

Answer Set Programming: From Theory to Practice

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University of Potsdam



Outline

- 1 Introduction
- 2 Foundations
- 3 Grounding
- 4 Solving
- 5 Modeling
- 6 Engineering
- 7 Applications

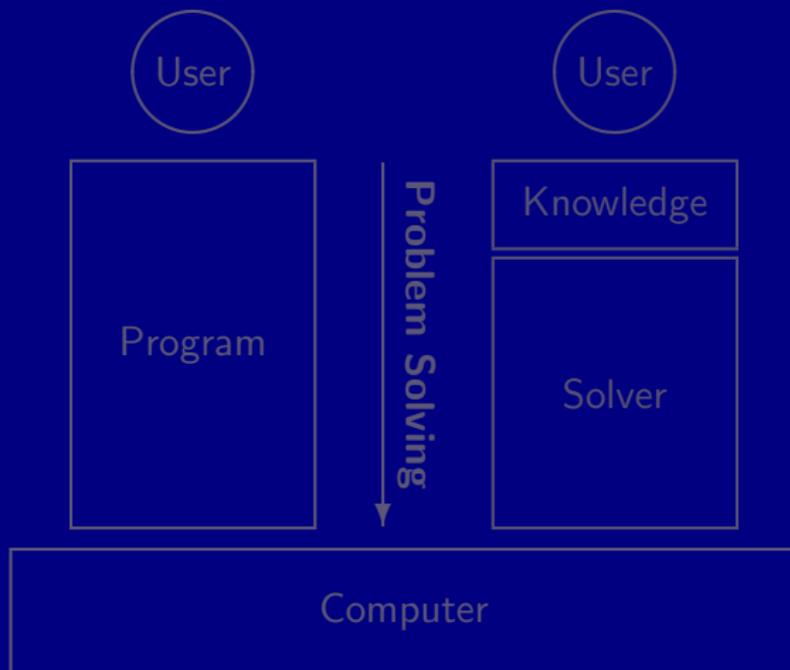
Introduction: Overview

- 1 Motivation
- 2 Nutshell
- 3 Evolution
- 4 Workflow
- 5 Usage

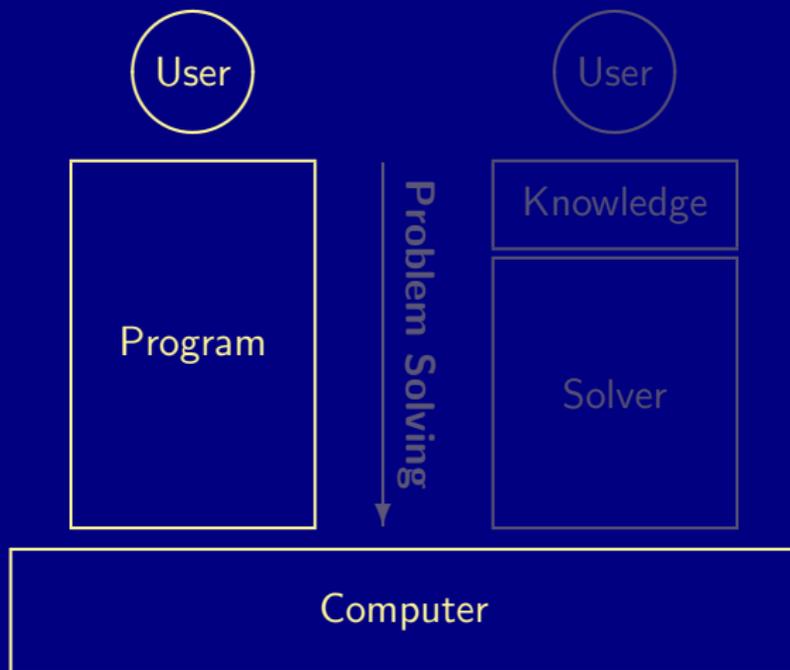
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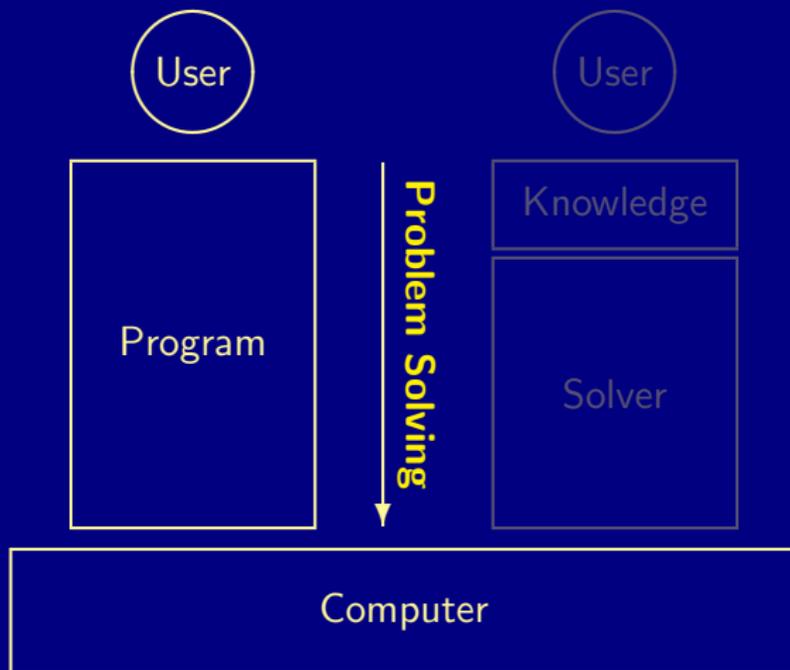
Traditional Software



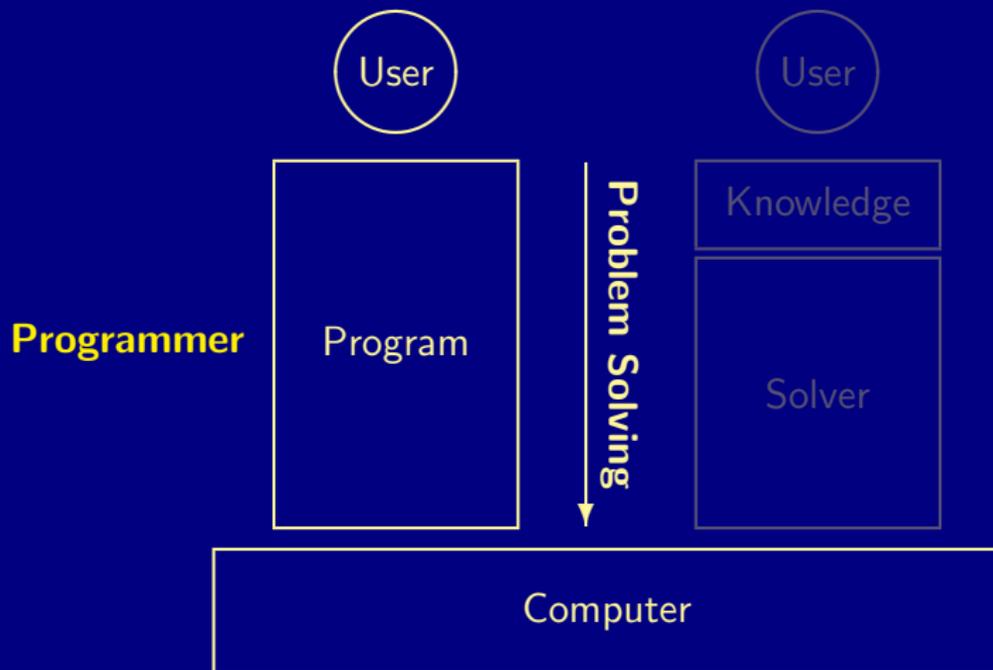
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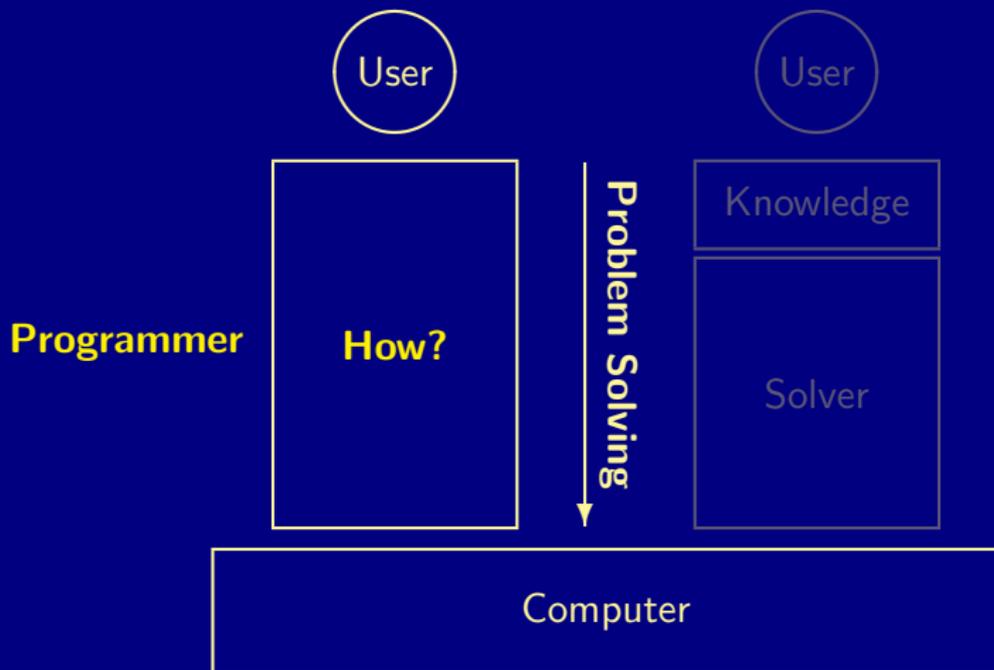
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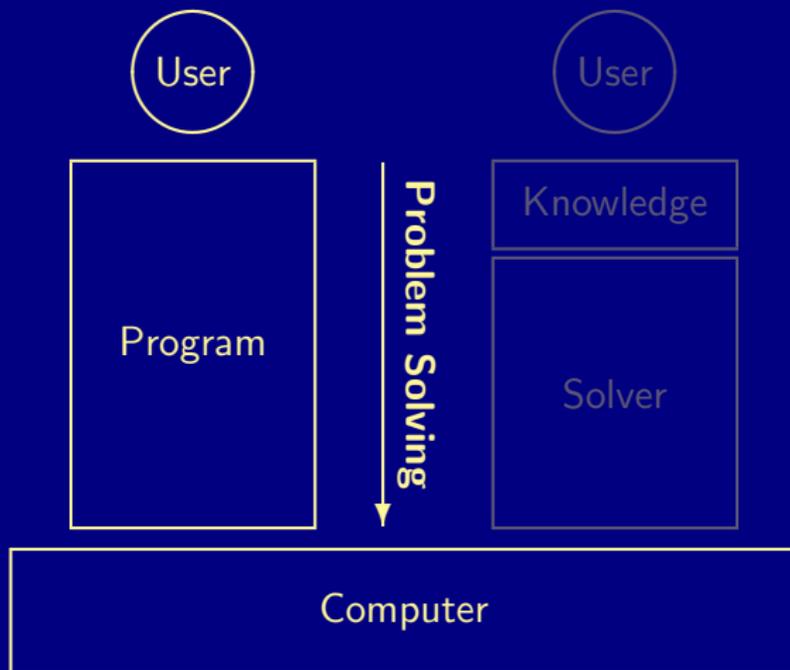
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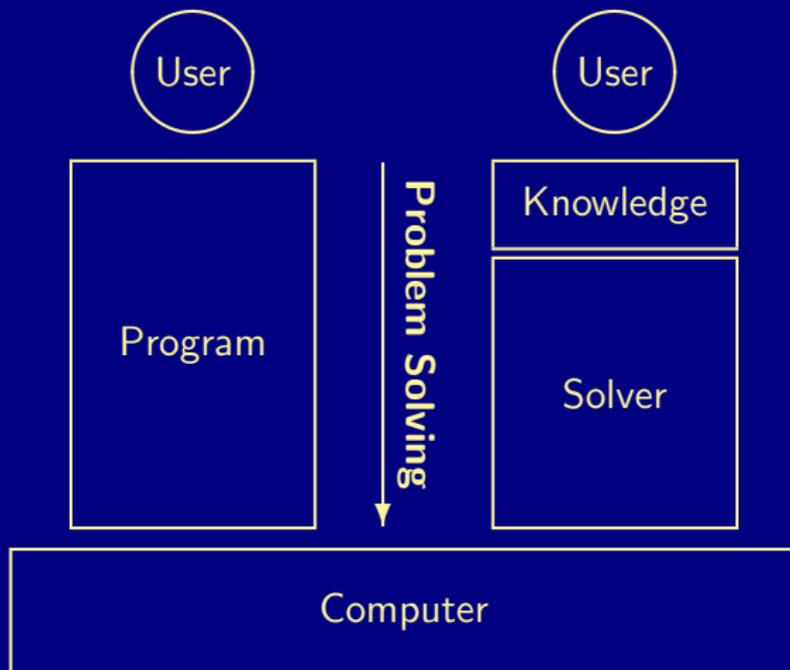
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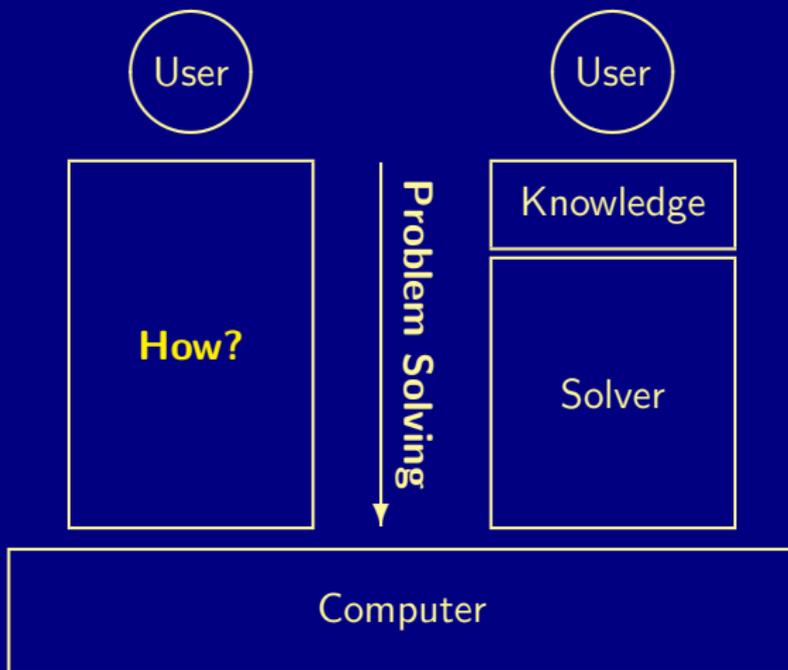
Knowledge-driven Software



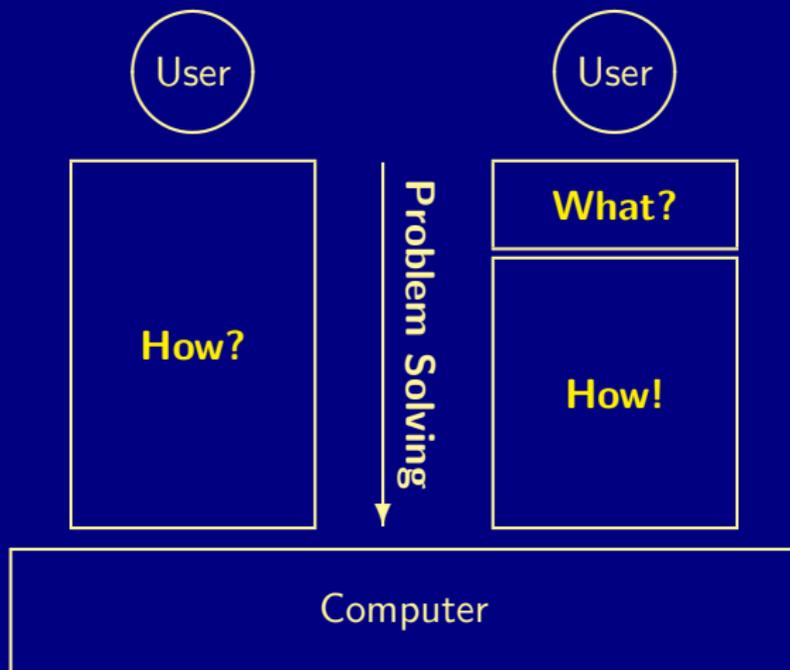
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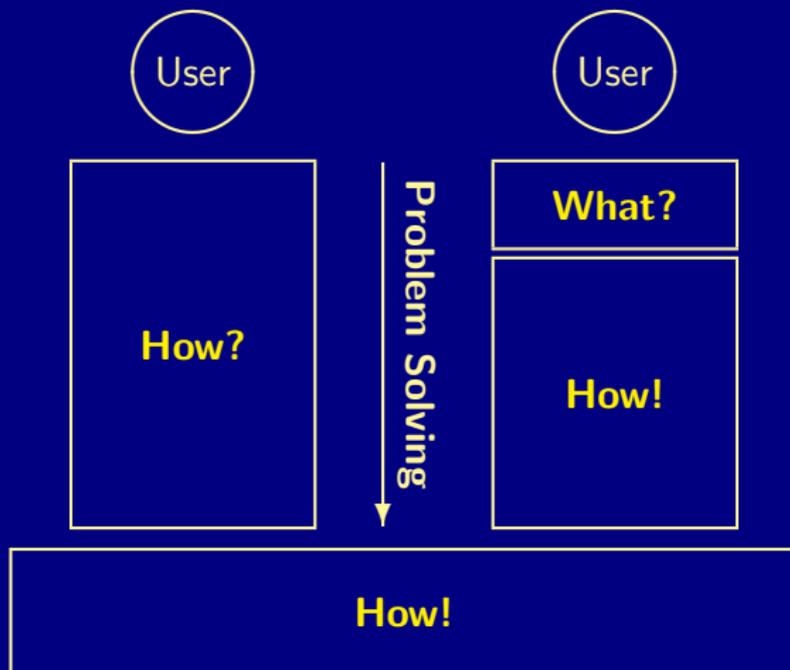
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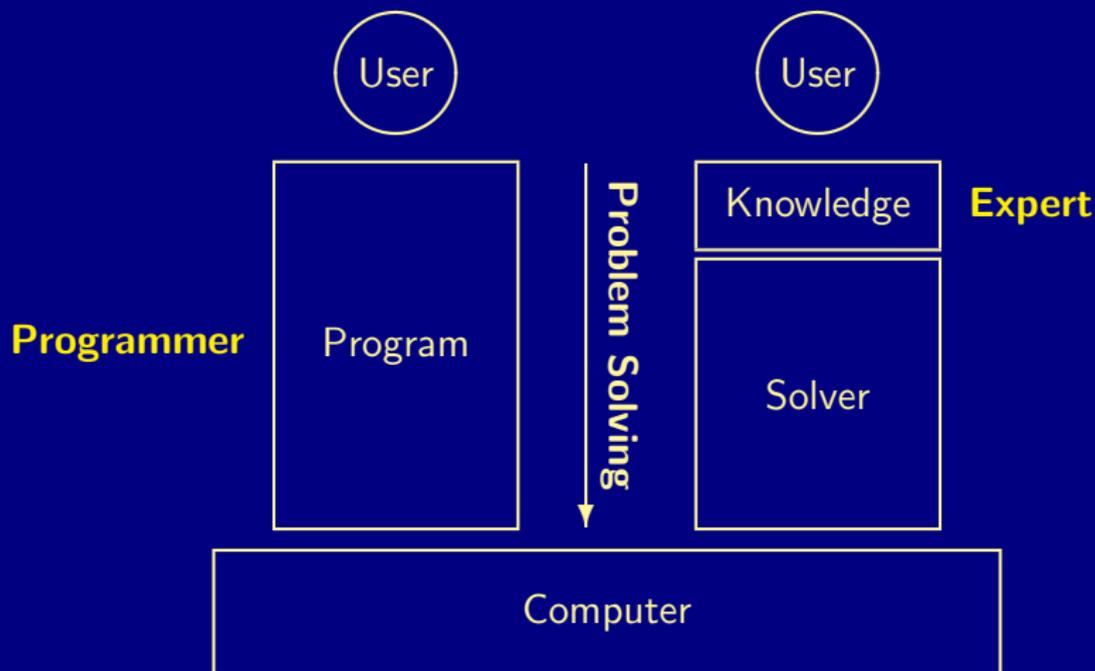
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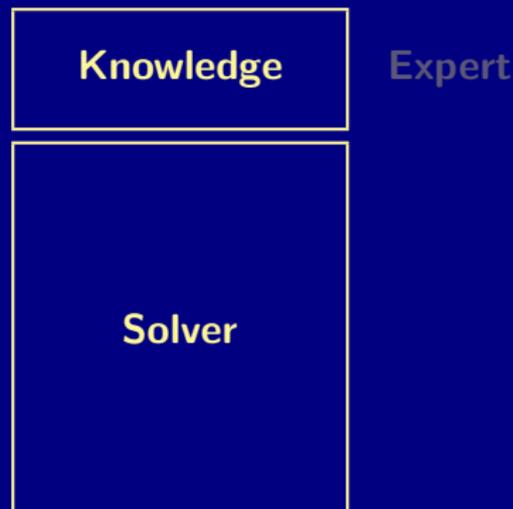
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What is the benefit?

