



PROBLEM SOLVING AND SEARCH IN ARTIFICIAL INTELLIGENCE

Lecture 6+7 ASP II * slides adapted from Torsten Schaub [Gebser et al.(2012)]

Sarah Gaggl

Dresden, 25th Nov, 3rd Dec 2019



Agenda

- 1 Introduction
- 2 Uninformed Search versus Informed Search (Best First Search, A* Search, Heuristics)
- 3 Local Search, Stochastic Hill Climbing, Simulated Annealing
- 4 Tabu Search
- 5 Answer-set Programming (ASP)
- 6 Constraint Satisfaction (CSP)
- 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

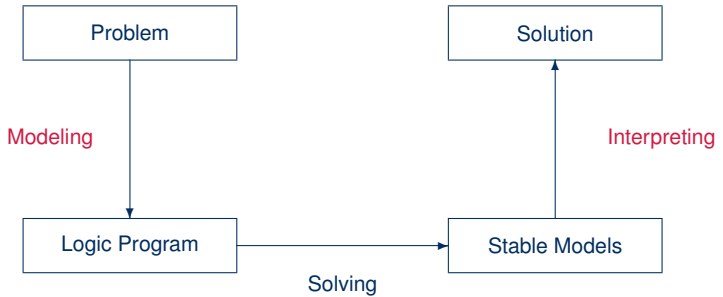
1 Basic Modeling

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

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

Satisfiability testing

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

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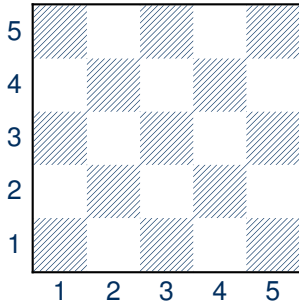
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- 1 Basic Modeling

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 - **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

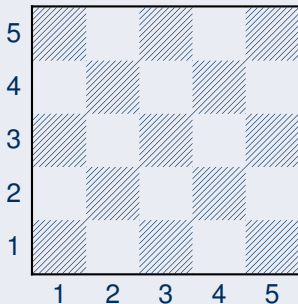
Running ...

```
$ 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+
...
```

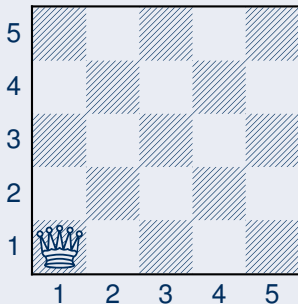
Placing some Queens: Answer 1

Answer 1



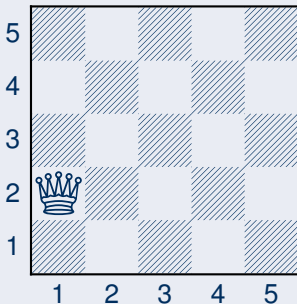
Placing some Queens: Answer 2

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

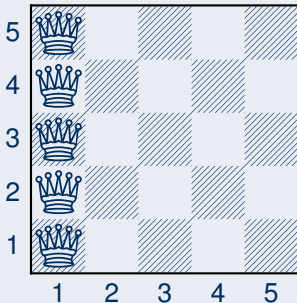
Placing n Queens

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)
...
```

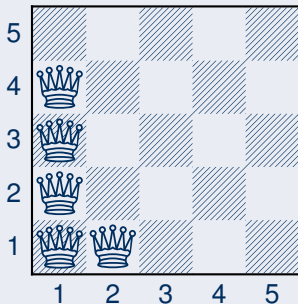
Placing n Queens: Answer 1

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

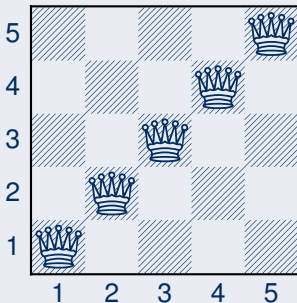
Horizontal and Vertical Attack

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

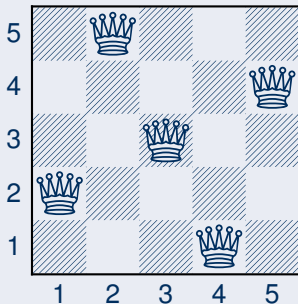
Running ...

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$ 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%)
```

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- 1 Basic Modeling
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 - 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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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

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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).`

Integrity constraint

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- **Example** $\leftarrow \text{edge}(3, 7), \text{color}(3, \text{red}), \text{color}(7, \text{red}) .$
- **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, $x \notin \mathcal{A}$.

