



# PROBLEM SOLVING AND SEARCH IN ARTIFICIAL INTELLIGENCE

## Lecture 7 ASP II \* slides adapted from Torsten Schaub [Gebser et al.(2012)]

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Dresden, 26th May and 2nd June 2017

# Agenda

- 1 Introduction
- 2 Constraint Satisfaction (CSP)
- 3 Uninformed Search versus Informed Search (Best First Search, A\* Search, Heuristics)
- 4 Local Search, Stochastic Hill Climbing, Simulated Annealing
- 5 Tabu Search
- 6 Answer-set Programming (ASP)
- 7 Structural Decomposition Techniques (Tree/Hypertree Decompositions)
- 8 Evolutionary Algorithms/ Genetic Algorithms

# Overview ASP II

- Modeling
  - ① Basic Modeling
  - ② Methodology
- Language
  - ③ Motivation
  - ④ Core language
  - ⑤ Extended language
- Language Extensions
  - ⑥ Two kinds of negation
  - ⑦ Disjunctive logic programs
- Computational Aspects
  - ⑧ Complexity

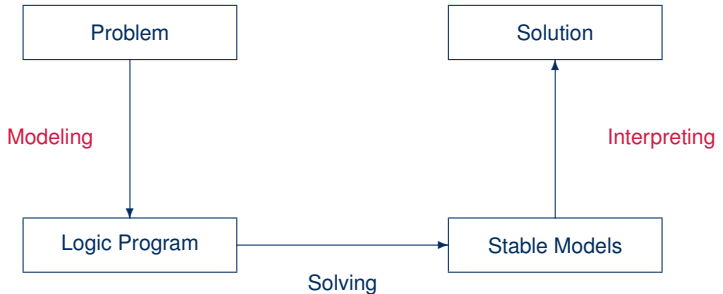
# Modeling: Overview

- 1 Basic Modeling
- 2 Methodology

# Outline

- 1 Basic Modeling
- 2 Methodology

# Modeling and Interpreting



# Modeling

- For solving a problem class  $\mathbf{C}$  for a problem instance  $\mathbf{I}$ , encode
  - 1 the problem instance  $\mathbf{I}$  as a set  $P_{\mathbf{I}}$  of facts and
  - 2 the problem class  $\mathbf{C}$  as a set  $P_{\mathbf{C}}$  of rulessuch that the solutions to  $\mathbf{C}$  for  $\mathbf{I}$  can be (polynomially) extracted from the stable models of  $P_{\mathbf{I}} \cup P_{\mathbf{C}}$
- $P_{\mathbf{I}}$  is (still) called **problem instance**
- $P_{\mathbf{C}}$  is often called the **problem encoding**
- An **encoding**  $P_{\mathbf{C}}$  is **uniform**, if it can be used to solve all its problem instances  
That is,  $P_{\mathbf{C}}$  encodes the solutions to  $\mathbf{C}$  for any set  $P_{\mathbf{I}}$  of facts

# Outline

- 1 Basic Modeling
- 2 Methodology



# 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)

# Basic methodology

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## Nutshell

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

# Outline

## 1 Basic Modeling

## 2 Methodology

- Satisfiability
- Queens
- Traveling Salesperson

# Satisfiability testing

- **Problem Instance:** A propositional formula  $\phi$  in CNF
- **Problem Class:** Is there an assignment of propositional variables to true and false such that a given formula  $\phi$  is true

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- **Example:** Consider formula

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

- **Logic Program:**

## Generator

$\{a, b\} \leftarrow$

## Tester

$\leftarrow \text{not } a, b$   
 $\leftarrow a, \text{not } b$

## Stable models

$X_1 = \{a, b\}$   
 $X_2 = \{\}$

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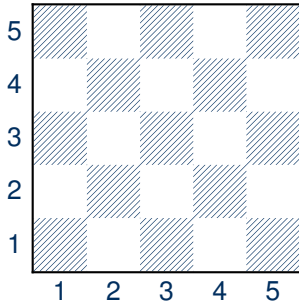
$X_2 = \{\}$



