

Answer Set Solving in Practice

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¹Standing on the shoulders of a great research group and community!

Organization: Overview

- 1 Roadmap
- 2 Resources
- 3 Literature
- 4 Systems

Outline

- 1 Roadmap
- 2 Resources
- 3 Literature
- 4 Systems

What's on the menu?

- 1 Motivation
 - 2 Introduction
 - 3 Modeling
 - 4 Language
 - 5 Grounding
 - 6 Foundations
 - 7 Solving
 - 8 Multi-shot solving
 - 9 Theory solving
 - 10 Heuristic-driven solving
 - 11 Systems
 - 12 Advanced modeling
 - 13 Preferences and Optimization
 - 14 Applications
- Bibliography

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Resources

■ Course material

- <https://github.com/potassco-asp-course>
- <https://potassco.org/teaching>

■ Videos

- <https://youtube.com/c/potassco-live>

■ Mailing lists

- <https://sourceforge.net/projects/potassco/lists/potassco-users>
- <https://sourceforge.net/projects/potassco/lists/potassco-announce>

■ Contact

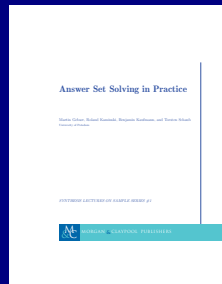
- asp@lists.cs.uni-potsdam.de

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- 1 Roadmap
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The Potassco Book

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

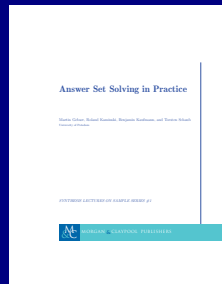


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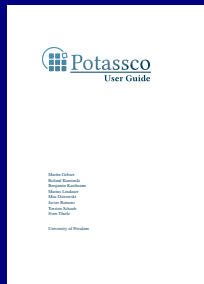
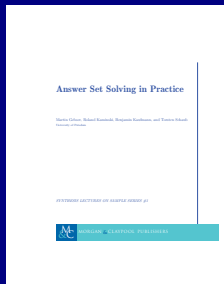
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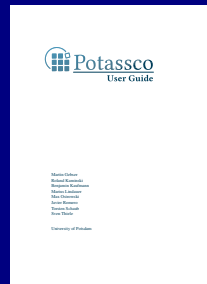
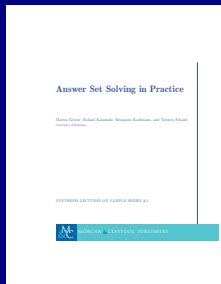
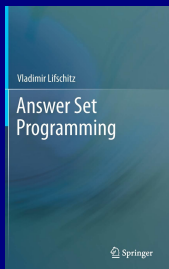
The Potassco Book and Guide



Resources

- <http://potassco.org/book>
- <http://potassco.org/teaching>

The ASP and Potassco Book, and Guide



Resources

- <http://potassco.org/book>
- <http://potassco.org/teaching>

Literature

- Books [5], [25], [32], [38], [43]
- Surveys [40], [31], [19], [10], [36], [47]
- Magazines [9], [48]
- Articles [35], [34], [7], [45], [44], [39], [33], [23], etc.
- Guide [29]
- See also our bibliography on [github](#)

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Systems

■ Systems

- *clingo* [27]

<https://potassco.org>

- *dlv* [37, 2]

<http://www.dlvsystem.com>

■ Grounders

- *lparse* [50]

- *gringo* [22, 28]+[24, 12]

<https://potassco.org>

- *idlv* [14]+[12]

<http://www.dlvsystem.com>

■ Solvers

- *smodels* [46, 49]

- *clasp* [26, 21]

<https://potassco.org>

- *wasp* [4]

<https://www.mat.unical.it/ricca/wasp>

■ Encodings

- *asparagus* [8]

<https://asparagus.cs.uni-potsdam.de>

- competitions [30, 18, 16, 3, 15]

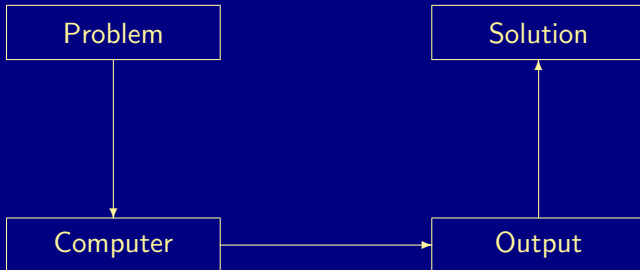
Motivation: Overview

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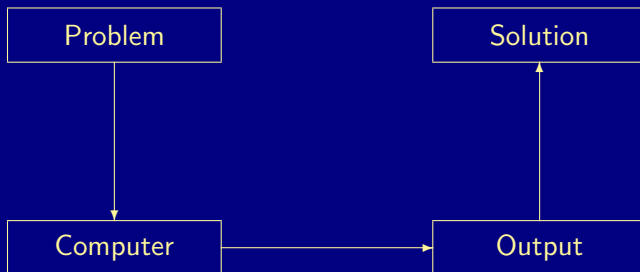
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Informatics



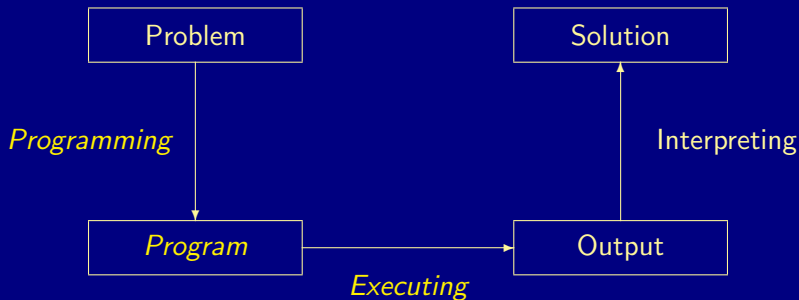
Traditional programming

"How to solve the problem?"



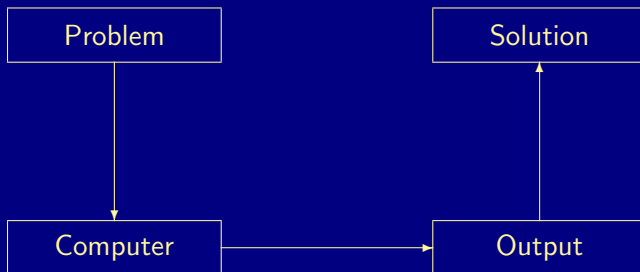
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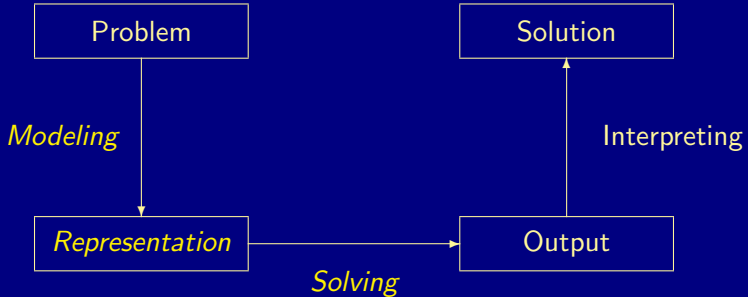
Declarative programming

“What is the problem?”



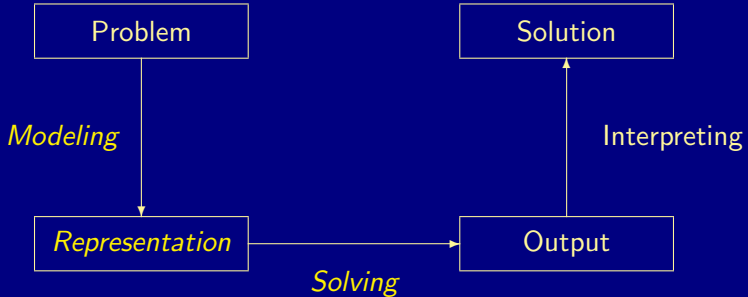
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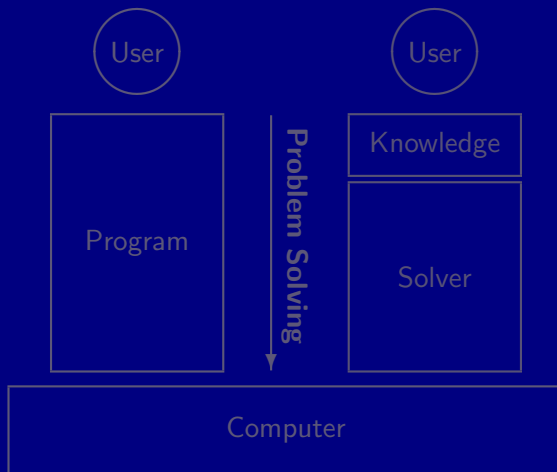


Declarative problem solving

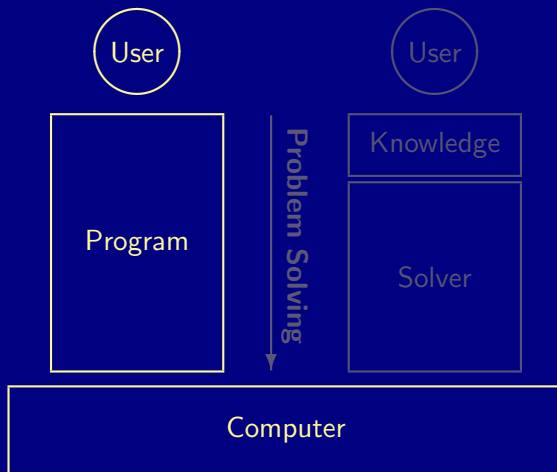
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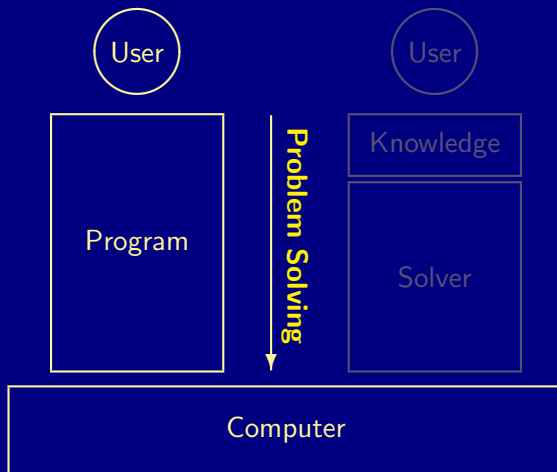
Traditional Software



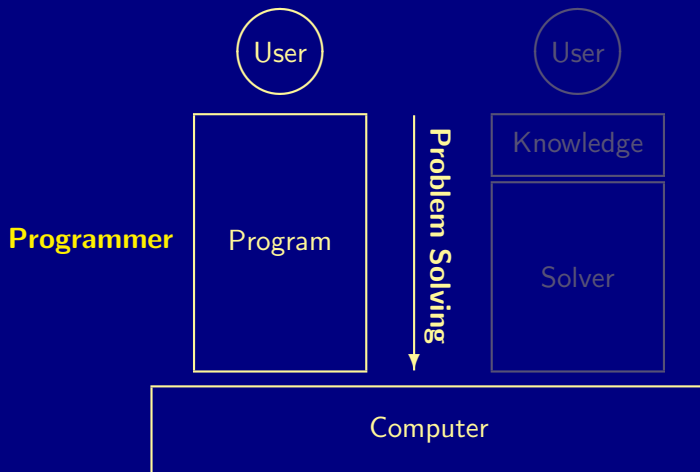
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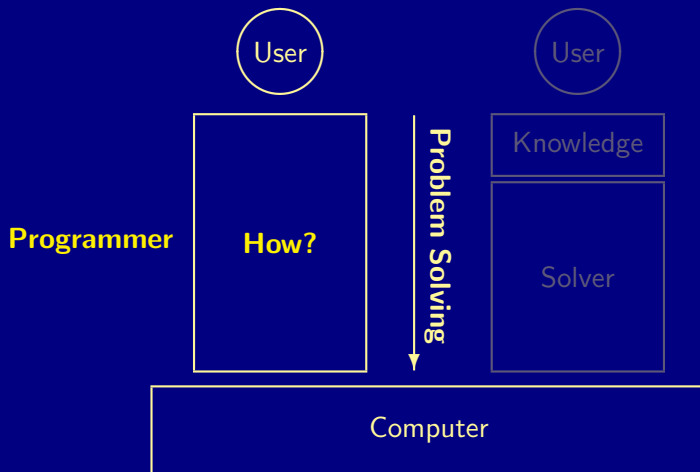
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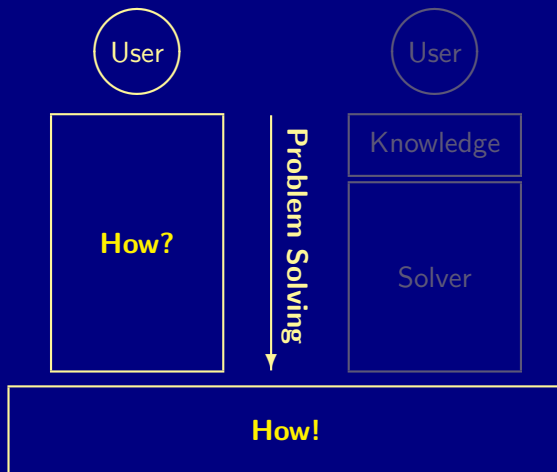
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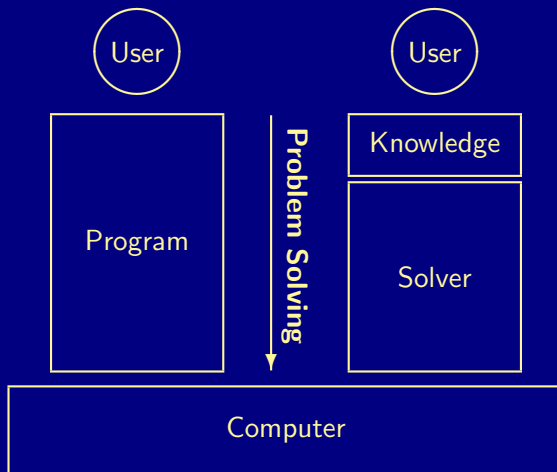
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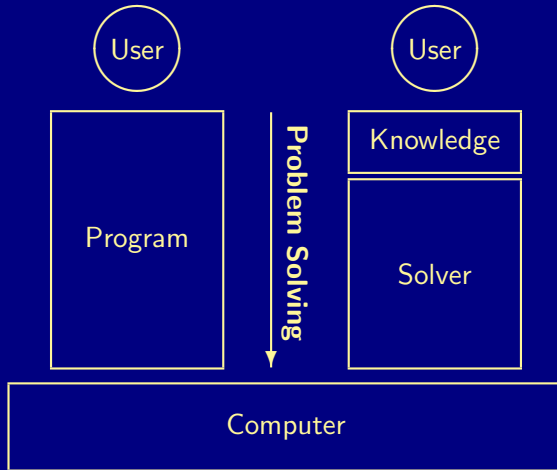
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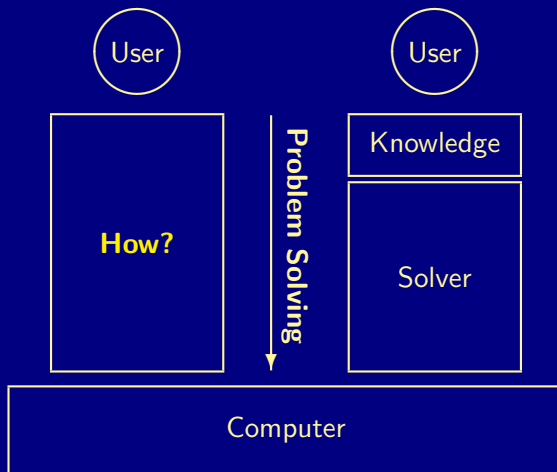
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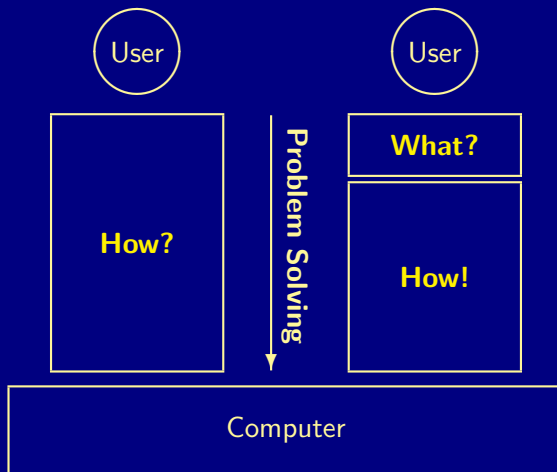
Knowledge-driven Software



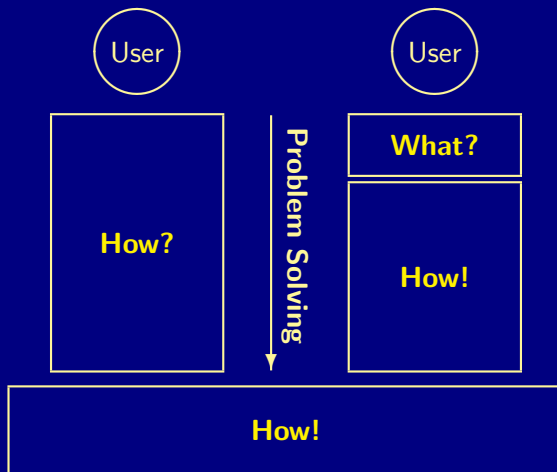
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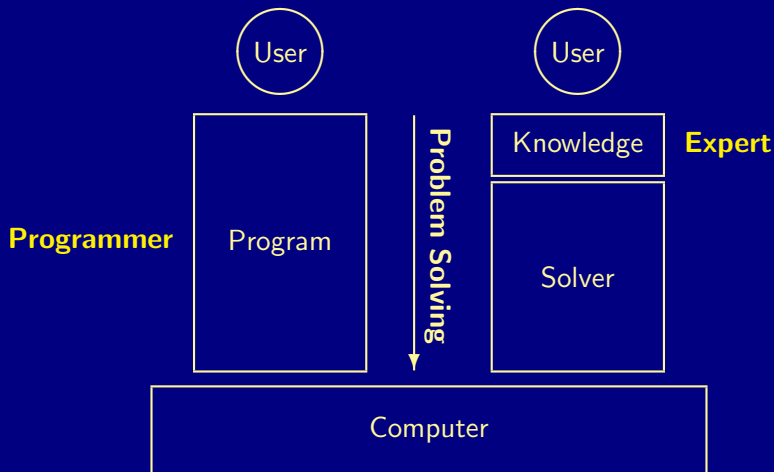
Knowledge-driven Software



Knowledge-driven Software



Knowledge-driven Software



What is the benefit?

- + Transparency
- + Flexibility
- + Maintainability
- + Reliability

- + Generality
- + Efficiency
- + Optimality
- + Availability

Knowledge

Expert

Solver

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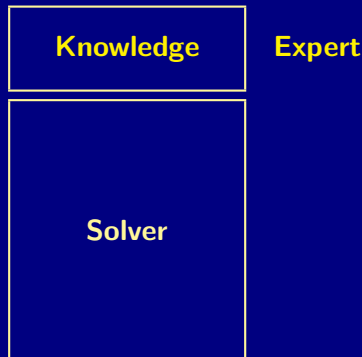
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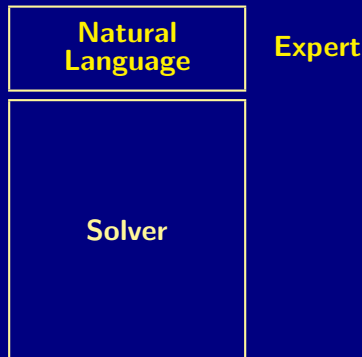
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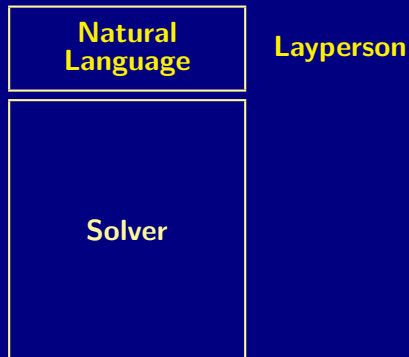
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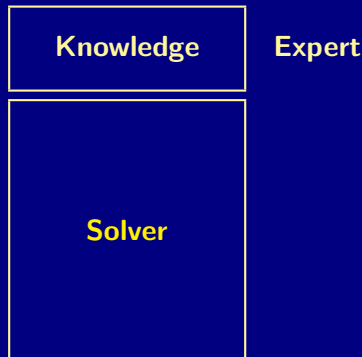
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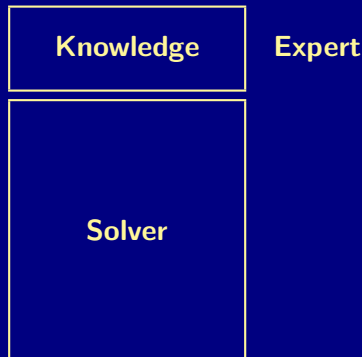
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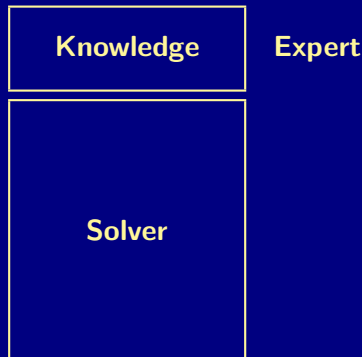
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Answer Set Programming (ASP)

- What is ASP?

ASP is an approach for declarative problem solving

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- Where is ASP from?

- Databases
- Logic programming
- Knowledge representation and reasoning
- Satisfiability solving

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Solving knowledge-intense combinatorial (optimization) problems

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Examples Sudoku, Configuration, Diagnosis, Music composition, Planning, System design, Time tabling, etc.

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- Debian, Ubuntu: Linux package configuration
- Exeura: Call routing
- Fcc: Radio frequency auction
- Gioia Tauro: Workforce management
- Nasa: Decision support for Space Shuttle
- Siemens: Partner units configuration
- Variantum: Product configuration
- US Navy: risk assessment

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Over 13 months in 2016–17 the **US Federal Communications Commission** conducted an “incentive auction” to repurpose radio spectrum from broadcast television to wireless internet. In the end, the auction yielded **\$19.8 billion**, \$10.05 billion of which was paid to 175 broadcasters for voluntarily relinquishing their licenses across 14 UHF channels. Stations that continued broadcasting were assigned potentially new channels to fit as densely as possible into the channels that remained. The government netted more than **\$7 billion** (used to pay down the national debt) after covering costs. A crucial element of the auction design was the construction of a **solver**, dubbed SATFC, **that determined whether sets of stations could be “repacked” in this way; it needed to run every time a station was given a price quote.** This

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Problems consisting of (many) decisions and constraints

- What are ASP's distinguishing features?

- High level, versatile modeling language
- High performance solvers
- Qualitative and quantitative optimization

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- What are ASP's distinguishing features?

- High level, versatile modeling language
- High performance solvers
- Qualitative and quantitative optimization

- Any industrial impact?

