

# CSE 544

# Principles of Database Management Systems

Fall 2016

Lecture 5 – Datalog (1)

# Announcements

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## Project

- Proposal due tomorrow
- Try to include **concrete deliverables** to keep track of progress

Homework 2 posted, due Friday, Nov. 4<sup>th</sup>

- SimpleDB

# References

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- R&G Chapter 24
- Phokion Kolaitis' tutorial on database theory at Simon's <https://simons.berkeley.edu/sites/default/files/docs/5241/simons16-21.pdf>
- Reading for Thursday:  
Joe Hellerstein, "The Declarative Imperative,"  
SIGMOD Record 2010

# Datalog

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- Alternative notation for queries
- Designed for recursive queries in the 80s
- Modern implementations: commercial (LogicBlox), networking (Overlog), programming languages, ...
- Topics
  - Syntax of datalog
  - How to evaluate
  - Datalog semantics

# Running Datalog

---

How to try out datalog quickly:

- Download DLV from: <http://www.dlvsystem.com/dlvdb/>
- Run DLV on this file:

```
parent(william, john).
parent(john, james).
parent(james, bill).
parent(sue, bill).
parent(james, carol).
parent(sue, carol).

male(john).
male(james).
female(sue).
male(bill).
female(carol).

grandparent(X, Y) :- parent(X, Z), parent(Z, Y).
father(X, Y) :- parent(X, Y), male(X).
mother(X, Y) :- parent(X, Y), female(X).
brother(X, Y) :- parent(P, X), parent(P, Y), male(X), X != Y.
sister(X, Y) :- parent(P, X), parent(P, Y), female(X), X != Y.
```

Actor(id, fname, lname)

Casts(pid, mid)

Movie(id, name, year)

# Facts and Rules

---

**Facts** = tuples in the database

Actor(344759, 'Douglas', 'Fowley').  
Casts(344759, 29851).  
Casts(355713, 29000).  
Movie(7909, 'A Night in Armour', 1910).  
Movie(29000, 'Arizona', 1940).  
Movie(29445, 'Ave Maria', 1940).

**Rules** = queries

Q1(y) :- Movie(x,y,z), z='1940'.

Find Movies made in 1940

Actor(id, fname, lname)

Casts(pid, mid)

Movie(id, name, year)

# Facts and Rules

---

**Facts** = tuples in the database

Actor(344759, 'Douglas', 'Fowley').  
Casts(344759, 29851).  
Casts(355713, 29000).  
Movie(7909, 'A Night in Armour', 1910).  
Movie(29000, 'Arizona', 1940).  
Movie(29445, 'Ave Maria', 1940).

**Rules** = queries

Q1(y) :- Movie(x,y,z), z='1940'.

Q2(f, l) :- Actor(z,f,l), Casts(z,x),  
Movie(x,y,'1940').

Find Actors who acted in Movies made in 1940

Actor(id, fname, lname)

Casts(pid, mid)

Movie(id, name, year)

# Facts and Rules

---

**Facts** = tuples in the database

Actor(344759, 'Douglas', 'Fowley').  
Casts(344759, 29851).  
Casts(355713, 29000).  
Movie(7909, 'A Night in Armour', 1910).  
Movie(29000, 'Arizona', 1940).  
Movie(29445, 'Ave Maria', 1940).

**Rules** = queries

Q1(y) :- Movie(x,y,z), z='1940'.

Q2(f, l) :- Actor(z,f,l), Casts(z,x),  
Movie(x,y,'1940').

Q3(f,l) :- Actor(z,f,l), Casts(z,x1), Movie(x1,y1,1910),  
Casts(z,x2), Movie(x2,y2,1940)

Find Actors who acted in a Movie in 1940 and in one in 1910

Actor(id, fname, lname)

Casts(pid, mid)

Movie(id, name, year)

# Facts and Rules

---

**Facts** = tuples in the database

Actor(344759, 'Douglas', 'Fowley').  
Casts(344759, 29851).  
Casts(355713, 29000).  
Movie(7909, 'A Night in Armour', 1910).  
Movie(29000, 'Arizona', 1940).  
Movie(29445, 'Ave Maria', 1940).

**Rules** = queries

Q1(y) :- Movie(x,y,z), z='1940'.

Q2(f, l) :- Actor(z,f,l), Casts(z,x),  
Movie(x,y,'1940').

Q3(f,l) :- Actor(z,f,l), Casts(z,x1), Movie(x1,y1,1910),  
Casts(z,x2), Movie(x2,y2,1940)

Extensional Database Predicates = EDB = Actor, Casts, Movie

Intensional Database Predicates = IDB = Q1, Q2, Q3

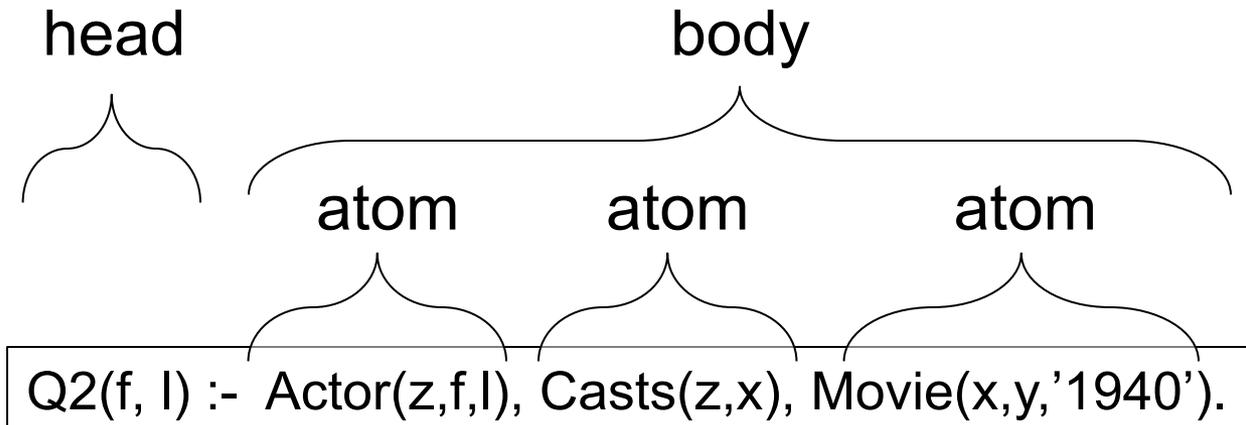
Actor(id, fname, lname)