# Outline

- 1 Basic Modeling
- 2 Methodology
  - Satisfiability
  - **Queens**
  - Traveling Salesperson

# 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 ...

```
$ gringo queens.lp --const n=5 | clasp
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
  Prepare   : 0.000
  Prepro.   : 0.000
  Solving   : 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

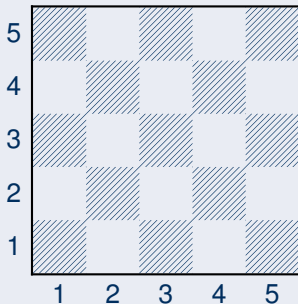
## Running ...

```
$ gringo queens.lp --const n=5 | clasp 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+
...
```

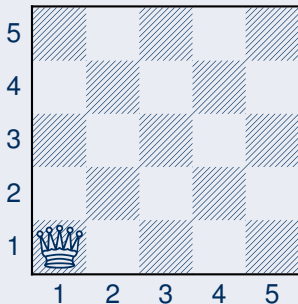
# Placing some Queens: Answer 1

## Answer 1



# Placing some Queens: Answer 2

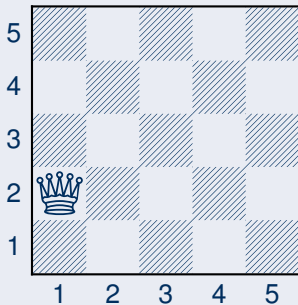
## Answer 2





# Placing some Queens: Answer 3

## Answer 3



# Placing $n$ Queens

queens.lp

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- not n { queen(I,J) } n.
```

- Place exactly  $n$  queens on the board

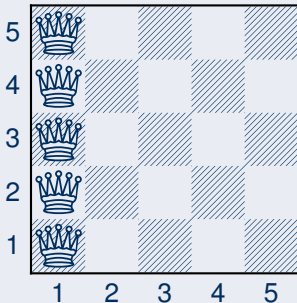
# Placing $n$ Queens

## Running ...

```
$ gringo queens.lp --const n=5 | clasp 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)
...
```

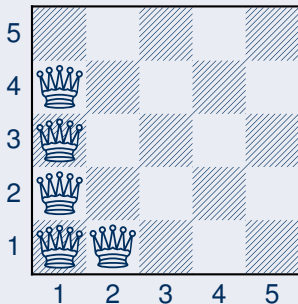
# Placing $n$ Queens: Answer 1

## Answer 1



# Placing $n$ Queens: Answer 2

## Answer 2



# Horizontal and Vertical Attack

queens.lp

```
row(1..n).  
col(1..n).  
{ queen(I,J) : row(I), col(J) }.  
:- not n { queen(I,J) } n.  
:- queen(I,J), queen(I,J'), J != J'.
```

- Forbid horizontal attacks

# Horizontal and Vertical Attack

queens.lp

```
row(1..n).
col(1..n).
{ queen(I,J) : row(I), col(J) }.
:- not n { 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

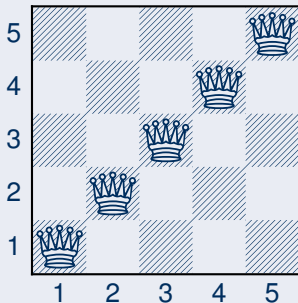
## Running ...

```
$ gringo queens.lp --const n=5 | clasp
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



# Diagonal Attack

```
queens.lp
```

```
row(1..n).
col(1..n).
{ queen(I,J) : row(I), col(J) }.
:- not n { 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

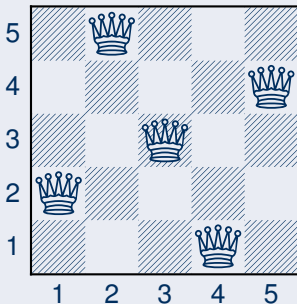
## Running ...

