

Answer Set Programming: Boolean Constraint Solving for Knowledge Representation and Reasoning

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University of Potsdam



Outline

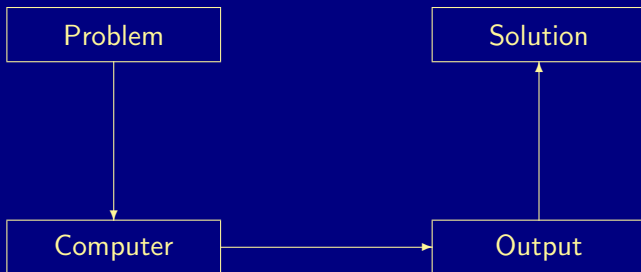
- 1 Motivation
- 2 Introduction
- 3 Modeling
- 4 (Grounding)
- 5 Solving
- 6 Potassco
- 7 Summary

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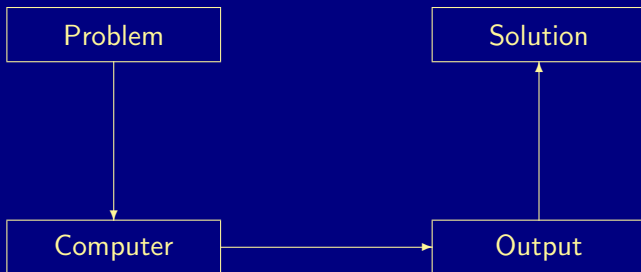
Informatics

“What is the problem?” versus *“How to solve the problem?”*



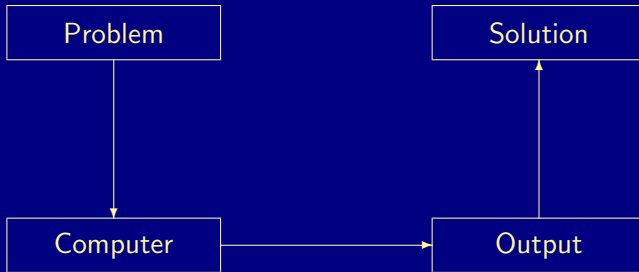
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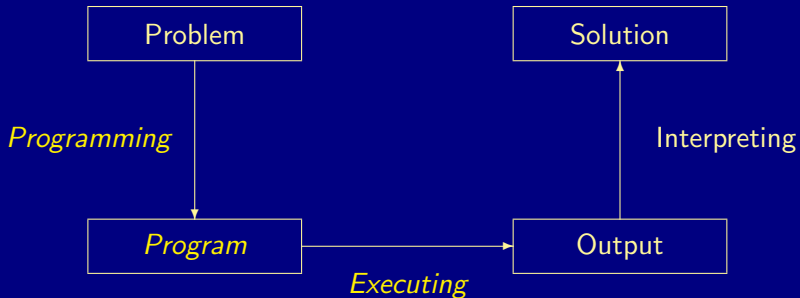
Traditional programming

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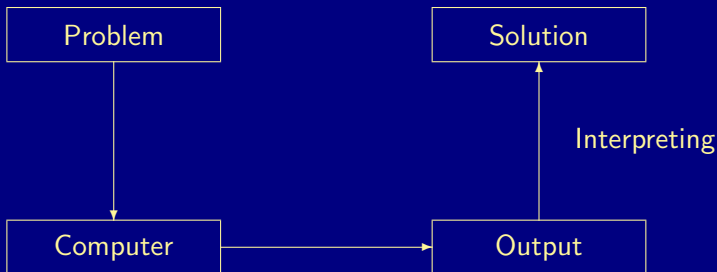
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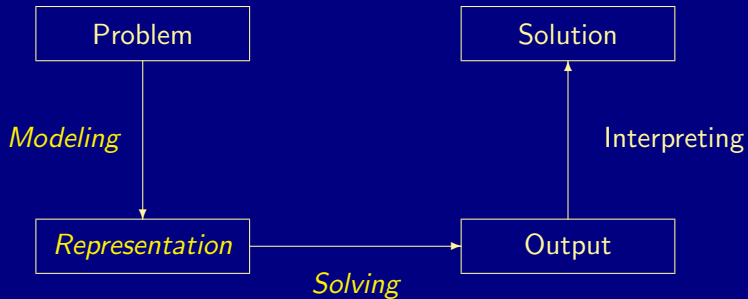
Declarative problem solving

“What is the problem?” versus *“How to solve the problem?”*



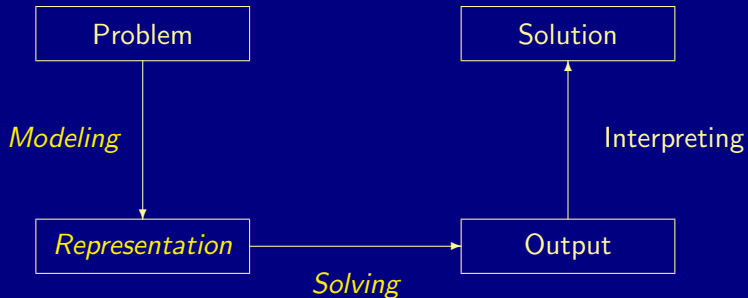
Declarative problem solving

“What is the problem?” versus *“How to solve the problem?”*



Declarative problem solving

“What is the problem?” versus *“How to solve the problem?”*



Answer Set Programming

in a Nutshell

ASP is an approach to declarative problem solving, combining
a rich yet simple modeling language
with high-performance solving capacities

ASP has its roots in

- (deductive) databases

- logic programming (with negation)

- (logic-based) knowledge representation and (nonmonotonic) reasoning
- constraint solving (in particular, SATisfiability testing)

ASP allows for solving all search problems in NP (and NP^{NP})
in a uniform way

ASP is versatile as reflected by the ASP solver *clasp*, winning
first places at ASP, CASC, MISC, PB, and SAT competitions

ASP embraces many emerging application areas, and users

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KR's shift of paradigm

Theorem Proving based approach (eg. Prolog)

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

Model Generation based approach (eg. SATisfiability testing)

- 1 Provide a representation of the problem
- 2 A solution is given by a model of the representation

Automated planning, Kautz and Selman (ECAI'92)

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

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Model Generation based Problem Solving

Representation	Solution
constraint satisfaction problem	assignment
propositional horn theories	smallest model
propositional theories	models
propositional theories	minimal models
propositional theories	stable models
propositional programs	minimal models
propositional programs	supported models
propositional programs	stable models
first-order theories	models
first-order theories	minimal models
first-order theories	stable models
first-order theories	Herbrand models
auto-epistemic theories	expansions
default theories	extensions

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Model Generation based Problem Solving

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Answer Set Programming *in general*

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Answer Set Programming *in practice*

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LP-style playing with blocks

Prolog program

```
on(a,b).  
on(b,c).  
  
above(X,Y) :- on(X,Y).  
above(X,Y) :- on(X,Z), above(Z,Y).
```

Prolog queries

```
?- above(a,c).  
true.  
  
?- above(c,a).  
no.
```

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Prolog queries (testing entailment)

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LP-style playing with blocks

