formula $\varphi(\bar{x})$, and constructs a d -equivalent formula $\psi(\bar{x})$ in Hanf normal form.

The algorithm's runtime is $f(\varphi, d) = 3\text{-exp}(\|\varphi\| + \lg \lg d)$.

Main result for Boolean queries

Theorem

There is a dynamic algorithm that receives as input

- ▶ *a degree bound $d \geq 2$,*
- ▶ *a Boolean FO+MOD[σ]-query φ , and*
- ▶ *a db D of degree $\leq d$,*

and computes

- ▶ *within $f(\varphi, d)\|D\|$ preprocessing time a data structure*
- ▶ *that can be updated in time $f(\varphi, d)$*

and allows to return the query result $\varphi(D)$ with answer time $O(1)$.

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Proof Idea: *Step 1: Transform φ into Hanf normal form ψ .*

Proof idea (by example)

$$\psi = \exists^{0 \bmod 2} y \text{ sph}_{\tau}(y) \wedge \exists^{0 \bmod 2} y \text{ sph}_{\rho}(y)$$

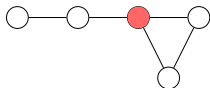
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Let τ be the type with 1 center and radius 2:



Let ρ be the type with 1 center and radius 2:



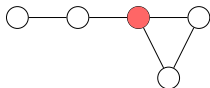
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Data structure: $A[\tau] = 0$, $A[\rho] = 0$

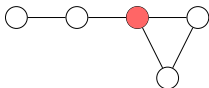
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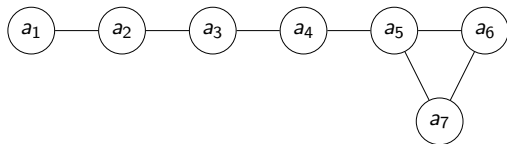


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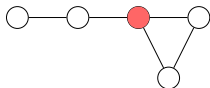
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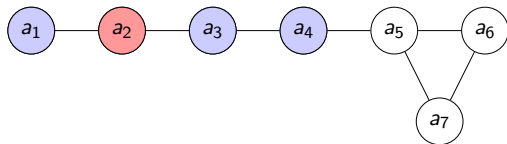


Let ρ be the type with 1 center and radius 2:



Data structure: $A[\tau] = 1$, $A[\rho] = 0$

Database:



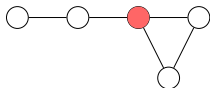
Proof idea (by example)

$$\psi = \exists^{0 \bmod 2} y \text{ sph}_\tau(y) \wedge \exists^{0 \bmod 2} y \text{ sph}_\rho(y)$$

Let τ be the type with 1 center and radius 2:

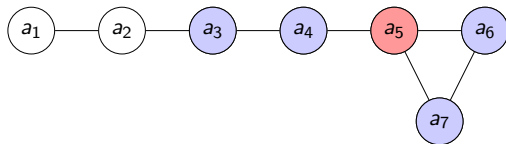


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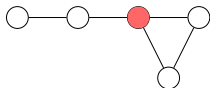
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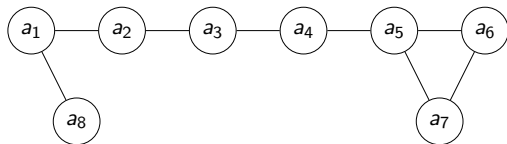


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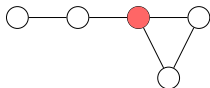
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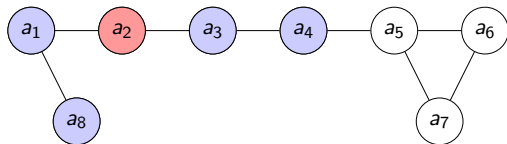


Let ρ be the type with 1 center and radius 2:



Data structure: $A[\tau] = 0$, $A[\rho] = 1$

Database:



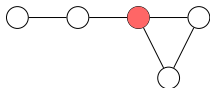
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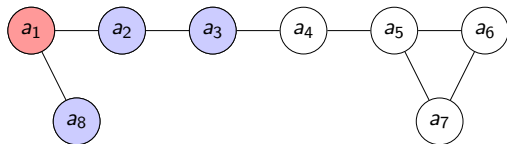


