# CS294-248 Special Topics in Database Theory Unit 9: Datalog

Dan Suciu

University of Washington

#### Announcement

• Project presentations: Thursday, Nov. 30th, 9:30am, Calvin 146

 By Monday: please add your tentative topic here: https://tinyurl.com/43mdvwzy

• You can change the topic later, as you wish.

### Outline

• Today: Basic Datalog

• Thursday: Extensions with Negation

### Review

#### Motivation

• FO and its fragments cannot express simple, "easy" queries:

► Transitive closure

► Parity ("Is |*R*| even?")

• Datalog: extends CQs with recursion

### Datalog Syntax

Review

- A program P = set of rules.
- A rule is a CQ:  $H := A_1 \wedge A_2 \wedge \cdots$
- Extensional Database Predicates
   EDBs
- Intensional Database Predicates IDBs

```
T(X,Y) := E(X,Y)

T(X,Y) := T(X,Z) \land E(Z,X)
```

Poset (partially ordered set)  $(P, \leq)$ . We assume P has a minimal element  $\perp$ .

 $f: P \to P$  is monotone if  $x \le y \Rightarrow f(x) \le f(y)$ .

x is a *pre-fixpoint* if  $f(x) \leq x$ 

x is a post-fixpoint if  $f(x) \ge x$ 

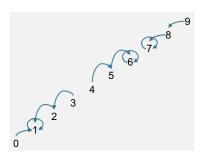
x is a *fixpoint* if f(x) = x;

What are the pre-, post-, fixpoints?

Pre-fixpoints:

Post-fixpoints:

Fixpoints:

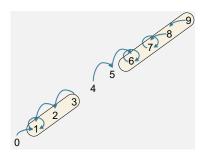


What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints:

Fixpoints:

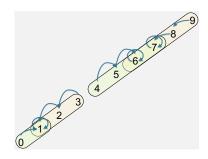


What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints:

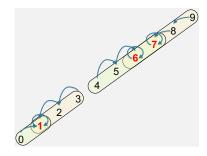


What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7

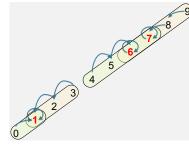


What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



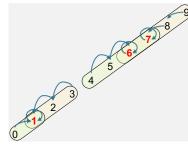
Does every monontone f have a fixpoint? A pre-fixpoint?

What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



Does every monontone f have a fixpoint? A pre-fixpoint?

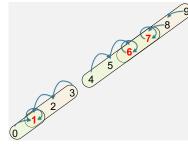
No: 
$$f:(\mathbb{N},\leq)\to(\mathbb{N},\leq)$$
,  $f(x)=x+1$ .

What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



Does every monontone f have a fixpoint? A pre-fixpoint?

No:  $f:(\mathbb{N},\leq)\to(\mathbb{N},\leq)$ , f(x)=x+1.

#### Theorem

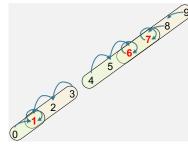
If the least pre-fixpoint exits then the least fixpoint exists and they are equal.

What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



Does every monontone f have a fixpoint? A pre-fixpoint?

No: 
$$f:(\mathbb{N},\leq)\to(\mathbb{N},\leq), f(x)=x+1.$$

#### Theorem

If the least pre-fixpoint exits then the least fixpoint exists and they are equal.

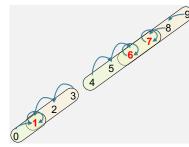
Proof:  $z \stackrel{\text{def}}{=} \text{least pre-fixpoint.}$ 

What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



Does every monontone f have a fixpoint? A pre-fixpoint?

No: 
$$f:(\mathbb{N},\leq)\to(\mathbb{N},\leq), f(x)=x+1.$$

#### Theorem

If the least pre-fixpoint exits then the least fixpoint exists and they are equal.

Proof:  $z \stackrel{\text{def}}{=} \text{least pre-fixpoint.}$ 

$$f(z) \leq z$$

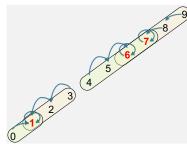
Dan Suciu

What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



Does every monontone f have a fixpoint? A pre-fixpoint?

No: 
$$f:(\mathbb{N},\leq)\to(\mathbb{N},\leq)$$
,  $f(x)=x+1$ .

#### Theorem

If the least pre-fixpoint exits then the least fixpoint exists and they are equal.

Proof:  $z \stackrel{\text{def}}{=} \text{least pre-fixpoint.}$ 

$$f(z) \le z$$
  $f(f(z)) \le f(z)$ 

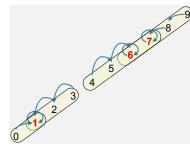
Dan Suciu

What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



Does every monontone f have a fixpoint? A pre-fixpoint?

No: 
$$f:(\mathbb{N},\leq)\to(\mathbb{N},\leq)$$
,  $f(x)=x+1$ .

#### Theorem

If the least pre-fixpoint exits then the least fixpoint exists and they are equal.

Proof:  $z \stackrel{\text{def}}{=} \text{least pre-fixpoint.}$ 

$$f(z) \le z$$
  $f(f(z)) \le f(z)$ 

f(z) pre-fixpoint

Dan Suciu

Topics in DB Theory: Unit 97

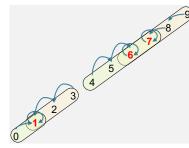
Fall 2023

What are the pre-, post-, fixpoints?

Pre-fixpoints: 1,2,3,6,7,8,9

Post-fixpoints: 0,1,4,5,6,7

Fixpoints: 1,6,7



Does every monontone f have a fixpoint? A pre-fixpoint?

No: 
$$f: (\mathbb{N}, \leq) \to (\mathbb{N}, \leq)$$
,  $f(x) = x + 1$ .