Casts(pid, mid)

Movie(id, name, year)

# Terminology

---



f, l = head variables

x, y, z = existential variables

Actor(id, fname, lname)

Casts(pid, mid)

Movie(id, name, year)

# Safe Datalog Rules

---

Here are unsafe datalog rules. What's "unsafe" about them ?

$U1(x,y) :- \text{Movie}(x,z,1994), y > 1910$

$U2(x) :- \text{Movie}(x,z,1994), \text{not Casts}(u,x)$

A datalog rule is safe if every variable appears in some positive relational atom

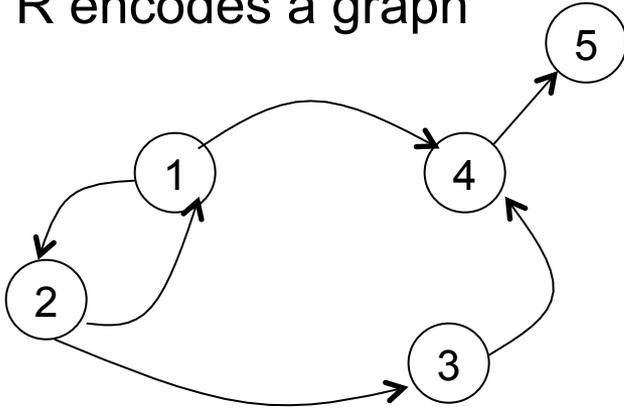
# Datalog v.s. SQL

---

- Non-recursive datalog with negation is a cleaned-up, core of SQL
- You should be able to translate easily between non-recursive datalog with negation and SQL

# Simple datalog programs

R encodes a graph



R=

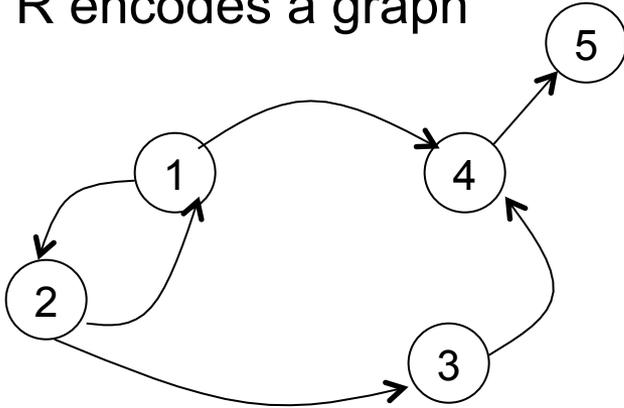
1	2
2	1
2	3
1	4
3	4
4	5

```
T(x,y) :- R(x,y)
T(x,y) :- R(x,z), T(z,y)
```

What does it compute?

# Simple datalog programs

R encodes a graph



R=

1	2
2	1
2	3
1	4
3	4
4	5

Initially:  
T is empty.

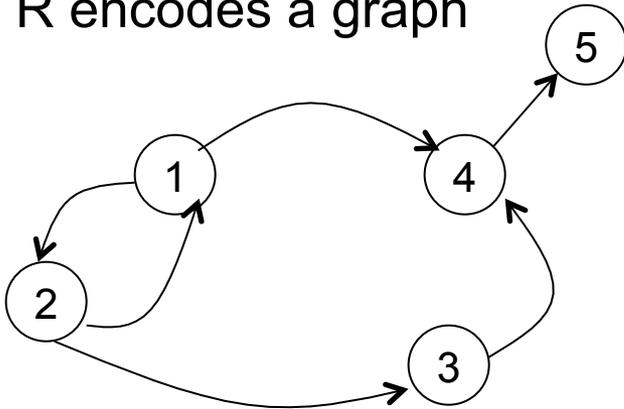


$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

What does  
it compute?

# Simple datalog programs

R encodes a graph



R=

1	2
2	1
2	3
1	4
3	4
4	5

Initially:  
T is empty.



First iteration:  
T =

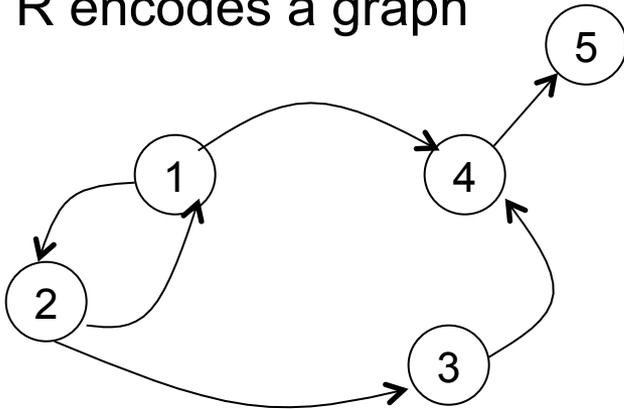
1	2
2	1
2	3
1	4
3	4
4	5

$T(x,y) \text{ :- } R(x,y)$   
 $T(x,y) \text{ :- } R(x,z), T(z,y)$

What does  
it compute?

