A Dichotomy

for Queries with Negation

in Probabilistic Databases



Dan Olteanu Joint work with Robert Fink

Uncertainty in Computation Simons Institute for the Theory of Computing Berkeley Oct 5, 2016

# Outline



#### Probabilistic Databases 101

The Dichotomy

The Interesting but Hard Queries

The Easy Queries

Leftovers

### Tuple-Independent Probabilistic Databases

Tuple-independent database of *n* tuples  $(t_i)_{i \in [n]}$ :

- Each tuple *t<sub>i</sub>* associated with an independent Boolean random variable *x<sub>i</sub>*.
- $P(x_i = true)$  gives the probability that  $t_i$  exists in the database.

Possible-worlds semantics:

- Each possible world defined by an assignment  $\theta$  of the variables  $(x_i)_{i \in [n]}$ :
  - It consists of all tuples  $t_i$  for which  $\theta(x_i) = true$ .
  - It has probability  $P(\theta) = \prod_{i \in [n]} P(x_i = \theta(x_i))$ .
- A tuple-independent database with *n* tuples has 2<sup>*n*</sup> possible worlds.

### Relational Algebra

Popular database query language since Codd times.

- Algebra carrier: set of all finite relations
- Algebra operations:  $\pi$  (projection),  $\times$  (Cartesian product), (set difference),  $\bowtie$  (join),  $\sigma$  (selection),  $\cup$  (set union),  $\delta$  (renaming)
- As expressive as domain relational calculus (RC)

#### In this talk: Relational algebra fragment 1RA<sup>-</sup>

- Included: Equality joins, selections, projections, difference
- Excluded: Repeating relation symbols, unions

#### Relational Algebra

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- Excluded: Repeating relation symbols, unions

Examples of (Boolean) 1RA<sup>-</sup> queries:

• Are there combinations of tuples in (R, T) that are not in (U, V)?

 $\pi_{\emptyset} [ (R(A) \times T(B)) - (U(A) \times V(B)) ]$  $\exists_{A} \exists_{B} [ (R(A) \wedge T(B)) \land \neg (U(A) \wedge V(B)) ]$ (in RC)

Does relation S "hold hands" with both R and T?

 $\pi_{\emptyset} [R(A) \bowtie S(A, B) \bowtie T(B)]$  $\exists_{A} \exists_{B} [R(A) \land S(A, B) \land T(B)]$ (in RC)

### The Query Evaluation Problem

For any Boolean  $1RA^-$  query Q and tuple-independent database D:

Compute the probability that Q is true in a random world of D.

The case of non-Boolean queries can be reduced to the Boolean case.

We are interested in the *data complexity* of this problem.

Fix the query Q and take the database D as input.

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Data complexity of any  $1RA^-$  query Q in tuple-independent databases:

Polynomial time if Q is hierarchical and #P-hard otherwise.

(Boolean)  $1RA^-$  query Q is hierarchical if

- For every pair of distinct query variables A and B in Q,
- there is no triple of relation symbols R, S, and T in Q such that:
- **R** has A but not B, S has both A and B, and T has B but not A.

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R(A) S(A, B) T(B)

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- R has A but not B, S has both A and B, and T has B but not A.





## Examples

Hierarchical queries:

$$\pi_{\emptyset} [(R(A) \bowtie S(A, B)) - T(A, B)]$$

$$\pi_{\emptyset} [(R(A) \times T(B)) - (U(A) \times V(B))]$$

$$\pi_{\emptyset} [(M(A) \times N(B)) - [(R(A) \times T(B)) - (U(A) \times V(B))]]$$

$$\pi_{\emptyset} [\pi_{A} [M(A) \times N(B)] - \pi_{A} [(R(A) \times T(B)) - (U(A) \times V(B))]]$$

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Non-hierarchical queries:

$$\pi_{\emptyset} \left[ R(A) \bowtie S(A, B) \bowtie T(B) \right]$$

$$\pi_{\emptyset} \left[ \pi_{B} \left( R(A) \bowtie S(A, B) \right) - T(B) \right]$$

$$\pi_{\emptyset} \left[ T(B) - \pi_{B} \left( R(A) \bowtie S(A, B) \right) \right]$$

$$\pi_{\emptyset} \left[ X(A) \bowtie \left[ R(A) - \pi_{A} \left( T(B) \bowtie S(A, B) \right) \right] \right]$$

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#### Hardness Proof Idea

Reduction from #P-hard model counting problem for positive bipartite DNF:

- Given a non-hierarchical  $1RA^-$  query Q and
- Any positive bipartite DNF formula Ψ over disjoint sets **X** and **Y** of random variables.
- $\#\Psi$  can be computed using linearly many calls to an oracle for P(Q), where Q is evaluated on tuple-independent databases of sizes linear in the size of  $\Psi$ .

