

# **Processing of Declarative Knowledge**

## **–Datalog–**

Francesco Ricca

Computational Intelligence Curriculum  
Institute of Information Systems

# What is Datalog?

## Datalog:

- **A logic language for querying databases**
- Overcomes some limits of Relational Algebra and SQL
  - Recursive definitions

## Why Datalog?

- The basic fragment of ASP
  - Deductive database applications, query answering

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# Datalog Syntax

**Rule:**

$head(\bar{H}) \text{ :- } body_1(\bar{X}_1), \dots, body_n(\bar{X}_n).$

**Intuitively:**

infer  $head(\bar{h})$  if  $body_1(\bar{x}_1), \dots, body_n(\bar{x}_n)$  is true.

**Fact:**

A rule with empty body ( :- symbol is omitted)

→ *Facts are true and model the input database* ←

**Variables:**

are allowed in atom's arguments, Prolog-like syntax

**Safety:**

all variables must occur in the body



# Datalog Syntax

## Example

### Program and query:

*father(X) :- parent(X, Y), male(X).*

### Database:

*male(rob).*

*parent(rob, ann).*

*parent(rob, ann).*

*parent(mary, ann).*

### Query Result:

*father(rob).*

# Recursive Example Datalog

## Example (Reachable airports)

**Input:** A set of direct connections between some cities represented by *connected*(\_, \_).

**Query:** Retrieve all the cities reachable by flight from Vienna airport, through a direct or undirect connection.

*...can you write an SQL query?*

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**Query:** Retrieve all the cities reachable by flight from Vienna airport, through a direct or undirect connection.

**Datalog:**

*reaches*(*vienna*, *B*) :- *connected*(*vienna*, *B*).

*reaches*(*vienna*, *C*) :- *reaches*(*vienna*, *B*), *connected*(*B*, *C*).

# Datalog Programs (1)

## Datalog Program:

- A set of rules
- **EDB**: predicates appearing only in bodies or in facts
- **IDB** : predicates defined (also) by rules

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## Example (Reachability)

**Input:** a graph encoded by relation *edge*(\_, \_).

**Problem:** Find all pairs of reachable nodes.

% if there is an edge from X to Y

% then X is reachable from Y

*reachable*(X, Y) :- *edge*(X, Y).

% Reachability is transitive

*reachable*(X, Y) :- *reachable*(X, Z), *edge*(Z, Y).

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# Datalog Programs

## Example (Reachability)

**Input:** a graph encoded by relation  $edge(\_, \_)$ .

**Problem:** Find all pairs of reachable nodes.

% if there is an edge from X to Y

% then X is reachable from Y

$reachable(X, Y) :- edge(X, Y).$

% Reachability is transitive

$reachable(X, Y) :- reachable(X, Z), edge(Z, Y).$

**Intuitive meaning:** *(bottom-up evaluation)*

*“Start with the facts in the EDB and iteratively derive facts for IDBs until no new fact is derived.”*



# Fully Declarative Language

## Example (Ancestor)

**Input:** parent relation modeled by *parent*(\_, \_).

**Problem:** Define the relation of arbitrary ancestors.

**Solution 1:**

*ancestor*(A, B) :- *parent*(A, B).

*ancestor*(A, C) :- *ancestor*(A, B), *ancestor*(B, C).

# Fully Declarative Language

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**Solution 1:**

*ancestor*(A, B) :- *parent*(A, B).

*ancestor*(A, C) :- *ancestor*(A, B), *ancestor*(B, C).

**Solution 2:**

*ancestor*(A, B) :- *parent*(A, B).

*ancestor*(A, C) :- *ancestor*(A, B), *parent*(B, C).

# Fully Declarative Language

## Example (Ancestor)

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*ancestor*(A, C) :- *ancestor*(A, B), *ancestor*(B, C).

**Solution 3:** Declarative: Atoms' and Rules' order is immaterial!

*ancestor*(A, C) :- *ancestor*(A, B), *parent*(B, C). ← No LOOP!

