# CS294-248 Special Topics in Database Theory Unit 9: Datalog (Part 2)

Dan Suciu

University of Washington

#### **Announcements**

• Next Tuesday, Nov. 28: office hours 2pm-4:30pm.

• Please submit a short report on your project by Wednesday, Nov. 29.

 Project presentations: Thursday, Nov. 30, 9:30am, Calvin 116. More details TBD.

# Recursion and Negation

## Recap: Datalog

- Datalog = set of rules.
- Immediate consequence operator
- Least fixpoint semantics
- Naive algorithm  $J^{(0)} \subset J^{(1)} \subset \cdots$

#### Example:

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

#### Non-example:

$$C(X) := A(X) \wedge \neg B(X)$$

What happens if we allow negation?

#### Three Examples

Transitive closure of the complement graph:

$$EC(X,Y) := V(X) \wedge V(Y) \wedge \neg E(X,Y)$$
  
 $T(X,Y) := EC(X,Y)$   
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Transitive closure of the complement graph:

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Complement of the transitive closure:

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

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The Win-Move Game:

$$W(X) := E(X, Y) \land \neg W(Y)$$

(will explain it later)

EDB is 
$$S = \{1\}$$

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Fixpoint 1: 
$$A = \{1\}, B = \emptyset$$
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Pre-fixpoint 3: 
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Pre-fixpoint 3: 
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A simpler example:

Recursion and Negation

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$$A := \neg B$$
  
 $B := \neg \Delta$ 

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EDB is 
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A simpler example:

Recursion and Negation

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ICO not monotone! Need new semantics

Semi-positive, stratified datalog

• Semantics motivated by logic.

• Semantics motivated by computation.

Mostly based on [Abiteboul et al., 1995].

## Semi-positive Datalog

EDB atoms may be positive or negated.

IDB atoms can only be positive.

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The Immediate Consequence Operator is monotone.

Semantics: least fixpoint of the ICO.

• Stratification: assign to each IDB predicate a stratum  $s(R) \in \mathbb{N}$ .

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  - ▶ For positive atoms  $A(X) :- \cdots \land B(Y) \land \cdots$ :  $s(A) \geq s(B)$ .
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- Semantics: for each stratum s = 1, 2, ..., view it as a semi-positive datalog program, compute its fixpoint.
- The output is called perfect model; it is not a minimal model!

#### Example

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Stratum 1: T

Stratum 2: Answ

<sup>&</sup>lt;sup>1</sup>Assuming no isolated nodes

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#### Semantics:

T = transitive closure, Answ = its complement

This is not the least fixpoint (minimal model) why??

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The following is also a fixpoint:<sup>1</sup>

$$T = V \times V$$
, Answ =  $\emptyset$ 

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#### Discussion

 Stratified datalog is by far the most popular extension of datalog with negation.

 It is limited: it completely prevents the interleaving of recursion and negation. The following is not allowed:

$$A := \neg B$$

# Logic-Based Extensions

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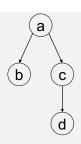
Stable Models

Well Founded Model

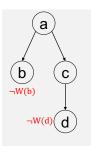
Representative example: the Win-Move Game (next)

- Players I, II take turns moving a pebble in a graph.
- Player who cannot move loses.
- For each node X, does Player I have a winning strategy?

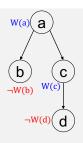
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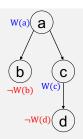
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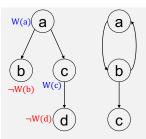


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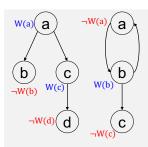
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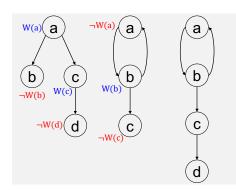
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Logic-Based Extensions 00000000

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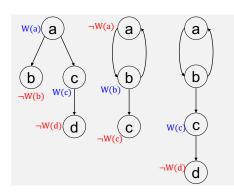
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Logic-Based Extensions 000000000

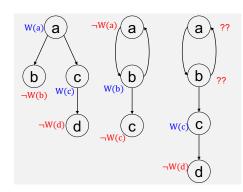
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# The Win-Move Game

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$$W(X) := E(X,Y) \wedge \neg W(Y)$$

• Least Fixpoint Logic (LFP) is FO extended with monotone fixpoint. E.g. the win-move game:

$$\boxed{\mathsf{lfp}_{W(x)}(\exists y(E(x,y) \land \forall z(E(y,z) \Rightarrow W(z))))}$$

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Before that we discuss two simple technical constructs:

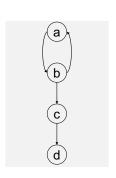
grounded program and reduct

# The Grounded Datalog Program

A grounded atom, or a fact, is an atom without variables

A grounded rule is a rule whose atoms are grounded.