- ASP Tech companies: DLV Systems and **Potassco Solutions**
- Increasing interest in (large) companies

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Some (biased) moments in time

- '80 Capturing incomplete information
- '90 Amalgamation and computation
- '00 Applications and semantic rediscoveries
- '10 Customization and integration

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- '80 Capturing incomplete information
 - Databases Closed world assumption
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 - Non-monotonic reasoning
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 - Complex reasoning modes APIs, multi-shot solving
 - Hybridization Constraint ASP, theory solving

Paradigm shift '90

Theorem Proving based approach (eg. Prolog)

- 1 Provide a representation of the problem
- 2 A solution is given by a **derivation** of a query

Model Generation based approach (eg. SATisfiability testing)

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- 2 A solution is given by a **model** of the representation

Automated planning, Kautz and Selman (ECAI'92)

Represent planning problems as propositional theories so that models not proofs describe solutions

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Logic programs

- A logic program is a set of rules of the form

$$\underbrace{a}_{\text{head}} \leftarrow \underbrace{b_1, \dots, b_m, \neg c_1, \dots, \neg c_n}_{\text{body}}$$

where

- a and all b_i, c_j are atoms (propositional variables)
 - $\leftarrow, ,, \neg$ denote if, and, and negation
 - intuitive reading: head must be true if body holds
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- Semantics given by stable models, informally,
 - 1 (classical) models of the logic program
 - 2 requiring that each true atom is provable

**Closed world
assumption**

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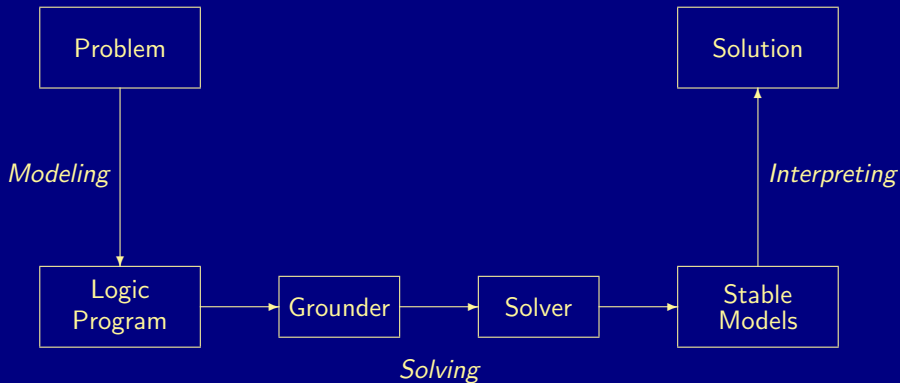
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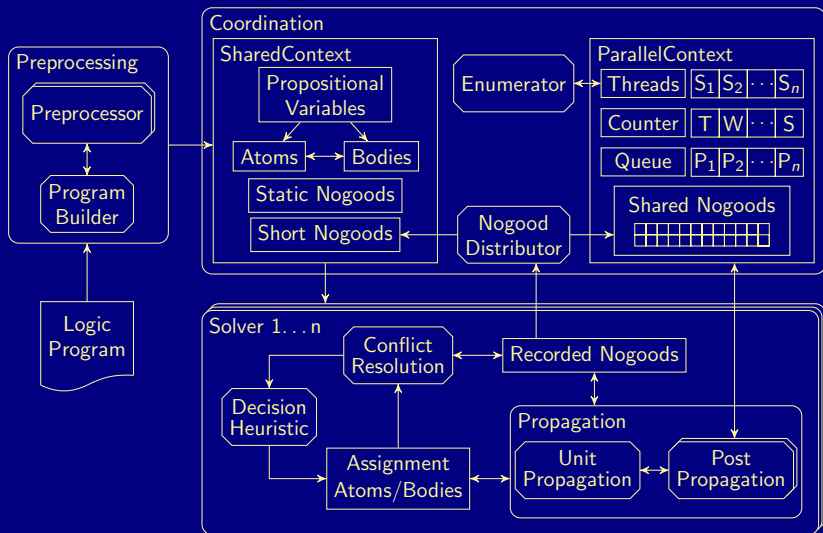
Modeling, grounding, and solving



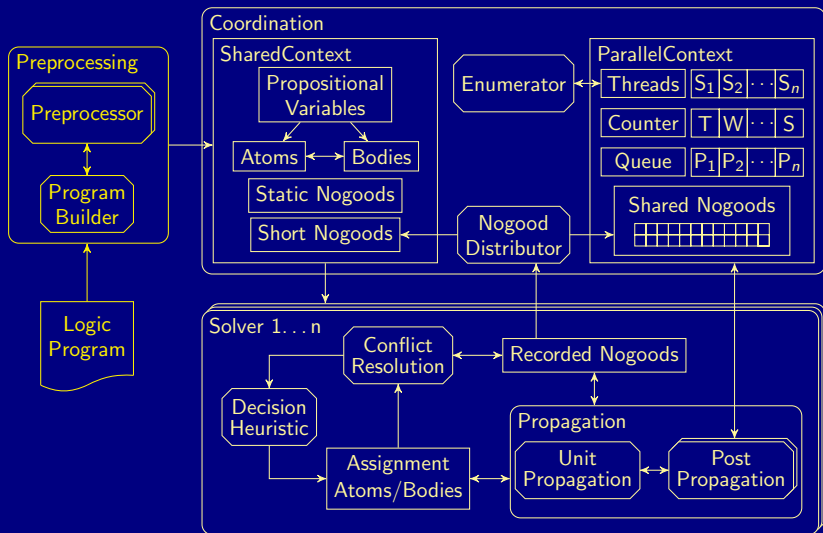
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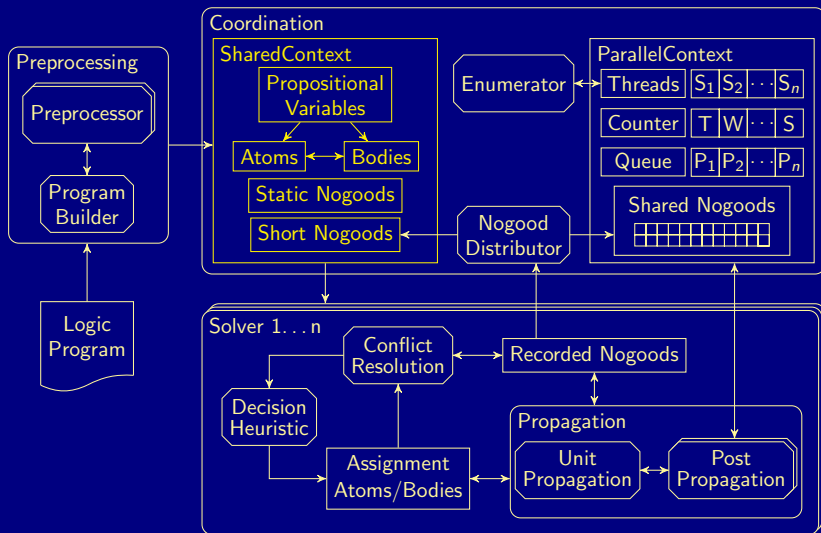
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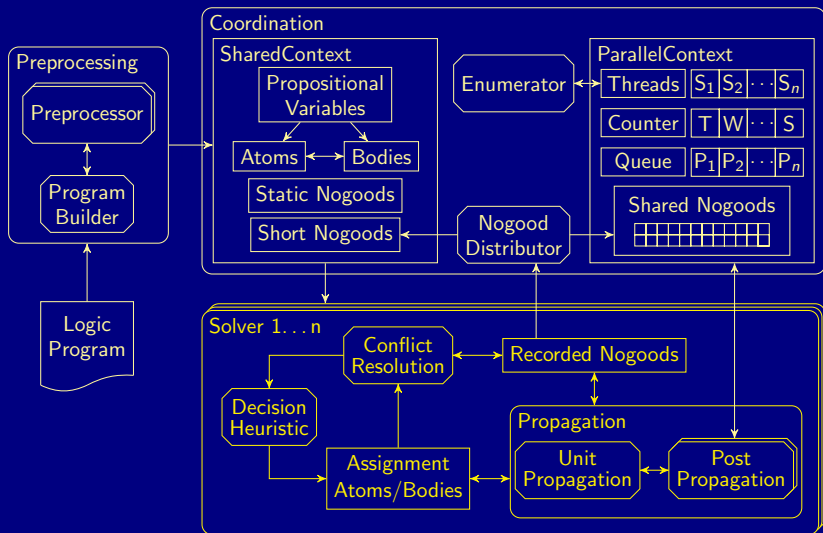
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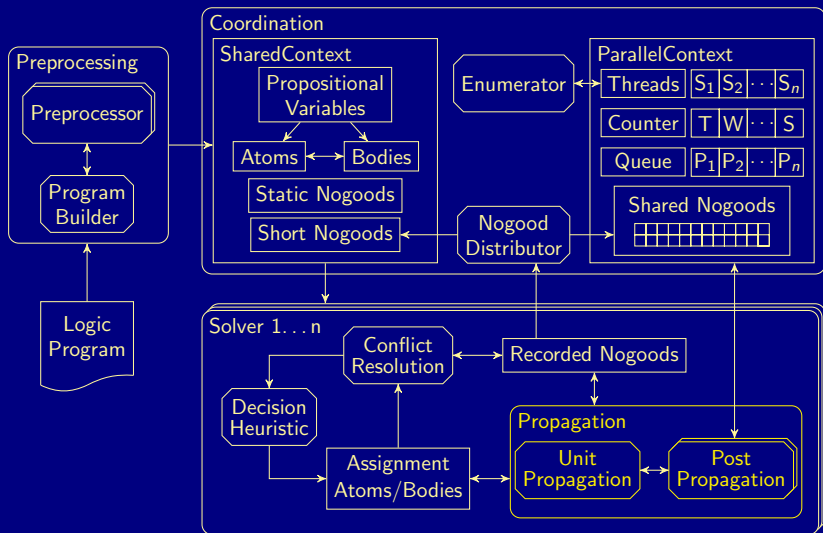
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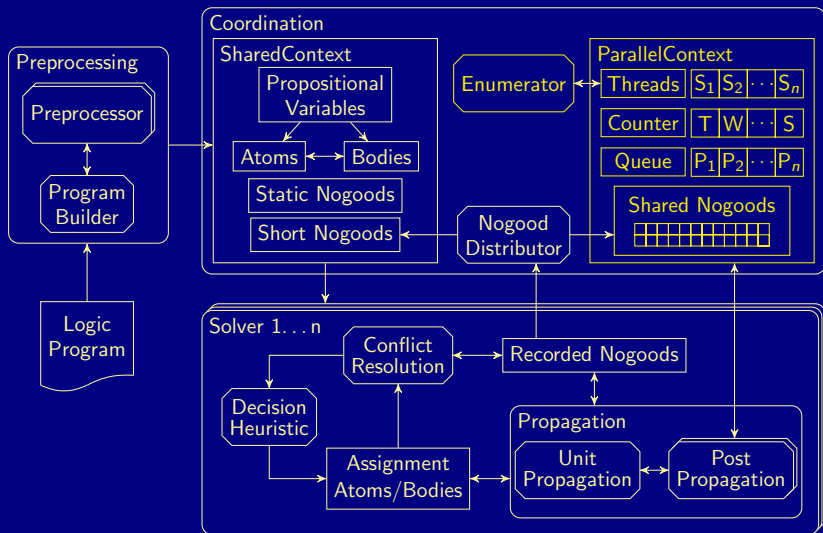
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Outline

- 5 Motivation
- 6 Nutshell
- 7 Evolution
- 8 Foundation
- 9 Workflow
- 10 Engine
- 11 Usage**
- 12 Summary

Two sides of a coin

- ASP as High-level Language
 - Express problem instance as sets of facts
 - Encode problem class as a set of rules
 - Read off solutions from stable models of facts and rules
- ASP as Low-level Language
 - Compile a problem into a set of facts and rules
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Upcoming experience

- ASP is a viable tool for Knowledge Representation and Reasoning
 - Integration of DB, LP, KR, and SAT techniques
 - Combinatorial search problems in the realm of NP and NP^{NP}
 - Succinct, elaboration-tolerant problem representations
 - rapid application development tool
 - Easy handling of knowledge-intensive applications
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Introduction: Overview

13 Syntax

14 Semantics

15 Reasoning

16 Language

17 Variables

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Outline

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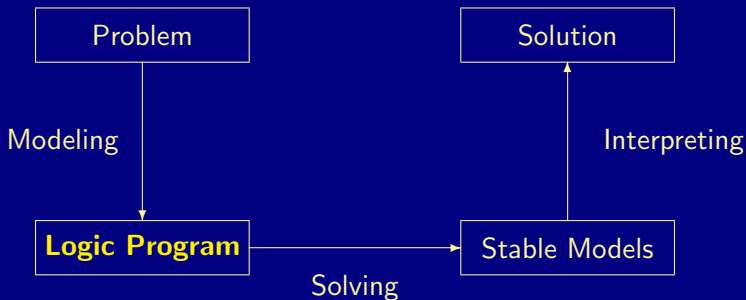
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Syntax



Normal logic programs

- A logic program, P , over a set \mathcal{A} of atoms is a finite set of rules
- A (normal) rule, r , is of the form

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

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$$h(r) = a_0$$

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Examples

■ Example rules

- $a \leftarrow b, \neg c$

- $a \leftarrow \neg c, b$

- $a \leftarrow$

- $a \leftarrow b$

- $a \leftarrow \neg c$

- $bachelor(joe) \leftarrow male(joe), \neg married(joe)$

■ Example literals

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Notational conventions

	false, true	if	and	or	iff	(default) negation	strong negation
source code		<code>:-</code>	<code>,</code>	<code>;</code>		<code>not</code>	<code>-</code>
logic program		\leftarrow	<code>,</code>	<code>;</code>		\neg	\sim
formula	\perp, \top	\rightarrow	\wedge	\vee	\leftrightarrow	\neg	\sim

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13 Syntax

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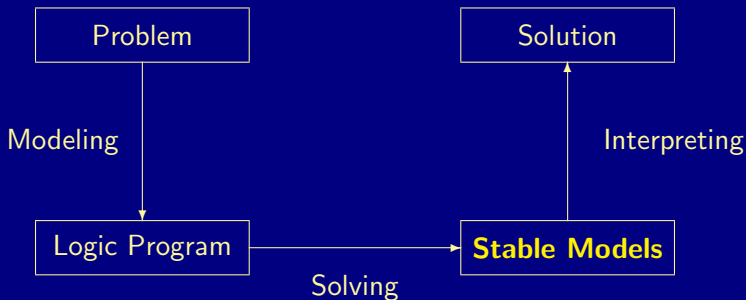
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Semantics



What is a model?

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- **Solution** An assignment satisfying a set of constraints

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 - **Example** A is a solution of $\{2x < z, x + y < 2z\}$

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 - Example We use $\{a, b\}$ to represent $\{a \mapsto \mathbf{T}, b \mapsto \mathbf{T}, c \mapsto \mathbf{F}\}$

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We often denote interpretations by the set of their true atoms and (alternatively) use sets as the semantic cornerstones (rather than assignments or interpretations)

What is a stable model?

- Reduct-based characterization
- Logical characterization
- Axiomatic characterization
- Operational characterization
- Proof-theoretic characterization
- Constraint-based characterization
- Algorithmic characterization
- C++-based characterization

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What is the meaning of a logic program?

- Idea: The set of atoms derivable from the rules in program
- Question: How to characterize X given P ?

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     $X := \{h(r) \mid r \in P, B(r)^+ \subseteq X\}$   
return  $X$ 
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- Positive rules are also referred to as definite clauses
 - Definite clauses are disjunctions with exactly one positive atom:

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- Horn clauses are clauses with at most one positive atom
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- Logical attempt
 - Hypothesis The smallest model of P ?

What is the meaning of a logic program?

- Lessons learned from positive programs
 - $Cn(P)$ is the smallest model of P , eliminating all others
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What is the meaning of a logic program?

- Lessons learned from positive programs
 - $Cn(P)$ is the smallest model of P , eliminating all others
 - Every atom in $Cn(P)$ is justified by a proof
- Logical attempt — failed
 - Hypothesis The smallest model of P ?
 - Example $P = \{a \leftarrow \neg b\}$ yields (stable model) $\{a\}$ but P has two minimal models, $\{a\}$ and $\{b\}$

What is the meaning of a logic program?

- Lessons learned from positive programs
 - $Cn(P)$ is the smallest set closed under P , eliminating all others
 - Every atom in $Cn(P)$ is justified by a proof
- Procedural attempt

What is the meaning of a logic program?

■ Procedural attempt

```
let  $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X\}$  in  
   $X := \emptyset$   
  while  $T_P X \neq X$   
     $X := T_P X$   
return  $X$ 
```


What is the meaning of a logic program?

■ Procedural attempt

```
let  $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap X = \emptyset\}$  in  
   $X := \emptyset$   
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■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$

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```

■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$

$$X_0 = \emptyset$$

What is the meaning of a logic program?

■ Procedural attempt

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```

■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$

$$X_0 = \emptyset$$

$$T_P X_0 = \{a\}$$

What is the meaning of a logic program?

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```

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     $X := T_P X$ 
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```

■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$

$$\begin{aligned}
 X_0 &= \emptyset & X_1 &= \{a\} \\
 T_P X_0 &= \{a\}
 \end{aligned}$$

What is the meaning of a logic program?

■ Procedural attempt

```

let  $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap X = \emptyset\}$  in
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```

■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c, c \leftarrow b\}$

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$$X_0 = \emptyset$$

$$T_P X_0 = \{a\}$$

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return X

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 \end{array}$$

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 $X := T_P X$
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 $X := T_P X$
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 $X := T_P X$
return X

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$$\begin{array}{llll}
 X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X_3 = \{a, b, c\} \\
 T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b, c\} & T_P X_3 = \{a, c\} \\
 & & & X_4 = \{a, c\}
 \end{array}$$

What is the meaning of a logic program?

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap X = \emptyset\}$ **in**
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$$\begin{array}{l} X_4 = \{a, c\} \\ T_P X_4 = \{a\} \end{array}$$

What is the meaning of a logic program?

■ Procedural attempt

let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap X = \emptyset\}$ **in**
 $X := \emptyset$
while $T_P X \neq X$
 $X := T_P X$
return X

■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c, c \leftarrow b\}$

$$\begin{array}{llll}
 X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X_3 = \{a, b, c\} \\
 T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b, c\} & T_P X_3 = \{a, c\} \\
 & & & \\
 X_5 = \{a\} & & & X_4 = \{a, c\} \\
 & & & T_P X_4 = \{a\}
 \end{array}$$

What is the meaning of a logic program?

■ Procedural attempt

let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap X = \emptyset\}$ **in**
 $X := \emptyset$
while $T_P X \neq X$
 $X := T_P X$
return X

■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c, c \leftarrow b\}$

$X_0 = \emptyset$	$X_1 = \{a\}$	$X_2 = \{a, b\}$	$X_3 = \{a, b, c\}$
$T_P X_0 = \{a\}$	$T_P X_1 = \{a, b\}$	$T_P X_2 = \{a, b, c\}$	$T_P X_3 = \{a, c\}$
	$X_5 = \{a\}$		$X_4 = \{a, c\}$
	...		$T_P X_4 = \{a\}$

What is the meaning of a logic program?

■ Procedural attempt

```

let  $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap X = \emptyset\}$  in
   $X := \emptyset$ 
  while  $T_P X \neq X$ 
     $X := T_P X$ 
  return  $X$ 

```

■ Example $P = \{a \leftarrow, b \leftarrow a, \neg c, c \leftarrow b\}$

What is the meaning of a logic program?

- Procedural attempt — failed

```
let  $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap X = \emptyset\}$  in  
   $X := \emptyset$   
  while  $T_P X \neq X$   
     $X := T_P X$   
return  $X$ 
```

What is the meaning of a logic program?

- Procedural attempt patched

guess Y

let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

$X := \emptyset$

while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

What is the meaning of a logic program?

- Procedural attempt patched

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

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while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$

What is the meaning of a logic program?

- Procedural attempt patched

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$X := T_P X$

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- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \emptyset$

What is the meaning of a logic program?

- Procedural attempt patched

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

$X := \emptyset$

while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \emptyset$

$$X_0 = \emptyset \quad X_1 = \{a\} \quad X_2 = \{a, b\}$$

$$T_P X_0 = \{a\} \quad T_P X_1 = \{a, b\} \quad T_P X_2 = \{a, b\}$$

What is the meaning of a logic program?

- Procedural attempt patched

guess Y

let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

$X := \emptyset$

while $T_P X \neq X$

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if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \emptyset$

$$\begin{array}{llll} X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X = \{a, b\} \\ T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b\} & \end{array}$$

What is the meaning of a logic program?

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

$X := \emptyset$

while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \emptyset$

$$\begin{array}{llll}
 X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X = \{a, b\} \neq Y \\
 T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b\} &
 \end{array}$$

What is the meaning of a logic program?

- Procedural attempt patched

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

$X := \emptyset$

while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \emptyset$ ✗

$$\begin{array}{llll}
 X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X = \{a, b\} \neq Y \\
 T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b\} &
 \end{array}$$

What is the meaning of a logic program?

- Procedural attempt patched

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if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{a\}$

What is the meaning of a logic program?

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What is the meaning of a logic program?

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

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while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{a\}$ ✗

$$\begin{array}{llll}
 X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X = \{a, b\} \neq Y \\
 T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b\} &
 \end{array}$$

What is the meaning of a logic program?

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

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while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{b\}$

What is the meaning of a logic program?

- Procedural attempt patched

guess Y

let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

$X := \emptyset$

while $T_P X \neq X$

$X := T_P X$

if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{b\}$

$$X_0 = \emptyset$$

$$X_1 = \{a\}$$

$$X_2 = \{a, b\}$$

$$X = \{a, b\} \neq Y$$

$$T_P X_0 = \{a\}$$

$$T_P X_1 = \{a, b\}$$

$$T_P X_2 = \{a, b\}$$

What is the meaning of a logic program?

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let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

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if $X = Y$ **then return** X **else fail**

- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{b\}$ **✗**

$$\begin{array}{llll} X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X = \{a, b\} \neq Y \\ T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b\} & \end{array}$$

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- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{a, b\}$

What is the meaning of a logic program?

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- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{a, b\}$

$$\begin{array}{llll} X_0 = \emptyset & X_1 = \{a\} & X_2 = \{a, b\} & X = \{a, b\} = Y \\ T_P X_0 = \{a\} & T_P X_1 = \{a, b\} & T_P X_2 = \{a, b\} & \end{array}$$

What is the meaning of a logic program?

- Procedural attempt patched

guess Y

let $T_P X = \{h(r) \mid r \in P, B(r)^+ \subseteq X, B(r)^- \cap Y = \emptyset\}$ **in**

$X := \emptyset$

while $T_P X \neq X$

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- Example $P = \{a \leftarrow, b \leftarrow a, \neg c\}$ $Y = \{a, b\}$ ✓

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- Procedural attempt patched — interesting ...

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1 guess Y

2 evaluate $B(r)^-$ of each rule $r \in P$ wrt to Y

if $B(r)^- \cap Y \neq \emptyset$ drop r

if $B(r)^- \cap Y = \emptyset$ replace $B(r)$ by $B(r)^+$ from r

3 if $Y = Cn(P')$ for the resulting (positive) program P' then ✓ else ✗

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What is the meaning of a logic program?

- The reduct, P^X , of a program P relative to a set X of atoms is

$$P^X = \{h(r) \leftarrow B(r)^+ \mid r \in P, B(r)^- \cap X = \emptyset\}$$

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- 1 deleting each rule r satisfying $B(r)^- \cap X \neq \emptyset$
and then
- 2 deleting all negative body literals from the remaining rules

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Only negative body literals are evaluated!

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What is the meaning of a logic program?

Stable models!