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- **Choice rule**
- Cardinality 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 $1 \leq i \leq o$

Choice rule

- **Idea** Choices over subsets
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- **Example**
`{ buy(pizza); buy(wine); buy(corn) } :- at(grocery) .`

Choice rule

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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

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

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

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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

- Conditional literal
- Optimization statement

Cardinality rule

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

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

where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$;
 l is a non-negative integer.

Cardinality rule

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where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$;
 l is a non-negative integer.

- **Informal meaning** The head atom belongs to the stable model, if at least l elements of the body are included in the stable model
- **Note** l acts as a **lower bound** on the body

Cardinality rule

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

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

where $0 \leq m \leq n$ and each a_i is an atom for $1 \leq i \leq n$;
 l is a non-negative integer.

- **Informal meaning** The head atom belongs to the stable model, if at least l elements of the body are included in the stable model
- **Note** l acts as a **lower bound** on the body
- **Example**
`pass(c42) :- 2 { pass(a1); pass(a2); pass(a3) }.`

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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 $\{a, b\}$

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

where atom $\text{ctr}(i, j)$ 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 $\text{ctr}/2$ is given for $0 \leq k \leq l$ by the rules

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

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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 $1 \leq i \leq n$;
 l and u are non-negative integers

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stands for

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

where b and c are new symbols

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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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- **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

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

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

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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 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 $0 \leq i \leq n$

Weight constraints

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

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

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

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- **Note** The expansion of conditional literals is context dependent
- **Example** Given ' $p(1..3) . \quad q(2) .$ '

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

is instantiated to

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

Conditional literals

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

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

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 $\{\ell \mid \ell_1, \dots, \ell_n\}$
- **Note** The expansion of conditional literals is context dependent
- **Example** Given ' $p(1..3) . \quad q(2) .$ '

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

is instantiated to

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Outline

3

Motivation

4

Core language

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

5

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

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

where each ℓ_i is a literal; and w_i and p_i are integers for $1 \leq i \leq n$

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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

Optimization statement

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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

- 6 Two kinds of negation
- 7 Disjunctive logic programs

Motivation

- Classical versus default negation

– Symbol \neg and *not*

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

Motivation

- Classical versus default negation

- Symbol \neg and *not*

- Idea

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

- $\text{not } a \approx a \notin X$

- Example

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

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

Classical negation

- We consider logic programs in negation normal form
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Classical negation

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

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

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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

- The program

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

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induces

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

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

Properties

- The only inconsistent stable “model” is $X = \mathcal{A} \cup \overline{\mathcal{A}}$

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- **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

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 - stable model: $\{cross\}$

Train spotting

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Train spotting

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- $P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow\}$
 - stable model: $\{cross, \neg train\}$

Train spotting

- $P_4 = \{cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow\}$

Train spotting

- $P_4 = \{cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow\}$
 - stable model: $\{cross, \neg cross, train, \neg train\}$ inconsistent as $\mathcal{A} \cup \bar{\mathcal{A}}$

Train spotting

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

Train spotting

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 - stable model: $\{cross, \neg train\}$

Train spotting

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Train spotting

- $P_6 = \{cross \leftarrow \neg train, \neg train \leftarrow not\ train, \neg cross \leftarrow\}$
 - no stable model

Train spotting

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 - stable model: $\{cross\}$
- $P_2 = \{cross \leftarrow \neg train\}$
 - stable model: \emptyset
- $P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow\}$
 - stable model: $\{cross, \neg train\}$
- $P_4 = \{cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow\}$
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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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- A **disjunctive logic program** is a finite set of disjunctive rules
- **Notation**

