

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>

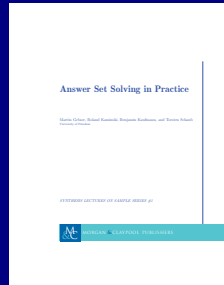
■ Systems

- **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: Overview

- 1 Motivation
- 2 Core language
 - Integrity constraint
 - Choice rule
 - Cardinality rule
 - Weight rule
- 3 Extended language
 - Conditional literal
 - Optimization statement
- 4 smodels format
- 5 ASP language standard

Outline

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Basic language extensions

- The expressiveness of a language can be enhanced by introducing new constructs
- To this end, we must address the following issues:
 - What is the **syntax** of the new language construct?
 - What is the **semantics** of the new language construct?
 - How to **implement** the new language construct?
- A way of providing semantics is to furnish a translation removing the new constructs, eg. classical negation
- This translation might also be used for implementing the language extension

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Integrity constraint

- Idea Eliminate unwanted solution candidates
- Syntax An **integrity constraint** is of the form

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

where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$

- Example `:- edge(3,7), color(3,red), color(7,red).`
- Embedding The above integrity constraint can be turned into the normal rule

$$x \leftarrow a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n, \sim x$$

where x is a new symbol, that is, $x \notin \mathcal{A}$.

- Another example $P = \{a \leftarrow \sim b, b \leftarrow \sim a\}$
versus $P' = P \cup \{\leftarrow a\}$ and $P'' = P \cup \{\leftarrow \sim a\}$

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Choice rule

- Idea Choices over subsets
- Syntax A **choice rule** 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 $1 \leq i \leq o$

- Informal meaning If the body is satisfied by the stable model at hand, then any subset of $\{a_1, \dots, a_m\}$ can be included in the stable model
- Example $\{ \text{buy}(\text{pizza}), \text{buy}(\text{wine}), \text{buy}(\text{corn}) \} \text{ :- } \text{at}(\text{grocery})$.
- Another Example $P = \{ \{a\} \leftarrow b, b \leftarrow \}$ has two stable models: $\{b\}$ and $\{a, b\}$

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Embedding in normal rules

- A choice rule of form

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

can be translated into $2m + 1$ normal rules

$$\begin{array}{l} a' \leftarrow a_{m+1}, \dots, a_n, \sim a_{n+1}, \dots, \sim a_o \\ a_1 \leftarrow a', \sim \overline{a_1} \quad \dots \quad a_m \leftarrow a', \sim \overline{a_m} \\ \overline{a_1} \leftarrow \sim a_1 \quad \dots \quad \overline{a_m} \leftarrow \sim a_m \end{array}$$

by introducing new atoms $a', \overline{a_1}, \dots, \overline{a_m}$.

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Cardinality rule

- Idea Control (lower) cardinality of subsets
- Syntax A **cardinality rule** is the form

$$a_0 \leftarrow l \{ a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n \}$$

where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$;
 l is a non-negative integer.

- Informal meaning The head atom belongs to the stable model, if at least l elements of the body are included in the stable model
- Note l acts as a lower bound on the body
- Example `pass(c42) :- 2 { pass(a1), pass(a2), pass(a3) }.`
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Embedding in normal rules

- Replace each cardinality rule

$$a_0 \leftarrow l \{ a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n \}$$

by $a_0 \leftarrow ctr(1, l)$

where atom $ctr(i, j)$ represents the fact that at least j of the literals having an equal or greater index than i , are in a stable model

- The definition of $ctr/2$ is given for $0 \leq k \leq l$ by the rules

$$\begin{aligned} ctr(i, k+1) &\leftarrow ctr(i+1, k), a_i \\ ctr(i, k) &\leftarrow ctr(i+1, k) \end{aligned} \quad \text{for } 1 \leq i \leq m$$

$$\begin{aligned} ctr(j, k+1) &\leftarrow ctr(j+1, k), \sim a_j \\ ctr(j, k) &\leftarrow ctr(j+1, k) \end{aligned} \quad \text{for } m+1 \leq j \leq n$$

$$ctr(n+1, 0) \leftarrow$$

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An example

- Program $\{a \leftarrow, c \leftarrow 1 \{a, b\}\}$ has the stable model $\{a, c\}$
- Translating the cardinality rule yields the rules

$$\begin{array}{l}
 a \leftarrow \\
 c \leftarrow ctr(1, 1) \\
 ctr(1, 2) \leftarrow ctr(2, 1), a \\
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 ctr(2, 2) \leftarrow ctr(3, 1), b \\
 ctr(2, 1) \leftarrow ctr(3, 1) \\
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having stable model $\{a, ctr(3, 0), ctr(2, 0), ctr(1, 0), ctr(1, 1)\}$