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

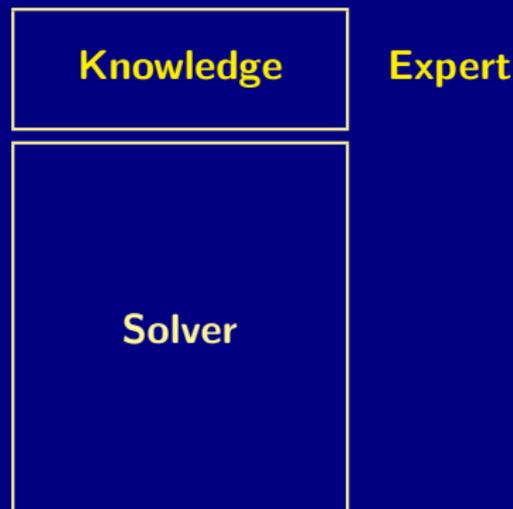
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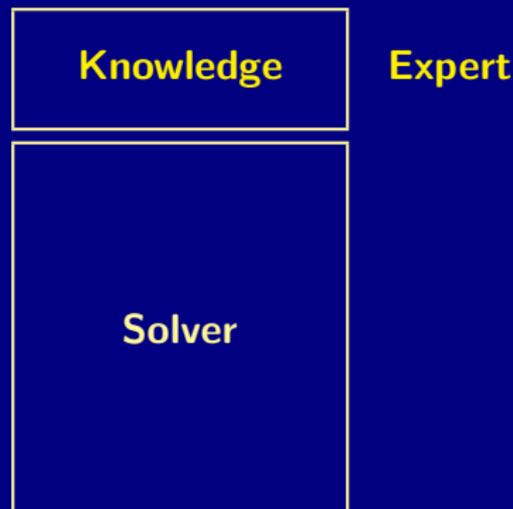
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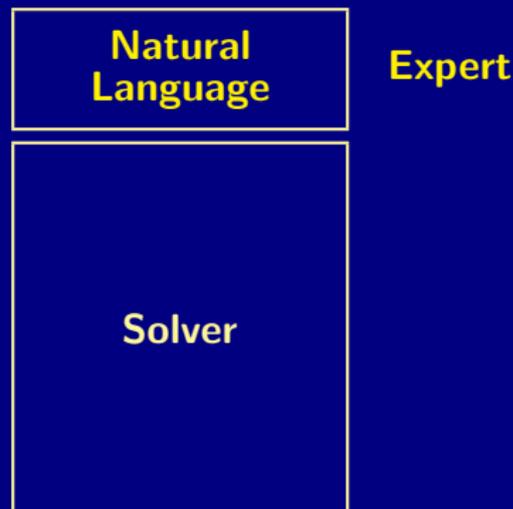
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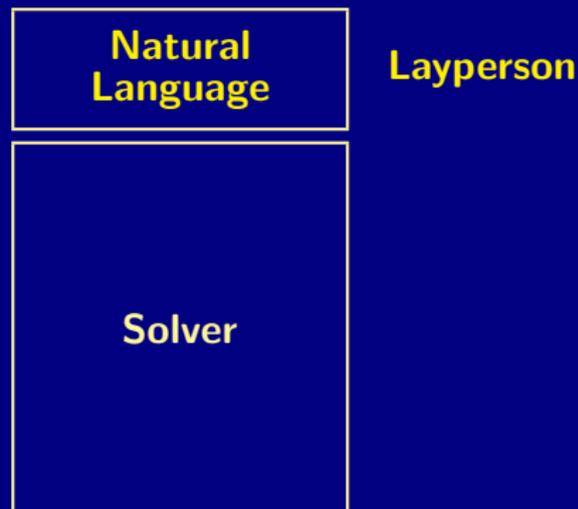
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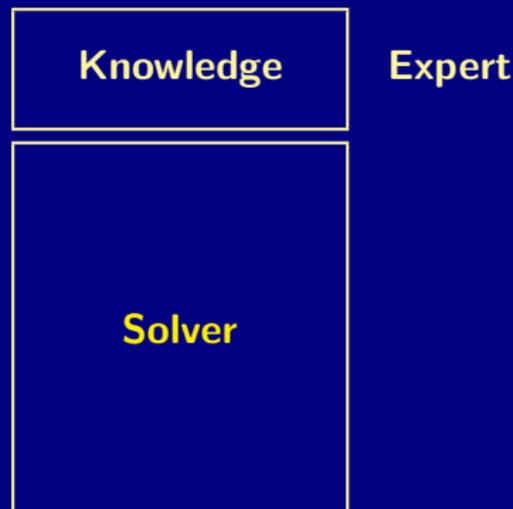
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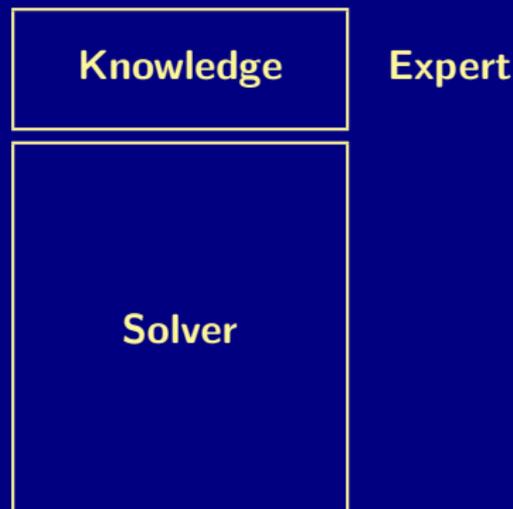
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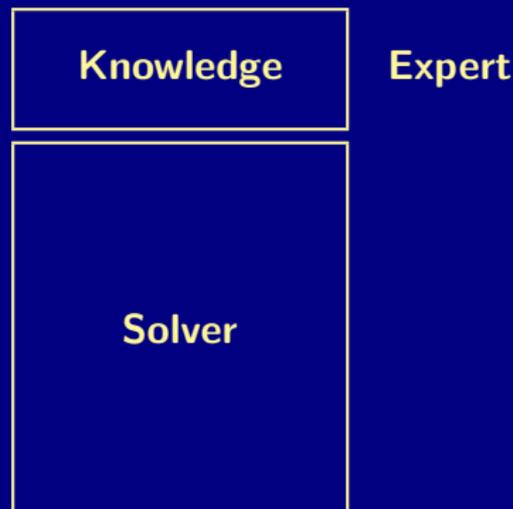
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 - Logic programming
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 - Satisfiability solving

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 - Debian, Ubuntu: Linux package configuration
 - Exeura: Call routing
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 - Nasa: Decision support for Space Shuttle
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- What problems are this? — **And indeed**

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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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- 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
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 - Qualitative and quantitative optimization
- Any industrial impact?
 - ASP Tech companies: DLV Systems and **Potassco Solutions**
 - Increasing interest in (large) companies

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 - Databases Closed world assumption
 - Logic programming Negation as failure
 - Non-monotonic reasoning
 - Auto-epistemic and Default logics, Circumscription

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 - Fix-point characterizations
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 - Auto-epistemic and Default logics, Circumscription
 - Extensions of first-order logic
 - Modalities, fix-points, second-order logic

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 - Well-founded and stable models semantics
 - ASP solving
 - “Stable models = Well-founded semantics + Branch”

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 - Modeling — Grounding — Solving
 - Icebreakers: `lparse` and `smodels`

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 - Bio-informatics, Linux Package Configuration, Music composition, Robotics, System Design, etc
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 - Roots: Logic of Here-and-There , G3

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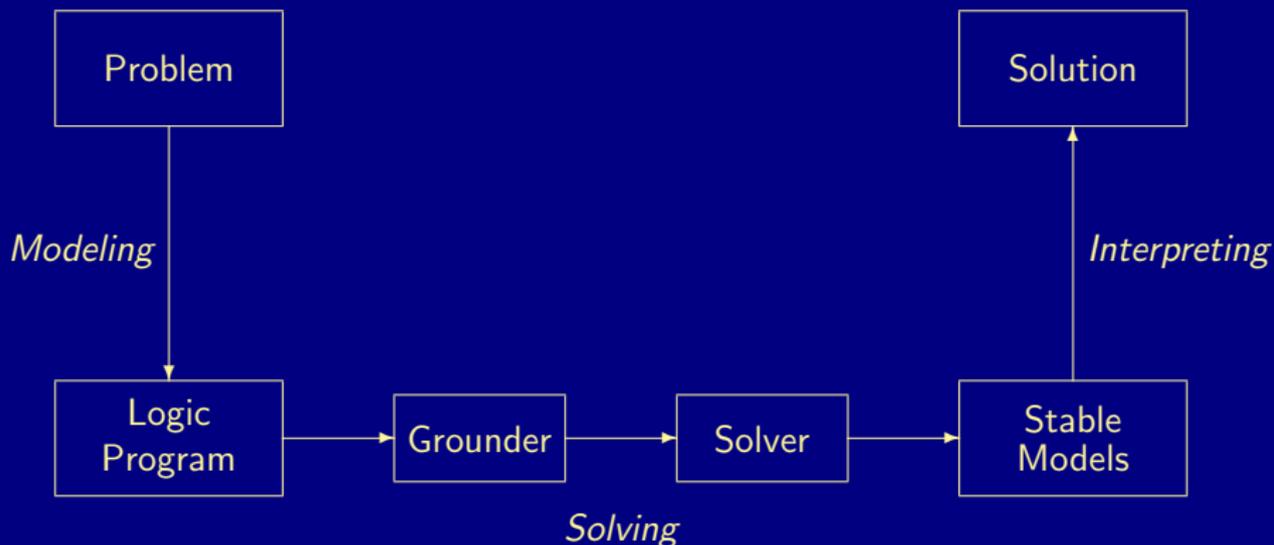
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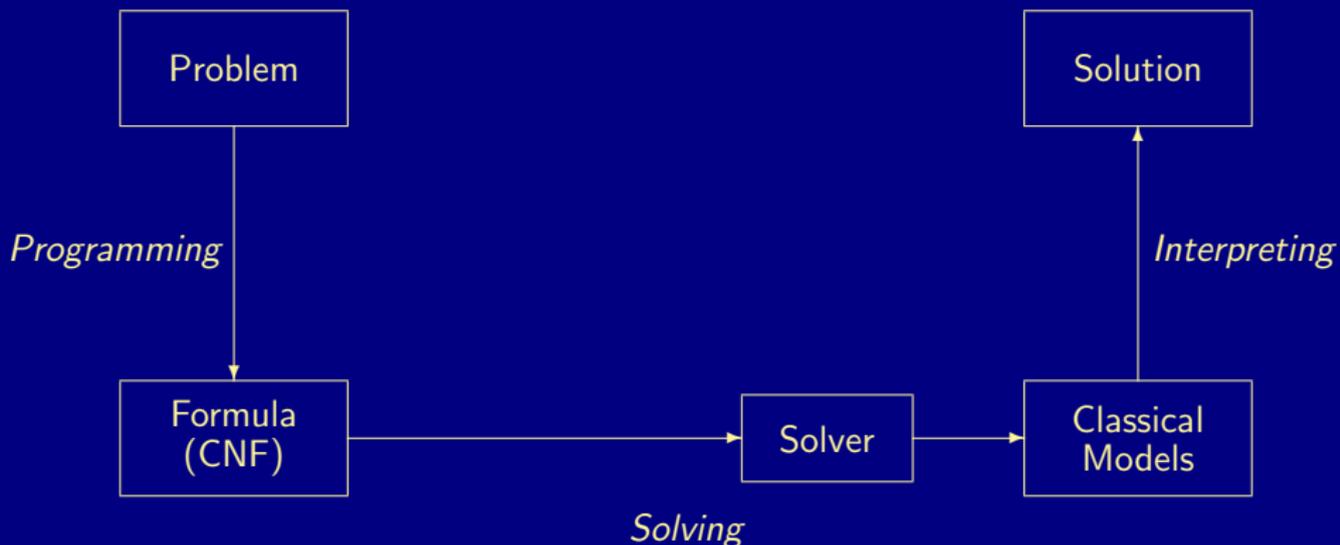
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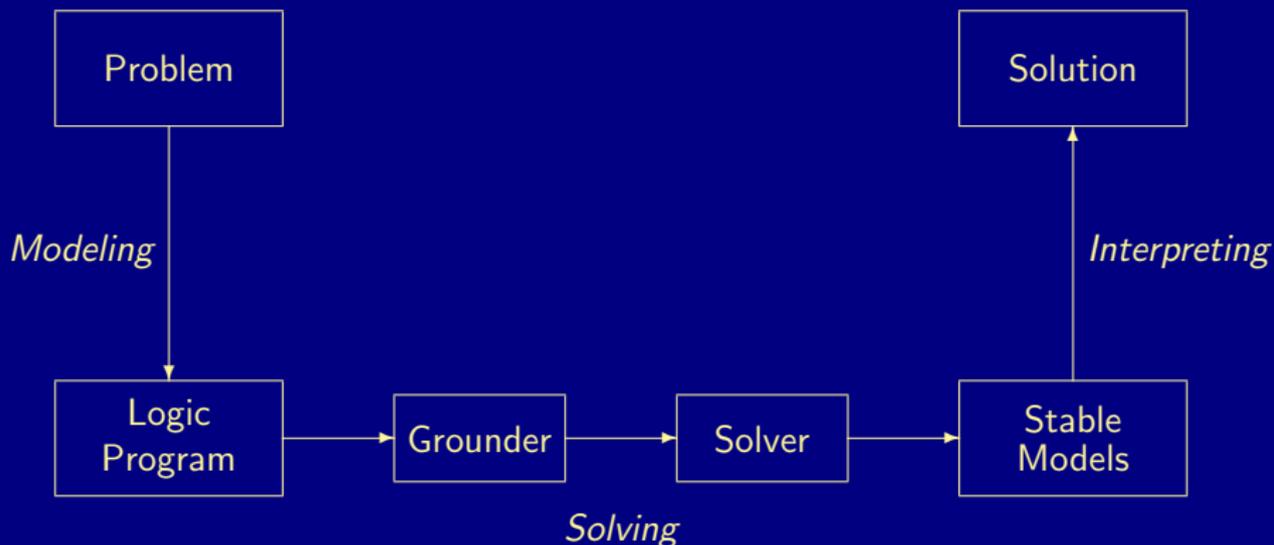
ASP modeling, grounding, and solving



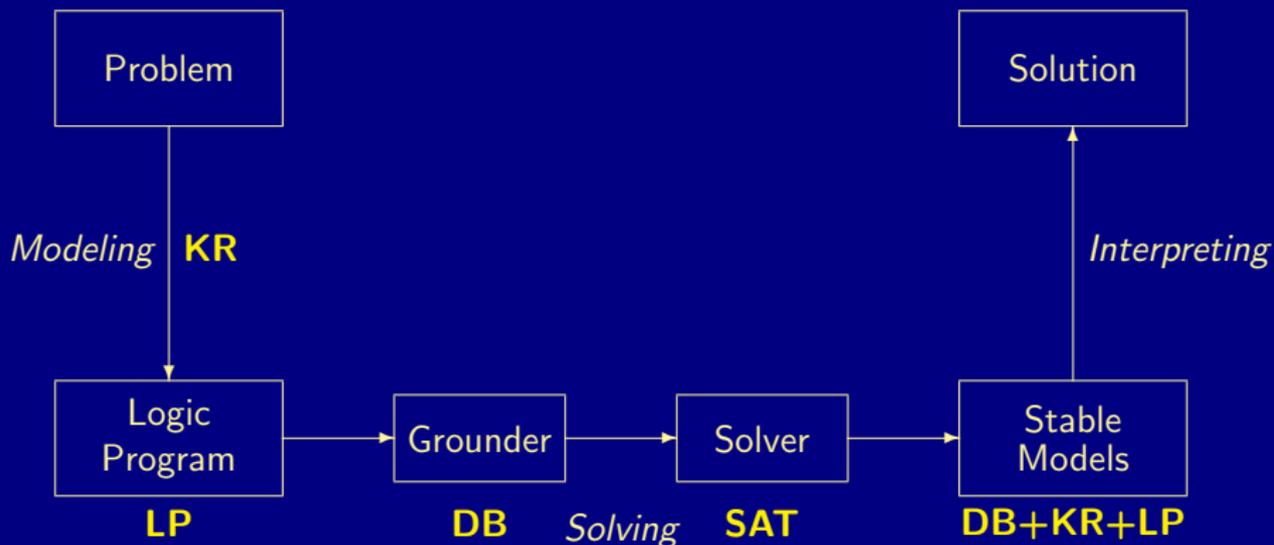
SAT solving



Rooting ASP solving



Rooting ASP solving



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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
 - Solve the original problem by solving its compilation
- ASP and Imperative language
 - Control continuously changing logic programs

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Two and a half sides of a coin

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- ASP and Imperative language
 - Control continuously changing logic programs

Foundations: Overview

6 Reduct-based characterization

7 Axiomatic characterization

8 Logical characterization

What is a stable model?

- Vladimir Lifschitz, Twelve Definitions of a Stable Model, [3]
- Reduct-based characterization
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Propositional Normal Logic Programs

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

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

- 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
- Semantics given by stable models, informally, models of P justifying each true atom by some rule in P

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Normal Logic Programs

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 - intuitive reading: **head** must be true if **body** holds
- Semantics given by **stable models**, informally, models of P justifying each true atom by some rule in P
- Disclaimer The following formalities apply to normal logic programs

Some truth tabling, back to SAT

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | |
| F | F | T | |
| F | T | F | |
| F | T | T | |
| T | F | F | |
| T | F | T | |
| T | T | F | |
| T | T | T | |

Some truth tabling, back to SAT

| <i>a</i> | <i>b</i> | <i>c</i> | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|---|
| F | F | F | $(\neg \mathbf{F} \rightarrow \mathbf{F}) \wedge (\mathbf{F} \rightarrow \mathbf{F})$ |
| F | F | T | $(\neg \mathbf{F} \rightarrow \mathbf{F}) \wedge (\mathbf{F} \rightarrow \mathbf{T})$ |
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| T | T | T | $(\neg \mathbf{T} \rightarrow \mathbf{T}) \wedge (\mathbf{T} \rightarrow \mathbf{T})$ |

Some truth tabling, back to SAT

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|--|
| F | F | F | (T \rightarrow F) \wedge (F \rightarrow F) |
| F | F | T | (T \rightarrow F) \wedge (F \rightarrow T) |
| F | T | F | (F \rightarrow F) \wedge (T \rightarrow F) |
| F | T | T | (F \rightarrow F) \wedge (T \rightarrow T) |
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| T | T | T | (F \rightarrow T) \wedge (T \rightarrow T) |

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| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|--|
| F | F | F | (T \rightarrow F) \wedge (F \rightarrow F) |
| F | F | T | (T \rightarrow F) \wedge (F \rightarrow T) |
| F | T | F | (F \rightarrow F) \wedge (T \rightarrow F) |
| F | T | T | (F \rightarrow F) \wedge (T \rightarrow T) |
| T | F | F | (T \rightarrow T) \wedge (F \rightarrow F) |
| T | F | T | (T \rightarrow T) \wedge (F \rightarrow T) |
| T | T | F | (F \rightarrow T) \wedge (T \rightarrow F) |
| T | T | T | (F \rightarrow T) \wedge (T \rightarrow T) |

Some truth tabling, back to SAT

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|--|
| F | F | F | F \wedge (F \rightarrow F) |
| F | F | T | F \wedge (F \rightarrow T) |
| F | T | F | (F \rightarrow F) \wedge F |
| F | T | T | (F \rightarrow F) \wedge (T \rightarrow T) |
| T | F | F | (T \rightarrow T) \wedge (F \rightarrow F) |
| T | F | T | (T \rightarrow T) \wedge (F \rightarrow T) |
| T | T | F | (F \rightarrow T) \wedge F |
| T | T | T | (F \rightarrow T) \wedge (T \rightarrow T) |

Some truth tabling, back to SAT

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|--|
| F | F | F | F \wedge (F \rightarrow F) |
| F | F | T | F \wedge (F \rightarrow T) |
| F | T | F | (F \rightarrow F) \wedge F |
| F | T | T | (F \rightarrow F) \wedge (T \rightarrow T) |
| T | F | F | (T \rightarrow T) \wedge (F \rightarrow F) |
| T | F | T | (T \rightarrow T) \wedge (F \rightarrow T) |
| T | T | F | (F \rightarrow T) \wedge F |
| T | T | T | (F \rightarrow T) \wedge (T \rightarrow T) |

Some truth tabling, back to SAT

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | F \wedge T |
| F | F | T | F \wedge T |
| F | T | F | T \wedge F |
| F | T | T | T \wedge T |
| T | F | F | T \wedge T |
| T | F | T | T \wedge T |
| T | T | F | T \wedge F |
| T | T | T | T \wedge T |

Some truth tabling, back to SAT

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | F |
| F | F | T | F |
| F | T | F | F |
| F | T | T | T |
| T | F | F | T |
| T | F | T | T |
| T | T | F | F |
| T | T | T | T |

Some truth tabling, back to SAT

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | F |
| F | F | T | F |
| F | T | F | F |
| F | T | T | T |
| T | F | F | T |
| T | F | T | T |
| T | T | F | F |
| T | T | T | T |

- We get four models: $\{b, c\}$, $\{a\}$, $\{a, c\}$, and $\{a, b, c\}$

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | |
| F | F | T | |
| F | T | F | |
| F | T | T | |
| T | F | F | |
| T | F | T | |
| T | T | F | |
| T | T | T | |

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|--|
| F | F | F | $(\neg \mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |
| F | F | T | $(\neg \mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |
| F | T | F | $(\neg \mathbf{T} \rightarrow a) \wedge (b \rightarrow c)$ |
| F | T | T | $(\neg \mathbf{T} \rightarrow a) \wedge (b \rightarrow c)$ |
| T | F | F | $(\neg \mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |
| T | F | T | $(\neg \mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |
| T | T | F | $(\neg \mathbf{T} \rightarrow a) \wedge (b \rightarrow c)$ |
| T | T | T | $(\neg \mathbf{T} \rightarrow a) \wedge (b \rightarrow c)$ |

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|---|
| F | F | F | (T $\rightarrow a$) \wedge (b $\rightarrow c$) |
| F | F | T | (T $\rightarrow a$) \wedge (b $\rightarrow c$) |
| F | T | F | (F $\rightarrow a$) \wedge (b $\rightarrow c$) |
| F | T | T | (F $\rightarrow a$) \wedge (b $\rightarrow c$) |
| T | F | F | (T $\rightarrow a$) \wedge (b $\rightarrow c$) |
| T | F | T | (T $\rightarrow a$) \wedge (b $\rightarrow c$) |
| T | T | F | (F $\rightarrow a$) \wedge (b $\rightarrow c$) |
| T | T | T | (F $\rightarrow a$) \wedge (b $\rightarrow c$) |

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|---|
| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $(\mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |
| F | T | T | $(\mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |
| T | F | F | $a \wedge (b \rightarrow c)$ |
| T | F | T | $a \wedge (b \rightarrow c)$ |
| T | T | F | $(\mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |
| T | T | T | $(\mathbf{F} \rightarrow a) \wedge (b \rightarrow c)$ |

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $(F \rightarrow a) \wedge (b \rightarrow c)$ |
| F | T | T | $(F \rightarrow a) \wedge (b \rightarrow c)$ |
| T | F | F | $a \wedge (b \rightarrow c)$ |
| T | F | T | $a \wedge (b \rightarrow c)$ |
| T | T | F | $(F \rightarrow a) \wedge (b \rightarrow c)$ |
| T | T | T | $(F \rightarrow a) \wedge (b \rightarrow c)$ |

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|----------|----------|----------|---|
| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $\mathbf{T} \wedge (b \rightarrow c)$ |
| F | T | T | $\mathbf{T} \wedge (b \rightarrow c)$ |
| T | F | F | $a \wedge (b \rightarrow c)$ |
| T | F | T | $a \wedge (b \rightarrow c)$ |
| T | T | F | $\mathbf{T} \wedge (b \rightarrow c)$ |
| T | T | T | $\mathbf{T} \wedge (b \rightarrow c)$ |

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| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $(b \rightarrow c)$ |
| F | T | T | $(b \rightarrow c)$ |
| T | F | F | $a \wedge (b \rightarrow c)$ |
| T | F | T | $a \wedge (b \rightarrow c)$ |
| T | T | F | $(b \rightarrow c)$ |
| T | T | T | $(b \rightarrow c)$ |

Reduct

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $(b \rightarrow c)$ |
| F | T | T | $(b \rightarrow c)$ |
| T | F | F | $a \wedge (b \rightarrow c)$ |
| T | F | T | $a \wedge (b \rightarrow c)$ |
| T | T | F | $(b \rightarrow c)$ |
| T | T | T | $(b \rightarrow c)$ |

Reduct

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $(b \rightarrow c)$ |
| F | T | T | $(b \rightarrow c) \models$ |
| T | F | F | $a \wedge (b \rightarrow c) \models a$ |
| T | F | T | $a \wedge (b \rightarrow c) \models a$ |
| T | T | F | $(b \rightarrow c)$ |
| T | T | T | $(b \rightarrow c) \models$ |