#### Theorem

If the least pre-fixpoint exits then the least fixpoint exists and they are equal.

Proof:  $z \stackrel{\text{def}}{=} \text{least pre-fixpoint.}$ 

$$f(z) \le z$$
  $f(f(z)) \le f(z)$ 

$$f(z)$$
 pre-fixpoint

$$f(z) = z$$

Dan Suciu

### Kleene's Sequence

$$\left| f^{(0)} \stackrel{\mathsf{def}}{=} \bot \qquad f^{(t+1)} \stackrel{\mathsf{def}}{=} f(f^{(t)}) \right| \quad f^{(0)} \le f^{(1)} \le f^{(2)} \le \cdots$$

#### Fact

If z is any pre-fixpoint, then  $f^{(t)} < z$  for all t.

Proof by induction:  $\perp \leq z$  and  $f^{(t+1)} = f(f^{(t)}) \leq f(z) \leq z$ .

### Kleene's Sequence

$$f^{(0)} \stackrel{\mathsf{def}}{=} \bot \qquad f^{(t+1)} \stackrel{\mathsf{def}}{=} f(f^{(t)}) \qquad f^{(0)} \le f^{(1)} \le f^{(2)} \le \cdots$$

#### **Fact**

If z is any pre-fixpoint, then  $f^{(t)} \le z$  for all t.

Proof by induction:  $\bot \le z$  and  $f^{(t+1)} = f(f^{(t)}) \le f(z) \le z$ .

Is  $\bigvee_{t>0} f^{(t)}$  the least fixpoint?

### Kleene's Sequence

$$f^{(0)} \stackrel{\text{def}}{=} \bot \qquad f^{(t+1)} \stackrel{\text{def}}{=} f(f^{(t)}) \qquad f^{(0)} \le f^{(1)} \le f^{(2)} \le \cdots$$

#### **Fact**

If z is any pre-fixpoint, then  $f^{(t)} \le z$  for all t.

Proof by induction:  $\bot \le z$  and  $f^{(t+1)} = f(f^{(t)}) \le f(z) \le z$ .

Is  $\bigvee_{t>0} f^{(t)}$  the least fixpoint?

Not always. Two problems:

- $\bigvee_{t>0} f^{(t)}$  may not exists.
- Even if it exists, we may have  $f(\bigvee_{t\geq 0} f^{(t)}) \neq \bigvee_{t\geq 0} f^{(t)}$ .

We will circumvent by requiring finite rank



[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

The rank of a poset  $(P, \leq)$  is the largest k s.t. there exists a chain of length k in P.

Note: rank may be  $\infty$ .

[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

The rank of a poset  $(P, \leq)$  is the largest k s.t. there exists a chain of length k in P.

Note: rank may be  $\infty$ .

• What is the rank of  $(\{0,1\},\leq)$ ?

[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

The rank of a poset  $(P, \leq)$  is the largest k s.t. there exists a chain of length k in P.

Note: rank may be  $\infty$ .

• What is the rank of  $(\{0,1\},\leq)$ ?

r=1.

[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

The rank of a poset  $(P, \leq)$  is the largest k s.t. there exists a chain of length k in P.

Note: rank may be  $\infty$ .

- What is the rank of  $(\{0,1\},\leq)$ ?
- What is the rank of  $(\mathcal{P}(A), \subseteq)$ ?

r = 1.

Dan Suciu

[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

The rank of a poset  $(P, \leq)$  is the largest k s.t. there exists a chain of length k in P.

Note: rank may be  $\infty$ .

- What is the rank of  $(\{0,1\},\leq)$ ?
- What is the rank of  $(\mathcal{P}(A), \subseteq)$ ?

r = 1.

$$r = |A|$$
.

Dan Suciu

[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

The rank of a poset  $(P, \leq)$  is the largest k s.t. there exists a chain of length k in P.

Note: rank may be  $\infty$ .

• What is the rank of  $(\{0,1\},\leq)$ ?

r = 1.

• What is the rank of  $(\mathcal{P}(A), \subseteq)$ ?

$$r = |A|$$
.

• What is the rank of  $(P_1, <_1) \times (P_2, <_2)$ ?

[Stanley, 1999]

A chain of rank r is a sequence  $x_0 < x_1 < \cdots < x_r$ .

The rank of a poset  $(P, \leq)$  is the largest k s.t. there exists a chain of length k in P.

Note: rank may be  $\infty$ .

- What is the rank of  $(\{0,1\},\leq)$ ?
- What is the rank of  $(\mathcal{P}(A), \subseteq)$ ?
- What is the rank of  $(P_1, \leq_1) \times (P_2, \leq_2)$ ?

$$r = 1$$
.

$$r = |A|$$
.

$$r = r(P_1) + r(P_2).$$

### Fixpoints in Posets of Finite Ranks

$$f^{(0)} \stackrel{\mathsf{def}}{=} \bot \qquad f^{(t+1)} \stackrel{\mathsf{def}}{=} f(f^{(t)}) \qquad f^{(0)} < f^{(1)} < f^{(2)} < \cdots \le f^{(r)} = f^{(r+1)}$$

#### Theorem

If P has finite rank r then  $lfp(f) = f^{(r)}$ .

### Least Fixpoint Semantics of a Datalog Program P

 $I = \text{an EDB instance}, A \stackrel{\text{def}}{=} ADom(I).$ 

If R has arity k, then an instance is  $R \in \mathcal{P}(A^k)$ .

### Least Fixpoint Semantics of a Datalog Program P

 $I = \text{an EDB instance}, A \stackrel{\text{def}}{=} ADom(I).$ 

If R has arity k, then an instance is  $R \in \mathcal{P}(A^k)$ .