# Simple datalog programs

R encodes a graph



R =

1	2
2	1
2	3
1	4
3	4
4	5

Initially:  
T is empty.



First iteration:  
T =

1	2
2	1
2	3
1	4
3	4
4	5

Second iteration:

T =

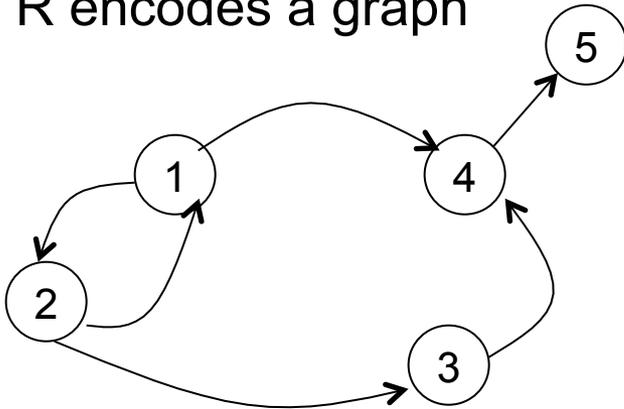
1	2
2	1
2	3
1	4
3	4
4	5
1	1
2	2
1	3
2	4
1	5
3	5

$T(x,y) \text{ :- } R(x,y)$   
 $T(x,y) \text{ :- } R(x,z), T(z,y)$

What does it compute?

# Simple datalog programs

R encodes a graph



R =

1	2
2	1
2	3
1	4
3	4
4	5

Initially:  
T is empty.



First iteration:  
T =

1	2
2	1
2	3
1	4
3	4
4	5

Second iteration:  
T =

1	2
2	1
2	3
1	4
3	4
4	5
1	1
2	2
1	3
2	4
1	5
3	5

Third iteration:  
T =

1	2
2	1
2	3
1	4
3	4
4	5
1	1
2	2
1	3
2	4
1	5
3	5
2	5

```

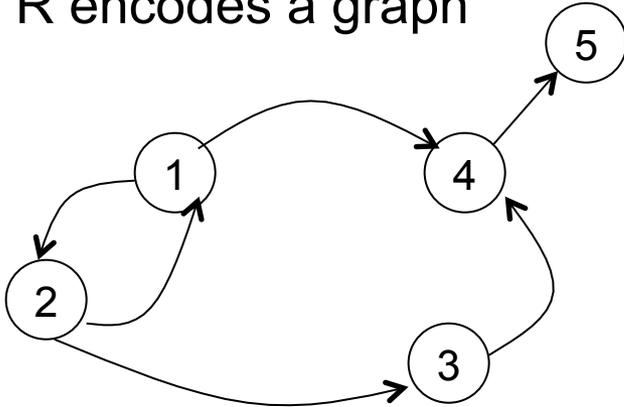
T(x,y) :- R(x,y)
T(x,y) :- R(x,z), T(z,y)
    
```

What does it compute?

Done

# Simple datalog programs

R encodes a graph



R =

1	2
2	1
2	3
1	4
3	4
4	5

Initially:  
T is empty.



First iteration:  
T =

1	2
2	1
2	3
1	4
3	4
4	5

Second iteration:  
T =

1	2
2	1
2	3
1	4
3	4
4	5
1	1
2	2
1	3
2	4
1	5
3	5

Third iteration:  
T =

1	2
2	1
2	3
1	4
3	4
4	5
1	1
2	2
1	3
2	4
1	5
3	5
2	5

Discovered  
3 times!

Discovered  
twice

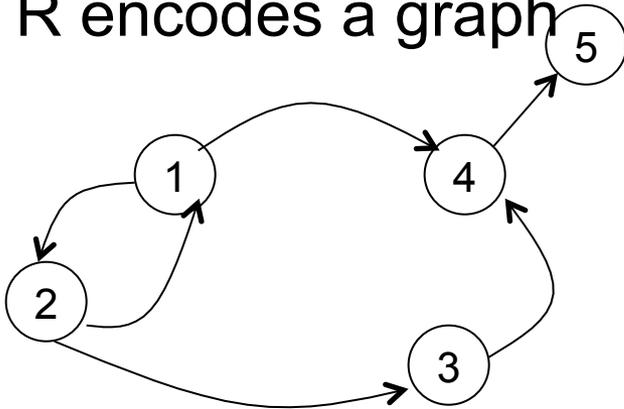
Done

What does  
it compute?

$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

# Simple datalog programs

R encodes a graph



R=

1	2
2	1
2	3
1	4
3	4
4	5

Alternative ways to compute TC:

$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

Right linear

$T(x,y) :- R(x,y)$   
 $T(x,y) :- T(x,z), R(z,y)$

Left linear

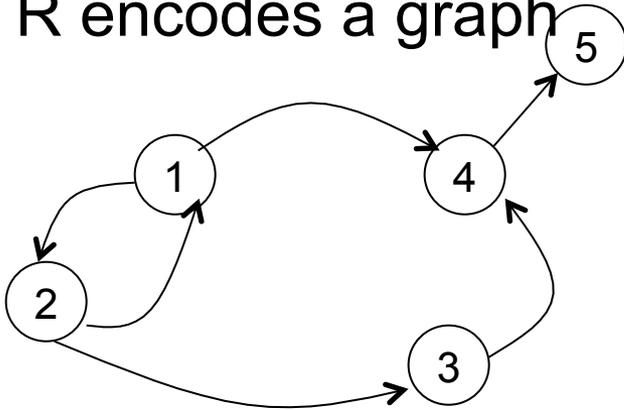
$T(x,y) :- R(x,y)$   
 $T(x,y) :- T(x,z), T(z,y)$

Non-linear

What are the pros / cons?