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having stable model $\{a, ctr(3, 0), ctr(2, 0), ctr(1, 0), ctr(1, 1), c\}$

... and vice versa

- A normal rule

$$a_0 \leftarrow a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n,$$

can be represented by the cardinality rule

$$a_0 \leftarrow n \{a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n\}$$

Cardinality rules with upper bounds

- A rule of the form

$$a_0 \leftarrow l \{ a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n \} u$$

where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$;
 l and u are non-negative integers

stands for

$$\begin{aligned} a_0 &\leftarrow b, \sim c \\ b &\leftarrow l \{ a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n \} \\ c &\leftarrow u+1 \{ a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n \} \end{aligned}$$

where b and c are new symbols

- The single constraint in the body of the above cardinality rule is referred to as a cardinality constraint

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Cardinality constraints

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where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$;
 l and u are non-negative integers

- Informal meaning A cardinality constraint is satisfied by a stable model X , if the number of its contained literals satisfied by X is between l and u (inclusive)
- In other words, if

$$l \leq |(\{a_1, \dots, a_m\} \cap X) \cup (\{a_{m+1}, \dots, a_n\} \setminus X)| \leq u$$

Cardinality constraints

- Syntax A **cardinality constraint** is of the form

$$l \{ a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n \} u$$

where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$;
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Cardinality constraints as heads

- A rule of the form

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

where $0 \leq m \leq n \leq o \leq p$ and each a_i is an atom for $1 \leq i \leq p$;
 l and u are non-negative integers

stands for

$$\begin{aligned} b &\leftarrow a_{n+1}, \dots, a_o, \sim a_{o+1}, \dots, \sim a_p \\ \{a_1, \dots, a_m\} &\leftarrow b \\ c &\leftarrow l \{a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n\} u \\ &\leftarrow b, \sim c \end{aligned}$$

where b and c are new symbols

- Example $1 \{ \text{color}(v42, \text{red}), \text{color}(v42, \text{green}), \text{color}(v42, \text{blue}) \} 1.$



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Full-fledged cardinality rules

- A rule of the form

$$l_0 S_0 u_0 \leftarrow l_1 S_1 u_1, \dots, l_n S_n u_n$$

where for $0 \leq i \leq n$ each $l_i S_i u_i$

stands for $0 \leq i \leq n$

$$a \leftarrow b_1, \dots, b_n, \sim c_1, \dots, \sim c_n$$

$$S_0^+ \leftarrow a$$

$$\leftarrow a, \sim b_0$$

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$$b_i \leftarrow l_i S_i$$

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Weight rule

- Syntax A **weight rule** is the form

$$a_0 \leftarrow l \{ a_1 = w_1, \dots, a_m = w_m, \sim a_{m+1} = w_{m+1}, \dots, \sim a_n = w_n \}$$

where $0 \leq m \leq n$ and each a_i is an atom;

l and w_i are integers for $1 \leq i \leq n$

- A weighted literal, $\ell_i = w_i$, associates each literal ℓ_i with a weight w_i
- Note A cardinality rule is a weight rule where $w_i = 1$ for $0 \leq i \leq n$

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l, u and w_i are integers for $1 \leq i \leq n$

- Meaning A weight constraint is satisfied by a stable model X , if

$$l \leq \left(\sum_{1 \leq i \leq m, a_i \in X} w_i + \sum_{m < i \leq n, a_i \notin X} w_i \right) \leq u$$

- Note (Cardinality and) weight constraints amount to constraints on (count and) sum aggregate functions
- Example 10 `[course(db)=6, course(ai)=6, course(project)=8, course(xml)=3]` 20

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Conditional literals (in *lparse* & *gringo* 3)

- Syntax A **conditional literal** is of the form

$$l : l_1 : \dots : l_n$$

where l and l_i are literals for $0 \leq i \leq n$

- Informal meaning A conditional literal can be regarded as the list of elements in the set $\{l \mid l_1, \dots, l_n\}$
- Note The expansion of conditional literals is context dependent
- Example Given 'p(1). p(2). p(3). q(2).'