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 - $Cn(P^X)$ is the \subseteq -smallest set closed under P^X
 - Each atom in X is justified by a proof from P^X

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- Procedural counterpart

guess Y

let $T_{P^Y}X = \{h(r) \mid r \in P^Y, B(r)^+ \subseteq X\}$ **in**

$X := \emptyset$

while $T_{P^Y}X \neq X$

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if $X = Y$ **then return** X **else fail**

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Example one

■ $P = \{p \leftarrow p, q \leftarrow \neg p\}$

X	P^X	$Cn(P^X)$
$\{\}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$
$\{p\}$	$p \leftarrow p$	\emptyset
$\{q\}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$
$\{p, q\}$	$p \leftarrow p$	\emptyset



stable model



no stable model

Example one

■ $P = \{p \leftarrow p, q \leftarrow \neg p\}$

X	P^X	$Cn(P^X)$
$\{ \}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$ ✗
$\{p\}$	$p \leftarrow p$	\emptyset ✗
$\{q\}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$ ✓
$\{p, q\}$	$p \leftarrow p$	\emptyset ✗



stable model



no stable model

Example one

$$\blacksquare P = \{p \leftarrow p, q \leftarrow \neg p\}$$

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$$\blacksquare P = \{p \leftarrow p, q \leftarrow \neg p\}$$

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$\{ \}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$ ✗
$\{p\}$	$p \leftarrow p$	\emptyset ✓
$\{q\}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$ ✓
$\{p, q\}$	$p \leftarrow p$	\emptyset ✓



model



no model

Example two

■ $P = \{p \leftarrow \neg q, q \leftarrow \neg p\}$

X	P^X	$Cn(P^X)$
$\{\quad\}$	$p \leftarrow$ $q \leftarrow$	$\{p, q\}$
$\{p\}$	$p \leftarrow$	$\{p\}$
$\{q\}$	$q \leftarrow$	$\{q\}$
$\{p, q\}$		\emptyset



stable model



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Example two

$$\blacksquare P = \{p \leftarrow \neg q, q \leftarrow \neg p\}$$

X	P^X	$Cn(P^X)$
$\{ \}$	$p \leftarrow$ $q \leftarrow$	$\{p, q\}$ ✗
$\{p\}$	$p \leftarrow$	$\{p\}$ ✓
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$\{q\}$	$q \leftarrow$	$\{q\}$ ✓
$\{p, q\}$		\emptyset ✗



stable model



no stable model

Example two

$$\blacksquare P = \{p \leftarrow \neg q, q \leftarrow \neg p\}$$

X	P^X	$Cn(P^X)$
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model



no model

Example three

■ $P = \{p \leftarrow \neg p\}$

\mathcal{X}	$P^{\mathcal{X}}$	$Cn(P^{\mathcal{X}})$
$\{\}$	$p \leftarrow$	$\{p\}$
$\{p\}$		\emptyset



stable model



no stable model

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stable model



no stable model

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stable model



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stable model



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stable model



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model



no model

Some properties

- A logic program may have zero, one, or multiple stable models
- If X is a stable model of a logic program P ,
then $X \subseteq H(P)$
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Exemplars

Logic program	Stable models
<code>a.</code>	$\{a\}$
<code>a :- b.</code>	$\{\}$
<code>a :- b. b.</code>	$\{a, b\}$
<code>a :- b. b :- a.</code>	$\{\}$
<code>a :- not c.</code>	$\{a\}$
<code>a :- not c. c.</code>	$\{c\}$
<code>a :- not c. c :- not a.</code>	$\{a\}, \{c\}$
<code>a :- not a.</code>	

Outline

13 Syntax

14 Semantics

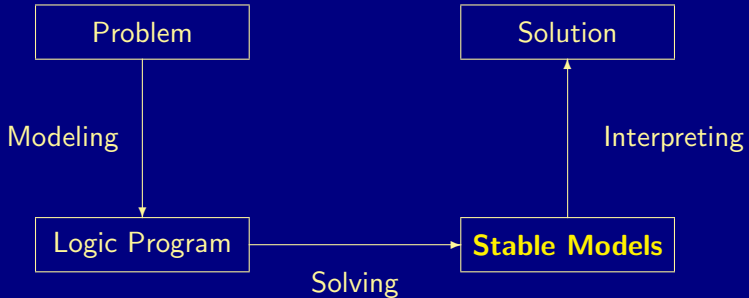
15 Reasoning

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17 Variables

18 Summary

Semantics



Reasoning modes

- Satisfiability
- Enumeration[†]
- Projection[†]
- Intersection[‡]
- Union[‡]
- Optimization
- and combinations of them

[†] without solution recording

[‡] without solution enumeration

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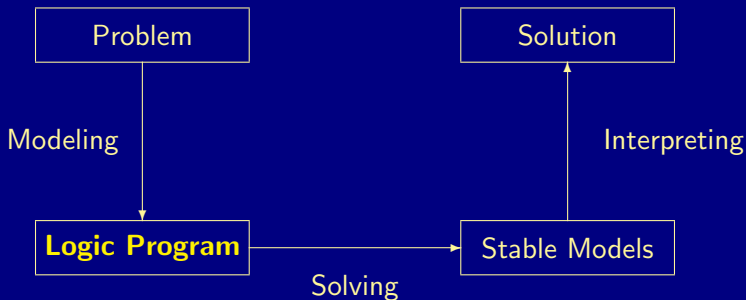
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Language constructs

- Facts `q(42).`
- Rules `p(X) :- q(X), not r(X).`
- Conditional literals `p :- q(X) : r(X).`
- Disjunction `p(X) ; q(X) :- r(X).`
- Integrity constraints `:- q(X), p(X).`
- Choice `2 { p(X,Y) : q(X) } 7 :- r(Y).`
- Aggregates `s(Y) :- r(Y), 2 #sum{ X : p(X,Y), q(X) } 7.`
- Multi-objective optimization `:~ q(X), p(X,C). [C@42]`
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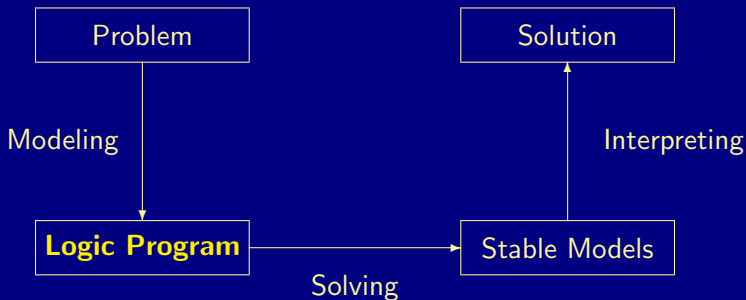
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Ground instantiation

- Let \mathcal{T} be a set of (variable-free) terms
- Let \mathcal{A} be a set of (variable-free) atoms constructible from \mathcal{T}
- A variable-free atom is also called ground
- Ground instances of a rule r are obtained by replacing all variables in r by elements from \mathcal{T} :

$$\text{ground}(r) = \{r\theta \mid \theta : \text{var}(r) \rightarrow \mathcal{T} \text{ and } \text{var}(r\theta) = \emptyset\}$$

where $\text{var}(r)$ stands for the set of all variables occurring in r and θ is a (ground) substitution

- Ground instantiation of logic program P

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- Let \mathcal{T} be a set of (variable-free) **terms**
 - Examples 42, “coucou”, Zorro, *grandfather(leon)*, $3 + X$
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Ground instantiation

- Let \mathcal{T} be a set of variable-free terms (also called **Herbrand universe**)
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$$\text{ground}(P) = \bigcup_{r \in P} \text{ground}(r)$$

An example

$$P = \{ r(a, b) \leftarrow, r(b, c) \leftarrow, t(X, Y) \leftarrow r(X, Y) \}$$

$$\mathcal{T} = \{a, b, c\}$$

$$\mathcal{A} = \left\{ \begin{array}{l} r(a, a), r(a, b), r(a, c), r(b, a), r(b, b), r(b, c), r(c, a), r(c, b), r(c, c), \\ t(a, a), t(a, b), t(a, c), t(b, a), t(b, b), t(b, c), t(c, a), t(c, b), t(c, c) \end{array} \right\}$$

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Grounding aims at reducing the ground instantiation
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by applying semantic principles

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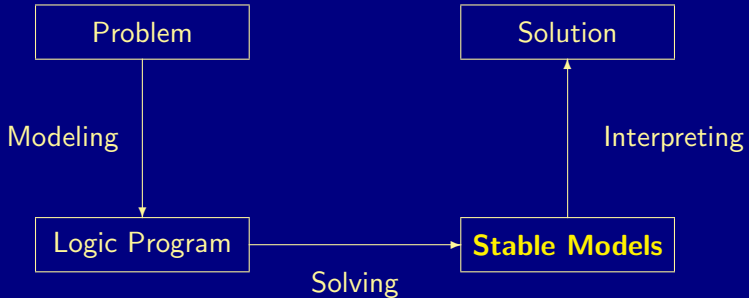
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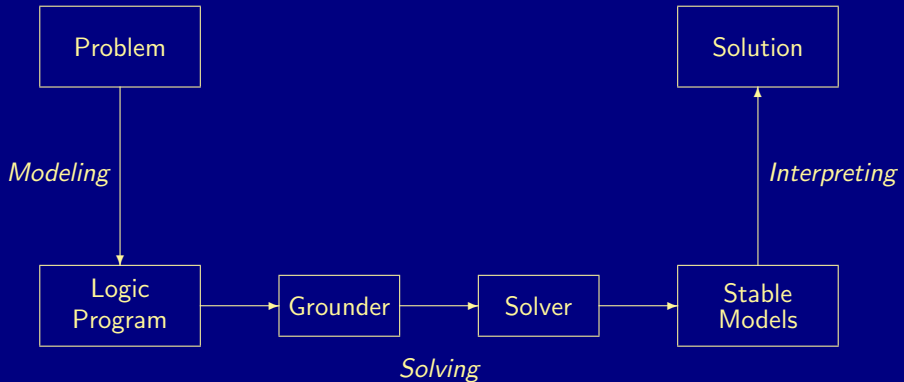
Stable models of programs with Variables

- Let P be a normal logic program with variables
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Modeling, grounding, and solving



Safety

- A normal rule is **safe**, if all its variables occur in its positive body

- Examples

- $p(a) \leftarrow$
- $p(X) \leftarrow$
- $p(X) \leftarrow q(X)$
- $p(X) \leftarrow \neg q(X)$
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Grounding safe programs

■ $P = \{ r(a, b) \leftarrow , r(b, c) \leftarrow , t(X, Y) \leftarrow r(X, Y) \}$

■ Grounding intuitively

1. Partition program along predicate dependencies

$$P_1 = \{ r(a, b) \leftarrow , r(b, c) \leftarrow \}$$

$$P_2 = \{ t(X, Y) \leftarrow r(X, Y) \}$$

Ground P_1

Rules: $\{ r(a, b) \leftarrow , r(b, c) \leftarrow \}$

Atoms: $\{ r(a, b), r(b, c) \}$

Ground P_2 relative to $\{ r(a, b), r(b, c) \}$

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Resulting ground rules

$$\{ r(a, b) \leftarrow , r(b, c) \leftarrow \} \cup \{ t(a, b) \leftarrow r(a, b), t(b, c) \leftarrow r(b, c) \}$$

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Outline

- 13 Syntax
- 14 Semantics
- 15 Reasoning
- 16 Language
- 17 Variables
- 18 Summary

Things to remember

■ Syntax

- atoms, literals, heads, bodies, rules, positive rules, positive and normal **logic programs**
- terms, variables, safety, ground terms, atoms, and rules
- integrity constraints, cardinality and weight constraints, optimization statements

■ Semantics

- assignments/solutions, interpretations/models, sets/closed sets
- T_P operator, fixpoint
- closure, reduct, **stable models**

ASP's syntax and semantics in a nutshell

■ Syntax

- A **logic program**, P , over a set \mathcal{A} of atoms is a finite **set** of rules
- A (normal) **rule**, r , is of the form

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

where $0 \leq m \leq n$ and each $a_i \in \mathcal{A}$ is an **atom** for $0 \leq i \leq n$

■ Semantics

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

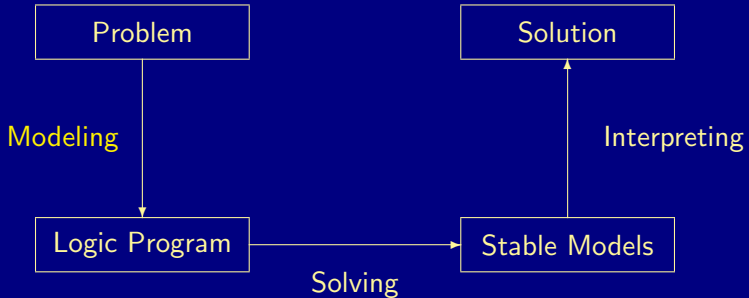
$$P^X = \{h(r) \leftarrow B(r)^+ \mid r \in P, B(r)^- \cap X = \emptyset\}$$

- A set X of atoms is a **stable model** of a program P if $Cn(P^X) = X$

Basic Modeling: Overview

- 19 Elaboration tolerance
- 20 ASP workflow
- 21 Methodology
- 22 Case studies
- 23 Summary

Modeling



Outline

19 Elaboration tolerance

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Guiding principle

- Elaboration Tolerance (McCarthy, 1998)

*“A formalism is **elaboration tolerant** [if] it is convenient to modify a set of facts expressed in the formalism to take into account new phenomena or changed circumstances.”*

- Uniform problem representation

For solving a problem instance I of a problem class C ,

- I is represented as a set of facts P_I ,
- C is represented as a set of rules P_C , and
- P_C can be used to solve all problem instances in C

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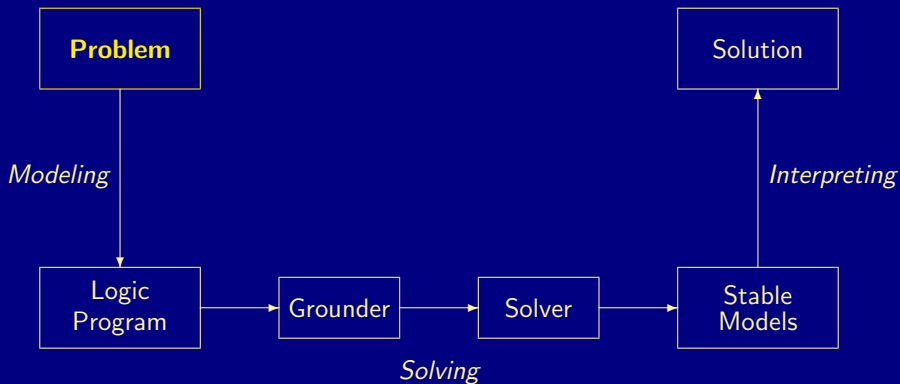
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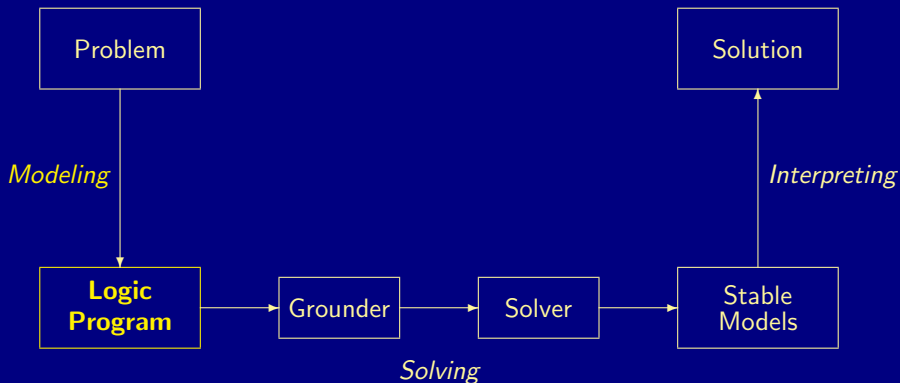
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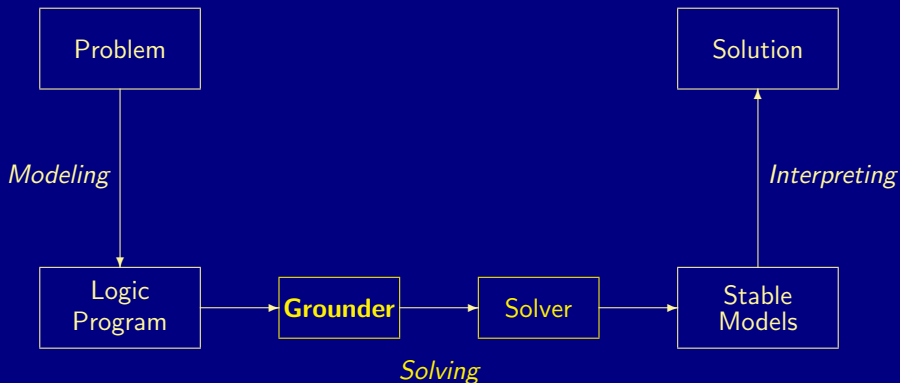
ASP workflow



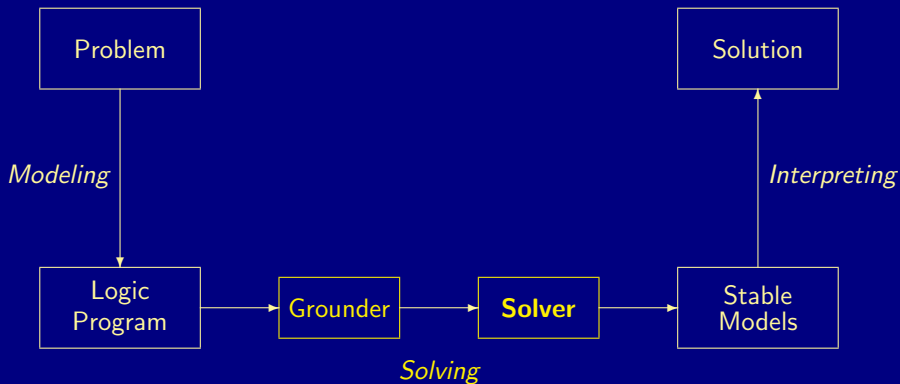
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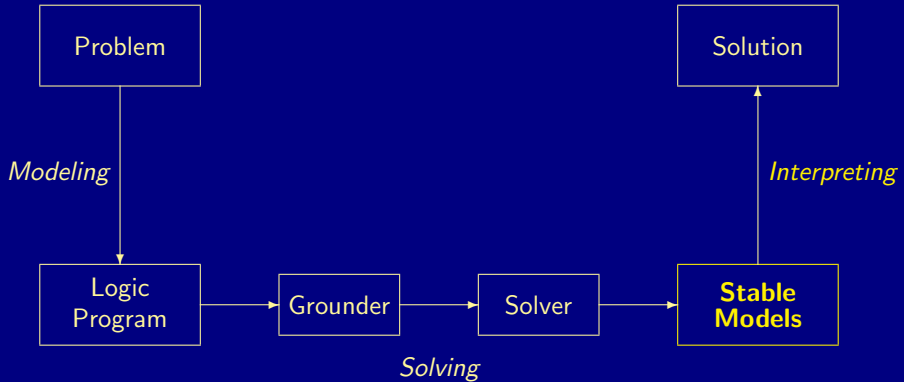
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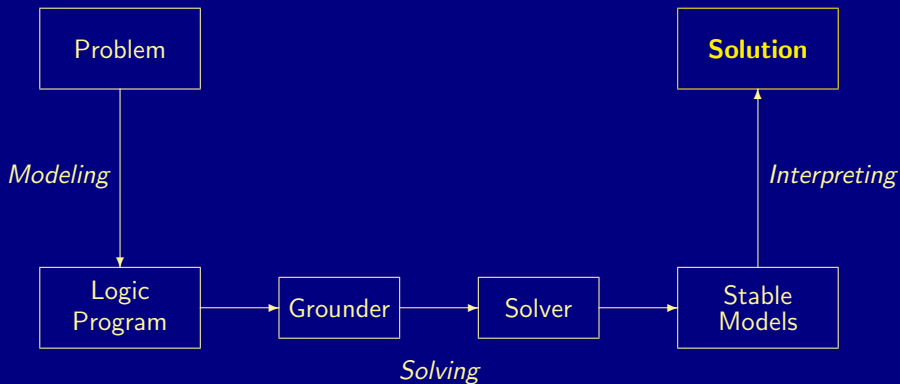
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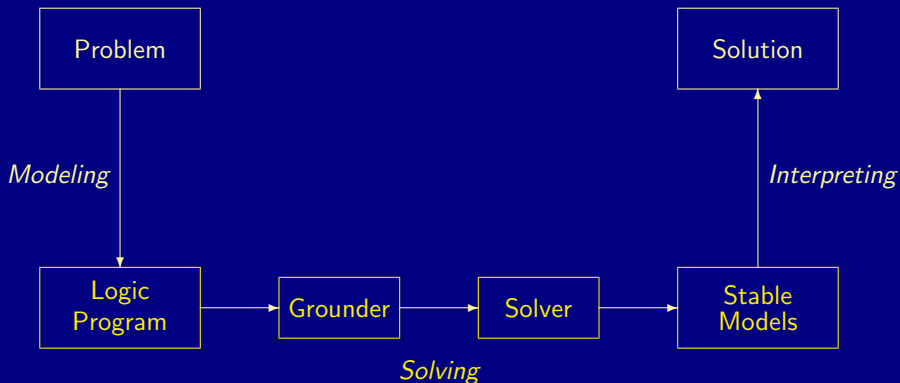
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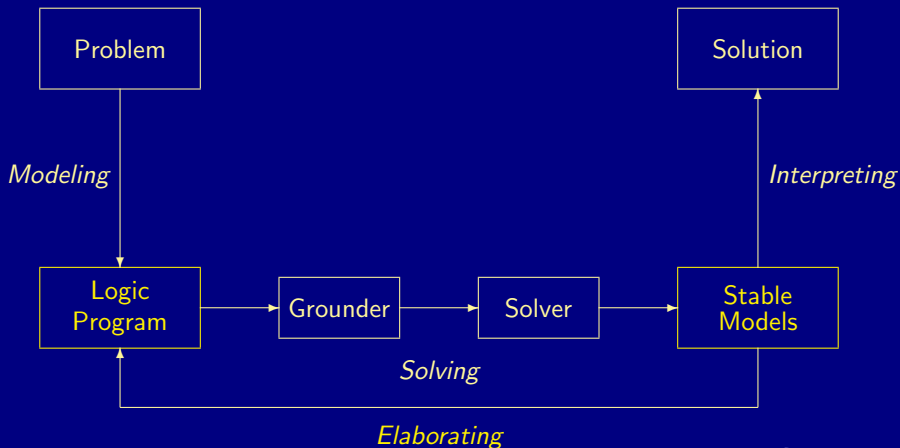
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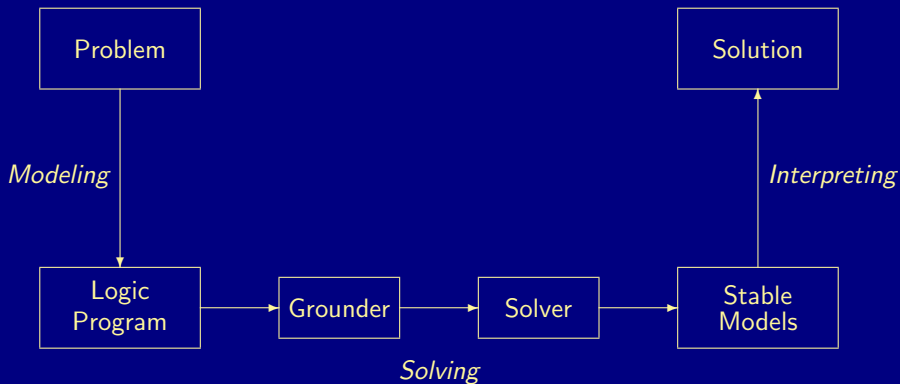
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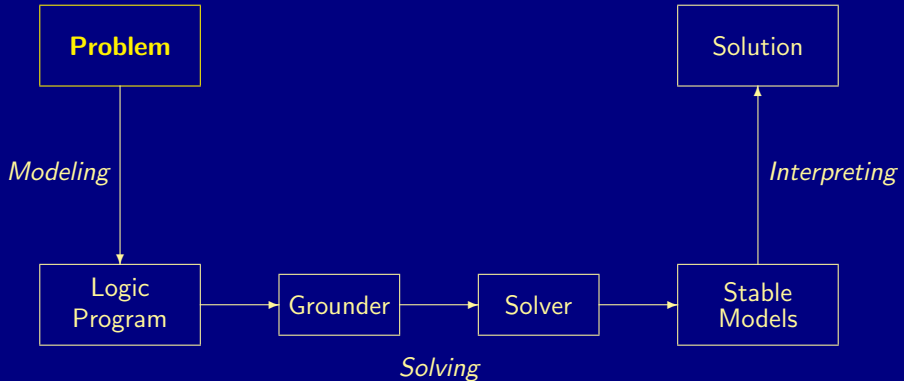
ASP workflow



ASP workflow



ASP workflow: Problem



A case-study: Graph coloring

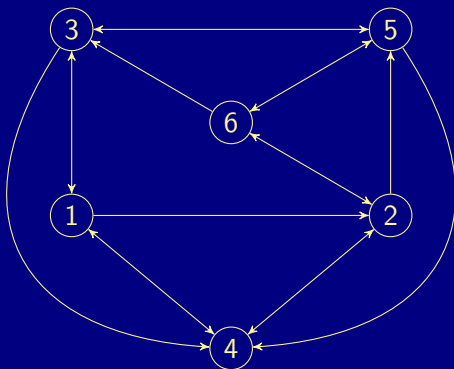
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A case-study: Graph coloring