Shuffled Prolog program

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Prolog queries

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?- above(a,c).
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Fatal Error: local stack overflow.
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Prolog queries (answered via fixed execution)

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SAT-style playing with blocks

Formula

$$\begin{aligned}
 & on(a, b) \\
 \wedge & on(b, c) \\
 \wedge & (on(X, Y) \rightarrow above(X, Y)) \\
 \wedge & (on(X, Z) \wedge above(Z, Y) \rightarrow above(X, Y))
 \end{aligned}$$

Herbrand model

$$\left\{ \begin{array}{cccccc}
 on(a, b), & on(b, c), & on(a, c), & on(b, b), & & \\
 above(a, b), & above(b, c), & above(a, c), & above(b, b), & above(c, b) &
 \end{array} \right\}$$

SAT-style playing with blocks

Formula

$$\begin{aligned}
 & on(a, b) \quad \text{LATEX} \\
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$on(a, b)$
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p cnf 18 38

2 0

6 0

-1 10 0

-2 11 0

-3 12 0

-4 13 0

-5 14 0

-6 15 0

-7 16 0

-8 17 0

-9 18 0

-1 -10 -10 0

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DIMACS

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Herbrand model (among 426!)

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➡ **Answer Set Programming (ASP)**

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Stable Herbrand model

{ on(a,b), on(b,c), above(b,c), above(a,b), above(a,c) }

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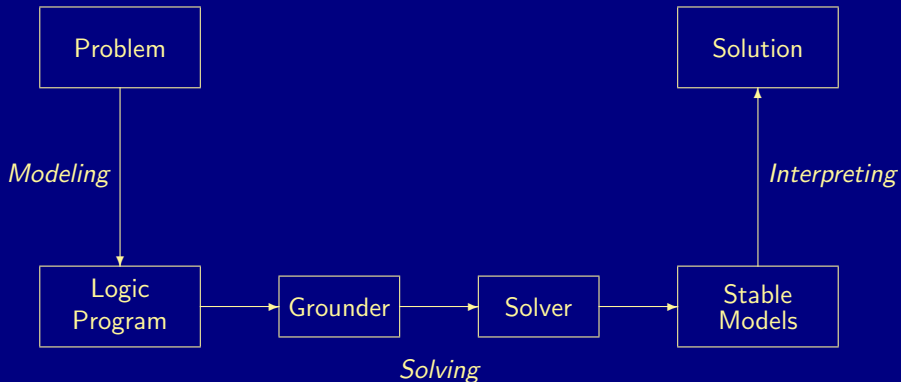
ASP versus LP

ASP	Prolog
Model generation	Query orientation
Bottom-up	Top-down
Modeling language	Programming language
Rule-based format	
Instantiation	Unification
Flat terms	Nested terms
(Turing +) $NP^{(NP)}$	Turing

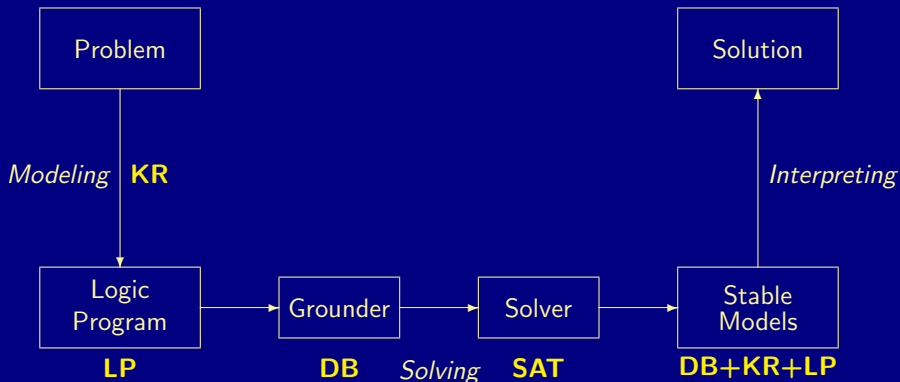
ASP versus SAT

ASP	SAT
Model generation	
Bottom-up	
Constructive Logic	Classical Logic
Closed (and open) world reasoning	Open world reasoning
Modeling language	—
Complex reasoning modes	Satisfiability testing
Satisfiability	Satisfiability
Enumeration/Projection	—
Intersection/Union	—
Optimization	—
(Turing +) $NP(NP)$	NP

ASP solving



Rooting ASP solving



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- **Declarativity** ASP does separate a problem's representation from the algorithms used for solving it
- **Scalability** ASP does not separate a problem's representation from its induced combinatorics

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$$\mathbf{ASP = DB + LP + KR + SAT}$$

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Semantic idea

Consider the logical formula Φ and its three (classical) models:

$$\{p, q\}, \{q, r\}, \text{ and } \{p, q, r\}$$

Formula Φ has one stable model, often called answer set:

$$\{p, q\}$$

$$\Phi \quad \boxed{q \wedge (q \wedge \neg r \rightarrow p)}$$

$$P_\Phi \quad \boxed{\begin{array}{l} q \leftarrow \\ p \leftarrow q, \sim r \end{array}}$$

Informally, a set X of atoms is a stable model of a logic program P if X is a (classical) model of P and if all atoms in X are justified by some rule in P (rooted in intuitionistic logics HT (Heyting, 1930) and G3 (Gödel, 1932))

Semantic idea

Consider the logical formula Φ and its three (classical) models:

$$\{p, q\}, \{q, r\}, \text{ and } \{p, q, r\}$$

Formula Φ has one stable model, often called answer set:

$$\{p, q\}$$

$$\Phi \quad q \wedge (q \wedge \neg r \rightarrow p)$$

$$P_\Phi \quad \begin{array}{l} q \leftarrow \\ p \leftarrow q, \sim r \end{array}$$

Informally, a set X of atoms is a stable model of a logic program P

- if X is a (classical) model of P and
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(rooted in intuitionistic logics HT (Heyting, 1930) and G3 (Gödel, 1932))

Semantic idea

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p	\mapsto	1
q	\mapsto	1
r	\mapsto	0

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Making things precise

■ Syntax

- A (normal) rule is an expression of the form

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

where $0 \leq m, n$ and each a, b_i, c_j is an atom

- A logic program is a finite set of rules

■ Semantics

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

$$P^X = \{ a \leftarrow b_1, \dots, b_m \mid r \in P \text{ and } \{c_1, \dots, c_n\} \cap X = \emptyset \}$$

The \subseteq -smallest model of P^X is denoted by $C_n(P^X)$

A set X of atoms is an stable model of a program P , if

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A simple example

$$P = \{p \leftarrow p, q \leftarrow \sim p\}$$

X	P^X	$Cn(P^X)$ <i>stable?</i>
$\{ \}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$
$\{p \}$	$p \leftarrow p$	\emptyset
$\{ q \}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$
$\{p, q\}$	$p \leftarrow p$	\emptyset

A simple example

$$P = \{p \leftarrow p, q \leftarrow \sim p\}$$

X	P^X	$Cn(P^X)$	<i>stable?</i>
$\{\}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$	\times
$\{p\}$	$p \leftarrow p$	\emptyset	\times
$\{q\}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$	\checkmark
$\{p, q\}$	$p \leftarrow p$	\emptyset	\times