#### A Simple Case

Input formula and query:

• 
$$\Psi = x_1 y_1 \lor x_1 y_2 \lor x_2 y_1$$
 over sets  $\mathbf{X} = \{x_1, x_2\}, \mathbf{Y} = \{y_1, y_2\}$   
•  $Q = \pi_{\emptyset} \Big[ R(A) \bowtie S(A, B) \bowtie T(B) \Big]$ 

Construct a database D such that  $\Psi$  becomes the grounding of Q wrt D:

Column Φ holds formulas over random variables.

• We use  $\top$  for *true* and  $\bot$  for *false*.

Variables also used as constants for A and B.

■ 
$$S(x_i, y_j, \top)$$
:  $x_i y_j$  is a clause in  $\Psi$ .

•  $R(x_i, x_i)$  and  $T(y_j, y_j)$ :  $x_i$  is a variable in **X** and  $y_j$  is a variable in **Y**.

R	T	5	$R \bowtie S \bowtie T$	$\pi_{\emptyset}[R$	$P \bowtie S \bowtie T$ ]
ΑΦ	ΒΦ	ΑΒΦ	ΑΒΦ		Φ
x1 x1	y1 <i>y</i> 1	$x_1 \; y_1 \top$	x <sub>1</sub> y <sub>1</sub> x <sub>1</sub> y <sub>1</sub>	()	Ψ
$x_2 x_2$	y <sub>2</sub> y <sub>2</sub>	$\mathtt{x_1} \ \mathtt{y_2} \top$	$x_1 y_2 x_1 y_2$		
		$x_2 y_1 \top$	x <sub>2</sub> y <sub>1</sub> x <sub>2</sub> y <sub>1</sub>		

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 over sets  $\mathbf{X} = \{x_1, x_2\}, \mathbf{Y} = \{y_1, y_2\}$   
•  $Q = \pi_{\emptyset} \left[ R(A) \bowtie S(A, B) \bowtie T(B) \right]$ 

Construct a database D such that  $\Psi$  becomes the grounding of Q wrt D:

Column Φ holds formulas over random variables.

• We use  $\top$  for *true* and  $\perp$  for *false*.

■ Variables also used as constants for *A* and *B*.

■ 
$$S(x_i, y_j, \top)$$
:  $x_i y_j$  is a clause in  $\Psi$ .

•  $R(x_i, x_i)$  and  $T(y_j, y_j)$ :  $x_i$  is a variable in **X** and  $y_j$  is a variable in **Y**.

R	T	5	$R \bowtie S \bowtie T$	$\pi_{\emptyset}[R$	$\mathbb{N} \boxtimes S \boxtimes T$ ]
ΑΦ	ΒΦ	ΑΒΦ	ΑΒΦ		Φ
x <sub>1</sub> x <sub>1</sub>	y1 <i>y</i> 1	$x_1 y_1 \top$	x <sub>1</sub> y <sub>1</sub> x <sub>1</sub> y <sub>1</sub>	()	Ψ
$x_2 x_2$	y <sub>2</sub> y <sub>2</sub>	$\mathtt{x_1} \ \mathtt{y_2} \top$	$x_1 y_2 x_1 y_2$		
		$x_2 y_1 \top$	x <sub>2</sub> y <sub>1</sub> x <sub>2</sub> y <sub>1</sub>		

This is the only minimal hard pattern for *positive* 1RA<sup>-</sup> queries!