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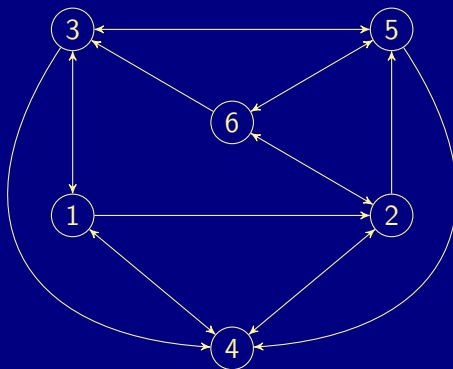
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- Problem class Assign each node one color such that no two nodes connected by an edge have the same color

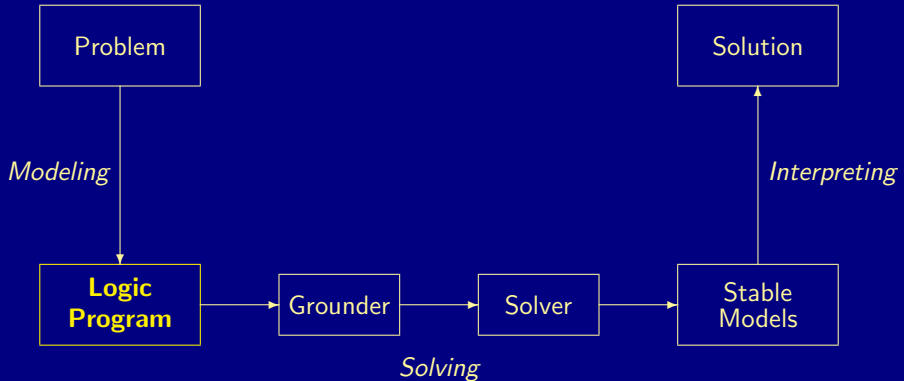
A case-study: Graph coloring

- **Problem instance** A graph consisting of nodes and edges
 - facts formed by predicates `node/1` and `edge/2`
 - facts formed by predicate `color/1`
- **Problem class** Assign each node one color such that no two nodes connected by an edge have the same color

In other words,

- 1 Each node has one color
- 2 Two connected nodes must not have the same color

ASP workflow: Problem representation



Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
edge(2,4).  edge(2,5).  edge(2,6).
edge(3,1).  edge(3,4).  edge(3,5).
edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).    color(b).    color(g).
```

Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

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edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).   color(b).   color(g).
```

Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
edge(2,4).  edge(2,5).  edge(2,6).
edge(3,1).  edge(3,4).  edge(3,5).
edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).    color(b).    color(g).
```

Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
edge(2,4).  edge(2,5).  edge(2,6).
edge(3,1).  edge(3,4).  edge(3,5).
edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).    color(b).    color(g).
```

Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
edge(2,4).  edge(2,5).  edge(2,6).
edge(3,1).  edge(3,4).  edge(3,5).
edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).  color(b).  color(g).
```

**Problem
instance**

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
```

```
edge(2,4).  edge(2,5).  edge(2,6).
```

```
edge(3,1).  edge(3,4).  edge(3,5).
```

```
edge(4,1).  edge(4,2).
```

```
edge(5,3).  edge(5,4).  edge(5,6).
```

```
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).   color(b).   color(g).
```

Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
```

```
edge(2,4).  edge(2,5).  edge(2,6).
```

```
edge(3,1).  edge(3,4).  edge(3,5).
```

```
edge(4,1).  edge(4,2).
```

```
edge(5,3).  edge(5,4).  edge(5,6).
```

```
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).   color(b).   color(g).
```

Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
edge(2,4).  edge(2,5).  edge(2,6).
edge(3,1).  edge(3,4).  edge(3,5).
edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).    color(b).    color(g).
```

} Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

} **Problem
encoding**

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
```

```
edge(2,4).  edge(2,5).  edge(2,6).
```

```
edge(3,1).  edge(3,4).  edge(3,5).
```

```
edge(4,1).  edge(4,2).
```

```
edge(5,3).  edge(5,4).  edge(5,6).
```

```
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).   color(b).   color(g).
```

Problem
instance

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Problem
encoding

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
edge(2,4).  edge(2,5).  edge(2,6).
edge(3,1).  edge(3,4).  edge(3,5).
edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).  color(b).  color(g).
```

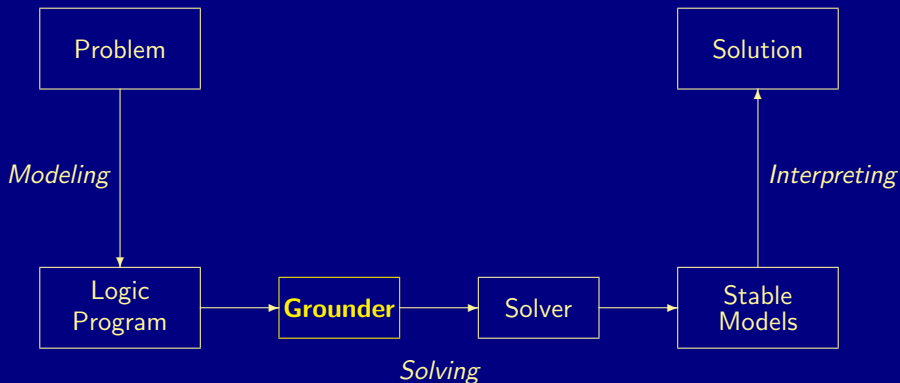
} graph.lp

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

} color.lp

ASP workflow: Grounding



Graph coloring: Grounding

```
$ gringo --text graph.lp color.lp
```

```
node(1). node(2). node(3). node(4). node(5). node(6).
```

```
edge(1,2). edge(2,4). edge(3,1). edge(4,1). edge(5,3). edge(6,2).
edge(1,3). edge(2,5). edge(3,4). edge(4,2). edge(5,4). edge(6,3).
edge(1,4). edge(2,6). edge(3,5). edge(5,6). edge(6,5).
```

```
color(r). color(b). color(g).
```

```
{assign(1,r); assign(1,b); assign(1,g)} = 1. {assign(4,r); assign(4,b); assign(4,g)} = 1.
{assign(2,r); assign(2,b); assign(2,g)} = 1. {assign(5,r); assign(5,b); assign(5,g)} = 1.
{assign(3,r); assign(3,b); assign(3,g)} = 1. {assign(6,r); assign(6,b); assign(6,g)} = 1.
```

```
:- assign(1,r), assign(2,r). :- assign(2,r), assign(4,r). [...] :- assign(6,r), assign(2,r).
:- assign(1,b), assign(2,b). :- assign(2,b), assign(4,b). :- assign(6,b), assign(2,b).
:- assign(1,g), assign(2,g). :- assign(2,g), assign(4,g). :- assign(6,g), assign(2,g).
:- assign(1,r), assign(3,r). :- assign(2,r), assign(5,r). :- assign(6,r), assign(3,r).
:- assign(1,b), assign(3,b). :- assign(2,b), assign(5,b). :- assign(6,b), assign(3,b).
:- assign(1,g), assign(3,g). :- assign(2,g), assign(5,g). :- assign(6,g), assign(3,g).
:- assign(1,r), assign(4,r). :- assign(2,r), assign(6,r). :- assign(6,r), assign(5,r).
:- assign(1,b), assign(4,b). :- assign(2,b), assign(6,b). :- assign(6,b), assign(5,b).
:- assign(1,g), assign(4,g). :- assign(2,g), assign(6,g). :- assign(6,g), assign(5,g).
```

Graph coloring: Grounding

```
$ gringo --text graph.lp color.lp
```

```
node(1). node(2). node(3). node(4). node(5). node(6).
```

```
edge(1,2). edge(2,4). edge(3,1). edge(4,1). edge(5,3). edge(6,2).
edge(1,3). edge(2,5). edge(3,4). edge(4,2). edge(5,4). edge(6,3).
edge(1,4). edge(2,6). edge(3,5). edge(5,6). edge(6,5).
```

```
color(r). color(b). color(g).
```

```
{ assign(1,r); assign(1,b); assign(1,g) } = 1. { assign(4,r); assign(4,b); assign(4,g) } = 1.
{ assign(2,r); assign(2,b); assign(2,g) } = 1. { assign(5,r); assign(5,b); assign(5,g) } = 1.
{ assign(3,r); assign(3,b); assign(3,g) } = 1. { assign(6,r); assign(6,b); assign(6,g) } = 1.
```

```
:- assign(1,r), assign(2,r). :- assign(2,r), assign(4,r). [...] :- assign(6,r), assign(2,r).
:- assign(1,b), assign(2,b). :- assign(2,b), assign(4,b). :- assign(6,b), assign(2,b).
:- assign(1,g), assign(2,g). :- assign(2,g), assign(4,g). :- assign(6,g), assign(2,g).
:- assign(1,r), assign(3,r). :- assign(2,r), assign(5,r). :- assign(6,r), assign(3,r).
:- assign(1,b), assign(3,b). :- assign(2,b), assign(5,b). :- assign(6,b), assign(3,b).
:- assign(1,g), assign(3,g). :- assign(2,g), assign(5,g). :- assign(6,g), assign(3,g).
:- assign(1,r), assign(4,r). :- assign(2,r), assign(6,r). :- assign(6,r), assign(5,r).
:- assign(1,b), assign(4,b). :- assign(2,b), assign(6,b). :- assign(6,b), assign(5,b).
:- assign(1,g), assign(4,g). :- assign(2,g), assign(6,g). :- assign(6,g), assign(5,g).
```

Graph coloring: Grounding

```
$ gringo --text graph.lp color.lp
```

```
node(1). node(2). node(3). node(4). node(5). node(6).
```

```
edge(1,2). edge(2,4). edge(3,1). edge(4,1). edge(5,3). edge(6,2).
edge(1,3). edge(2,5). edge(3,4). edge(4,2). edge(5,4). edge(6,3).
edge(1,4). edge(2,6). edge(3,5). edge(5,6). edge(6,5).
```

```
color(r). color(b). color(g).
```

```
{assign(1,r); assign(1,b); assign(1,g)} = 1. {assign(4,r); assign(4,b); assign(4,g)} = 1.
{assign(2,r); assign(2,b); assign(2,g)} = 1. {assign(5,r); assign(5,b); assign(5,g)} = 1.
{assign(3,r); assign(3,b); assign(3,g)} = 1. {assign(6,r); assign(6,b); assign(6,g)} = 1.
```

```
:- assign(1,r), assign(2,r). :- assign(2,r), assign(4,r). [...] :- assign(6,r), assign(2,r).
:- assign(1,b), assign(2,b). :- assign(2,b), assign(4,b). :- assign(6,b), assign(2,b).
:- assign(1,g), assign(2,g). :- assign(2,g), assign(4,g). :- assign(6,g), assign(2,g).
:- assign(1,r), assign(3,r). :- assign(2,r), assign(5,r). :- assign(6,r), assign(3,r).
:- assign(1,b), assign(3,b). :- assign(2,b), assign(5,b). :- assign(6,b), assign(3,b).
:- assign(1,g), assign(3,g). :- assign(2,g), assign(5,g). :- assign(6,g), assign(3,g).
:- assign(1,r), assign(4,r). :- assign(2,r), assign(6,r). :- assign(6,r), assign(5,r).
:- assign(1,b), assign(4,b). :- assign(2,b), assign(6,b). :- assign(6,b), assign(5,b).
:- assign(1,g), assign(4,g). :- assign(2,g), assign(6,g). :- assign(6,g), assign(5,g).
```


Graph coloring: Grounding

```
$ gringo --text graph.lp color.lp
```

```
node(1). node(2). node(3). node(4). node(5). node(6).
```

```
edge(1,2). edge(2,4). edge(3,1). edge(4,1). edge(5,3). edge(6,2).
edge(1,3). edge(2,5). edge(3,4). edge(4,2). edge(5,4). edge(6,3).
edge(1,4). edge(2,6). edge(3,5). edge(5,6). edge(6,5).
```

```
color(r). color(b). color(g).
```

```
{ assign(1,r); assign(1,b); assign(1,g) } = 1. { assign(4,r); assign(4,b); assign(4,g) } = 1.
{ assign(2,r); assign(2,b); assign(2,g) } = 1. { assign(5,r); assign(5,b); assign(5,g) } = 1.
{ assign(3,r); assign(3,b); assign(3,g) } = 1. { assign(6,r); assign(6,b); assign(6,g) } = 1.
```

```
:- assign(1,r), assign(2,r). :- assign(2,r), assign(4,r). [...] :- assign(6,r), assign(2,r).
:- assign(1,b), assign(2,b). :- assign(2,b), assign(4,b). :- assign(6,b), assign(2,b).
:- assign(1,g), assign(2,g). :- assign(2,g), assign(4,g). :- assign(6,g), assign(2,g).
:- assign(1,r), assign(3,r). :- assign(2,r), assign(5,r). :- assign(6,r), assign(3,r).
:- assign(1,b), assign(3,b). :- assign(2,b), assign(5,b). :- assign(6,b), assign(3,b).
:- assign(1,g), assign(3,g). :- assign(2,g), assign(5,g). :- assign(6,g), assign(3,g).
:- assign(1,r), assign(4,r). :- assign(2,r), assign(6,r). :- assign(6,r), assign(5,r).
:- assign(1,b), assign(4,b). :- assign(2,b), assign(6,b). :- assign(6,b), assign(5,b).
:- assign(1,g), assign(4,g). :- assign(2,g), assign(6,g). :- assign(6,g), assign(5,g).
```

Graph coloring: Grounding

```
$ clingo --text graph.lp color.lp
```

```
node(1). node(2). node(3). node(4). node(5). node(6).
```

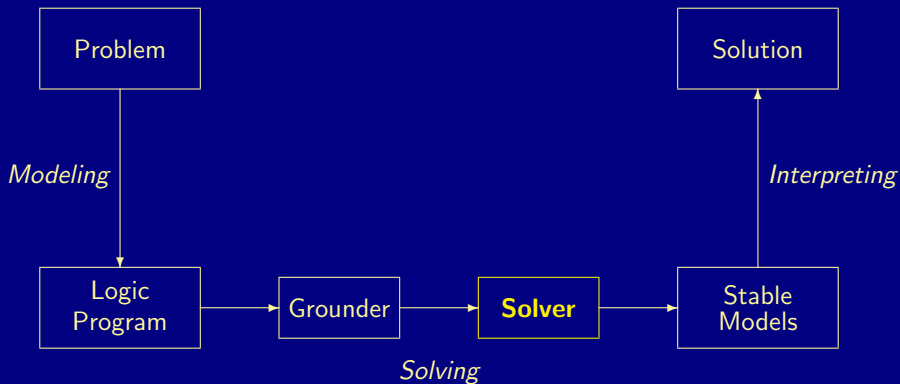
```
edge(1,2). edge(2,4). edge(3,1). edge(4,1). edge(5,3). edge(6,2).
edge(1,3). edge(2,5). edge(3,4). edge(4,2). edge(5,4). edge(6,3).
edge(1,4). edge(2,6). edge(3,5). edge(5,6). edge(6,5).
```

```
color(r). color(b). color(g).
```

```
{ assign(1,r); assign(1,b); assign(1,g) } = 1. { assign(4,r); assign(4,b); assign(4,g) } = 1.
{ assign(2,r); assign(2,b); assign(2,g) } = 1. { assign(5,r); assign(5,b); assign(5,g) } = 1.
{ assign(3,r); assign(3,b); assign(3,g) } = 1. { assign(6,r); assign(6,b); assign(6,g) } = 1.
```

```
:- assign(1,r), assign(2,r). :- assign(2,r), assign(4,r). [...] :- assign(6,r), assign(2,r).
:- assign(1,b), assign(2,b). :- assign(2,b), assign(4,b). :- assign(6,b), assign(2,b).
:- assign(1,g), assign(2,g). :- assign(2,g), assign(4,g). :- assign(6,g), assign(2,g).
:- assign(1,r), assign(3,r). :- assign(2,r), assign(5,r). :- assign(6,r), assign(3,r).
:- assign(1,b), assign(3,b). :- assign(2,b), assign(5,b). :- assign(6,b), assign(3,b).
:- assign(1,g), assign(3,g). :- assign(2,g), assign(5,g). :- assign(6,g), assign(3,g).
:- assign(1,r), assign(4,r). :- assign(2,r), assign(6,r). :- assign(6,r), assign(5,r).
:- assign(1,b), assign(4,b). :- assign(2,b), assign(6,b). :- assign(6,b), assign(5,b).
:- assign(1,g), assign(4,g). :- assign(2,g), assign(6,g). :- assign(6,g), assign(5,g).
```

ASP workflow: Solving



Graph coloring: Solving

```
$ gringo graph.lp color.lp | clasp 0
```

```
clasp version 2.1.0
Reading from stdin
Solving...
Answer: 1
node(1) [...] assign(6,b) assign(5,g) assign(4,b) assign(3,r) assign(2,r) assign(1,g)
Answer: 2
node(1) [...] assign(6,r) assign(5,g) assign(4,r) assign(3,b) assign(2,b) assign(1,g)
Answer: 3
node(1) [...] assign(6,g) assign(5,b) assign(4,g) assign(3,r) assign(2,r) assign(1,b)
Answer: 4
node(1) [...] assign(6,r) assign(5,b) assign(4,r) assign(3,g) assign(2,g) assign(1,b)
Answer: 5
node(1) [...] assign(6,g) assign(5,r) assign(4,g) assign(3,b) assign(2,b) assign(1,r)
Answer: 6
node(1) [...] assign(6,b) assign(5,r) assign(4,b) assign(3,g) assign(2,g) assign(1,r)
SATISFIABLE
```

```
Models      : 6
Time        : 0.002s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
```

Graph coloring: Solving

```
$ gringo graph.lp color.lp | clasp 0
```

```
clasp version 2.1.0
```

```
Reading from stdin
```

```
Solving...
```

```
Answer: 1
```

```
node(1) [...] assign(6,b) assign(5,g) assign(4,b) assign(3,r) assign(2,r) assign(1,g)
```

```
Answer: 2
```

```
node(1) [...] assign(6,r) assign(5,g) assign(4,r) assign(3,b) assign(2,b) assign(1,g)
```

```
Answer: 3
```

```
node(1) [...] assign(6,g) assign(5,b) assign(4,g) assign(3,r) assign(2,r) assign(1,b)
```

```
Answer: 4
```

```
node(1) [...] assign(6,r) assign(5,b) assign(4,r) assign(3,g) assign(2,g) assign(1,b)
```

```
Answer: 5
```

```
node(1) [...] assign(6,g) assign(5,r) assign(4,g) assign(3,b) assign(2,b) assign(1,r)
```

```
Answer: 6
```

```
node(1) [...] assign(6,b) assign(5,r) assign(4,b) assign(3,g) assign(2,g) assign(1,r)
```

```
SATISFIABLE
```

```
Models      : 6
```

```
Time        : 0.002s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
```

Graph coloring: Solving

```
$ clingo graph.lp color.lp 0
```

```
clasp version 2.1.0
```

```
Reading from stdin
```

```
Solving...
```

```
Answer: 1
```

```
node(1) [...] assign(6,b) assign(5,g) assign(4,b) assign(3,r) assign(2,r) assign(1,g)
```

```
Answer: 2
```

```
node(1) [...] assign(6,r) assign(5,g) assign(4,r) assign(3,b) assign(2,b) assign(1,g)
```

```
Answer: 3
```

```
node(1) [...] assign(6,g) assign(5,b) assign(4,g) assign(3,r) assign(2,r) assign(1,b)
```

```
Answer: 4
```

```
node(1) [...] assign(6,r) assign(5,b) assign(4,r) assign(3,g) assign(2,g) assign(1,b)
```

```
Answer: 5
```

```
node(1) [...] assign(6,g) assign(5,r) assign(4,g) assign(3,b) assign(2,b) assign(1,r)
```

```
Answer: 6
```

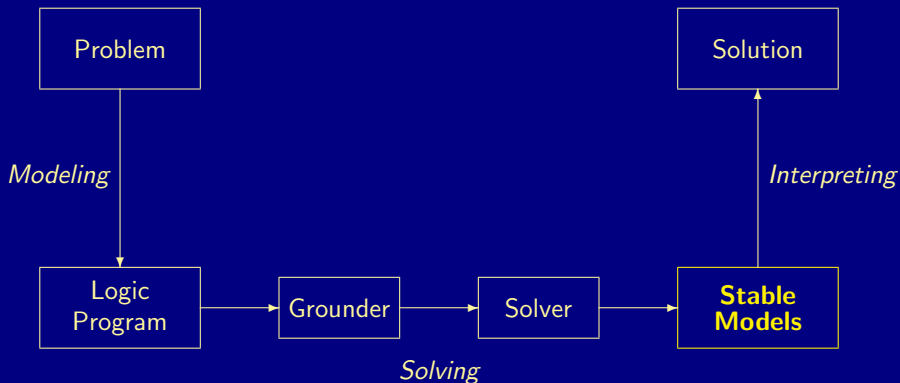
```
node(1) [...] assign(6,b) assign(5,r) assign(4,b) assign(3,g) assign(2,g) assign(1,r)
```

```
SATISFIABLE
```

```
Models      : 6
```

```
Time        : 0.002s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
```

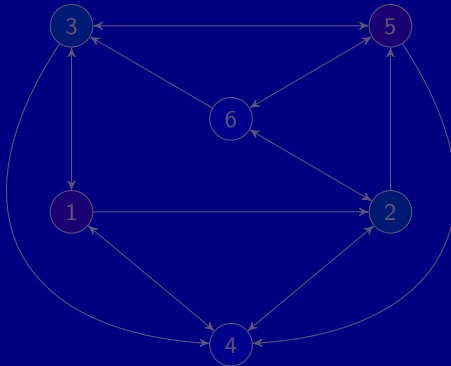
ASP workflow: Stable models



A coloring

Answer: 6

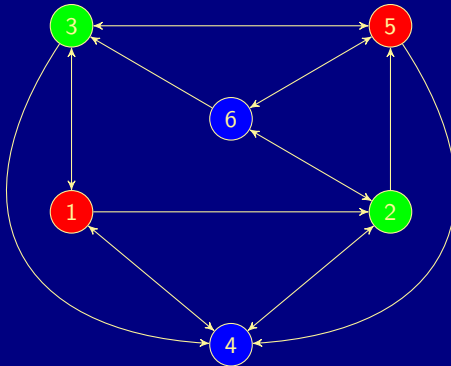
```
node(1)    [...]    \  
assign(6,b) assign(5,r) assign(4,b) assign(3,g) assign(2,g) assign(1,r)
```



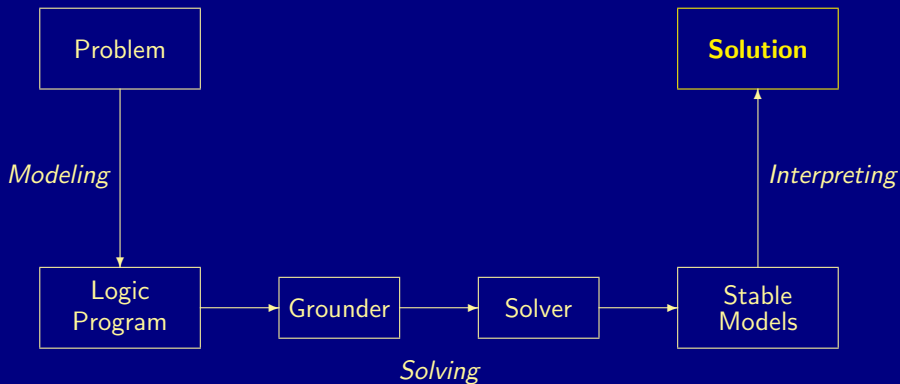
A coloring

Answer: 6

```
node(1)    [...]    \  
assign(6,b) assign(5,r) assign(4,b) assign(3,g) assign(2,g) assign(1,r)
```



ASP workflow: Solutions



Outline

19 Elaboration tolerance

20 ASP workflow

21 Methodology

22 Case studies

23 Summary

Basic methodology

Methodology

Generate and Test (or: Guess and Check)

- Generator Generate potential stable model candidates
(typically through non-deterministic constructs)
- Tester Eliminate invalid candidates
(typically through integrity constraints)