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$\{ \}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$	\times
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$\{ \}$	$p \leftarrow p$ $q \leftarrow$	$\{q\}$	\times
$\{p\}$	$p \leftarrow p$	\emptyset	\times
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A simple example

$$P = \{p \leftarrow p, q \leftarrow \sim p\}$$

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

A simple example

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

A simple example

$$P = \{p \leftarrow p, q \leftarrow \sim p\}$$

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

A simple example

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

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

A simple example

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

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

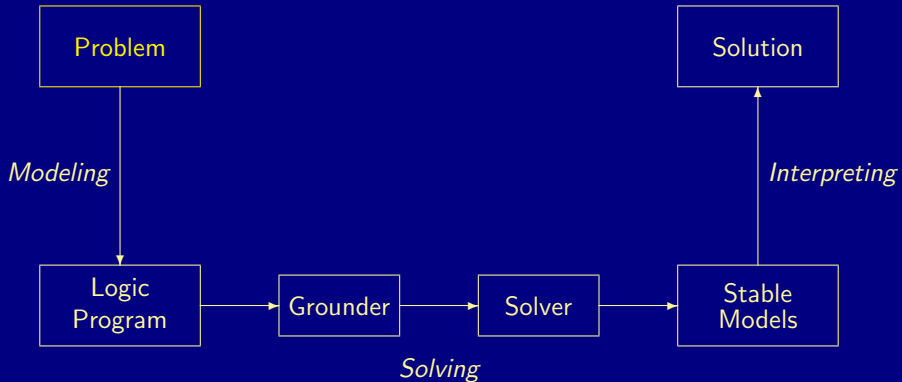
Some properties

- A logic program may have zero, one, or multiple stable models
- If X is a stable model of a logic program P , then X is a model of P (seen as a formula)
- If X and Y are stable models of a *normal* program P , then $X \not\subseteq Y$

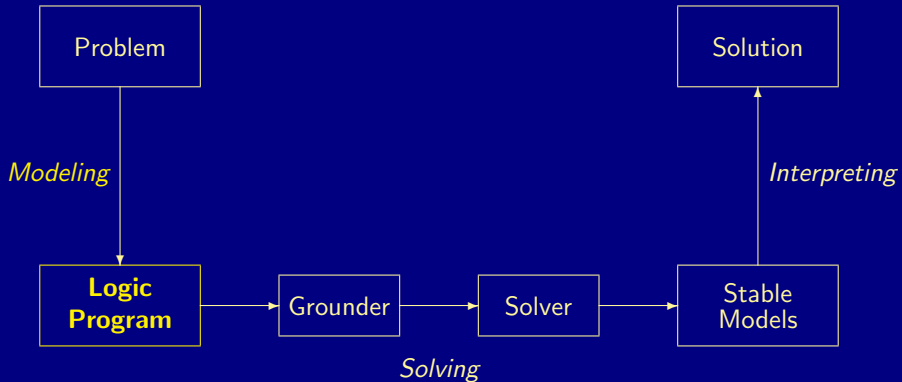
Outline

- 1 Motivation
- 2 Introduction
- 3 Modeling**
- 4 (Grounding)
- 5 Solving
- 6 Potassco
- 7 Summary