```
$ gringo queens.lp --const n=5 | clasp
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
  Prepare   : 0.000
  Prepro.   : 0.000
  Solving   : 0.000
```

# Diagonal Attack: Answer 1

## Answer 1



# Optimizing

```
queens-opt.lp
```

```
1 { queen(I,1..n) } 1 :- I = 1..n.  
1 { queen(1..n,J) } 1 :- J = 1..n.  
:- 2 { queen(D-J,J) }, D = 2..2*n.  
:- 2 { queen(D+J,J) }, 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=3
```

```
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%)
```

# Outline

- 1 Basic Modeling
- 2 Methodology
  - Satisfiability
  - Queens
  - **Traveling Salesperson**

# Traveling Salesperson



# Traveling Salesperson

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

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

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

# Traveling Salesperson

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

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

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

# Traveling Salesperson

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

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

```
edge(X,Y) :- cost(X,Y,_).
```

# Traveling Salesperson

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

# Traveling Salesperson

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

# Traveling Salesperson

```
1 { cycle(X,Y) : edge(X,Y) } 1 :- node(X).  
1 { cycle(X,Y) : edge(X,Y) } 1 :- node(Y).  
  
reached(Y) :- cycle(1,Y).  
reached(Y) :- cycle(X,Y), reached(X).  
  