Nutshell

Logic program = Data + Generator + Tester (+ Optimizer)

Basic methodology

Methodology

Generate and Test (or: Guess and Check)

Generator Generate potential stable model candidates
(typically through non-deterministic constructs)

Tester Eliminate invalid candidates
(typically through integrity constraints)

Nutshell

Logic program = Data + Generator + Tester (+ Optimizer)

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
edge(2,4).  edge(2,5).  edge(2,6).
edge(3,1).  edge(3,4).  edge(3,5).
edge(4,1).  edge(4,2).
edge(5,3).  edge(5,4).  edge(5,6).
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).  color(b).  color(g).
```

**Problem
instance**

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

**Problem
encoding**

Graph coloring

```
node(1..6).
```

```
edge(1,2).    edge(1,3).    edge(1,4).
```

```
edge(2,4).    edge(2,5).    edge(2,6).
```

```
edge(3,1).    edge(3,4).    edge(3,5).
```

```
edge(4,1).    edge(4,2).
```

```
edge(5,3).    edge(5,4).    edge(5,6).
```

```
edge(6,2).    edge(6,3).    edge(6,5).
```

```
color(r).    color(b).    color(g).
```

Data

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```

```
:- edge(N,M), assign(N,C), assign(M,C).
```

**Problem
encoding**

Graph coloring

```
node(1..6).
```

```
edge(1,2).  edge(1,3).  edge(1,4).
```

```
edge(2,4).  edge(2,5).  edge(2,6).
```

```
edge(3,1).  edge(3,4).  edge(3,5).
```

```
edge(4,1).  edge(4,2).
```

```
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```

```
edge(6,2).  edge(6,3).  edge(6,5).
```

```
color(r).  color(b).  color(g).
```


Data

```
{ assign(N,C) : color(C) } = 1 :- node(N).
```


Generator

```
:- edge(N,M), assign(N,C), assign(M,C).
```

Tester

Outline

- 19 Elaboration tolerance
- 20 ASP workflow
- 21 Methodology
- 22 Case studies**
- 23 Summary

Outline

- 19 Elaboration tolerance
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 - Satisfiability testing
 - Queens
 - Traveling salesperson
 - Reviewer assignment
 - Planning
- 23 Summary

Satisfiability testing

- **Problem Instance** A propositional formula ϕ in CNF
- **Problem Class** Is there an assignment of propositional variables to true and false such that a given formula ϕ is true
- **Example** Consider formula

$$(a \vee \neg b) \wedge (\neg a \vee b)$$

- **Logic Program**

Generator

Tester

Stable models

$\{a\}.$

$:- \text{not } a, b.$

$X_1 = \{a, b\}$

$\{b\}.$

$:- a, \text{not } b.$

$X_2 = \{\}$

- **Note** The generator puts a and b under the open world assumption
The tester eliminates interpretations; it is expressed negatively

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$$\neg(\neg a \wedge b) \wedge \neg(a \wedge \neg b)$$

- **Logic Program**

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Satisfiability testing

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- **Problem Class** Is there an assignment of propositional variables to true and false such that a given formula ϕ is true
- **Example** Consider formula

$$(\neg a \wedge b \rightarrow \perp) \wedge (a \wedge \neg b \rightarrow \perp)$$

- **Logic Program**

Generator

$\{a\}.$
 $\{b\}.$

Tester

$:- \text{not } a, b.$
 $:- a, \text{not } b.$

Stable models

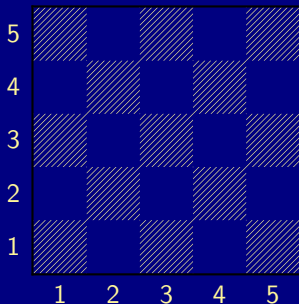
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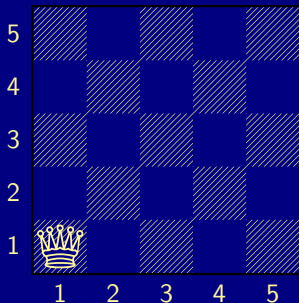
The n-queens problem



- Place n queens on an $n \times n$ chess board
- Queens must not attack one another
- Example $n = 5$



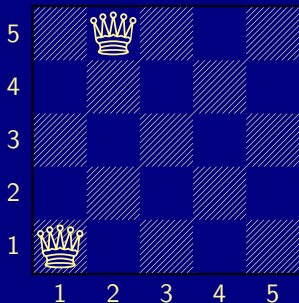
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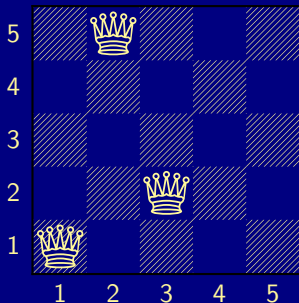
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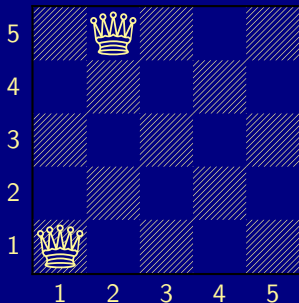
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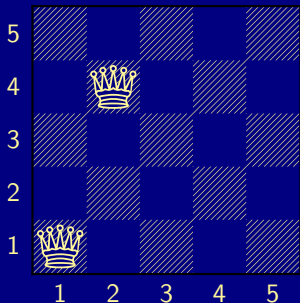
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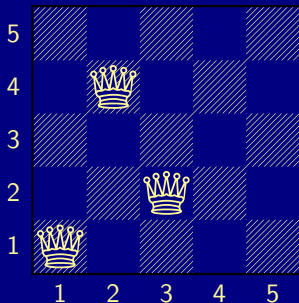
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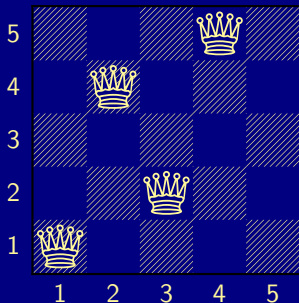
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- Queens must not attack one another
- Example $n = 5$



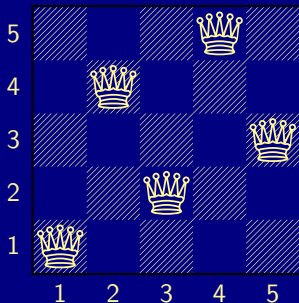
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The n -queens problem



- Place n queens on an $n \times n$ chess board
- Queens must not attack one another
- Example $n = 5$

Defining the field

```
queens.lp
```

```
row(1..n).
```

```
col(1..n).
```

➡ Define the field

- n rows
- n columns

Defining the field

```
queens.lp
```

```
row(1..n).
```

```
col(1..n).
```

➡ Define the field

- `n` rows
- `n` columns

Defining the field

Running ...

```
$ clingo queens.lp --const n=5
Answer: 1
row(1) row(2) row(3) row(4) row(5) \
col(1) col(2) col(3) col(4) col(5)
SATISFIABLE

Models      : 1
Time        : 0.000
```

Placing some queens

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.
```

- ➡ Guess a solution candidate
by placing some queens on the board

Placing some queens

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.
```

- ➡ Guess a solution candidate
by placing some queens on the board

Placing some queens

Running ...

```
$ clingo queens.lp --const n=5 3
```

```
Answer: 1
```

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5)
```

```
Answer: 2
```

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) queen(1,1)
```

```
Answer: 3
```

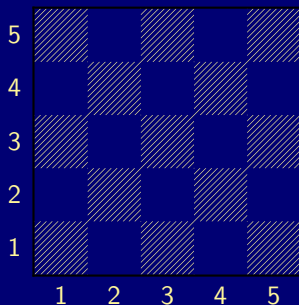
```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) queen(2,1)
```

```
SATISFIABLE
```

```
Models      : 3+
```


Placing some queens

Answer: 1

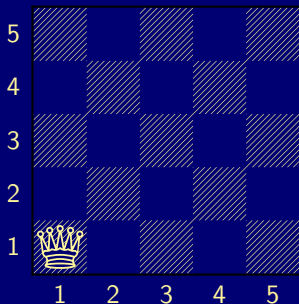


Answer: 1

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5)
```

Placing some queens

Answer: 2

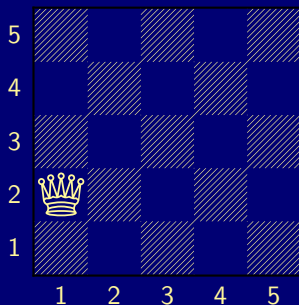


Answer: 2

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) \  
queen(1,1)
```

Placing some queens

Answer: 3



Answer: 3

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) \  
queen(2,1)
```

Placing n queens

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- { queen(I,J) } != n.
```

➡ Place exactly n queens on the board

Placing n queens

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- { queen(I,J) } != n.
```

➡ Place exactly n queens on the board

Placing n queens directly

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) } = n.
```

➡ Place exactly n queens on the board

Placing n queens

Running ...

```
$ clingo queens.lp --const n=5 2
```

```
Answer: 1
```

```
row(1) row(2) row(3) row(4) row(5) \
```

```
col(1) col(2) col(3) col(4) col(5) \
```

```
queen(5,1) queen(4,1) queen(3,1) queen(2,1) queen(1,1)
```

```
Answer: 2
```

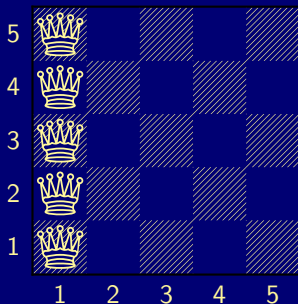
```
row(1) row(2) row(3) row(4) row(5) \
```

```
col(1) col(2) col(3) col(4) col(5) \
```

```
queen(1,2) queen(4,1) queen(3,1) queen(2,1) queen(1,1)
```

Placing n queens

Answer: 1

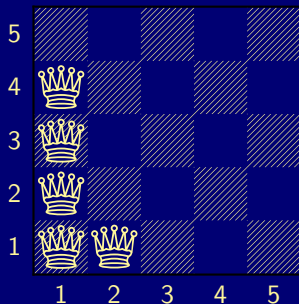


Answer: 1

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) \  
queen(5,1) queen(4,1) queen(3,1) \  
queen(2,1) queen(1,1)
```


Placing n queens

Answer: 2



Answer: 2

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) \  
queen(1,2) queen(4,1) queen(3,1) \  
queen(2,1) queen(1,1)
```

Horizontal and vertical attack

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- { queen(I,J) } != n.  
:- queen(I,J), queen(I,J'), J != J'.  
:- queen(I,J), queen(I',J), I != I'.
```

➡ Forbid horizontal and vertical attacks

Horizontal and vertical attack

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- { queen(I,J) } != n.  
:- queen(I,J), queen(I,J'), J != J'.  
:- queen(I,J), queen(I',J), I != I'.
```

➡ Forbid horizontal and vertical attacks

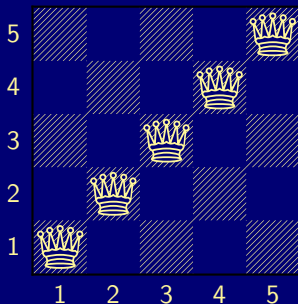
Horizontal and vertical attack

Running ...

```
$ clingo queens.lp --const n=5
Answer: 1
row(1) row(2) row(3) row(4) row(5) \
col(1) col(2) col(3) col(4) col(5) \
queen(5,5) queen(4,4) queen(3,3) queen(2,2) queen(1,1)
```

Horizontal and vertical attack

Answer: 1



Answer: 1

```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) \  
queen(5,5) queen(4,4) queen(3,3) \  
queen(2,2) queen(1,1)
```

Diagonal attack

queens.lp

```
row(1..n).
col(1..n).
{ queen(I,J) : row(I), col(J) }.
:- { queen(I,J) } != n.
:- queen(I,J), queen(I,J'), J != J'.
:- queen(I,J), queen(I',J), I != I'.
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I-J == I'-J'.
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I+J == I'+J'.
```

➡ Forbid diagonal attacks

Diagonal attack

queens.lp

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- { queen(I,J) } != n.  
:- queen(I,J), queen(I,J'), J != J'.  
:- queen(I,J), queen(I',J), I != I'.  
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I-J == I'-J'.  
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I+J == I'+J'.
```

➡ Forbid diagonal attacks

Diagonal attack

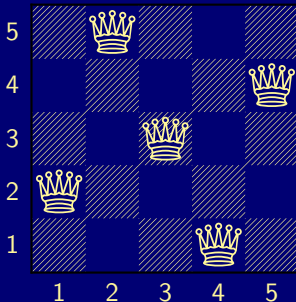
Running ...

```
$ clingo queens.lp --const n=5
Answer: 1
row(1) row(2) row(3) row(4) row(5) \
col(1) col(2) col(3) col(4) col(5) \
queen(4,5) queen(1,4) queen(3,3) queen(5,2) queen(2,1)
SATISFIABLE

Models      : 1+
Time        : 0.000
```


Diagonal attack

Answer: 1



Answer: 1

```

row(1) row(2) row(3) row(4) row(5) \
col(1) col(2) col(3) col(4) col(5) \
queen(4,5) queen(1,4) queen(3,3) \
queen(5,2) queen(2,1)

```

Optimizing

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- { queen(I,J) } != n.  
:- queen(I,J), queen(I,J'), J != J'.  
:- queen(I,J), queen(I',J), I != I'.  
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I-J == I'-J'.  
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I+J == I'+J'.
```

- Encoding can be optimized
- Much faster to solve

Optimizing

```
queens.lp
```

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- { queen(I,J) } != n.  
:- queen(I,J), queen(I,J'), J != J'.  
:- queen(I,J), queen(I',J), I != I'.  
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I-J == I'-J'.  
:- queen(I,J), queen(I',J'), (I,J) != (I',J'), I+J == I'+J'.
```

- Encoding can be optimized
- Much faster to solve

Optimizing

```
queens-opt.lp
```

```
{ queen(I,1..n) } = 1 :- I = 1..n.  
{ queen(1..n,J) } = 1 :- J = 1..n.  
:- { queen(D-J,J) } > 1, D = 2..2*n.  
:- { queen(D+J,J) } > 1, D = 1-n..n-1.
```

- Encoding can be optimized
- Much faster to solve

And sometimes it rocks

```
$ clingo -c n=5000 queens-opt-diag.lp --config=jumpy -q --stats=2
```

```
clingo version 4.1.0
```

```
Solving...
```

```
SATISFIABLE
```

```
Models      : 1+
Time        : 3758.143s (Solving: 1905.22s 1st Model: 1896.20s Unsat: 0.00s)
CPU Time    : 3758.320s

Choices     : 288594554
Conflicts   : 3442   (Analyzed: 3442)
Restarts    : 17     (Average: 202.47 Last: 3442)
Model-Level : 7594728.0
Problems    : 1      (Average Length: 0.00 Splits: 0)
Lemmas      : 3442   (Deleted: 0)
  Binary    : 0      (Ratio: 0.00%)
  Ternary   : 0      (Ratio: 0.00%)
  Conflict  : 3442   (Average Length: 229056.5 Ratio: 100.00%)
  Loop      : 0      (Average Length: 0.0 Ratio: 0.00%)
  Other     : 0      (Average Length: 0.0 Ratio: 0.00%)

Atoms       : 75084857 (Original: 75069989 Auxiliary: 14868)
Rules       : 100129956 (1: 50059992/100090100 2: 39990/29856 3: 10000/10000)
Bodies      : 25090103
Equivalences : 125029999 (Atom=Atom: 50009999 Body=Body: 0 Other: 75020000)
Tight       : Yes
Variables   : 25024868 (Eliminated: 11781 Frozen: 25000000)
Constraints  : 66664   (Binary: 35.6% Ternary: 0.0% Other: 64.4%)

Backjumps   : 3442   (Average: 681.19 Max: 169512 Sum: 2344658)
  Executed   : 3442   (Average: 681.19 Max: 169512 Sum: 2344658 Ratio: 100.00%)
  Bounded    : 0      (Average: 0.00 Max: 0 Sum: 0 Ratio: 0.00%)
```

And sometimes it rocks

```
$ clingo -c n=5000 queens-opt-diag.lp --config=jumpy -q --stats=2
```

```
clingo version 4.1.0
```

```
Solving...
```

```
SATISFIABLE
```

```
Models      : 1+
Time        : 3758.143s (Solving: 1905.22s 1st Model: 1896.20s Unsat: 0.00s)
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  Conflict  : 3442   (Average Length: 229056.5 Ratio: 100.00%)
  Loop      : 0      (Average Length: 0.0 Ratio: 0.00%)
  Other     : 0      (Average Length: 0.0 Ratio: 0.00%)

Atoms       : 75084857 (Original: 75069989 Auxiliary: 14868)
Rules       : 100129956 (1: 50059992/100090100 2: 39990/29856 3: 10000/10000)
Bodies      : 25090103
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Tight       : Yes
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Constraints  : 66664   (Binary: 35.6% Ternary: 0.0% Other: 64.4%)

Backjumps   : 3442   (Average: 681.19 Max: 169512 Sum: 2344658)
  Executed   : 3442   (Average: 681.19 Max: 169512 Sum: 2344658 Ratio: 100.00%)
  Bounded    : 0      (Average: 0.00 Max: 0 Sum: 0 Ratio: 0.00%)
```

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The traveling salesperson problem (TSP)

- Problem Instance A set of cities and distances among them, or simply a weighted graph
- Problem Class What is the shortest possible route visiting each city once and returning to the city of origin?
- Note
 - TSP extends the Hamiltonian cycle problem:
Is there a cycle in a graph visiting each node exactly once
 - TSP is relevant to applications in logistics, planning, chip design, and the core of the vehicle routing problem

The traveling salesperson problem (TSP)

- **Problem Instance** A set of cities and distances among them, or simply a weighted graph
- **Problem Class** What is the shortest possible route visiting each city once and returning to the city of origin?
- **Note**
 - TSP extends the Hamiltonian cycle problem:
Is there a cycle in a graph visiting each node exactly once
 - TSP is relevant to applications in logistics, planning, chip design, and the core of the vehicle routing problem

Traveling salesperson

Problem instance, cities.lp

```
start(a).
```

```
city(a). city(b). city(c). city(d).
```

```
road(a,b,10). road(b,c,20). road(c,d,25). road(d,a,40).  
road(b,d,30). road(d,c,25). road(c,a,35).
```

Traveling salesperson

Problem encoding, tsp.lp

```
{ travel(X,Y) } :- road(X,Y,_).  
  
visited(Y) :- travel(X,Y), start(X).  
visited(Y) :- travel(X,Y), visited(X).  
  
:- city(X), not visited(X).  
  
:- city(X), 2 { travel(X,Y) }.  
:- city(X), 2 { travel(Y,X) }.
```

Traveling salesperson

Problem encoding, tsp.lp

```
{ travel(X,Y) } :- road(X,Y,_).  
  
visited(Y) :- travel(X,Y), start(X).  
visited(Y) :- travel(X,Y), visited(X).  
  
:- city(X), not visited(X).  
  
:- city(X), 2 { travel(X,Y) }.  
:- city(X), 2 { travel(Y,X) }.  
  
:~ travel(X,Y), road(X,Y,D). [D,X,Y]
```

Traveling salesperson

Problem encoding, tsp.lp

```
{ travel(X,Y) } :- road(X,Y,_).  
  
visited(Y) :- travel(X,Y), start(X).  
visited(Y) :- travel(X,Y), visited(X).  
  
:- city(X), not visited(X).  
  
:- city(X), 2 { travel(X,Y) }.  
:- city(X), 2 { travel(Y,X) }.  
  
#minimize { D,X,Y : travel(X,Y), road(X,Y,D) }.
```

Running salesperson

```
$ clingo tsp.lp cities.lp
```

```
clingo version 5.3.1
```

```
Reading...
```

```
Solving...
```

```
Answer: 1
```

```
start(a) [...] road(c,a,35)
```

```
travel(a,b) travel(b,d) travel(d,c) travel(c,a)
```

```
visited(b) visited(c) visited(d) visited(a)
```

```
Optimization: 100
```

```
Answer: 2
```

```
start(a) [...] road(c,a,35)
```

```
travel(a,b) travel(b,d) travel(d,c) travel(c,a)
```

```
visited(b) visited(c) visited(d) visited(a)
```

```
Optimization: 95
```

```
OPTIMUM FOUND
```

```
Models          : 2
```

```
  Optimum       : yes
```

```
Optimization    : 95
```

```
Calls           : 1
```

```
Time            : 0.005s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
```

```
CPU Time        : 0.002s
```

Running salesperson

```
$ clingo tsp.lp cities.lp
```

```
clingo version 5.3.1
```

```
Reading...
```

```
Solving...
```

```
Answer: 1
```

```
start(a) [...] road(c,a,35)
```

```
travel(a,b) travel(b,d) travel(d,c) travel(c,a)
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```
visited(b) visited(c) visited(d) visited(a)
```

```
Optimization: 100
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```
Answer: 2
```

```
start(a) [...] road(c,a,35)
```

```
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```

```
visited(b) visited(c) visited(d) visited(a)
```

```
Optimization: 95
```

```
OPTIMUM FOUND
```

```
Models          : 2
```

```
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```

```
Optimization    : 95
```

```
Calls           : 1
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Running salesperson

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```

```
Reading...
```

```
Solving...
```

```
Answer: 1
```

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```

```
travel(a,b) travel(b,d) travel(d,c) travel(c,a)
```

```
visited(b) visited(c) visited(d) visited(a)
```

```
Optimization: 100
```

```
Answer: 2
```

```
start(a) [...] road(c,a,35)
```

```
travel(a,b) travel(b,c) travel(c,d) travel(d,a)
```

```
visited(b) visited(c) visited(d) visited(a)
```

```
Optimization: 95
```

```
OPTIMUM FOUND
```

```
Models          : 2
```

```
  Optimum       : yes
```

```
Optimization    : 95
```

```
Calls           : 1
```

```
Time            : 0.005s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
```

```
CPU Time        : 0.002s
```


Running salesperson

```
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clingo version 5.3.1
Reading...
Solving...
Answer: 1
start(a) [...] road(c,a,35)
travel(a,b) travel(b,d) travel(d,c) travel(c,a)
visited(b) visited(c) visited(d) visited(a)
Optimization: 100
Answer: 2
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visited(b) visited(c) visited(d) visited(a)
Optimization: 95
OPTIMUM FOUND

Models          : 2
  Optimum       : yes
Optimization    : 95
Calls           : 1
Time            : 0.005s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
CPU Time        : 0.002s
```

Running salesperson

```
$ clingo tsp.lp cities.lp
clingo version 5.3.1
Reading...
Solving...
Answer: 1
start(a) [...] road(c,a,35)
travel(a,b) travel(b,d) travel(d,c) travel(c,a)
visited(b) visited(c) visited(d) visited(a)
Optimization: 100
Answer: 2
start(a) [...] road(c,a,35)
travel(a,b) travel(b,c) travel(c,d) travel(d,a)
visited(b) visited(c) visited(d) visited(a)
Optimization: 95
OPTIMUM FOUND
```

```
Models      : 2
  Optimum   : yes
Optimization : 95
Calls       : 1
Time        : 0.005s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
CPU Time    : 0.002s
```

Traveling salesperson

Alternative problem encoding

```
{ travel(X,Y) : road(X,Y,_) } = 1 :- city(X).  
{ travel(X,Y) : road(X,Y,_) } = 1 :- city(Y).
```

```
visited(Y) :- travel(X,Y), start(X).  
visited(Y) :- travel(X,Y), visited(X).
```

```
:- city(X), not visited(X).
```

```
#minimize { D,X,Y : travel(X,Y), road(X,Y,D) }.
```

Outline

- 19 Elaboration tolerance
- 20 ASP workflow
- 21 Methodology
- 22 Case studies
 - Satisfiability testing
 - Queens
 - Traveling salesperson
 - Reviewer assignment
 - Planning
- 23 Summary

Reviewer Assignment

- Problem Instance A set of papers and a set of reviewers along with their first and second choices of papers and conflict of interests
- Problem Class A nice assignment of three reviewers to each paper

Reviewer Assignment

- Problem Instance A set of papers and a set of reviewers along with their first and second choices of papers and conflict of interests
- Problem Class A “nice” assignment of three reviewers to each paper