The grounding of a program *P* consists of all possible groundings of its rules



$$W(X) := E(X,Y) \land \neg W(Y)$$

$$W(a) := E(a,b) \land \neg W(b)$$

$$W(b) := E(b,a) \land \neg W(a)$$

$$W(b) := E(b,c) \land \neg W(c)$$

$$W(c) := E(c,d) \wedge \neg W(d)$$

 $P \stackrel{\text{def}}{=}$  the grounded program, J =any set of grounded atoms;

The reduct,  $P_I$  is obtained as follows:

- Remove all rules with a negated atom in J.
- Remove all remaining negated atoms.

 $P_J$  is monotone; Ifp $(P_J)$  exists;  $J_1 \subseteq J_2$  implies Ifp $(P_{J_1}) \supseteq$ Ifp $(P_{J_2})$ 

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$$lfp(P_I) = \{W(a), W(b)\}\$$

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$$W(a) := E(a,b) \land \neg W(b)$$
  
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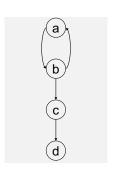
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$$\mathsf{lfp}(P_J) = \{W(a), W(b)\}$$

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J is a stable model if  $J = Ifp(P_J)$ 



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J is a stable model if  $J = Ifp(P_I)$ 

Example:  $J = \{W(a), W(c)\}$ 

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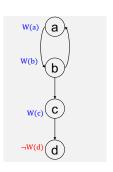
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J is a stable model if  $J = lfp(P_J)$ 

Example:  $J = \{W(a), W(c)\}$ 

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Non-example:  $J = \{W(a), W(b), W(c)\}$  why??



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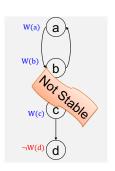
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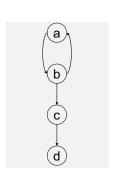
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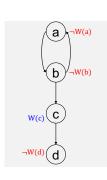
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- Stable models introduced by [Gelfond and Lifschitz, 1988]
- Elegant, principled definition.
- But: NP-hard to check if there exists any stable model.

A stratified program has a unique stable model, which is the perfect model.

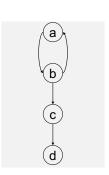
```
A(1) := Perfect model: J = \{A(1), C(1)\}
B(1) := \neg A(1)
C(1) := A(1)
Not stable: J = \{A(1), B(1), C(1)\} why?
```

 $C(1) := C(1) \land \neg B(1)$ 

$$J^{(0)} \stackrel{\text{def}}{=} \emptyset, \ J^{(t+1)} \stackrel{\text{def}}{=} \mathsf{lfp}(P_{J^{(t)}})$$

$$J^{(0)} \subseteq J^{(2)} \subseteq J^{(4)} \subseteq \cdots \subseteq J^{(5)} \subseteq J^{(3)} \subseteq J^{(1)}.$$

$$\bigcup_{t} J^{(2t)} \stackrel{\text{def}}{=} \text{certain-} \bigcap_{t} J^{(2t+1)} \stackrel{\text{def}}{=} \text{possible-facts}$$



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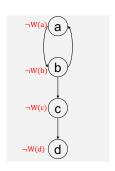
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$$\bigcup_{t} J^{(2t)} \stackrel{\text{def}}{=} \text{certain-} \bigcap_{t} J^{(2t+1)} \stackrel{\text{def}}{=} \text{possible-facts}$$



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

$$W(a) := E(a,b) \land \neg W(b)$$

$$W(b) := E(b, a) \land \neg W(a)$$

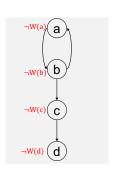
$$W(b) := E(b,c) \land \neg W(c)$$

$$W(c) := E(c,d) \land \neg W(d)$$

$$J^{(0)} \stackrel{\text{def}}{=} \emptyset, \ J^{(t+1)} \stackrel{\text{def}}{=} \mathsf{lfp}(P_{J^{(t)}})$$

$$J^{(0)} \subseteq J^{(2)} \subseteq J^{(4)} \subseteq \cdots \subseteq J^{(5)} \subseteq J^{(3)} \subseteq J^{(1)}.$$