# Other Interesting Programs

R encodes a graph



Non 2-colorability:

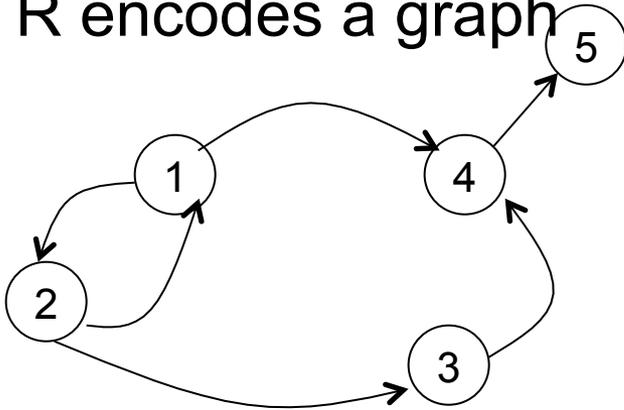
```
ODD(x,y) :- R(x,y)
ODD(x,y) :- R(x,z), EVEN(z,y)
EVEN(x,y) :- R(x,z), ODD(z,y)
Q :- ODD(x,x)
```

Path Systems (PTIME-complete)

```
T(x) :- A(x)
T(x) :- R(x,y,z), T(y), T(z)
```

# Other Interesting Programs

R encodes a graph



Cousins:

```
Cousins(y,z) :- Parent(x,y),Parent(x,z)
Cousins(u,v) :- Cousins(x,y),Parent(x,u),Parent(y,v)
```

# Syntax of Datalog Programs

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The schema consists of two sets of relations:

- Extensional Database (EDB):  $R_1, R_2, \dots$
- Intentional Database (IDB):  $P_1, P_2, \dots$

A datalog program  $\mathbf{P}$  has the form:

$\mathbf{P}$ : 
$$\begin{array}{l} P_{i1}(x_{11}, x_{12}, \dots) \text{ :- body}_1 \\ P_{i2}(x_{21}, x_{22}, \dots) \text{ :- body}_2 \\ \dots \end{array}$$

- Each head predicate  $P_i$  is an IDB
- Each body is a conjunction of IDB and/or EDB predicates

Note: no negation (yet)! Recursion OK.

# Naïve Datalog Evaluation Algorithm

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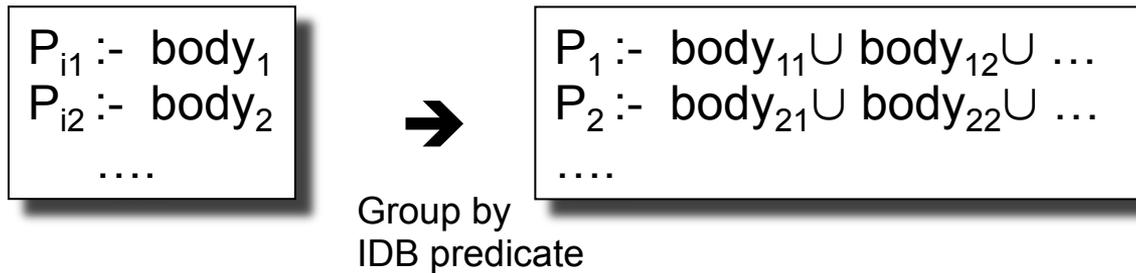
Datalog program:

```
Pi1 :- body1  
Pi2 :- body2  
.....
```

# Naïve Datalog Evaluation Algorithm

---

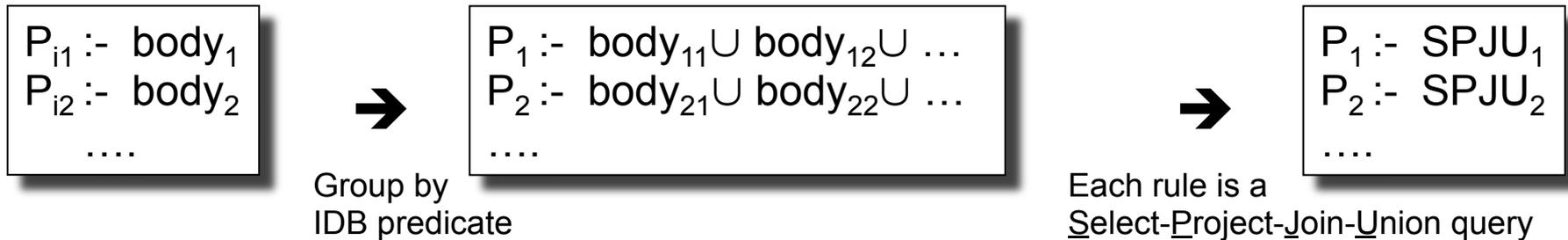
Datalog program:



# Naïve Datalog Evaluation Algorithm

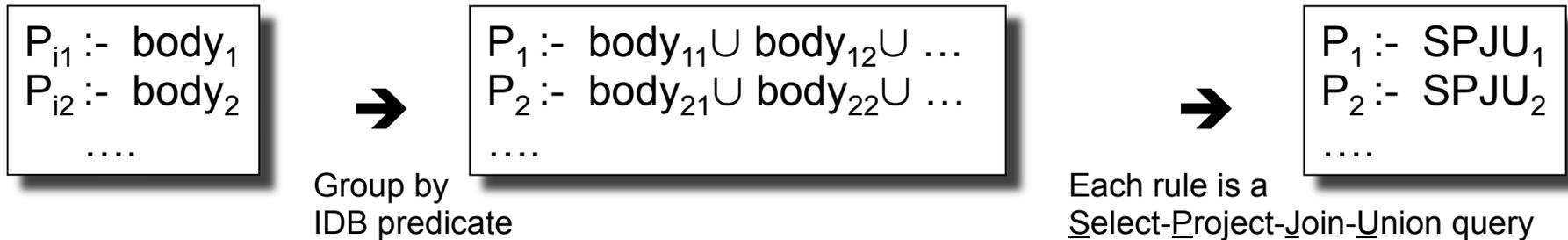
---

Datalog program:



# Naïve Datalog Evaluation Algorithm

Datalog program:

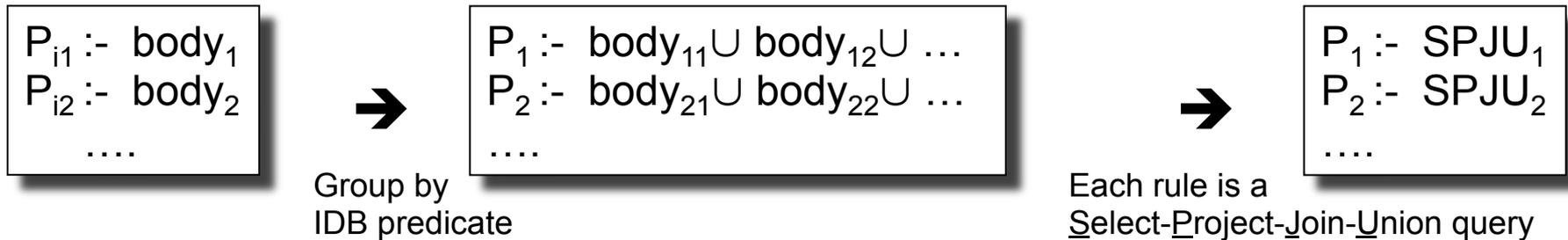


Example:  $T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

→ ?

# Naïve Datalog Evaluation Algorithm

Datalog program:

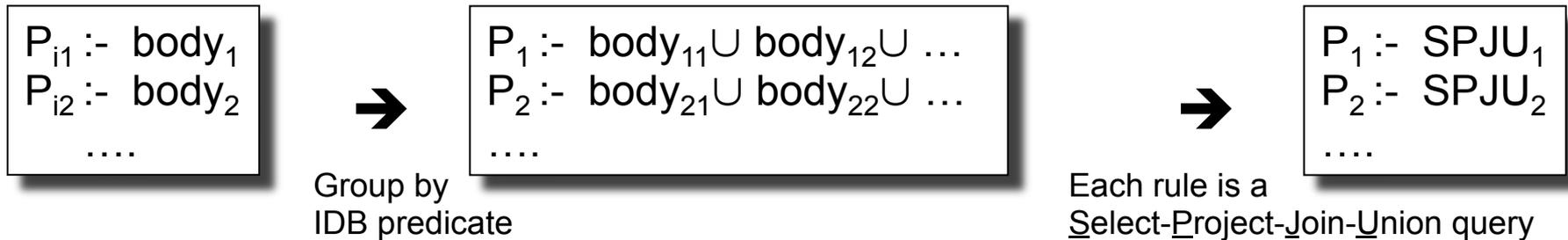


Example:  $T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

$\rightarrow T(x,y) :- R(x,y) \cup \Pi_{xy}(R(x,z) \bowtie T(z,y))$

# Naïve Datalog Evaluation Algorithm

Datalog program:



Naïve datalog evaluation algorithm:

$P_1 = P_2 = \dots = \emptyset$

Loop

New $P_1 = \text{SPJU}_1$ ; New $P_2 = \text{SPJU}_2$ ; ...  
if (New $P_1 = P_1$  and New $P_2 = P_2$  and ...) then exit

$P_1 = \text{New}P_1$ ;  $P_2 = \text{New}P_2$ ; ...

Endloop

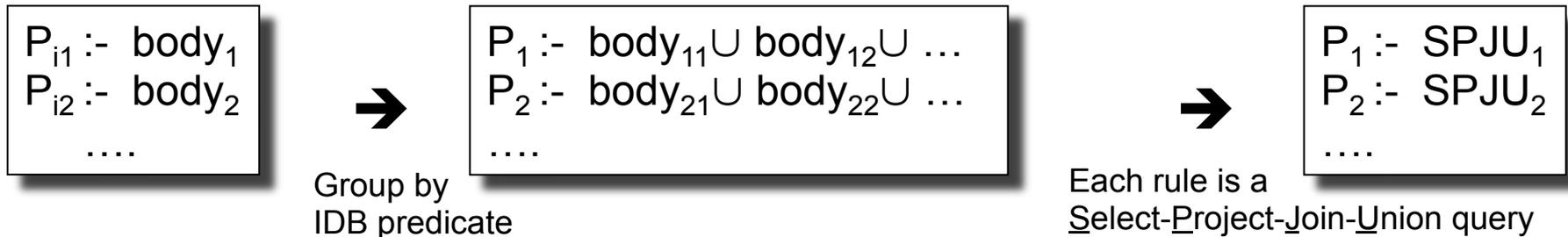
Example:

$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

→  $T(x,y) :- R(x,y) \cup \Pi_{xy}(R(x,z) \bowtie T(z,y))$

# Naïve Datalog Evaluation Algorithm

Datalog program:



Naïve datalog evaluation algorithm:

$P_1 = P_2 = \dots = \emptyset$

Loop

New $P_1 = \text{SPJU}_1$ ; New $P_2 = \text{SPJU}_2$ ; ...  
if (New $P_1 = P_1$  and New $P_2 = P_2$  and ...)  
then exit

$P_1 = \text{New}P_1$ ;  $P_2 = \text{New}P_2$ ; ...