Reviewer Assignment

by Ilkka Niemelä

```
paper(p1).  reviewer(r1).  classA(r1,p1).  classB(r1,p2).  coi(r1,p3).  
paper(p2).  reviewer(r2).  classA(r2,p3).  classB(r2,p4).  coi(r2,p6).  
[...]
```

```
{ assigned(P,R) : reviewer(R) } = 3 :-  paper(P).
```

```
:- assigned(P,R), coi(R,P).
```

```
:- assigned(P,R), not classA(R,P), not classB(R,P).
```

```
:- not 6 { assigned(P,R) : paper(P) } 9, reviewer(R).
```

```
assignedB(P,R) :-  classB(R,P), assigned(P,R).
```

```
:- 3 { assignedB(P,R) : paper(P) }, reviewer(R).
```

```
#minimize { 1,P,R : assignedB(P,R), paper(P), reviewer(R) }.
```

Reviewer Assignment

by Ilkka Niemelä

```
paper(p1).  reviewer(r1).  classA(r1,p1).  classB(r1,p2).  coi(r1,p3).  
paper(p2).  reviewer(r2).  classA(r2,p3).  classB(r2,p4).  coi(r2,p6).  
[...]
```

```
{ assigned(P,R) : reviewer(R) } = 3 :-  paper(P).
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```

```
:- not 6 { assigned(P,R) : paper(P) } 9, reviewer(R).
```

```
assignedB(P,R) :-  classB(R,P), assigned(P,R).
```

```
:- 3 { assignedB(P,R) : paper(P) }, reviewer(R).
```

```
#minimize { 1,P,R : assignedB(P,R), paper(P), reviewer(R) }.
```


Reviewer Assignment

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```
paper(p1).  reviewer(r1).  classA(r1,p1).  classB(r1,p2).  coi(r1,p3).  
paper(p2).  reviewer(r2).  classA(r2,p3).  classB(r2,p4).  coi(r2,p6).  
[...]
```

```
{ assigned(P,R) : reviewer(R) } = 3 :-  paper(P).
```

```
:- assigned(P,R), coi(R,P).
```

```
:- assigned(P,R), not classA(R,P), not classB(R,P).
```

```
:- not 6 { assigned(P,R) : paper(P) } 9, reviewer(R).
```

```
assignedB(P,R) :-  classB(R,P), assigned(P,R).
```

```
:- 3 { assignedB(P,R) : paper(P) }, reviewer(R).
```

```
#minimize { 1,P,R : assignedB(P,R), paper(P), reviewer(R) }.
```

Reviewer Assignment

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```
paper(p1).  reviewer(r1).  classA(r1,p1).  classB(r1,p2).  coi(r1,p3).  
paper(p2).  reviewer(r2).  classA(r2,p3).  classB(r2,p4).  coi(r2,p6).  
[...]
```

```
{ assigned(P,R) : reviewer(R) } = 3 :-  paper(P).
```

```
:- assigned(P,R), coi(R,P).
```

```
:- assigned(P,R), not classA(R,P), not classB(R,P).
```

```
:- not 6 { assigned(P,R) : paper(P) } 9, reviewer(R).
```

```
assignedB(P,R) :-  classB(R,P), assigned(P,R).
```

```
:- 3 { assignedB(P,R) : paper(P) }, reviewer(R).
```

```
#minimize { 1,P,R : assignedB(P,R), paper(P), reviewer(R) }.
```

Reviewer Assignment

by Ilkka Niemelä

```
paper(p1).  reviewer(r1).  classA(r1,p1).  classB(r1,p2).  coi(r1,p3).  
paper(p2).  reviewer(r2).  classA(r2,p3).  classB(r2,p4).  coi(r2,p6).  
[...]
```

```
{ assigned(P,R) : reviewer(R) } = 3 :-  paper(P).
```

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```

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:- assigned(P,R), not classA(R,P), not classB(R,P).
```

```
:- not 6 { assigned(P,R) : paper(P) } 9, reviewer(R).
```

```
assignedB(P,R) :-  classB(R,P), assigned(P,R).
```

```
:- 3 { assignedB(P,R) : paper(P) }, reviewer(R).
```

```
#minimize { 1,P,R : assignedB(P,R), paper(P), reviewer(R) }.
```

Reviewer Assignment

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```
paper(p1).  reviewer(r1).  classA(r1,p1).  classB(r1,p2).  coi(r1,p3).  
paper(p2).  reviewer(r2).  classA(r2,p3).  classB(r2,p4).  coi(r2,p6).  
[...]
```

```
#count { P,R : assigned(P,R) , reviewer(R) } = 3 :-  paper(P).
```

```
:- assigned(P,R), coi(R,P).
```

```
:- assigned(P,R), not classA(R,P), not classB(R,P).
```

```
:- not 6 <= #count { P,R : assigned(P,R), paper(P) } <= 9, reviewer(R).
```

```
assignedB(P,R) :-  classB(R,P), assigned(P,R).
```

```
:- 3 <= #count { P,R : assignedB(P,R), paper(P) }, reviewer(R).
```

```
#minimize { 1,P,R : assignedB(P,R), paper(P), reviewer(R) }.
```

Reviewer Assignment

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```
paper(p1).  reviewer(r1).  classA(r1,p1).  classB(r1,p2).  coi(r1,p3).  
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[...]
```

```
#count { P,R : assigned(P,R) , reviewer(R) } = 3 :-  paper(P).
```

```
:- assigned(P,R), coi(R,P).
```

```
:- assigned(P,R), not classA(R,P), not classB(R,P).
```

```
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```

```
assignedB(P,R) :-  classB(R,P), assigned(P,R).
```

```
:- 3 <= #count { P,R : assignedB(P,R), paper(P) }, reviewer(R).
```

```
#minimize { 1,P,R : assignedB(P,R), paper(P), reviewer(R) }.
```

Outline

- 19 Elaboration tolerance
- 20 ASP workflow
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Simplified STRIPS² Planning

■ Problem Instance

- set of fluents
- initial and goal state
- set of actions, consisting of pre- and postconditions
- number k of allowed actions

■ Problem Class Find a plan, that is, a sequence of k actions leading from the initial state to the goal state

■ Example

- fluents $\{p, q, r\}$
- initial state $\{p, \neg q, \neg r\}$
- goal state $\{r\}$
- actions $a = (\{p\}, \{q, \neg p\})$ and $b = (\{q\}, \{r, \neg q\})$
- length 2

plan $\langle a, b \rangle \quad \{p, \neg q, \neg r\} \xrightarrow{a} \{\neg p, q, \neg r\} \xrightarrow{b} \{\neg p, \neg q, r\}$

²Stanford Research Institute Problem Solver, 1971

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- actions $a = (\{p\}, \{q, \neg p\})$ and $b = (\{q\}, \{r, \neg q\})$
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- plan $\langle a, b \rangle \quad \{p, \neg q, \neg r\} \xrightarrow{a} \{\neg p, q, \neg r\} \xrightarrow{b} \{\neg p, \neg q, r\}$

Simplistic STRIPS Planning

Problem instance

```
time(1..k).
```

```
fluent(p).
```

```
fluent(q).
```

```
fluent(r).
```

```
action(a).
```

```
    pre(a,p).
```

```
    add(a,q).
```

```
    del(a,p).
```

```
action(b).
```

```
    pre(b,q).
```

```
    add(b,r).
```

```
    del(b,q).
```

```
init(p).
```

```
query(r).
```

Simplistic STRIPS Planning

Problem encoding

```
holds(P,0) :- init(P).
```

```
{ occ(A,T) : action(A) } = 1 :- time(T).
```

```
:- occ(A,T), pre(A,F), not holds(F,T-1).
```

```
holds(F,T) :- occ(A,T), add(A,F).
```

```
holds(F,T) :- holds(F,T-1), time(T), not occ(A,T) : del(A,F).
```

```
:- query(F), not holds(F,k).
```

Simplistic STRIPS Planning

Solving

```
$ clingo planning-encoding.lp planning-instance.lp -c k=2 0
clingo version 5.5.0
Reading from planning-encoding.lp ...
Solving...
Answer: 1
[...] occ(a,1) occ(b,2)
SATISFIABLE
```

```
Models          : 1
Time            : 0.001s (Solving: 0.00s)
CPU Time       : 0.001s
```

Simplistic STRIPS Planning

Solving

```
$ clingo planning-encoding.lp planning-instance.lp -c k=2 0
clingo version 5.5.0
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Answer: 1
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```
Models      : 1
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CPU Time    : 0.001s
```

Outline

19 Elaboration tolerance

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Things to remember

- Elaboration tolerance, uniform problem representation, problem instance, problem encoding
- Generate and test methodology
- ASP's workflow, modeling, grounding, solving (and optimizing)
- *clingo* = *gringo*+*clasp*+...

Things to remember

- Elaboration tolerance, uniform problem representation, problem instance, problem encoding
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- *clingo* = *gringo*+*clasp*+...

Language: Overview

24 Base language

25 Optimization

26 Formats

27 Summary

Outline

24 Base language

25 Optimization

26 Formats

27 Summary

Outline

- 24 Base language
 - Motivation
 - Integrity constraint
 - Choice rule
 - Cardinality rule
 - Weight rule
 - Conditional literal
- 25 Optimization
- 26 Formats
- 27 Summary

Basic language extensions

- **Fact** The expressiveness of a language can be enhanced by adding interesting language constructs
- **Questions**
 - 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?
- **Answer**
 - A way of providing semantics is to furnish a translation removing the new constructs
 - This translation might also be used for implementing the language extension

Basic language extensions

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Outline

- 24 Base language
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Integrity constraint

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

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

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

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- Example

```
:- edge(3,7), color(3,red), color(7,red).
```

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- Example programs

$$\begin{array}{l} \{ a \leftarrow \neg b, b \leftarrow \neg a \} \qquad \{a\} \quad \{b\} \\ \{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow a \} \\ \{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow \neg a \} \end{array}$$

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$$\begin{array}{ll}
 \{ a \leftarrow \neg b, b \leftarrow \neg a \} & \{a\} \quad \{b\} \\
 \{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow a \} & \{b\} \\
 \{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow \neg a \} &
 \end{array}$$

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$\{ a \leftarrow \neg b, b \leftarrow \neg a \}$	$\{a\}$	$\{b\}$
$\{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow a \}$		$\{b\}$
$\{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow \neg a \}$	$\{a\}$	

Embedding in normal rules

- Translation An integrity constraint of form

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

can be translated into the normal rule

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

where x is a new symbol

- Example programs

$$\begin{array}{ll} \{ a \leftarrow \neg b, b \leftarrow \neg a \} & \{a\} \quad \{b\} \\ \{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow a \} & \\ \{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ \leftarrow \neg a \} & \end{array}$$

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- Example programs

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$\{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ x \leftarrow a, \neg x \}$		$\{b\}$
$\{ a \leftarrow \neg b, b \leftarrow \neg a \} \cup \{ x \leftarrow \neg a, \neg x \}$	$\{a\}$	

Outline

- 24 Base language
 - Motivation
 - Integrity constraint
 - Choice rule
 - Cardinality rule
 - Weight rule
 - Conditional literal
- 25 Optimization
- 26 Formats
- 27 Summary

Choice rule

- Purpose Provide choices over subsets of atoms
- Syntax A **choice rule** is of the form

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

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, any subset of $\{a_1, \dots, a_m\}$ can be included in the stable model

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- ```
{ buy(pizza); buy(wine); buy(corn) } :- at(grocery).
```

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

$$\begin{array}{llll} x & \leftarrow & a_{m+1}, \dots, a_n, \neg a_{n+1}, \dots, \neg a_o & \\ a_1 & \leftarrow & x, \neg x_1 & \dots \quad a_m \leftarrow x, \neg x_m \\ x_1 & \leftarrow & \neg a_1 & \dots \quad x_m \leftarrow \neg a_m \end{array}$$

by introducing new atoms  $x, x_1, \dots, x_m$

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- Example program

$$\left\{ \begin{array}{l} x \leftarrow b \\ a \leftarrow x, \neg x_1 \\ x_1 \leftarrow \neg a \end{array} \right\} \cup \{ b \leftarrow \} \quad \{b, x, x_1\} \quad \{a, b, x\}$$

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## Cardinality rule

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

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

where  $0 \leq m \leq n$  and each  $a_i$  is an atom for  $1 \leq i \leq n$   
and  $l$  is a non-negative integer called **lower bound**

- Informal meaning The head belongs to the stable model, if at least  $l$  positive/negative body literals are in/excluded in the stable model

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```
pass(c42) :- 2 { pass(a1); pass(a2); pass(a3) }.
```

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for  $0 \leq k \leq l$  the rules

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

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

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

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## An example

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

$$\begin{array}{ll}
 a \leftarrow x(1, 1) & b \leftarrow \\
 x(1, 2) \leftarrow x(2, 1), b & \\
 x(1, 1) \leftarrow x(2, 1) & \\
 x(2, 2) \leftarrow x(3, 1), c & \\
 x(2, 1) \leftarrow x(3, 1) & \\
 x(1, 1) \leftarrow x(2, 0), b & \\
 x(1, 0) \leftarrow x(2, 0) & \\
 x(2, 1) \leftarrow x(3, 0), c & \\
 x(2, 0) \leftarrow x(3, 0) & \\
 x(3, 0) \leftarrow & 
 \end{array}$$

having stable model  $\{a, b, x(3, 0), x(2, 0), x(1, 0), x(1, 1)\}$

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 \end{array}$$

having stable model  $\{a, b, x(3, 0), x(2, 0), x(1, 0), x(1, 1)\}$

... and vice versa

■ A normal rule

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

can be represented by the cardinality rule

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

## Cardinality rules with upper bounds

- A rule of the form

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

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



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and  $l$  and  $u$  are non-negative integers

stands for

$$a_0 \leftarrow x, \neg y$$

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

$$y \leftarrow u+1 \{ a_1, \dots, a_m, \neg a_{m+1}, \dots, \neg a_n \}$$

where  $x$  and  $y$  are new symbols

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and  $l$  and  $u$  are non-negative integers

- Note The expression in the body of the cardinality rule is referred to as a **cardinality constraint** with lower and upper bound  $l$  and  $u$

## Cardinality constraints as heads

- A rule of the form

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

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

```
1 {color(2,red); color(2,green); color(2,blue)} 1.
```

# Cardinality constraints as heads

- A rule of the form

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

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

and  $l$  and  $u$  are non-negative integers

stands for

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

where  $x$  and  $y$  are new symbols

## 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 each  $l_i \ S_i \ u_i$  is a cardinality constraint for  $0 \leq i \leq n$

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stands for

$$x \leftarrow y_1, \dots, y_n, \neg z_1, \dots, \neg z_n$$

$$S_0^+ \leftarrow x$$

$$\leftarrow x, \neg y_0$$

$$\leftarrow x, z_0$$

$$y_i \leftarrow l_i \ S_i$$

$$z_i \leftarrow u_i + 1 \ S_i$$

where  $x, y_i, z_i$  are new symbols and  $S_0^+$  gives all atoms in  $S_0$

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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 each  $l_i \ S_i \ u_i$  is a cardinality constraint for  $0 \leq i \leq n$   
stands for

$$x \leftarrow y_1, \dots, y_n, \neg z_1, \dots, \neg z_n$$

$$S_0^+ \leftarrow x$$

$$\leftarrow x, \neg y_0$$

$$\leftarrow x, z_0$$

$$y_i \leftarrow l_i \ S_i$$

$$z_i \leftarrow u_i + 1 \ S_i$$

where  $x, y_i, z_i$  are new symbols and  $S_0^+$  gives all atoms in  $S_0$

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# Outline

- 24 Base language
  - Motivation
  - Integrity constraint
  - Choice rule
  - Cardinality rule
  - Weight rule
  - Conditional literal
- 25 Optimization
- 26 Formats
- 27 Summary

## Weight rule

- Purpose Bound (lower) sum of subsets of literal weights
- Syntax A **weighted literal**  $w : k$  associates weight  $w$  with literal  $k$
- Syntax A **weight rule** is the form

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

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

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

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- Informal meaning The head belongs to the stable model, if the sum of weights associated with positive/negative body literals in/excluded in the stable model is at least  $l$

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- Note
  - A cardinality rule is a weight rule where  $w_i = 1$  for  $0 \leq i \leq n$

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- Example

`5 { 4:course(db); 6:course(ai); 3:course(xml) } 10`

# Outline

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  - ➡ The expansion of this collection is context dependent

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  - $r(X) : p(X), \text{not } q(X)$  yields  $r(1)$  and  $r(3)$

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- **Example** Assume 'p(1..3). q(2).'

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The rule

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

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

is instantiated to

$$r(1); r(3) \text{ :- } r(1), r(3), 1 \{ r(1); r(3) \}.$$

# Outline

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# Optimization statement

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

$$\textit{minimize} \{ w_1 @ p_1 : l_{1_1}, \dots, l_{m_1}; \dots; w_n @ p_n : l_{1_n}, \dots, l_{m_n} \}.$$

where each  $l_{j_i}$  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 sum of weights (by descending levels)

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$$\text{maximize } \{ w_1 @ p_1 : l_1, \dots, w_n @ p_n : l_n \}$$

stands for *minimize*  $\{ -w_1 @ p_1 : l_1, \dots, -w_n @ p_n : l_n \}$

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

```
#maximize { 250@1:hd(1); 500@1:hd(2); 750@1:hd(3) }.
#minimize { 30@2:hd(1); 40@2:hd(2); 60@2:hd(3) }.
```

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

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```
#maximize { C@1:hd(I,P,C) }.
#minimize { P@2:hd(I,P,C) }.
```

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

# Weak constraints

- Weak constraints are an alternative to minimize statements
- Syntax  $\Leftarrow l_1, \dots, l_n [w@p]$   
where each  $l_i$  is a literal for  $1 \leq i \leq n$ ; and  $w$  and  $p$  are integers
- Example
  - :~ hd(1). [30@2]
  - :~ hd(2). [40@2]
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- Example  
 $: \sim \text{hd}(I, P, C) . [P@2]$

# Outline

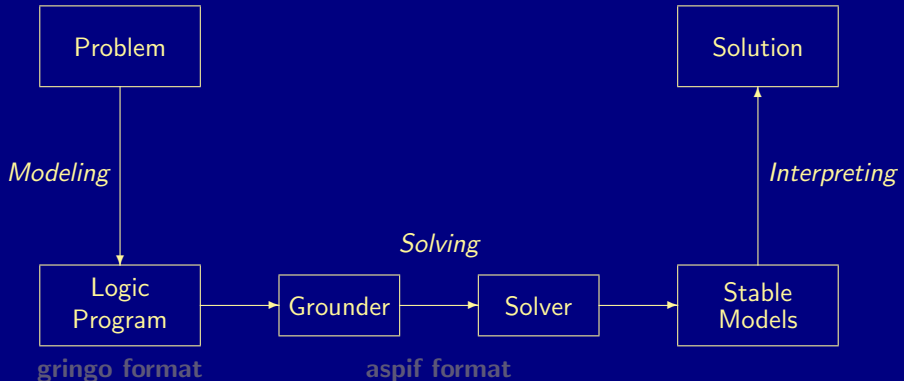
24 Base language

25 Optimization

26 **Formats**

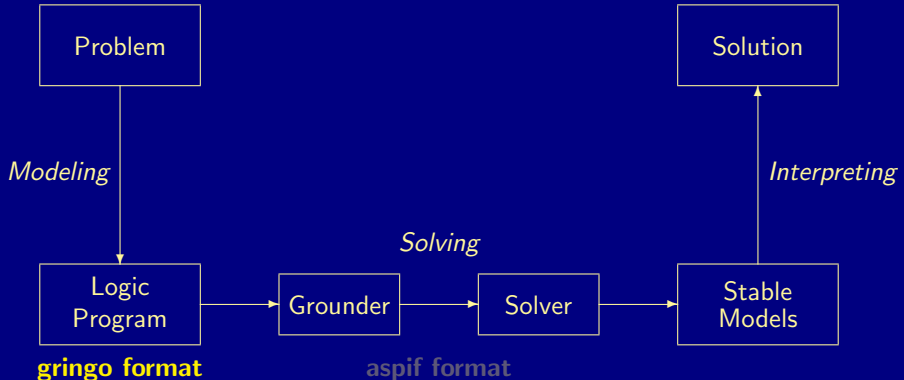
27 Summary

# Gringo's input and output language



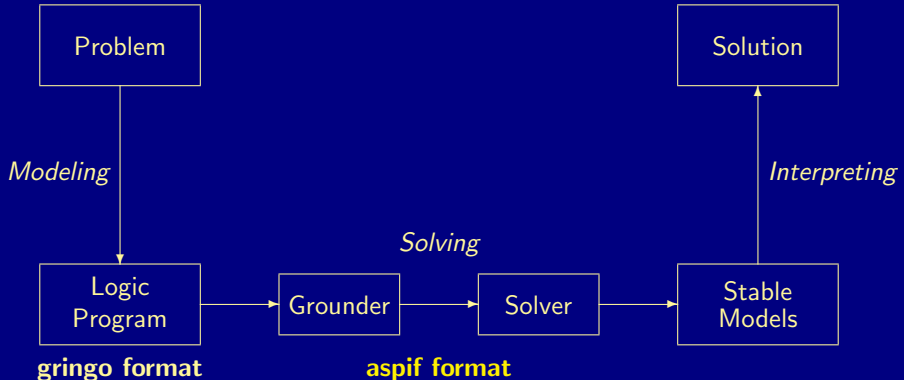
- **gringo format** is a user-oriented language for (non-ground) programs extending the ASP language standard *ASP-Core-2* [13]
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# Gringo's input and output language



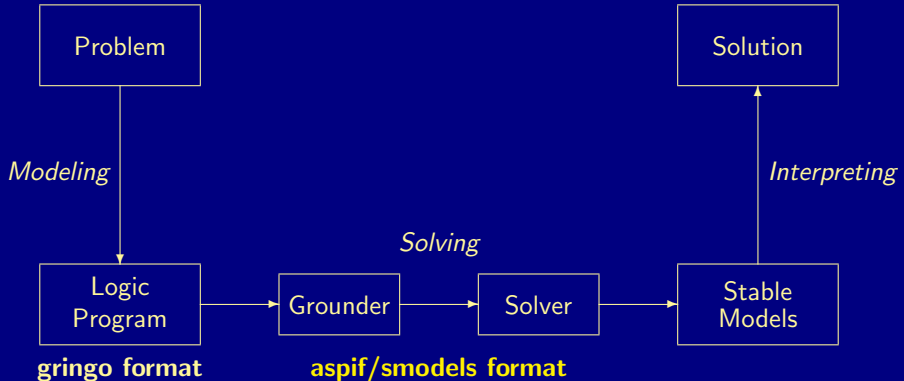
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# Outline

- 24 Base language
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  - Input format
  - Intermediate format
- 27 Summary



# Terms and literals

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

# Terms and literals

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

## ■ Atoms $a, \neg a$

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

## ■ Tuples $t$

## ■ Atoms $a, -a$

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## ■ Aggregate literals $a, \neg a, \neg\neg a$

## ■ Literals

# Terms and literals

- Terms  $t$
- Tuples  $\mathbf{t}$  of terms
- Atoms  $a, \neg a$
- Symbolic literals  $a, \neg a, \neg\neg a$
- Arithmetic literals  $t_1 \prec t_2$
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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  $a, \neg a, \neg\neg a$
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eg  $\neg p(f(3,c,Z),g(42,-,-))$  or  $q()$  written as  $q$

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viz `#false` and `#true`
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eg  $p(a,X)$ , 'not  $p(a,X)$ ', 'not not  $p(a,X)$ '
- Arithmetic literals  $t_1 \prec t_2$
- Conditional literals  $I : L$
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  - $t_1$  and  $t_2$  are terms
  - $\prec$  is a comparison symbol
- Conditional literals  $/ : L$
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 eg  $3 < 1$  or  $f(42) = X$
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- Conditional literals  $l : \mathbf{L}$  where
  - $l$  is a symbolic or arithmetic literal
  - $\mathbf{L}$  is a tuple of symbolic or arithmetic literals
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- Aggregate literals  $a, \neg a, \neg\neg a$
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 eg 'p(X,Y):q(X),r(Y)' or p(42) or '#false:q'
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  - omitting  $\prec_1$  or  $\prec_2$  defaults to  $\leq$
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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

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

where

- $l_i$  is a conditional literal for  $1 \leq i \leq m$  and
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- Note Semicolons ';' must be used in (1) instead of commas ',' whenever some  $l_i$  is a (genuine) conditional literal for  $1 \leq i \leq n$
- Example  $a(X) :- b(X) : c(X), d(X); e(x).$
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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

- $\alpha, \prec_i, s_i, \mathbf{t}_j$  are as given above for  $i = 1, 2$  and  $1 \leq j \leq k$
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- $l_i$  is a literal for  $m+1 \leq i \leq n$  (as in (1))

is a shorthand for the following  $k+1$  rules

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

- Example  $10 < \# \text{sum} \{ C, X, Y : \text{edge}(X, Y) : \text{cost}(X, Y, C) \}.$

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# Shortcuts

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$$s_1 \{l_1 : \mathbf{L}_1; \dots; l_k : \mathbf{L}_k\} s_2$$

is a shortcut for

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if it appears in the head of a rule and

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where  $t_i \neq t_j$  whenever  $l_i \neq l_j$  for  $i \neq j$  and  $1 \leq i, j \leq k$

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- Example `sum(S) :- S = #sum { C,X,Y : travel(X,Y,C) }`.