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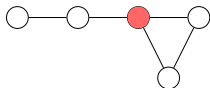
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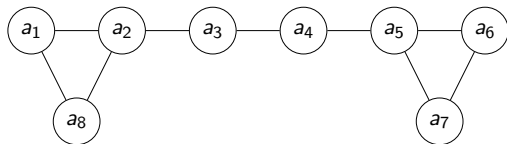


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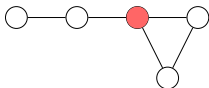
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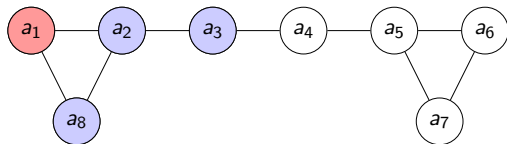


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Data structure: $A[\tau] = 0$, $A[\rho] = 1$

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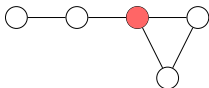
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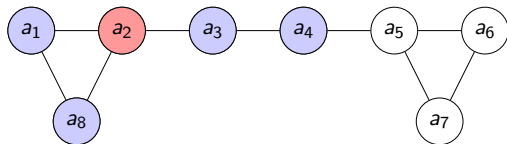


Let ρ be the type with 1 center and radius 2:



Data structure: $A[\tau] = 0$, $A[\rho] = 2$

Database:



Main result for Boolean queries

Theorem

There is a dynamic algorithm that receives as input

- ▶ *a degree bound $d \geq 2$,*
- ▶ *a Boolean FO+MOD[σ]-query φ , and*
- ▶ *a db D of degree $\leq d$,*

and computes

- ▶ *within $f(\varphi, d)\|D\|$ preprocessing time a data structure*
- ▶ *that can be updated in time $f(\varphi, d)$*

and allows to return the query result $\varphi(D)$ with answer time $O(1)$.

$$f(\varphi, d) = 3 \cdot \exp(\|\varphi\| + \lg \lg d)$$

Main result for enumeration

Theorem

There is a dynamic algorithm that receives as input

- ▶ *a degree bound $d \geq 2$,*
- ▶ *a k -ary FO+MOD[σ]-query $\varphi(\bar{x})$, and*
- ▶ *a db D of degree $\leq d$,*

and computes

- ▶ *within $f(\varphi, d)\|D\|$ preprocessing time a data structure*
- ▶ *that can be updated in time $f(\varphi, d)$*

and allows to enumerate $\varphi(D)$ with delay $O(k^3)$.

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Proof Idea:

Proof idea: Reduction to coloured graphs

Input:

Database D

FO+MOD-query $\varphi(x_1, \dots, x_k)$

Same approach as in [Durand, S., Segoufin, PODS'14],
but now we have to take care of updates!

Proof idea: Reduction to coloured graphs

Input:
Database D
FO+MOD-query $\varphi(x_1, \dots, x_k)$

$\sigma_k := \{E, C_1, \dots, C_k\}$
 σ_k -structure \mathcal{G}

$\bar{v} \in \psi_k(\mathcal{G})$

$$\psi_k(x_1, \dots, x_k) := \bigwedge_{i=1}^k C_i(x_i) \wedge \bigwedge_{i \neq j} \neg E(x_i, x_j)$$

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Proof idea: Reduction to coloured graphs

Input:
Database D
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Enumerate:
 $\bar{a} \in \varphi(D)$

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Representing Databases by Coloured Graphs

$$\varphi(x_1, \dots, x_k) \equiv_d \bigvee_{i \in \mathcal{I}} \text{sph}_{\tau_i}(x_1, \dots, x_k) \quad \& \text{ sentences}$$

Representing Databases by Coloured Graphs

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Updating the graph (1)

Claim

If D_{new} is obtained from D_{old} by one update step, then $\mathcal{G}_{D_{new}}$ can be obtained from $\mathcal{G}_{D_{old}}$ by $d^{\mathcal{O}(k^2r+k\|\sigma\|)}$ update steps and additional computing time $2^{\mathcal{O}(\|\sigma\|k^2d^{2r+2})}$.