:- node(Y), not reached(Y).
```

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

reached(Y) :- cycle(1,Y).
reached(Y) :- cycle(X,Y), reached(X).

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

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

# Language: Overview

- 3 Motivation
- 4 Core language
- 5 Extended language



# Outline

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- 5 Extended language

# Basic language extensions

- The expressiveness of a language can be enhanced by introducing new constructs
- To this end, we must address the following issues:
  - What is the **syntax** of the new language construct?
  - What is the **semantics** of the new language construct?
  - How to **implement** the new language construct?

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  - How to **implement** the new language construct?
- A way of providing semantics is to furnish a translation removing the new constructs, eg. classical negation
- This translation might also be used for implementing the language extension

# Outline

- 3 Motivation
- 4 Core language**
- 5 Extended language

# Outline

3

Motivation

4

Core language

- Integrity constraint
- Choice rule
- Cardinality rule
- Weight rule

5

Extended language

- Conditional literal
- Optimization statement

# Integrity constraint

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

$$\leftarrow a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n$$

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

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

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- **Example**  $\text{:- edge}(3, 7), \text{color}(3, \text{red}), \text{color}(7, \text{red}).$
- **Embedding** The above integrity constraint can be turned into the normal rule

$$x \leftarrow a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n, \text{not } x$$

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



# Outline

3 Motivation

4 Core language

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- **Choice rule**
- Cardinality rule
- Weight rule

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# Choice rule

- **Idea** Choices over subsets
- **Syntax** A **choice rule** is of the form

$$\{a_1, \dots, a_m\} \leftarrow a_{m+1}, \dots, a_n, \text{not } a_{n+1}, \dots, \text{not } a_o$$

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

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- **Example**  
`{ buy(pizza); buy(wine); buy(corn) } :- at(grocery) .`
- **Another Example**  $P = \{\{a\} \leftarrow b, b \leftarrow\}$  has two stable models:  $\{b\}$  and  $\{a, b\}$

# Embedding in normal rules

- A choice rule of form

$$\{a_1, \dots, a_m\} \leftarrow a_{m+1}, \dots, a_n, \text{not } a_{n+1}, \dots, \text{not } a_o$$

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

$$\begin{array}{l} b \leftarrow a_{m+1}, \dots, a_n, \text{not } a_{n+1}, \dots, \text{not } a_o \\ a_1 \leftarrow b, \text{not } a'_1 \quad \dots \quad a_m \leftarrow b, \text{not } a'_m \\ a'_1 \leftarrow \text{not } a_1 \quad \dots \quad a'_m \leftarrow \text{not } a_m \end{array}$$

by introducing new atoms  $b, a'_1, \dots, a'_m$ .

# Embedding in normal rules

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# Embedding in normal rules

- A choice rule of form

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$$\begin{array}{llll} b & \leftarrow & a_{m+1}, \dots, a_n, \text{not } a_{n+1}, \dots, \text{not } a_o & \\ a_1 & \leftarrow & b, \text{not } a'_1 & \dots \quad a_m \leftarrow b, \text{not } a'_m \\ a'_1 & \leftarrow & \text{not } a_1 & \dots \quad a'_m \leftarrow \text{not } a_m \end{array}$$

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Motivation

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Core language

- Integrity constraint
- Choice rule
- **Cardinality rule**
- Weight rule

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Extended language

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- Optimization statement

# Cardinality rule

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

$$a_0 \leftarrow l \{ a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n \}$$

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`pass(c42) :- 2 { pass(a1); pass(a2); pass(a3) }.`
- **Another Example**  $P = \{a \leftarrow 1\{b, c\}, b \leftarrow\}$  has stable model  $\{a, b\}$

# Embedding in normal rules

- Replace each cardinality rule

$$a_0 \leftarrow l \{ a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n \}$$

by  $a_0 \leftarrow \text{ctr}(1, l)$

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## ... and vice versa

- A normal rule

$$a_0 \leftarrow a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n$$

can be represented by the cardinality rule

$$a_0 \leftarrow n \{a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n\}$$

# Cardinality rules with upper bounds

- A rule of the form

$$a_0 \leftarrow l \{ a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n \} u \quad (1)$$

where  $0 \leq m \leq n$  and each  $a_i$  is an atom for  $1 \leq i \leq n$ ;  
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$$\begin{aligned} a_0 &\leftarrow b, \text{not } c \\ b &\leftarrow l \{ a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n \} \\ c &\leftarrow u+1 \{ a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n \} \end{aligned}$$

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- **Note** The single constraint in the body of the cardinality rule (1) is referred to as a **cardinality constraint**

# Cardinality constraints

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

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- In other words, if

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

# Cardinality constraints as heads

- A rule of the form

$$l \{a_1, \dots, a_m, \text{not } a_{m+1}, \dots, \text{not } a_n\} u \leftarrow a_{n+1}, \dots, a_o, \text{not } a_{o+1}, \dots, \text{not } a_p$$

where  $0 \leq m \leq n \leq o \leq p$  and each  $a_i$  is an atom for  $1 \leq i \leq p$ ;  
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- **Example** `1{ color(v42,red); color(v42,green); color(v42,blue) }1.`

# Outline

3

Motivation

4

Core language

- Integrity constraint
- Choice rule
- Cardinality rule
- **Weight rule**

5

Extended language

- Conditional literal
- Optimization statement

# Weight rule

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

$$a_0 \leftarrow l \{ w_1 : a_1, \dots, w_m : a_m, w_{m+1} : \text{not } a_{m+1}, \dots, w_n : \text{not } a_n \}$$

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

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

- A weighted literal  $w_i : \ell_i$  associates each literal  $\ell_i$  with a weight  $w_i$

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



# Weight constraints

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

$$l \{ w_1 : a_1, \dots, w_m : a_m, w_{m+1} : \text{not } a_{m+1}, \dots, w_n : \text{not } a_n \} u$$

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

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

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- **Example**

`10 { 4:course(db); 6:course(ai); 8:course(project); 3:course(xml) } 20`

# Outline

- 3 Motivation
- 4 Core language
- 5 Extended language**

# Outline

3

Motivation

4

Core language

- Integrity constraint
- Choice rule
- Cardinality rule
- Weight rule

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Extended language

- **Conditional literal**
- Optimization statement

# Conditional literals

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

$$\ell : \ell_1, \dots, \ell_n$$

where  $\ell$  and  $\ell_i$  are literals for  $0 \leq i \leq n$

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Motivation

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Extended language

- Conditional literal
- **Optimization statement**

# Optimization statement

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

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

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

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

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

