

Artificial Intelligence, Computational Logic

PROBLEM SOLVING AND SEARCH IN ARTIFICIAL INTELLIGENCE

Lecture 6 ASP II * slides adapted from Torsten Schaub [Gebser et al.(2012)]

Sarah Gaggl

Dresden, 13th and 27th May 2016



Agenda



Introduction

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

Overview ASP II

Modeling



Language

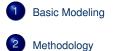


- 3 Motivation
 - Core language
- Extended language 5
- Language Extensions



- Two kinds of negation
 - Disjunctive logic programs
- Computational Aspects
 - Occupies Complexity

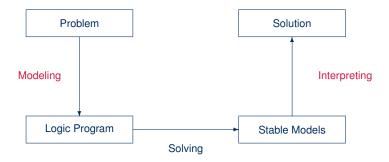
Modeling: Overview



Outline



Modeling and Interpreting



Modeling

• For solving a problem class C for a problem instance I, encode

the problem instance I as a set P_I of facts and the problem class C as a set P_C of rules such that the solutions to C for I can be (polynomially) extracted from the stable models of $P_I \cup P_C$

- *P*_I is (still) called problem instance
- Pc is often called the problem encoding
- An encoding P_C is uniform, if it can be used to solve all its problem instances
 That is, P_C encodes the solutions to C for any set P₁ of facts

Outline





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

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)

Outline





Traveling Salesperson

- 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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- Example: Consider formula

$$(a \lor \neg b) \land (\neg a \lor b)$$

Generator	•	Tester		Stable	e mo	dels
$\{a, b\}$	\leftarrow	\leftarrow	not a, b	X_1	=	$\{a,b\}$
		\leftarrow	a, not b	X_2	=	{}

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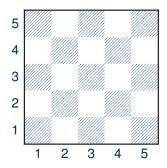
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Outline





The n-Queens Problem



- Place *n* queens on an *n* × *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
```

11000010	•	-
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) }.
```

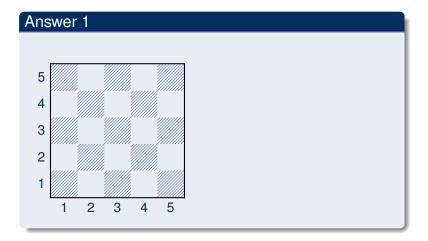
• Guess a solution candidate by placing some queens on the board

Placing some Queens

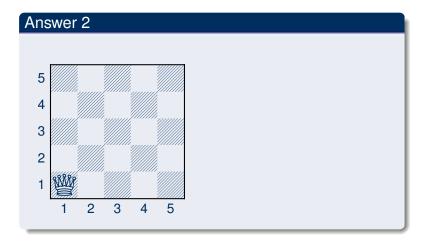
Running ...

```
$ 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+
...
```

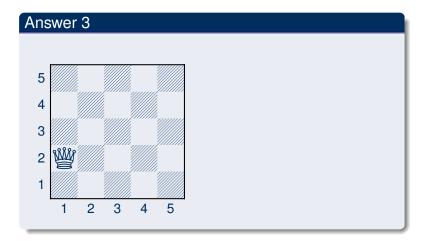
Placing some Queens: Answer 1



Placing some Queens: Answer 2



Placing some Queens: 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.
```

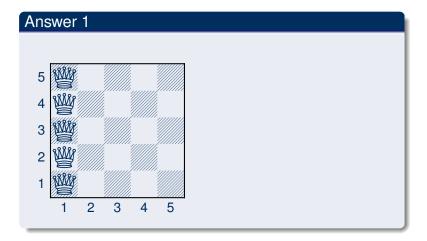
• Place exactly *n* queens on the board

Placing n Queens

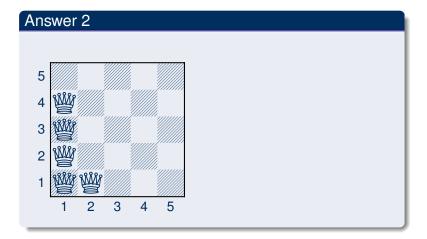
Running ...