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- ▶ Updating the colours:
Before: $C_j = C_j^{\mathcal{G}_{D_{old}}}$
 - 1: **for** $j = 1$ **to** c **do**
 - 2: **for** every tuple $\bar{b} \in \bigcup_{\ell=1}^k U^\ell$ **do**
 - 3: **if** $(\mathcal{N}_r^{D_{new}}(\bar{b}), \bar{b}) \cong \tau_j$ **then** $C_j \leftarrow C_j \cup \{v_{\bar{b}}\}$
 - 4: **else** $C_j \leftarrow C_j \setminus \{v_{\bar{b}}\}$

Afterwards: $C_j = C_j^{\mathcal{G}_{D_{new}}}$

Updating the graph (2)

► $E^{\mathcal{G}^D} := \{(v_{\bar{a}}, v_{\bar{b}}) \in V^2 : \text{dist}^D(\bar{a}, \bar{b}) \leq 2r + 1\}$

Updating the graph (2)

▶ $E^{\mathcal{G}_D} := \{(v_{\bar{a}}, v_{\bar{b}}) \in V^2 : \text{dist}^D(\bar{a}, \bar{b}) \leq 2r + 1\}$

▶ Updating the edges:

Before: $E = E^{\mathcal{G}_D_{old}}$

```
1: for every tuple  $\bar{b} \in \bigcup_{\ell=1}^k U^\ell$  do
2:   for every tuple  $\bar{b}' \in \bigcup_{\ell=1}^k U^\ell$  do
3:     if condition (1), (2) and (3) holds then
4:        $E \leftarrow E \cup \{(v_{\bar{b}}, v_{\bar{b}'})\}$ 
5:     else
6:        $E \leftarrow E \setminus \{(v_{\bar{b}}, v_{\bar{b}'})\}$ 
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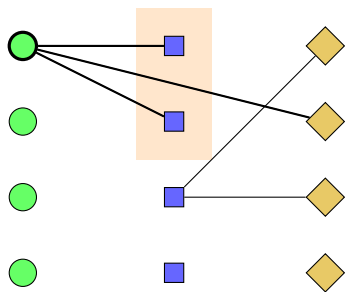
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▶ Conditions:

- (1) There is a $j \in \{1, \dots, c\}$ such that $\bar{b} \in C_j^{\mathcal{G}}$
- (2) There is a $j' \in \{1, \dots, c\}$ such that $\bar{b}' \in C_{j'}^{\mathcal{G}}$
- (3) $\text{dist}^{D_{new}}(\bar{b}, \bar{b}') \leq 2r + 1$

Enumeration with delay $O(k^3d)$

$$\psi_k(x_1, \dots, x_k) := \bigwedge_{i=1}^k C_i(x_i) \wedge \bigwedge_{i \neq j} \neg E(x_i, x_j)$$