Reduct

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|---|
| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $(b \rightarrow c)$ |
| F | T | T | $(b \rightarrow c) \models$ |
| T | F | F | $a \wedge (b \rightarrow c) \models a$ |
| T | F | T | $a \wedge (b \rightarrow c) \models a$ |
| T | T | F | $(b \rightarrow c)$ |
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Reduct

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ | |
|-----|-----|-----|---|-------------|
| F | F | F | $a \wedge (b \rightarrow c)$ | $\models a$ |
| F | F | T | $a \wedge (b \rightarrow c)$ | $\models a$ |
| F | T | F | $(b \rightarrow c)$ | \models |
| F | T | T | $(b \rightarrow c)$ | \models |
| T | F | F | $a \wedge (b \rightarrow c)$ | $\models a$ |
| T | F | T | $a \wedge (b \rightarrow c)$ | $\models a$ |
| T | T | F | $(b \rightarrow c)$ | \models |
| T | T | T | $(b \rightarrow c)$ | \models |

Reduct

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ | |
|-----|-----|-----|---|--------------------------|
| F | F | F | $a \wedge (b \rightarrow c)$ | |
| F | F | T | $a \wedge (b \rightarrow c)$ | |
| F | T | F | $(b \rightarrow c)$ | |
| F | T | T | $(b \rightarrow c)$ | |
| T | F | F | $a \wedge (b \rightarrow c)$ | $\models a$ Stable model |
| T | F | T | $a \wedge (b \rightarrow c)$ | |
| T | T | F | $(b \rightarrow c)$ | |
| T | T | T | $(b \rightarrow c)$ | |

Reduct

- We get one stable model: $\{a\}$

Some truth tabling, and now ASP

| a | b | c | $(\neg b \rightarrow a) \wedge (b \rightarrow c)$ |
|-----|-----|-----|--|
| F | F | F | $a \wedge (b \rightarrow c)$ |
| F | F | T | $a \wedge (b \rightarrow c)$ |
| F | T | F | $(b \rightarrow c)$ |
| F | T | T | $(b \rightarrow c)$ |
| T | F | F | $a \wedge (b \rightarrow c) \models a$ Stable model |
| T | F | T | $a \wedge (b \rightarrow c)$ |
| T | T | F | $(b \rightarrow c)$ |
| T | T | T | $(b \rightarrow c)$ |

Reduct

- We get one stable model: $\{a\}$
- Stable models = Smallest models of (respective) reducts

Outline

6 Reduct-based characterization

7 Axiomatic characterization

8 Logical characterization

Propositional Normal Logic Programs

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

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

- a and all b_i, c_j are atoms (propositional variables)
 - $\leftarrow, ,, \neg$ denote if, and, and default negation
 - intuitive reading: head must be true if body holds
- Semantics given by stable models, informally, sets X of atoms such that
 - X is a (classical) model of P and
 - each atom in X is justified by some rule in P

Logic Programs

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 - each atom in X is justified by some rule in P

Logic Programs as Propositional Formulas

$$P = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow a, \neg c \quad x \leftarrow y \quad y \leftarrow x, b\}$$

$$CF(P) = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow (a \wedge \neg c) \vee y \quad y \leftarrow x \wedge b\} \\ \cup \{c \leftrightarrow \perp\}$$

$$LF(P) = \{(x \vee y) \rightarrow a \wedge \neg c\}$$

Classical models of $CF(P)$:

$\{b\}, \{b, c\}, \{b, x, y\}, \{b, c, x, y\}, \{a, c\}, \{a, b, c\}, \{a, x\}, \{a, c, x\},$
 $\{a, x, y\}, \{a, c, x, y\}, \{a, b, x, y\}, \{a, b, c, x, y\}$

Unsupported atoms

Unfounded atoms

Logic Programs as Propositional Formulas

$$P = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow a, \neg c \quad x \leftarrow y \quad y \leftarrow x, b\}$$

$$RF(P) = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow (a \wedge \neg c) \vee y \quad y \leftarrow x \wedge b\} \\ \cup \{c \leftrightarrow \perp\}$$

$$LF(P) = \{(x \vee y) \rightarrow a \wedge \neg c\}$$

Classical models of $RF(P)$: (only true atoms shown)

$\{b\}, \{b, c\}, \{b, x, y\}, \{b, c, x, y\}, \{a, c\}, \{a, b, c\}, \{a, x\}, \{a, c, x\},$
 $\{a, x, y\}, \{a, c, x, y\}, \{a, b, x, y\}, \{a, b, c, x, y\}$

- Unsupported atoms
- Unfounded atoms

Logic Programs as Propositional Formulas

$$P = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow a, \neg c \quad x \leftarrow y \quad y \leftarrow x, b\}$$

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$\{b\}$, $\{b, c\}$, $\{b, x, y\}$, $\{b, c, x, y\}$, $\{a, c\}$, $\{a, b, c\}$, $\{a, x\}$, $\{a, c, x\}$,
 $\{a, x, y\}$, $\{a, c, x, y\}$, $\{a, b, x, y\}$, $\{a, b, c, x, y\}$

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- Unfounded atoms

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$$P = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow a, \neg c \quad x \leftarrow y \quad y \leftarrow x, b\}$$

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Classical models of $RF(P)$:

$\{b\}$, $\{b, c\}$, $\{b, x, y\}$, $\{b, c, x, y\}$, $\{a, c\}$, $\{a, b, c\}$, $\{a, x\}$, $\{a, c, x\}$,
 $\{a, x, y\}$, $\{a, c, x, y\}$, $\{a, b, x, y\}$, $\{a, b, c, x, y\}$

- **Unsupported atoms**
- **Unfounded atoms**

Logic Programs as Propositional Formulas

$$P = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow a, \neg c \quad x \leftarrow y \quad y \leftarrow x, b\}$$

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 $\{a, x, y\}, \{a, c, x, y\}, \{a, b, x, y\}, \{a, b, c, x, y\}$

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Logic Programs as Propositional Formulas

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$\{b\}, \{b, c\}, \{b, x, y\}, \{b, c, x, y\}, \{a, c\}, \{a, b, c\}, \{a, x\}, \{a, c, x\},$
 $\{a, x, y\}, \{a, c, x, y\}, \{a, b, x, y\}, \{a, b, c, x, y\}$

- ~~Unsupported atoms~~
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Logic Programs as Propositional Formulas

$$P = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow a, \neg c \quad x \leftarrow y \quad y \leftarrow x, b\}$$

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$$LF(P) = \{(x \vee y) \rightarrow a \wedge \neg c\}$$

Classical models of $CF(P) \cup LF(P)$:

$\{b\}$, $\{b, c\}$, $\{b, x, y\}$, $\{b, c, x, y\}$, $\{a, c\}$, $\{a, b, c\}$, $\{a, x\}$, $\{a, c, x\}$,
 $\{a, x, y\}$, $\{a, c, x, y\}$, $\{a, b, x, y\}$, $\{a, b, c, x, y\}$

- ~~Unsupported atoms~~
- ~~Unfounded atoms~~

Logic Programs as Propositional Formulas

$$P = \{a \leftarrow \neg b \quad b \leftarrow \neg a \quad x \leftarrow a, \neg c \quad x \leftarrow y \quad y \leftarrow x, b\}$$

$$CF(P) = \{a \leftrightarrow (\bigvee_{(a \leftarrow B) \in P} BF(B)) \mid a \in A(P)\}$$

$$BF(B) = \bigwedge_{b \in B \cap A(P)} b \wedge \bigwedge_{\neg c \in B} \neg c$$

$$LF(P) = \{(\bigvee_{a \in L} a) \rightarrow (\bigvee_{a \in L, (a \leftarrow B) \in P, B \cap L = \emptyset} BF(B)) \mid L \in loop(P)\}$$

Classical models of $CF(P) \cup LF(P)$:

Theorem (Lin and Zhao)

Let P be a normal logic program and $X \subseteq A(P)$.

Then, X is a stable model of P iff $X \models CF(P) \cup LF(P)$.

- Size of $CF(P)$ is **linear** in the size of P
- Size of $LF(P)$ may be **exponential** in the size of P

ASP and SAT

- $SAT = ASP + \text{Law of the excluded middle}$
- $ASP = SAT + \text{Completion and Loop formulas}$
- Note: Checking whether a propositional formula has a stable model is Σ_P^2 -complete

ASP and SAT

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ASP and SAT

- $SAT = ASP + \text{Law of the excluded middle}^1$
- $ASP = SAT + \text{Completion and Loop formulas}$
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¹For instance, ' $\{a\}$.' stands for ' $a \vee \neg a$ '.

ASP and SAT

- $SAT = ASP + \text{Law of the excluded middle}$
- $ASP = SAT + \text{Completion and Loop formulas}$
- Note Checking whether a propositional formula has a stable model is Σ_P^2 -complete

Outline

- 6 Reduct-based characterization
- 7 Axiomatic characterization
- 8 Logical characterization**

The logic of Here-and-There (HT)

- An **interpretation** is a pair $\langle H, T \rangle$ of sets of atoms with $H \subseteq T$
 - H is called “here” and
 - T is called “there”
- Note $\langle H, T \rangle$ is a simplified Kripke structure
- Intuition
 - H represents provably true atoms
 - T represents possibly true atoms
 - atoms not in T are false
- Idea
 - $\langle H, T \rangle \models \varphi \sim \varphi$ is provably true
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Satisfaction

- $\langle H, T \rangle \models a$ if $a \in H$ for any atom a
- $\langle H, T \rangle \models \varphi \wedge \psi$ if $\langle H, T \rangle \models \varphi$ and $\langle H, T \rangle \models \psi$
- $\langle H, T \rangle \models \varphi \vee \psi$ if $\langle H, T \rangle \models \varphi$ or $\langle H, T \rangle \models \psi$
- $\langle H, T \rangle \models \varphi \rightarrow \psi$ if $\langle X, T \rangle \models \varphi$ implies $\langle X, T \rangle \models \psi$
for both $X = H, T$

- Note $\langle H, T \rangle \models \neg\varphi$ if $\langle T, T \rangle \not\models \varphi$ since $\neg\varphi = \varphi \rightarrow \perp$

- An interpretation $\langle H, T \rangle$ is a model of φ , if $\langle H, T \rangle \models \varphi$

Satisfaction

- $\langle H, T \rangle \models a$ if $a \in H$ for any atom a
- $\langle H, T \rangle \models \varphi \wedge \psi$ if $\langle H, T \rangle \models \varphi$ and $\langle H, T \rangle \models \psi$
- $\langle H, T \rangle \models \varphi \vee \psi$ if $\langle H, T \rangle \models \varphi$ or $\langle H, T \rangle \models \psi$
- $\langle H, T \rangle \models \varphi \rightarrow \psi$ if $\langle X, T \rangle \models \varphi$ implies $\langle X, T \rangle \models \psi$
for both $X = H, T$
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Satisfaction

- $\langle H, T \rangle \models a$ if $a \in H$ for any atom a
- $\langle H, T \rangle \models \varphi \wedge \psi$ if $\langle H, T \rangle \models \varphi$ and $\langle H, T \rangle \models \psi$
- $\langle H, T \rangle \models \varphi \vee \psi$ if $\langle H, T \rangle \models \varphi$ or $\langle H, T \rangle \models \psi$
- $\langle H, T \rangle \models \varphi \rightarrow \psi$ if $\langle X, T \rangle \models \varphi$ implies $\langle X, T \rangle \models \psi$
for both $X = H, T$
- Note $\langle H, T \rangle \models \neg\varphi$ if $\langle T, T \rangle \not\models \varphi$ since $\neg\varphi = \varphi \rightarrow \perp$
- An interpretation $\langle H, T \rangle$ is a **model** of φ , if $\langle H, T \rangle \models \varphi$

Classical tautologies

| H | T | a | $\neg a$ | $a \vee \neg a$ | $\neg \neg a$ | $a \leftarrow \neg \neg a$ |
|-------------|-------------|-----|----------|-----------------|---------------|----------------------------|
| $\{a\}$ | $\{a\}$ | T | F | T | T | T |
| \emptyset | $\{a\}$ | F | F | F | T | F |
| \emptyset | \emptyset | F | T | T | F | T |

Equilibrium models

- A total interpretation $\langle T, T \rangle$ is an **equilibrium model** of a formula φ , if
 - 1 $\langle T, T \rangle \models \varphi$,
 - 2 $\langle H, T \rangle \not\models \varphi$ for all $H \subset T$
- T is called a **stable model** of φ
- Note $\langle T, T \rangle$ acts as a classical model
- Note $\langle H, T \rangle \models P$ iff $H \models P^T$ (P^T is the reduct of P by T)

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Grounding: Overview

Example

$d(a)$

$d(c)$

$d(d)$

$p(a, b)$

$p(b, c)$

$p(c, d)$

$p(X, Z) \leftarrow p(X, Y), p(Y, Z)$

$q(a)$

$q(b)$

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Ground instantiation

Let P be a logic program

- 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 $r \in P$: Set of variable-free rules obtained by replacing all variables in r by elements from \mathcal{T} :

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

where $var(r)$ stands for the set of all variables occurring in r ;
 θ is a (ground) substitution

- Ground instantiation of P : $ground(P) = \bigcup_{r \in P} ground(r)$

Ground instantiation

Let P be a logic program

- Let \mathcal{T} be a set of variable-free terms (also called Herbrand universe)
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$$P = \{ r(a, b) \leftarrow, r(b, c) \leftarrow, t(X, Y) \leftarrow r(X, Y) \}$$

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

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Grounding aims at reducing the ground instantiation

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- **Grounding** aims at reducing the ground instantiation

Stable models of programs with Variables

Let P be a normal logic program with variables

- A set X of (ground) atoms is a stable model of P ,
if $Cn(\text{ground}(P)^X) = X$

Stable models of programs with Variables

Let P be a normal logic program with variables

- A set X of (**ground**) atoms is a **stable model** of P ,
if $Cn(\mathit{ground}(P)^X) = X$

Safety

- ASP systems require that all rules in a program are safe
- A normal rule is **safe**, if each of its variables also occurs in some positive body literal
- A normal program is safe, if all of its rules are safe

Example

$d(a)$

$d(c)$

$d(d)$

$p(a, b)$

$p(b, c)$

$p(c, d)$

$p(X, Z) \leftarrow p(X, Y), p(Y, Z)$

$q(a)$

$q(b)$

$q(X) \leftarrow \neg r(X)$

$r(X) \leftarrow \neg q(X), d(X)$

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Example

Safe ?

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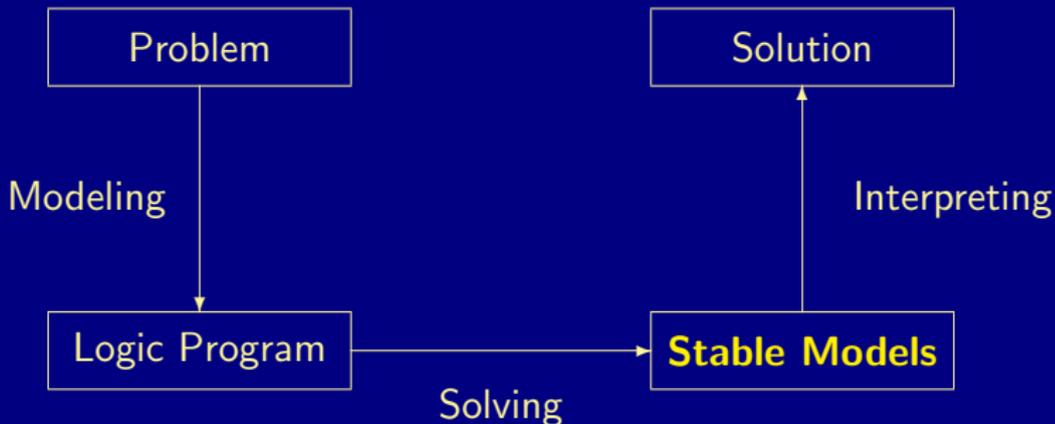


Solving: Overview

9 Conflict-driven constraint learning

10 Engine

Reasoning modes



Reasoning modes

- Satisfiability
- Enumeration[†]
- Projection[†]
- Intersection[‡]
- Union[‡]
- Optimization

- and combinations of them

[†] without solution recording

[‡] without solution enumeration

Outline

9 Conflict-driven constraint learning

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Towards conflict-driven search

Boolean constraint solving algorithms pioneered for SAT led to:

- Traditional DPLL-style approach
(DPLL stands for 'Davis-Putnam-Logemann-Loveland')
 - (Unit) propagation
 - (Chronological) backtracking
 - in ASP, eg *smodels*
- Modern CDCL-style approach
(CDCL stands for 'Conflict-Driven Constraint Learning')
 - (Unit) propagation
 - Conflict analysis (via resolution)
 - Learning + Backjumping + Assertion
 - in ASP, eg *clasp*

DPLL-style solving

loop

```

propagate                                // deterministically assign literals
if no conflict then
    if all variables assigned then return solution
    else decide                            // non-deterministically assign some literal
else
    if top-level conflict then return unsatisfiable
    else
        backtrack                          // unassign literals propagated after last decision
        flip                                // assign complement of last decision literal