## A Surprising Case

Input formula and query:

• 
$$\Psi = x_1 y_1 \lor x_1 y_2$$
 over sets  $\mathbf{X} = \{x_1\}, \mathbf{Y} = \{y_1, y_2\}$   
•  $Q = \pi_{\emptyset} \Big[ R(A) - \pi_A (T(B) \bowtie S(A, B)) \Big]$ 

Construct a database D such that  $\Psi$  becomes the grounding of Q wrt D:

■  $S(a, b, \top)$ : Clause *a* has variable *b* in  $\Psi$ .

• R(a, T) and  $T(b, \neg b)$ : *a* is a clause and *b* is a variable in  $\Psi$ .

R	Т	S	$T \bowtie S$	$\pi_A(T \bowtie S)$	R –	$\pi_A(T \bowtie S)$
<b>Α</b> Φ	ΒΦ	<b>ΑΒ</b> Φ	<b>ΑΒ</b> Φ	Α Φ	Α	Φ
1 ⊤ 2 ⊤	$x_1 \neg x_1$ $v_1 \neg v_1$	$ 1 x_1 \top $ $ 1 v_1 \top $	$\begin{array}{c} 1 \\ x_1 \\ \neg x_1 \\ 1 \\ y_1 \\ \neg y_1 \end{array}$	$ \begin{array}{c} 1 \neg x_1 \lor \neg y_1 \\ 2 \neg x_1 \lor \neg y_2 \end{array} $	1 2	X1 Y1 X1 Y2
	$y_2 \neg y_2$	$\begin{array}{c} 2 \\ x_1 \\ 2 \\ y_2 \\ \end{array}$	$\begin{array}{c} 2  \mathbf{x}_1  \neg \mathbf{x}_1 \\ 2  \mathbf{y}_2  \neg \mathbf{y}_2 \end{array}$	1 92		1,2

## A Surprising Case

Input formula and query:

• 
$$\Psi = x_1 y_1 \lor x_1 y_2$$
 over sets  $\mathbf{X} = \{x_1\}, \mathbf{Y} = \{y_1, y_2\}$   
•  $Q = \pi_{\emptyset} \Big[ R(A) - \pi_A (T(B) \bowtie S(A, B)) \Big]$ 

Construct a database D such that  $\Psi$  becomes the grounding of Q wrt D:

- $S(a, b, \top)$ : Clause *a* has variable *b* in  $\Psi$ .
- R(a, T) and  $T(b, \neg b)$ : *a* is a clause and *b* is a variable in  $\Psi$ .

R	<u> </u>	5	$T \bowtie S$	$\pi_A(T \bowtie S)$	R –	$\pi_A(T \bowtie S)$
<b>Α</b> Φ	ΒΦ	ΑΒΦ	ΑΒΦ	ΑΦ	Α	Φ
1 ⊤	$x_1 \neg x_1$	$1 x_1 \top$	$1 x_1 \neg x_1$	$1 \neg x_1 \lor \neg y_1$	1	<i>x</i> <sub>1</sub> <i>y</i> <sub>1</sub>
2 ⊤	$y_1 \neg y_1$	$1 y_1 \top$	$1 y_1 \neg y_1$	$2 \neg x_1 \lor \neg y_2$	2	<i>x</i> <sub>1</sub> <i>y</i> <sub>2</sub>
	y <sub>2</sub> ¬ <i>y</i> <sub>2</sub>	$2 x_1 \top$	$2 x_1 \neg x_1$			
		2 y <sub>2</sub> ⊤	2 y <sub>2</sub> ¬y <sub>2</sub>			

This query is already hard when T is the only probabilistic input relation!

#### A More Involved Case

Input formula and query:

• 
$$\Psi = x_1 y_1 \lor x_1 y_2 \lor x_2 y_1$$
 over sets  $\mathbf{X} = \{x_1, x_2\}, \mathbf{Y} = \{y_1, y_2\}$   
•  $Q = \pi_{\emptyset} \Big[ S(\mathbf{A}, \mathbf{B}) - R(\mathbf{A}) \times T(\mathbf{B}) \Big]$ 

We need a different reduction gadget:

- Use additional random variables  $\mathbf{Z} = \{z_1, \dots, z_{|E|}\}$ , one per clause in  $\Psi = \psi_1 \vee \cdots \vee \psi_{|E|}$ .
- Construct a database *D* such that the grounding of *Q* wrt *D* is  $\neg \Upsilon = \neg \left[ \bigvee_{i=1}^{|E|} \neg z_i \neg \psi_i \right] = \bigwedge_{i=1}^{|E|} (z_i \lor \psi_i).$