```
$ 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



Placing *n* Queens: 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'.
```

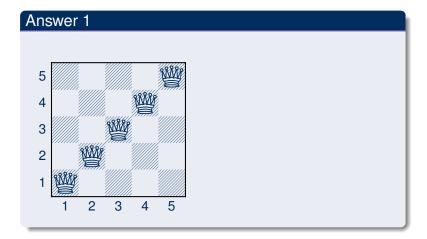
- Forbid horizontal attacks
- Forbid vertical attacks

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



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'.
```

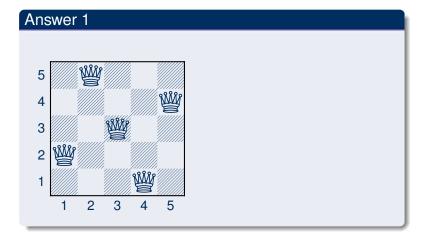
• Forbid diagonal attacks

Diagonal Attack

Running ...

Moders	•	1 +
Time	:	0.000
Prepare	:	0.000
Prepro.	:	0.000
Solving	:	0.000

Diagonal Attack: Answer 1



Optimizing

queens-opt.lp

```
 \begin{array}{l} 1 & \{ \mbox{ queen} (I,1,.n) \ \} \ 1 & :- \ I = 1..n. \\ 1 & \{ \mbox{ queen} (1..n,J) \ \} \ 1 & :- \ J = 1..n. \\ :- \ 2 & \{ \mbox{ queen} (D-J,J) \ \}, \ D = 2..2*n. \\ :- \ 2 & \{ \mbox{ queen} (D+J,J) \ \}, \ D = 1-n..n-1. \end{array}
```

- Encoding can be optimized
- Much faster to solve

And sometimes it rocks

```
$ clingo -c n=5000 gueens-opt-diag.lp -config=jumpy -g -stats=3
clingo version 4.1.0
Solving...
SATISFIABLE
            : 1+
Models
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)
         : 100129956 (1: 50059992/100090100 2: 39990/29856 3: 10000/10000)
Rules
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%)
TU Dresden, 13th and 27th May 2016 0.00 Max:
                                          0 Sum:
                                                     0 Ratio:
                                                                0.00%)
slide 38 of 200
```

Outline





```
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)).
```

```
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).
```

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

```
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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, _).
```

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

```
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).
```

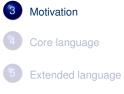
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reached(Y) :- cycle(1,Y).
reached(Y) :- cycle(X,Y), reached(X).
:- node(Y), not reached(Y).
```

```
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



Outline



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?

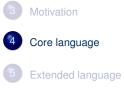
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 - What is the semantics of the new language construct?
 - How to implement the new language construct?
- A way of providing semantics is to furnish a translation removing the new constructs, eg. classical negation
- This translation might also be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

Outline



Outline

3





Core languageIntegrity constraint

- Choice rule
- Cardinality rule
- Weight rule



Extended language

- Conditional literal
- Optimization statement

Integrity constraint

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

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

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

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

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- Embedding The above integrity constraint can be turned into the normal rule

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where *x* is a new symbol, that is, $x \notin A$.

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• Another example $P = \{a \leftarrow not b, b \leftarrow not a\}$ versus $P' = P \cup \{\leftarrow a\}$ and $P'' = P \cup \{\leftarrow not a\}$

Outline

Motivation
 Core language

 Integrity constraint
 Choice rule
 Cardinality rule
 Weight rule

 Extended language

 Conditional literal
 Optimization statement

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

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

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

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- Example

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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 $\{a, b\}$

• A choice rule of form

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

can be translated into 2m + 1 normal rules

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

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

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$$a_1 \leftarrow b, \text{not } a'_1 \quad \dots \quad a_m \leftarrow b, \text{not } a'_m$$

$$a'_1 \leftarrow \text{not } a_1 \quad \dots \quad a'_m \leftarrow \text{not } a_m$$

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Outline

Core language
 Integrity constraint
 Choice rule
 Cardinality rule
 Weight rule

5

Extended language
Conditional literal

Optimization statement

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

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```
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```

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

• Replace each cardinality rule

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

by $a_0 \leftarrow ctr(1, l)$

where atom ctr(i,j) represents the fact that at least *j* of the literals having an equal or greater index than *i*, are in a stable model

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• The definition of ctr/2 is given for $0 \le k \le l$ by the rules

$$\begin{array}{rcl} ctr(i,k+1) & \leftarrow & ctr(i+1,k), a_i \\ ctr(i,k) & \leftarrow & ctr(i+1,k) \end{array} & \qquad \text{for } 1 \leq i \leq m \\ ctr(j,k+1) & \leftarrow & ctr(j+1,k), not a_j \\ ctr(j,k) & \leftarrow & ctr(j+1,k) \end{array} & \qquad \text{for } m+1 \leq j \leq n \\ ctr(n+1,0) & \leftarrow \end{array}$$