If IDB predicates have arities  $k_1, k_2, ...$ then an IDB instance is  $J \in \mathcal{P}(A^{k_1}) \times \mathcal{P}(A^{k_2}) \times \cdots$ 

### Least Fixpoint Semantics of a Datalog Program P

 $I = \text{an EDB instance}, A \stackrel{\text{def}}{=} ADom(I).$ 

If R has arity k, then an instance is  $R \in \mathcal{P}(A^k)$ .

If IDB predicates have arities  $k_1, k_2, ...$ then an IDB instance is  $J \in \mathcal{P}(A^{k_1}) \times \mathcal{P}(A^{k_2}) \times \cdots$ 

#### Immediate Consequence Operator:

$$T_P: \mathcal{P}(A^{k_1}) \times \mathcal{P}(A^{k_2}) \times \cdots \to \mathcal{P}(A^{k_1}) \times \mathcal{P}(A^{k_2}) \times \cdots$$

The semantics of the datalog program P is  $lfp(T_p)$ .

### Naive Evaluation Algorithm

$$J^{(0)}:=\emptyset$$
 for  $t=0,\infty$   $J^{(t+1)}:=T_P(J^{(t)})$  if  $J^{(t+1)}=J^{(t)}$  break

Notice:  $J^{(0)} \subseteq J^{(1)} \subseteq \cdots$  is Kleene's sequence.

#### Theorem

The Naive Algorithm takes  $O(ADom(I)^k)$  iterations, where I is the EDB instance and k is the largest arity of any IDB.

Data complexity is in PTIME.

## Examples in Datalog

#### Overview

• We have seen Transitive Closure. Can we write something different?

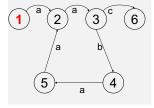
Regular expressions, CFGs.

• Same generation.

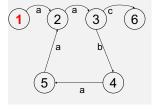
AND/OR reachability.

Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$ 

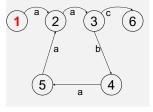
Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$  EDB graph:

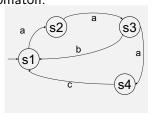


Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$  EDB graph:

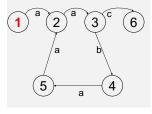


Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$  EDB graph: Automaton:





Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$  EDB graph: Automaton:



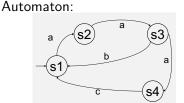
a s2 a s3 a s3 a c s4

EDB graph:

### Regular Expressions

Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$ 

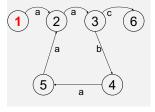
1 a 2 a 3 c 6



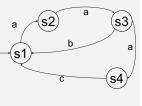
$$Q2(Y) := Q1(X) \wedge E(X, Y, 'a')$$

Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$ 

EDB graph:



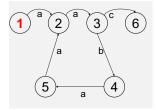
Automaton:

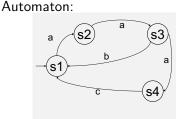


$$Q1(1) := Q3(Y) := Q2(X) \land E(X, Y, 'a')$$

$$Q2(Y) := Q1(X) \wedge E(X, Y, 'a')$$

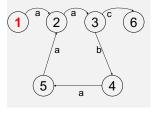
Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$  EDB graph: Automaton:

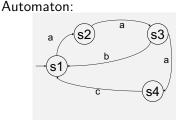




$$Q1(1) := Q2(Y) := Q1(X) \land E(X, Y, 'a')$$
  
 $Q3(Y) := Q2(X) \land E(X, Y, 'a')$   $Q1(Y) := Q3(X) \land E(X, Y, 'b')$ 

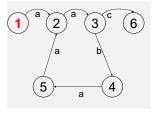
Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$  EDB graph: Automaton:

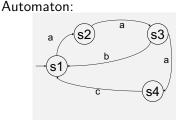




$$Q1(1)$$
:-  $Q2(Y)$ :-  $Q1(X) \wedge E(X, Y, 'a')$   
 $Q3(Y)$ :-  $Q2(X) \wedge E(X, Y, 'a')$   $Q1(Y)$ :-  $Q3(X) \wedge E(X, Y, 'b')$   
 $Q4(Y)$ :-  $Q3(X) \wedge E(X, Y, 'a')$   $Q1(Y)$ :-  $Q4(X) \wedge E(X, Y, 'c')$ 

Find nodes reachable from 1 with a path labeled with  $(aab|aaac)^*$  EDB graph: Automaton:





$$Q1(1):$$
  $Q2(Y):=Q1(X) \land E(X,Y,'a')$   $Q3(Y):=Q2(X) \land E(X,Y,'a')$   $Q1(Y):=Q3(X) \land E(X,Y,'b')$   $Q4(Y):=Q3(X) \land E(X,Y,'a')$   $Q1(Y):=Q4(X) \land E(X,Y,'c')$  Answer(X):=Q1(X)

• Automaton need not be deterministic.

- Automaton need not be deterministic.
- Also CFG. E.g language of parentheses:  $S \to \epsilon |SS| aSb$ . How?

- Automaton need not be deterministic.
- Also CFG. E.g language of parentheses:  $S \to \epsilon |SS| aSb$ . How?

$$S(X,X)$$
 :- Node(X)  
 $S(X,Y)$  :-  $S(X,Z) \land S(Z,Y)$   
 $S(X,Y)$  :-  $E(X,U,'a') \land S(U,V) \land E(V,Y,'b')$ 

- Automaton need not be deterministic.
- Also CFG. E.g language of parentheses:  $S \to \epsilon |SS| aSb$ . How?

$$S(X,X) := Node(X)$$
  
 $S(X,Y) := S(X,Z) \land S(Z,Y)$   
 $S(X,Y) := E(X,U,'a') \land S(U,V) \land E(V,Y,'b')$ 

• Exercise\*\*: non-CFG, e.g. the language  $\{a^nb^nc^n \mid n \in \mathbb{N}\}$ .