Assume `'travel(a,b,3). travel(b,c,3).'`

We get `sum(3)!`

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```
#sum { 3,a,b : travel(a,b,3); 3,b,c : travel(b,c,3) }
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```
$ gringo --text <(echo "{a;b}.")
```

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#count{ 1,0,a : a; 1,0,b : b }.
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- Example Why “;”?

$$\{ q(X,Y) : p(X), p(Y), X < Y; q(X,X) : p(X) \} = 1$$

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## Weak constraints

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

$$\Leftarrow l_1, \dots, l_n [w@p, t_1, \dots, t_m]$$

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

```
:~ hd(I,P,C). [P@42,I]
```

amounts to the minimize statement

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#minimize{ P@42,I : hd(I,P,C) }.
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## Some more directives

### ■ Output

`#show.`      `#show  $p/n$ .`      `#show  $t : l_1, \dots, l_n$ .`

### ■ Projection

`#project  $p/n$ .`      `#project  $a : l_1, \dots, l_n$ .`

### ■ Heuristics

`#heuristic  $a : l_1, \dots, l_n$ . [ $k@p, m$ ]`

### ■ Acyclicity

`#edge  $(u, v) : l_1, \dots, l_n$ .`

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# Outline

- 24 Base language
- 25 Optimization
- 26 Formats
  - Input format
  - Intermediate format
- 27 Summary

## *smodels* format

- The *smodels* format consists of
  - normal rules
  - choice rules
  - cardinality rules
  - weight rules
  - minimization statements
- Block-oriented format

# *models* format in detail

| Type/Format                                                                                                                                                                                                                                                     |
|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Normal rule Slide 92<br>$1 \sqcup (a_0) \sqcup n \sqcup n - m \sqcup \iota(a_{m+1}) \sqcup \dots \sqcup \iota(a_n) \sqcup \iota(a_1) \sqcup \dots \sqcup \iota(a_m)$                                                                                            |
| Cardinality rule Slide 553<br>$2 \sqcup (a_0) \sqcup n \sqcup n - m \sqcup \iota(a_{m+1}) \sqcup \dots \sqcup \iota(a_n) \sqcup \iota(a_1) \sqcup \dots \sqcup \iota(a_m)$                                                                                      |
| Choice rule Slide 541<br>$3 \sqcup m \sqcup (a_1) \sqcup \dots \sqcup \iota(a_m) \sqcup o - m \sqcup o - n \sqcup \iota(a_{n+1}) \sqcup \dots \sqcup \iota(a_o) \sqcup \iota(a_{m+1}) \sqcup \dots \sqcup \iota(a_n)$                                           |
| Weight rule Slide 581<br>$5 \sqcup (a_0) \sqcup \iota \sqcup n \sqcup n - m \sqcup \iota(a_{m+1}) \sqcup \dots \sqcup \iota(a_n) \sqcup \iota(a_1) \sqcup \dots \sqcup \iota(a_m) \sqcup w_{m+1} \sqcup \dots \sqcup w_n \sqcup w_1 \sqcup \dots \sqcup w_m$    |
| Minimize statement <sup>2</sup> Slide 593<br>$6 \sqcup 0 \sqcup n \sqcup n - m \sqcup \iota(a_{m+1}) \sqcup \dots \sqcup \iota(a_n) \sqcup \iota(a_1) \sqcup \dots \sqcup \iota(a_m) \sqcup w_{m+1} \sqcup \dots \sqcup w_n \sqcup w_1 \sqcup \dots \sqcup w_m$ |
| Disjunctive rule Slide 723<br>$8 \sqcup m \sqcup (a_1) \sqcup \dots \sqcup \iota(a_m) \sqcup o - m \sqcup o - n \sqcup \iota(a_{n+1}) \sqcup \dots \sqcup \iota(a_o) \sqcup \iota(a_{m+1}) \sqcup \dots \sqcup \iota(a_n)$                                      |

- The function  $\iota$  represents a mapping of atoms to numbers

# Example

```
{a}.
b :- a.
c :- not a.
```

```
3 1 2 0 0
1 3 1 1 2
1 4 1 0 2
0
2 a
3 c
4 b
0
B+
0
B-
1
0
1
```

# Example

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1
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```

## *aspif* format

- The *aspif* format consists of
  - rule statements
  - minimize statements
  - projection statements
  - output statements
  - external statements
  - assumption statements
  - heuristic statements
  - edge statements
  - theory terms and atoms
  - comments
- Line-oriented format



# Rule statements

Rule statements have the form

$1 \_ H \_ B$

■ Head  $H$  has form

$h \_ m \_ a_1 \_ \dots \_ a_m$

- $h \in \{0, 1\}$  determines whether the head is a disjunction or choice,
- $m \geq 0$  is the number of head elements, and
- each  $a_i$  is a positive literal

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- Note Heads are disjunctions or choices, including the special case of singular disjunctions for representing normal rules

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■ Body  $B$  has one of two forms

■ normal bodies have form

$$0 \_ n \_ l_1 \_ \dots \_ l_n$$

- $n \geq 0$  is the length of the rule body, and
- each  $l_i$  is a literal

■ weight bodies have form

$$1 \_ l \_ n \_ l_1 \_ w_1 \_ \dots \_ l_n \_ w_n$$

- $l$  is a positive integer to denote the lower bound,
- $n \geq 0$  is the number of literals in the rule body, and
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# Rule statements

Rule statements have the form

$$1 \sqcup H \sqcup B$$

■ Head  $H$  has form

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- $h \in \{0, 1\}$  determines whether the head is a disjunction or choice,
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■ Note All types of rules are included in the above rule format

# Example

```
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```
asp 1 0 0
1 1 1 1 0 0
1 0 1 2 0 1 1
1 0 1 3 0 1 -1
4 1 a 1 1
4 1 b 1 2
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0
```

# Outline

24 Base language

25 Optimization

26 Formats

27 Summary



# Things to remember

- integrity constraints
- choice rules
- cardinality rules, cardinality constraints
- weight rules, weight constraints
- conditional literals
- optimization statements and weak constraints
- input languages, *gringo*, ASP-Core-2
- intermediate languages, *smodels*, *aspif*

# Language Extensions: Overview

- 28 Two kinds of negation
- 29 Disjunctive logic programs
- 30 Propositional theories
- 31 Aggregates

# Outline

- 28 Two kinds of negation
- 29 Disjunctive logic programs
- 30 Propositional theories
- 31 Aggregates

# Motivation

## ■ Classical versus default negation

### ■ Symbol $-$ and $\neg$

#### ■ Idea

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

- $\neg a \approx a \notin X$

#### ■ Example

- $cross \leftarrow -train$

- $cross \leftarrow \neg train$

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$$\blacksquare -a \approx -a \in X$$

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### ■ Symbol $\neg$ and $\neg$

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$$\blacksquare \neg a \approx \neg a \in X$$

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$$\blacksquare \textit{cross} \leftarrow \neg \textit{train}$$

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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  $\overline{\mathcal{A}} = \{-a \mid a \in \mathcal{A}\}$  such that  $\mathcal{A} \cap \overline{\mathcal{A}} = \emptyset$
- Given a program  $P$  over  $\mathcal{A}$ , classical negation is encoded by adding

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

- 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  $P \cup P^-$

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# An example

## ■ The program

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

induces

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

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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

- $P_1 = \{cross \leftarrow \neg train\}$ 
  - stable model:  $\{cross\}$
- $P_2 = \{cross \leftarrow \neg train\}$ 
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- $P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow\}$ 
  - stable model:  $\{cross, \neg train\}$
- $P_4 = \{cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow\}$ 
  - stable model:  $\{cross, \neg cross, train, \neg train\}$
- $P_5 = \{cross \leftarrow \neg train, \neg train \leftarrow \neg train\}$ 
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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

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

- A set  $X$  of atoms is a stable model of a program  $P$  (with default negation in rule heads) over  $\mathcal{A}$ ,  
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# Outline

- 28 Two kinds of negation
- 29 Disjunctive logic programs
- 30 Propositional theories
- 31 Aggregates

# Disjunctive logic programs

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

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

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

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

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

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

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

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

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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)

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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

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

- The sets  $\{a, b\}$ ,  $\{a, c\}$ , and  $\{a, b, c\}$  are closed under  $P$
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## Graph coloring (reloaded)

```
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)).
```

```
assign(X,r) ; assign(X,b) ; assign(X,g) :- node(X).
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```
:- edge(X,Y), assign(X,C), assign(Y,C).
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```
color(r). color(b). color(g).
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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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## Some properties

- A disjunctive logic program may have zero, one, or multiple stable models
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- If  $a \in X$  for some stable model  $X$  of a disjunctive logic program  $P$ , then there is a rule  $r \in P$  such that  $B(r)^+ \subseteq X$ ,  $B(r)^- \cap X = \emptyset$ , and  $H(r) \cap X = \{a\}$

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## An example with variables

$$P = \left\{ \begin{array}{ll} a(1, 2) & \leftarrow \\ b(X) ; c(Y) & \leftarrow a(X, Y), \neg c(Y) \end{array} \right\}$$

$$\text{ground}(P) = \left\{ \begin{array}{ll} a(1, 2) & \leftarrow \\ b(1) ; c(1) & \leftarrow a(1, 1), \neg c(1) \\ b(1) ; c(2) & \leftarrow a(1, 2), \neg c(2) \\ b(2) ; c(1) & \leftarrow a(2, 1), \neg c(1) \\ b(2) ; c(2) & \leftarrow a(2, 2), \neg c(2) \end{array} \right\}$$

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- Consider  $X = \{a(1, 2), b(1)\}$
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- Consider disjunctive rules of the form

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

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

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- A set  $X$  of atoms is a stable model of a disjunctive 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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# Outline

- 28 Two kinds of negation
- 29 Disjunctive logic programs
- 30 Propositional theories**
- 31 Aggregates

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- Formulas are formed from

- atoms in  $\mathcal{A}$
- $\perp$

using

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

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

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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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## Two examples

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

■ For  $X = \{p, q, r\}$ , we get

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For  $X = \emptyset$ , we get

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# Logic programs as propositional theories

- The normal logic program  $P = \{p \leftarrow \neg q, q \leftarrow \neg p\}$  corresponds to  $\tau[P] = \{\neg q \rightarrow p, \neg p \rightarrow q\}$ 
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# Outline

- 28 Two kinds of negation
- 29 Disjunctive logic programs
- 30 Propositional theories
- 31 Aggregates**

# 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:

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

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  $\text{sum } \{30 : \text{hd}(a), \dots, 50 : \text{hd}(m)\} \leq 300$

## 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  $\text{sum}\{1 : p, 1 : q\} \neq 1$

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  $\text{sum}\{1 : p, 1 : q\} = 1$

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## ■ Monotone aggregates

### ■ For instance,

- $B(r)^+$

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

- We get a simpler characterization:  $\bigwedge_{I \subseteq \{1, \dots, n\}, \alpha \{w_i | i \in I\} \not\preceq_k \bigvee_{i \in \bar{I}} a_i$

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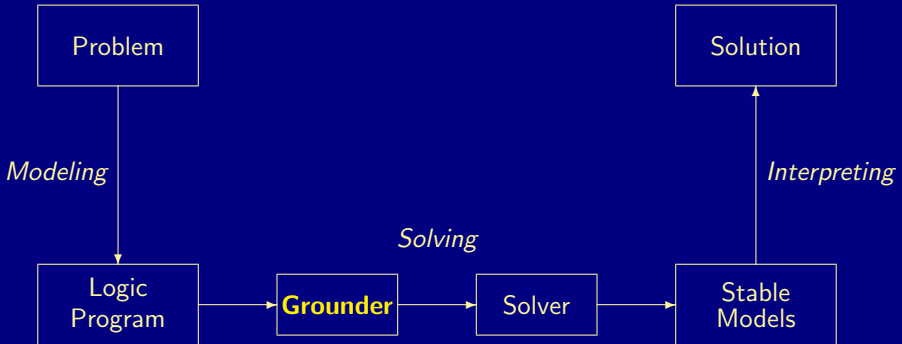
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# Grounding: Overview

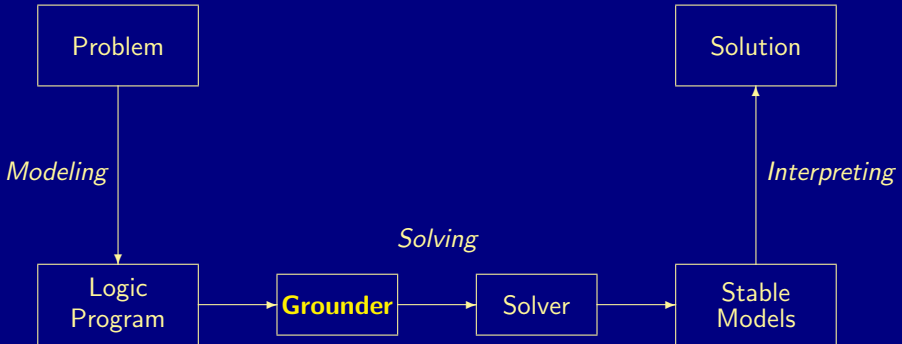
- 32 Naive grounding
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- Disclaimer: Grounding algorithms for normal logic programs

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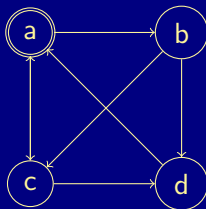
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# Hamiltonian cycle

Instance

```
% vertices
node(a). node(b).
node(c). node(d).

% edges
edge(a,b). edge(a,c).
edge(b,c). edge(b,d).
edge(c,a). edge(c,d).
edge(d,a).
```





# Hamiltonian cycle

## Encoding

```
% generate path
path(X,Y) :- not omit(X,Y), edge(X,Y).
omit(X,Y) :- not path(X,Y), edge(X,Y).

% at most one incoming/outgoing edge
:- path(X,Y), path(X',Y), X < X'.
:- path(X,Y), path(X,Y'), Y < Y'.

% at least one incoming/outgoing edge
on_path(Y) :- path(X,Y), path(Y,Z).
:- node(X), not on_path(X).

% connectedness
reach(X) :- start(X).
reach(Y) :- reach(X), path(X,Y).
:- node(X), not reach(X).
```

# Essential concepts

- Safety of a rule
  - each variable must occur in a positive body literal
- Term universe of a program
  - all constants in the program
- Atom base of a program
  - all ground atoms over predicates in program
- Ground instance of a rule
  - all variables replaced with elements from term universe
- Ground instantiation of a logic program
  - $ground(P)$  is the union of all instances of rules in  $P$
- Unifier of two expressions
  - a (variable) substitution making two expressions equal once applied to both

# Size of grounding

## Hamiltonian cycle

```
% term universe: {a,b,c,d}
12 facts from instance
% path(X,Y) :- not omit(X,Y), edge(X,Y).
% omit(X,Y) :- not path(X,Y), edge(X,Y).
% reach(Y) :- reach(X), path(X,Y).
16 rules + 16 rules + 16 rules
% on_path(Y) :- path(X,Y), path(Y,Z).
% :- path(X,Y), path(X',Y), X < X'.
% :- path(X,Y), path(X,Y'), Y < Y'.
64 rules + 64 rules + 64 rules
% reach(X) :- start(X).
% :- node(X), not on_path(X).
% :- node(X), not reach(X).
4 rules + 4 rules + 4 rules
```

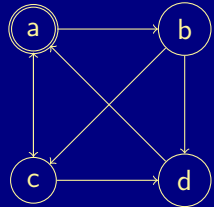
## Unnecessary rules, I

Hamiltonian cycle

```

% path(X,Y) :- not omit(X,Y), edge(X,Y).
path(a,a) :- not omit(a,a), edge(a,a).
path(a,b) :- not omit(a,b), edge(a,b).
path(a,c) :- not omit(a,c), edge(a,c).
path(a,d) :- not omit(a,d), edge(a,d).
 :
path(d,a) :- not omit(d,a), edge(d,a).
path(d,b) :- not omit(d,b), edge(d,b).
path(d,c) :- not omit(d,c), edge(d,c).
path(d,d) :- not omit(d,d), edge(d,d).

```



## Unnecessary rules, II

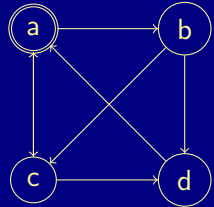
Hamiltonian cycle

```

% :- path(X,Y), path(X',Y), X < X'.
:- path(a,a), path(a,a), a < a.
:- path(a,b), path(a,b), a < a.
:- path(a,c), path(a,c), a < a.
:- path(a,d), path(a,d), a < a.

:- path(a,a), path(b,a), a < b.
:- path(a,b), path(b,b), a < b.
:- path(a,c), path(b,c), a < b.
:- path(a,d), path(b,d), a < b.
 ⋮
:- path(d,d), path(d,d), d < d.

```



# Outline

- 32 Naive grounding
- 33 Bottom-up grounding
- 34 Semi-naive grounding
- 35 On-the-fly simplifications
- 36 Rule instantiation
- 37 Summary

# Bottom up grounding

- Idea Ground **relevant** rules by gradually extending the atom base
- Relative ground instantiation of a logic program  $P$  wrt a set of ground atoms  $D$

$$\text{ground}_D(P) = \{r \in \text{ground}(P) \mid B(r)^+ \subseteq D\}$$

# Bottom up grounding

- Idea Ground **relevant** rules by gradually extending the atom base
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- Algorithm

---

```

function ground_bottom_up(P, D)
 $G \leftarrow \text{ground}_D(P)$;
 if $H(G) \not\subseteq D$ then
 \quad return ground_bottom_up($P, D \cup H(G)$);
 return G ;

```

---

# Bottom up grounding

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```

---

- Property Given a safe normal program  $P$  and set of ground facts  $I$ ,  $P \cup I$  is equivalent to  $\text{ground\_bottom\_up}(P, H(I)) \cup I$

# Bottom up grounding, step 1

Hamiltonian cycle

```
% Step 1
path(a,b) :- not omit(a,b), edge(a,b).
 : % 7 rules total
path(d,a) :- not omit(d,a), edge(d,a).

omit(a,b) :- not path(a,b), edge(a,b).
 : % 7 rules total
omit(d,a) :- not path(d,a), edge(d,a).

:- node(a), not on_path(a). :- node(b), not on_path(b).
:- node(c), not on_path(c). :- node(d), not on_path(d).

:- node(a), not reach(a). :- node(b), not reach(b).
:- node(c), not reach(c). :- node(d), not reach(d).

reach(a) :- start(a).
```

# Bottom up grounding, step 2

Hamiltonian cycle

```
% Step 2 and rules of Step 1
:- path(a,c), path(b,c), a < b.
:- path(b,d), path(c,d), b < c.
:- path(c,a), path(d,a), c < d.

:- path(a,b), path(a,c), b < c.
:- path(c,a), path(c,d), a < d.
:- path(b,c), path(b,d), c < d.

on_path(a) :- path(c,a), path(a,c).
 : % 12 rules total
on_path(d) :- path(c,d), path(d,a).

reach(b) :- reach(a), path(a,b).
reach(c) :- reach(a), path(a,c).
```

# Bottom up grounding, step 3 and 4

Hamiltonian cycle

```
% Step 3 and rules of Step 2
reach(c) :- reach(b), path(b,c).
reach(d) :- reach(b), path(b,d).
reach(a) :- reach(c), path(c,a).
reach(d) :- reach(c), path(c,d).
```

```
% Step 4 and rules of Step 3
reach(a) :- reach(d), path(d,a).
```

# Properties of bottom-up grounding

- Grounds only **relevant** rules
  - each positive body literal has a non-cyclic derivation (ignoring negative literals)
- **Re-grounds** rules from previous steps
- Performs no **simplifications**

# Improving bottom-up grounding

- Use dependencies to **focus** grounding
  - begin with partial atom base given by facts
  - use rule dependency graph of program to obtain **components** that can be **grounded successively**
- Adapt **semi-naive evaluation** put forward in the database field
  - avoids redundancies when grounding
- Perform **simplifications** during grounding
  - remove literals from rule bodies if possible
  - omit rules if body cannot be satisfied

# Program dependencies

- Dependency graph of program  $P$ 
  - rule  $r_2$  depends on rule  $r_1$   
if  $b \in B(r_2)^+ \cup B(r_2)^-$  unifies with  $h \in h(r_1)$
  - $G_P = (P, E)$  where  $E = \{(r_1, r_2) \mid r_2 \text{ depends on } r_1\}$
- Positive dependency graph of program  $P$ 
  - rule  $r_2$  positively depends on rule  $r_1$   
if  $b \in B(r_2)^+$  unifies with  $h \in H(r_1)$
  - $G_P^+ = (P, E)$  where  $E = \{(r_1, r_2) \mid r_2 \text{ positively depends on } r_1\}$



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  - $(C_1, \dots, C_n)$  is a topological ordering of  $G_P$

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  - $(C_1, \dots, C_n)$  is a topological ordering of  $G_P$ ,  
that is,  $(r_1, r_2) \in E, r_1 \in C_i, r_2 \in C_j$ , implies  $i \leq j$

# Program dependencies

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  - $(C_1, \dots, C_n)$  is a topological ordering of  $G_P$
  - $(C_{i,1}, \dots, C_{i,m_i})$  is a topological ordering of each  $G_{C_i}^+$

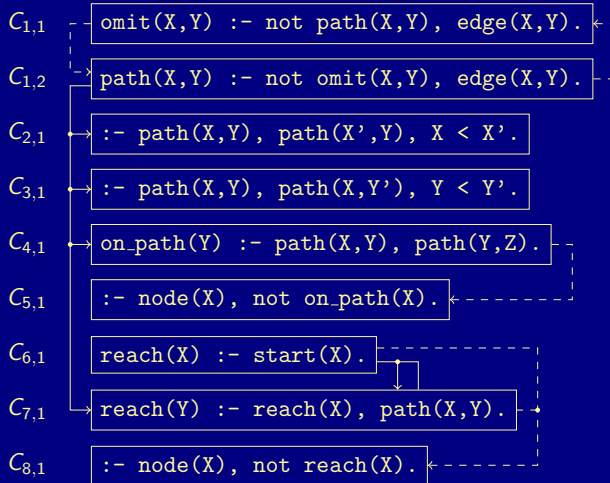
# Program dependencies

- Dependency graph of program  $P$ 
    - $G_P = (P, E)$  where  $E = \{(r_1, r_2) \mid r_2 \text{ depends on } r_1\}$
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    - $(C_{i,1}, \dots, C_{i,m_i})$  is a topological ordering of each  $G_{C_i}^+$
- ↳  $L_P = (C_{1,1}, \dots, C_{1,m_1}, \dots, C_{n,1}, \dots, C_{n,m_n})$



## Dependencies

Hamiltonian cycle



# Grounding with dependencies

---

```

function ground_with_dependencies(P, D)
 $G \leftarrow \emptyset$;
 foreach C in L_P do
 $G' \leftarrow \text{ground_bottom_up}(C, D)$;
 $(G, D) \leftarrow (G \cup G', D \cup H(G'))$;
 return G ;

```

---

- **Property** Given a safe normal program  $P$  and set of facts  $I$ ,  
 $P \cup I$  is equivalent to  $\text{ground\_with\_dependencies}(P, H(I)) \cup I$

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# Grounding with dependencies

Hamiltonian cycle

```
% component $C_{1,1}$
omit(a,b) :- not path(a,b), edge(a,b).
 : % 7 rules total
omit(d,a) :- not path(d,a), edge(d,a).
```

```
% component $C_{1,2}$
path(a,b) :- not omit(a,b), edge(a,b).
 : % 7 rules total
path(d,a) :- not omit(d,a), edge(d,a).
```

...