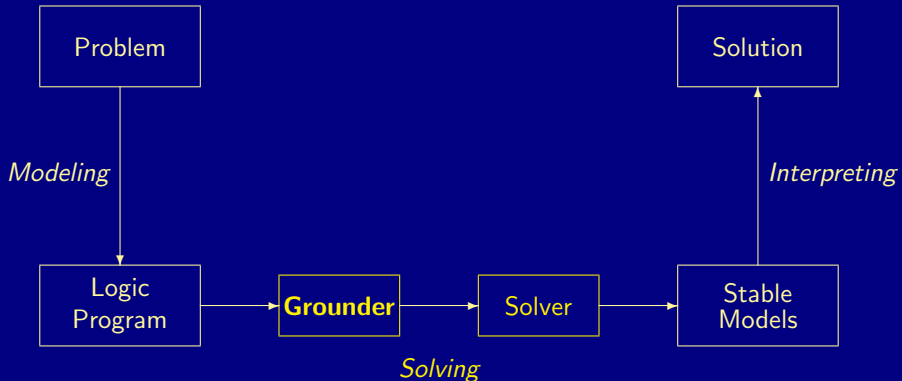
ASP solving process



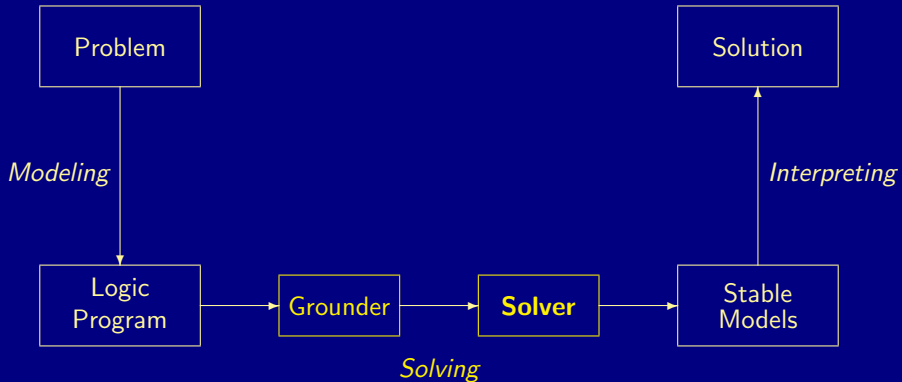
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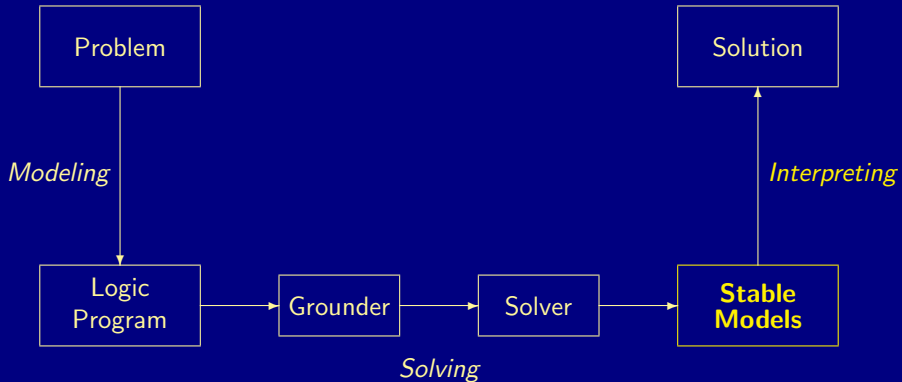
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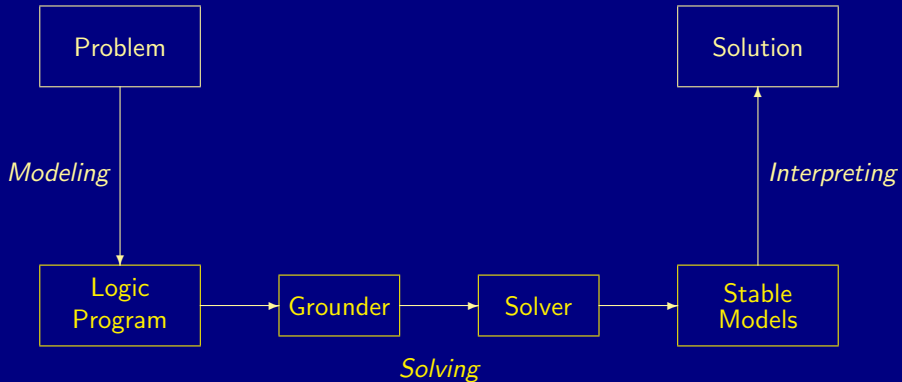
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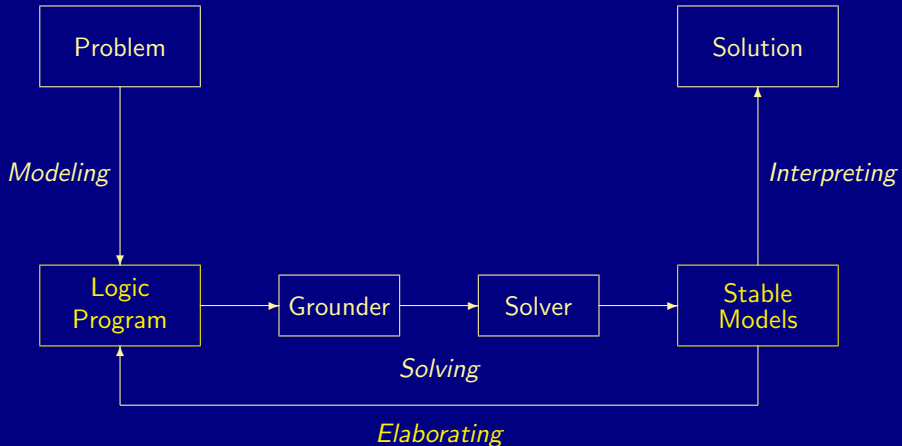
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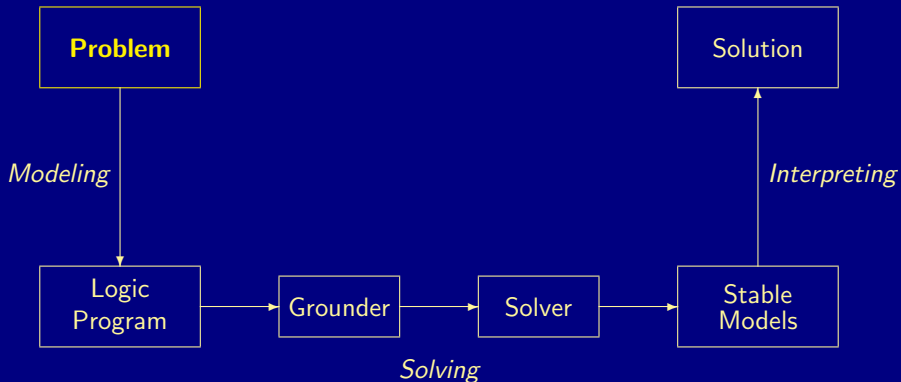
ASP solving process



Some language constructs

- Variables (over the Herbrand universe)
 - $p(X) :- q(X)$ over constants $\{a, b, c\}$ stands for
 $p(a) :- q(a), p(b) :- q(b), p(c) :- q(c)$
- Conditional Literals
 - $p :- q(X) : r(X)$ given $r(a), r(b), r(c)$ stands for
 $p :- q(a), q(b), q(c)$
- Disjunction
 - $p(X) \mid q(X) :- r(X)$
- Integrity Constraints
 - $:- q(X), p(X)$
- Choice
 - $2 \{ p(X,Y) : q(X) \} 7 :- r(Y)$
- Aggregates
 - $s(Y) :- r(Y), 2 \#count \{ p(X,Y) : q(X) \} 7$
 - also: $\#sum, \#avg, \#min, \#max, \#even, \#odd$

A case-study: Graph coloring



Graph coloring

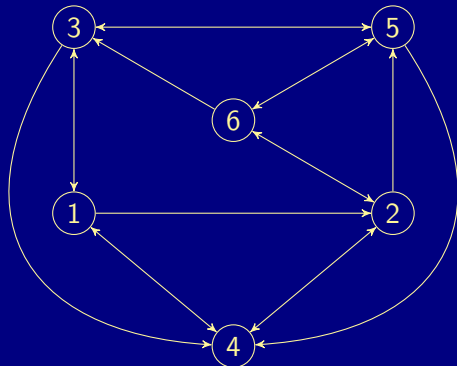
- Problem instance A graph consisting of nodes and edges

Graph coloring

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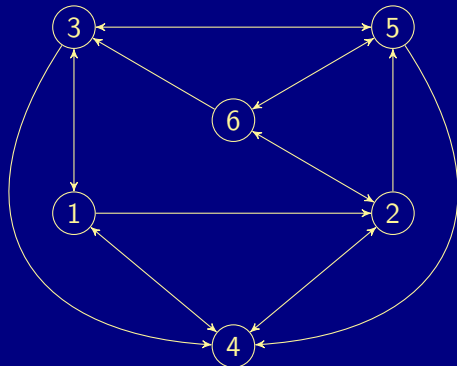
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Graph coloring

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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

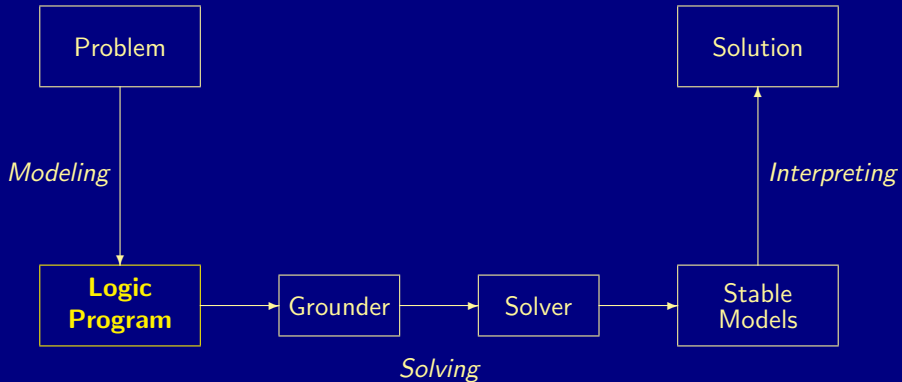
Graph coloring

- Problem instance A graph consisting of nodes and edges
 - facts formed by predicates `node/1` and `edge/2`
 - facts formed by predicate `col/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 a unique 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).
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col(r).  col(b).  col(g).
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```
1 { color(X,C) : col(C) } 1 :- node(X).
:- edge(X,Y), color(X,C), color(Y,C).
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Problem
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Problem
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Problem
instance