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$$\begin{array}{rcl} ctr(i,k+1) & \leftarrow & ctr(i+1,k), a_i \\ ctr(i,k) & \leftarrow & ctr(i+1,k) \end{array} & \qquad \text{for } 1 \leq i \leq m \\ ctr(j,k+1) & \leftarrow & ctr(j+1,k), not a_j \\ ctr(j,k) & \leftarrow & ctr(j+1,k) \end{array} & \qquad \text{for } m+1 \leq j \leq n \\ ctr(n+1,0) & \leftarrow \end{array}$$

• Replace each cardinality rule

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• Program $\{a \leftarrow, c \leftarrow 1 \{a, b\}\}$ has the stable model $\{a, c\}$

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- Translating the cardinality rule yields the rules

$$a \leftarrow c \leftarrow ctr(1, 1)$$

$$ctr(1, 2) \leftarrow ctr(2, 1), a$$

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$$ctr(2, 2) \leftarrow ctr(3, 1), b$$

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$$ctr(3, 0) \leftarrow$$

... and vice versa

• A normal rule

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

can be represented by the cardinality rule

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

Cardinality rules with upper bounds

• A rule of the form

$$a_0 \leftarrow l \{ a_1, \ldots, a_m, not \ a_{m+1}, \ldots, not \ a_n \} u \tag{1}$$

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

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$$a_0 \leftarrow b, not c$$

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

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

where b and c are new symbols

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

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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

 $l \{ a_1, \ldots, a_m, not \ a_{m+1}, \ldots, not \ a_n \} u$

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- Informal meaning A cardinality constraint is satisfied by a stable model *X*, if the number of its contained literals satisfied by *X* is between *l* and *u* (inclusive)
- In other words, if

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

Cardinality constraints as heads

• A rule of the form

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

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

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

stands for

$$\begin{cases} b & \leftarrow a_{n+1}, \dots, a_o, \text{ not } a_{o+1}, \dots, \text{ not } a_p \\ \{a_1, \dots, a_m\} & \leftarrow b \\ c & \leftarrow b \\ c & \leftarrow l \{a_1, \dots, a_m, \text{ not } a_{m+1}, \dots, \text{ not } a_n\} u \\ \leftarrow b, \text{ not } c \end{cases}$$

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 $\begin{array}{rcl}
b &\leftarrow & a_{n+1}, \dots, a_o, \text{ not } a_{o+1}, \dots, \text{ not } a_p \\
\{a_1, \dots, a_m\} &\leftarrow & b \\
c &\leftarrow & l \{a_1, \dots, a_m, \text{ not } a_{m+1}, \dots, \text{ not } a_n\} u \\
&\leftarrow & b, \text{ not } c
\end{array}$

where b and c are new symbols

• Example 1{ color(v42,red); color(v42,green); color(v42,blue) }1.

Outline

Motiva



Core language

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



Extended language

- Conditional literal
- Optimization statement

Weight rule

• Syntax A weight rule is the form

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

```
where 0 \le m \le n and each a_i is an atom;
 l and w_i are integers for 1 \le i \le n
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• A weighted literal $w_i : \ell_i$ associates each literal ℓ_i with a weight w_i

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```

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

• Syntax A weight constraint is of the form

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

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• Meaning A weight constraint is satisfied by a stable model X, if

$$l \le \left(\sum_{1 \le i \le m, a_i \in X} w_i + \sum_{m < i \le n, a_i \notin X} w_i\right) \le u$$

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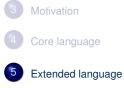
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- Example

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

Outline



Outline

Motivation



- Core languageIntegrity constraint
- Choice rule
- Cardinality rule
- Weight rule



Optimization statement

• Syntax A conditional literal is of the form

 $\ell:\ell_1,\ldots,\ell_n$

where ℓ and ℓ_i are literals for $0 \leq i \leq n$

 Informal meaning A conditional literal can be regarded as the list of elements in the set { l | l1,..., ln}

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- Example Given 'p(1..3). q(2).'

 $r(X) : p(X), notq(X) := r(X) : p(X), notq(X); 1 {r(X) : p(X), notq(X)}.$

is instantiated to

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

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Motivation



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- Choice rule
- Cardinality rule
- Weight rule



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