- Automaton need not be deterministic.
- Also CFG. E.g language of parentheses:  $S \to \epsilon |SS| aSb$ . How?

$$S(X,X)$$
 :- Node(X)  
 $S(X,Y)$  :-  $S(X,Z) \wedge S(Z,Y)$   
 $S(X,Y)$  :-  $E(X,U,'a') \wedge S(U,V) \wedge E(V,Y,'b')$ 

• Exercise\*\*: non-CFG, e.g. the language  $\{a^nb^nc^n\mid n\in\mathbb{N}\}$ . (won't discuss in class)

$$T(X,X,Y,Y,Z,Z) := \mathsf{Node}(X) \land \mathsf{Node}(Y) \land \mathsf{Node}(Z) \\ T(X_1,X_2,Y_1,Y_2,Z_1,Z_2) := T(X_1,X_3,Y_1,Y_3,Z_1,Z_3) \land E(X_3,X_2,'a') \\ \land E(Y_3,Y_2,'b') \land E(Z_3,Z_2,'c') \\ \mathsf{Answer}(X,Y) := T(X,U,V,U,V,Y)$$

x, y are at the same generation if they have a common ancestor z at the same distance.

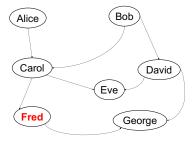
x, y are at the same generation if they have a common ancestor z at the same distance.

Find people at the same generation with Fred.

x, y are at the same generation if they have a common ancestor z at the same distance.

Find people at the same generation with Fred.

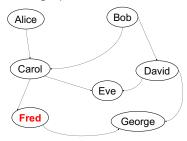
#### EDB graph:



x, y are at the same generation if they have a common ancestor z at the same distance.

Find people at the same generation with Fred.

#### EDB graph:

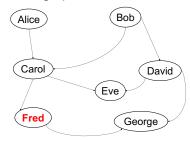


x, y are at the same generation if they have a common ancestor z at the same distance.

Find people at the same generation with Fred.

$$SG(X, X)$$
 :- Person(X)  
 $SG(X, Y)$  :-

#### EDB graph:

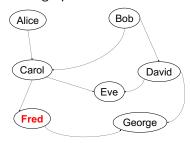


x, y are at the same generation if they have a common ancestor z at the same distance.

Find people at the same generation with Fred.

$$SG(X,X)$$
:- Person(X)  
 $SG(X,Y)$ :- ????

#### EDB graph:

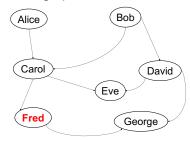


x, y are at the same generation if they have a common ancestor z at the same distance.

Find people at the same generation with Fred.

$$SG(X,X)$$
 :- Person(X)  
 $SG(X,Y)$  :-  $SG(U,V) \wedge E(U,X) \wedge E(V,Y)$ 

#### EDB graph:

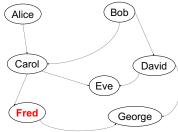


x, y are at the same generation if they have a common ancestor z at the same distance.

Find people at the same generation with Fred.

$$SG(X,X)$$
:-  $Person(X)$   
 $SG(X,Y)$ :-  $SG(U,V) \wedge E(U,X) \wedge E(V,Y)$   
 $Answer(X)$ :-  $SG('Fred',X)$ 

### EDB graph:



• The examples so far are still just transitive at their essence! why?

 Recall that transitive closure is in NLOGSPACE. The next example goes beyond NLOGSPACE.

OR-nodes: unlimited AND-children

AND-nodes: two OR-children

OR-nodes: unlimited AND-children

AND-nodes: two OR-children

T(X, Y, Z):

OR-node X has AND-child with children Y, Z.

OR-nodes: unlimited AND-children

AND-nodes: two OR-children

T(X, Y, Z):

OR-node X has AND-child with children Y, Z.

Find all accessible nodes from a

OR-nodes: unlimited AND-children

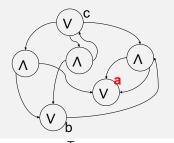
AND-nodes: two OR-children

T(X, Y, Z):

OR-node X has AND-child with children Y, Z.

Find all accessible nodes from a

#### EDB graph:



1		
X	Y	Z
С	а	Ь
С	Ь	С
С	a	a
Ь	a	a

OR-nodes: unlimited AND-children

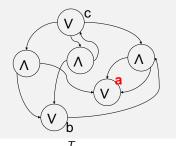
AND-nodes: two OR-children

T(X, Y, Z):

OR-node X has AND-child with children Y, Z.

Find all accessible nodes from a

#### EDB graph:



1		
X	Y	Z
С	а	Ь
С	ь	с
С	a	а
Ь	a	a

Answer: a, b, c.

OR-nodes: unlimited AND-children

AND-nodes: two OR-children

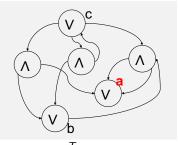
$$T(X, Y, Z)$$
:

OR-node X has AND-child with children Y, Z.

Find all accessible nodes from a

$$A(a)$$
:-
$$A(X) := T(X, Y, Z) \land A(Y) \land A(Z)$$

#### EDB graph:



I		
Χ	Y	Z
С	а	Ь
С	Ь	c
С	a	a
Ь	a	a

Answer: a, b, c.

- AGAP is PTIME-complete. Recall: NLOGSPACE ⊆ PTIME and inclusion is conjecture to be strict.
- It follows that datalog can express strictly more than transitive closure.
- The data complexity of datalog is in PTIME.
- Limitation of "pure" datalog: monotone queries only.
- Montone queries have huge potential for optimizations (next).

# Optimizing Monotone Datalog

### Outline

Semi-naive evaluation.

Asynchronous execution: also discuss grounding.

