

Answer Set Programming: A Primer



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Unit 1 – Basic Concepts & Properties

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Unit Outline

1. Introduction

1.1 Roots of ASP

1.2 Prolog

2. Horn Logic Programming

2.1 Positive Logic Programs

2.2 Minimal Model Semantics

3. Negation in Logic Programs

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4. Stable Logic Programming

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5.2 Strong Negation

5.3 Disjunction

Introduction

- Answer Set Programming (ASP) is a recent problem solving approach
- The term was coined by Vladimir Lifschitz [1999,2002]
- Proposed by other people at about the same time, e.g. [Marek and Truszczyński, 1999],[Niemelä, 1999]
- It has roots in KR, logic programming, and nonmonotonic reasoning
- At an abstract level, relates to SAT solving and CSP.
- Book: [Baral, 2003]

French Phrases, Italian Soda (*Dell Logic Puzzles*)

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Question:

What is each of them drinking, and where is each of them going ?

Sudoku

	6		1		4		5	
		8	3		5	6		
2								1
8			4		7			6
		6				3		
7			9		1			4
5								2
		7	2		6	9		
	4		5		8		7	

Task:

Fill in the grid so that every row, every column, and every 3x3 box contains the digits 1 through 9.

Wanted!

A general-purpose approach for modeling and solving these and many other problems.

Issues:

- Diverse domains
- Spatial and temporal reasoning
- Constraints
- Incomplete information
- Preferences and priority

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Proposal:

Answer Set Programming (ASP) paradigm!

Roots of ASP – Knowledge Representation (KR)

How to model

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- Commonsense reasoning
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Approach

- use a logic-based formalism
- Inherent feature: nonmonotonicity

Many logic-based KR formalisms have been developed.

Logic Programming – Prolog revisited

1960s/70s: Logic as a Programming Language (?)

- Breakthrough in Computational Logic by Robinson's discovery of the Resolution Principle (1965)

Kowalski (1979):

ALGORITHM = LOGIC + CONTROL

- Knowledge for problem solving (LOGIC)
- “Processing” of the knowledge (CONTROL)

Prolog

Prolog = “Programming in Logic”

- Basic data structures: terms
- Programs: rules and facts
- Computing: Queries (goals)
 - Proofs provide answers
 - SLD-resolution
 - unification - basic mechanism to manipulate data structures
- Extensive use of recursion

Example (Recursion)

```
append([], X, X) .
```

```
append([X|Y], Z, [X|T]) :- append(Y, Z, T) .
```

```
reverse([], []).
```

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reverse([X|Y], Z) :- append(U, [X], Z), reverse(Y, U) .
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- Both relations are defined recursively.
- Terms represent complex objects: lists, sets, ...

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Problem:

Reverse the list `[a, b, c]`

Ask query: `?- reverse([a, b, c], X) .`

- A proof of the query yields a substitution: $X = [c, b, a]$
- The substitution constitutes an answer

The key: Techniques to search for proofs

- Understanding of the resolution mechanism is important
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Query: ?- reverse([a|X], [b,c,d,b])

Is this truly declarative programming?

Desiderata

Relieve the programmer from several concerns.

It is desirable that

- the order of program rules does not matter;
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“Pure” declarative programming

- Prolog does not satisfy these desiderata
- Satisfied e.g. by the answer set semantics of logic programs

Positive Logic Programs

Definition (Positive Logic Program)

A *positive logic program* P is a finite set of clauses (rules) in the form

$$a \leftarrow b_1, \dots, b_m, \quad (1)$$

where a, b_1, \dots, b_m are atoms of a first-order language L .

- a is the *head* of the rule
- b_1, \dots, b_m is the *body* of the rule.
- If $m = 0$, the rule is a *fact* (written shortly a)

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Example

$$\begin{aligned} \text{connected}(\text{cagliari}) &\leftarrow \text{hub}(\text{rome}), \text{link}(\text{rome}, \text{cagliari}) \\ \text{connected}(X) &\leftarrow \text{hub}(Y), \text{link}(Y, X) \end{aligned}$$

Herbrand Semantics

Definition (Herbrand Universe, Base, Interpretation)

Given a logic program P , the **Herbrand universe** of P , $HU(P)$, is the set of all terms which can be formed from constants and functions symbols in P (resp. the vocabulary, if explicitly known).

The **Herbrand base** of P , $HB(P)$, is the set of all ground atoms which can be formed from predicates and terms $t \in HU(P)$.

A (Herbrand) **interpretation** is a first-order interpretation $I = (D, \cdot^I)$ of the vocabulary with domain $D = HU(P)$ where each term $t \in HU(P)$ is interpreted by itself, i.e., $t^I = t$.