Endloop

Example:

$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

$\rightarrow T(x,y) :- R(x,y) \cup \Pi_{xy}(R(x,z) \bowtie T(z,y))$

$T = \emptyset$

Loop

New $T(x,y) = R(x,y) \cup \Pi_{xy}(R(x,z) \bowtie T(z,y))$   
if (New $T = T$ )  
then exit

$T = \text{New}T$

Endloop

# Discussion

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- A datalog program always terminates (why?)
- A datalog program always runs in PTIME in the size of the database (why?)

# Problem with the Naïve Algorithm

---

- The same facts are discovered over and over again
- The semi-naïve algorithm tries to reduce the number of facts discovered multiple times

# Background: Incremental View Maintenance

---

Let  $V$  be a view computed by one datalog rule (no recursion)

$V \text{ :- body}$

If (some of) the relations are updated:  $R_1 \leftarrow R_1 \cup \Delta R_1, R_2 \leftarrow R_2 \cup \Delta R_2, \dots$

Then the view is also modified as follows:  $V \leftarrow V \cup \Delta V$

**Incremental view maintenance:**

Compute  $\Delta V$  without having to recompute  $V$

# Background: Incremental View Maintenance

---

Example 1:

$V(x,y) :- R(x,z), S(z,y)$

If  $R \leftarrow R \cup \Delta R$  then what is  $\Delta V(x,y)$  ?

# Background: Incremental View Maintenance

---

Example 1:

$V(x,y) :- R(x,z), S(z,y)$

If  $R \leftarrow R \cup \Delta R$  then what is  $\Delta V(x,y)$  ?

$\Delta V(x,y) :- \Delta R(x,z), S(z,y)$

# Background: Incremental View Maintenance

---

Example 2:

$V(x,y) :- R(x,z), S(z,y)$

If  $R \leftarrow R \cup \Delta R$  and  $S \leftarrow S \cup \Delta S$   
then what is  $\Delta V(x,y)$  ?

# Background: Incremental View Maintenance

---

Example 2:

$V(x,y) :- R(x,z), S(z,y)$

If  $R \leftarrow R \cup \Delta R$  and  $S \leftarrow S \cup \Delta S$   
then what is  $\Delta V(x,y)$  ?

$\Delta V(x,y) :- \Delta R(x,z), S(z,y)$

$\Delta V(x,y) :- R(x,z), \Delta S(z,y)$

$\Delta V(x,y) :- \Delta R(x,z), \Delta S(z,y)$

# Background: Incremental View Maintenance

---

Example 3:

$V(x,y) :- T(x,z), T(z,y)$

If  $T \leftarrow T \cup \Delta T$   
then what is  $\Delta V(x,y)$  ?

# Background: Incremental View Maintenance

---

Example 3:

$V(x,y) :- T(x,z), T(z,y)$

If  $T \leftarrow T \cup \Delta T$   
then what is  $\Delta V(x,y)$  ?

$\Delta V(x,y) :- \Delta T(x,z), T(z,y)$

$\Delta V(x,y) :- T(x,z), \Delta T(z,y)$

$\Delta V(x,y) :- \Delta T(x,z), \Delta T(z,y)$

# Semi-naïve Evaluation Algorithm

---

Separate the Datalog program into the non-recursive, and the recursive part.

Each  $P_i$  defined by non-recursive-SPJU $_i$  and (recursive-)SPJU $_i$ .

$P_1 = \Delta P_1 = \text{non-recursive-SPJU}_1,$

$P_2 = \Delta P_2 = \text{non-recursive-SPJU}_2,$

...

Loop

$\Delta P_1 = \Delta \text{SPJU}_1 - P_1; \Delta P_2 = \Delta \text{SPJU}_2 - P_2; \dots$

if ( $\Delta P_1 = \emptyset$  and  $\Delta P_2 = \emptyset$  and ...)

then break

$P_1 = P_1 \cup \Delta P_1; P_2 = P_2 \cup \Delta P_2; \dots$

Endloop

# Semi-naïve Evaluation Algorithm

Separate the Datalog program into the non-recursive, and the recursive part.  
Each  $P_i$  defined by non-recursive-SPJU $_i$  and (recursive-)SPJU $_i$ .

$P_1 = \Delta P_1 = \text{non-recursive-SPJU}_1, P_2 = \Delta P_2 = \text{non-recursive-SPJU}_2, \dots$

Loop

$\Delta P_1 = \Delta \text{SPJU}_1 - P_1; \Delta P_2 = \Delta \text{SPJU}_2 - P_2; \dots$

if ( $\Delta P_1 = \emptyset$  and  $\Delta P_2 = \emptyset$  and ...)

then break

$P_1 = P_1 \cup \Delta P_1; P_2 = P_2 \cup \Delta P_2; \dots$

Endloop

Example:

$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

$T = \Delta T = ?$  (non-recursive rule)

Loop

$\Delta T(x,y) = ?$  (recursive  $\Delta$ -rule)

if ( $\Delta T = \emptyset$ )

then break

$T = T \cup \Delta T$

Endloop

# Semi-naïve Evaluation Algorithm

Separate the Datalog program into the non-recursive, and the recursive part.  
Each  $P_i$  defined by non-recursive-SPJU $_i$  and (recursive-)SPJU $_i$ .