R		5	$S - R \times T$	$\pi_{\emptyset} \big[ S - R \times T \big]$
ΑΦ	<u>Β</u> Φ	ΑΒΦ	ΑΒΦ	Φ
x <sub>1</sub> x <sub>1</sub>	y <sub>1</sub> x <sub>1</sub>	$x_1 y_1 \neg z_1$	$x_1 y_1 \neg z_1 \neg (x_1 y_1)$	() $\bigvee_{i=1}^{ E } \neg z_i \neg \psi_i$
x <sub>2</sub> x <sub>2</sub>	y <sub>2</sub> y <sub>2</sub>	$x_1 y_2 \neg z_2$	$x_1 y_2 \neg z_2 \neg (x_1 y_2)$	
		x₂ y₁ ¬ <i>z</i> ₃	$x_2 y_1 \neg z_3 \neg (x_2 y_1)$	

• Compute  $\#\Psi$  using linearly many calls to the oracle for  $P_Q = 1 - P(\Upsilon)$ .

The Small Print (1/2)

•  $\Psi = \bigvee_{(i,j)\in E} x_i y_j = \psi_1 \lor \cdots \lor \psi_{|E|}$  over disjoint variable sets X and Y

• Let  $\Theta$  be the set of assignments of variables  $X \cup Y$  that satisfy  $\Psi$ :

$$\#\Psi = \sum_{ heta \in \Theta: heta \models \Psi} 1.$$

Partition  $\Theta$  into disjoint sets  $\Theta_0 \cup \cdots \cup \Theta_{|E|}$ , such that  $\theta \in \Theta_i$  if and only if  $\theta$  satisfies exactly *i* clauses of  $\Psi$ :

$$\#\Psi = \sum_{\theta \in \Theta_1: \theta \models \Psi} 1 + \dots + \sum_{\theta \in \Theta_{|E|}: \theta \models \Psi} 1 = |\Theta_1| + \dots + |\Theta_{|E|}|.$$

•  $|\Theta_1|, \ldots, |\Theta_{|E|}|$  can be computed using an oracle for  $P_{\Upsilon}$ :

$$\Upsilon = \bigvee_{i=1}^{|\mathcal{E}|} \neg z_i \wedge \neg \psi_i \qquad \text{ or, equivalently } \qquad \neg \Upsilon = \bigwedge_{i=1}^{|\mathcal{E}|} (z_i \vee \psi_i)$$

The Small Print (2/2)

Express the probability of  $\neg \Upsilon = \bigwedge_{i=1}^{|E|} (z_i \lor \psi_i)$  as a function of  $|\Theta_1|, \ldots, |\Theta_{|E|}|$ :

Fix the probabilities of variables in  $X \cup Y$  to 1/2 and of variables in Z to  $p_z$ . Then:

$$P_{\neg\Upsilon} = \sum_{k=0}^{|E|} P\left(\neg\Upsilon \middle| \begin{array}{c} \text{exactly } k \text{ clauses} \\ \text{of } \Psi \text{ are satisfied} \end{array}\right) \cdot P\left( \begin{array}{c} \text{exactly } k \text{ clauses} \\ \text{of } \Psi \text{ are satisfied} \end{array}\right)$$
$$\underbrace{P_z^{|E|-k}}_{p_z^{|E|-k}} \underbrace{\frac{1}{2}^{|\mathbf{X}|+|\mathbf{Y}|} \cdot |\Theta_k|}_{\frac{1}{2} |\mathbf{X}|+|\mathbf{Y}|} \\ = \frac{1}{2} \sum_{k=0}^{|\mathbf{X}|+|\mathbf{Y}|} \sum_{k=0}^{|E|} p_z^{|E|-k} |\Theta_k|$$

This is a polynomial in  $p_z$  of degree |E|, with coefficients  $|\Theta_0|, \ldots, |\Theta_{|E|}|$ .

■ The coefficients can be derived from |E| + 1 pairs (p<sub>z</sub>, P<sub>T</sub>) using Lagrange's polynomial interpolation formula.