I is identified with the set $\{ p(t_1, \dots, t_n) \in HB(P) \mid \langle t_1^I, \dots, t_n^I \rangle \in p^I \}$.

Informally, a (Herbrand) interpretation can be seen as a set denoting which ground atoms are true in a given scenario.

Example (Program P_1)

$$h(0, 0).$$
$$t(a, b, r).$$
$$p(0, 0, b).$$
$$p(f(X), Y, Z) \leftarrow p(X, Y, Z'), h(X, Y), t(Z, Z', r).$$
$$h(f(X), f(Y)) \leftarrow p(X, Y, Z'), h(X, Y), t(Z, Z', r).$$

Constant symbols: $0, a, b, r$; Function symbols: f .

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Constant symbols: $0, a, b, r$; Function symbols: f .

$$HU(P_1): \{ 0, a, b, r, f(0), f(f(0)), \dots f^i(0), \dots f(a), f(f(a)), \dots \}$$

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Some Herbrand interpretations:

$$I_1 = \emptyset;$$

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$$I_1 = \emptyset; \quad I_2 = HB(P_1);$$

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Some Herbrand interpretations:

$$I_1 = \emptyset; \quad I_2 = HB(P_1); \quad I_3 = \{ h(0, 0), t(a, b, r), p(0, 0, b) \}; \dots$$

Grounding

The semantics of positive logic programs is defined in terms of grounding.

Definition (ground instance, grounding)

A *ground instance* of a clause C of the form (1) is any clause C' obtained from C by applying a substitution

$$\theta: \text{Var}(C) \rightarrow HU(P)$$

to the variables in C , denoted as $\text{Var}(C)$.

- $\text{grnd}(C)$ denotes the set of all possible ground instances of C
- for any program P , the *grounding* of P is $\text{grnd}(P) = \bigcup_{C \in P} \text{grnd}(C)$.

Roughly speaking, C is a shortcut denoting $\text{grnd}(C)$, and each variable appearing in C ranges over the Herbrand universe.

Example (Program P_2)

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The ground instances of the first rule are

$$p(f(0), 0, 0) \leftarrow p(0, 0, 0), h(0, 0), t(0, 0, r). \quad X = Y = Z = Z' = 0$$

...

$$p(f(0), r, 0) \leftarrow p(0, r, 0), h(0, r), t(0, 0, r). \quad X = Z = Z' = 0, Y = r$$

...

$$p(f(r), r, r) \leftarrow p(r, r, r), h(r, r), t(r, r, r). \quad X = Y = Z = Z' = r$$

...

$$p(f(f(0)), 0, 0) \leftarrow p(f(0), 0, 0), h(f(0), 0), t(0, 0, r). \quad X = Y = Z = Z' = 0$$

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$$p(f(r), r, r) \leftarrow p(r, r, r), h(r, r), t(r, r, r). \quad X = Y = Z = Z' = r$$

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$$p(f(f(0)), 0, 0) \leftarrow p(f(0), 0, 0), h(f(0), 0), t(0, 0, r). \quad X = Y = Z = Z' = 0$$

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The single ground instance of the second rule is

$$h(0, 0).$$

Herbrand Models

Definition (Model, satisfaction)

An interpretation I is a (Herbrand) *model* of a

- a ground (variable-free) clause $C = a \leftarrow b_1, \dots, b_m$, if either $\{b_1, \dots, b_m\} \not\subseteq I$ or $a \in I$; $(I \models C)$
- a clause C , if $I \models C'$ for every $C' \in \text{grnd}(C)$; $(I \models C)$
- a program P , if $I \models C$ for every clause C in P . $(I \models C)$

Example (Program P_2 cont'd)

$$p(f(X), Y, Z) \leftarrow p(X, Y, Z'), h(X, Y), t(Z, Z', r). \\ h(0, 0).$$

Which of the following interpretations are models of P_2 ?

- $I_1 = \emptyset$
- $I_2 = HB(P_2)$
- $I_3 = \{h(0, 0), t(a, b, r), p(0, 0, b)\}$

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Which of the following interpretations are models of P_2 ?

- $I_1 = \emptyset$ **no**
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Which of the above interpretations are models of P_1 ?

Note:

Proposition

For every positive logic program P , $HB(P)$ is a model of P .

Minimal Model Semantics

- A logic program has multiple models in general.
- Select one of these models as the canonical model.
- Commonly accepted: truth of an atom in model I should be “founded” by clauses.

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Given

$$P_3 = \{a \leftarrow b. \quad b \leftarrow c. \quad c\},$$

truth of a in the model $I = \{a, b, c\}$ is “founded.”

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truth of a in the model $I = \{a, b, c\}$ is “founded.”

Given

$$P_4 = \{a \leftarrow b. \quad b \leftarrow a. \quad c\},$$

truth of a in the model $I = \{a, b, c\}$ is not founded.

