

Answer Set Solving in Practice

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Rough Roadmap

- 1 Introduction
- 2 Language
- 3 Modeling
- 4 Grounding
- 5 Foundations
- 6 Solving
- 7 Systems
- 8 Applications

Resources

■ Course material

- <http://www.cs.uni-potsdam.de/wv/lehre>
- <http://moodle.cs.uni-potsdam.de>
- <http://potassco.sourceforge.net/teaching.html>

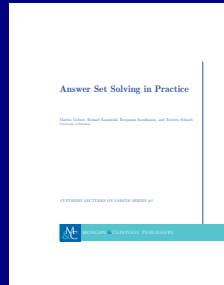
■ Systems

- **clasp** <http://potassco.sourceforge.net>
- **dlv** <http://www.dlvsystem.com>
- **smodels** <http://www.tcs.hut.fi/Software/smodels>
- **gringo** <http://potassco.sourceforge.net>
- **lparse** <http://www.tcs.hut.fi/Software/smodels>
- **clingo** <http://potassco.sourceforge.net>
- **iclingo** <http://potassco.sourceforge.net>
- **oclingo** <http://potassco.sourceforge.net>

- **asparagus** <http://asparagus.cs.uni-potsdam.de>

The Potassco Book

1. Motivation
2. Introduction
3. Basic modeling
4. Grounding
5. Characterizations
6. Solving
7. Systems
8. Advanced modeling
9. Conclusions



Resources

- <http://potassco.sourceforge.net/book.html>
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Literature

Books [4], [29], [53]

Surveys [50], [2], [39], [21], [11]

Articles [41], [42], [6], [61], [54], [49], [40], etc.

Language Extensions: Overview

- 1 Two kinds of negation
- 2 Disjunctive logic programs
- 3 Propositional theories

Outline

- 1 Two kinds of negation
- 2 Disjunctive logic programs
- 3 Propositional theories

Motivation

■ Classical versus default negation

■ Symbol \neg and \sim

■ Idea

- $\neg a \approx \neg a \in X$

- $\sim a \approx a \notin X$

■ Example

- $cross \leftarrow \neg train$

- $cross \leftarrow \sim train$

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Classical negation

- We consider logic programs in negation normal form
 - That is, classical negation is applied to atoms only
- Given an alphabet \mathcal{A} of atoms, let $\overline{\mathcal{A}} = \{\neg a \mid a \in \mathcal{A}\}$ such that $\mathcal{A} \cap \overline{\mathcal{A}} = \emptyset$
- Given a program P over \mathcal{A} , classical negation is encoded by adding

$$P^\neg = \{a \leftarrow b, \neg b \mid a \in (\mathcal{A} \cup \overline{\mathcal{A}}), b \in \mathcal{A}\}$$

- A set X of atoms is a stable model of a program P over $\mathcal{A} \cup \overline{\mathcal{A}}$, if X is a stable model of $P \cup P^\neg$

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An example

■ The program

$$P = \{a \leftarrow \sim b, b \leftarrow \sim a\} \cup \{c \leftarrow b, \neg c \leftarrow b\}$$

induces

$$P^\neg = \left\{ \begin{array}{lll} a \leftarrow a, \neg a & a \leftarrow b, \neg b & a \leftarrow c, \neg c \\ \neg a \leftarrow a, \neg a & \neg a \leftarrow b, \neg b & \neg a \leftarrow c, \neg c \\ b \leftarrow a, \neg a & b \leftarrow b, \neg b & b \leftarrow c, \neg c \\ \neg b \leftarrow a, \neg a & \neg b \leftarrow b, \neg b & \neg b \leftarrow c, \neg c \\ c \leftarrow a, \neg a & c \leftarrow b, \neg b & c \leftarrow c, \neg c \\ \neg c \leftarrow a, \neg a & \neg c \leftarrow b, \neg b & \neg c \leftarrow c, \neg c \end{array} \right\}$$