*ancestor*(A, B) :- *parent*(A, B).

# Arithmetic Expressions and Builtins

## Arithmetic and comparison operators

- $<, >, <=, >=, =$
- $+, -, *, /$

### Example (Fibonacci numbers)

*fib*(0, 1).

*fib*(1, 1).

*fib*( $N + 2$ ,  $Y1 + Y2$ )  $\text{:-}$  *fib*( $N$ ,  $Y1$ ), *fib*( $N + 1$ ,  $Y2$ ).

For recursive definitions an upper bound for integers (system setting) or a domain has to be specified.

# Example pure Datalog limits

## Example (No Peroni here!)

**Input:** Information about bars and beers represented by facts of the form

*beers(name, manufacturer). sells(bar, beer)*

**Query:** Retrieve all bars that **do not** sell Peroni

*...can you write an Datalog query?*

# Example pure Datalog limits

## Example (No Peroni here!)

**Input:** Information about bars and beers represented by facts of the form

*beers(name, manufacturer). sells(bar, beer)*

**Query:** Retrieve all bars that **do not** sell Peroni

**Datalog:**

*noPeroni(Bar) :- sells(Bar, Beer),*

*not sellsPeroni(Bar).*

*sellsPeroni(Bar) :- sells(Bar, Beer), beer(Beer, peroni).*

# Datalog with Negation

## Rule:

$$\text{head}(\overline{H}) \text{ :- } \text{body}_1(\overline{X}_1), \dots, \text{body}_n(\overline{X}_n), \\ \text{not } \text{body}_{n+1}(\overline{X}_{n+1}), \dots, \text{not } \text{body}_m(\overline{X}_m).$$

## Positive and Negative Body:

$$\text{body}_1(\overline{x}_1), \dots, \text{body}_n(\overline{x}_n) \leftarrow \text{positive body} \\ \text{body}_{n+1}(\overline{x}_{n+1}), \dots, \text{body}_m(\overline{x}_m) \leftarrow \text{negative body}$$

## Intuitively:

infer  $\text{head}(\overline{h})$  if all atoms in the positive body are true  
and all atoms in the negative body are false

## Safety:

all variables must occur in a positive body literal

## Stratification (intuitive):

negation must not be involved in recursive definitions!

# Stratification (i.e., no recursion through negation)

## Example (Unstratified Program)

$$\begin{aligned}p(X) &:- I(X), \text{ not } q(X). \\ q(X) &:- I(X), \text{ not } p(X).\end{aligned}$$

## Example (Stratified Program)

$$\begin{aligned}p(X) &:- p(X), \text{ not } q(X). \\ q(X) &:- I(X), \text{ not } m(b).\end{aligned}$$



# Needed Restrictions

## Safety:

$s(X) \text{ :- } r(Y).$

$s(X) \text{ :- } \textit{notr}(X).$

$s(X) \text{ :- } r(Y), X < Y.$

## Intuitively:

In each of these cases the result is infinite!?!

## Stratification:

Negation wrapped inside recursion is not that obvious

$a \text{ :- not } b. \quad b \text{ :- not } a.$

More on this later...

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More on this later...

# Practice

## **Download a Datalog implementation**

`http://www.dlvsystem.com`

## **Download a GUI**

`http://www.mat.unical.it/ricca/aspid`

# Syntax & Notation

**Terms:** Constants and Variables

**Atoms:** of the form  $\text{predicate}(t_1, \dots, t_n)$

**Literals:** atoms  $a$  (pos.) and negated atoms  $\text{not } a$  (neg.)

**Rules:**  $h \text{ :- } p_1, \dots, p_n, \text{not } n_1, \dots \text{not } n_n.$

**Head:**  $H(r) = h$

**Body:**  $B(r) = \{p_1, \dots, p_n, \text{not } n_1, \dots \text{not } n_n.\}$

**Positive Body:**  $B^+(r) = \{p_1, \dots, p_n\}$

**Negative Body:**  $B^-(r) = \{\text{not } n_1, \dots \text{not } n_n\}$

**Program:** A set of rules

**Safety:** All variables occur in some positive body atom

**Ground:** no variable occurs in it

**Positive Program:** all rules are such that  $B^-(r) = \emptyset$

# Semantics Positive Programs

**Interpretation:** a set  $I$  of ground atoms

- atom  $a$  is true w.r.t.  $I$  if  $a \in I$ , it is false otherwise, and
- negative literal  $\text{not } a$  is true w.r.t.  $I$  if  $a \notin I$ , it is false otherwise.