Problem
encoding

color.lp

```
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).
```

```
col(r). col(b). col(g).
```

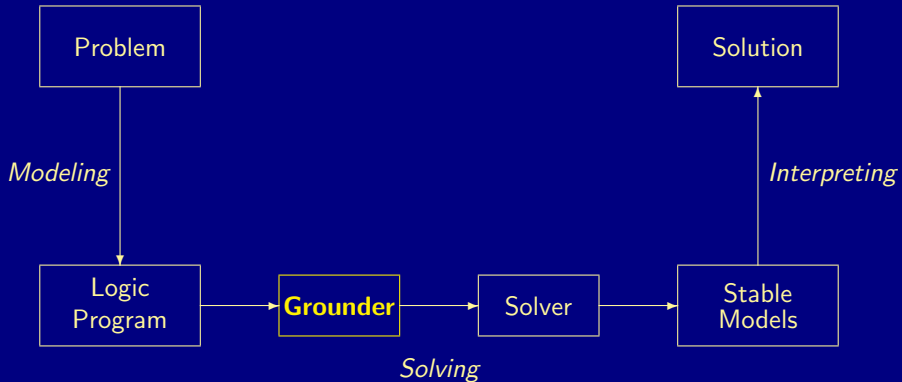
```
1 { color(X,C) : col(C) } 1 :- node(X).
```

```
:- edge(X,Y), color(X,C), color(Y,C).
```

Problem
instance

Problem
encoding

ASP solving process



Graph coloring: Grounding

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

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

```
col(r). col(b). col(g).
```

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

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

Graph coloring: Grounding

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

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

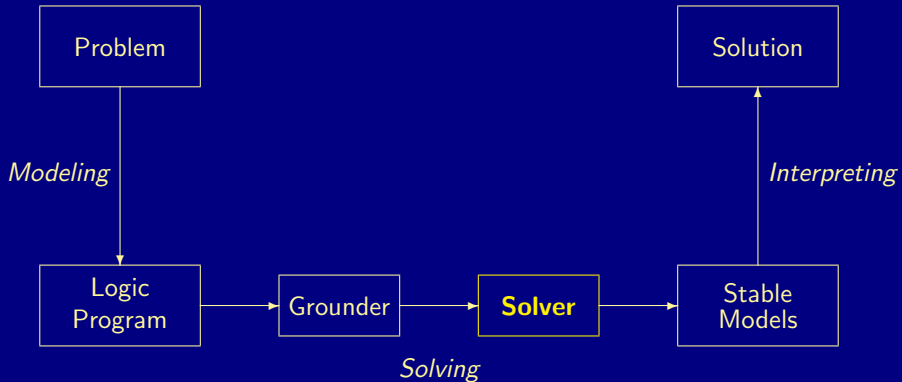
```
edge(1,2). edge(1,3). edge(1,4). edge(2,4). edge(2,5). edge(2,6).
edge(3,1). edge(3,4). edge(3,5). edge(4,1). edge(4,2). edge(5,3).
edge(5,4). edge(5,6). edge(6,2). edge(6,3). edge(6,5).
```

```
col(r). col(b). col(g).
```

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

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


ASP solving process



Graph coloring: Solving

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

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

Models      : 6
Time        : 0.002s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
CPU Time    : 0.000s
```

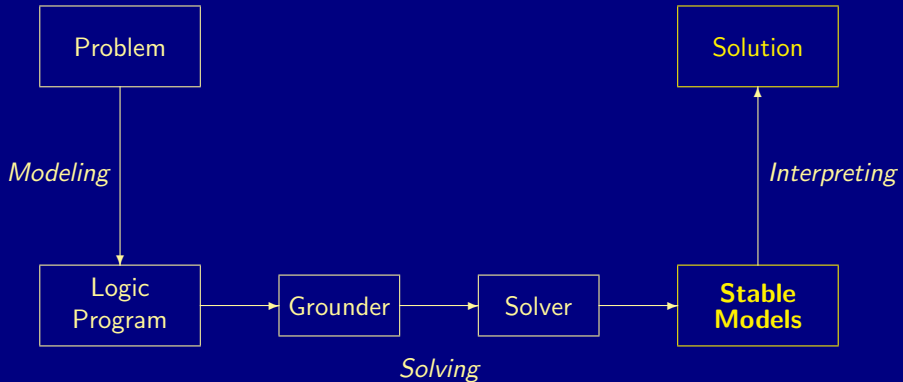
Graph coloring: Solving

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

```
clasp version 2.1.0
Reading from stdin
Solving...
Answer: 1
edge(1,2) ... col(r) ... node(1) ... color(6,b) color(5,g) color(4,b) color(3,r) color(2,r) color(1,g)
Answer: 2
edge(1,2) ... col(r) ... node(1) ... color(6,r) color(5,g) color(4,r) color(3,b) color(2,b) color(1,g)
Answer: 3
edge(1,2) ... col(r) ... node(1) ... color(6,g) color(5,b) color(4,g) color(3,r) color(2,r) color(1,b)
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Models      : 6
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```

ASP solving process

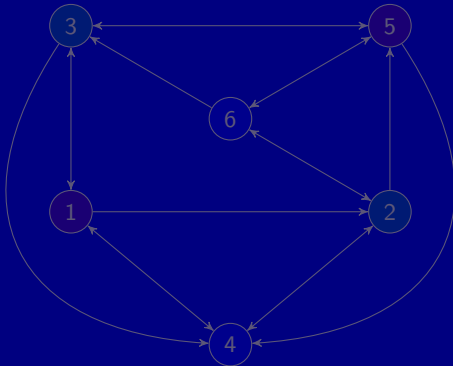


A coloring

Answer: 6

```
edge(1,2) ... col(r) ... node(1) ...
```

```
color(6,b) color(5,r) color(4,b) color(3,g) color(2,g) color(1,r)
```

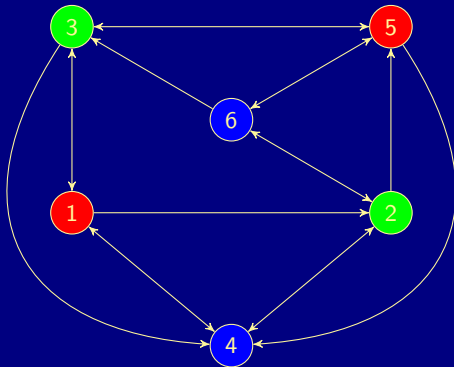


A coloring

Answer: 6

```
edge(1,2) ... col(r) ... node(1) ...
```

```
color(6,b) color(5,r) color(4,b) color(3,g) color(2,g) color(1,r)
```



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

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Nutshell

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

Satisfiability testing

$$(a \leftrightarrow b) \wedge c$$

Satisfiability testing

$$(a \leftrightarrow b) \wedge c$$

```
{ a ; b ; c }.
```

```
:- not a, b.
```

```
:- a, not b.
```

```
:- not c.
```