Minimal Model Semantics (cont'd)

Semantics: Prefer models with true-part as small as possible.

Definition

A model I of P is *minimal*, if there exists no model J of P such that $J \subset I$.

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Theorem

Every logic program P has a single minimal model (called the least model), denoted $LM(P)$.

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Every logic program P has a single minimal model (called the least model), denoted $LM(P)$.

This is entailed by the following property:

Proposition (Intersection closure)

If I and J are models of P , then also $I \cap J$ is a model of P .

Example

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- For $P_2 = \{ p(f(X), Y, Z) \leftarrow p(X, Y, Z'), h(X, Y), t(Z, Z', r). \quad h(0, 0) \}$, we have $LM(P_2) = \{h(0, 0)\}$.

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- For $P_2 = \{ p(f(X), Y, Z) \leftarrow p(X, Y, Z'), h(X, Y), t(Z, Z', r). \quad h(0, 0) \}$, we have $LM(P_2) = \{h(0, 0)\}$.
- For P_1 above, we have

$$LM(P_1) = \{h(0, 0), t(a, b, r), p(0, 0, b), p(f(0), 0, a), h(f(0), f(0))\}.$$

Computation

The minimal model can be computed via fixpoint iteration.

Definition (T_P Operator)

Let $T_P: 2^{HB(P)} \rightarrow 2^{HB(P)}$ be defined as

$$T_P(I) = \left\{ a \mid \begin{array}{l} \text{there exists some } a \leftarrow b_1, \dots, b_m \\ \text{in } \textit{grnd}(P) \text{ such that } \{b_1, \dots, b_m\} \subseteq I \end{array} \right\} .$$

We let denote $T_P^0 = \emptyset, \quad T_P^{i+1} = T_P(T_P^i), \quad i \geq 0.$

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Fundamental result:

Theorem

T_P has a least fixpoint, $\textit{lfp}(T_P)$, and the sequence $\langle T_P^i \rangle, \quad i \geq 0,$ converges to $\textit{lfp}(T_P)$.

Proof: Use the fixpoint theorems of Knaster-Tarski and Kleene.

Example

■ For $P_3 = \{ a \leftarrow b. \quad b \leftarrow c. \quad c \}$, we have

$$T_{P_3}^0 = \{\},$$

Example

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Example (cont'd)

- For program P_1 above, we have

$$T_{P_1}^0 = \emptyset,$$

$$T_{P_1}^1 = \{h(0, 0), t(a, b, r), p(0, 0, b)\},$$

$$T_{P_1}^2 = \{h(0, 0), t(a, b, r), p(0, 0, b), p(f(0), 0, b), h(f(0), f(0))\},$$

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Hence

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$$lfp(T_{P_1}) = \{h(0, 0), t(a, b, r), p(0, 0, b), p(f(0), 0, b), h(f(0), f(0))\}.$$

- For program $P = \{p(0). \quad p(f(X)) \leftarrow p(X)\}$, we have

$$T_P^0 = \emptyset, \quad T_P^1 = \{p(0)\}, \dots, T_P^i = \{p(0), \dots, p(f^{i-1}(0))\}, \quad i \geq 0;$$

hence $lfp(T_P) = \{p(f^i(0)) \mid i \geq 0\}$ is infinite.

Negation in Logic Programs

Why negation?

- Natural linguistic concept
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Definition

A *normal logic program* is a set of rules of the form

$$a \leftarrow b_1, \dots, b_m, \text{not } c_1, \dots, \text{not } c_n \quad (n, m \geq 0) \quad (2)$$

where a and all b_i, c_j are atoms in a first-order language L .

not is called “negation as failure”, “default negation”, or “weak negation”

Things get more complex!

Programs with Negation

Prolog: “*not* $\langle X \rangle$ ” means “Negation as Failure (to prove to $\langle X \rangle$)”

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Query:

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Answer:

$X = \text{dilbert} .$

Example (cont'd)

Modifying the last rule of P_5 , we get P_6 :

man(dilbert).

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Problem: not a single intuitive model!

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Result in Prolog ????

Problem: not a single intuitive model!

Two intuitive Herbrand models:

$$\begin{aligned} M_1 &= \{\text{man}(\text{dilbert}), \text{single}(\text{dilbert})\}, \text{ and} \\ M_2 &= \{\text{man}(\text{dilbert}), \text{husband}(\text{dilbert})\} . \end{aligned}$$

Which one to choose?

Semantics of Logic Programs With Negation

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■ Agreement for so-called “stratified programs”

Different selection principles for non-stratified programs

Stratified Negation

Intuition: To evaluate a rule $r: a \leftarrow \dots, \text{not } p(\vec{t}), \dots$, the value of $p(\vec{t})$ should be known.