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

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

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

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

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

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

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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 $\min_{\subseteq}(P)$
 - $\min_{\subseteq}(P)$ 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

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A “positive” example

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- 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)).
```

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

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

Graph coloring (reloaded)

```
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```

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```

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

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

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

More Examples

- $P_1 = \{a ; b ; c \leftarrow\}$

More Examples

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

More Examples

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

More Examples

- $P_2 = \{a ; b ; c \leftarrow , \leftarrow a\}$
 - stable models $\{b\}$ and $\{c\}$

More Examples

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

More Examples

- $P_3 = \{a ; b ; c \leftarrow , \leftarrow a , b \leftarrow c , c \leftarrow b\}$
 - stable model $\{b, c\}$

More Examples

- $P_4 = \{a ; b \leftarrow c, b \leftarrow \text{not } a, \text{not } c, a ; c \leftarrow \text{not } b\}$

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More Examples

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 - stable models $\{a\}$, $\{b\}$, and $\{c\}$
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 - stable model $\{b, c\}$
- $P_4 = \{a ; b \leftarrow c , b \leftarrow \text{not } a, \text{not } c , a ; c \leftarrow \text{not } b\}$
 - 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 $X \not\subseteq Y$

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- 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 $X \not\subseteq Y$
- 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

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

An example with variables

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

An example with variables

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For every stable model X of P , we have

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

An example with variables

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

An example with variables

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- Consider $X = \{a(1,2), b(1)\}$

An example with variables

$$\mathit{ground}(P)^X = \left\{ \begin{array}{llll} a(1,2) & \leftarrow & & \\ b(1) ; c(1) & \leftarrow & a(1,1) & \\ b(1) ; c(2) & \leftarrow & a(1,2) & \\ b(2) ; c(1) & \leftarrow & a(2,1) & \\ b(2) ; c(2) & \leftarrow & a(2,2) & \end{array} \right\}$$

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$$\mathit{ground}(P)^X = \left\{ \begin{array}{llll} a(1,2) & \leftarrow & & \\ b(1);c(1) & \leftarrow & a(1,1) & \\ b(1);c(2) & \leftarrow & a(1,2) & \\ b(2);c(1) & \leftarrow & a(2,1) & \\ b(2);c(2) & \leftarrow & a(2,2) & \end{array} \right\}$$

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

An example with variables

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

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- Consider $X = \{a(1,2), c(2)\}$
- We get $\min_{\subseteq}(\mathit{ground}(P)^X) = \{ \{a(1,2)\} \}$
- X is no stable model of P because $X \notin \min_{\subseteq}(\mathit{ground}(P)^X)$

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 $\mathcal{A} \cup \tilde{\mathcal{A}}$

An example

- The program

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

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- \tilde{P} has two stable models, $\{a\}$ and $\{\tilde{a}\}$

An example

- The program

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

yields

$$\tilde{P} = \{a \leftarrow \text{not } \tilde{a}\} \cup \{\tilde{a} \leftarrow \text{not } a\}$$

- \tilde{P} has two stable models, $\{a\}$ and $\{\tilde{a}\}$
- This induces the stable models $\{a\}$ and \emptyset of P

Computational Aspects: Overview

Complexity

Outline

8 Complexity

Complexity

Let a be an atom and X be a set of atoms

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

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

Complexity

Let a be an atom and X be a set of atoms

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

Complexity

Let a be an atom and X be a set of atoms

- For a positive disjunctive logic program P :
 - Deciding whether X is a stable model of P is co-NP-complete
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- For a disjunctive logic program P :
 - Deciding whether X is a stable model of P is co-NP-complete
 - Deciding whether a is in a stable model of P is NP^{NP} -complete
- For a disjunctive logic program P with optimization statements:
 - Deciding whether X is an optimal stable model of P is co- NP^{NP} -complete
 - Deciding whether a is in an optimal stable model of P is Δ_3^{P} -complete

Complexity

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- For a disjunctive logic program P with optimization statements:
 - Deciding whether X is an optimal stable model of P is co- NP^{NP} -complete
 - Deciding whether a is in an optimal stable model of P is 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



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