Maximum satisfiability testing

“(~~$a \leftrightarrow b$~~)” + $(a \leftrightarrow b) \wedge c$

```
{ a ; b ; c }.
```

```
:- not a, b.
```

```
:- a, not b.
```

```
:- not c.
```

```
:~ a, b. [42@1]
```

```
:~ not a, not b. [69@2]
```

n-queens

Basic encoding

```
{ queen(1..n,1..n) }.
```

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

```
:- queen(I,J), queen(I,JJ), J != JJ.
```

```
:- queen(I,J), queen(II,J), I != II.
```

```
:- queen(I,J), queen(II,JJ), (I,J) != (II,JJ), I-J == II-JJ.
```

```
:- queen(I,J), queen(II,JJ), (I,J) != (II,JJ), I+J == II+JJ.
```

n-queens

Advanced encoding

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

n-queens

(Experimental) constraint encoding

```
1 $<= $queen(1..n) $<= n.  
  
#disjoint { X : $queen(X) $+ 0 : X=1..n }.  
#disjoint { X : $queen(X) $+ X : X=1..n }.  
#disjoint { X : $queen(X) $- X : X=1..n }.
```

Traveling salesperson

Basic encoding

```
1 { cycle(X,Y) : edge(X,Y) } 1 :- node(X).
1 { cycle(X,Y) : edge(X,Y) } 1 :- node(Y).

reached(X) :- X = #min { Y : node(Y) }.
reached(Y) :- cycle(X,Y), reached(X).

:- node(Y), not reached(Y).

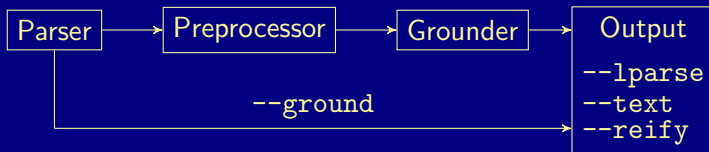
#minimize { C,X,Y : cycle(X,Y), cost(X,Y,C) }.
```

Outline

- 1 Motivation
- 2 Introduction
- 3 Modeling
- 4 (Grounding)**
- 5 Solving
- 6 Potassco
- 7 Summary

gringo

- Accepts safe programs with aggregates
- Tolerates unrestricted use of function symbols (as long as it yields a finite ground instantiation :)
- Expressive power of a Turing machine
- Basic architecture of *gringo*:



Grounding by example

 $d(a)$ $d(c)$ $d(d)$ $p(a, b)$ $p(b, c)$ $p(c, d)$ $p(X, Z) \leftarrow p(X, Y), p(Y, Z)$ $q(a)$ $q(b)$ $q(X) \leftarrow \sim r(X), d(X)$ $r(X) \leftarrow \sim q(X), d(X)$ $s(X) \leftarrow \sim r(X), p(X, Y), q(Y)$

Grounding by example

Safe ?

 $d(a)$ $d(c)$ $d(d)$ $p(a, b)$ $p(b, c)$ $p(c, d)$ $p(X, Z) \leftarrow p(X, Y), p(Y, Z)$ $q(a)$ $q(b)$ $q(X) \leftarrow \sim r(X), d(X)$ $r(X) \leftarrow \sim q(X), d(X)$ $s(X) \leftarrow \sim r(X), p(X, Y), q(Y)$

Grounding by example

	Safe ?
$d(a)$	✓
$d(c)$	✓
$d(d)$	✓
$p(a, b)$	✓
$p(b, c)$	✓
$p(c, d)$	✓
$p(X, Z) \leftarrow p(X, Y), p(Y, Z)$	
$q(a)$	✓
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$q(X) \leftarrow \sim r(X), d(X)$	
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Grounding by example

	Safe ?
$d(a)$	✓
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$p(a, b)$	✓
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$p(c, d)$	✓
$p(X, Z) \leftarrow p(X, Y), p(Y, Z)$	
$q(a)$	✓
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$q(X) \leftarrow \sim r(X), d(X)$	
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Grounding by example

	Safe ?
$d(a)$	✓
$d(c)$	✓
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$p(X, Z) \leftarrow p(X, Y), p(Y, Z)$	✓
$q(a)$	✓
$q(b)$	✓
$q(X) \leftarrow \sim r(X), d(X)$	✓
$r(X) \leftarrow \sim q(X), d(X)$	✓
$s(X) \leftarrow \sim r(X), p(X, Y), q(Y)$	✓

Match

- A **substitution** is a mapping from variables to terms
- Given sets B and D of atoms, a substitution θ is a **match** of B in D , if $B\theta \subseteq D$
- Given a set B of atoms and a set D of ground atoms, define

$$\Theta(B, D) = \{ \theta \mid \theta \text{ is a } \subseteq\text{-minimal match of } B \text{ in } D \}$$

- Example $\{X \mapsto 1\}$ and $\{X \mapsto 2\}$ are \subseteq -minimal matches of $\{p(X)\}$ in $\{p(1), p(2), p(3)\}$, while match $\{X \mapsto 1, Y \mapsto 2\}$ is not

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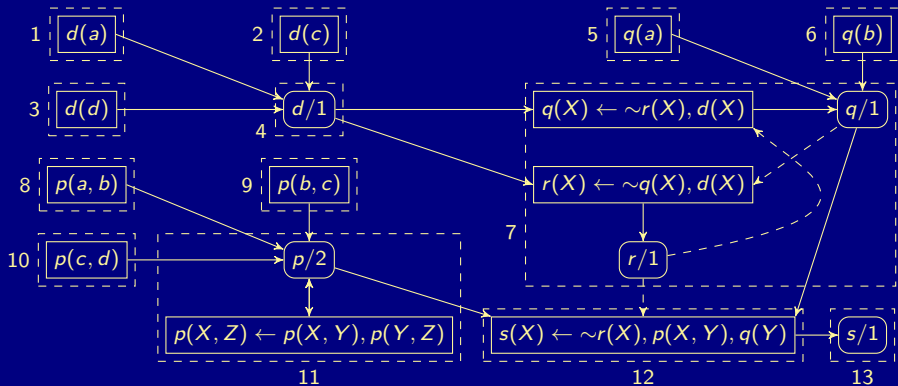
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Naive instantiation

Algorithm 1: NAIVEINSTANTIATION**Input** : A safe (first-order) logic program P **Output** : A ground logic program P' $D := \emptyset$ $P' := \emptyset$ **repeat** $D' := D$ **foreach** $r \in P$ **do** $B := \text{body}(r)^+$ **foreach** $\theta \in \Theta(B, D)$ **do** $D := D \cup \{\text{head}(r)\theta\}$ $P' := P' \cup \{r\theta\}$ **until** $D = D'$

Predicate-rule dependency graph



Instantiation

SCC	$\Theta(B, D)$	D	P'
1	$\{\emptyset\}$	$d(a)$	$d(a) \leftarrow$
2	$\{\emptyset\}$	$d(c)$	$d(c) \leftarrow$
3	$\{\emptyset\}$	$d(d)$	$d(d) \leftarrow$
5	$\{\emptyset\}$	$q(a)$	$q(a) \leftarrow$
6	$\{\emptyset\}$	$q(b)$	$q(b) \leftarrow$
7	$\{\{X \mapsto a\},$ $\{X \mapsto c\},$ $\{X \mapsto d\},$ $\{X \mapsto a\},$ $\{X \mapsto c\},$ $\{X \mapsto d\}\}$	$q(c)$ $q(d)$ $r(c)$ $r(d)$	$q(a) \leftarrow \sim r(a), d(a)$ $q(c) \leftarrow \sim r(c), d(c)$ $q(d) \leftarrow \sim r(d), d(d)$ $r(a) \leftarrow \sim q(a), d(a)$ $r(c) \leftarrow \sim q(c), d(c)$ $r(d) \leftarrow \sim q(d), d(d)$