```
#maximize { 250@1:hd(1), 500@1:hd(2), 750@1:hd(3), 1000@1:hd(4) }.  
#minimize { 30@2:hd(1), 40@2:hd(2), 60@2:hd(3), 80@2:hd(4) }.
```

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

# Language Extensions: Overview

- 6 Two kinds of negation
- 7 Disjunctive logic programs

# Outline

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# Motivation

- Classical versus default negation
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- Example

- $\text{cross} \leftarrow \neg \text{train}$

- $\text{cross} \leftarrow \text{not train}$



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- Given a program  $P$  over  $\mathcal{A}$ , classical negation is encoded by adding

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

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- A set  $X$  of atoms is a **stable model** of a program  $P$  over  $\mathcal{A} \cup \overline{\mathcal{A}}$ , if  $X$  is a stable model of  $P \cup P^\neg$

# An example

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- The stable models of  $P$  are given by the ones of  $P \cup P^\neg$ , viz  $\{a\}$

# Properties

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# Properties

- The only inconsistent stable “model” is  $X = \mathcal{A} \cup \overline{\mathcal{A}}$
- **Note** Strictly speaking, an inconsistent set like  $\mathcal{A} \cup \overline{\mathcal{A}}$  is not a model
- For a logic program  $P$  over  $\mathcal{A} \cup \overline{\mathcal{A}}$ , exactly one of the following two cases applies:
  - 1 All stable models of  $P$  are consistent or
  - 2  $X = \mathcal{A} \cup \overline{\mathcal{A}}$  is the only stable model of  $P$

# Train spotting

- $P_1 = \{cross \leftarrow not\ train\}$
- $P_2 = \{cross \leftarrow \neg train\}$
- $P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow\}$
- $P_4 = \{cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow\}$
- $P_5 = \{cross \leftarrow \neg train, \neg train \leftarrow not\ train\}$
- $P_6 = \{cross \leftarrow \neg train, \neg train \leftarrow not\ train, \neg cross \leftarrow\}$

# Train spotting

- $P_1 = \{cross \leftarrow not\ train\}$ 
  - stable model:  $\{cross\}$

# Train spotting

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# Train spotting

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# Train spotting

- $P_6 = \{cross \leftarrow \neg train, \neg train \leftarrow not\ train, \neg cross \leftarrow\}$ 
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# Default negation in rule heads

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

$$\begin{aligned}\tilde{P} = & \{r \in P \mid \text{head}(r) \neq \text{not } a\} \\ & \cup \{\leftarrow \text{body}(r) \cup \{\text{not } \tilde{a}\} \mid r \in P \text{ and } \text{head}(r) = \text{not } a\} \\ & \cup \{\tilde{a} \leftarrow \text{not } a \mid r \in P \text{ and } \text{head}(r) = \text{not } a\}\end{aligned}$$

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- A set  $X$  of atoms is a **stable model** of a program  $P$  (with default negation in rule heads) over  $\mathcal{A}$ ,  
if  $X = Y \cap \mathcal{A}$  for some stable model  $Y$  of  $\tilde{P}$  over  $\mathcal{A} \cup \tilde{\mathcal{A}}$

# Outline

- 6 Two kinds of negation
- 7 Disjunctive logic programs**

# Disjunctive logic programs

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

$$a_1 ; \dots ; a_m \leftarrow a_{m+1}, \dots, a_n, \text{not } a_{n+1}, \dots, \text{not } a_o$$

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

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- **Notation**

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

$$\text{body}(r) = \{a_{m+1}, \dots, a_n, \text{not } a_{n+1}, \dots, \text{not } a_o\}$$

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

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

$$\text{atom}(P) = \bigcup_{r \in P} (\text{head}(r) \cup \text{body}(r)^+ \cup \text{body}(r)^-)$$

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- A program is called **positive** if  $\text{body}(r)^- = \emptyset$  for all its rules

# Stable models

- Positive programs
  - A set  $X$  of atoms is **closed under** a positive program  $P$  iff for any  $r \in P$ ,  $head(r) \cap X \neq \emptyset$  whenever  $body(r)^+ \subseteq X$ 
    - $X$  corresponds to a model of  $P$  (seen as a formula)
  - The set of all  $\subseteq$ -minimal sets of atoms being closed under a positive program  $P$  is denoted by  $\min_{\subseteq}(P)$ 
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# A “positive” example

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- We have  $\min_{\subseteq}(P) = \{\{a, b\}, \{a, c\}\}$

# Graph coloring (reloaded)

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

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