Will not discuss: Magic Set optimization

### Naive, and Semi-naive

#### Naive

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for } t = 0, \infty \ J^{(t+1)} &:= T_P(J^{(t)}) \ & ext{if } J^{(t+1)} &= J^{(t)} \ & ext{break} \end{aligned}$$

### Naive, and Semi-naive

#### Naive

## $J^{(0)} := \emptyset$

for 
$$t = 0, \infty$$

$$J^{(t+1)} := T_P(J^{(t)})$$
if  $J^{(t+1)} = J^{(t)}$ 

break

#### Semi-naive

$$egin{aligned} egin{aligned} egin{aligned\\ egin{aligned} egi$$

#### Naive

$$J^{(0)} := \emptyset$$
for  $t = 0, \infty$ 
 $J^{(t+1)} := T_P(J^{(t)})$ 
if  $J^{(t+1)} = J^{(t)}$ 

break

#### Semi-naive

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for} \ t = 0, \infty \ & \Delta^{(t)} &:= T_P(J^{(t)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if} \ \Delta^{(t)} &= \emptyset \ ext{break} \end{aligned}$$

#### w/ incremental computation

Optimizing Monotone Datalog

$$egin{aligned} J^{(0)} &:= \emptyset, \quad \Delta^{(0)} := T_P(\emptyset) \ & ext{for } t = 1, \infty \ & \Delta^{(t)} &:= T_P(J^{(t-1)} \cup \Delta^{(t-1)}) - J^{(t)} \ &= \Delta T_P(J^{(t-1)}, \Delta^{(t-1)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} = \emptyset ext{ break} \end{aligned}$$

#### Naive

#### . .

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for} \ \ t = 0, \infty \ & ext{} J^{(t+1)} &:= T_P(J^{(t)}) \ & ext{if} \ \ J^{(t+1)} &= J^{(t)} \end{aligned}$$

break

#### Semi-naive

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for } t = 0, \infty \ & \Delta^{(t)} &:= T_P(J^{(t)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} &= \emptyset ext{ break} \end{aligned}$$

## w/ incremental computation

$$egin{aligned} \overline{J^{(0)}} := \emptyset, & \Delta^{(0)} := T_P(\emptyset) \ & ext{for } t = 1, \infty \ & \Delta^{(t)} := T_P(J^{(t-1)} \cup \Delta^{(t-1)}) - J^{(t)} \ & = \Delta T_P(J^{(t-1)}, \Delta^{(t-1)}) - J^{(t)} \ & J^{(t+1)} := J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} = \emptyset ext{ break} \end{aligned}$$

#### Transitive Closure:

$$T(X,Y) := E(X,Y)$$
  
 $T(X,Y) := T(X,Z) \wedge E(Z,Y)$ 

#### Naive

#### valve

$$J^{(0)}:=\emptyset$$
 for  $t=0,\infty$   $J^{(t+1)}:=T_P(J^{(t)})$  if  $J^{(t+1)}=J^{(t)}$  break

#### Semi-naive

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for} \ \ t = 0, \infty \ & \Delta^{(t)} &:= T_P(J^{(t)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if} \ \ \Delta^{(t)} &= \emptyset \ \ ext{break} \end{aligned}$$

## w/ incremental computation

$$egin{aligned} J^{(0)} &:= \emptyset, \quad \Delta^{(0)} := T_P(\emptyset) \ & ext{for } t = 1, \infty \ & \Delta^{(t)} &:= T_P(J^{(t-1)} \cup \Delta^{(t-1)}) - J^{(t)} \ &= \Delta T_P(J^{(t-1)}, \Delta^{(t-1)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} = \emptyset ext{ break} \end{aligned}$$

Transitive Closure:

$$T^{(0)}(X,Y) := ext{false}, \; \Delta^{(0)}(X,Y) := E(X,Y)$$

$$T(X, Y) := E(X, Y)$$
  
 $T(X, Y) := T(X, Z) \wedge E(Z, Y)$ 

 $T^{(0)}(X,Y) := \text{false}, \ \Delta^{(0)}(X,Y) := E(X,Y)$ 

# Naive, and Semi-naive

#### Naive

## -(0) -

$$J^{(0)}:=\emptyset$$
 for  $t=0,\infty$   $J^{(t+1)}:=T_P(J^{(t)})$  if  $J^{(t+1)}=J^{(t)}$  break

#### Semi-naive

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for} \ t = 0, \infty \ & \Delta^{(t)} &:= \mathcal{T}_P(J^{(t)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if} \ \Delta^{(t)} &= \emptyset \ ext{break} \end{aligned}$$

#### w/ incremental computation

$$egin{aligned} J^{(0)} &:= \emptyset, \quad \Delta^{(0)} := T_P(\emptyset) \ & ext{for } t = 1, \infty \ & \Delta^{(t)} &:= T_P(J^{(t-1)} \cup \Delta^{(t-1)}) - J^{(t)} \ &= \Delta T_P(J^{(t-1)}, \Delta^{(t-1)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} &= \emptyset ext{ break} \end{aligned}$$

T(X,Y) := E(X,Y)

 $T(X,Y) := T(X,Z) \wedge E(Z,Y)$ 

$$\begin{array}{c} \text{for } t=1,\infty \\ \Delta^{(t)}(X,Y) := \end{array}$$

#### Naive

## varve

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for} \ \ t = 0, \infty \ & ext{} J^{(t+1)} &:= T_P(J^{(t)}) \ & ext{if} \ \ J^{(t+1)} &= J^{(t)} \end{aligned}$$

break

#### Semi-naive

$$J^{(0)} := \emptyset$$
 for  $t = 0, \infty$   $\Delta^{(t)} := T_P(J^{(t)}) - J^{(t)}$   $J^{(t+1)} := J^{(t)} \cup \Delta^{(t)}$  if  $\Delta^{(t)} = \emptyset$  break

