

# Answer Set Solving in Practice

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# Language Extensions: Overview

- 1 Two kinds of negation
- 2 Disjunctive logic programs
- 3 Propositional theories
- 4 Aggregates
- 5 Gringo language

# Outline

- 1 Two kinds of negation
- 2 Disjunctive logic programs
- 3 Propositional theories
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# 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  $X = \mathcal{A} \cup \overline{\mathcal{A}}$

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:
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## Train spotting

- $P_1 = \{cross \leftarrow \sim train\}$ 
  - stable model:  $\{cross\}$
- $P_2 = \{cross \leftarrow \neg train\}$ 
  - stable model:  $\emptyset$
- $P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow\}$ 
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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

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

- 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

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

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

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

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$$A(P) = \bigcup_{r \in P} (H(r) \cup B(r)^+ \cup B(r)^-)$$

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# Stable models

## ■ Positive programs

- A set  $X$  of atoms is **closed under** a positive program  $P$  iff for any  $r \in P$ ,  $H(r) \cap X \neq \emptyset$  whenever  $B(r)^+ \subseteq X$ 
  - $X$  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

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

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

## Stable models

- Positive programs

- A set  $X$  of atoms is **closed under** a positive program  $P$  iff for any  $r \in P$ ,  $H(r) \cap X \neq \emptyset$  whenever  $B(r)^+ \subseteq X$ 
    - $X$  corresponds to a model of  $P$  (seen as a formula)
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- Disjunctive programs

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

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

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

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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).
```

```
edge(1,(2;3;4)).   edge(2,(4;5;6)).   edge(3,(1;4;5)).  
edge(4,(1;2)).     edge(5,(3;4;6)).   edge(6,(2;3;5)).
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assign(X,r) ; assign(X,b) ; assign(X,g) :- node(X).
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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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$$P = \left\{ \begin{array}{l} a(1, 2) \leftarrow \\ b(X) ; c(Y) \leftarrow a(X, Y), \sim c(Y) \end{array} \right\}$$

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# Outline

- 1 Two kinds of negation
- 2 Disjunctive logic programs
- 3 Propositional theories**
- 4 Aggregates
- 5 Gringo language

# Propositional theories

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

using

- conjunction ( $\wedge$ )
  - disjunction ( $\vee$ )
  - implication ( $\rightarrow$ )
- Notation

$$\top = (\perp \rightarrow \perp)$$

$$\sim\phi = (\phi \rightarrow \perp)$$

A propositional theory is a finite set of formulas

# Propositional theories

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- The satisfaction relation  $X \models \phi$  between a set  $X$  of atoms and a (set of) formula(s)  $\phi$  is defined as in propositional logic
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$$\phi^X = \perp \quad \text{if } X \not\models \phi$$

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

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

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

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## Relationship to logic programs

- The translation,  $\tau[(\phi \leftarrow \psi)]$ , of a rule  $(\phi \leftarrow \psi)$  is defined as follows:
  - $\tau[(\phi \leftarrow \psi)] = (\tau[\psi] \rightarrow \tau[\phi])$
  - $\tau[\perp] = \perp$
  - $\tau[\top] = \top$
  - $\tau[\phi] = \phi$  if  $\phi$  is an atom
  - $\tau[\sim\phi] = \sim\tau[\phi]$
  - $\tau[(\phi, \psi)] = (\tau[\phi] \wedge \tau[\psi])$
  - $\tau[(\phi; \psi)] = (\tau[\phi] \vee \tau[\psi])$

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,  
 $X$  is a stable model of  $P$  iff  $X$  is a stable model of  $\tau[P]$

## Relationship to logic programs

- The translation,  $\tau[(\phi \leftarrow \psi)]$ , of a rule  $(\phi \leftarrow \psi)$  is defined as follows:
  - $\tau[(\phi \leftarrow \psi)] = (\tau[\psi] \rightarrow \tau[\phi])$
  - $\tau[\perp] = \perp$
  - $\tau[\top] = \top$
  - $\tau[\phi] = \phi$  if  $\phi$  is an atom
  - $\tau[\sim\phi] = \sim\tau[\phi]$
  - $\tau[(\phi, \psi)] = (\tau[\phi] \wedge \tau[\psi])$
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# Logic programs as propositional theories