• |E| + 1 oracle calls to  $P_{\Upsilon}$  suffice to determine  $\#\Psi = \sum_{i=0}^{|E|} |\Theta_i|$ .

### Hard Query Patterns

There are 48 (!) minimal non-hierarchical query patterns.

- Binary trees with leaves A, AB, and B and inner nodes  $\bowtie$  or -.
  - Some are symmetric and need not be considered separately: A and B can be exchanged, joins are commutative and associative.
  - > Still, many cases left to consider due to the difference operator.



- There is a database construction scheme for each pattern.
- Each non-hierarchical query Q matches a pattern P<sub>x.y</sub>.

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- There is a database construction scheme for each pattern.
- Each non-hierarchical query Q matches a pattern  $P_{x.y.}$

In the absence of negation,  $P_{1.1}$  is the only hard pattern to consider!

#### Non-hierarchical Queries Match Minimal Hard Patterns

Each non-hierarchical query Q matches a pattern  $P_{x.y}$ :

- There is a total mapping from **P**<sub>x,y</sub> to *Q*'s parse tree that
  - ▶ is identity on inner nodes  $\bowtie$  and -,
  - preserves ancestor-descendant relationships,
  - maps leaves to relations: A to R(A); AB to S(A, B); and B to T(B).



The match "preserves" the grounding of the query pattern: Q and P<sub>x,y</sub> have the same grounding for any database using our construction scheme.

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#### Evaluation of Hierarchical 1RA<sup>-</sup> Queries

Approach based on knowledge compilation

- For any database D, the probability P<sub>Q(D)</sub> of a 1RA<sup>−</sup> query Q is the probability P<sub>Ψ</sub> of Q's grounding Ψ.
- Compile  $\Psi$  into OBDD( $\Psi$ ) in polynomial time.
- Compute probability of  $OBDD(\Psi)$  in time linear in its size.

#### Evaluation of Hierarchical 1RA<sup>-</sup> Queries

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- For any database D, the probability P<sub>Q(D)</sub> of a 1RA<sup>−</sup> query Q is the probability P<sub>Ψ</sub> of Q's grounding Ψ.
- Compile  $\Psi$  into OBDD( $\Psi$ ) in polynomial time.
- Compute probability of OBDD( $\Psi$ ) in time linear in its size.

Distinction from existing tractability results [O. & Huang 2008]:

- 1RA<sup>−</sup> without negation: Grounding formulas are read-once.
  - Read-once formulas admit linear-size OBBDs.
- 1RA<sup>-</sup>: Grounding formulas are <u>not</u> read-once.
  - They admit OBBDs of sizes linear in the database size <u>but</u> exponential in the query size.

## The Inner Workings

From hierarchical 1RA<sup>-</sup> to RC-hierarchical  $\exists$ -consistent RC<sup> $\exists$ </sup>:

- Translate query Q into an equivalent disjunction of disjunction-free existential relational calculus queries  $Q_1 \lor \cdots \lor Q_k$ .
  - k can be very large for queries with projection under difference!

#### RC-hierarchical:

- For each  $\exists_X(Q')$ , every relation symbol in Q' has variable X.
  - Each of the disjuncts yields a poly-size OBDD.

#### ∃-consistent:

The nesting order of the quantifiers is the same in  $Q_1, \cdots, Q_k$ .

- All OBDDs have compatible variable orders and their disjunction is a poly-size OBDD.
- The OBDD width grows exponentially with k, its height stays linear in the size of the database.
  - Width = maximum number of edges crossing the section between any two consecutive levels.

## Query Evaluation Example (1/3)

Consider the following query and tuple-independent database:

$$Q = \pi_{\emptyset} \Big[ \big( R(A) \times T(B) \big) - \big( U(A) \times V(B) \big) \Big]$$

R	T	U	V	$R \times T$	$R \times T - U \times V$
ΑΦ	ВΦ	ΑΦ	ВΦ	ΑΒΦ	ΑΒ Φ
1 r <sub>1</sub> 2 r <sub>2</sub>	1 t <sub>1</sub> 2 t <sub>2</sub>	1 u <sub>1</sub> 2 u <sub>2</sub>	1 v <sub>1</sub> 2 v <sub>2</sub>	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$
				2 2 r2 t2	$2 2 r_2 t_2 \neg (u_2 v_2)$