```

CDCL-style solving

loop

```

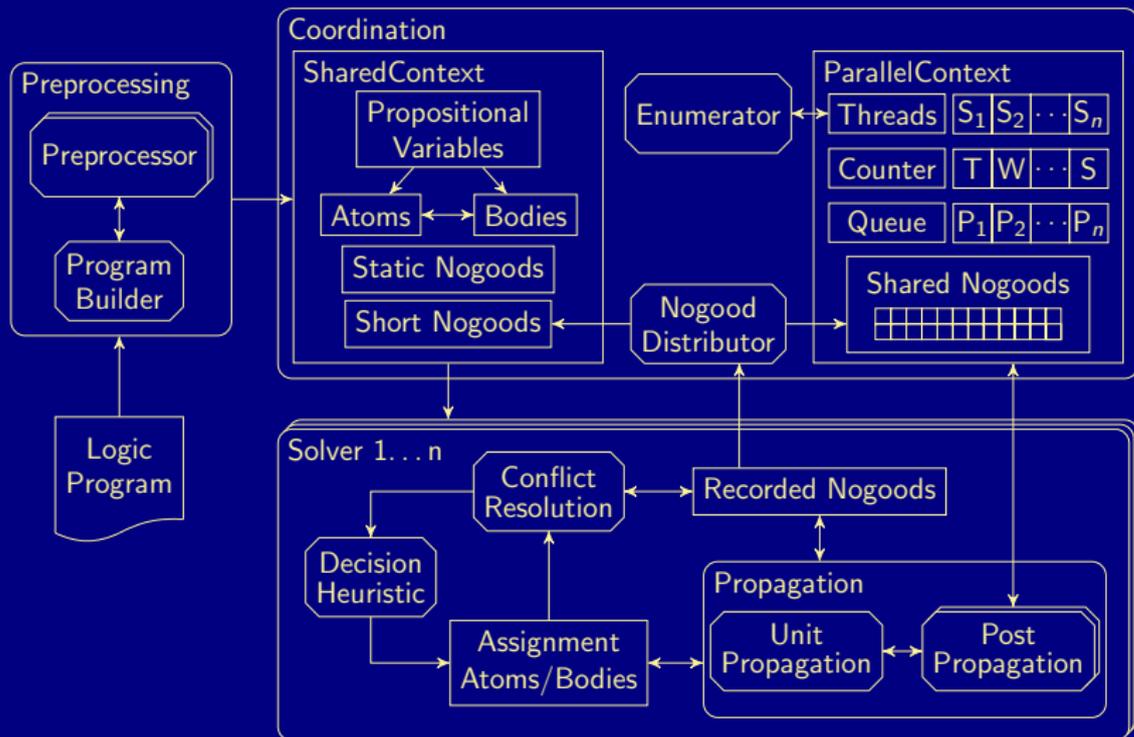
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if no conflict then
    if all variables assigned then return solution
    else decide                                 // non-deterministically assign some literal
else
    if top-level conflict then return unsatisfiable
    else
        analyze                                 // analyze conflict and add conflict constraint
        backjump                               // unassign literals until conflict constraint is unit

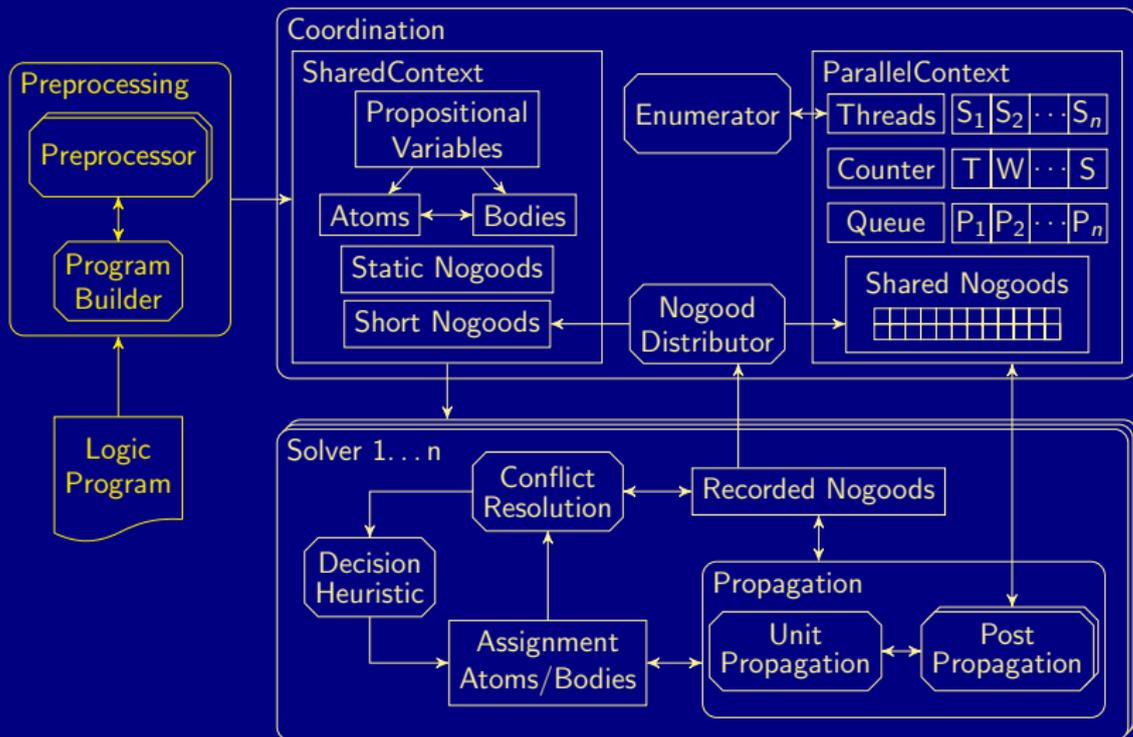
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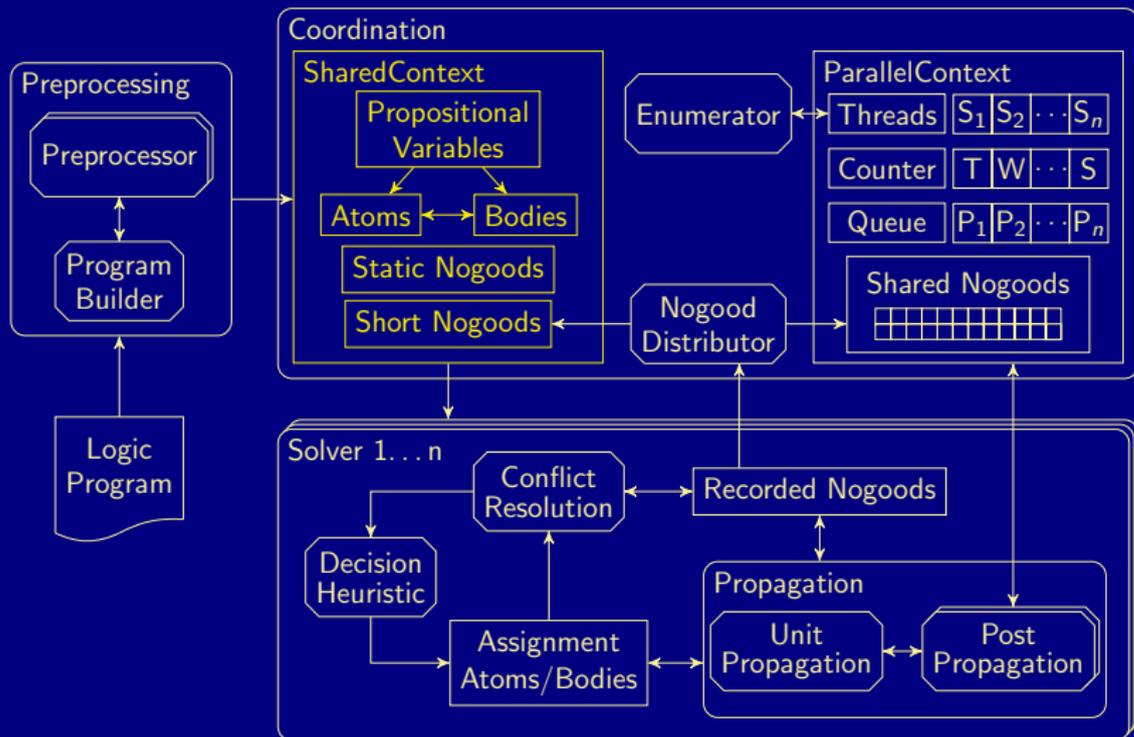
Outline

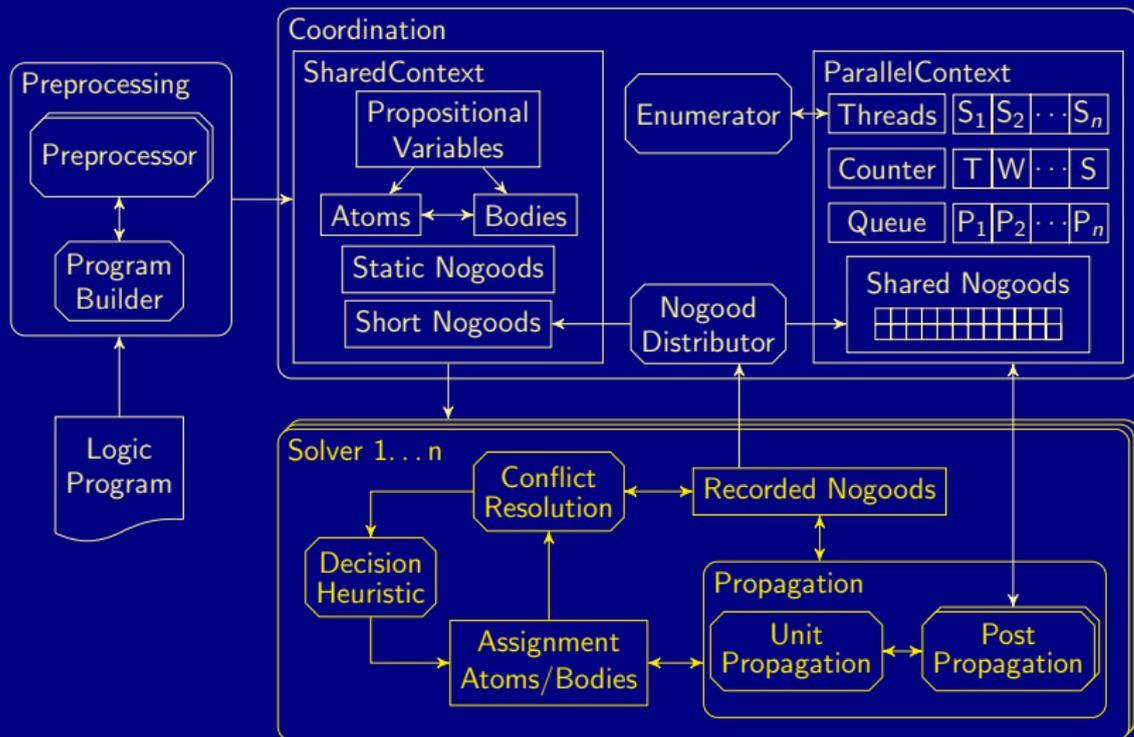
9 Conflict-driven constraint learning

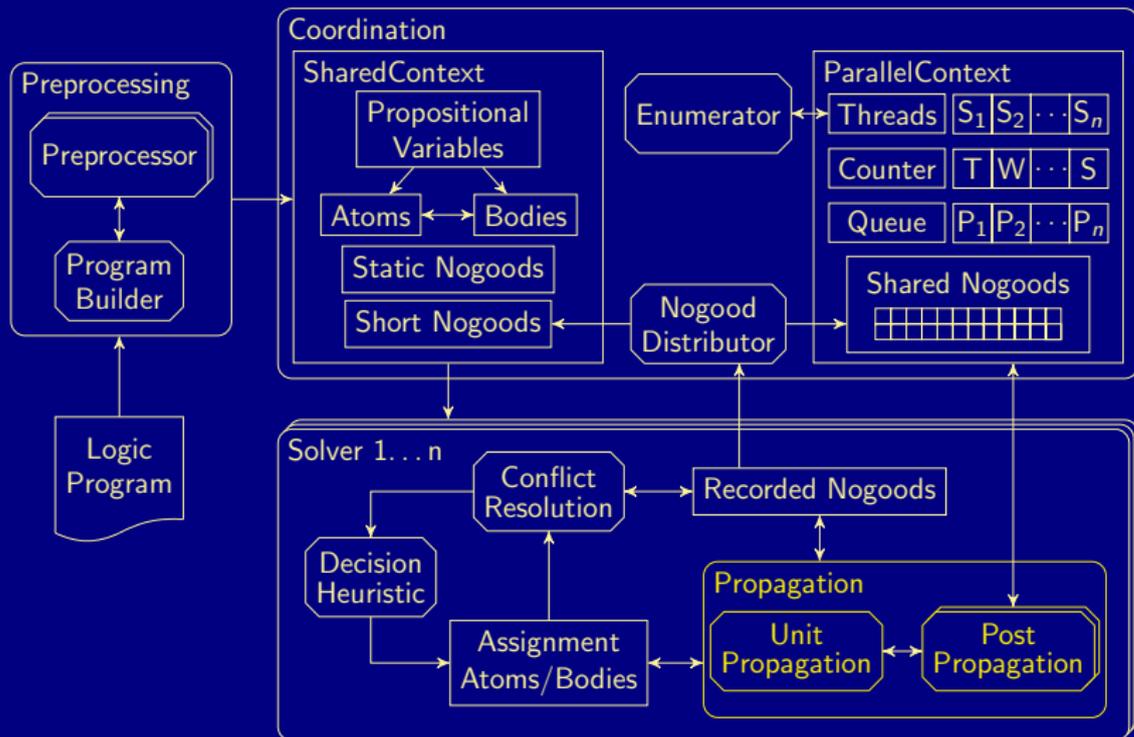
10 Engine

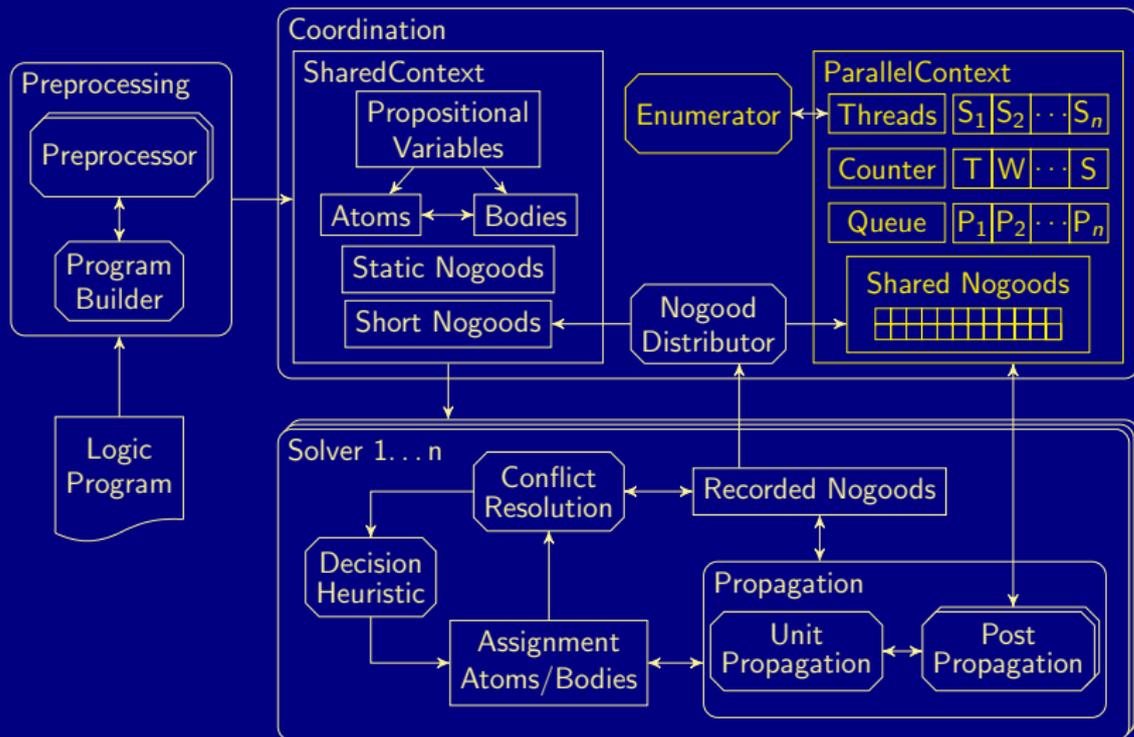
Multi-threaded architecture of *clasp*

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Modeling: Overview

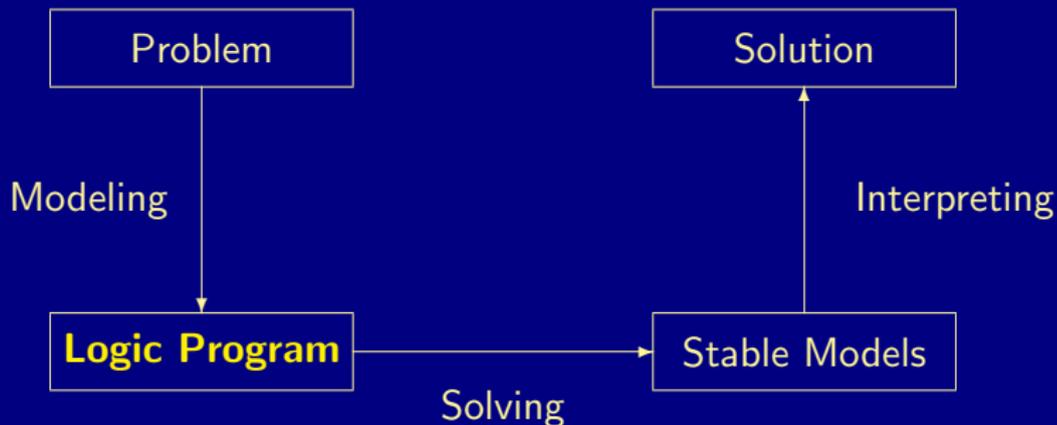
11 Elaboration tolerance

12 ASP solving process

13 Methodology

14 Case studies

Extended syntax



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]`
`#minimize { C@42 : q(X), p(X,C) }`

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`s(Y) :- r(Y), 2 #sum{ X : p(X,Y), q(X) } 7.`

- Multi-objective optimization

`:~ q(X), p(X,C). [C@42]`

`#minimize { C@42 : q(X), p(X,C) }`

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

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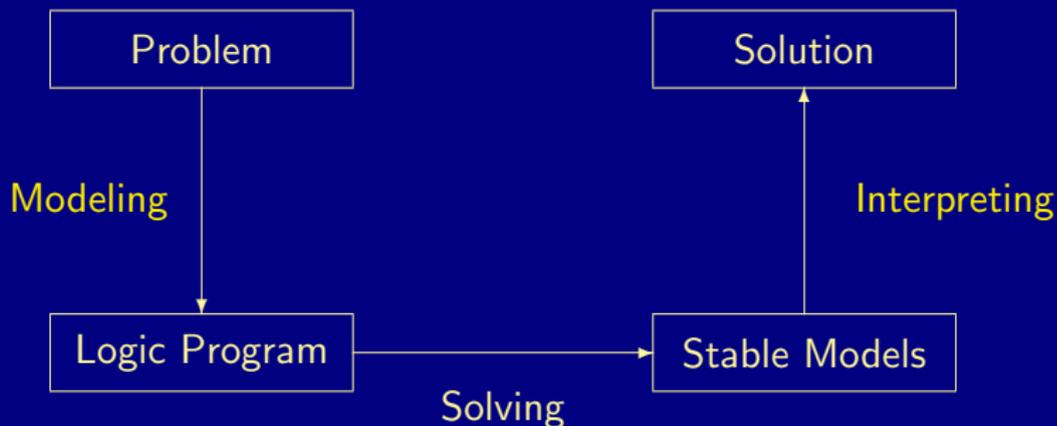
Language constructs

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- Rules `p(X) :- q(X), not r(X).`
- Conditional literals `p :- q(X) : r(X).`
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Language constructs

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Modeling and Interpreting



Outline

11 Elaboration tolerance

12 ASP solving process

13 Methodology

14 Case studies

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

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Outline

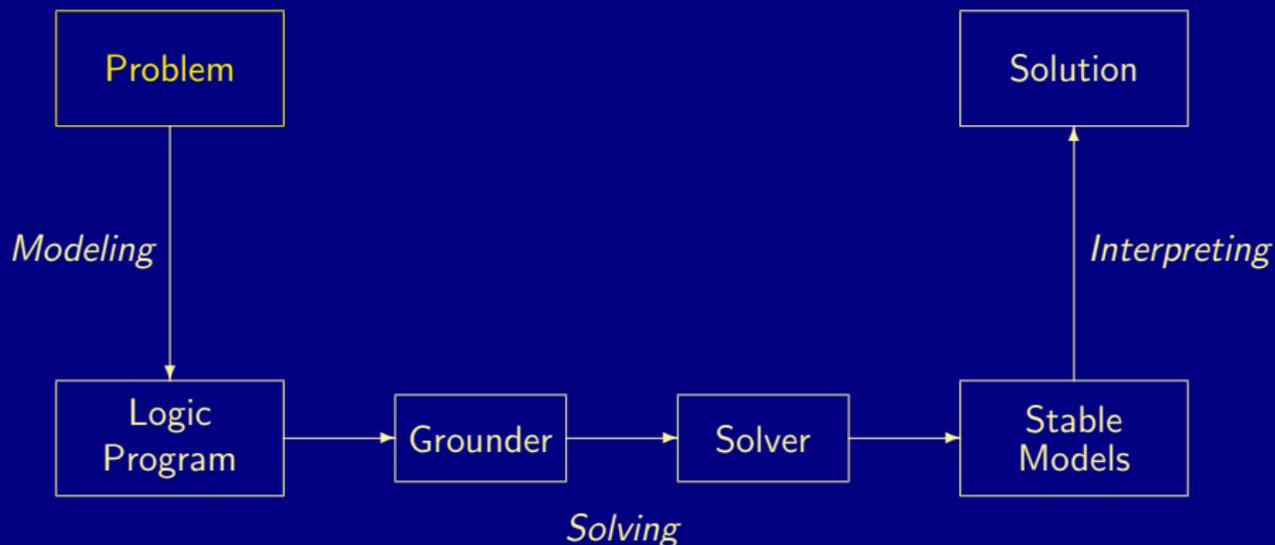
11 Elaboration tolerance

12 ASP solving process

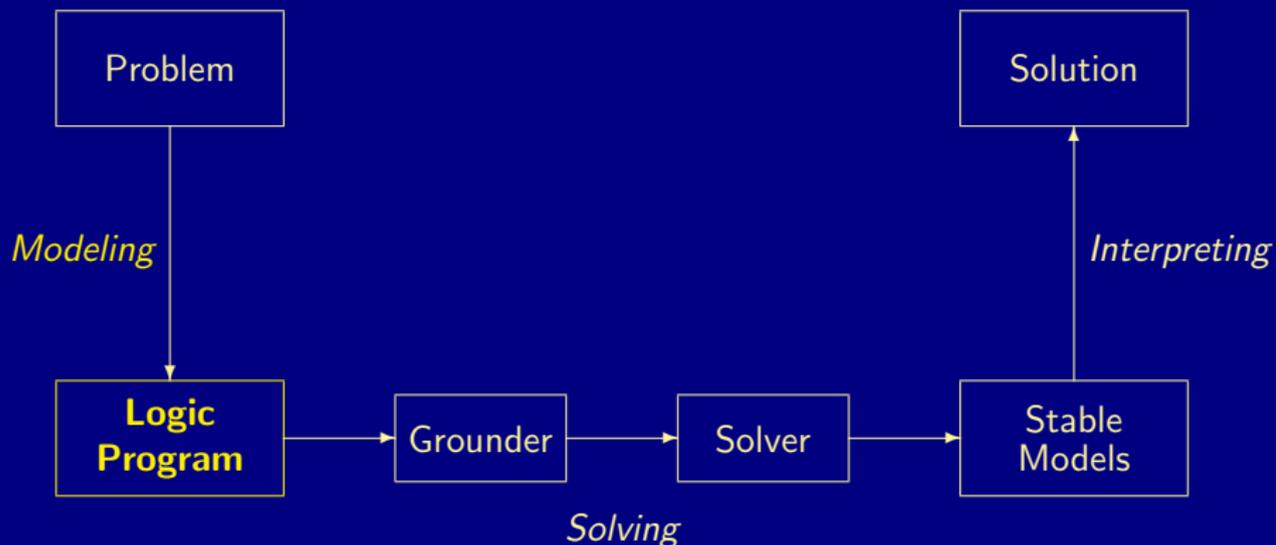
13 Methodology

14 Case studies

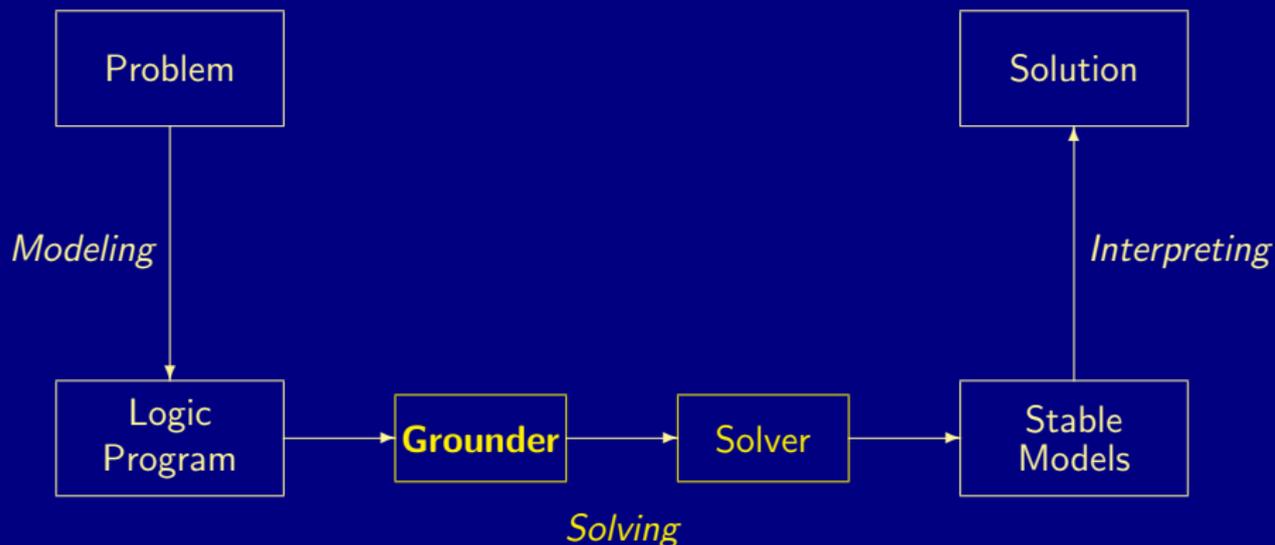
ASP solving process



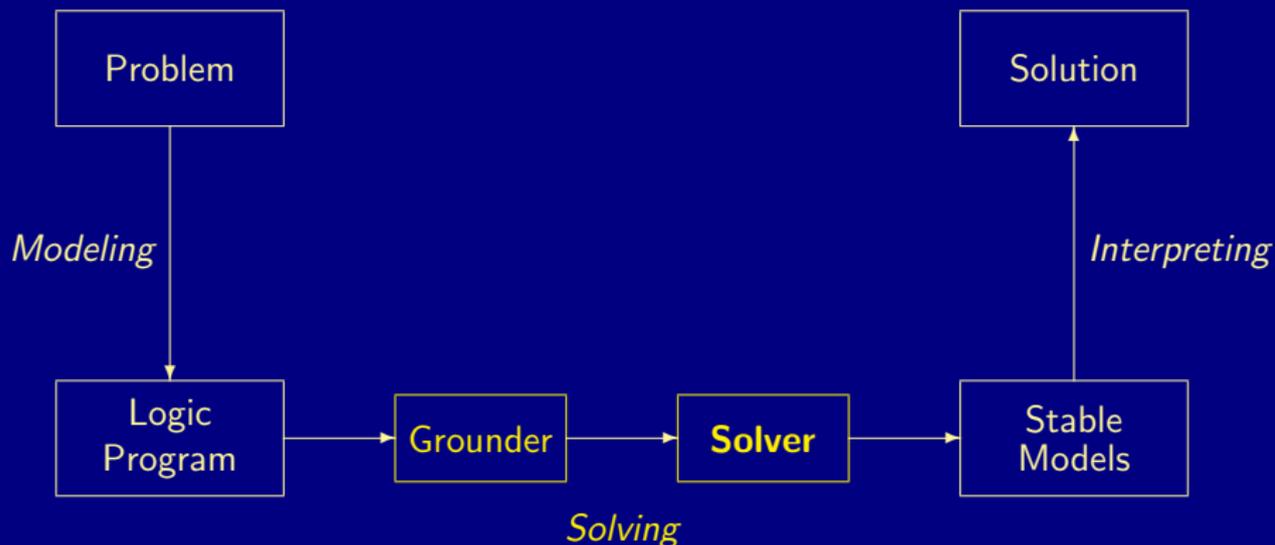
ASP solving process



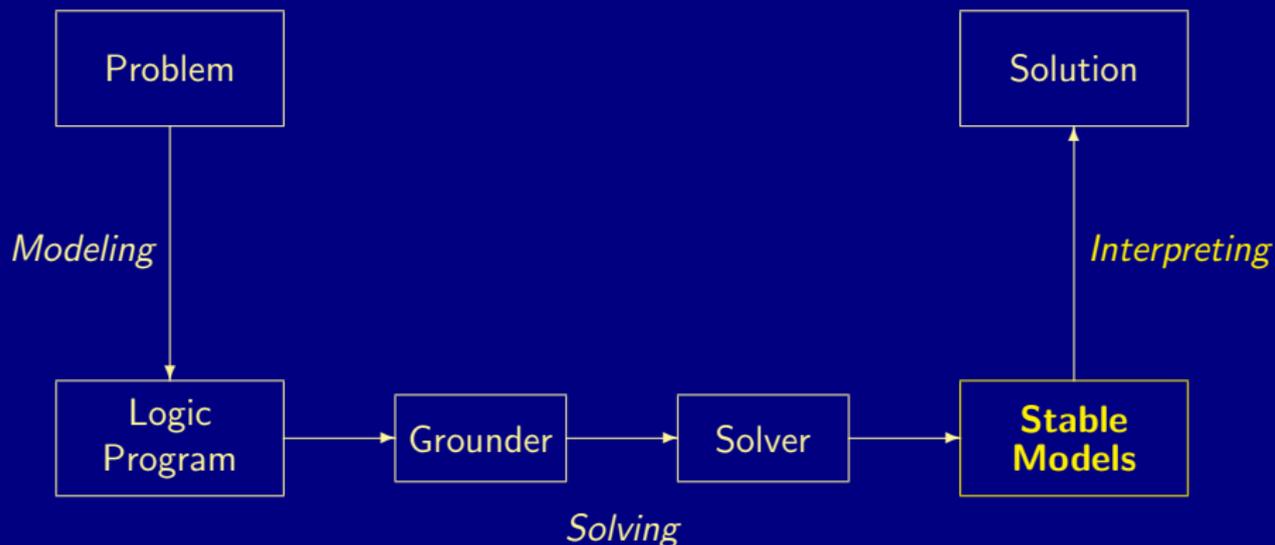
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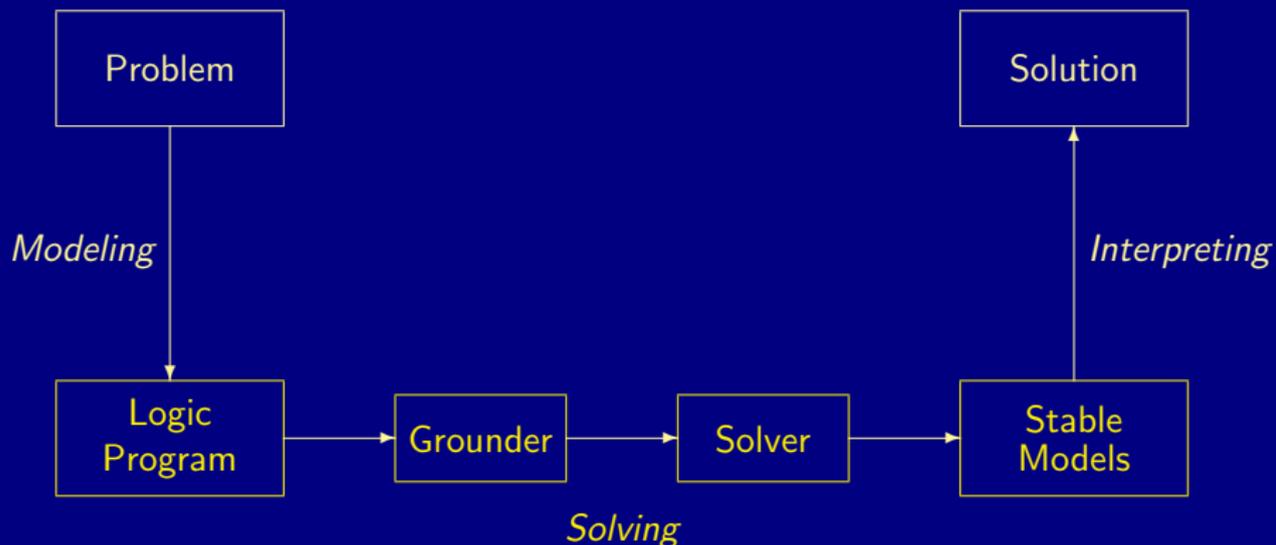
ASP solving process



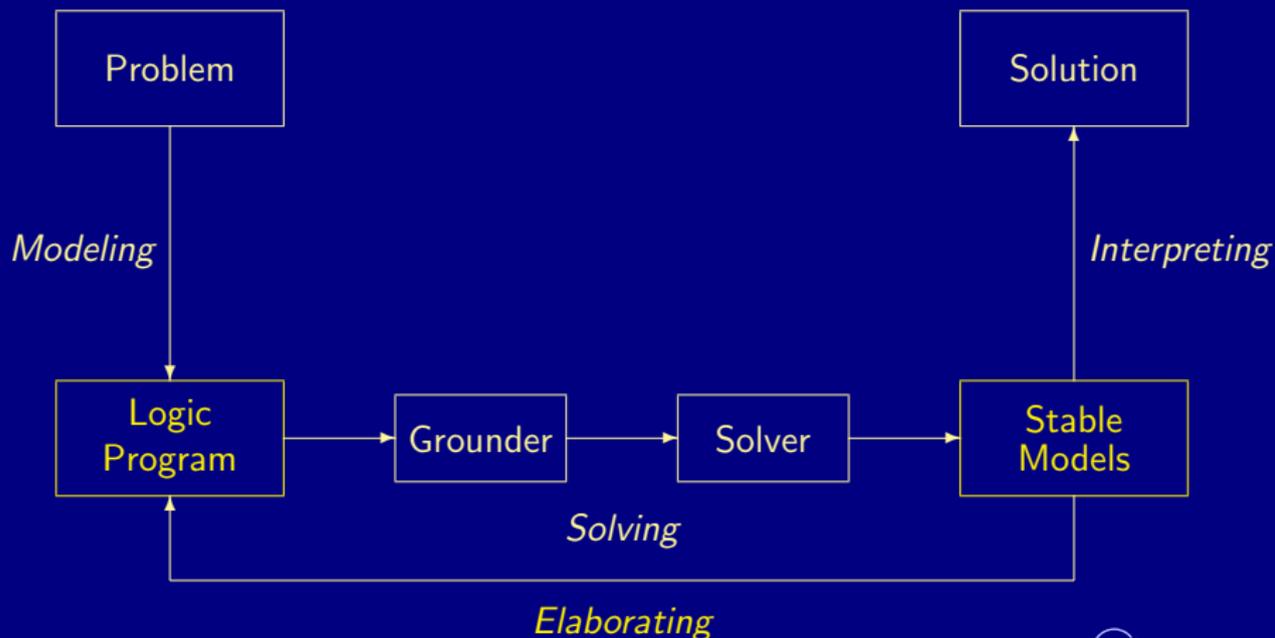
ASP solving process



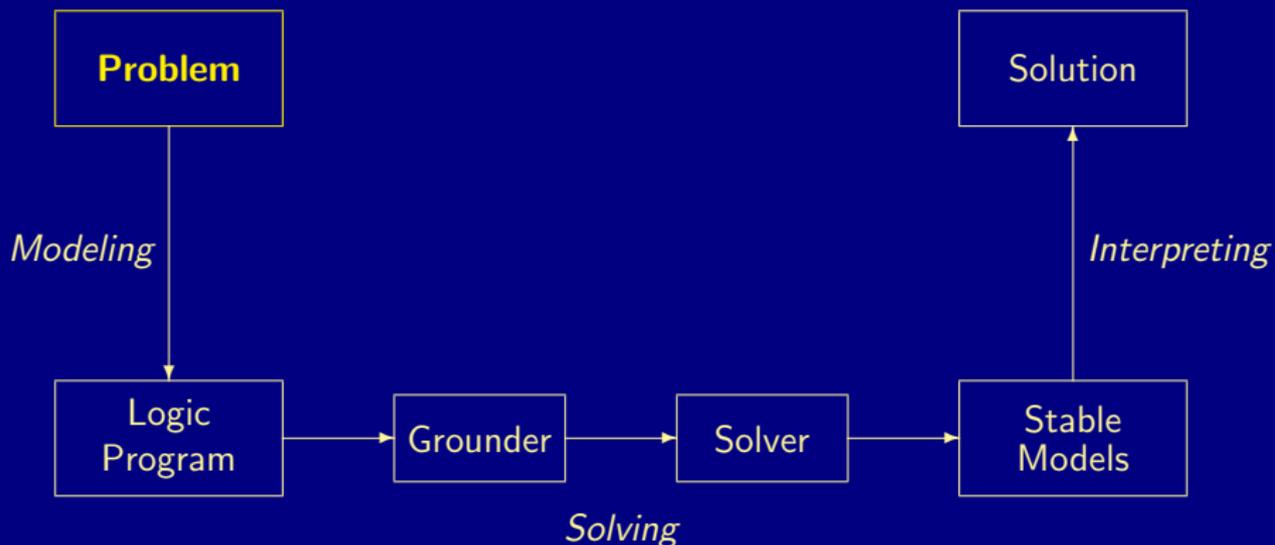
ASP solving process



ASP solving process



A case-study: Graph coloring



Graph coloring

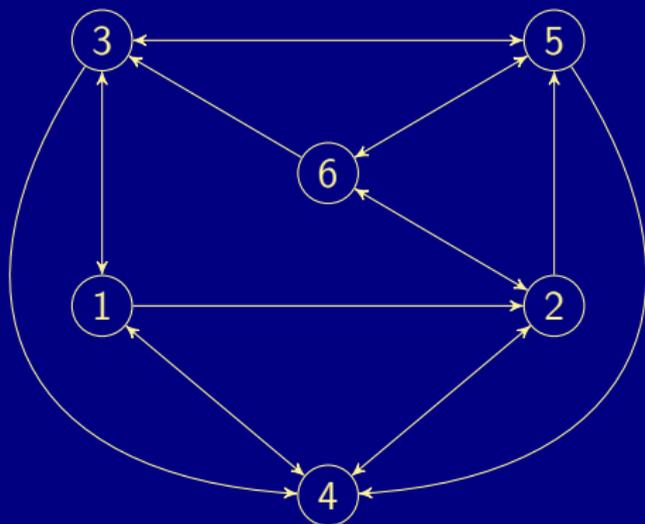
- Problem instance A graph consisting of nodes and edges

Graph coloring

- Problem instance A graph consisting of nodes and edges

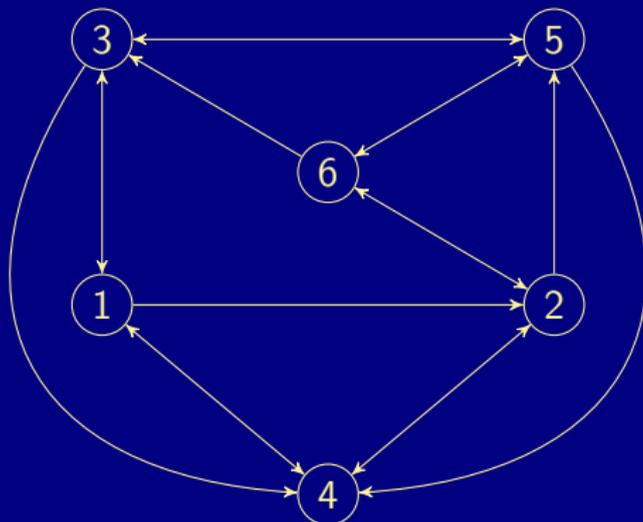
Graph coloring

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Graph coloring

- Problem instance A graph consisting of nodes and edges
 - facts formed by predicates `node/1` and `edge/2`



Graph coloring

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Graph coloring

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

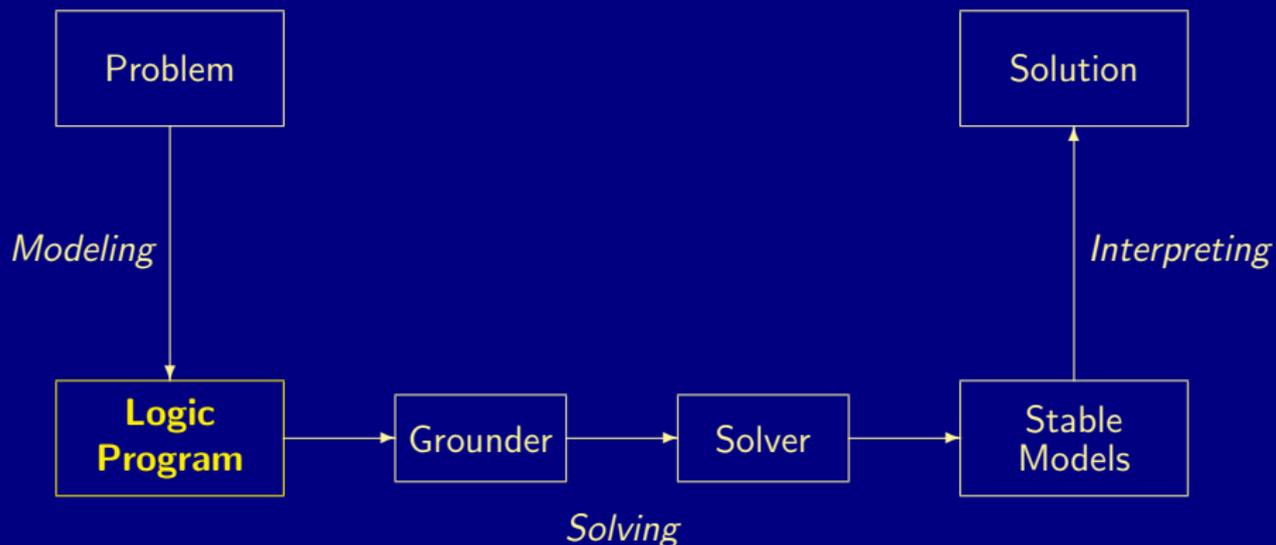
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- Problem instance A graph consisting of nodes and edges
 - facts formed by predicates `node/1` and `edge/2`
 - facts formed by predicate `color/1`
- Problem class Assign each node one color such that no two nodes connected by an edge have the same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

In other words,

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

ASP solving process



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

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encoding**



Graph coloring

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

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

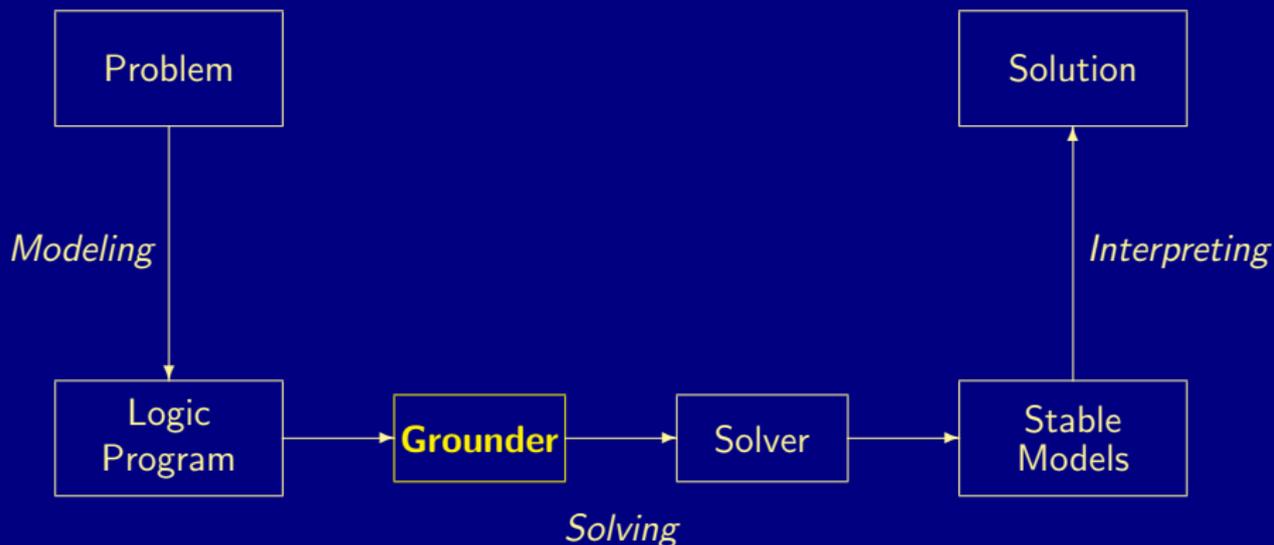
} graph.lp

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

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

} color.lp

ASP solving process



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).
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:- 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).
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:- assign(1,r), assign(4,r). :- assign(2,r), assign(6,r). :- assign(6,r), assign(5,r).
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```

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).
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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).
```