**Satisfaction:** a rule  $r$  is satisfied w.r.t.  $I$  if  $H(r) \in I$  whenever all literals  $\ell \in B(r)$  are true w.r.t.  $I$

**Model:** an interpretation  $I$  is a model for program  $P$  if all rules in  $P$  are satisfied by  $I$

**Least Model:** an interpretation  $I$  is the least or minimal model for program  $P$  if every  $I' \subset I$  is not a model for  $P$

# Example Models

**Given:**

$a : \neg b, c.$

$c : \neg d.$

$d.$

**Interpretations and Models:**

$I_1 = \{b, c, d\}, I_2 = \{a, b, c, d\} \quad I_3 = \{c, d\}$

→ only  $I_2$  and  $I_3$  are models!

# Example Models

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**Interpretations and Models:**

$I_1 = \{b, c, d\}, I_2 = \{a, b, c, d\} \quad I_3 = \{c, d\}$

→ only  $I_2$  and  $I_3$  are models!

→  $I_3$  is minimal!

# Semantics Positive Programs

**Rule Instantiation:**  $I(r)$  is the set ground rules that can be obtained by replacing every variable in  $r$  by a constant occurring in  $P$

**Instantiation:**  $G(P) = \cup_{r \in P} I(r)$

**Model:** an interpretation  $M$  is a model for program  $P$  if  $M$  is a model of  $G(P)$

**Least Model:** an interpretation  $M$  is the least model of program  $P$  if  $M$  is the least model of  $G(P)$



# Operational Semantics Positive Programs (Ground case)

**Immediate Consequence Operator:** Given ground program  $P$ , and Interpretation  $I$

$$T_p(I) = \{a \mid \exists r \in P \text{ s.t. } H(r) = a \vee \forall l \in B(r) \text{ are true in } I\}$$

**Example:**  $a :- b. \quad c :- d. \quad e :- a. \quad I = \{b\} \quad T_p(I) = \{a\}.$

**Fixpoint procedure:**

- Start with  $I = \emptyset$
- Repeatedly apply  $T_p$  until a fixpoint  $T_p(I) = I$  is reached.

**Least Model:** The least fixpoint  $T_p$ .

**Theorem:** A positive Datalog program  $P$  has a unique least model, which is the minimal model corresponding to the intersection of all models of  $P$ .

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# Operational Semantics

## Ground + Fixpoint:

Given  $P$ , build  $G(P)$ , apply operator to compute fixpoint  
 $T_{G(P)}(M) = M$

### Consider:

$a(X) : \neg b(X), c(X).$

$b(a). b(b). c(a). c(c).$

### Instantiation:

$a(a) : \neg b(a), c(a).$

$a(b) : \neg b(b), c(b).$

$a(c) : \neg b(c), c(c).$

...

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... **Do we need all ground rules?**

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... **Do they have any chance to be satisfied?**

# Operational Semantics

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## Instantiation:

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$a(c) : \neg b(c), c(c).$

... Start from facts, match bodies, apply ... fixpoint!



# Example Semantics

## Consider:

*grandParent*(*X*, *Y*) :- *parent*(*X*, *Z*), *parent*(*Z*, *Y*).  
*parent*(*a*, *b*). *parent*(*b*, *c*).

## Evaluation:

- 1  $I = \{parent(a, b), parent(b, c)\}$
- 2 the body can be instantiated  
(*parent*(*a*, *b*), *parent*(*b*, *c*))  
 $I := I \cup \{grandParent(a, c)\}$
- 3 no body can be matched with atoms in *I* ... STOP!