$r(X):p(X):not\ q(X) :- r(X):p(X):not\ q(X), 1\ \{r(X):p(X):not\ q(X)\}.$

is instantiated to

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Optimization statement

- Idea Express cost functions subject to minimization and/or maximization
- Syntax A **minimize statement** is of the form

$$\textit{minimize}\{ \ell_1 = w_1 @ p_1, \dots, \ell_n = w_n @ p_n \}.$$

where each ℓ_i is a literal; and w_i and p_i are integers for $1 \leq i \leq n$

Priority levels, p_i , allow for representing lexicographically ordered minimization objectives

- Meaning A minimize statement is a directive that instructs the ASP solver to compute optimal stable models by minimizing a weighted sum of elements

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- A maximize statement of the form

$$\textit{maximize}\{ \ell_1 = w_1 @ p_1, \dots, \ell_n = w_n @ p_n \}$$

stands for $\textit{minimize}\{ \ell_1 = -w_1 @ p_1, \dots, \ell_n = -w_n @ p_n \}$

- Example When configuring a computer, we may want to maximize hard disk capacity, while minimizing price

```
#maximize[ hd(1)=250@1, hd(2)=500@1, hd(3)=750@1, hd(4)=1000@1 ].  
#minimize[ hd(1)=30@2, hd(2)=40@2, hd(3)=60@2, hd(4)=80@2 ].
```

The priority levels indicate that (minimizing) price is more important than (maximizing) capacity

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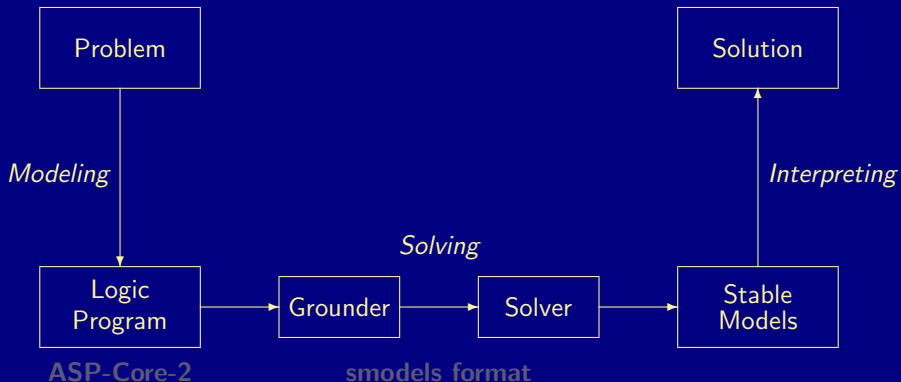
smodels format

- Logic programs in *smodels* format consist of
 - normal rules
 - choice rules
 - cardinality rules
 - weight rules
 - optimization statements
- Such a format is obtained by grounders *lparse* and *gringo*

Outline

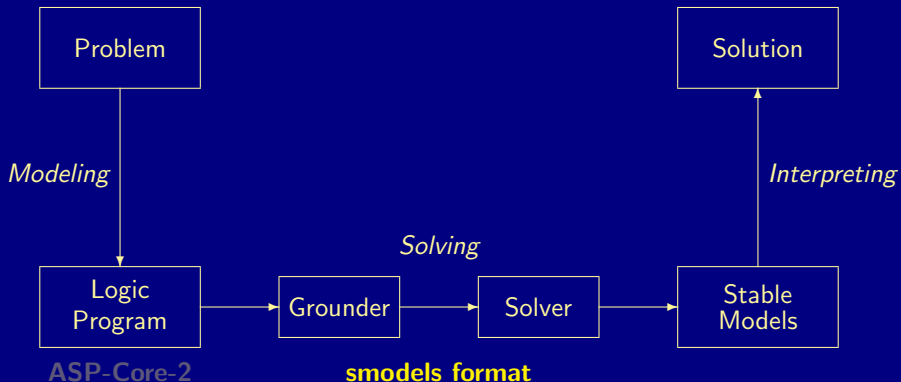
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ASP-Core-2



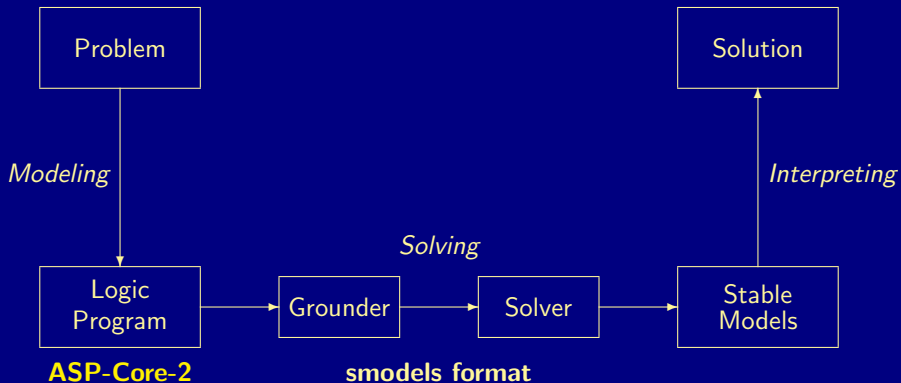
- *smodels format* is a machine-oriented standard for ground programs
- ASP-Core-2 is a user-oriented standard for (non-ground) programs, extending the input languages of *dlv* and *gringo* series 3