- 1 Evaluate first $p(\vec{t})$.
- 2 if $p(\vec{t})$ is $\begin{cases} \text{false,} & \text{then } \text{not } p(\vec{t}) \text{ is true,} \\ \text{true,} & \text{then } \text{not } p(\vec{t}) \text{ is false and } r \text{ is not applicable.} \end{cases}$

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Example

$$P = \{ \text{boring}(\text{chess}) \leftarrow \text{not interesting}(\text{chess}) \}$$

- $\text{interesting}(\text{chess})$ is false $\Rightarrow \text{not interesting}(\text{chess})$ is true.
- hence, r is applied and $\text{boring}(\text{chess})$ is true.
- This leads to the Herbrand model $H = \{\text{boring}(\text{chess})\}$ of P .

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Note: this introduces *procedurality* (violates declarativity)!

Dependency Graph

Restriction:

- The method works if there is no cyclic negation.
- Need a syntactic criterion to ensure this property.

Definition (Dependency graph)

The *dependency graph* of a set P of rules, is a directed graph $dep(P) = \langle N, E \rangle$ where

- $N = \{ \text{predicate } p \mid p \text{ occurs in } P, p \text{ is not a built-in} \} (=:\text{pred}(S))$
- E contains $p \rightarrow q$, iff P contains some rule $a \leftarrow \dots, \ell, \dots$ with $a = p(\dots)$ and $\ell = q(\dots)$ or $\ell = \text{not } q(\dots)$

Label $p \rightarrow q$ with “*”, if $\ell = \text{not } q(\dots)$

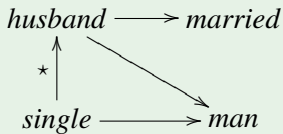
Example (Program P_7)

$man(dilbert).$

$husband(X) \leftarrow man(X), married(X).$

$single(X) \leftarrow man(X), not\ husband(X).$

$dep(P_7):$



Stratification

Definition (Stratification)

A *stratification* of a set P of rules is a partitioning

$$\Sigma = \{S_1, \dots, S_n\}$$

of $\text{pred}(P)$ into n nonempty, pairwise disjoint sets such that

- (a) if $p \in S_i$, $q \in S_j$, and $p \rightarrow q$ is in $\text{dep}(P)$, then $i \geq j$; and
- (b) if $p \in S_i$, $q \in S_j$, and $p \rightarrow^* q$ is in $\text{dep}(P)$ then $i > j$.

The sets S_1, \dots, S_n are the *strata* of P w.r.t. Σ .

P is *stratified*, if it has some stratification Σ .

- Informally, Σ specifies an *order of evaluation* for the predicates in P
- The sequential evaluation of S_1, S_2, \dots, S_n can be done by computing a series of *iterative least models*.

Semantics

Definition (iterative least model)

Suppose P is a logic program with stratification $\Sigma = \{S_1, \dots, S_k\}$, $k \geq 1$. Then

- $P_{S_i} = \{a \leftarrow b_1, \dots, b_n \in P \mid a = p(\dots), p \in S_i\}$, and
- $HB^*(P_{S_i}) = \bigcup_{j \leq i} \{p(\mathbf{t}) \in HB(P) \mid p \in S_j\}$.

The *iterative least models* $M_i \subseteq HB(P)$, $1 \leq i \leq k$, are such that

- (i) M_1 is the least model of P_{S_1} ;
- (ii) if $i > 1$, then M_i is the least subset M of $HB(P)$ such that
 - M is a model of P_{S_i} , and
 - $M \cap HB^*(P_{S_{i-1}}) = M_{i-1} \cap HB^*(P_{S_{i-1}})$.

The *iterative least model* of P is $M_{P,\Sigma} = M_k$.

Example (P_7 cont'd)

man(dilbert).

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Stratification: $\Sigma = \{S_1 = \{man, married\}, S_2 = \{husband\}, S_3 = \{single\}\}$

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$$HB^*(P_{S_1}) = \{man(dilbert), married(dilbert)\}.$$

Then, $M_2 = \{man(dilbert)\}$ is a model of P_{S_2} and

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■ $P_{S_3} = \{single(X) \leftarrow man(X), not\ husband(X)\}$.

Thus $M_3 = \{single(dilbert)\} \cup M_2$ is the least model of P_{S_3} such that

$M_3 \cap HB^*(P_{S_2}) = M_2 \cap HB^*(P_{S_2})$.

Stratification Theorem

Note: stratifications are not unique.

Example (P_7 cont'd)

Other stratification: $\Sigma' = \{S'_1 = \{man, married, husband\}, S'_2 = \{single\}\}$.

Evaluation with Σ' yields same result!

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This is not accidental:

Theorem (Apt *et al.* [1988])

Let P be a stratified program. Then for every stratifications Σ and Σ' of P , it holds that $M_{P,\Sigma} = M_{P,\Sigma'}$.