- 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  $\tau[P] = \{\top \rightarrow p \vee q\}$ 
  - stable models:  $\{p\}$  and  $\{q\}$
- The nested logic program  $P = \{p \leftarrow \sim\sim p\}$  corresponds to  $\tau[P] = \{\sim\sim p \rightarrow p\}$ 
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# Outline

- 1 Two kinds of negation
- 2 Disjunctive logic programs
- 3 Propositional theories
- 4 Aggregates**
- 5 Gringo language

# Motivation

- Aggregates provide a general way to obtain a single value from a collection of input values
- Popular aggregate (functions)
  - average
  - count
  - maximum
  - minimum
  - sum
- Cardinality and weight constraints rely on count and sum aggregates

## Syntax

- An **aggregate** has the form:

$$\alpha \{w_1 : a_1, \dots, w_m : a_m, w_{m+1} : \sim a_{m+1}, \dots, w_n : \sim a_n\} \prec k$$

where for  $1 \leq i \leq n$

- $\alpha$  stands for a function mapping multisets over  $\mathbb{Z}$  to  $\mathbb{Z} \cup \{+\infty, -\infty\}$
  - $\prec$  stands for a relation between  $\mathbb{Z} \cup \{+\infty, -\infty\}$  and  $\mathbb{Z}$
  - $k \in \mathbb{Z}$
  - $a_i$  are atoms and
  - $w_i$  are integers
- Example  $sum \{30 : hd(a), \dots, 50 : hd(m)\} \leq 300$

## Semantics

- A (positive) aggregate  $\alpha \{w_1 : a_1, \dots, w_n : a_n\} \prec k$  can be represented by the formula:

$$\bigwedge_{I \subseteq \{1, \dots, n\}, \alpha \{w_i | i \in I\} \not\prec k} \left( \bigwedge_{i \in I} a_i \rightarrow \bigvee_{i \in \bar{I}} a_i \right)$$

where  $\bar{I} = \{1, \dots, n\} \setminus I$  and  $\not\prec$  is the complement of  $\prec$

- Then,  $\alpha \{w_1 : a_1, \dots, w_n : a_n\} \prec k$  is true in  $X$  iff the above formula is true in  $X$

## Example

- Consider  $sum\{1 : p, 1 : q\} \neq 1$   
That is,  $a_1 = p$ ,  $a_2 = q$  and  $w_1 = 1$ ,  $w_2 = 1$
- Calculemus!

$I$	$\{w_i \mid i \in I\}$	$\sum\{w_i \mid i \in I\}$	$\sum\{w_i \mid i \in I\} = 1$
$\emptyset$	$\{\}$	0	<i>false</i>
$\{1\}$	$\{1\}$	1	<i>true</i>
$\{2\}$	$\{1\}$	1	<i>true</i>
$\{1, 2\}$	$\{1, 1\}$	2	<i>false</i>

- We get  $(p \rightarrow q) \wedge (q \rightarrow p)$
- Analogously, we obtain  $(p \vee q) \wedge \neg(p \wedge q)$  for  $sum\{1 : p, 1 : q\} = 1$

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# Monotonicity

## ■ Monotone aggregates

- For instance,

- $B(r)^+$

- $sum\{1 : p, 1 : q\} > 1$  amounts to  $p \wedge q$

- We get a simpler characterization:  $\bigwedge_{I \subseteq \{1, \dots, n\}, \alpha \{w_i | i \in I\} \neq k} \bigvee_{i \in \bar{I}} a_i$