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Properties

- The only inconsistent stable “model” is $X = \mathcal{A} \cup \overline{\mathcal{A}}$
- Note Strictly speaking, an inconsistent set like $\mathcal{A} \cup \overline{\mathcal{A}}$ is not a model
- For a logic program P over $\mathcal{A} \cup \overline{\mathcal{A}}$, exactly one of the following two cases applies:
 - 1 All stable models of P are consistent or
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Train spotting

- $P_1 = \{cross \leftarrow \sim train\}$
 - stable model: $\{cross\}$
- $P_2 = \{cross \leftarrow \neg train\}$
 - stable model: \emptyset
- $P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow\}$
 - stable model: $\{cross, \neg train\}$
- $P_4 = \{cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow\}$
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Default negation in rule heads

- We consider logic programs with default negation in rule heads
- Given an alphabet \mathcal{A} of atoms, let $\tilde{\mathcal{A}} = \{\tilde{a} \mid a \in \mathcal{A}\}$ such that $\mathcal{A} \cap \tilde{\mathcal{A}} = \emptyset$

- Given a program P over \mathcal{A} , consider the program

$$\begin{aligned} \tilde{P} = & \{r \in P \mid \text{head}(r) \neq \sim a\} \\ & \cup \{\leftarrow \text{body}(r) \cup \{\sim \tilde{a}\} \mid r \in P \text{ and } \text{head}(r) = \sim a\} \\ & \cup \{\tilde{a} \leftarrow \sim a \mid r \in P \text{ and } \text{head}(r) = \sim a\} \end{aligned}$$

- A set X of atoms is a stable model of a program P (with default negation in rule heads) over \mathcal{A} ,
if $X = Y \cap \mathcal{A}$ for some stable model Y of \tilde{P} over $\mathcal{A} \cup \tilde{\mathcal{A}}$

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Disjunctive logic programs

- A **disjunctive rule**, r , is of the form

$$a_1 ; \dots ; a_m \leftarrow a_{m+1}, \dots, a_n, \sim a_{n+1}, \dots, \sim a_o$$

where $0 \leq m \leq n \leq o$ and each a_i is an atom for $0 \leq i \leq o$

- A **disjunctive logic program** is a finite set of disjunctive rules

- Notation

$$head(r) = \{a_1, \dots, a_m\}$$


$$body(r) = \{a_{m+1}, \dots, a_n, \sim a_{n+1}, \dots, \sim a_o\}$$

$$body(r)^+ = \{a_{m+1}, \dots, a_n\}$$

$$body(r)^- = \{a_{n+1}, \dots, a_o\}$$

$$atom(P) = \bigcup_{r \in P} (head(r) \cup body(r)^+ \cup body(r)^-)$$

$$body(P) = \{body(r) \mid r \in P\}$$

- A program is called **positive** if $body(r)^- = \emptyset$ for all its rules  Potassco

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Stable models

■ Positive programs

- A set X of atoms is **closed under** a positive program P iff for any $r \in P$, $head(r) \cap X \neq \emptyset$ whenever $body(r)^+ \subseteq X$
 - X corresponds to a model of P (seen as a formula)
- The set of all \subseteq -minimal sets of atoms being closed under a positive program P is denoted by $\min_{\subseteq}(P)$
 - $\min_{\subseteq}(P)$ corresponds to the \subseteq -minimal models of P (ditto)

■ Disjunctive programs

The reduct, P^X , of a disjunctive program P relative to a set X of atoms is defined by

$$P^X = \{head(r) \leftarrow body(r)^+ \mid r \in P \text{ and } body(r)^- \cap X = \emptyset\}$$

A set X of atoms is a stable model of a disjunctive program P , if $X \in \min_{\subseteq}(P^X)$

Stable models

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A “positive” example

$$P = \left\{ \begin{array}{l} a \leftarrow \\ b; c \leftarrow a \end{array} \right\}$$

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Graph coloring (reloaded)