Hence, we can simplify $M_{P,\Sigma}$ to $M_P = M_{P,\Sigma}$ (for arbitrary Σ of choice)

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Corollary

*Stratified programs have a canonical model, also called **perfect model**.*

Stable model semantics

First, for variable-free (ground) programs P

- Treat “*not*” specially
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- Consider $M' = \{man(dilbert)\}$.

If as in M' , $man(dilbert)$ were true and $husband(dilbert)$ false, by r_1 also $single(dilbert)$ should be true. This is not coherent.

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If as in M' , $man(dilbert)$ were true and $husband(dilbert)$ false, by r_1 also $single(dilbert)$ should be true. This is not coherent.

- Consider $M'' = \{man(dilbert), single(dilbert), husband(dilbert)\}$.

The bodies of r_2 and R_2 are not true wrt M'' , hence there is no evidence for $single(dilbert)$ and $husband(dilbert)$ being true.

Stable Models

Definition (Gelfond-Lifschitz Reduct P^M 1988)

The *GL-reduct* (simply *reduct*) of a ground program P w.r.t. an interpretation M , denoted P^M , is the program obtained from P by

- 1 removing rules with *not* a in the body for each $a \in M$; and
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- 2 removing literals *not* a from all other rules.

Intuition:

- M makes an **assumption** about what is true and what is false.
- The reduct P^M incorporates this assumptions.
- As a “*not*”-free program, P^M derives positive facts, given by $LM(P^M)$.
- If this coincides with M , then the assumption of M is “stable”.

Stable Models (cont'd)

Definition (stable model)

An interpretation M of P is a *stable* model of P , if

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Stable Models (cont'd)

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Observe:

- $P^M = P$ for any “*not*”-free program P .
- Thus, for any positive program $LM(P)$ ($=LM(P^M)$) is its single stable model.

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Candidate interpretations:

- $M_1 = \{man(dilbert), single(dilbert)\},$
- $M_2 = \{man(dilbert), husband(dilbert)\},$
- $M_3 = \{man(dilbert), single(dilbert), husband(dilbert)\}$
- $M_4 = \{man(dilbert)\},$

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M_1 and M_2 are stable models.

Example (P_6 cont'd)

$man(dilbert).$ (f_1)

$single(dilbert) \leftarrow man(dilbert), not husband(dilbert).$ (r_1)

$husband(dilbert) \leftarrow man(dilbert), not single(dilbert).$ (r_2)

■ $M_1 = \{man(dilbert), single(dilbert)\}$:

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- $M_2 = \{man(dilbert), husband(dilbert)\}$: by symmetry of *husband* and *single*, also M_2 is stable.

Example (P_6 cont'd)

$$man(dilbert). \quad (f_1)$$
$$single(dilbert) \leftarrow man(dilbert), not husband(dilbert). \quad (r_1)$$
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■ $M_3 = \{man(dilbert), single(dilbert), husband(dilbert)\}$:

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$P_6^{M_3}$ is $man(dilbert).$

$$LM(P_6^{M_3}) = \{man(dilbert)\} \neq M_3.$$

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- $M_4 = \{man(dilbert)\}$:

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- $M_4 = \{man(dilbert)\}$:

$$P_6^{M_4} \text{ is } man(dilbert).$$

$$single(dilbert) \leftarrow man(dilbert).$$

$$husband(dilbert) \leftarrow man(dilbert).$$

$$LM(P_6^{M_4}) = \{man(dilbert), single(dilbert), husband(dilbert)\} \neq M_4.$$

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- Each normal logic program has some Herbrand model.
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Note:

- If p does not occur in P , then $P \cup \{p \leftarrow \text{not } p\}$ has no stable model.
- Adding $p \leftarrow \text{not } p$ to P “kills” all stable models of P !

Programs with Variables

- Consider, like in Prolog, only Herbrand interpretations.
- As for positive programs, view a program clause as a shorthand for all its ground instances.
- Recall: $grnd(P)$ is the grounding of program P .

Definition (stable model, general case)

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An interpretation M of P is a *stable model* of P , if M is a stable model of $grnd(P)$.

- Alternative way: Perform grounding in the GL-reduct, i.e., require $M = LM(P^M)$ where $P^M =_{def} grnd(P)^M$ for non-ground P .