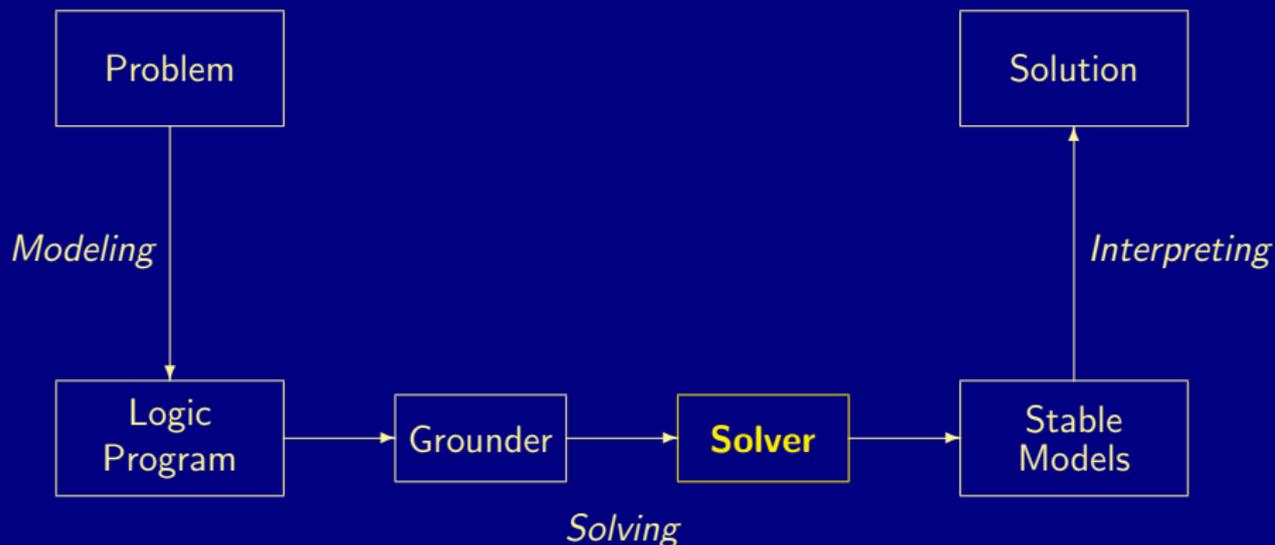
```
edge(1,2). edge(2,4). edge(3,1). edge(4,1). edge(5,3). edge(6,2).
edge(1,3). edge(2,5). edge(3,4). edge(4,2). edge(5,4). edge(6,3).
edge(1,4). edge(2,6). edge(3,5). edge(5,6). edge(6,5).
```

```
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 solving process



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
```

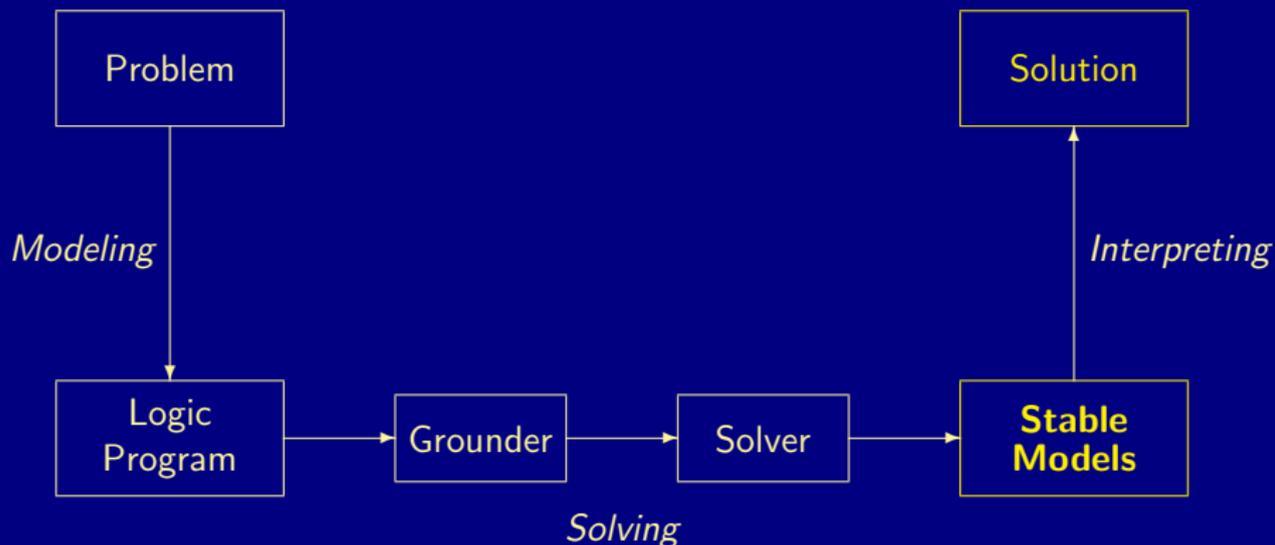
```
node(1) [...] assign(6,b) assign(5,r) assign(4,b) assign(3,g) assign(2,g) assign(1,r)
```

```
SATISFIABLE
```

```
Models      : 6
```

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

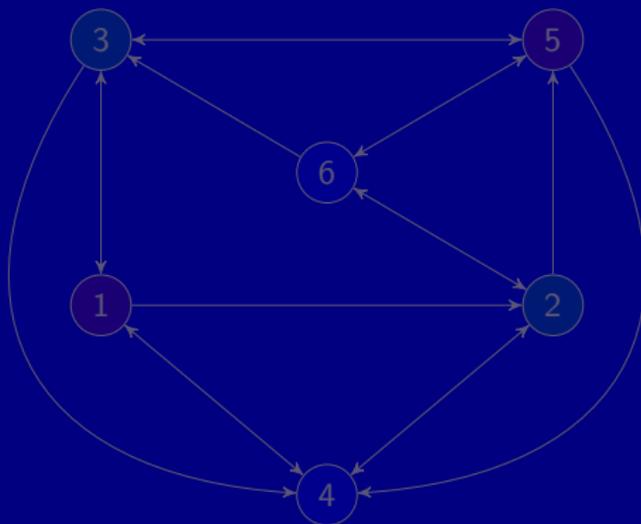
ASP solving process



A coloring

Answer: 6

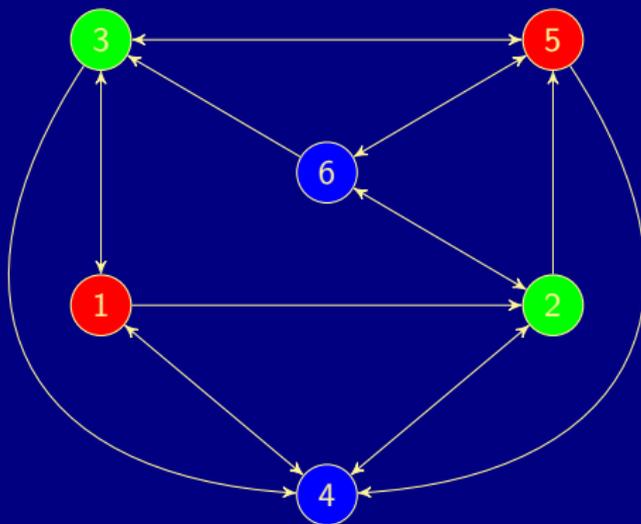
```
node(1) [...] \
assign(6,b) assign(5,r) assign(4,b) assign(3,g) assign(2,g) assign(1,r)
```