```
minimize { w_1 @ p_1 : \ell_1, \ldots, 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

• A maximize statement of the form

maximize {
$$w_1@p_1 : \ell_1, ..., w_n@p_n : \ell_n$$
 }

stands for *minimize* { $-w_1@p_1 : \ell_1, ..., -w_n@p_n : \ell_n$ }

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stands for *minimize* $\{ -w_1 @ p_1 : \ell_1, ..., -w_n @ p_n : \ell_n \}$

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

```
#maximize { 25001:hd(1), 50001:hd(2), 75001:hd(3), 100001:hd(4) }.
#minimize { 3002:hd(1), 4002:hd(2), 6002:hd(3), 8002:hd(4) }.
```

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

Language Extensions: Overview



Two kinds of negation



Disjunctive logic programs

Outline



Disjunctive logic programs

Motivation

- Classical versus default negation
 - Symbol ¬ and not

Motivation

• Classical versus default negation

- Symbol ¬ and not
- Idea
 - $\neg a \approx \neg a \in X$ • not $a \approx a \notin X$

Motivation

• Classical versus default negation

- Symbol ¬ and not
- Idea
 - $\neg a \approx \neg a \in X$
 - not $a \approx a \notin X$
- Example
 - $cross \leftarrow \neg train$
 - $cross \leftarrow not train$

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A set *X* of atoms is a stable model of a program *P* over *A* ∪ *A*, if *X* is a stable model of *P* ∪ *P*[¬]

An example

• The program

$$P = \{a \leftarrow not \ b, \ b \leftarrow not \ a\} \cup \{c \leftarrow b, \ \neg c \leftarrow b\}$$

An example

• The program

$$P = \{a \leftarrow not \ b, \ b \leftarrow not \ a\} \cup \{c \leftarrow b, \ \neg c \leftarrow b\}$$

induces

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

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• The stable models of *P* are given by the ones of *P* ∪ *P*[¬], viz {*a*}

Properties

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- The only inconsistent stable "model" is $X = A \cup \overline{A}$
- Note Strictly speaking, an inconsistent set like $A \cup \overline{A}$ is not a model
- For a logic program *P* over $A \cup \overline{A}$, exactly one of the following two cases applies:



All stable models of *P* are consistent or $X = A \cup \overline{A}$ is the only stable model of *P*

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

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

• $P_2 = \{cross \leftarrow \neg train\}$

- $P_2 = \{cross \leftarrow \neg train\}$
 - stable model: Ø

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$$P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow \}$$

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 - stable model: {*cross*, \neg *cross*, *train*, \neg *train*} inconsistent as $\mathcal{A} \cup \overline{\mathcal{A}}$

• $P_5 = \{cross \leftarrow \neg train, \neg train \leftarrow not train\}$

*P*₅ = {cross ← ¬train, ¬train ← not train}
stable model: {cross, ¬train}

• $P_6 = \{cross \leftarrow \neg train, \neg train \leftarrow not train, \neg cross \leftarrow \}$

P₆ = {cross ← ¬train, ¬train ← not train, ¬cross ←}
 no stable model

- $P_1 = \{cross \leftarrow not train\}$
 - stable model: {cross}
- $P_2 = \{cross \leftarrow \neg train\}$
 - − stable model: Ø

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$$P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow \}$$

- stable model: {cross, ¬train}
- $P_4 = \{ cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow \}$
 - stable model: {*cross*, \neg *cross*, *train*, \neg *train*} inconsistent as $\mathcal{A} \cup \overline{\mathcal{A}}$
- *P*₅ = {cross ← ¬train, ¬train ← not train}
 stable model: {cross, ¬train}
- $P_6 = \{ cross \leftarrow \neg train, \neg train \leftarrow not train, \neg cross \leftarrow \}$ - no stable model

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- Given an alphabet A of atoms, let $\widetilde{A} = \{\widetilde{a} \mid a \in A\}$ such that $A \cap \widetilde{A} = \emptyset$

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$$\widetilde{P} = \{r \in P \mid head(r) \neq not \ a\}$$
$$\cup \{\leftarrow body(r) \cup \{not \ \widetilde{a}\} \mid r \in P \text{ and } head(r) = not \ a\}$$
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A set X of atoms is a stable model of a program P (with default negation in rule heads) over A,
 if X = Y ∩ A for some stable model Y of P over A ∪ A