Example (Variant P'_6 of P_6)

man(dilbert). (r_1)

woman(alice). (r_2)

single(X) ← man(X), not husband(X). (r_3)

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We have that, for instance,

$$\textit{grnd}(r_3) = \{ \textit{single}(\textit{dilbert}) \leftarrow \textit{man}(\textit{dilbert}), \textit{not husband}(\textit{dilbert}). \\ \textit{single}(\textit{alice}) \leftarrow \textit{man}(\textit{alice}), \textit{not husband}(\textit{alice}). \};$$

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Example (P'_6 cont'd)

The program $grnd(P'_6)$, and thus P'_6 , has the following stable models:

- $M_1 = \{man(dilbert), woman(alice), single(dilbert)\}$
- $M_2 = \{man(dilbert), woman(alice), husband(dilbert)\}$

Indeed,

- the rule instances of r_3 and r_4 for *dilbert* generate two possible scenarios;
- the rule instances of r_3 and r_4 for *alice* are inapplicable.

Properties of Stable Models

- Stable model semantics has a strong theoretical basis, many properties are known.
- We consider here some elementary ones.
- See e.g.
 - [Lifschitz, 2008]
 - [Ferraris and Lifschitz, 2005]
 - [Gelfond, 2008]

for other insights, alternative definitions and properties of stable models.

Relationship to Classical Models

How do stable models of P relate to classical models of P ?

Definition (classical model of normal logic program)

An interpretation I is a *model* of

- a ground clause $C : a \leftarrow b_1, \dots, b_m, \text{not } c_1, \dots, \text{not } c_n$, if either
 $\{b_1, \dots, b_m\} \not\subseteq I$ or $\{a, c_1, \dots, c_n\} \cap I \neq \emptyset$ $(I \models C)$;
- a clause C , if $I \models C'$ for every $C' \in \text{grnd}(C)$ $(I \models C)$;
- a set P of rules, if $I \models C$ for every clause C in P $(I \models P)$.

This complies with Herbrand models satisfying the clause

$$a \vee \text{not } b_1 \vee \dots \vee \text{not } b_m \vee c_1 \vee \dots \vee c_n,$$

where *not* is interpreted as classical negation (“ \neg ”).

Relationship to Classical Models (cont'd)

The following holds:

Theorem

- 1 *Every stable model M of P is a model of P .*
- 2 *A stable model M does not contain any model M' of P properly ($M' \not\subseteq M$), i.e., is a minimal model of P (w.r.t. \subseteq).*

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Corollary

Stable models are incomparable w.r.t. \subseteq , i.e., if M_1 and M_2 are different stable models of P , then $M_1 \not\subseteq M_2$ and $M_2 \not\subseteq M_1$.

Thus, stable models adhere to minimality of positive information.

Supportedness

- Note: each atom a in a stable model M must be derived from some rule of P .
- Extend the immediate consequence operator T_P to *not*.

Definition (T_P for normal P)

Given a normal program P and an interpretation I , let

$$T_P(I) = \left\{ a \mid \begin{array}{l} \text{there is some } r = a \leftarrow b_1, \dots, b_m, \text{not } c_1, \dots, \text{not } c_n \in \text{grnd}(P) \\ \text{such that } \{b_1, \dots, b_m\} \subseteq I, \{c_1, \dots, c_n\} \cap I = \emptyset \end{array} \right\}.$$

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- Problem with M_2 : Self-supportedness of b (via c)

Unfounded sets

Stable models amount to supported models with no (cyclic) self-support

Definition (cf. [Van Gelder *et al.*, 1991],[Leone *et al.*, 1997])

A set $U \subseteq HB_P$ is an *unfounded set* of P relative to interpretation I , if for every $a \in U$ and $r : a \leftarrow b_1, \dots, b_m, \text{not } c_1, \dots, \text{not } c_n$ in $\text{grnd}(P)$, either

- 1 for some $i \in \{1, \dots, m\}$, either $b_i \notin I$ or $b_i \in U$, or
- 2 for some $j \in \{1, \dots, n\}$, $c_j \in I$.

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- Every P has a *greatest unfounded set* relative to I , denoted $U_P(I)$.
- Intuitively, if I is compatible with P , all atoms in $U_P(I)$ can be safely switched to false while maintaining compatibility.

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Definition (unfounded-freeness)

I is called *unfounded-free*, if $I \cap U = \emptyset$ for each unfounded set U of P rel. to I .

Note: I is *unfounded-free* iff $I \cap U_P(I) = \{\}$.

Unfounded sets (cont'd)

Theorem (implicit in [Leone *et al.*, 1997])

Given a program P , a model M of P is stable iff M is unfounded-free.

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Example (P_s cont'd)

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$$c \leftarrow b.$$

■ $M_2 = \{b, c\}$: $U_{P_s}(M_2) = \{b, c\}$, thus $M_2 \cap U_{P_s}(M_2) \neq \emptyset$.

■ $M_1 = \{a\}$: $U_{P_s}(M_1) = \emptyset$, thus $M_1 \cap U_{P_s}(M_1) = \emptyset$.