Instantiation

SCC	$\Theta(B, D)$	D	P'
8	$\{\emptyset\}$	$p(a, b)$	$p(a, b) \leftarrow$
9	$\{\emptyset\}$	$p(b, c)$	$p(b, c) \leftarrow$
10	$\{\emptyset\}$	$p(c, d)$	$p(c, d) \leftarrow$
11	$\{\{X \mapsto a, Y \mapsto b, Z \mapsto c\},$ $\{X \mapsto b, Y \mapsto c, Z \mapsto d\}\}$	$p(a, c)$ $p(b, d)$	$p(a, c) \leftarrow p(a, b), p(b, c)$ $p(b, d) \leftarrow p(b, c), p(c, d)$
	$\{\{X \mapsto a, Y \mapsto c, Z \mapsto d\},$ $\{X \mapsto a, Y \mapsto b, Z \mapsto d\}\}$	$p(a, d)$	$p(a, d) \leftarrow p(a, c), p(c, d)$ $p(a, d) \leftarrow p(a, b), p(b, d)$
12	$\{X \mapsto a, Y \mapsto b\},$	$s(a)$	$s(a) \leftarrow \sim r(a), p(a, b), q(b)$
	$\{X \mapsto a, Y \mapsto c\},$		$s(a) \leftarrow \sim r(a), p(a, c), q(c)$
	$\{X \mapsto a, Y \mapsto d\},$		$s(a) \leftarrow \sim r(a), p(a, d), q(d)$
	$\{X \mapsto b, Y \mapsto c\},$	$s(b)$	$s(b) \leftarrow \sim r(b), p(b, c), q(c)$
	$\{X \mapsto b, Y \mapsto d\},$		$s(b) \leftarrow \sim r(b), p(b, d), q(d)$
	$\{X \mapsto c, Y \mapsto d\}$	$s(c)$	$s(c) \leftarrow \sim r(c), p(c, d), q(d)$

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- 3 Modeling
- 4 (Grounding)
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clasp

- *clasp* is a native ASP solver combining conflict-driven search with sophisticated reasoning techniques:
 - advanced preprocessing, including equivalence reasoning
 - lookback-based decision heuristics
 - restart policies
 - nogood deletion
 - progress saving
 - dedicated data structures for binary and ternary nogoods
 - lazy data structures (watched literals) for long nogoods
 - dedicated data structures for cardinality and weight constraints
 - lazy unfounded set checking based on “source pointers”
 - tight integration of unit propagation and unfounded set checking
 - various reasoning modes
 - parallel search
 - ...

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Reasoning modes of *clasp*

- Beyond deciding (stable) model existence, *clasp* allows for:
 - Optimization
 - Enumeration (without solution recording)
 - Projective enumeration (without solution recording)
 - Intersection and Union (linear solution computation)
 - and combinations thereof
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 - ASP solving (*smodels* format)
 - MaxSAT and SAT solving (extended *dimacs* format)
 - PB solving (*opb* and *wbo* format)

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Parallel search in *clasp*

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 - pursues a coarse-grained, task-parallel approach to parallel search via shared memory multi-threading
 - up to 64 configurable (non-hierarchic) threads
 - allows for parallel solving via search space splitting and/or competing strategies
 - both supported by solver portfolios
 - features different nogood exchange policies

Parallel search in *clasp*

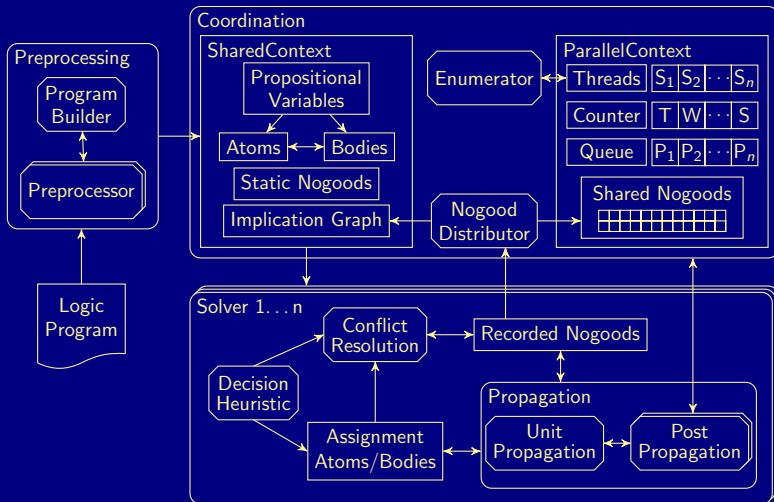
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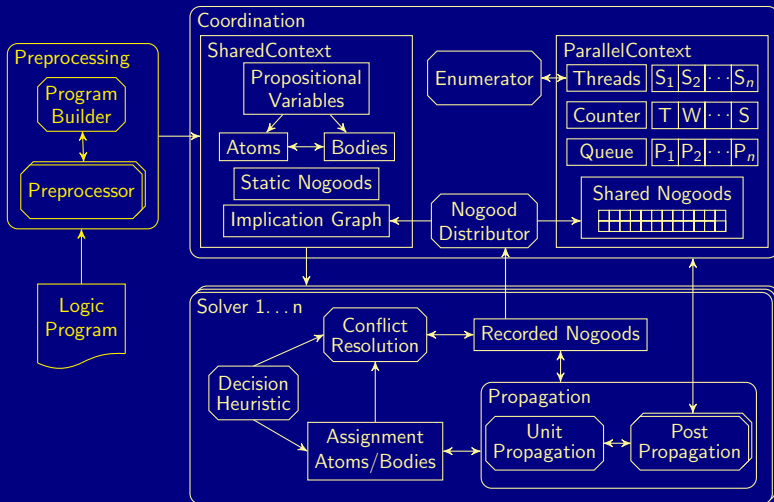
Parallel search in *clasp*