**Results:**  $\{parent(a, b), parent(b, c), grandParent(a, c)\}$  is the least model

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(*parent*(*a*, *b*), *parent*(*b*, *c*))  
 $I := I \cup \{grandParent(a, c)\}$
- 3 no body can be matched with atoms in *I* ... **STOP!**

**Results:**  $\{parent(a, b), parent(b, c), grandParent(a, c)\}$  is the least model

# Semantics c.t.d.

## Immediate Consequence Operator:

Given ground program  $P$ , and Interpretation  $I$

$$T_p(I) = \{H(r_g) \mid \exists r_g \text{ instantiating } r \in P \text{ s.t.} \\ \text{the body of } r_g \text{ is true w.r.t. } I\}$$

## Operational Semantics:

Compute  $M = T_p(M)$  by repeatedly applying  $T_p$  starting from EDB.

# Stratified Programs

**Dependency Graph:** Given program  $P$ , graph  $DG(P)$  is as follows:

- a node  $p$  in  $V$  for each predicate in  $p$  occurring in  $P$
- positive edge  $p \leftarrow q$  if there is rule  $r$  s.t.  $p$  occurs in  $H(r)$  and  $q$  occurs in  $B^+(r)$
- negative edge  $p \leftarrow_n q$  if there is rule  $r$  s.t.  $p$  occurs in  $H(r)$  and  $q$  occurs in  $B^-(r)$

**Recursive Program:**  $P$  is recursive if  $DG(P)$  is cyclic.

**Stratified Program:**  $P$  is stratified if no cycle in  $DG(P)$  contains a negative edge.

# Negation and Recursion

## Consider:

$p(X) \text{ :- } q(X), \text{ not } p(X).$

$q(1). q(2).$

## Evaluation:

- 1  $q = \{(1), (2)\}, p = \{\}$
- 2  $q = \{(1), (2)\}, p = \{(1), (2)\}$
- 3  $q = \{(1), (2)\}, p = \{\}$
- 4 ...

# Stratified Program

## Consider:

$r_1 : \text{reach}(X) : \neg \text{source}(X).$

$r_2 : \text{reach}(X) : \neg \text{reach}(Y), \text{arc}(Y, X).$

$r_3 : \text{noReach}(X) : \neg \text{target}(X), \text{notreach}(X).$

## Dependency Graph:

- $V = \{\text{reach}, \text{source}, \text{target}, \text{noReach}, \text{arc}\}$
- $E = \{(\text{reach}, \text{source}), (\text{reach}, \text{reach}), (\text{reach}, \text{arc}), (\text{noReach}, \text{target}), (\text{noReach}, \text{reach})_n\}$
- cyclic, but stratified!

# Stratified Program

## Consider:

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## Dependency Graph:

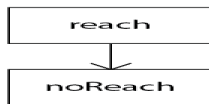
- $V = \{\text{reach}, \text{source}, \text{target}, \text{noReach}, \text{arc}\}$
- $E = \{(\text{reach}, \text{source}), (\text{reach}, \text{reach}), (\text{reach}, \text{arc}), (\text{noReach}, \text{target}), (\text{noReach}, \text{reach})_n\}$
- cyclic, but stratified!



# Stratified Program - components and modules

## Components and Subprograms:

- Let  $\text{Comp}(DG)$  be the set of the strongly connected components of  $DG$
- Given  $C \in \text{Comp}(DG)$  the subprogram associated to  $C$  is  $\text{Sub}(P, C) = \{r \in P \text{ s.t. } H(r) \in C\}$
- Given  $C'$  depends on  $C''$  if there is some (negative) arc in  $DG$  from a node in  $C''$  to a node in  $C'$



## Example ctd:

- $\text{Comp}(DG) = \{\{reach\}, \{noReach\}\}$
- $\text{Sub}(P, \{reach\}) = \{r_1, r_2\}$
- $\text{Sub}(P, \{noReach\}) = \{r_3\}$