$P_1 = \Delta P_1 = \text{non-recursive-SPJU}_1, P_2 = \Delta P_2 = \text{non-recursive-SPJU}_2, \dots$

Loop

$\Delta P_1 = \Delta \text{SPJU}_1 - P_1; \Delta P_2 = \Delta \text{SPJU}_2 - P_2; \dots$

if ( $\Delta P_1 = \emptyset$  and  $\Delta P_2 = \emptyset$  and ...)

then break

$P_1 = P_1 \cup \Delta P_1; P_2 = P_2 \cup \Delta P_2; \dots$

Endloop

Example:

$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

$T(x,y) = R(x,y), \Delta T(x,y) = R(x,y)$

Loop

$\Delta T(x,y) = R(x,z), \Delta T(z,y) - R(x,y)$

if ( $\Delta T = \emptyset$ )

then break

$T = T \cup \Delta T$

Endloop

# Semi-naïve Evaluation Algorithm

Separate the Datalog program into the non-recursive, and the recursive part.  
Each  $P_i$  defined by non-recursive-SPJU $_i$  and (recursive-)SPJU $_i$ .

$P_1 = \Delta P_1 = \text{non-recursive-SPJU}_1, P_2 = \Delta P_2 = \text{non-recursive-SPJU}_2, \dots$

Loop

$\Delta P_1 = \Delta \text{SPJU}_1 - P_1; \Delta P_2 = \Delta \text{SPJU}_2 - P_2; \dots$

if ( $\Delta P_1 = \emptyset$  and  $\Delta P_2 = \emptyset$  and ...)

then break

$P_1 = P_1 \cup \Delta P_1; P_2 = P_2 \cup \Delta P_2; \dots$

Endloop

Example:

$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

Note: for any linear datalog programs,  
the semi-naïve algorithm has only  
one  $\Delta$ -rule for each rule!

$T(x,y) = R(x,y), \Delta T(x,y) = R(x,y)$

Loop

$\Delta T(x,y) = R(x,z), \Delta T(z,y) = R(x,y)$

if ( $\Delta T = \emptyset$ )

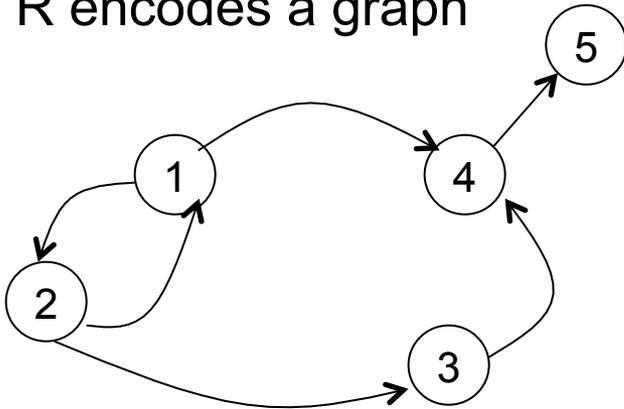
then break

$T = T \cup \Delta T$

Endloop

# Simple datalog programs

R encodes a graph



$T(x,y) :- R(x,y)$   
 $T(x,y) :- R(x,z), T(z,y)$

```

T = R, ΔT = R
Loop
  ΔT(x,y) = R(x,z), ΔT(z,y)
  -- R(x,y)
  if (ΔT = ∅)
    then break
  T = T ∪ ΔT
Endloop
  
```

R =

1	2
1	4
2	1
2	3
3	4
4	5

Initially:

ΔT =

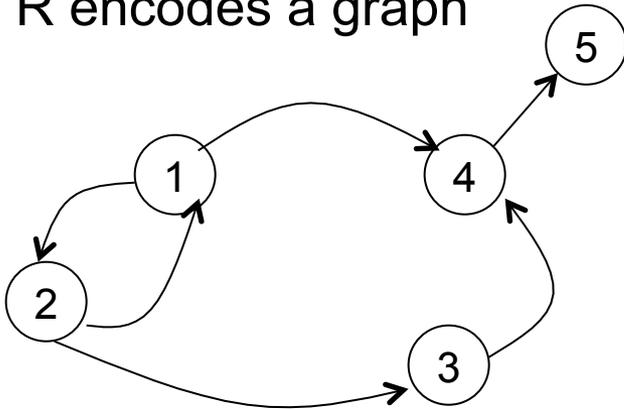
1	2
1	4
2	1
2	3
3	4
4	5

T =

1	2
1	4
2	1
2	3
3	4
4	5

# Simple datalog programs

R encodes a graph



```

T(x,y) :- R(x,y)
T(x,y) :- R(x,z), T(z,y)
  
```

```

T = R, ΔT = R
Loop
  ΔT(x,y) = R(x,z), ΔT(z,y)
  -- R(x,y)
  if (ΔT = ∅)
    then break
  T = T ∪ ΔT
Endloop
  
```

First iteration:

R =

1	2
1	4
2	1
2	3
3	4
4	5

Initially:

ΔT =

1	2
1	4
2	1
2	3
3	4
4	5

T =

1	2
1	4
2	1
2	3
3	4
4	5

T =

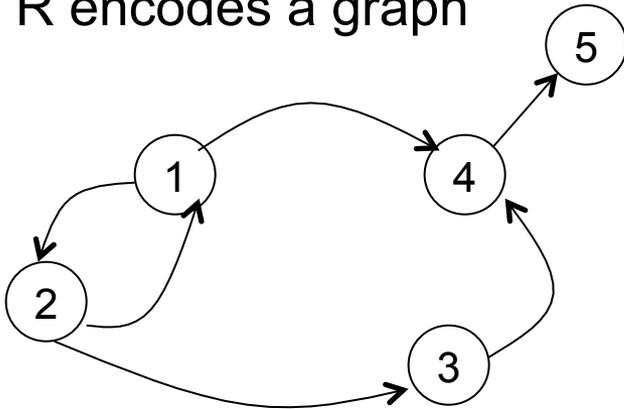
1	2
1	4
2	1
2	3
3	4
4	5
1	1
1	3
1	5
2	2
2	4
3	5