```
color(X,r) ; color(X,b) ; color(X,g) :- node(X).
```

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

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

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

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

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

# More Examples

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# Some properties

- A disjunctive logic program may have zero, one, or multiple stable models
- If  $X$  is a stable model of a disjunctive logic program  $P$ , then  $X$  is a model of  $P$  (seen as a formula)
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- If  $X$  and  $Y$  are stable models of a disjunctive logic program  $P$ , then  $X \not\subseteq Y$
- If  $A \in X$  for some stable model  $X$  of a disjunctive logic program  $P$ , then there is a rule  $r \in P$  such that  $body(r)^+ \subseteq X$ ,  $body(r)^- \cap X = \emptyset$ , and  $head(r) \cap X = \{A\}$

# An example with variables

$$P = \left\{ \begin{array}{ll} a(1, 2) & \leftarrow \\ b(X) ; c(Y) & \leftarrow a(X, Y), \text{not } c(Y) \end{array} \right\}$$

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$$\text{ground}(P) = \left\{ \begin{array}{ll} a(1, 2) & \leftarrow \\ b(1) ; c(1) & \leftarrow a(1, 1), \text{not } c(1) \\ b(1) ; c(2) & \leftarrow a(1, 2), \text{not } c(2) \\ b(2) ; c(1) & \leftarrow a(2, 1), \text{not } c(1) \\ b(2) ; c(2) & \leftarrow a(2, 2), \text{not } c(2) \end{array} \right\}$$

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For every stable model  $X$  of  $P$ , we have

- $a(1, 2) \in X$  and
- $\{a(1, 1), a(2, 1), a(2, 2)\} \cap X = \emptyset$

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- We get  $\min_{\subseteq}(\mathit{ground}(P)^X) = \{ \{a(1,2), b(1)\}, \{a(1,2), c(2)\} \}$
- $X$  is a stable model of  $P$  because  $X \in \min_{\subseteq}(\mathit{ground}(P)^X)$

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- $X$  is no stable model of  $P$  because  $X \notin \min_{\subseteq}(\text{ground}(P)^X)$

# Default negation in rule heads

- Consider disjunctive rules of the form

$$a_1 ; \dots ; a_m ; \text{not } a_{m+1} ; \dots ; \text{not } a_n \leftarrow a_{n+1}, \dots, a_o, \text{not } a_{o+1}, \dots, \text{not } a_p$$

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



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

$$\begin{aligned} \tilde{P} = & \{ \text{head}(r)^+ \leftarrow \text{body}(r) \cup \{ \text{not } \tilde{a} \mid a \in \text{head}(r)^- \} \mid r \in P \} \\ & \cup \{ \tilde{a} \leftarrow \text{not } a \mid r \in P \text{ and } a \in \text{head}(r)^- \} \end{aligned}$$

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- A set  $X$  of atoms is a **stable model** of a disjunctive program  $P$  (with default negation in rule heads) over  $\mathcal{A}$ , if  $X = Y \cap \mathcal{A}$  for some stable model  $Y$  of  $\tilde{P}$  over  $\mathcal{A} \cup \tilde{\mathcal{A}}$

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- This induces the stable models  $\{a\}$  and  $\emptyset$  of  $P$

# Computational Aspects: Overview

## 8 Complexity

# Outline

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- For a propositional theory  $\Phi$ :
  - Deciding whether  $X$  is a stable model of  $\Phi$  is co-NP-complete
  - Deciding whether  $a$  is in a stable model of  $\Phi$  is  $\text{NP}^{\text{NP}}$ -complete

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- See also: <http://potassco.sourceforge.net>