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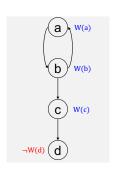
$$W(b) := E(b,c) \land \neg W(c)$$

$$|W(c)| - E(c,d) \wedge \neg W(d)$$
Topics in DB Theory: Unit 9b Fall 2023

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$$J^{(0)} = \emptyset$$
  $J^{(1)} = \{W(a), W(b), W(c)\}$ 

$$W(a) := E(a,b) \land \neg W(b)$$

$$W(b) := E(b,a) \land \neg W(a)$$

$$W(b) := E(b,c) \land \neg W(c)$$

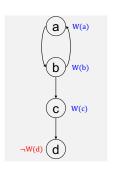
$$W(c) := E(c,d) \land \neg W(d)$$
Topics in DB Theory: Unit 9b Fall 2023

### Alternating Fixpoint:

$$\int_{0}^{1} \int_{0}^{1} \int_{0$$

$$J^{(0)} \subseteq J^{(2)} \subseteq J^{(4)} \subseteq \cdots \subseteq J^{(5)} \subseteq J^{(3)} \subseteq J^{(1)}.$$

$$\bigcup_{t} J^{(2t)} \stackrel{\text{def}}{=} \text{certain-} \bigcap_{t} J^{(2t+1)} \stackrel{\text{def}}{=} \text{possible-facts}$$



$$J^{(0)} = \emptyset \qquad \qquad J^{(1)} = \{W(a),$$

$$J^{(1)} = \{W(a), W(b), W(c)\}$$

$$W(a) := E(a,b) \land \neg W(b)$$

$$W(b) := E(b,a) \land \neg W(a)$$

$$W(b) := E(b,c) \land \neg W(c)$$

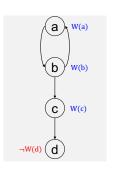
$$W(c) := E(c,d) \land \neg W(d)$$

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$$J^{(0)} \stackrel{\mathsf{def}}{=} \emptyset, \ J^{(t+1)} \stackrel{\mathsf{def}}{=} \mathsf{lfp}(P_{J^{(t)}})$$

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

$$J^{(1)} = \{W(a), W(b), W(c)\}$$

$$W(a) := E(a,b) \land \neg W(b)$$

$$W(b) := E(b,a) \land \neg W(a)$$

$$W(b) := E(b,c) \land \neg W(c)$$

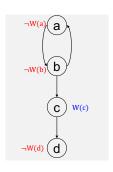
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$$J^{(0)} = \emptyset$$
  $J^{(1)} = \{W(a), W(b), W(c)\}$   
 $J^{(2)} = \{W(c)\}$ 

$$W(a) := E(a,b) \land \neg W(b)$$

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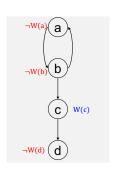
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$$\int_{0}^{(0)} \stackrel{\text{def}}{=} \emptyset, \ \int_{0}^{(t+1)} \stackrel{\text{def}}{=} \mathsf{lfp}(P_{J^{(t)}})$$

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$$\bigcup_{t} J^{(2t)} \stackrel{\text{def}}{=} \text{certain-} \bigcap_{t} J^{(2t+1)} \stackrel{\text{def}}{=} \text{possible-facts}$$



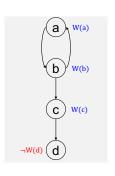
$$J^{(0)} = \emptyset$$
  $J^{(1)} = \{W(a), W(b), W(c)\}$   
 $J^{(2)} = \{W(c)\}$ 

$$\begin{array}{c|c} W(a) := E(a,b) & \land \neg W(b) \\ \hline W(b) := E(b,a) & \land \neg W(a) \\ \hline W(b) := E(b,c) & \land \neg W(c) \\ \hline W(c) := E(c,d) & \land \neg W(d) \\ \hline \end{array}$$