## w/ incremental computation

$$egin{aligned} J^{(0)} &:= \emptyset, \quad \Delta^{(0)} := T_P(\emptyset) \ & ext{for } t = 1, \infty \ & \Delta^{(t)} &:= T_P(J^{(t-1)} \cup \Delta^{(t-1)}) - J^{(t)} \ &= \Delta T_P(J^{(t-1)}, \Delta^{(t-1)}) - J^{(t)} \ J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} = \emptyset ext{ break} \end{aligned}$$

$$T(X,Y) := E(X,Y)$$
  
 $T(X,Y) := T(X,Z) \wedge E(Z,Y)$ 

$$T^{(0)}(X,Y):= exttt{false},\ \Delta^{(0)}(X,Y):=E(X,Y)$$
 for  $t=1,\infty$ 

$$\Delta^{(t)}(X,Y) := \Delta^{(t-1)}(X,Z) \wedge E(Z,Y) \wedge \neg T^{(t)}(X,Y)$$

#### Naive

#### · · · · ·

$$egin{aligned} J^{(0)} &:= \emptyset \ & ext{for} \ \ t = 0, \infty \ & ext{} J^{(t+1)} &:= T_P(J^{(t)}) \ & ext{if} \ \ J^{(t+1)} &= J^{(t)} \end{aligned}$$

break

#### Semi-naive

$$egin{aligned} \overline{J^{(0)}} &:= \emptyset \ & ext{for } t = 0, \infty \ & \Delta^{(t)} &:= T_P(J^{(t)}) - J^{(t)} \ & J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} &= \emptyset ext{ break} \end{aligned}$$

## w/ incremental computation

$$egin{aligned} \overline{J^{(0)}} &:= \emptyset, \quad \Delta^{(0)} := T_P(\emptyset) \ & ext{for } t = 1, \infty \ & \Delta^{(t)} &:= T_P(J^{(t-1)} \cup \Delta^{(t-1)}) - J^{(t)} \ &= \Delta T_P(J^{(t-1)}, \Delta^{(t-1)}) - J^{(t)} \ J^{(t+1)} &:= J^{(t)} \cup \Delta^{(t)} \ & ext{if } \Delta^{(t)} = \emptyset ext{ break} \end{aligned}$$

$$T(X,Y) := E(X,Y)$$
  
 $T(X,Y) := T(X,Z) \wedge E(Z,Y)$ 

$$T^{(0)}(X,Y):= exttt{false},\ \Delta^{(0)}(X,Y):=E(X,Y)$$
 for  $t=1,\infty$ 

$$\Delta^{(t)}(X,Y) := \Delta^{(t-1)}(X,Z) \wedge E(Z,Y) \wedge \neg T^{(t)}(X,Y)$$

$$T^{(t+1)}(X,Y) := T^{(t)}(X,Y) \vee \Delta^{(t)}(X,Y)$$

if 
$$\Delta^{(t)} = \emptyset$$
 break

Semi-naive is implemented by virtually all datalog systems.

Non-linear datalog rules have more complex delta-queries:

Exponential number of queries:

$$(A \cup \Delta A) \bowtie (B \cup \Delta B) \bowtie (C \cup \Delta C)$$

Semi-naive is implemented by virtually all datalog systems.

Non-linear datalog rules have more complex delta-queries:

Exponential number of queries:

$$(A \cup \Delta A) \bowtie (B \cup \Delta B) \bowtie (C \cup \Delta C) = (A \bowtie B \bowtie C) \cup \\ \cup (\Delta A \bowtie B \bowtie C) \cup \cdots \cup (\Delta A \bowtie \Delta B \bowtie C) \cup \cdots \cup (\Delta A \bowtie \Delta B \bowtie \Delta C)$$

Semi-naive is implemented by virtually all datalog systems.

Non-linear datalog rules have more complex delta-queries:

Exponential number of queries:

$$(A \cup \Delta A) \bowtie (B \cup \Delta B) \bowtie (C \cup \Delta C) = (A \bowtie B \bowtie C) \cup \cup (\Delta A \bowtie B \bowtie C) \cup \cdots \cup (\Delta A \bowtie \Delta B \bowtie C) \cup \cdots \cup (\Delta A \bowtie \Delta B \bowtie \Delta C)$$

Mix of old/new tables (issue: new tables are bigger):

$$(A \cup \Delta A) \bowtie (B \cup \Delta B) \bowtie (C \cup \Delta C)$$

Semi-naive is implemented by virtually all datalog systems.

Non-linear datalog rules have more complex delta-queries:

Exponential number of queries:

$$(A \cup \Delta A) \bowtie (B \cup \Delta B) \bowtie (C \cup \Delta C) = (A \bowtie B \bowtie C) \cup \cup (\Delta A \bowtie B \bowtie C) \cup \cdots \cup (\Delta A \bowtie \Delta B \bowtie C) \cup \cdots \cup (\Delta A \bowtie \Delta B \bowtie \Delta C)$$

Mix of old/new tables (issue: new tables are bigger):

```
(A \cup \Delta A) \bowtie (B \cup \Delta B) \bowtie (C \cup \Delta C) = (A \bowtie B \bowtie C) \cup \\ \cup (\Delta A \bowtie B \bowtie C) \cup ((A \cup \Delta A) \bowtie \Delta B \bowtie C) \cup ((A \cup \Delta A) \bowtie (B \cup \Delta B) \bowtie \Delta C)
```

# Asynchronous Execution

• (Semi-) naive is synchronous: apply all rules to all tuples.

Asynchronous execution: apply some rules to some tuples.

• Simple principle: fair computation of a fixpoint.

# Asynchronous Sequence

Posets 
$$(P_1, \leq), (P_2, \leq)$$
, finite ranks,  $f: P_1 \times P_2 \rightarrow P_1$ ,  $g: P_1 \times P_2 \rightarrow P_2$ .

Goal compute 
$$lfp(f, g)$$
:

$$f(f(x,y),g(x,y)) = f(x,y)$$

# Asynchronous Sequence

Posets  $(P_1, \leq), (P_2, \leq)$ , finite ranks,  $f: P_1 \times P_2 \rightarrow P_1$ ,  $g: P_1 \times P_2 \rightarrow P_2$ .