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node(1..6).
```

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More Examples

- $P_1 = \{a ; b ; c \leftarrow\}$

- stable models $\{a\}$, $\{b\}$, and $\{c\}$

- $P_2 = \{a ; b ; c \leftarrow , \leftarrow a\}$

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- A disjunctive logic program may have zero, one, or multiple stable models
- If X is a stable model of a disjunctive logic program P , then X is a model of P (seen as a formula)
- If X and Y are stable models of a disjunctive logic program P , then $X \not\subseteq Y$
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$$P = \left\{ \begin{array}{l} a(1, 2) \quad \leftarrow \\ b(X) ; c(Y) \quad \leftarrow \quad a(X, Y), \sim c(Y) \end{array} \right\}$$

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- Consider disjunctive rules of the form

$$a_1 ; \dots ; a_m ; \sim a_{m+1} ; \dots ; \sim a_n \leftarrow a_{n+1}, \dots, a_o, \sim a_{o+1}, \dots, \sim a_p$$

where $0 \leq m \leq n \leq o \leq p$ and each a_i is an atom for $0 \leq i \leq p$

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Outline

- 1 Two kinds of negation
- 2 Disjunctive logic programs
- 3 Propositional theories

Propositional theories

- Formulas are formed from
 - atoms in \mathcal{A}
 - \perp

using

- conjunction (\wedge)
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 - implication (\rightarrow)
- Notation

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- The satisfaction relation $X \models \phi$ between a set X of atoms and a (set of) formula(s) ϕ is defined as in propositional logic
- The reduct, ϕ^X , of a formula ϕ relative to a set X of atoms is defined recursively as follows:

$$\phi^X = \perp \quad \text{if } X \not\models \phi$$

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$$\phi^X = (\psi^X \circ H^X) \quad \text{if } X \models \phi \text{ and } \phi = (\psi \circ H) \text{ for } \circ \in \{\wedge, \vee, \rightarrow\}$$

$$\text{If } \phi = \sim\psi = (\psi \rightarrow \perp),$$

$$\text{then } \phi^X = (\perp \rightarrow \perp) = \top, \text{ if } X \not\models \psi, \text{ and } \phi^X = \perp, \text{ otherwise}$$

The reduct, Φ^X , of a propositional theory Φ relative to a set X of atoms is defined as $\Phi^X = \{\phi^X \mid \phi \in \Phi\}$

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$$\phi^X = (\psi^X \circ H^X) \quad \text{if } X \models \phi \text{ and } \phi = (\psi \circ H) \text{ for } \circ \in \{\wedge, \vee, \rightarrow\}$$

$$\text{If } \phi = \sim\psi = (\psi \rightarrow \perp),$$

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Stable models

- A set X of atoms satisfies a propositional theory Φ , written $X \models \Phi$, if $X \models \phi$ for each $\phi \in \Phi$
- The set of all \subseteq -minimal sets of atoms satisfying a propositional theory Φ is denoted by $\min_{\subseteq}(\Phi)$
- A set X of atoms is a stable model of a propositional theory Φ , if $X \in \min_{\subseteq}(\Phi^X)$
- If X is a stable model of Φ , then
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Two examples

- $\Phi_1 = \{p \vee (p \rightarrow (q \wedge r))\}$

- For $X = \{p, q, r\}$, we get

$$\Phi_1^{\{p,q,r\}} = \{p \vee (p \rightarrow (q \wedge r))\} \text{ and } \min_{\subseteq}(\Phi_1^{\{p,q,r\}}) = \{\emptyset\}$$

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The translation of a logic program P is $\tau[P] = \{\tau[r] \mid r \in P\}$

Given a logic program P and a set X of atoms,
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- The normal logic program $P = \{p \leftarrow \sim q, q \leftarrow \sim p\}$ corresponds to $\tau[P] = \{\sim q \rightarrow p, \sim p \rightarrow q\}$
 - stable models: $\{p\}$ and $\{q\}$
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- The normal logic program $P = \{p \leftarrow \sim q, q \leftarrow \sim p\}$ corresponds to $\tau[P] = \{\sim q \rightarrow p, \sim p \rightarrow q\}$
 - stable models: $\{p\}$ and $\{q\}$
- The disjunctive logic program $P = \{p ; q \leftarrow\}$ corresponds to $\tau[P] = \{\top \rightarrow p \vee q\}$
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- The nested logic program $P = \{p \leftarrow \sim\sim p\}$ corresponds to $\tau[P] = \{\sim\sim p \rightarrow p\}$
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