Outline





Disjunctive logic programs

Disjunctive logic programs

• A disjunctive rule, r, is of the form

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

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

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- A disjunctive logic program is a finite set of disjunctive rules
- Notation

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

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

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

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

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

$$body(P) = \{body(r) \mid r \in P\}$$

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A program is called positive if *body*(*r*)[−] = Ø 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 $\min_{\subseteq}(P)$
 - $\min_{\subseteq}(P)$ corresponds to the \subseteq -minimal models of *P* (ditto)

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

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

Stable models

- Positive programs
 - A set X of atoms is closed under a positive program P iff for any r ∈ P, head(r) ∩ X ≠ Ø whenever body(r)⁺ ⊆ X
 - X corresponds to a model of P (seen as a formula)
 - The set of all ⊆-minimal sets of atoms being closed under a positive program *P* is denoted by min_⊂(*P*)
 - $\min_{\subset}(P)$ corresponds to the \subseteq -minimal models of P (ditto)
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$$P^{X} = \{head(r) \leftarrow body(r)^{+} \mid r \in P \text{ and } body(r)^{-} \cap X = \emptyset\}$$

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

A "positive" example

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

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A "positive" example

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

- The sets $\{a, b\}$, $\{a, c\}$, and $\{a, b, c\}$ are closed under *P*
- 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)).
```

```
\operatorname{color}(X,r) ; \operatorname{color}(X,b) ; \operatorname{color}(X,g) :- \operatorname{node}(X).
```