## ■ Anti-monotone aggregates

- For instance,

- $B(r)^-$

- $sum\{1 : p, 1 : q\} < 1$  amounts to  $\neg p \wedge \neg q$

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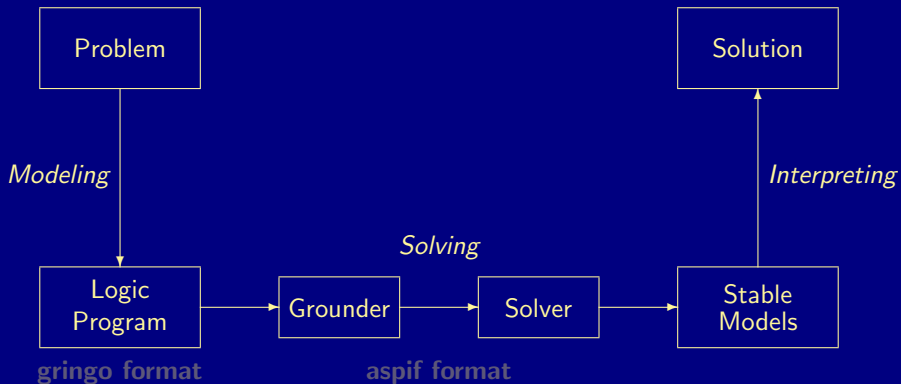
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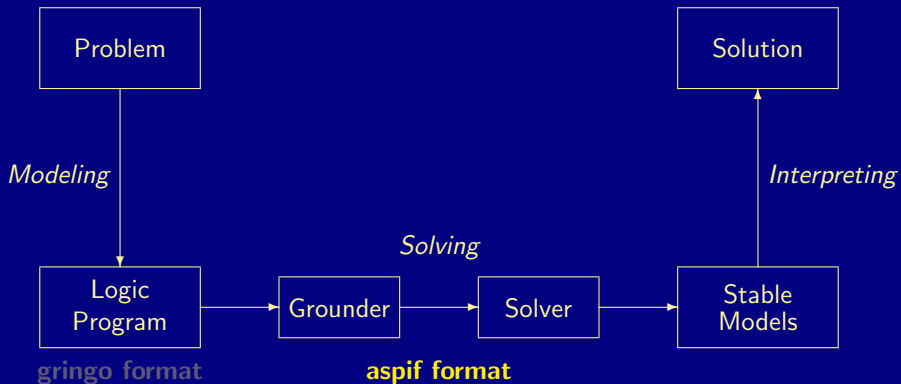
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## Gringo language



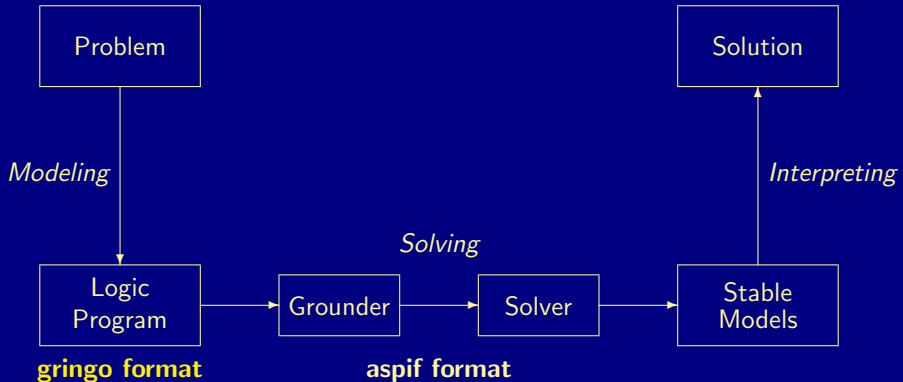
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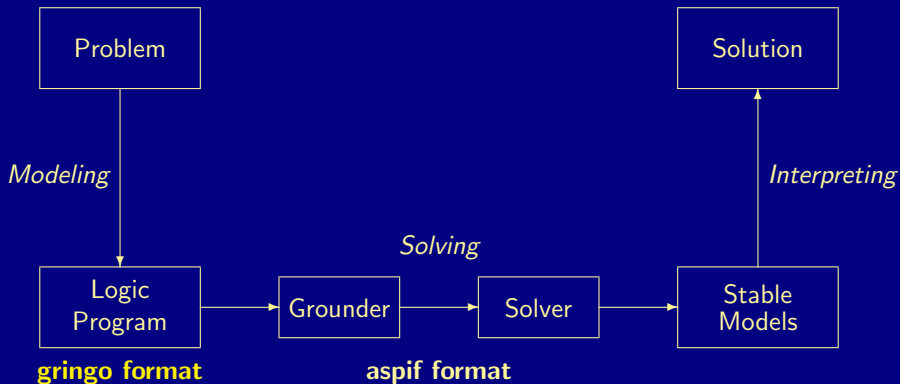
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# Terms and literals