ASP-Core-2



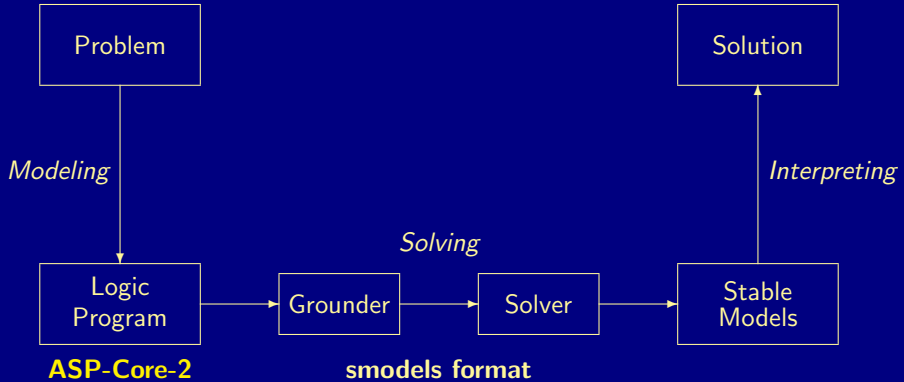
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ASP-Core-2



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ASP-Core-2



- smodels format is a machine-oriented standard for ground programs
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Aggregates

- Syntax **ASP-Core-2 aggregates** are of the form

$$t_1 \prec_1 \#A\{t_{1_1}, \dots, t_{m_1} : l_{1_1}, \dots, l_{n_1}\} \prec_2 t_2$$

where

- $\#A \in \{\#count, \#sum, \#max, \#min\}$
 - $\prec_1, \prec_2 \in \{<, \leq, =, \neq, >, \geq\}$
 - t_{1_1}, \dots, t_{m_1} and t_1, t_2 are terms
 - l_{1_1}, \dots, l_{n_1} are literals
- Example Weight constraint

10 [course(db)=6, course(ai)=6, course(project)=8, course(xml)=3] 20

is written as an ASP-Core-2 aggregate as

$$10 \leq \#sum\{6, db:course(db); 6, ai:course(ai); 8, project:course(project); 3, xml:course(xml)\} \leq 20$$

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$$t_1 \prec_1 \#A\{t_{1_1}, \dots, t_{m_1} : l_{1_1}, \dots, l_{n_1}; \dots; t_{1_k}, \dots, t_{m_k} : l_{1_k}, \dots, l_{n_k}\} \prec_2 t_2$$

where

- $\#A \in \{\#count, \#sum, \#max, \#min\}$
 - $\prec_1, \prec_2 \in \{<, \leq, =, \neq, >, \geq\}$
 - $t_{1_1}, \dots, t_{m_1}, \dots, t_{1_k}, \dots, t_{m_k}$, and t_1, t_2 are terms
 - $l_{1_1}, \dots, l_{n_1}, \dots, l_{1_k}, \dots, l_{n_k}$ are literals
- Example Weight constraint

10 [course(db)=6,course(ai)=6,course(project)=8,course(xml)=3] 20

is written as an ASP-Core-2 aggregate as

$$10 \leq \#sum\{6,db:course(db); 6,ai:course(ai); 8,project:course(project); 3,xml:course(xml)\} \leq 20$$

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Weak constraints

- Syntax A **weak constraint** is of the form

$$:\sim a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n. [w@p, t_1, \dots, t_m]$$

where

- a_1, \dots, a_n are atoms
- t_1, \dots, t_m, w , and p are terms
- a_1, \dots, a_n may contain ASP-Core-2 aggregates
- w and p stand for a weight and priority level ($p = 0$ if '@ p ' is omitted)
- Example Minimize statement

```
#minimize [ hd(1)=30@2, hd(2)=40@2, hd(3)=60@2, hd(4)=80@2 ] .
```

can be written in terms of weak constraints as

$$\begin{array}{ll} :\sim \text{hd}(1). [30@2, 1] & :\sim \text{hd}(3). [60@2, 3] \\ :\sim \text{hd}(2). [40@2, 2] & :\sim \text{hd}(4). [80@2, 4] \end{array}$$

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gringo 4

- The input language of *gringo* series 4 comprises
 - ASP-Core-2
 - concepts from *gringo* 3 (conditional literals, #show directives, ...)
- Example The *gringo* 3 rule

```
r(X):p(X):not q(X) :- r(X):p(X):not q(X), 1 {r(X):p(X):not q(X)}.
```

can be written as follows in the language of *gringo* 4:

```
r(X):p(X),not q(X) :- r(X):p(X),not q(X);
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Term-based #show directives as in

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#show. #show hello. #show X : p(X). 1 {p(earth);p(mars);p(venus)}1.
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