A coloring

Answer: 6

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



Outline

11 Elaboration tolerance

12 ASP solving process

13 Methodology

14 Case studies

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).
```

```
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).
```


Generator

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


Tester

Outline

11 Elaboration tolerance

12 ASP solving process

13 Methodology

14 Case studies

Outline

11 Elaboration tolerance

12 ASP solving process

13 Methodology

14 Case studies

- Satisfiability

- Queens

- Traveling salesperson

- Reviewer Assignment

- Planning

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

$\{a\} \leftarrow$

$\{b\} \leftarrow$

Tester

$\leftarrow \neg a, b$

$\leftarrow a, \neg b$

Stable models

$X_1 = \{a, b\}$

$X_2 = \{\}$

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
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Satisfiability testing

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- Example: Consider formula

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- Logic Program

Generator

$$\{a\} \leftarrow$$

$$\{b\} \leftarrow$$

Tester

$$\leftarrow \neg a, b$$

$$\leftarrow a, \neg b$$

Stable models

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

$$X_2 = \{\}$$

Outline

11 Elaboration tolerance

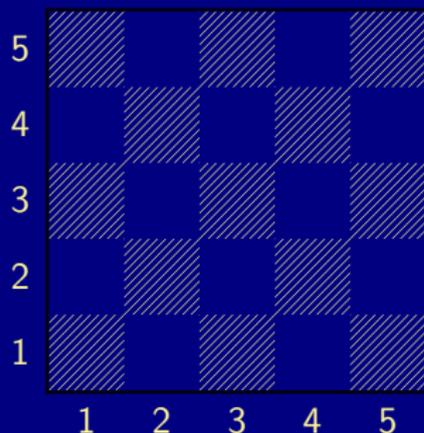
12 ASP solving process

13 Methodology

14 Case studies

- Satisfiability
- **Queens**
- Traveling salesperson
- Reviewer Assignment
- Planning

The n-queens problem



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



Defining the field

```
queens.lp
```

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

- Create file `queens.lp`
- 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

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
```

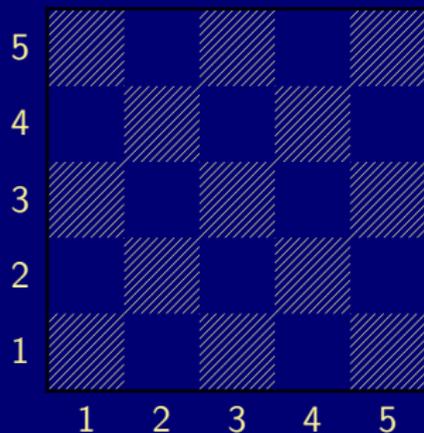
```
row(1) row(2) row(3) row(4) row(5) \  
col(1) col(2) col(3) col(4) col(5) queen(2,1)
```

```
SATISFIABLE
```

```
Models      : 3+
```

Placing some queens

Answer: 1

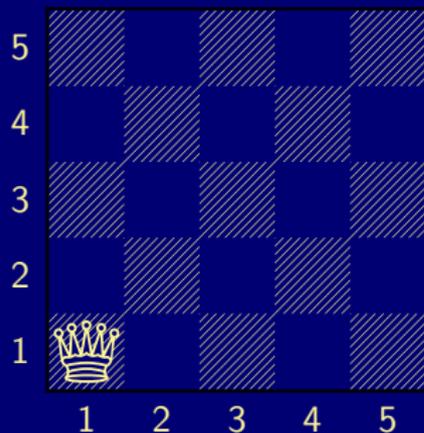


Answer: 1

```
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

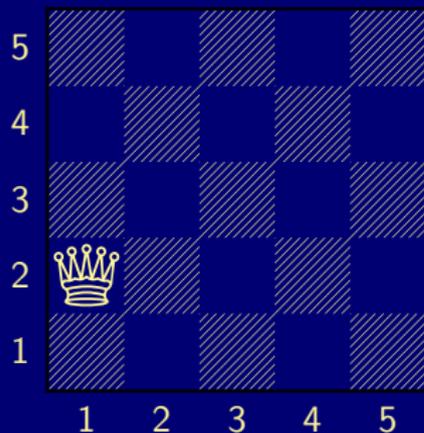


Answer: 2

```
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) }.  
:- not { queen(I,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
```

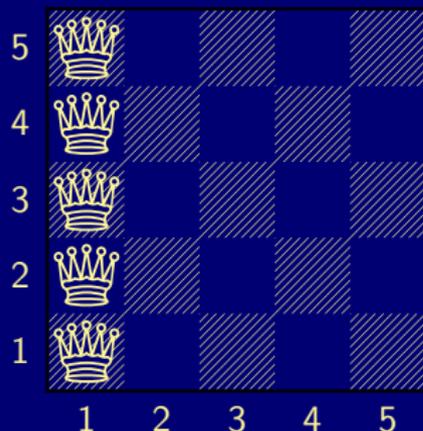
```
row(1) row(2) row(3) row(4) row(5) \
```

```
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

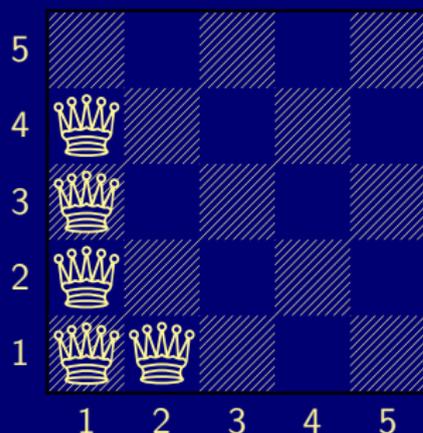


Answer: 1

```
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 attacks
- Forbid 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 attacks
- Forbid vertical attacks

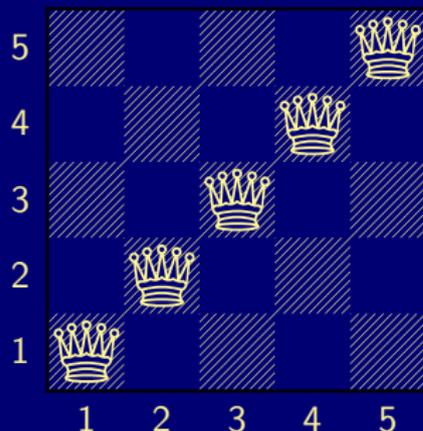
Horizontal and vertical attack

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

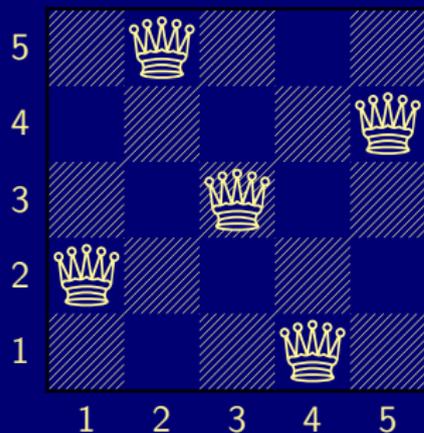
Running ...

```
$ 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-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
```

```
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```

Outline

- 11 Elaboration tolerance
- 12 ASP solving process
- 13 Methodology
- 14 Case studies
 - Satisfiability
 - Queens
 - **Traveling salesperson**
 - Reviewer Assignment
 - Planning

Traveling salesperson

```
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

```
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

```
{ 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) }.

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

```
{ 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]

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

```
{ 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).
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:- city(X), 2 { travel(Y,X) }.

:~ travel(X,Y), road(X,Y,D). [D,X,Y]
```

tsp.lp

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).
```

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
```

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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
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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) }.
```

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

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

```
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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) }.
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Reviewer Assignment

by Ilkka Niemelä

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

by Ilkka Niemelä

```
reviewer(r1). paper(p1). classA(r1,p1). classB(r1,p2). coi(r1,p3).
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Simplified STRIPS¹ 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 $\{p\}$
 - goal state $\{r\}$
 - actions $a = (\{p\}, \{q, \neg p\})$ and $b = (\{q\}, \{r, \neg q\})$
 - length 2
 - plan $\langle a, b \rangle$

¹Stanford Research Institute Problem Solver, 1971

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Simplistic STRIPS Planning

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

```
fluent(p).      action(a).      action(b).      init(p).  
fluent(q).      pre(a,p).        pre(b,q).  
fluent(r).      add(a,q).         add(b,r).       query(r).  
                del(a,p).         del(b,q).
```

```
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

```
time(1..k).
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```
fluent(p).      action(a).      action(b).      init(p).  
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```

```
holds(P,0) :- init(P).
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Simplistic STRIPS Planning

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                del(a,p).      del(b,q).
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Simplistic STRIPS Planning

```
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fluent(p).      action(a).      action(b).      init(p).
fluent(q).      pre(a,p).      pre(b,q).
fluent(r).      add(a,q).      add(b,r).      query(r).
                del(a,p).      del(b,q).
```

```
holds(P,0) :- init(P).
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{ occ(A,T) : action(A) } = 1 :- time(T).
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```

```
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).
```

Engineering: Overview

15 Meta programming

16 Controlling

17 Multi-shot solving

18 Theory solving

19 Heuristic-driven solving

Do it yourself!

- Roland Kaminski, Javier Romero, Torsten Schaub, Philipp Wanko:
How to build your own ASP-based system?!
CoRR abs/2008.06692 (2020)

Outline

15 Meta programming

16 Controlling

17 Multi-shot solving

18 Theory solving

19 Heuristic-driven solving

What is this?

- A technique in which programs treat other programs as data
- Workflow
 - 1 Translate a logic program into facts
 - 2 Combine the obtained facts with a meta logic program
- The first step is called reification

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What is this?

- A technique in which programs treat other programs as data
- Workflow
 - 1 Translate a logic program into facts
 - 2 Combine the obtained facts with a meta logic program
- The first step is called **reification**

An example

- Logic program `ezy.lp`

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

- Implicit statements

```
#show a.  
#show b.  
#show c.
```

- Reification

```
$ clingo --output=reify ezy.lp
```

An example

- Logic program `ezy.lp`

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- Implicit statements

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- Implicit statements

```
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#show b.  
#show c.
```

- Reification

```
$ clingo --output=reify ezy.lp
```

An example, reified

```
$ clingo --output=reify ezy.lp
atom_tuple(0).
atom_tuple(0,1).
literal_tuple(0).
rule(choice(0),normal(0)).
atom_tuple(1).
atom_tuple(1,2).
literal_tuple(1).
literal_tuple(1,-1).
rule(disjunction(1),normal(1)).
atom_tuple(2).
atom_tuple(2,3).
literal_tuple(2).
literal_tuple(2,1).
rule(disjunction(2),normal(2)).
output(a,2).
literal_tuple(3).
literal_tuple(3,3).
output(b,3).
literal_tuple(4).
literal_tuple(4,2).
output(c,4).
```

An example, again

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

An example, reified again

```

rule(choice(0),normal(0)).      atom_tuple(0).      literal_tuple(0).
                                atom_tuple(0,1).
rule(disjunction(1),normal(1)). atom_tuple(1).      literal_tuple(1).
                                atom_tuple(1,2). literal_tuple(1,-1).
rule(disjunction(2),normal(2)). atom_tuple(2).      literal_tuple(2).
                                atom_tuple(2,3). literal_tuple(2,1).

output(a,2).
output(b,3).                    literal_tuple(3).
                                literal_tuple(3,3).

output(c,4).                    literal_tuple(4).
                                literal_tuple(4,2).

```

Meta encoding, or ASP in ASP

```

conjunction(B) :- literal_tuple(B),
    hold(L) : literal_tuple(B, L), L > 0;
    not hold(L) : literal_tuple(B,-L), L > 0.

body(normal(B)) :- rule(_,normal(B)), conjunction(B).
body(sum(B,G))  :- rule(_,sum(B,G)),
    #sum { W,L :      hold(L), weighted_literal_tuple(B, L,W), L > 0 ;
          W,L : not hold(L), weighted_literal_tuple(B,-L,W), L > 0 } >= G.

    hold(A) : atom_tuple(H,A)   :- rule(disjunction(H),B), body(B).
{ hold(A) : atom_tuple(H,A) } :- rule(      choice(H),B), body(B).

#show.
#show T : output(T,B), conjunction(B).

```

An example, running

- Logic program `ezy.lp`

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

- Running

```
$ clingo ezy.lp 0  
clingo version 5.5.0  
Reading from ezy.lp  
Solving...  
Answer: 1  
c  
Answer: 2  
a b  
SATISFIABLE
```

An example, running

- Logic program `ezy.lp`

```
{a}.  
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```

- Running

```
$ clingo ezy.lp 0  
clingo version 5.5.0  
Reading from ezy.lp  
Solving...  
Answer: 1  
c  
Answer: 2  
a b  
SATISFIABLE
```

An example, running reified

- Logic program `ezy.lp`

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

- Running reified

```
$ clingo --output=reify ezy.lp | clingo - meta.lp 0  
clingo version 5.5.0  
Reading from - ...  
Solving...  
Answer: 1  
c  
Answer: 2  
a b  
SATISFIABLE
```

An example, running reified

- Logic program `ezy.lp`

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

- Running reified

```
$ clingo --output=reify ezy.lp | clingo - meta.lp 0  
clingo version 5.5.0  
Reading from - ...  
Solving...  
Answer: 1  
c  
Answer: 2  
a b  
SATISFIABLE
```

Some applications

- Re-interpretation of language constructs
- Interpretation of new language constructs
- Computation of classical models
- Computation of supported models
- Computation of here-and-there models
- Computation of diverse stable models
- Computation of preferred models
- Checking program equivalence
- Guess-and-check programming

Outline

15 Meta programming

16 Controlling

17 Multi-shot solving

18 Theory solving

19 Heuristic-driven solving

Taming the ASP system, imperatively

- Three alternative ways of combining ASP with other languages, either via
 - embedded script
 - module import
 - application class

- We use Python, although other choices exist

An example

- Input program `example.lp`

```
num(3).  
num(6).  
div(N,@divisors(N)) :- num(N).
```

- Resulting program

```
num(3).  
num(6).  
div(3,(1;3)).  
div(6,(1;2;3;6)).
```

An example

- Input program `example.lp`

```
num(3).  
num(6).  
div(N,@divisors(N)) :- num(N).
```

- Resulting program

```
num(3).  
num(6).  
div(3,(1;3)).  
div(6,(1;2;3;6)).
```

Embedded script (embedded.lp)

```
#script (python)
import clingo
def divisors(a):
    a = a.number
    for i in range(1, a+1):
        if a % i == 0:
            yield clingo.Number(i)
#end.
```

Embedded script, running

```
$ clingo example.lp embedded.lp
clingo version 5.5.0
Reading from example.lp ...
Solving...
Answer: 1
num(3) num(6) div(3,1) div(3,3) \
div(6,1) div(6,2) div(6,3) div(6,6)
SATISFIABLE
```

Module import (module.py)

```
import clingo

class ExampleApp:
    @staticmethod
    def divisors(a):
        a = a.number
        for i in range(1, a+1):
            if a % i == 0:
                yield clingo.Number(i)

    def run(self):
        ctl = clingo.Control()
        ctl.load("example.lp")
        ctl.ground([("base", [])], context=self)
        ctl.solve(on_model=print)

if __name__ == "__main__":
    ExampleApp().run()
```

Embedded script, running

```
$ python module.py  
num(3) num(6) div(3,1) div(3,3) \  
div(6,1) div(6,2) div(6,3) div(6,6)
```

Application class (app.py)

```
import sys
import clingo

class ExampleApp(clingo.Application):
    program_name = "example"
    version = "1.0"

    @staticmethod
    def divisors(a):
        a = a.number
        for i in range(1, a+1):
            if a % i == 0:
                yield clingo.Number(i)

    def main(self, ctl, files):
        for path in files: ctl.load(path)
        if not files:
            ctl.load("-")
        ctl.ground([("base", [])], context=self)
        ctl.solve()

if __name__ == "__main__":
    clingo.clingo_main(ExampleApp(), sys.argv[1:])
```

Application class, running

```
$ python app.py example.lp
example version 1.0
Reading from example.lp
Solving...
Answer: 1
num(3) num(6) div(3,1) div(3,3) \
div(6,1) div(6,2) div(6,3) div(6,6)
SATISFIABLE
```

What to use when...?

- embedded script
 - suitable for small amendments to the logic program, anything on the term level during grounding
 - perform calculations that are hard or inconvenient to express in ASP
- module import
 - convenient way to use *clingo* as part of a larger project
 - provides high level functions to control grounding and solving
 - surrounding application is in charge of the control flow and ASP is used to perform specific computations
- application class
 - aims at building custom systems based on *clingo*
 - similar to module import but with more customization capabilities
 - 👉 constitutes the cornerstone of recent *clingo*-based systems such as *clingcon*, *clingo*[DL], *eclingo*, and *telingo*

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Motivation

- Multi-shot solving allows for solving continuously changing logic programs in an operative way
 - Single-shot solving: *ground* | *solve*
 - Multi-shot solving: *ground* | *solve*

Agents, Assisted Living, Robotics, Planning, Query-answering, etc

clingo = ASP + Control

Extend ASP with dedicated directives

Provide powerful API (here: Python)

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Structuring logic programs

- Program directive

```
#program <name> [ (<parameters>) ]
```

where

- <name> is a term
 - (<parameters>) is a tuple of terms
-
- Example `#program play(p,t).`
 - Default `#program base.`

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 - Default `#program base.`

An example (chemistry.lp)

```
a(1).  
#program acid(k).  
b(k).  
c(X,k) :- a(X).  
#program base.  
a(2).
```

The example, processing (control-base.py)

```
import clingo
ctl = clingo.Control()
ctl.load("chemistry.lp")
ctl.ground([("base", [])])
ctl.solve(on_model=print)
```

```
a(1).
#program acid(k).
b(k).
c(X,k) :- a(X).
#program base.
a(2).
```

```
$ python control-base.py
a(1) a(2)
```

The example, processing (control-base.py)

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ctl = clingo.Control()
ctl.load("chemistry.lp")
ctl.ground([( "base", [])])
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```

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```
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#program acid(k).
b(k).
c(X,k) :- a(X).
#program base.
a(2).
```

```
$ python control-base.py
a(1) a(2)
```

The example, processing (control-acid.py)

```
import clingo
ctl = clingo.Control()
ctl.load("chemistry.lp")
ctl.ground([( "acid", [42])])
ctl.solve(on_model=print)
```

```
a(1).
#program acid(k).
b(k).
c(X,k) :- a(X).
#program base.
a(2).
```

```
$ python control-acid.py
b(42)
```

The example, processing (control-acid.py)

```
import clingo
ctl = clingo.Control()
ctl.load("chemistry.lp")
ctl.ground([( "acid", [42] )])
ctl.solve(on_model=print)
```

```
a(1).
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b(k).
c(X,k) :- a(X).
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$ python control-acid.py
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```

The example, processing (control-acid.py)

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import clingo
ctl = clingo.Control()
ctl.load("chemistry.lp")
ctl.ground([( "acid", [42] )])
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```

```
a(1).
#program acid(k).
b(k).
c(X,k) :- a(X).
#program base.
a(2).
```

```
$ python control-acid.py
b(42)
```

External atoms

- External directive

```
#external <atom> [ : <body> ]
```

where

- `<atom> [: <body>]` is a (conditional) literal
- Example `#external mark(X,Y,p,t) : field(X,Y).`
- Note External atoms are
 - protected from program simplifications
 - assigned truth values via API (default: false)
 and can be
 - overwritten by adding rules defining the atom
 - permanently set to false

External atoms

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- overwritten by adding rules defining the atom
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An example (chemistry-external.lp)

```

a(1).
#program acid(k).
b(k).
c(X,k) :- a(X).
#program base.
a(2).

#program acid(k).
#external d(X,k) : c(X,k).
e(X,k) :- d(X,k).

```

Note: Grounding both base and acid(42) yields two externals

An example (chemistry-external.lp)

```

a(1).
#program acid(k).
b(k).
c(X,k) :- a(X).
#program base.
a(2).

#program acid(k).
#external d(X,k) : c(X,k).
e(X,k) :- d(X,k).

```

Note: Grounding both base and acid(42) yields two externals

An example (chemistry-external.lp)

```

a(1).
#program acid(k).
b(k).
c(X,k) :- a(X).
#program base.
a(2).

#program acid(k).
#external d(X,k) : c(X,k).
e(X,k) :- d(X,k).