- Terms  $t$
- Tuples  $\mathbf{t}$
- Atoms  $a, \neg a$
- Symbolic literals  $a, \sim a, \sim\sim a$
- Arithmetic literals  $t_1 \prec t_2$
- Conditional literals  $l : L$
- Aggregate atoms  $s_1 \prec_1 \alpha\{\mathbf{t}_1 : L_1; \dots; \mathbf{t}_n : L_n\} \prec_2 s_2$
- Aggregate literals  $a, \sim a, \sim\sim a$
- Literals

# Terms and literals

- Terms  $t$  are formed from
  - constant symbols, eg  $c, d, \dots$
  - function symbols, eg  $f, g, \dots$
  - numeric symbols, eg  $1, 2, \dots$
  - variable symbols, eg  $X, Y, \dots, _$
  - parentheses  $(, )$
  - tuple delimiters  $\langle, \rangle$  (omitted whenever possible)
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- tuple delimiters  $\langle, \rangle$

eg  $f(3,c,Z), g(42,-,-)$ , or  $f(\langle 3,c \rangle, X)$

- Tuples  $t$

- Atoms  $a, \neg a$

- Symbolic literals  $a, \sim a, \sim\sim a$

- Arithmetic literals  $t_1 \prec t_2$

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- Aggregate atoms  $s_1 \prec_1 \alpha\{t_1 : L_1; \dots; t_n : L_n\} \prec_2 s_2$

- Aggregate literals  $a, \sim a, \sim\sim a$

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# Terms and literals

- Terms  $t$
- Tuples  $\mathbf{t}$  of terms
- Atoms  $a, \neg a$
- Symbolic literals  $a, \sim a, \sim\sim a$
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# Terms and literals

- Terms  $t$
- Tuples  $\mathbf{t}$
- (Negated) Atoms  $a, \neg a$  are formed from
  - predicate symbols, eg  $p, q, \dots$
  - parentheses  $(, )$
  - tuples of terms
- Symbolic literals  $a, \sim a, \sim\sim a$
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  - predicates, eg  $p, q, \dots$
  - parentheses  $(, )$
  - tuples of terms

eg  $\neg p(f(3, c, Z), g(42, -, -))$  or  $q()$  written as  $q$

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# Terms and literals

- Terms  $t$
- Tuples  $\mathbf{t}$
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viz **#false** and **#true**
- Symbolic literals  $a, \sim a, \sim\sim a$
- Arithmetic literals  $t_1 < t_2$
- Conditional literals  $l : L$
- Aggregate atoms  $s_1 <_1 \alpha\{\mathbf{t}_1 : L_1; \dots; \mathbf{t}_n : L_n\} <_2 s_2$
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- Tuples  $\mathbf{t}$
- Atoms  $a, \neg a, \perp, \top$
- Symbolic literals  $a, \sim a, \sim\sim a$   
eg  $p(a,X), \text{'not } p(a,X)\text{'}, \text{'not not } p(a,X)\text{'}$
- Arithmetic literals  $t_1 < t_2$
- Conditional literals  $I : L$
- Aggregate atoms  $s_1 <_1 \alpha\{t_1 : L_1; \dots; t_n : L_n\} <_2 s_2$
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  - $t_1$  and  $t_2$  are terms
  - $\prec$  is a comparison symbol
- Conditional literals  $l : L$
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  - $\prec$  is a comparison symbol
 eg  $3 < 1$  or  $f(42) = X$
- Conditional literals  $l : L$
- Aggregate atoms  $s_1 \prec_1 \alpha\{t_1 : L_1; \dots; t_n : L_n\} \prec_2 s_2$
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- Literals