- Unfounded-freeness is exploited for computing stable models (DLV)
- It corresponds to loop formulas [Lin and Zhao, 2002], [Lee, 2005].

Stratified Programs

Stable model semantics gracefully generalizes stratified semantics:

Theorem

If a program P is stratified, then P has a single stable model, which coincides with the perfect (i.e., the iterative least) model of P .

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Notes:

- A stratified P may have several minimal models; only one is stable

E.g., $P = \{boring(chess) \leftarrow not\ interesting(chess)\}$

has two minimal models:

$M_1 = \{boring(chess)\}$ and $M_2 = \{interesting(chess)\}$.

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The perfect model is $M_P = M_1$.

- Stratified programs can only express deterministic scenarios, no “alternatives” are possible!

Non-Cumulativity

- In classical logic, adding consequences of a theory T to T preserves its semantics.
- This property is known as cumulativity (or lemma support).
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Proposition

Suppose P and atom a fulfill $M \models a$, for each stable model M of P . Then P and $P \cup \{a\}$ need not have the same stable models (even if P is consistent).

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Example

$$b \leftarrow \text{not } c. \quad c \leftarrow \text{not } b.$$
$$a \leftarrow b. \quad a \leftarrow \text{not } a.$$

P has the stable model $M = \{a, b\}$; $P \cup \{a\}$ has in addition $N = \{a, c\}$.

Note: the property holds for stratified programs.

Computational Properties

How difficult is it to compute some stable model?

Decision problem CONS:

Given a program P , does P have some stable model?

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Theorem

For normal logic programs P , problem CONS is

- *NP-complete in the propositional and ground case;*
- *NEXPTIME-complete in the datalog (function-free) case;*
- *Σ_1^1 -complete in the general first-order case.*

Recall: NP (NEXPTIME) = class of problems solvable in polynomial (exponential) time on a non-deterministic Turing machine.

Σ_1^1 is a class in the Analytic Hierarchy

Computational Properties (cont'd)

Lower complexity holds for fragments:

- For positive and stratified propositional programs, CONS is polynomial (in fact, trivial).
 - Still solvable in linear time if constraints and strong negation are allowed (P-complete).
 - For datalog programs, complexity increases to EXPTIME.
- For programs with function symbols, several decidable program classes are known (up to 3-EXPTIME).

More on basic complexity: [Dantsin *et al.*, 2001].

Extensions

- Many extensions exist, partly motivated by applications
- Some are syntactic sugar, other strictly add expressiveness
- Incomplete list:
 - constraints
 - strong negation
 - disjunction
 - nested expressions
 - cardinality constraints (Smodels)
 - optimization: weight constraints, *minimize* (Smodels); weak constraints (DLV)
 - aggregates (Smodels, DLV)
 - templates (for macros), external functions (DLVHEX)
 - Frame Logic syntax (for Semantic Web)
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Constraints

■ Adding

$$p \leftarrow q_1, \dots, q_m, \text{not } r_1, \dots, \text{not } r_n, \text{not } p.$$

to P “kills” all stable models of P that

- contain q_1, \dots, q_m , and
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- ## ■ This is convenient to eliminate scenarios which does not satisfy integrity constraints.

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- Short:

Constraint

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Example (Dilbert P_6 cont'd)

$man(dilbert).$ (f_1)

$single(dilbert) \leftarrow man(dilbert), not husband(dilbert).$ (r_1)

$husband(dilbert) \leftarrow man(dilbert), not single(dilbert).$ (r_2)

$\leftarrow husband(X), not wedding_ring(X).$ (c_1)

- The constraint c_1 eliminates models in which there is no evidence for a husband having a wedding ring.
- Single stable model: $M_1 = \{man(dilbert), single(dilbert)\}$

Strong Negation

- Weak negation “*not a*” means “*a* can not be proved (derived) using rules,” and that *a* is false by default (believed to be false).
- This is different from *knowing* (provably) that *a* is false; this is expressed by $\neg a$ (sometimes $\neg a$).
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Example (due to John McCarthy)

Consider an agent *A* with the following task:

- “At a railroad crossing, cross the rails if no train approaches.”

We may encode this scenario using one of the following two rules:

$$walk \leftarrow at(A, L), crossing(L), not\ train_approaches(L). \quad (r_1)$$

$$walk \leftarrow at(A, L), crossing(L), \neg train_approaches(L). \quad (r_2)$$

Extended Logic Programs

Definition

An *extended logic program (ELP)* is a finite set of rules

$$a \leftarrow b_1, \dots, b_m, \text{not } c_1, \dots, \text{not } c_n \quad (n, m \geq 0) \quad (3)$$

where a, b_i, c_j are atoms or strongly negated atoms in a f.o. language L .