## Query Evaluation Example (1/3)

Consider the following query and tuple-independent database:

$$Q = \pi_{\emptyset} \Big[ \big( R(A) \times T(B) \big) - \big( U(A) \times V(B) \big) \Big]$$

R	<u> </u>	U	V	$R \times T$	$R \times T - U \times V$
ΑΦ	ВΦ	ΑΦ	ВΦ	ΑΒΦ	ΑΒ Φ
1 r <sub>1</sub> 2 r <sub>2</sub>	1 t <sub>1</sub> 2 t <sub>2</sub>	1 u <sub>1</sub> 2 u <sub>2</sub>	1 v <sub>1</sub> 2 v <sub>2</sub>	$ \begin{array}{c} 1 & 1 & r_1 t_1 \\ 1 & 2 & r_1 t_2 \\ 2 & 1 & r_2 t_1 \\ 2 & 2 & r_2 t_2 \end{array} $	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$

The grounding of Q is:

$$\Psi = r_1 \big[ t_1 (\neg u_1 \lor \neg v_1) \lor t_2 (\neg u_1 \lor \neg v_2) \big] \lor r_2 \big[ t_1 (\neg u_2 \lor \neg v_1) \lor t_2 (\neg u_2 \lor \neg v_2) \big].$$

- Variables entangle in  $\Psi$  beyond read-once factorization.
- This is the pivotal intricacy introduced by the difference operator.

Query Evaluation Example (2/3)

 $\mathsf{Translate} \ \ \mathsf{Q} = \pi_{\emptyset} \Big[ \big( \mathsf{R}(\mathsf{A}) \times \mathsf{T}(\mathsf{B}) \big) - \big( \mathsf{U}(\mathsf{A}) \times \mathsf{V}(\mathsf{B}) \big) \Big] \ \mathsf{into} \ \mathsf{RC}^{\exists} :$ 

$$Q_{RC} = \underbrace{\exists_A (R(A) \land \neg U(A)) \land \exists_B T(B)}_{Q_1} \lor \underbrace{\exists_A R(A) \land \exists_B (T(B) \land \neg V(B))}_{Q_2}.$$

Both  $Q_1$  and  $Q_2$  are RC-hierarchical.

•  $Q_1 \lor Q_2$  is  $\exists$ -consistent: Same order  $\exists_A \exists_B$  for  $Q_1$  and  $Q_2$ .

Query grounding:

$$\Psi = \underbrace{(r_1 \neg u_1 \lor r_2 \neg u_2) \land (t_1 \lor t_2)}_{\Psi_1} \lor \underbrace{(r_1 \lor r_2) \land (t_1 \neg v_1 \lor t_2 \neg v_2)}_{\Psi_2}.$$

- Both  $\Psi_1$  and  $\Psi_2$  admit linear-size OBDDs.
- The two OBDDs have compatible orders and their disjunction is an OBDD whose width is the product of the widths of the two OBDDs.

### Query Evaluation Example (3/3)

Compile grounding formula into OBDD:



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## Dichotomies Beyond 1RA-

Some known dichotomies

- Non-repeating CQ, UCQ [Dalvi & Suciu 2004, 2010]
- Quantified queries, ranking queries

[O.& team 2011, 2012]

Non-repeating relational algebra =  $1RA^-$  + union.

- Hierarchical property not enough, consistency also needed.
- $\pi_{\emptyset}[(R(A) \bowtie S_1(A, B) \cup T(B) \bowtie S_2(A, B)) S(A, B)]$  is hard, though it is equivalent to a union of two hierarchical 1RA<sup>-</sup> queries.

Non-repeating relational calculus

- $S(x,y) \land \neg R(x)$  is tractable,  $S(x,y) \land (R(x) \lor T(y))$  is hard.
  - Both are non-repeating, yet not expressible in 1RA<sup>-</sup>.
- Possible (though expensive) approach:
  - ▶ Translate to  $RC^{\exists}$  and check RC-hierarchical and  $\exists$ -consistency.

Full relational algebra (or full relational calculus)

It is undecidable whether the union of two equivalent relational algebra queries, one hard and one tractable, is tractable.

## Thank you!