# Terms and literals

- Terms  $t$
- Tuples  $\mathbf{t}$ ,  $\mathbf{L}$  of literals
- Atoms  $a, \neg a, \perp, \top$
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- Arithmetic literals  $t_1 \prec t_2$
- Conditional literals  $l : \mathbf{L}$  where
  - $l$  is a symbolic or arithmetic literal
  - $\mathbf{L}$  is a tuple of symbol or arithmetic literals
- Aggregate atoms  $s_1 \prec_1 \alpha\{t_1 : \mathbf{L}_1; \dots; t_n : \mathbf{L}_n\} \prec_2 s_2$
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- Tuples  $\mathbf{t}, \mathbf{L}$
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- Conditional literals  $l : \mathbf{L}$  where
  - $l$  is a symbolic or arithmetic literal
  - $\mathbf{L}$  is a tuple of symbol or arithmetic literals
  - $l : \mathbf{L}$  is written as  $l$  whenever  $\mathbf{L}$  is empty
- Aggregate atoms  $s_1 \prec_1 \alpha\{t_1 : \mathbf{L}_1; \dots; t_n : \mathbf{L}_n\} \prec_2 s_2$
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- Atoms  $a, \neg a, \perp, \top$
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- Conditional literals  $l : \mathbf{L}$  where
  - $l$  is a symbolic or arithmetic literal
  - $\mathbf{L}$  is a tuple of symbol or arithmetic literals
 eg 'p(X,Y):q(X),r(Y)' or p(42) or '#false:q'
- Aggregate atoms  $s_1 \prec_1 \alpha\{t_1 : \mathbf{L}_1; \dots; t_n : \mathbf{L}_n\} \prec_2 s_2$
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eg  $10 \leq \#sum \{6, C:course(C); 3, S:seminar(S)\} \leq 20$
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- For a detailed account please consult the user's guide!

## Rules

- Rules are of the form

$$l_1 ; \dots ; l_m \leftarrow l_{m+1}, \dots, l_n \quad (2)$$

where

- $l_i$  is a conditional literal for  $1 \leq i \leq m$  and
  - $l_i$  is a literal for  $m + 1 \leq i \leq n$
- Note Semicolons ';' must be used in (2) instead of commas ',' whenever some  $l_i$  is a (genuine) conditional literal for  $1 \leq i \leq n$
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## Shortcuts

- A rule of the form

$$s_1 \prec_1 \alpha \{ \mathbf{t}_1 : l_1 : \mathbf{L}_1; \dots; \mathbf{t}_k : l_k : \mathbf{L}_k \} \prec_2 s_2 \leftarrow l_{m+1}, \dots, l_n$$

where

- $\alpha$ ,  $\prec_j$ ,  $s_j$ ,  $\mathbf{t}_j$  are as given above for  $i = 1, 2$  and  $1 \leq j \leq k$
- $l_j : \mathbf{L}_j$  is a conditional literal for  $1 \leq j \leq k$
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is a shorthand for the following  $k + 1$  rules

$$\begin{aligned} \{l_j\} \leftarrow l_{m+1}, \dots, l_n, \mathbf{L}_j & \quad \text{for } 1 \leq j \leq k \\ \leftarrow l_{m+1}, \dots, l_n, \sim s_1 \prec_1 \alpha \{ \mathbf{t}_1 : l_1, \mathbf{L}_1; \dots; \mathbf{t}_k : l_k, \mathbf{L}_k \} \prec_2 s_2 \end{aligned}$$

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# Examples

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*gringo* generates two distinct term tuples 1,0,a and 1,0,b

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## Weak constraints

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

$$:\sim l_1, \dots, l_n. [w@p, t_1, \dots, t_m]$$

where

- $l_1, \dots, l_n$  are literals
- $t_1, \dots, t_m$ ,  $w$ , and  $p$  are terms
- $w$  and  $p$  stand for a weight and priority level ( $p = 0$  if '@ $p$ ' is omitted)
- Example The weak constraint

$$:\sim \text{hd}(I,P,C) . [C@2,I]$$

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*gringo* 3 versus 4/5

- The input language of *gringo* series 4/5 comprises
  - ASP-Core-2
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- Example The *gringo* 3 rule
  - $r(X) : p(X) : \text{not } q(X) :- r(X) : p(X) : \text{not } q(X),$   
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*gringo* 3 versus 4/5

- The input language of *gringo* series 4/5 comprises
  - ASP-Core-2
  - concepts from *lparse* and *gringo* 3
- Example The *gringo* 3 rule
  - $r(X) : p(X) : \text{not } q(X) :- r(X) : p(X) : \text{not } q(X),$   
 $1 \{ r(X) : p(X) : \text{not } q(X) \}.$   
 can be written as follows in the language of *gringo* 4/5:
    - $r(X) : p(X), \text{not } q(X) :- r(X) : p(X), \text{not } q(X);$   
 $1 \{ r(X) : p(X), \text{not } q(X) \}.$
- Note Directives `#compute`, `#domain`, and `#hide` are discontinued
- Attention
  - The languages of *gringo* 3 and 4/5 are not fully compatible
  - Many example programs in the literature are written for *gringo* 3

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