for all $u_1 \in C_1^G$ do
 Enum(u_1).

Output **EOE**.

function ENUM(u_1, \dots, u_i)
 if $i = k$ then

 Output (u_1, \dots, u_i)

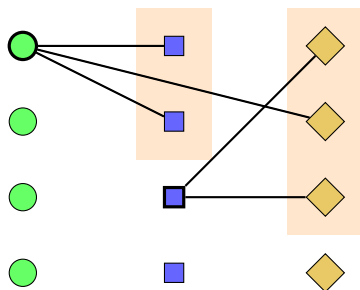
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 if $u_{i+1} \notin \bigcup_{j=1}^i N^G(u_j)$ then
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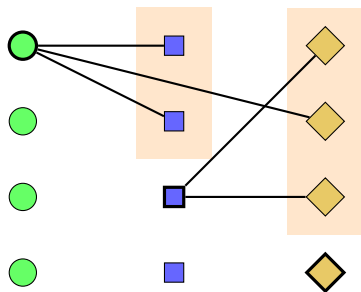
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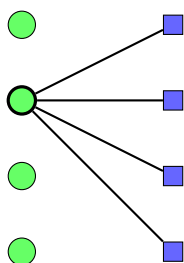
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 for all $u_{i+1} \in C_{i+1}^G$ do

 if $u_{i+1} \notin \bigcup_{j=1}^i N^G(u_j)$ then
 Enum(u_1, \dots, u_i, u_{i+1})

Enumeration with delay $O(k^3d)$

$$\psi_k(x_1, \dots, x_k) := \bigwedge_{i=1}^k C_i(x_i) \wedge \bigwedge_{i \neq j} \neg E(x_i, x_j)$$



for all $u_1 \in C_1^G$ do

 Enum(u_1).



Output EOE.



function ENUM(u_1, \dots, u_i)

 if $i = k$ then

 Output (u_1, \dots, u_i)

 else

 for all $u_{i+1} \in C_{i+1}^G$ do

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Problem: Too few blue nodes

Handling small colours

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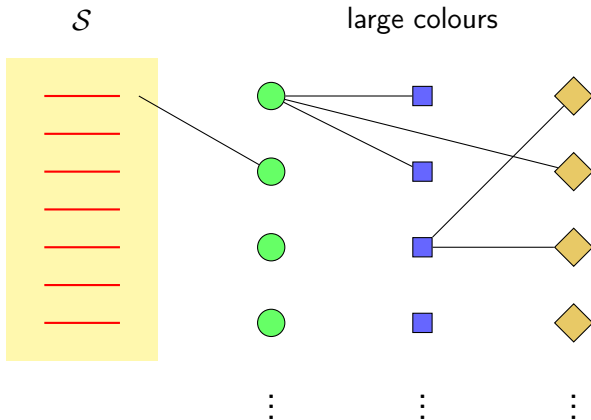
$$\bar{s} \in \mathcal{S} \iff \text{ex. } \bar{a} \text{ such that } (\bar{s}, \bar{a}) \in \varphi(D)$$

The enumeration procedure

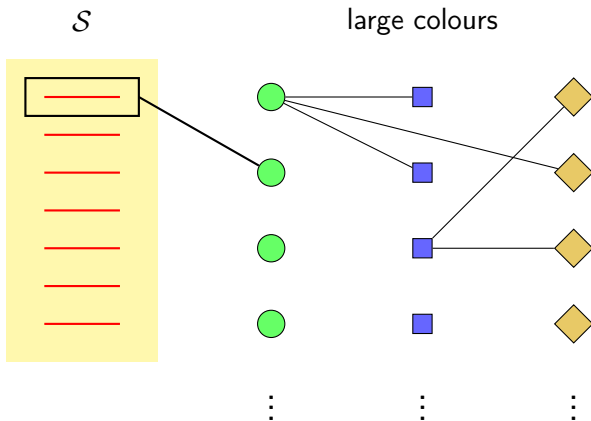
- 1: **for all** $(u_1, \dots, u_s) \in \mathcal{S}$ **do**
- 2: **Enum** (u_1, \dots, u_s) .
- 3: Output the end-of-enumeration message **EOE**.
- 4:
- 5: **function** **ENUM** (u_1, \dots, u_i)
- 6: **if** $i = k$ **then**
- 7: output the tuple (u_1, \dots, u_i)
- 8: **else**