```
:- edge(X,Y), color(X,C), color(Y,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)).
col(r). col(b). col(g).
color(X,C) : col(C) :- node(X).
:- edge(X,Y), color(X,C), color(Y,C).
```

•
$$P_1 = \{a ; b ; c \leftarrow\}$$

- $P_1 = \{a ; b ; c \leftarrow\}$
 - stable models $\{a\}$, $\{b\}$, and $\{c\}$

•
$$P_2 = \{a ; b ; c \leftarrow, \leftarrow a\}$$

•
$$P_2 = \{a ; b ; c \leftarrow, \leftarrow a\}$$

- stable models $\{b\}$ and $\{c\}$

•
$$P_3 = \{a ; b ; c \leftarrow, \leftarrow a, b \leftarrow c, c \leftarrow b\}$$

•
$$P_3 = \{a ; b ; c \leftarrow, \leftarrow a, b \leftarrow c, c \leftarrow b\}$$

- stable model $\{b, c\}$

• $P_4 = \{a : b \leftarrow c, b \leftarrow not a, not c, a : c \leftarrow not b\}$

- $P_4 = \{a : b \leftarrow c, b \leftarrow not a, not c, a : c \leftarrow not b\}$
 - stable models $\{a\}$ and $\{b\}$

- $P_1 = \{a ; b ; c \leftarrow\}$
 - stable models $\{a\}, \{b\}, \text{ and } \{c\}$
- *P*₂ = {*a* ; *b* ; *c* ← , ← *a*}
 stable models {*b*} and {*c*}
- $P_3 = \{a ; b ; c \leftarrow, \leftarrow a, b \leftarrow c, c \leftarrow b\}$
 - stable model $\{b, c\}$
- $P_4 = \{a : b \leftarrow c, b \leftarrow not a, not c, a : c \leftarrow not b\}$
 - stable models $\{a\}$ and $\{b\}$

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)
- If *X* and *Y* are stable models of a disjunctive logic program *P*, then *X* ⊄ *Y*

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)
- If *X* and *Y* are stable models of a disjunctive logic program *P*, then *X* ⊄ *Y*
- If A ∈ X for some stable model X of a disjunctive logic program P, then there is a rule r ∈ P such that body(r)⁺ ⊆ X, body(r)⁻ ∩ X = Ø, and head(r) ∩ X = {A}

$$P = \begin{cases} a(1,2) \leftarrow \\ b(X); c(Y) \leftarrow a(X,Y), not c(Y) \end{cases}$$

$$P = \begin{cases} a(1,2) \leftarrow \\ b(X); c(Y) \leftarrow a(X,Y), not c(Y) \end{cases}$$

$$ground(P) = \begin{cases} a(1,2) \leftarrow \\ b(1); c(1) \leftarrow a(1,1), not c(1) \\ b(1); c(2) \leftarrow a(1,2), not c(2) \\ b(2); c(1) \leftarrow a(2,1), not c(1) \\ b(2); c(2) \leftarrow a(2,2), not c(2) \end{cases}$$

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For every stable model X of P, we have

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

$$ground(P) = \begin{cases} a(1,2) \leftarrow \\ b(1);c(1) \leftarrow a(1,1), not c(1) \\ b(1);c(2) \leftarrow a(1,2), not c(2) \\ b(2);c(1) \leftarrow a(2,1), not c(1) \\ b(2);c(2) \leftarrow a(2,2), not c(2) \end{cases}$$

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• Consider $X = \{a(1,2), b(1)\}$

$$ground(P)^{X} = \begin{cases} a(1,2) \leftarrow \\ b(1);c(1) \leftarrow a(1,1) \\ b(1);c(2) \leftarrow a(1,2) \\ b(2);c(1) \leftarrow a(2,1) \\ b(2);c(2) \leftarrow a(2,2) \end{cases}$$

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- Consider $X = \{a(1,2), b(1)\}$
- We get $\min_{\subseteq}(ground(P)^X) = \{ \{a(1,2), b(1)\}, \{a(1,2), c(2)\} \}$

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

$$ground(P) = \begin{cases} a(1,2) \leftarrow \\ b(1); c(1) \leftarrow a(1,1), not c(1) \\ b(1); c(2) \leftarrow a(1,2), not c(2) \\ b(2); c(1) \leftarrow a(2,1), not c(1) \\ b(2); c(2) \leftarrow a(2,2), not c(2) \end{cases}$$

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$$X = \{a(1,2), c(2)\}$$

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- Consider $X = \{a(1,2), c(2)\}$
- We get $\min_{\subseteq}(ground(P)^X) = \{ \{a(1,2)\} \}$
- *X* is no stable model of *P* because $X \not\in \min_{\subset}(ground(P)^X)$

Default negation in rule heads

• Consider disjunctive rules of the form

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

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

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

• Given a program P over A, consider the program

$$\widetilde{P} = \{head(r)^+ \leftarrow body(r) \cup \{not \ \widetilde{a} \mid a \in head(r)^-\} \mid r \in P\} \\ \cup \{\widetilde{a} \leftarrow not \ a \mid r \in P \text{ and } a \in head(r)^-\}$$

Default negation in rule heads

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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 A, if X = Y ∩ A for some stable model Y of P̃ over A ∪ Ã

• The program

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• The program

$$P = \{a ; not \ a \leftarrow \}$$

yields

$$\widetilde{P} = \{a \leftarrow not \ \widetilde{a}\} \cup \{\widetilde{a} \leftarrow not \ a\}$$

- \widetilde{P} has two stable models, $\{a\}$ and $\{\widetilde{a}\}$
- This induces the stable models {*a*} and Ø of *P*

Computational Aspects: Overview



Outline



- For a positive normal logic program *P*:
 - Deciding whether X is the stable model of P is P-complete
 - Deciding whether *a* is in the stable model of *P* is P-complete

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- For a positive normal logic program *P*:
 - Deciding whether *X* is the stable model of *P* is P-complete
 - Deciding whether *a* is in the stable model of *P* is P-complete
- For a normal logic program *P*:
 - Deciding whether *X* is a stable model of *P* is P-complete
 - Deciding whether *a* is in a stable model of *P* is NP-complete
- For a normal logic program *P* with optimization statements:
 - Deciding whether X is an optimal stable model of P is co-NP-complete
 - Deciding whether *a* is in an optimal stable model of *P* is Δ_2^p -complete

- For a positive disjunctive logic program *P*:
 - Deciding whether *X* is a stable model of *P* is co-NP-complete
 - Deciding whether *a* is in a stable model of *P* is NP^{NP}-complete
- For a disjunctive logic program *P*:
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 - Deciding whether X is an optimal stable model of P is co-NP^{NP}-complete
 - Deciding whether *a* is in an optimal stable model of *P* is Δ_3^p -complete
- For a propositional theory Φ :
 - Deciding whether X is a stable model of Φ is co-NP-complete
 - Deciding whether a is in a stable model of Φ is NP^{NP}-complete

References

Martin Gebser, Benjamin Kaufmann Roland Kaminski, and Torsten Schaub. Answer Set Solving in Practice. Synthesis Lectures on Artificial Intelligence and Machine Learning. Morgan and Claypool Publishers, 2012. doi=10.2200/S00457ED1V01Y201211AIM019.

• See also: http://potassco.sourceforge.net