Goal compute lfp(f, g):

$$(f(x,y),g(x,y))=(x,y)$$

#### Kleene's sequence:

$$(x^{(0)}, y^{(0)}) \stackrel{\text{def}}{=} (\bot, \bot) (x^{(t+1)}, y^{(t+1)}) \stackrel{\text{def}}{=} (f(x^{(t)}, y^{(t)}), g(x^{(t)}, y^{(t)}))$$

Every step is an fg-step.

# Asynchronous Sequence

Posets  $(P_1, \leq), (P_2, \leq)$ , finite ranks,  $f: P_1 \times P_2 \rightarrow P_1$ ,  $g: P_1 \times P_2 \rightarrow P_2$ .

Goal compute lfp(f, g):

$$f(x,y),g(x,y))=(x,y)$$

#### Kleene's sequence:

$$(x^{(0)}, y^{(0)}) \stackrel{\text{def}}{=} (\bot, \bot) (x^{(t+1)}, y^{(t+1)}) \stackrel{\text{def}}{=} (f(x^{(t)}, y^{(t)}), g(x^{(t)}, y^{(t)}))$$

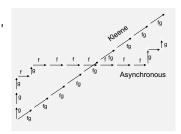
Every step is an fg-step.

#### Asynchronous sequence:

$$\begin{aligned} &(u^{(0)}, v^{(0)}) \overset{\text{def}}{=} (\bot, \bot) \\ &(u^{(k+1)}, v^{(k+1)}) \overset{\text{def}}{=} \\ & \qquad \qquad \left\{ \begin{aligned} &(f(u^{(k)}, v^{(k)}), g(u^{(k)}, v^{(k)})) & \text{or} \\ &(f(u^{(k)}, v^{(k)}), v^{(k)}) & \text{or} \\ &(u^{(k)}, g(u^{(k)}, v^{(k)})) \end{aligned} \right. \end{aligned}$$

fg-step, or f-step, or g-step.

Fact 1: for any pre-fixpoint (x, y) of (f, g),  $(u^{(k)}, v^{(k)}) \leq (x, y)$ .



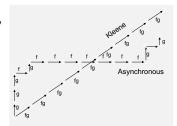
oboooooo

# Fair Computation of a Fixpoint

Fact 1: for any pre-fixpoint (x, y) of (f, g),  $(u^{(k)}, v^{(k)}) < (x, y).$ 

Sequence is fair if:  $\forall k \exists m > k \exists n > k$  s.t:

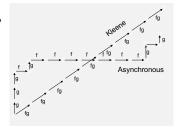
- m is an f-step or fg-step, and
- n is a g-step or fg-step.



Fact 1: for any pre-fixpoint (x, y) of (f, g),  $(u^{(k)}, v^{(k)}) \leq (x, y)$ .

Sequence is fair if:  $\forall k \exists m > k \exists n > k$  s.t:

- m is an f-step or fg-step, and
- n is a g-step or fg-step.

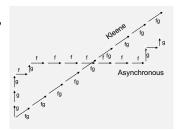


Fact 2: If the sequence is fair, then  $\exists k \text{ s.t. } (u^{(k)}, v^{(k)}) = \mathsf{lfp}(f, g)$ .

Fact 1: for any pre-fixpoint (x, y) of (f, g),  $(u^{(k)}, v^{(k)}) < (x, y).$ 

Sequence is fair if:  $\forall k \exists m > k \exists n > k$  s.t:

- m is an f-step or fg-step, and
- n is a g-step or fg-step.



Optimizing Monotone Datalog

0000000000

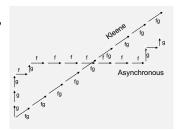
Fact 2: If the sequence is fair, then  $\exists k \text{ s.t. } (u^{(k)}, v^{(k)}) = \mathsf{lfp}(f, g)$ .

Proof suffices to prove:  $\forall t \exists k, (x^{(t)}, v^{(t)}) < (u^{(k)}, v^{(k)}).$ 

Fact 1: for any pre-fixpoint (x, y) of (f, g),  $(u^{(k)}, v^{(k)}) \leq (x, y)$ .

Sequence is fair if:  $\forall k \exists m > k \exists n > k$  s.t:

- m is an f-step or fg-step, and
- *n* is a *g*-step or *fg*-step.



Optimizing Monotone Datalog

0000000000

Fact 2: If the sequence is fair, then  $\exists k \text{ s.t. } (u^{(k)}, v^{(k)}) = \mathsf{lfp}(f, g)$ .

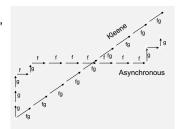
Proof suffices to prove:  $\forall t \exists k, (x^{(t)}, y^{(t)}) \leq (u^{(k)}, v^{(k)}).$ 

$$(x^{(t+1)},y^{(t+1)}) = (f(x^{(t)},y^{(t)}),g(x^{(t)},y^{(t)})) \leq (f(u^{(k)},v^{(k)}),g(u^{(k)},v^{(k)}))$$

Fact 1: for any pre-fixpoint (x, y) of (f, g),  $(u^{(k)}, v^{(k)}) < (x, y).$ 

Sequence is fair if:  $\forall k \exists m > k \exists n > k$  s.t:

- m is an f-step or fg-step, and
- n is a g-step or fg-step.