```

Note Grounding both base and acid(42) yields two externals

The example, processing (control-external.py)

```
ctl = clingo.Control()
ctl.load("chemistry-external.lp")
ctl.ground([("base", []), ("acid", [42])])
ctl.solve(on_model=print)
ctl.assign_external(Function("d", [2,42]), True)
ctl.solve(on_model=print)
```

```
$ python control-external.py
a(1) a(2) c(1,42) c(2,42) b(42)
a(1) a(2) c(1,42) c(2,42) b(42) d(2,42) e(2,42)
```

The example, processing (control-external.py)

```
ctl = clingo.Control()
ctl.load("chemistry-external.lp")
ctl.ground([("base", []), ("acid", [42])])
ctl.solve(on_model=print)
ctl.assign_external(Function("d", [2,42]), True)
ctl.solve(on_model=print)
```

```
$ python control-external.py
a(1) a(2) c(1,42) c(2,42) b(42)
a(1) a(2) c(1,42) c(2,42) b(42) d(2,42) e(2,42)
```

An example of incremental solving

```
#program base.  
p(0).  
#program step (t).  
p(t) :- p(t-1).  
#program check (t).  
#external query(t).           % added in python below  
:- not p(42), query(t).
```

An example of incremental solving

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:- not p(42), query(t).
```


Incremental solving, zoom on register_options

```
def register_options(self, options: ApplicationOptions):
    """
    Register program options.
    """
    group = "Inc-Example Options"
    options.add(
        group, "imin",
        "Minimum number of steps [{}]" .format(self._conf.imin),
        parse_int(self._conf, "imin", min_value=0),
        argument="<n>")
    options.add(
        group, "imax",
        "Maximum number of steps [{}]" .format(self._conf.imax),
        parse_int(self._conf, "imax", min_value=0, optional=True),
        argument="<n>")
    options.add(
        group, "istop",
        "Stop criterion [{}]" .format(self._conf.istop),
        parse_stop(self._conf, "istop"))
```

Incremental solving, zoom on main

```

def main(self, ctl: Control, files: Iterable[str]):
    '''
    The main function implementing incremental solving.
    '''
    if not files:
        files = ["-"]
    for file_ in files:
        ctl.load(file_)
    ctl.add("check", ["t"], "#external query(t).")
    conf = self._conf
    step = 0
    ret: Optional[SolveResult] = None
    while ((conf.imax is None or step < conf.imax) and
           (ret is None or step < conf.imin or (
               (conf.istop == "SAT" and not ret.satisfiable) or
               (conf.istop == "UNSAT" and not ret.unsatisfiable) or
               (conf.istop == "UNKNOWN" and not ret.unknown)))):
        parts = []
        parts.append(("check", [Number(step)]))
        if step > 0:
            ctl.release_external(Function("query", [Number(step - 1)]))
            parts.append(("step", [Number(step)]))
        else:
            parts.append(("base", []))
        ctl.ground(parts)
        ctl.assign_external(Function("query", [Number(step)]), True)
        ret, step = cast(SolveResult, ctl.solve()), step + 1

```

Optimization

- Imagine some Blocksworld planning problem ...
- Code snippet

```
ngoal(T) :- on(B,L,T), not goal_on(B,L).  
:- ngoal(n).
```

where `n` is a fixed horizon

- Optimization

```
_minimize(1,T) :- ngoal(T).
```

Optimization

- Imagine some Blocksworld planning problem ...
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          :- ngoal(n).
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where `n` is a fixed horizon

- Optimization

```
_minimize(1,T) :- ngoal(T).
```

Optimization

```

'''
Exmample to show branch and bound based optimization using multi-shot solving.
'''

import sys
from typing import Optional, Iterable, cast
from clingo import Model, Control, SolveResult, SymbolType, Application, Number, clingo_main

class OptApp(Application):
    '''
    Example application.
    '''
    program_name: str = "opt-example"
    version: str = "1.0"
    _bound: Optional[int]

    def __init__(self):
        self._bound = None

    def _on_model(self, model: Model):
        self._bound = 0
        for atom in model.symbols(atoms=True):
            if (atom.match("_minimize", 2) and
                atom.arguments[0].type is SymbolType.Number):
                self._bound += atom.arguments[0].number

    def main(self, ctl: Control, files: Iterable[str]):
        '''
        Main function implementing branch and bound optimization.
        '''
        if not files:
            files = ["-"]
        for file_ in files:
            ctl.load(file_)
        ctl.add("bound", ["b"],
              ":- #sum { V,I: _minimize(V,I) } >= b.")
        ctl.ground([("base", [])])
        while cast(SolveResult, ctl.solve(on_model=self._on_model)).satisfiable:
            print("Found new bound: {}".format(self._bound))
            ctl.ground([("bound", [Number(cast(int, self._bound))])])
        if self._bound is not None:
            print("Optimum found")

clingo_main(OptApp(), sys.argv[1:])

```

Optimization, zoom on main

```

def main(self, ctl: Control, files: Iterable[str]):
    '''
    Main function implementing branch and bound optimization.
    '''
    if not files:
        files = ["-"]
    for file_ in files:
        ctl.load(file_)
    ctl.add("bound", ["b"],
           ":- #sum { V,I: _minimize(V,I) } >= b.")
    ctl.ground([( "base", [])])
    while cast(SolveResult, ctl.solve(on_model=self._on_model)).satisfiable:
        print("Found new bound: {}".format(self._bound))
        ctl.ground([( "bound", [Number(cast(int, self._bound))])])
    if self._bound is not None:
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```

Optimization, zoom on `_on_model`

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def _on_model(self, model: Model):
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```

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Motivation

- Input: $ASP = DB + KRR + LP + SAT$
- Output: $ASPmT = DB + KRR + LP + S$
- ASP solving: *ground* | *solve*
- Challenge: Logic programs with elusive theory atoms
- Application areas:
Agents, Assisted Living, Robotics, Planning, Scheduling,
Bio- and Cheminformatics, etc

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- Output $ASPmT = DB + KRR + LP + SMT$
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Motivation

- Input $ASP = DB + KRR + LP + SAT$
- Output $ASPmT = DB + KRR + LP + SMT$ — **NO!**
- ASP solving: *ground* | *solve*
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Motivation

- Input $ASP = DB + KRR + LP + SAT$
- Output $ASPmT = (DB + KRR + LP + SAT)mT$
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- Challenge: Logic programs with elusive theory atoms
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Motivation

- Input $ASP = DB + KRR + LP + SAT$
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- **ASP solving:** *ground* | *solve*
- Challenge Logic programs with elusive theory atoms
- Application areas
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Motivation

- Input $ASP = DB + KRR + LP + SAT$
- Output $ASPmT = (DB + KRR + LP + SAT)mT$
- **ASP solving modulo theories:** *ground % theories | solve % theories*
- Challenge Logic programs with elusive theory atoms
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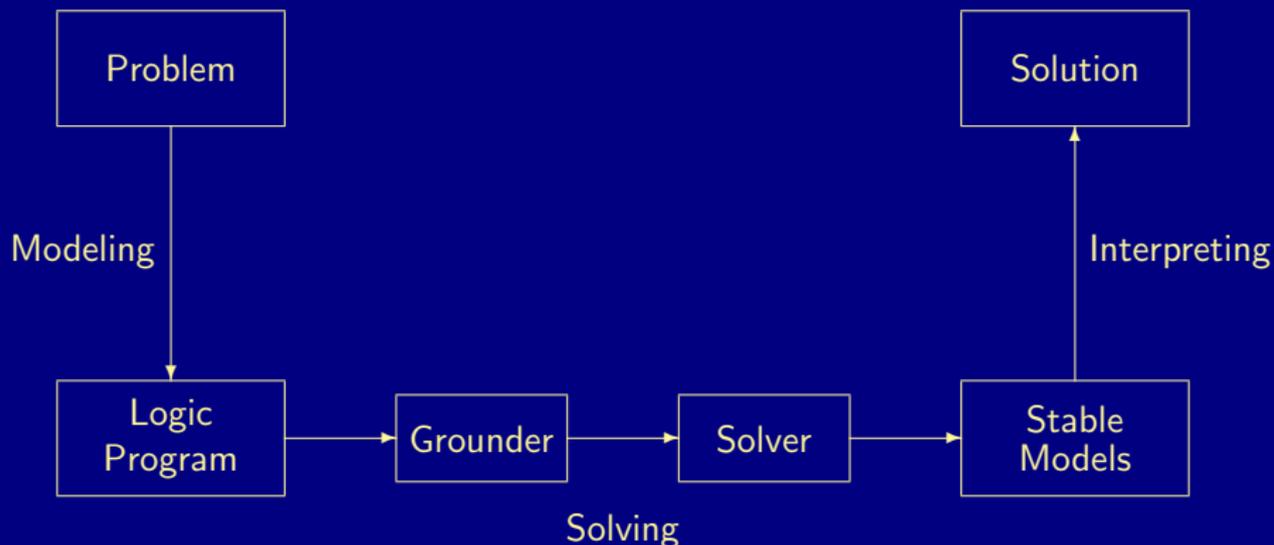
Motivation

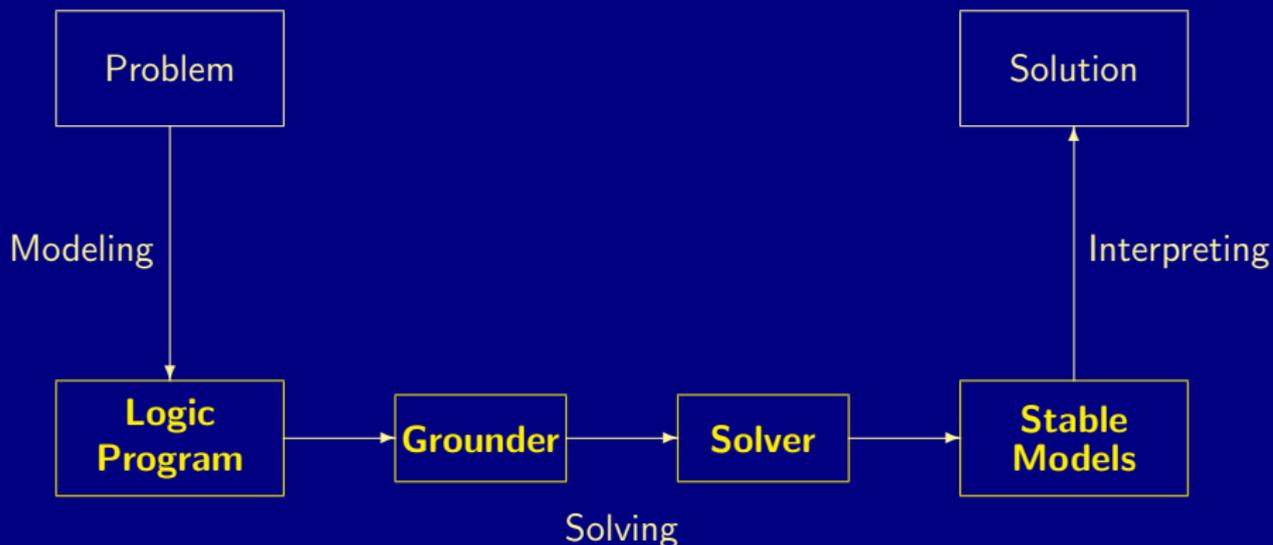
- Input $ASP = DB + KRR + LP + SAT$
- Output $ASPmT = (DB + KRR + LP + SAT)mT$
- ASP solving modulo theories: *ground % theories | solve % theories*
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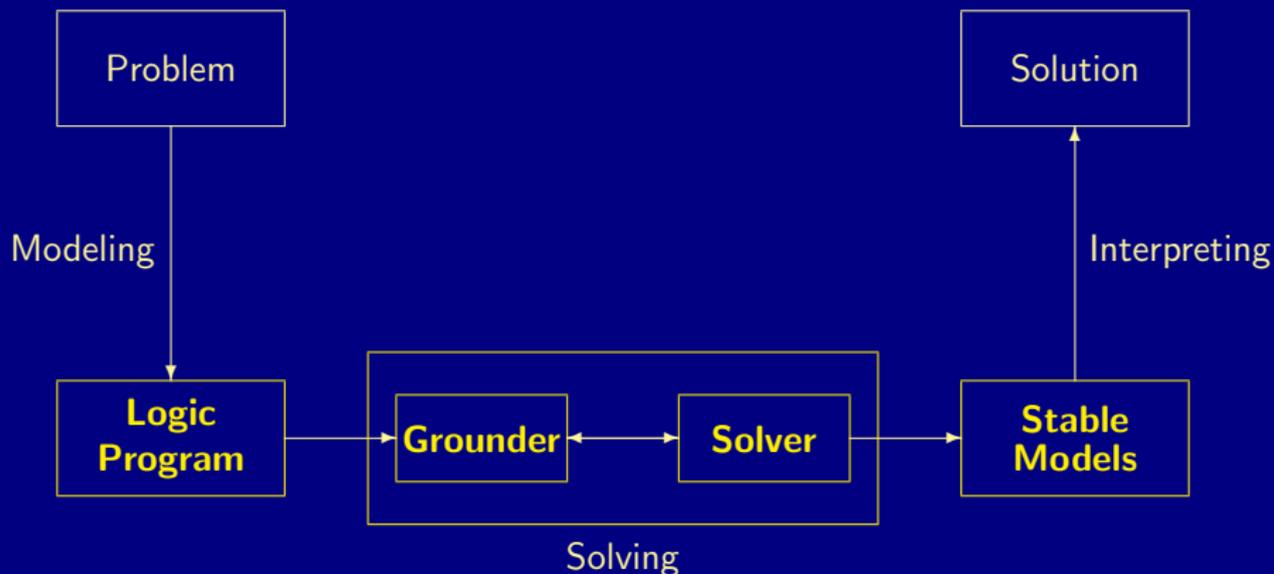
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- ASP solving modulo theories: *ground % theories | solve % theories*
- Challenge Logic programs with elusive theory atoms
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Agents, Assisted Living, Robotics, Planning, Scheduling,
Bio- and Cheminformatics, etc

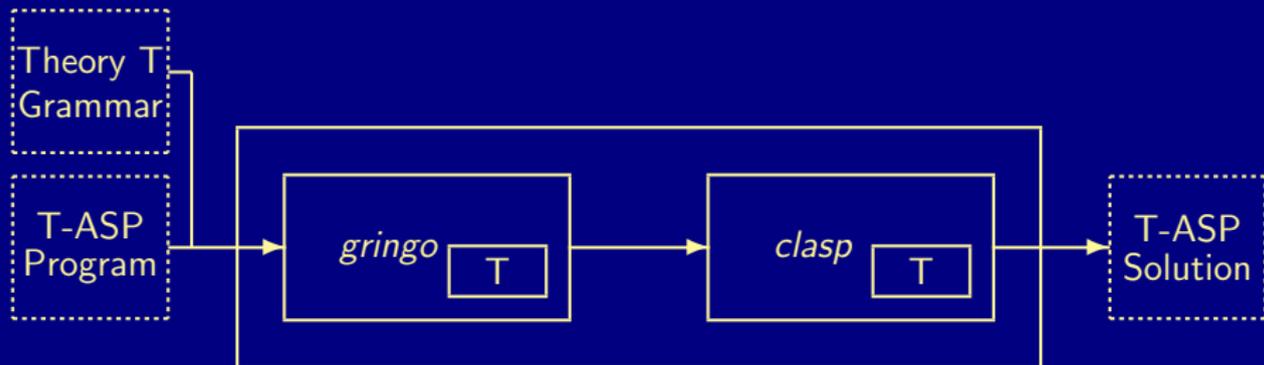
ASP solving process



ASP solving process **modulo theories**

ASP solving process modulo theories



clingo's approach

Outline

15 Meta programming

16 Controlling

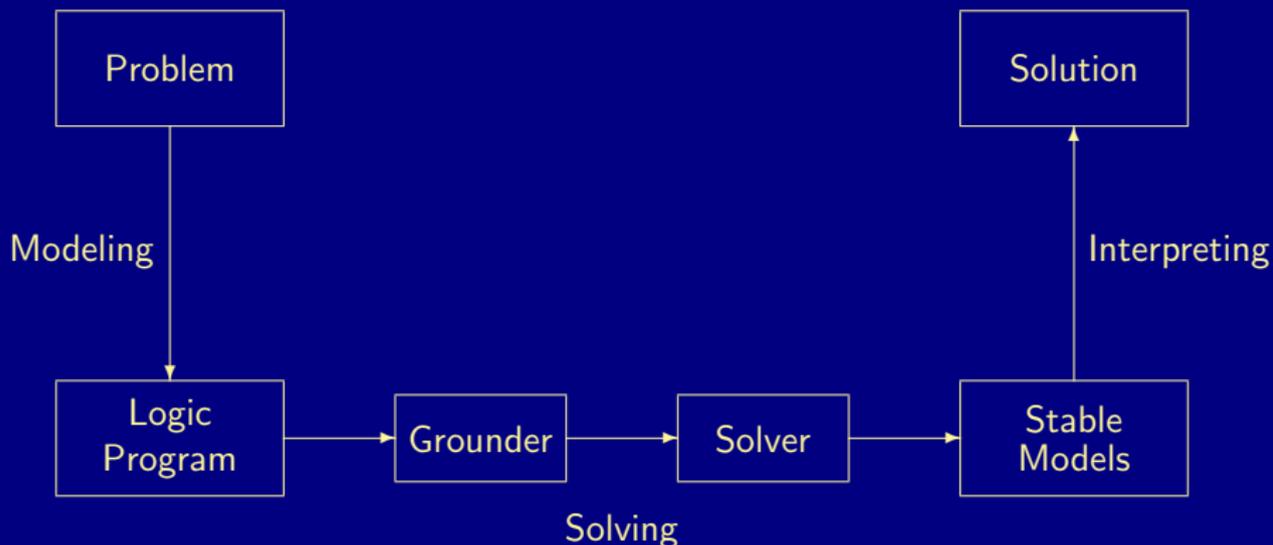
17 Multi-shot solving

18 Theory solving

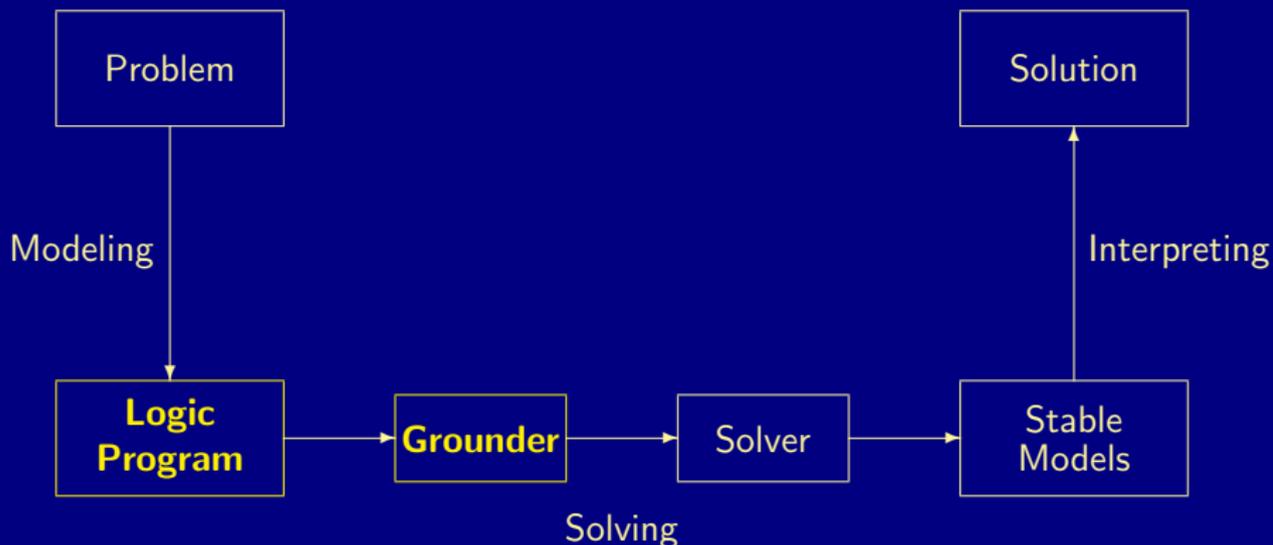
- Theory language
- Theory propagation

19 Heuristic-driven solving

ASP solving process **modulo theories**



ASP solving process **modulo theories**



Linear constraints

```

#theory csp {
  linear_term {
    + : 5, unary;
    - : 5, unary;
    * : 4, binary, left;
    + : 3, binary, left;
    - : 3, binary, left
  };

  dom_term {
    + : 5, unary;
    - : 5, unary;
    .. : 1, binary, left
  };

  show_term {
    / : 1, binary, left
  };

  minimize_term {
    + : 5, unary;
    - : 5, unary;
    * : 4, binary, left;
    + : 3, binary, left;
    - : 3, binary, left;
    @ : 0, binary, left
  };

  &dom/0 : dom_term, {=}, linear_term, any;
  &sum/0 : linear_term, {<=,=,>=,<,>,!}, linear_term, any;
  &show/0 : show_term, directive;
  &distinct/0 : linear_term, any;
  &minimize/0 : minimize_term, directive
}.

```

send+more=money

$$\begin{array}{rcccc}
 & s & e & n & d \\
 + & m & o & r & e \\
 \hline
 m & o & n & e & y
 \end{array}$$

Each letter corresponds exactly to one digit and all variables have to be pairwise distinct

$$\begin{array}{rcccc}
 & 9 & 5 & 6 & 7 \\
 + & 1 & 0 & 8 & 5 \\
 \hline
 1 & 0 & 6 & 5 & 2
 \end{array}$$

The example has exactly one solution

$$\{ s \mapsto 9, e \mapsto 5, n \mapsto 6, d \mapsto 7, m \mapsto 1, o \mapsto 0, r \mapsto 8, y \mapsto 2 \}$$

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$$\{ s \mapsto 9, e \mapsto 5, n \mapsto 6, d \mapsto 7, m \mapsto 1, o \mapsto 0, r \mapsto 8, y \mapsto 2 \}$$

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```

#include "csp.lp".

digit(1,3,s).    digit(2,3,m).    digit(sum,4,m).
digit(1,2,e).    digit(2,2,o).    digit(sum,3,o).
digit(1,1,n).    digit(2,1,r).    digit(sum,2,n).
digit(1,0,d).    digit(2,0,e).    digit(sum,1,e).
                                   digit(sum,0,y).

base(10).
exp(E) :- digit(_,E,_).

power(1,0).
power(B*P,E) :- base(B), power(P,E-1), exp(E), E>0.

number(N) :- digit(N,_,_), N!= sum.
high(D) :- digit(N,E,D), not digit(N,E+1,_).

&dom {0..9} = X :- digit(_,_,X).

&sum { M*D : digit(N,E,D),    power(M,E), number(N);
      -M*D : digit(sum,E,D), power(M,E)          } = 0.

&sum { D } > 0 :- high(D).

&distinct { D : digit(_,_,D) }.

&show { D : digit(_,_,D) }.

```

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```

#include "csp.lp".

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send+more=money

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digit(1,2,e).    digit(2,2,o).    digit(sum,3,o).
digit(1,1,n).    digit(2,1,r).    digit(sum,2,n).
digit(1,0,d).    digit(2,0,e).    digit(sum,1,e).
                                digit(sum,0,y).

base(10).
exp(0).  exp(1).  exp(2).  exp(3).  exp(4).
power(1,0).
power(10,1).  power(100,2).  power(1000,3).  power(10000,4).
number(1).  number(2).
high(s).  high(m).

&dom{0..9}=s. &dom{0..9}=m. &dom{0..9}=e. &dom{0..9}=o. &dom{0..9}=n. &d
&sum{ 1000*s;    100*e;    10*n;    1*d;
      1000*m;    100*o;    10*r;    1*e;
      -10000*m; -1000*o; -100*n; -10*e; -1*y } = 0.

&sum{s} > 0.  &sum{m} > 0.

&distinct{s; m; e; o; n; r; d; y}.

&show{s; m; e; o; n; r; d; y}.