- 9: **for all** $u_{i+1} \in C_{i+1}^{\mathcal{G}}$ **do**
- 10: **if** $u_{i+1} \notin \bigcup_{j=1}^i N^{\mathcal{G}}(u_j)$ **then**
- 11: **Enum** $(u_1, \dots, u_i, u_{i+1})$

where $N^{\mathcal{G}}(u_j) := \{v \in V^{\mathcal{G}} : (u_j, v) \in E^{\mathcal{G}}\}$.

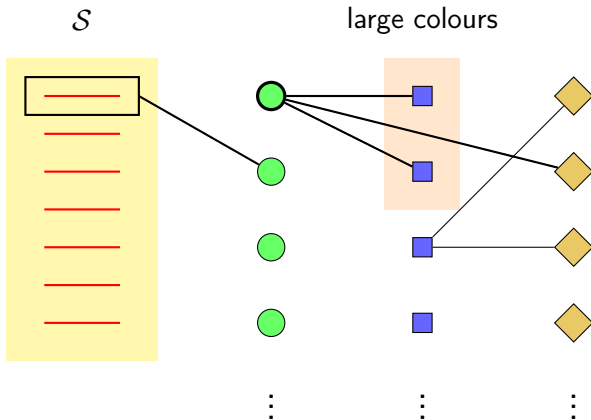
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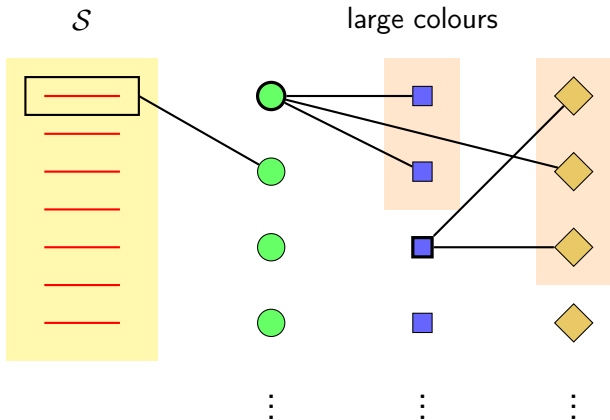
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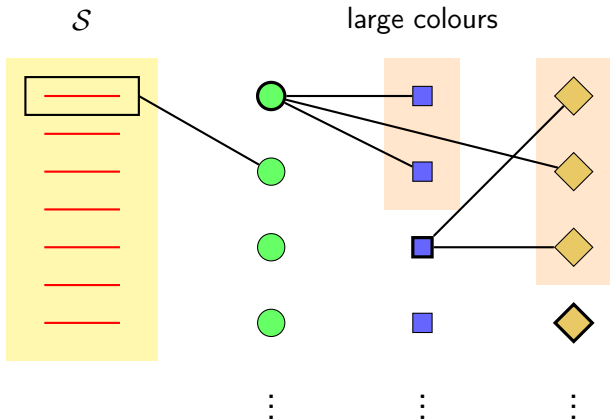
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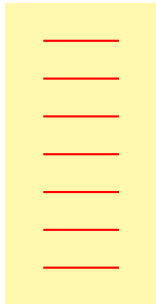


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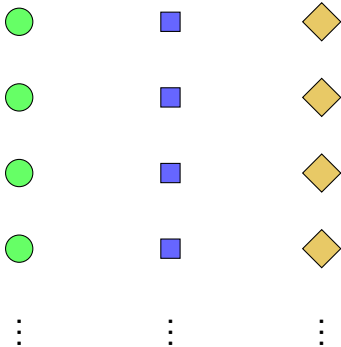


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\mathcal{S}



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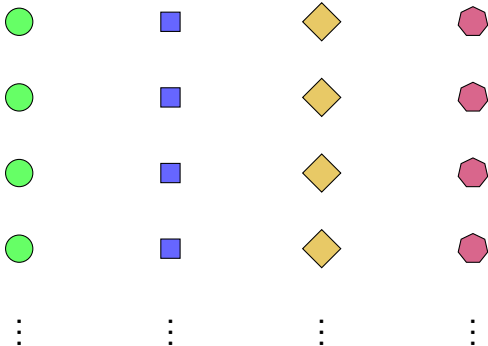


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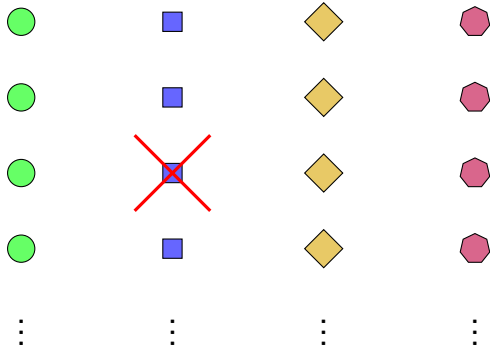


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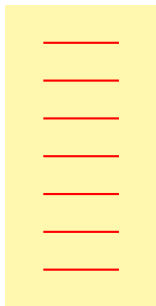


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large colours



⋮

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Main result for enumeration

Theorem

There is a dynamic algorithm that receives as input

- ▶ *a degree bound $d \geq 2$,*
- ▶ *a k -ary FO+MOD[σ]-query $\varphi(\bar{x})$, and*
- ▶ *a db D of degree $\leq d$,*

and computes

- ▶ *within $f(\varphi, d)\|D\|$ preprocessing time a data structure*
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[Berkholz, Keppeler, S., ICDT'17]

- ▶ **Input:** combined complexity
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