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The semantics of ELPs can be defined by program transformation:

- view literals “ $\neg p(\vec{X})$ ” as atoms with fresh predicate symbols “ $\neg p$ ”;
- add clauses $\text{falsity} \leftarrow \text{not falsity}, p(\vec{X}), \neg p(\vec{X})$
to P ($p(\vec{X})$ and $\neg p(\vec{X})$ are not simultaneously true); and
- select the stable models of the resulting program (called *answer sets* of P).

Extended Logic Programs

Definition

An *extended logic program (ELP)* is a finite set of rules

$$a \leftarrow b_1, \dots, b_m, \text{not } c_1, \dots, \text{not } c_n \quad (n, m \geq 0) \quad (3)$$

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Answer sets M : *three-valued view*

Atom a may be true ($a \in M$), false ($\neg a \in M$), or *unknown* ($a, \neg a \notin M$).

Strong negation (combined with weak negation) is e.g. helpful to express *default rules*.

Example (French speaking)

$$\text{french}(\text{luc}). \quad (f_1)$$

$$\text{speaks}(X, \text{french}) \leftarrow \text{french}(X), \text{not } \neg \text{speaks}(X, \text{french}). \quad (r_1)$$

$$\neg \text{speaks}(X, \text{french}) \leftarrow \text{thumb}(X). \quad (r_2)$$

- r_1 expresses that by default, French can speak French.
- Single answer set $M = \{\text{french}(\text{luc}), \text{speaks}(\text{luc}, \text{french})\}$.

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Note:

- ELPs are closely related to Default Logic [Reiter, 1980]
- The answer sets of P correspond 1-1 to the extensions of the default theory $T = (\emptyset, \{d(C) \mid C \in P\})$ ($d(C)$ casts C into a default rule).

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Example

- $ok(C) \vee \neg ok(C) \leftarrow component(C).$

Minimality

- Semantics: disjunction is *minimal* (different from classical logic):

$$a \vee b \vee c.$$

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Models $\{a\}$ and $\{a, b\}$, but only $\{a\}$ is minimal.

- but minimality is *not necessarily exclusive*:

$$a \vee b. \quad b \vee c. \quad a \vee c.$$

Minimal models: $\{a, b\}$, $\{a, c\}$, and $\{b, c\}$.

Disjunction vs. Unstratified Negation

Reconsider the Dilbert Program P_6 :

$$\begin{aligned} & \textit{man}(\textit{dilbert}). \\ \textit{single}(X) & \leftarrow \textit{man}(X), \textit{not husband}(X). \\ \textit{husband}(X) & \leftarrow \textit{man}(X), \textit{not single}(X). \end{aligned}$$

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is under stable semantics equivalent to the program P_{dd} :

$$\begin{aligned} & \textit{man}(\textit{dilbert}). \\ \textit{single}(X) \vee \textit{husband}(X) & \leftarrow \textit{man}(X). \end{aligned}$$

The use of disjunction is more intuitive!

Extended Logic Programs with Disjunctions

Definition

A *extended disjunctive logic program* (EDLP) is a finite set of rules

$$a_1 \vee \dots \vee a_k \leftarrow b_1, \dots, b_m, \text{not } c_1, \dots, \text{not } c_n \quad (k, m, n \geq 0) \quad (4)$$

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where all a_i, b_j, c_l are atoms or strongly negated atoms in f.o. language L .

Semantics:

- Answer sets of P are defined similarly as for an ELP
- Differences:
 - I is a model of ground (4), if either $\{b_1, \dots, b_m\} \not\subseteq I$ or $\{a_1, \dots, a_k, c_1, \dots, c_n\} \cap I \neq \emptyset$
 - “ M is **the least** model of P^M ” \leadsto “ M is a **minimal** model of P^M ” (P^M may have multiple minimal models).

Example (Disjunctive Dilbert P_{dd} , cont'd)

$man(dilbert).$

$single(X) \vee husband(X) \leftarrow man(X).$

As P_{dd} is “not”-free, $grnd(P_{dd})^M = grnd(P_{dd})$ for every M .

Answer sets:

■ $M_1 = \{man(dilbert), single(dilbert)\}$, and

■ $M_2 = \{man(dilbert), husband(dilbert)\}.$

Some Properties of EDLPs

- Every answer set of an EDLP P is a minimal model of P (models analogous as for ELPs)
- Different answer sets of an EDLP P are incomparable
- An EDLP may have no, a single or multiple answer sets
- For EDLPs without strong negation, answer sets are **models** that are unfounded-free [Leone *et al.*, 1997]
- Deciding whether a propositional EDLP P has some answer set is Σ_2^P -complete. $(\Sigma_2^P = \text{NP}^{\text{NP}})$

Disjunction adds higher problem solving capacity, it is not just syntactic sugar!

- **But:** EDLPs **can not** be regarded as a fragment of Reiter's Default Logic.

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