ΔT =  
paths of  
length 2

1	1
1	3
1	5
2	2
2	4
3	5

# Simple datalog programs

R encodes a graph



```

T(x,y) :- R(x,y)
T(x,y) :- R(x,z), T(z,y)
  
```

```

T = R, ΔT = R
Loop
  ΔT(x,y) = R(x,z), ΔT(z,y)
  -- R(x,y)
  if (ΔT = ∅)
    then break
  T = T ∪ ΔT
Endloop
  
```

R =

1	2
1	4
2	1
2	3
3	4
4	5

Initially:

ΔT =

1	2
1	4
2	1
2	3
3	4
4	5

T =

1	2
1	4
2	1
2	3
3	4
4	5

First iteration:

ΔT = paths of length 2

1	1
1	3
1	5
2	2
2	4
3	5

T =

1	2
1	4
2	1
2	3
3	4
4	5
1	1
1	3
1	5
2	2
2	4
3	5

Second iteration:

ΔT = paths of length 3

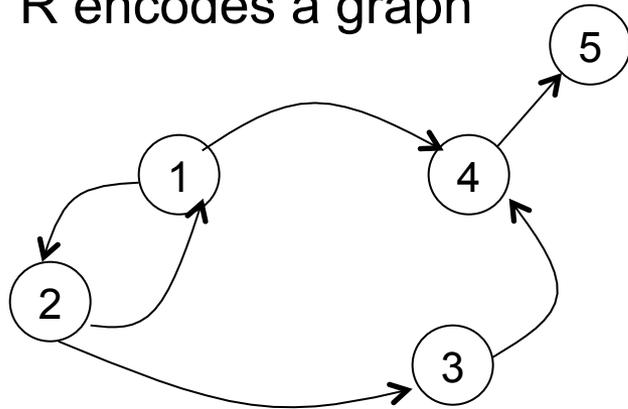
1	2
1	4
2	1
2	3
2	5

T =

1	2
1	4
2	1
2	3
3	4
4	5
1	1
1	3
1	5
2	2
2	4
3	5
2	5

# Simple datalog programs

R encodes a graph



```
T(x,y) :- R(x,y)
T(x,y) :- R(x,z), T(z,y)
```

```
T = R, ΔT = R
Loop
  ΔT(x,y) = R(x,z), ΔT(z,y)
  -- R(x,y)
  if (ΔT = ∅)
    then break
  T = T ∪ ΔT
Endloop
```

First iteration:

Second iteration:

Third iteration:

R =

1	2
1	4
2	1
2	3
3	4
4	5

Initially:

ΔT =

1	2
1	4
2	1
2	3
3	4
4	5

T =

1	2
1	4
2	1
2	3
3	4
4	5

ΔT = paths of length 2

1	1
1	3
1	5
2	2
2	4
3	5

T =

1	2
1	4
2	1
2	3
3	4
4	5
1	1
1	3
1	5
2	2
2	4
3	5

ΔT = paths of length 3

1	2
1	4
2	1
2	3
2	5

T =

1	2
1	4
2	1
2	3
3	4
4	5
1	1
1	3
1	5
2	2
2	4
3	5
2	5

ΔT = paths of length 4

--	--

# Discussion of Semi-Naïve Algorithm

---

- Avoids re-computing some tuples, but not all tuples
- Easy to implement, no disadvantage over naïve
- A rule is called linear if its body contains only one recursive IDB predicate:
  - A linear rule always results in a single incremental rule
  - A non-linear rule may result in multiple incremental rules

# Adding Negation: Datalog<sup>-</sup>

---

**Example:** compute the complement of the transitive closure

```
T(x,y) :- R(x,y)
T(x,y) :- T(x,z), R(z,y)
CT(x,y) :- Node(x), Node(y), not T(x,y)
```

What does this mean??

# Recursion and Negation Don't Like Each Other

---

EDB:  $I = \{ R(a) \}$

$S(x) :- R(x), \text{ not } T(x)$   
 $T(x) :- R(x), \text{ not } S(x)$

What are the possible outcomes of S and T?

# Recursion and Negation Don't Like Each Other

---

EDB:  $I = \{ R(a) \}$

$S(x) :- R(x), \text{ not } T(x)$   
 $T(x) :- R(x), \text{ not } S(x)$

What are the possible outcomes of S and T?

$$J_1 = \{ \}$$

$$J_2 = \{ S(a) \}$$

$$J_3 = \{ T(a) \}$$

$$J_4 = \{ S(a), T(a) \}$$

# Adding Negation: datalog<sup>-</sup>

---

- **Solution 1: Stratified Datalog<sup>-</sup>**
  - Rules must be partitioned into strata
  - IDB predicates defined in strata  $\leq k$  may be negated in strata  $\geq k+1$
- **Solution 2: Inflationary-fixpoint Datalog<sup>-</sup>**
  - Fire rules and always add facts (never retract)
  - Stop when nothing new is added
  - Always terminates (why ?)
- **Solution 3: Partial-fixpoint Datalog<sup>-,\*</sup>**
  - Fire rules, adding/retracting facts as needed
  - Stop when reaching a fixpoint
  - May not terminate
- **Solution 4: Well-founded semantics** (next lecture)

What semantics does the paper use?