- No re-grounding if there is no positive recursion in a component

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...

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Grounding component  $C_{7,1}$ 

Hamiltonian cycle

```
% Step 1
reach(b) :- reach(a), path(a,b).
reach(c) :- reach(a), path(a,c).
```

```
% Step 2 and rules of Step 1
reach(c) :- reach(b), path(b,c).
reach(d) :- reach(b), path(b,d).
reach(a) :- reach(c), path(c,a).
reach(d) :- reach(c), path(c,d).
```

```
% Step 3 and rules of Step 2
reach(a) :- reach(d), path(d,a).
```

```
% less re-grounding but still...
```

# Outline

- 32 Naive grounding
- 33 Bottom-up grounding
- 34 Semi-naive grounding
- 35 On-the-fly simplifications
- 36 Rule instantiation
- 37 Summary

# Semi-naive grounding

- Idea (originates from database systems [1])

To avoid recomputing the same atoms at each level,  
semi-naive grounding focuses on newly generated atoms:

Any new atom at step  $i$  relies on at least one atom  
newly derived at step  $i - 1$

- Recursive atoms Given  $L_P = (C_1, \dots, C_n)$ ,  
an atom  $a_1$  is recursive in component  $C_i$   
if  $a_1$  unifies  $a_2$  such that

- $r_1 \in C_i$  and  $r_2 \in C_j$  with  $i \leq j$
- $a_1 \in B(r_1)^+ \cup B(r_1)^-$
- $a_2 \in h(r_2)$



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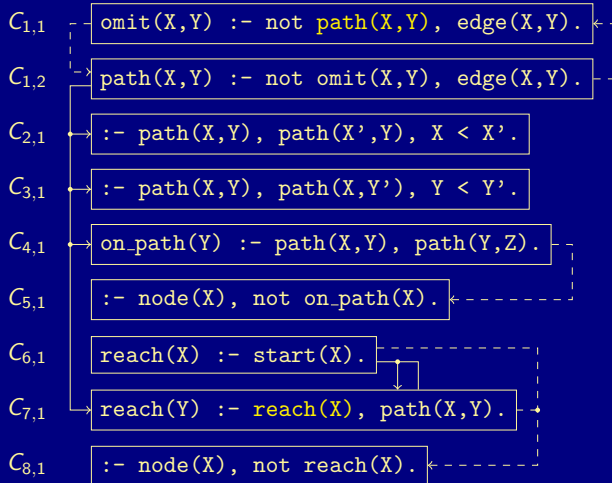
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if  $a_1$  unifies  $a_2$  such that

- $r_1 \in C_i$  and  $r_2 \in C_j$  with  $i \leq j$
- $a_1 \in B(r_1)^+ \cup B(r_1)^-$
- $a_2 \in h(r_2)$

## Recursive atoms

Hamiltonian cycle



## Preparing components

- The set of **prepared rules** for  $r \in C$  is

$$\left\{ \begin{array}{l} h \leftarrow n(b_1), a(b_2), \dots, a(b_{i-1}), a(b_i), B \\ h \leftarrow o(b_1), n(b_2), \dots, a(b_{i-1}), a(b_i), B \\ \vdots \quad \vdots \quad \quad \quad \ddots \quad \quad \quad \vdots \quad \vdots \\ h \leftarrow o(b_1), o(b_2), \dots, n(b_{i-1}), a(b_i), B \\ h \leftarrow o(b_1), o(b_2), \dots, o(b_{i-1}), n(b_i), B \end{array} \right\}$$

or  $\{ h \leftarrow n(b_{i+1}), \dots, n(b_j), b_{j+1}, \dots, b_n \}$  if  $i = 0$

where

- $B(r) = \{b_1, \dots, b_i, b_{i+1}, \dots, b_j, b_{j+1}, \dots, b_n\}$
- $b_k \in B(r)^+$  for  $1 \leq k \leq i$  is recursive
- $b_k \in B(r)^+$  for  $i < k \leq j$  is not recursive
- $B = a(b_{i+1}), \dots, a(b_j), b_{j+1}, \dots, b_n$
- A prepared component is the union of all its prepared rules

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- A **prepared component** is the union of all its prepared rules

## Preparing components

```
% prepared component $C_{1,1}$
omit(X,Y) :- n(edge(X,Y)), not path(X,Y).
% prepared component $C_{1,2}$
path(X,Y) :- n(edge(X,Y)), not omit(X,Y).
% prepared component $C_{2,1}$
:- n(path(X,Y)), n(path(X',Y)), X < X'.
...
% prepared component $C_{7,1}$
reach(Y) :- n(reach(X)), a(path(X,Y)).
...
```

# Semi-naive evaluation-based grounding

---

```

function ground_semi_naive(P, A)
 $G \leftarrow \emptyset$;
 foreach C in L_P do
 $(O, N) \leftarrow (\emptyset, A)$;
 repeat
 let $D_p = \{p(a) \mid a \in D\}$ for set D of atoms;
 $G' \leftarrow \text{ground}_{O_o \cup N_n \cup A_a}(\text{prepared } C)$;
 $N \leftarrow H(G') \setminus A$;
 $(G, O, A) \leftarrow (G \cup G', A, N \cup A)$;
 until $N = \emptyset$;
 return G with $o/1, n/1, a/1$ stripped from positive bodies;

```

---

- **Property** Given a safe normal program  $P$  and set of facts  $I$ ,  
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Grounding component  $C_{7,1}$ 

Hamiltonian cycle

```

% grounding of
% reach(Y) :- n(reach(X)), a(path(X,Y)).

% Step 1 with N = A from previous step (reach(a) ∈ A)
reach(b) :- n(reach(a)), a(path(a,b)).
reach(c) :- n(reach(a)), a(path(a,c)).

% Step 2 with N = { reach(b), reach(c) }
reach(c) :- n(reach(b)), a(path(b,c)).
reach(d) :- n(reach(b)), a(path(b,d)).
reach(a) :- n(reach(c)), a(path(c,a)).
reach(d) :- n(reach(c)), a(path(c,d)).

% Step 3 with N = { reach(d) }
reach(a) :- n(reach(d)), a(path(d,a)).

```



Grounding component  $C_{7,1}$ 

Hamiltonian cycle

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reach(d) :- reach(b), path(b,d).
reach(a) :- reach(c), path(c,a).
reach(d) :- reach(c), path(c,d).

% Step 3 with N = { reach(d) }
reach(a) :- reach(d), path(d,a).

% without n/1 and a/1 of course

```

# Nonlinear programs

Example

```
trans(U,V) :- edge(U,V).
trans(U,W) :- trans(U,V), trans(V,W).
```

```
% prepared Component 1:
trans(U,V) :- n(edge(U,V)).
```

```
% prepared Component 2:
trans(U,W) :- n(trans(U,V)), a(trans(V,W)).
trans(U,W) :- o(trans(U,V)), n(trans(V,W)).
```

# Nonlinear programs

## Example

```
trans(U,V) :- edge(U,V).
% trans(U,W) :- trans(U,V), trans(V,W).
% better written as:
trans(U,W) :- trans(U,V), edge(V,W).

% prepared Component 1:
trans(U,V) :- n(edge(U,V)).

% prepared Component 2:
trans(U,W) :- n(trans(U,V)), a(edge(V,W)).
```

# Outline

- 32 Naive grounding
- 33 Bottom-up grounding
- 34 Semi-naive grounding
- 35 On-the-fly simplifications**
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- 37 Summary

# Propagation of atoms

- Simplifications are performed **on-the-fly** relative to the atom base
  - ➡ rules are printed immediately but not stored in *gringo*
- Distinguish a set of true atoms among the atom base
  - ➡ each true atom amounts to a fact
- Simplifications
  - Remove facts from positive body
  - Discard rules with negative literals over a fact
  - Discard rules whenever the head is a fact
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# Propagation of true atoms

Hamiltonian cycle

```
...
path(a,b) :- not omit(a,b), edge(a,b).
...
reach(a) :- start(a).
```



# Propagation of true atoms

Hamiltonian cycle

```
...
path(a,b) :- not omit(a,b).
...
reach(a). % reach(a) is added as fact
```

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```
...
path(a,b) :- not omit(a,b).
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reach(a). % reach(a) is added as fact

...

:- node(a), not reach(a).
...
```

# Propagation of true atoms

Hamiltonian cycle

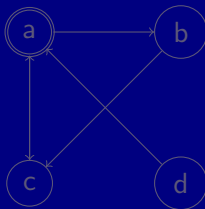
```
...
path(a,b) :- not omit(a,b).
...
reach(a). % reach(a) is added as fact

...

:- node(a), not reach(a). % rule is discarded
...
```

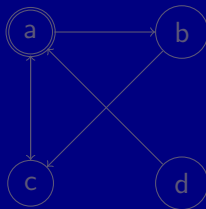
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- Non-recursive body negative literals whose atom is not in the current atom base can be removed from rule bodies
- Example Consider the instance where node *d* is not reachable



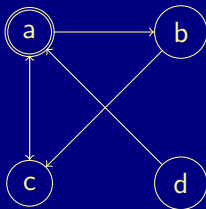
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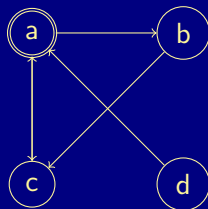
# Propagation of negative literals

Hamiltonian cycle

```

path(a,b) :- not omit(a,b).
path(a,c) :- not omit(a,c).
path(b,c) :- not omit(b,c).
path(c,a) :- not omit(c,a).
path(d,a) :- not omit(d,a).
...
reach(a).
reach(b) :- path(a,b).
reach(c) :- path(a,c).
reach(c) :- path(b,c), reach(b).
...
% reach(d) $\notin D$ (atom base)
:- not reach(b).
:- not reach(c).
:- not reach(d). % remove not reach(d) from body

```



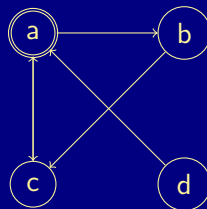
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reach(a).
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reach(c) :- path(a,c).
reach(c) :- path(b,c), reach(b).
...
% reach(d) $\notin D$ (atom base)
:- not reach(b).
:- not reach(c).
:- . % inconsistency detected during grounding

```





# Stratified logic programs

- A logic program  $P$  is stratified, if it has a partition  $(P_i)_{0 \leq i \leq n}$  such that for each predicate  $p$ 
  - $\text{def}_P(p) \subseteq P_i$  for some  $1 \leq i \leq n$
 and, for each  $1 \leq i \leq n$ ,
  - if  $p$  occurs in  $B(r)^+$  for some  $r \in P_i$ , then  $\text{def}_P(p) \subseteq \bigcup_{j \leq i} P_j$
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- Property Stratified logic programs
  - are completely evaluated during grounding
  - have a single stable model

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# Stratified logic programs

- A logic program  $P$  is **stratified**, if its dependency graph has no cycles containing negative edges
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# Outline

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## Safe body order

- Given safe rule  $r$ , the tuple  $(b_1, \dots, b_n)$  is a safe body order if
  - $\{b_1, \dots, b_n\} = B(r)$
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- Example Given rule `':- node(X), not reach(X).'`
  - `(node(X), not reach(X))` is a safe body order
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# Matching body literals

- $match_{F,D}(\sigma, b)$  is the set of all matches for literal  $b$ 
  - $\sigma$  is a (ground) substitution
  - $F$  are true atoms (set of ground atoms)
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- $\sigma' \in match_{F,D}(\sigma, b)$  if
  - $\sigma \subseteq \sigma'$  and  $vars(b) \subseteq vars(\sigma') \subseteq vars(b) \cup vars(\sigma)$
  - $b\sigma'$  holds if  $b$  is a comparison literal
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<sup>3</sup> $vars(a)$  gives all variables occurring in atom  $a$ .



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<sup>3</sup> $vars(\sigma)$  gives all variables mapped by  $\sigma$  on ground terms.

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# An Example

- Consider body order  $(p(X), q(X, Y), \text{not } r(Y))$ 
  - $F = \{r(3)\}$
  - $D = \{p(1), q(1, 2), q(1, 3), r(3)\}$
- $match_{F,D}(\emptyset, p(X)) = \{\{X \mapsto 1\}\}$
- $match_{F,D}(\{X \mapsto 1\}, q(X, Y)) = \{\{X \mapsto 1, Y \mapsto 2\}, \{X \mapsto 1, Y \mapsto 3\}\}$
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# Rule grounding by backtracking

```

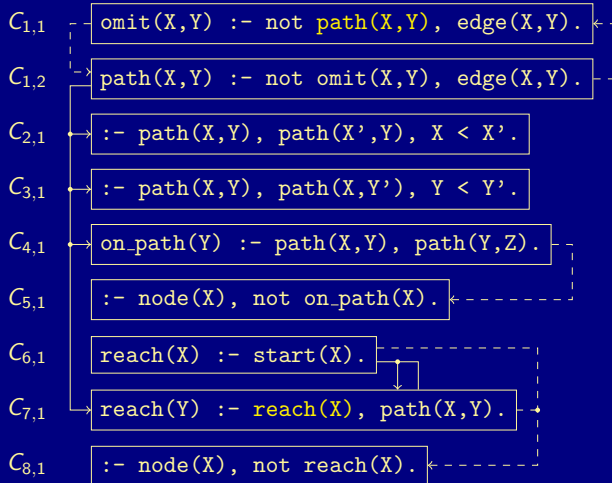
function ground_backtrackr,R,D($\sigma, F, (b_1, \dots, b_n)$)
 if $n = 0$ then
 let $H = H(r\sigma);$
 $B = B(r\sigma)^+ \setminus F \cup;$
 $\{\text{not } a\sigma \mid a \in B(r)^- \setminus R, a\sigma \in D\} \cup;$
 $\{\text{not } a\sigma \mid a \in B(r)^- \cap R\};$
 if $B = \emptyset$ then $F \leftarrow F \cup H;$
 return $(\{H \leftarrow B \mid B^- \cap F = \emptyset, H \cap F = \emptyset\}, F);$
 else
 $G \leftarrow \emptyset;$
 foreach $\sigma' \in \text{match}_{F,D}(\sigma, b_1)$ do
 $(G, F) \leftarrow (G, F) \sqcup \text{ground_backtrack}_{r,R,D}(\sigma', F, (b_2, \dots, b_n));$
 return $(G, F);$

```



## Recursive atoms

Hamiltonian cycle



# Ground rule one

Hamiltonian cycle

- $r = \text{'omit}(X,Y) \text{ :- not path}(X,Y), \text{ edge}(X,Y). \text{'}$
- $R = \{ \text{path}(X,Y) \}$
- $F = \left\{ \begin{array}{l} \text{edge}(a,b), \text{edge}(a,c), \text{edge}(b,c), \text{edge}(b,d), \\ \text{edge}(c,a), \text{edge}(c,d), \text{edge}(d,a), \dots \end{array} \right\}$

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**1**  $\text{ground\_backtrack}_{r,R,D}(\emptyset, F, (\text{edge}(X,Y), \text{not path}(X,Y)))$

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Hamiltonian cycle

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- 1  $\text{ground\_backtrack}_{r,R,D}(\emptyset, F, (\text{edge}(X,Y), \text{not path}(X,Y)))$
  - 2  $\text{ground\_backtrack}_{r,R,D}(\{X \mapsto a, Y \mapsto b\}, F, (\text{not path}(X,Y)))$

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Hamiltonian cycle

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1  $\text{ground\_backtrack}_{r,R,D}(\emptyset, F, (\text{edge}(X,Y), \text{not path}(X,Y)))$

2  $\text{ground\_backtrack}_{r,R,D}(\{X \mapsto a, Y \mapsto b\}, F, (\text{not path}(X,Y)))$

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## Ground rule one

Hamiltonian cycle

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# Special termination cases

$$\blacksquare a(42) \text{ :- not } b(52) \quad R = \emptyset \quad F = \emptyset \quad D = \emptyset$$

$$\Rightarrow (\{a(42) .\}, F \cup \{a(42)\})$$

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# Outline

- 32 Naive grounding
- 33 Bottom-up grounding
- 34 Semi-naive grounding
- 35 On-the-fly simplifications
- 36 Rule instantiation
- 37 Summary

## Things we ignored

- (Recursive) aggregates
- Conditional literals
- Optimization statements
- Disjunctions
- Arithmetic functions
- Syntactic sugar to write more compact encodings
- Safety of  $=$  relation (for aggregates and terms)
- Python/Lua integration



# Things to remember

- Safety and bottom-up grounding
- Grounding along topological ordering of sets of rule
- Semi-naive evaluation
- Simplifications
- Rule instantiation
- Impact of
  - atoms  $F$  found to be true
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# Computational aspects: Overview

- 38 Consequence operator
- 39 Computation from first principles
- 40 Complexity

# Outline

38 Consequence operator

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# Consequence operator

- Let  $P$  be a positive program and  $X$  a set of atoms
  - The **consequence operator**  $T_P$  is defined as follows:

$$T_P X = \{h(r) \mid r \in P \text{ and } B(r) \subseteq X\}$$

- Iterated applications of  $T_P$  are written as  $T_P^j$  for  $j \geq 0$ 
  - $T_P^0 X = X$
  - $T_P^i X = T_P T_P^{i-1} X$  for  $i \geq 1$
- **Properties** For any positive program  $P$ 
  - $Cn(P) = \bigcup_{i \geq 0} T_P^i \emptyset$
  - $X \subseteq Y$  implies  $T_P X \subseteq T_P Y$
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# An example

- Consider the positive program

$$P = \{p \leftarrow, q \leftarrow, r \leftarrow p, s \leftarrow q, t, t \leftarrow r, u \leftarrow v\}$$

- We get

$$\begin{aligned}
 T_P^0 \emptyset &= \emptyset \\
 T_P^1 \emptyset &= \{p, q\} &= T_P T_P^0 \emptyset &= T_P \emptyset \\
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# Approximating stable models

- First Idea Approximate a stable model  $X$  by two sets of atoms  $L$  and  $U$  such that  $L \subseteq X \subseteq U$ 
  - $L$  and  $U$  constitute lower and upper bounds on  $X$
  - $L$  and  $(\mathcal{A} \setminus U)$  describe a three-valued model of the program
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$$X \subseteq Y \text{ implies } P^Y \subseteq P^X \text{ implies } Cn(P^Y) \subseteq Cn(P^X)$$

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# Approximating stable models

## ■ Second Idea

**repeat**

**replace**  $L$  **by**  $L \cup Cn(P^U)$

**replace**  $U$  **by**  $U \cap Cn(P^L)$

**until**  $L$  and  $U$  do not change anymore

## ■ Observations

### ■ At each iteration step

■  $L$  becomes larger (or equal)

■  $U$  becomes smaller (or equal)

■  $L \subseteq X \subseteq U$  is invariant for every stable model  $X$  of  $P$

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# The simplistic expand algorithm

$\text{expand}_P(L, U)$

**repeat**

$L' \leftarrow L$

$U' \leftarrow U$

$L \leftarrow L' \cup Cn(P^{U'})$

$U \leftarrow U' \cap Cn(P^{L'})$

**if**  $L \not\subseteq U$  **then return**

**until**  $L = L'$  and  $U = U'$

## An example

$$P = \left\{ \begin{array}{l} a \leftarrow \\ b \leftarrow a, \neg c \\ d \leftarrow b, \neg e \\ e \leftarrow \neg d \end{array} \right\}$$

|   | $L'$        | $Cn(P^{U'})$ | $L$        | $U'$                | $Cn(P^{L'})$     | $U$              |
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# The simplistic expand algorithm

- $\text{expand}_P$ 
  - tightens the approximation on stable models
  - is stable model preserving

Let's expand with  $d$  !

$$P = \left\{ \begin{array}{l} a \leftarrow \\ b \leftarrow a, \neg c \\ d \leftarrow b, \neg e \\ e \leftarrow \neg d \end{array} \right\}$$

|   | $L'$          | $Cn(P^{U'})$  | $L$           | $U'$                | $Cn(P^{L'})$  | $U$           |
|---|---------------|---------------|---------------|---------------------|---------------|---------------|
| 1 | $\{d\}$       | $\{a\}$       | $\{a, d\}$    | $\{a, b, c, d, e\}$ | $\{a, b, d\}$ | $\{a, b, d\}$ |
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## A simplistic solving algorithm

 $solve_P(L, U)$ 

|                                                   |                       |
|---------------------------------------------------|-----------------------|
| $(L, U) \leftarrow expand_P(L, U)$                | <i>// propagation</i> |
| <b>if</b> $L \not\subseteq U$ <b>then failure</b> | <i>// failure</i>     |
| <b>if</b> $L = U$ <b>then output</b> $L$          | <i>// success</i>     |
| <b>else choose</b> $a \in U \setminus L$          | <i>// choice</i>      |
| $solve_P(L \cup \{a\}, U)$                        |                       |
| $solve_P(L, U \setminus \{a\})$                   |                       |

# A simplistic solving algorithm

- Close to the approach taken by the ASP solver `smodels`, inspired by the Davis-Putman-Logemann-Loveland (DPLL) procedure
  - backtracking search building a binary search tree
  - a node in the search tree corresponds to a three-valued interpretation
  - the search space is pruned by
    - deriving deterministic consequences and detecting conflicts (**expand**)
    - making one choice at a time by appeal to a heuristic (**choose**)
  - heuristic choices are made on atoms

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# Outline

38 Consequence operator

39 Computation from first principles

40 Complexity

# Complexity

Let  $a$  be an atom and  $X$  be a set of atoms

- For a positive normal logic program  $P$ 
  - deciding whether  $X$  is the stable model of  $P$  is  $P$ -complete
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- For a normal logic program  $P$  with optimization statements
  - deciding whether  $X$  is an optimal stable model of  $P$  is  $co-NP$ -complete
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Let  $a$  be an atom and  $X$  be a set of atoms

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  - deciding whether  $a$  is in a stable model of  $P$  is  $NP^{NP}$ -complete
- For a disjunctive logic program  $P$ 
  - deciding whether  $X$  is a stable model of  $P$  is  $co-NP$ -complete
  - deciding whether  $a$  is in a stable model of  $P$  is  $NP^{NP}$ -complete
- For a disjunctive logic program  $P$  with optimization statements
  - deciding whether  $X$  is an optimal stable model of  $P$  is  $co-NP^{NP}$ -complete
  - deciding whether  $a$  is in an optimal stable model of  $P$  is  $\Delta_3^P$ -complete
- For a propositional theory  $\Phi$ 
  - deciding whether  $X$  is a stable model of  $\Phi$  is  $co-NP$ -complete
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Let  $a$  be an atom and  $X$  be a set of atoms

- For a positive disjunctive logic program  $P$ 
  - deciding whether  $X$  is a stable model of  $P$  is  $co-NP$ -complete
  - deciding whether  $a$  is in a stable model of  $P$  is  $NP^{NP}$ -complete
- For a disjunctive logic program  $P$ 
  - deciding whether  $X$  is a stable model of  $P$  is  $co-NP$ -complete
  - deciding whether  $a$  is in a stable model of  $P$  is  $NP^{NP}$ -complete
- For a disjunctive logic program  $P$  with optimization statements
  - deciding whether  $X$  is an optimal stable model of  $P$  is  $co-NP^{NP}$ -complete
  - deciding whether  $a$  is in an optimal stable model of  $P$  is  $\Delta_3^P$ -complete
- For a propositional theory  $\Phi$ 
  - deciding whether  $X$  is a stable model of  $\Phi$  is  $co-NP$ -complete
  - deciding whether  $a$  is in a stable model of  $\Phi$  is  $NP^{NP}$ -complete

# Auf Wiedersehen!

*"Tomorrow isn't staying out  
I'll be back, without a doubt!"*

Pink panther

# Bibliography

- The following list of references is compiled from the open source bibliography available at

`https://github.com/krr-up/bibliography`

- Feel free to submit corrections via pull requests !

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