Fact 2: If the sequence is fair, then  $\exists k \text{ s.t. } (u^{(k)}, v^{(k)}) = \mathsf{lfp}(f, g).$ 

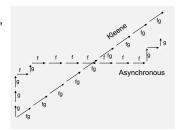
Proof suffices to prove:  $\forall t \exists k, (x^{(t)}, v^{(t)}) < (u^{(k)}, v^{(k)}).$ 

$$\begin{aligned} &(x^{(t+1)},y^{(t+1)}) = (f(x^{(t)},y^{(t)}),g(x^{(t)},y^{(t)})) \leq (f(u^{(k)},v^{(k)}),g(u^{(k)},v^{(k)})) \\ &\leq (f(u^{(m)},v^{(m)}),g(u^{(n)},v^{(n)})) \end{aligned}$$

Fact 1: for any pre-fixpoint (x, y) of (f, g),  $(u^{(k)}, v^{(k)}) \leq (x, y)$ .

Sequence is fair if:  $\forall k \exists m > k \exists n > k$  s.t:

- m is an f-step or fg-step, and
- *n* is a *g*-step or *fg*-step.



Fact 2: If the sequence is fair, then  $\exists k \text{ s.t. } (u^{(k)}, v^{(k)}) = \mathsf{lfp}(f, g)$ .

Proof suffices to prove:  $\forall t \exists k, (x^{(t)}, y^{(t)}) \leq (u^{(k)}, v^{(k)}).$ 

$$\begin{aligned} &(x^{(t+1)},y^{(t+1)}) = (f(x^{(t)},y^{(t)}),g(x^{(t)},y^{(t)})) \leq (f(u^{(k)},v^{(k)}),g(u^{(k)},v^{(k)})) \\ &\leq (f(u^{(m)},v^{(m)}),g(u^{(n)},v^{(n)})) = (u^{(m+1)},v^{(n+1)}) \leq (u^{(p)},v^{(p)}) \\ &\qquad \qquad \text{where } p = \max(m,n) + 1. \end{aligned}$$

- Kleene's sequence has rank  $(P_1) + \text{rank}(P_2)$ ; the asynchronous sequence could be as long as  $\text{rank}(P_1) \times \text{rank}(P_2)$
- Application: nested recursion

$$\begin{aligned} \mathsf{lfp}(f,g) = & \ \mathsf{let} \ u = \mathsf{lfp}(\lambda x. \ \ \mathsf{let} \ v = \mathsf{lfp}(\lambda y.g(x,y)) \\ & \ \ \mathsf{in} \ (f(x,v),v)) \\ & \ \ \mathsf{in} \ (u,\mathsf{lfp}(\lambda y.g(u,y))) \end{aligned}$$

RHS is asynchronous sequence with steps  $ggg\cdots fggg\cdots fggg\cdots$ 

• Immediate generalization to *n* posets  $(P_1, \leq) \times \cdots (P_n, \leq)$ .

# Grounding of a Datalog Program

What are the posets  $(P_1, \leq), (P_2, \leq), \ldots$  for a datalog program?

• Option 1:  $P_i$  is  $(ADom^{k_i}, \subseteq)$  represents an IDB predicate.

• Option 2 (better):  $P_i$  is  $(\{0,1\},\leq)$  represents an IDB tuple.

# Example

$$R(X) := E(a, X)$$
  
 $R(X) := R(Z) \land E(Z, X)$ 

$$R(X) := E(a, X)$$
  
 $R(X) := R(Z) \land E(Z, X)$ 

EDB input graph:

Optimizing Monotone Datalog



$$R(X) := E(a, X)$$
  
 $R(X) := R(Z) \wedge E(Z, X)$ 

## Grounded program:

$$R(a) := E(a, a)$$

$$R(a) := R(a) \wedge E(a, a)$$

$$R(a) := R(b) \wedge E(b, a)$$

$$R(b) := E(a,b)$$

$$R(b) := R(a) \wedge E(a,b)$$

$$R(b) := R(b) \wedge E(b, b)$$

## EDB input graph:

Optimizing Monotone Datalog



# Example

$$R(X) := E(a, X)$$
  
 $R(X) := R(Z) \wedge E(Z, X)$ 

### Grounded program:

$$R(a) := E(a, a)$$

$$R(a) := R(a) \wedge E(a, a)$$

$$R(a) := R(b) \wedge E(b, a)$$

$$R(b) := E(a,b)$$

$$R(b) := R(a) \wedge E(a,b)$$

$$R(b) := R(b) \wedge E(b,b)$$

$$R(a) := E(a, a) \vee R(a) \wedge E(a, a) \vee R(b) \wedge E(b, a),$$

$$R(b) := E(a,b) \vee R(a) \wedge E(a,b) \vee R(b) \wedge E(b,b)$$

# EDB input graph:

Optimizing Monotone Datalog



# Example

$$R(X) := E(a, X)$$
  
 $R(X) := R(Z) \wedge E(Z, X)$ 

#### Grounded program:

$$R(a) := E(a, a)$$

$$R(a) := R(a) \wedge E(a, a)$$

$$R(a) := R(b) \wedge E(b, a)$$

$$R(b) := E(a,b)$$

$$R(b) := R(a) \wedge E(a,b)$$

$$R(b) := R(b) \wedge E(b,b)$$

$$R(a) := E(a,a) \vee R(a) \wedge E(a,a) \vee R(b) \wedge E(b,a),$$

$$R(b) := E(a,b) \vee R(a) \wedge E(a,b) \vee R(b) \wedge E(b,b)$$

The grounded program allows more fine-grained asynchronous execution.

EDB input graph:



# Summary

- Main purpose of datalog is to add recursion.
- Least-fixpoint semantics; Kleene's sequence; Naive algorithm.
- Cool optimizations: semi-naive, magic-sets (difficult!), asynchronous evaluation.
- Can express PTIME-complete problems (AGAP).
- But limited to monotone queries.

Next lecture: adding negation to datalog.



Stanley, R. P. (1999).

Enumerative combinatorics. Vol. 2, volume 62 of Cambridge Studies in Advanced Mathematics. Cambridge University Press, Cambridge.

With a foreword by Gian-Carlo Rota and appendix 1 by Sergey Fomin.