

**Systems and Solving Techniques for Knowledge
Representation and Reasoning:
Datalog (part II)**

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Syntax & Notation

Terms: Constants and Variables

Atoms: of the form $predicate(t_1, \dots, t_n)$

Literals: atoms a (pos.) and negated atoms $not\ a$ (neg.)

Rules: $h :- p_1, \dots, p_n, not\ n_1, \dots, not\ n_n.$

Head: $H(r) = h$

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

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

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

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:

$a : -b, c.$

$c : -d.$

$d.$

Interpretations and Models:

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

→ only I_2 and I_3 are models!

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Interpretations and Models:

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→ only I_2 and I_3 are models!

→ I_3 is minimal!

Semantics Positive Programs

Rule Instantiation: given a rule r , $Inst(r)$ is the set of ground rules that can be obtained by replacing every variable in r by a constant occurring in a program P

Instantiation: given a program P , $G(P) = \cup_{r \in P} Inst(r)$

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 for Positive Programs (Ground case)

Immediate Consequence Operator: Given a ground program P , and an interpretation I

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

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

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 (Non-ground case)

Ground + Fixpoint:

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

Consider:

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

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

Instantiation:

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

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

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...

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Instantiation:

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

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

Consider:

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Instantiation:

$a(a) : -b(a), c(a)$.

$a(b) : -b(b), c(b)$.

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

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

Consider:

$a(X) : -b(X), c(X)$.

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Instantiation:

$a(a) : -b(a), c(a)$.

$a(b) : -b(b), c(b)$.

$a(c) : -b(c), c(c)$.

... 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 $I = \emptyset$
- 2 $I = \{parent(a, b), parent(b, c)\}$
- 3 body can be instantiated ($parent(a, b), parent(b, c)$)
Apply $T_P: I := I \cup \{grandParent(a, c)\}$
- 4 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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Semantics c.t.d.

Immediate Consequence Operator:

Given a non-ground program P , and an 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 a program P , the graph $DG(P) := (V, E)$ is defined as follows:

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

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

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

Evaluation:

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

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

3 ...

Stratified Program

Consider:

$r_1 : reach(X) : \neg source(X).$

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

$r_3 : noReach(X) : \neg target(X), not reach(X).$

Dependency Graph:

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

Stratified Program

Consider:

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

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

$r_3 : noReach(X) : -target(X), not reach(X).$

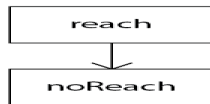
Dependency Graph:

- $V = \{reach, source, target, noReach, arc\}$
- $E = \{(reach, source), (reach, reach), (reach, arc), (noReach, target), (noReach, reach)_n\}$
- 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:

- 1 Start from the components that do not depend on other components
- 2 Evaluate subprograms associated to components as for positive programs
- 3 Remove evaluated components
- 4 Go to step 2. if still components are to be evaluated

Example ctd:

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

Example Stratified Program

Consider:

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

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

$r_3 : noReach(X) : -target(X), not reach(X).$

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:

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

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- 2 $I := I \cup \{reach(1)\}$
- 3 $I := I \cup \{reach(2)\} \dots$ **STOP!**

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

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**Thanks to Francesco Ricca for a preliminary
version of these slides**