$$\int_{0}^{1} \int_{0}^{1} \int_{0$$

$$J^{(0)} \subseteq J^{(2)} \subseteq J^{(4)} \subseteq \cdots \subseteq J^{(5)} \subseteq J^{(3)} \subseteq J^{(1)}$$
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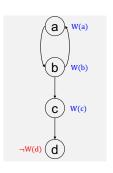
$$\begin{array}{ll} \emph{J}^{(0)} = \emptyset & \emph{J}^{(1)} = \{ \emph{W}(\emph{a}), \emph{W}(\emph{b}), \emph{W}(\emph{c}) \} \\ \emph{J}^{(2)} = \{ \emph{W}(\emph{c}) \} & \emph{J}^{(3)} = \{ \emph{W}(\emph{a}), \emph{W}(\emph{b}), \emph{W}(\emph{c}) \} \\ \end{array}$$

$$\begin{array}{c|c} \hline W(a) := E(a,b) \\ \hline W(b) := E(b,a) \\ \hline W(b) := E(b,c) \land \neg W(c) \\ \hline W(c) := E(c,d) \\ \hline \end{pmatrix} \land \neg W(d)$$

$$\int_{0}^{(0)} \stackrel{\text{def}}{=} \emptyset, \ \int_{0}^{(t+1)} \stackrel{\text{def}}{=} \mathsf{lfp}(P_{J^{(t)}})$$

$$J^{(0)} \subseteq J^{(2)} \subseteq J^{(4)} \subseteq \cdots \subseteq J^{(5)} \subseteq J^{(3)} \subseteq J^{(1)}.$$

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$$J^{(0)} = \emptyset$$
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$$W(a) := E(a,b) \land \neg W(b)$$

$$W(b) := E(b,a) \land \neg W(a)$$
  
 $W(b) := E(b,c) \land \neg W(c)$ 

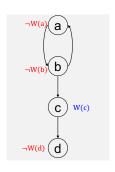
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$$J^{(0)} = \emptyset \qquad J^{(1)} = \{W(a), W(b), W(c)\}$$
  

$$J^{(2)} = \{W(c)\} \qquad J^{(3)} = \{W(a), W(b), W(c)\}$$
  

$$J^{(4)} = \{W(c)\} \qquad \dots$$

$$W(a) := E(a,b) \land \neg W(b)$$
  
 $W(b) := E(b,a) \land \neg W(a)$ 

$$W(b) := E(b,c) \land \neg W(c)$$

$$VV(D) := E(D,C) \land \neg VV(C)$$

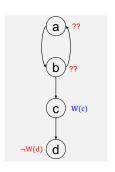
$$W(c) := E(c,d) \land \neg W(d)$$

### Alternating Fixpoint:

$$J^{(0)} \stackrel{\mathsf{def}}{=} \emptyset, \ J^{(t+1)} \stackrel{\mathsf{def}}{=} \mathsf{lfp}(P_{J^{(t)}})$$

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$$J^{(0)} = \emptyset$$
  $J^{(1)} = \{W(a), W(b), W(c)\}$   
 $J^{(2)} = \{W(c)\}$   $J^{(3)} = \{W(a), W(b), W(c)\}$   
 $J^{(4)} = \{W(c)\}$  ...  
Certain facts:  $W(c)$ :

$$W(b) := E(b,a) \land \neg W(a)$$
  
 $W(b) := E(b,c) \land \neg W(c)$   
 $W(c) := E(c,d) \land \neg W(d)$ 

 $W(a) := E(a,b) \wedge \neg W(b)$ 

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possible facts: W(a), W(b).

• Well-founded models can be computed in PTIME.

Yet, I don't know of any system that supports it.
 Maybe because of the 3-valued logic?

Next: two other semantics motivated by computation.

# Computation-Based Extensions

• Datalog with inflationary fixpoint semantics.

Datalog with partial fixpoint semantics.

# Inflationary Fixpoint

Let P be a datalog program,  $T_P$  its ICO.

The inflationary fixpoint is  $| ifp(P) \stackrel{\text{def}}{=} \bigcup_{t \geq 0} J_t |$ , where:

$$J_0 \stackrel{\mathsf{def}}{=} \emptyset$$
,  $J_{t+1} \stackrel{\mathsf{def}}{=} J_t \cup T_P(J_t)$ 

#### **Fact**

ifp(P) can be computed in PTIME in the size of the EDB 1.