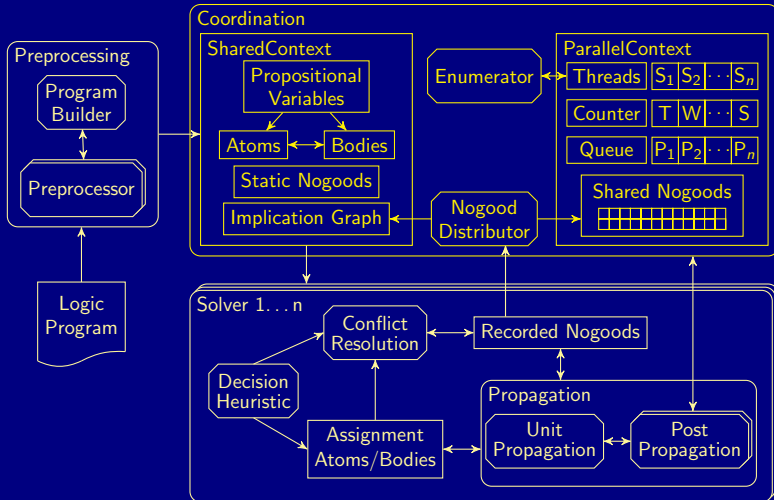
- *clasp*
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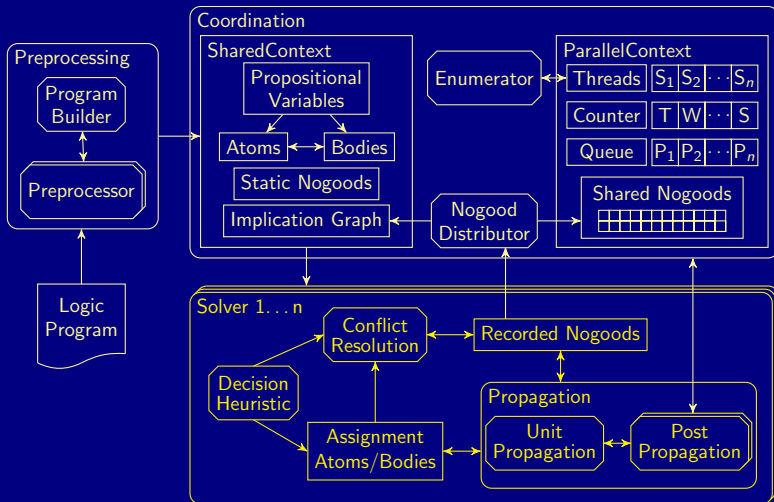
Parallel search in *clasp*

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Multi-threaded architecture of *clasp*

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Comparing configurations

on advanced n-queens encoding

n	frumpy	jumpy	handy	crafty	trendy	chatty
50	0.063	0.023	3.416	0.030	1.805	0.061
100	20.364	0.099	7.891	0.136	7.321	0.121
150	60.000	0.212	14.522	0.271	19.883	0.347
200	60.000	0.415	15.026	0.667	32.476	0.753
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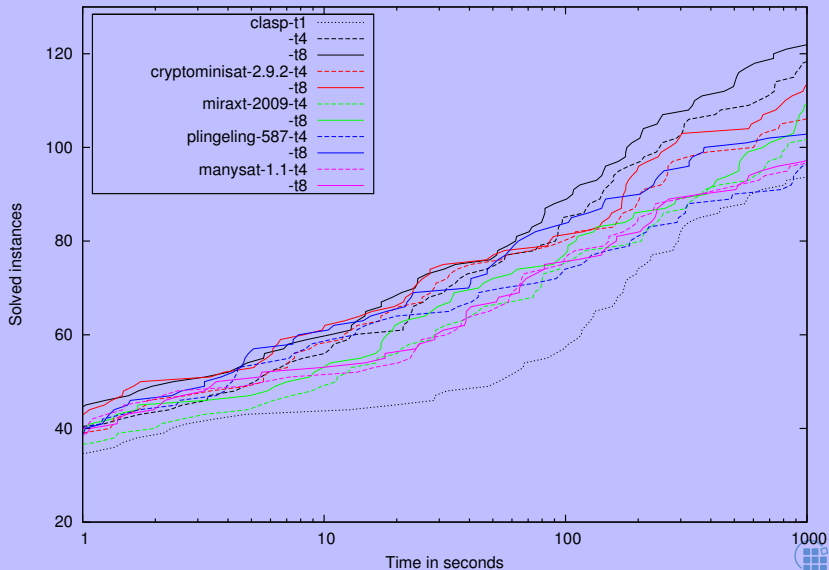
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clasp in context

- Compare *clasp* (2.0.5) to the multi-threaded SAT solvers
 - *cryptominisat* (2.9.2)
 - *manysat* (1.1)
 - *miraxt* (2009)
 - *plingeling* (587f)

all run with four and eight threads in their default settings

- 160/300 benchmarks from crafted category at SAT'11
 - all solvable by *ppfolio* in 1000 seconds
 - crafted SAT benchmarks are closest to ASP benchmarks

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clasp's default portfolio for parallel solving via `clasp --print-portfolio`

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[CRAFTY]: --heuristic=vsids --restarts=x,128,1.5 --deletion=3,75,10.0 --del-init-r=1000,9000 --del-grow=1.1,20.0
[TRENDY]: --heuristic=vsids --restarts=d,100,0.7 --deletion=3,50 --del-init=500,19500 --del-grow=1.1,20.0,x,100
[FRUMPY]: --heuristic=berkmin --restarts=x,100,1.5 --deletion=1,75 --del-init-r=200,40000 --del-max=400000 --de
[JUMPY]: --heuristic=vsids --restarts=l,100 --del-init-r=1000,20000 --del-algo=basic,2 --deletion=3,75 --del-g
[STRONG]: --heuristic=berkmin --restarts=x,100,1.5 --deletion=1,75 --del-init-r=200,40000 --del-max=400000 --de
[HANDY]: --heuristic=vsids --restarts=d,100,0.7 --deletion=2,50,20.0 --del-max=200000 --del-algo=sort,2 --del
[S2]: --heuristic=vsids --reverse-arcs=1 --otfs=1 --local-restarts --save-progress=0 --contraction=250 --counte
[S4]: --heuristic=vsids --restarts=l,256 --counter-restart=3 --strengthen=recursive --update-lbd --del-glue=2 --
[SLOW]: --heuristic=berkmin --berk-max=512 --restarts=f,16000 --lookahead=atom,50
[VMTF]: --heuristic=vmtf --str=no --contr=0 --restarts=x,100,1.3 --del-init-r=800,9200
[SIMPLE]: --heuristic=vsids --strengthen=recursive --restarts=x,100,1.5,15 --contraction=0
[LUBY-SP]: --heuristic=vsids --restarts=l,128 --save-p --otfs=1 --init-w=2 --contr=0 --opt-heu=3
[LOCAL-R]: --berk-max=512 --restarts=x,100,1.5,6 --local-restarts --init-w=2 --contr=0
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- *clasp's* portfolio is fully customizable
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Outline

- 1 Motivation
- 2 Introduction
- 3 Modeling
- 4 (Grounding)
- 5 Solving
- 6 Potassco**
- 7 Summary

potassco.sourceforge.net

Potassco, the Potsdam Answer Set Solving Collection,
bundles tools for ASP developed at the University of Potsdam,
for instance:

- **Grounder** gringo, lingo, pyngo
- **Solver** clasp, {a,h,un}clasp, claspD, claspfolio, claspar, aspeed
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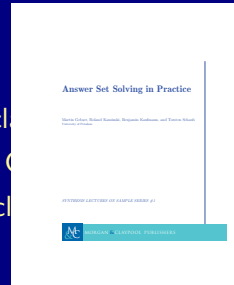
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