```

Outline

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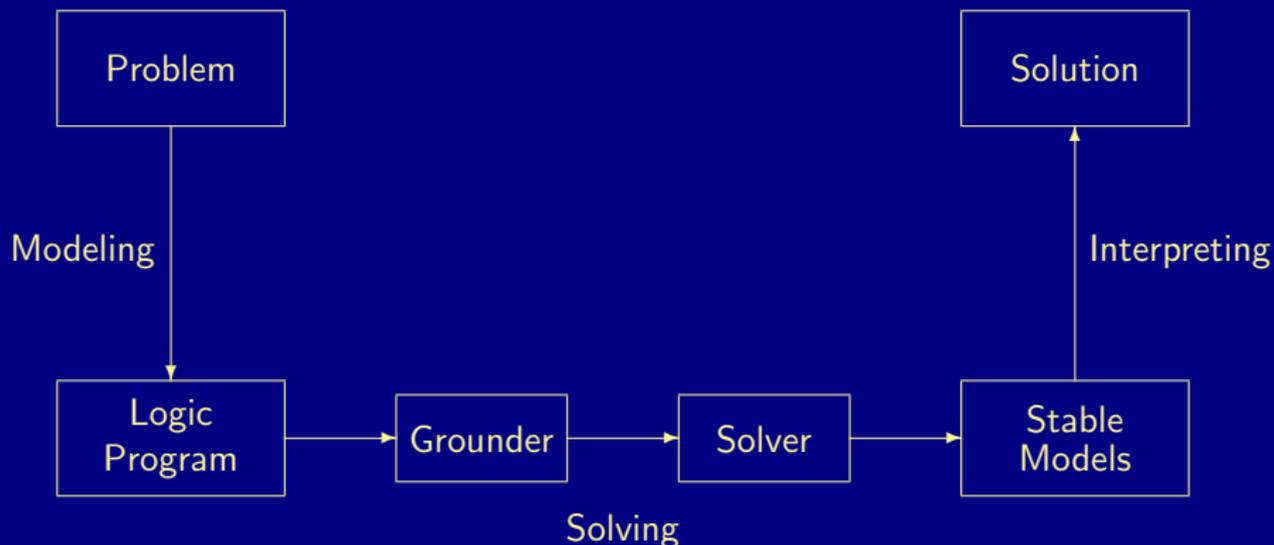
18 Theory solving

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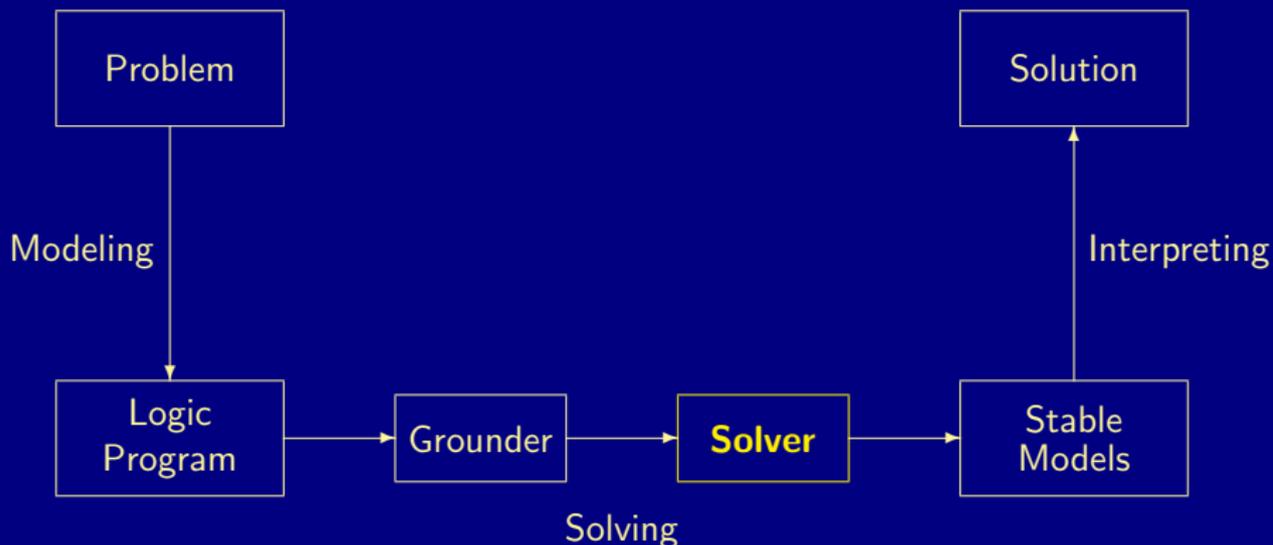
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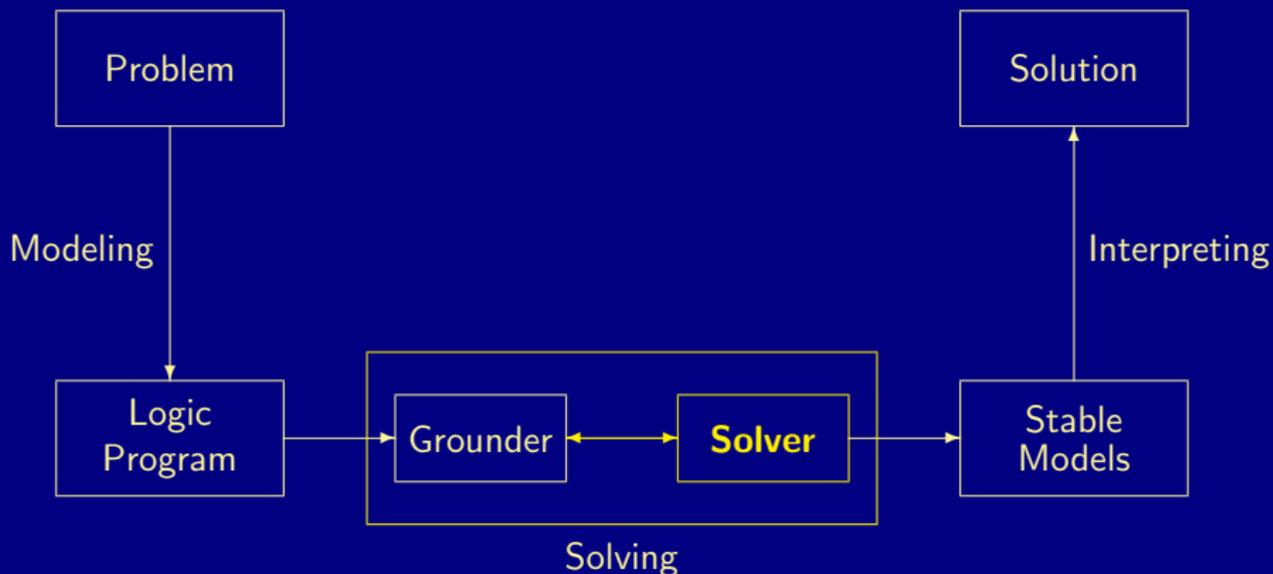
ASP solving process **modulo theories**

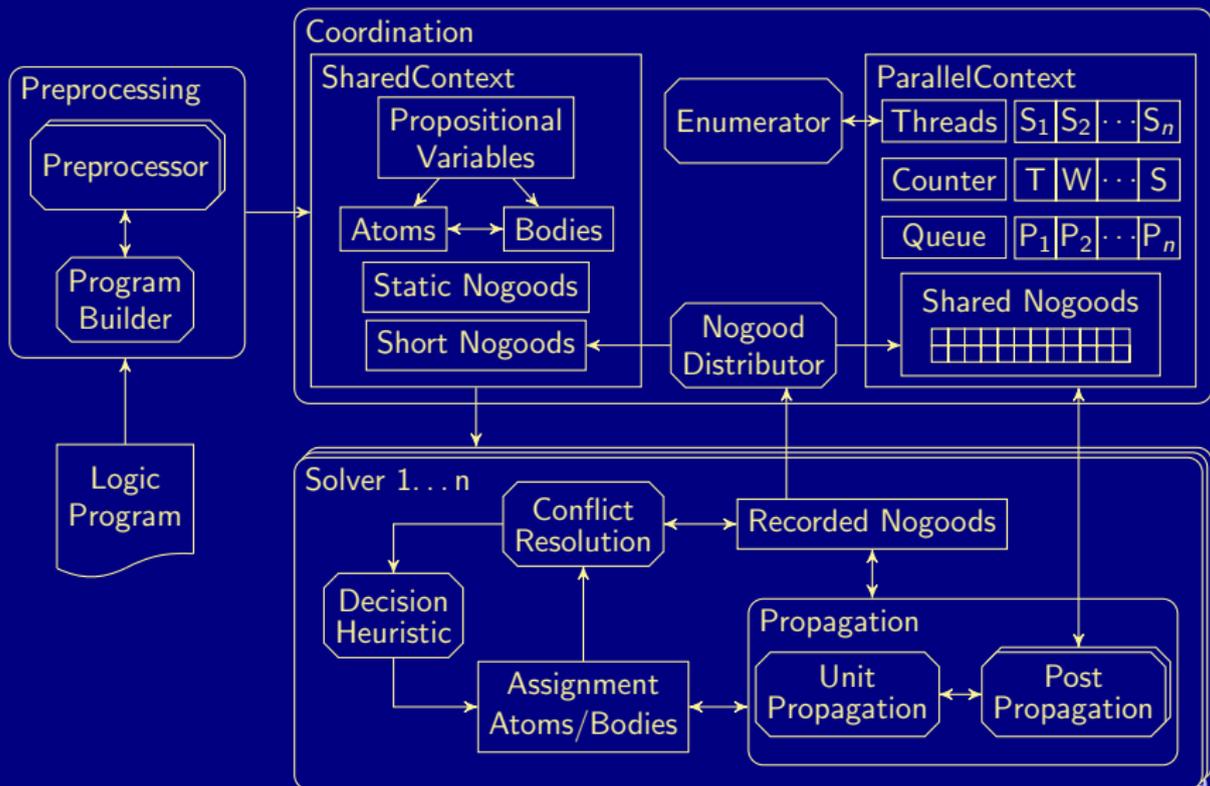


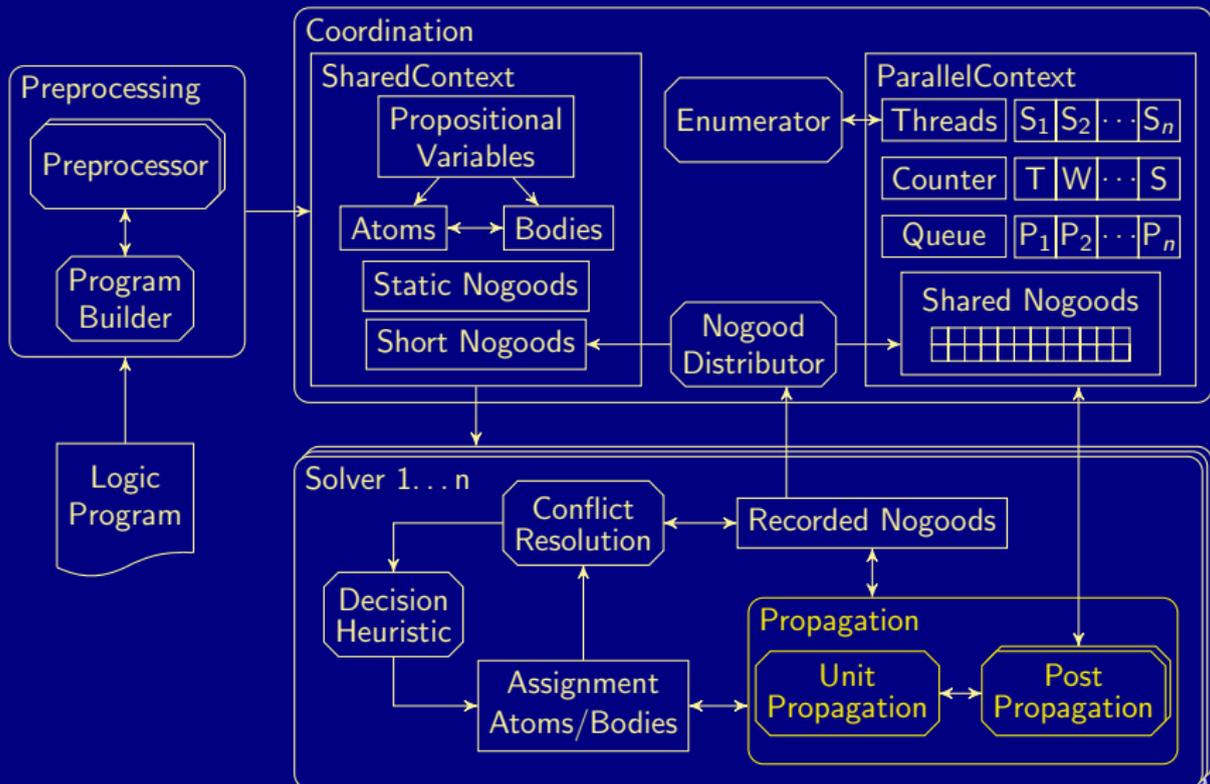
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Architecture of *clasp*

Architecture of *clasp*

Conflict-driven constraint learning modulo theories

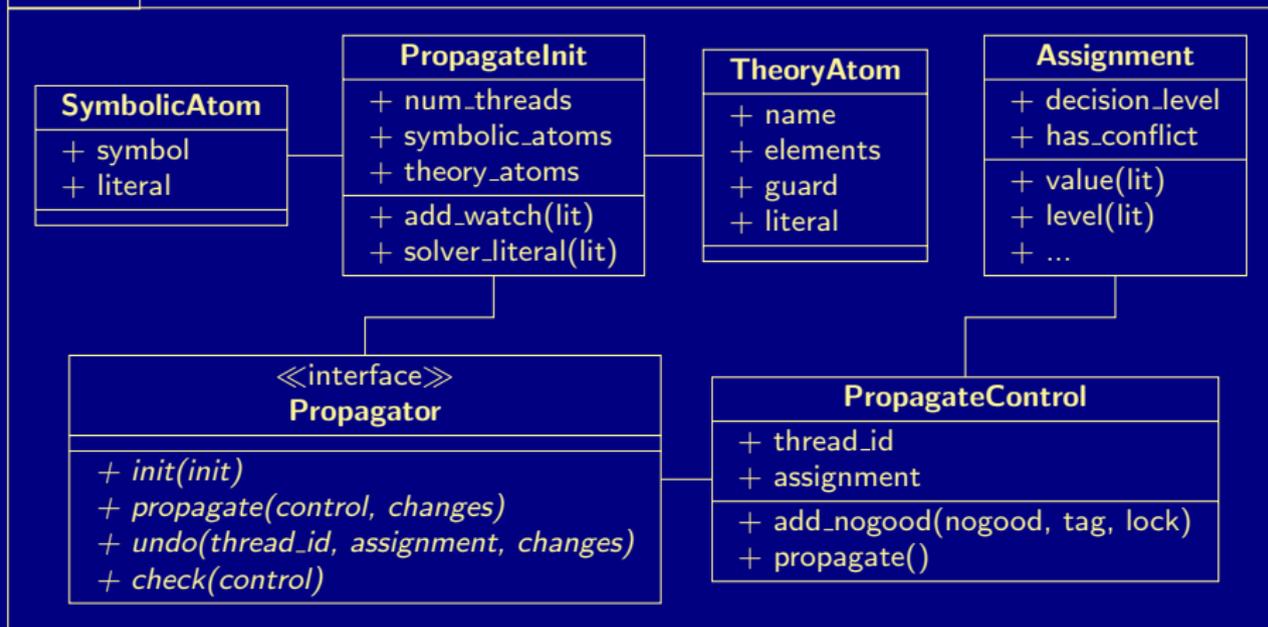
```

(I) initialize // register theory propagators and initialize watches
    loop
      propagate completion, loop, and recorded nogoods // deterministically assign literals
      if no conflict then
        if all variables assigned then
          (C) if some  $\delta \in \Delta_T$  is violated for  $T \in \mathbb{T}$  then record  $\delta$  // theory propagator's check
              else return variable assignment //  $\mathbb{T}$ -stable model found
          else
            (P) propagate theories  $T \in \mathbb{T}$  // theory propagators may record theory nogoods
                if no nogood recorded then decide // non-deterministically assign some literal
            else
              if top-level conflict then return unsatisfiable
              else
                (U) analyze // resolve conflict and record a conflict constraint
                    backjump // undo assignments until conflict constraint is unit

```

Propagator interface

clingo



The *dot* propagator

```
#script (python)
import sys
import time

class Propagator:
    def init(self, init):
        self.sleep = .1
        for atom in init.symbolic_atoms:
            init.add_watch(init.solver_literal(atom.literal))

    def propagate(self, ctl, changes):
        for l in changes:
            sys.stdout.write(".")
            sys.stdout.flush()
            time.sleep(self.sleep)
        return True

    def undo(self, solver_id, assign, undo):
        for l in undo:
            sys.stdout.write("\b \b")
            sys.stdout.flush()
            time.sleep(self.sleep)

def main(prg):
    prg.register_propagator(Propagator())
    prg.ground([("base", [])])
    prg.solve()
    sys.stdout.write("\n")

#end.
```

Outline

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19 Heuristic-driven solving

Motivation

- **Observation** Sometimes it is advantageous to take a more application-oriented approach by including domain-specific information
 - domain-specific knowledge can be added for improving propagation
 - domain-specific heuristics can be used for making better choices
- **Idea** Incorporation of domain-specific heuristics by extending
 - input language and/or solver options for expressing domain-specific heuristics
 - solving capacities for integrating domain-specific heuristics

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Basic CDCL decision algorithm

loop

```

propagate // compute deterministic consequences
if no conflict then
    if all variables assigned then return variable assignment
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else
    if top-level conflict then return unsatisfiable
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        analyze // analyze conflict and add a conflict constraint
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Heuristic language

■ Heuristic directive

```
#heuristic a : l1, ..., ln. [k@p, m]
```

where

- a is an atom, and l_1, \dots, l_n are literals
- k and p are integers
- m is a heuristic modifier

■ Heuristic modifiers

`init` for initializing the heuristic value of a with k

`factor` for amplifying the heuristic value of a by factor k

`level` for ranking all atoms; the rank of a is k

`sign` for attributing the sign of k as truth value to a

■ Example

```
#heuristic occurs(A,T) : action(A), time(T). [T, factor]
```

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Heuristic language

■ Heuristic directive

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

where

- *a* is an atom, and l_1, \dots, l_n are literals
- *k* and *p* are integers
- *m* is a heuristic modifier

■ Heuristic modifiers

`init` for initializing the heuristic value of *a* with *k*

`factor` for amplifying the heuristic value of *a* by factor *k*

`level` for ranking all atoms; the rank of *a* is *k*

`sign` for attributing the sign of *k* as truth value to *a*

`true/false` combine level and sign

■ Example

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■ Example

```
#heuristic occurs(mv,5) : action(mv), time(5). [5, factor]
```

Simple STRIPS planning

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

```
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).
```

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

#heuristic occurs(A,T) : action(A), time(T). [2, factor]

```

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#heuristic occurs(A,T) : action(A), time(T). [1, level]
```

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```

```
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```

```
#heuristic holds(F,T-1) :      holds(F,T). [t-T+1, true]
```

```
#heuristic holds(F,T-1) : not holds(F,T) [t-T+1, false]
                           fluent(F), time(T).
```

Heuristic options

■ Alternative for specifying structure-oriented heuristics in *clasp*

```
--dom-mod=<arg> : Default modification for
                  domain heuristic
```

```
<arg>: <mod>[,<pick>]
```

```
<mod>  : Modifier
```

```
{1=level|2=pos|3=true|4=neg|
 5=false|6=init|7=factor}
```

```
<pick> : Apply <mod> to
```

```
{0=all|1=scc|2=hcc|4=disj|
 8=min|16=show} atoms
```

Engage heuristic modifications (in both settings!)

```
--heuristic=Domain
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Inclusion-minimal stable models

- Consider a logic program containing a minimize statement of form
 - `#minimize{ a_1, \dots, a_n }`
- Computing one inclusion-minimal stable model can be done either via
 - `#heuristic a_i [1,false].` for $i = 1, \dots, n$, or
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- Computing all inclusion-minimal stable model can be done
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Applications: Overview

20 Train scheduling

21 Robotic intra-logistics

Outline

20 Train scheduling

21 Robotic intra-logistics

Motivation

- Increasing railway traffic demands global and flexible ways for scheduling trains in order to use railway networks to capacity
- Difficulty arises from dependencies among trains induced by connections and shared resources
- Train scheduling combines three distinct tasks
 - Routing
 - Conflict detection and resolution
 - Scheduling
- Solution operational at Swiss Federal Railway using *clingo*[DL]
 - ASP
 - Difference constraints
 - (Hybrid) Optimization
 - Heuristic directives
 - Multi-shot solving

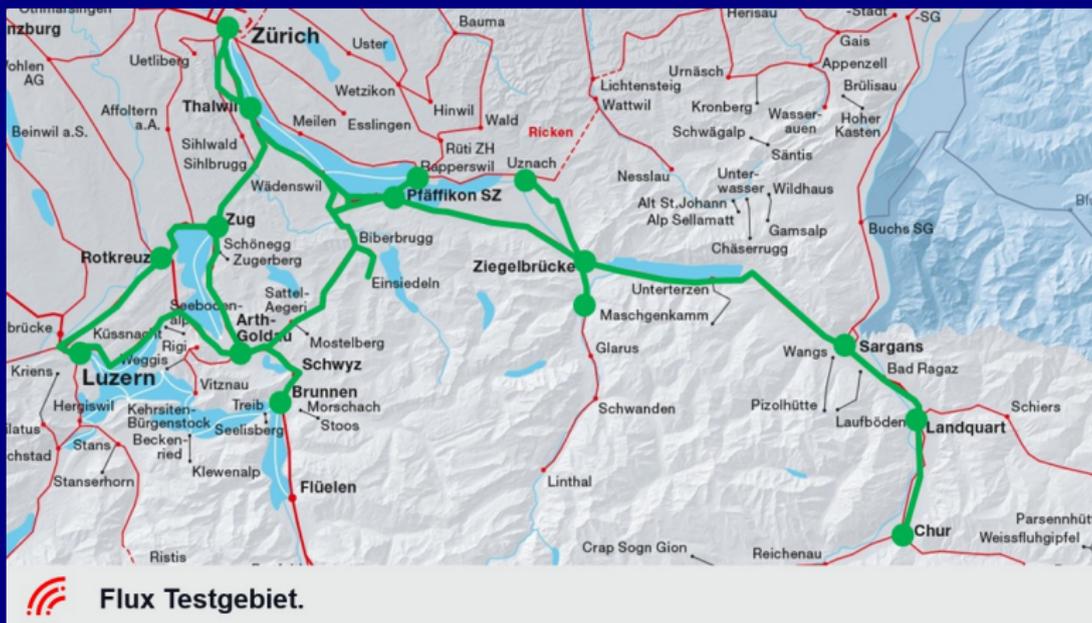
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Benchmark



We optimally solved the train scheduling problem on real-world railway networks spanning about 150 km with up to 467 trains within 5  minutes 

Outline

20 Train scheduling

21 Robotic intra-logistics

Motivation

- Objective How to develop robust and scalable AI technology for dealing with complex dynamic application scenarios?

- What's needed? — a fruit fly!

Robotic intra-logistics

- Why?
 - rich multi-faceted, full of variations
 - scalable layout, objects, granularity
 - measurable makespan, energy, quality of service
 - integrative mapf, data, constraints, decisions
 - relevant industry 4.0
- What for? — enabling research and teaching

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- What for? — enabling research and teaching

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- Objective How to develop robust and scalable KRR technology for dealing with complex dynamic application scenarios?

- What's needed? — a model scenario

Robotic intra-logistics

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Robotic intra-logistics

- Robotics systems for logistics and warehouse automation based on many
 - mobile robots
 - movable shelves
- Main tasks: order fulfillment, i.e.
 - routing
 - order picking
 - replenishment
- Many competing industry solutions:
 - Amazon, Dematic, Genzembach, Gray Orange, Swisslog



Robotic intra-logistics at Amazon

Robotic intra-logistics at Swisslog

What's (not) in the picture?

- Objects
floor, robots, shelves, products, people, etc.
- Relations
positions, carries/d, capacity, orientation, durations, etc.
- Actions
move, pickup, putdown, pick, charge, restock, etc.
- Objectives
deadlines, throughput, exploitation, energy management, human machine interaction, etc.

Outline

22 Summary

Potassco

- Potassco Systems — <http://potassco.org>
 - Academic branch
 - Freely available systems
 - Open source license (MIT)

- Potassco Solutions — <http://potassco.com>
 - Service branch
 - Consulting
 - Engineering
 - Maintenance
 - Training

Sites Germany (HQ@Potsdam), Australia, Austria, China, Cyprus, Finland, France, Japan, Portugal, Spain, Turkey

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The benefits of ASP

It's yours!

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

- + Generality
- + Effectiveness
- + Optimality
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Knowledge

Solver

ASP is a technology,
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