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#### **Fact**

ifp(P) can be computed in PTIME in the size of the EDB 1.

why? Because  $J_0 \subseteq J_1 \subseteq \cdots \subseteq (ADom(I))^k$ 

### The partial fixpoint is:

$$\mathsf{pfp}(P) \stackrel{\mathsf{def}}{=} egin{cases} J_{t_0} & \mathsf{if} \ J_{t_0} = J_{t_0+1} \ \emptyset & \mathsf{if} \ J_t 
eq J_{t+1}, orall t \end{cases}$$

where

$$J_0 \stackrel{\mathsf{def}}{=} \emptyset$$
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#### **Fact**

pfp(P) can be computed in PSPACE in the size of the EDB 1.

why?

# Partial Fixpoint

The partial fixpoint is:

$$\mathsf{pfp}(P) \stackrel{\mathsf{def}}{=} egin{cases} J_{t_0} & \mathsf{if} \ J_{t_0} = J_{t_0+1} \ \emptyset & \mathsf{if} \ J_t 
eq J_{t+1}, orall t \end{cases}$$

where

$$J_0 \stackrel{\mathsf{def}}{=} \emptyset$$
,  $J_{t+1} \stackrel{\mathsf{def}}{=} T_P(J_t)$ 

#### **Fact**

pfp(P) can be computed in PSPACE in the size of the EDB 1.

why? each  $|J_t|$  has size polynomial in ADom(1).

Detect non-termination using a counter.

It's harder than one may think!

### Complement of the TC:

$$T(X,Y) := E(X,Y)$$
 $T(X,Y) := T(X,Z) \land E(Z,X)$ 
Answ $(X,Y) := V(X) \land V(Y)$ 
 $\land \neg T(X,Y)$ 

ifp(P) is incorrect!

# How To Express Negation

It's harder than one may think!

#### Complement of the TC:

$$T(X,Y) := E(X,Y)$$
 $T(X,Y) := T(X,Z) \land E(Z,X)$ 
Answ $(X,Y) := V(X) \land V(Y)$ 
 $\land \neg T(X,Y)$ 

ifp(P) is incorrect!

Detect the last step [Abiteboul et al., 1995, Ex.14.4.2]

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

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

$$T_{\mathsf{prev}}(X,Y) := T(X,Y)$$

$$T_{\mathsf{prev-not-last}}(X,Y) := T(X,Y) \land \land T(X',Y') \land \land T(X',Y')$$

$$\mathsf{Answ}(X,Y) := V(X) \land V(Y) \land \neg T(X,Y)$$

$$\land T_{\mathsf{prev}}(X',Y') \land \neg T_{\mathsf{prev-not-last}}(X',Y')$$

# Descriptive Complexity

• Datalog¬ cannot express parity, no matter which semantics we adopt.

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<sup>&</sup>lt;sup>2</sup>Exercise: express succ(X, Y), min(X), max(Y) using <.

# Descriptive Complexity

- Datalog¬ cannot express parity, no matter which semantics we adopt.
- If we have access to an order relation < then we can express parity as:<sup>2</sup>

<sup>&</sup>lt;sup>2</sup>Exercise: express succ(X, Y), min(X), max(Y) using <.

# Descriptive Complexity

- Datalog cannot express parity, no matter which semantics we adopt.
- If we have access to an order relation < then we can express parity as:<sup>2</sup>

$$E(X,Y) := \operatorname{succ}(X,Z) \wedge \operatorname{succ}(Z,Y)$$
  
 $E(X,Y) := E(X,Z) \wedge E(Z,Y) \quad // \text{ even-length distance}$   
 $\operatorname{Even}() := R(X) \wedge \min(X) \wedge E(X,Y) \wedge \max(Y) \wedge R(Y)$ 

# Theorem (Descriptive Complexity [Vardi, 1982, Immerman, 1986])

- Datalog¬(<, ifp) expresses precisely queries in PTIME.</li>
- Datalog<sup>¬</sup>(<, pfp) expresses precisely queries in PSPACE.

<sup>&</sup>lt;sup>2</sup>Exercise: express succ(X, Y), min(X), max(Y) using <.

- Datalog: simple, elegant, appealing. New resurgence after a 40 years history.
- Stratified datalog is a simple and practical extension.
- Beyond that, it becomes questionable.
- But the theory is beautiful. A famous result:

```
Theorem ([Abiteboul et al., 1992])
datalog^{\neg}(ifp) = datalog^{\neg}(pfp) \ iff \ PTIME=PSPACE.
```



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