# Stratified Program - Evaluation

## Evaluation:

- Start from the components that do not depend on other components
- Evaluate subprograms associated to components as for positive programs
- Remove evaluated components
- repeat until all components are evaluated

## Example ctd:

- 1 Evaluate  $\{\{reach\}\}$
- 2 Evaluate  $\{\{noReach\}\}$

# Example Stratified Program

## Consider:

$r_1 : reach(X) : -source(X).$

$r_2 : reach(X) : -reach(Y), arc(Y, X).$

$r_3 : noReach(X) : -target(X), notreach(X).$

EDB:  $node(1).node(2).node(3).node(4).arc(1, 2).$   
 $arc(3, 4).arc(4, 3).source(1), target(2).target(3).$

**Evaluate**  $Sub(P, \{reach\}) = \{r_1, r_2\}$ :

- 1  $I = \{source(1), target(2), target(3), \dots\}$
- 2  $I := I \cup \{reach(1)\}$
- 3  $I := I \cup \{reach(2)\} \dots \text{STOP!}$

**Evaluate**  $Sub(P, \{noReach\}) = \{r_3\}$ :

- 1  $I := I \cup \{noReach(3)\} \dots \text{STOP!}$

# Example Stratified Program

## Consider:

$r_1 : \text{reach}(X) : -\text{source}(X).$

$r_2 : \text{reach}(X) : -\text{reach}(Y), \text{arc}(Y, X).$

$r_3 : \text{noReach}(X) : -\text{target}(X), \text{notreach}(X).$

EDB:  $\text{node}(1).\text{node}(2).\text{node}(3).\text{node}(4).\text{arc}(1, 2).$   
 $\text{arc}(3, 4).\text{arc}(4, 3).\text{source}(1), \text{target}(2).\text{target}(3).$

**Evaluate**  $\text{Sub}(P, \{\text{reach}\}) = \{r_1, r_2\}$ :

- ①  $I = \{\text{source}(1), \text{target}(2), \text{target}(3), \dots\}$
- ②  $I := I \cup \{\text{reach}(1)\}$
- ③  $I := I \cup \{\text{reach}(2)\} \dots \text{STOP!}$

**Evaluate**  $\text{Sub}(P, \{\text{noReach}\}) = \{r_3\}$ :

- ①  $I := I \cup \{\text{noReach}(3)\} \dots \text{STOP!}$

# Example Stratified Program

## Consider:

$r_1 : reach(X) : -source(X).$

$r_2 : reach(X) : -reach(Y), arc(Y, X).$

$r_3 : noReach(X) : -target(X), notreach(X).$

EDB:  $node(1).node(2).node(3).node(4).arc(1, 2).$   
 $arc(3, 4).arc(4, 3).source(1), target(2).target(3).$

**Evaluate**  $Sub(P, \{reach\}) = \{r_1, r_2\}$ :

- ①  $I = \{source(1), target(2), target(3), \dots\}$
- ②  $I := I \cup \{reach(1)\}$
- ③  $I := I \cup \{reach(2)\} \dots \text{STOP!}$

**Evaluate**  $Sub(P, \{noReach\}) = \{r_3\}$ :

- ①  $I := I \cup \{noReach(3)\} \dots \text{STOP!}$

## Exercise... limits

Given the following relational database schema

(\* indicates primary keys):

- *beers(name\*, manufacturer)*
- *sells(bar\*, beer\*, price)*
- *associate(bar, bar)*

Write the following (if possible) in SQL, and Datalog

- 1 find the manufacturers of the beers "John's bar" sells
- 2 find the number of beers that "John's bar" sells at a price higher than "Anns's bar"
- 3 find the bars that sell exactly two beers
- 4 find the bars that sell more than three beers
- 5 find the bars that are associated, directly or indirectly, through a chain of bar associations to "John's bar"
- 6 find the most expensive beer
- 7 find the bars that sell more beers