



# PROBLEM SOLVING AND SEARCH IN ARTIFICIAL INTELLIGENCE

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

Sarah Gaggl

Dresden

# 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 III

- Language
  - ⑤ Extended language
- Language Extensions
  - ⑥ Two kinds of negation
  - ⑦ Disjunctive logic programs
- Computational Aspects
  - ⑧ Complexity

# Language: Overview

## 1 Extended language

# Outline

## 1 Extended language

# Outline

- 1 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  $0 \leq i \leq n$

- **Informal meaning** A conditional literal can be regarded as the list of elements in the set  $\{\ell \mid \ell_1, \dots, \ell_n\}$

# Conditional literals

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

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

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

- **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



# Conditional literals

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

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

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

- **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

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

# Conditional literals

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

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

where  $\ell$  and  $\ell_i$  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

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

# Conditional literals

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

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

where  $\ell$  and  $\ell_i$  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) . q(2) .'

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

is instantiated to

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

# Conditional literals

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

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

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

- **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

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

# Conditional literals

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

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

where  $\ell$  and  $\ell_i$  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) . q(2) .'

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

is instantiated to

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

# Outline

- 1 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  $\ell_i$  is a literal; and  $w_i$  and  $p_i$  are integers for  $1 \leq i \leq n$

# 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  $\ell_i$  is a literal; and  $w_i$  and  $p_i$  are integers for  $1 \leq i \leq n$

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



# 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  $\ell_i$  is a literal; and  $w_i$  and  $p_i$  are integers for  $1 \leq i \leq n$

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

- A maximize statement of the form

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

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

# Optimization statement

- A maximize statement of the form

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

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

- **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

- 2 Two kinds of negation
- 3 Disjunctive logic programs

# Outline

- 2 Two kinds of negation
- 3 Disjunctive logic programs

# Motivation

- Classical versus default negation
  - Symbol  $\neg$  and *not*

# Motivation

- Classical versus default negation

- Symbol  $\neg$  and *not*

- Idea

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

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

# 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
  - That is, classical negation is applied to atoms only

# Classical negation

- We consider logic programs in negation normal form
  - That is, classical negation is applied to atoms only
- Given an alphabet  $\mathcal{A}$  of atoms, let  $\overline{\mathcal{A}} = \{\neg a \mid a \in \mathcal{A}\}$  such that  $\mathcal{A} \cap \overline{\mathcal{A}} = \emptyset$

# Classical negation

- We consider logic programs in negation normal form
  - That is, classical negation is applied to atoms only
- 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

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

# Classical negation

- 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

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

- 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

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

# An example

- The program

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

induces

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

# An example

- The program

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

induces

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

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

# Properties

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



# 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

# 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

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

# Train spotting

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

# Train spotting

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

# Train spotting

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

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

# Train spotting

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

# Train spotting

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

# Train spotting

- $P_1 = \{cross \leftarrow not\ train\}$ 
  - stable model:  $\{cross\}$
- $P_2 = \{cross \leftarrow \neg train\}$ 
  - stable model:  $\emptyset$
- $P_3 = \{cross \leftarrow \neg train, \neg train \leftarrow\}$ 
  - stable model:  $\{cross, \neg train\}$
- $P_4 = \{cross \leftarrow \neg train, \neg train \leftarrow, \neg cross \leftarrow\}$ 
  - stable model:  $\{cross, \neg cross, train, \neg train\}$  inconsistent as  $\mathcal{A} \cup \bar{\mathcal{A}}$
- $P_5 = \{cross \leftarrow \neg train, \neg train \leftarrow not\ train\}$ 
  - stable model:  $\{cross, \neg train\}$
- $P_6 = \{cross \leftarrow \neg train, \neg train \leftarrow not\ train, \neg cross \leftarrow\}$ 
  - no stable model

# Default negation in rule heads

- We consider logic programs with default negation in rule heads



# Default negation in rule heads

- We consider logic programs with default negation in rule heads
- Given an alphabet  $\mathcal{A}$  of atoms, let  $\tilde{\mathcal{A}} = \{\tilde{a} \mid a \in \mathcal{A}\}$  such that  $\mathcal{A} \cap \tilde{\mathcal{A}} = \emptyset$

# Default negation in rule heads

- We consider logic programs with default negation in rule heads
- Given an alphabet  $\mathcal{A}$  of atoms, let  $\tilde{\mathcal{A}} = \{\tilde{a} \mid a \in \mathcal{A}\}$  such that  $\mathcal{A} \cap \tilde{\mathcal{A}} = \emptyset$
- Given a program  $P$  over  $\mathcal{A}$ , consider the program

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

# Default negation in rule heads

- Given an alphabet  $\mathcal{A}$  of atoms, let  $\tilde{\mathcal{A}} = \{\tilde{a} \mid a \in \mathcal{A}\}$  such that  $\mathcal{A} \cap \tilde{\mathcal{A}} = \emptyset$
- Given a program  $P$  over  $\mathcal{A}$ , consider the program

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

- 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

- 2 Two kinds of negation
- 3 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  $0 \leq i \leq o$

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

# 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  $0 \leq i \leq o$

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

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

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

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

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

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

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

# 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  $0 \leq i \leq o$

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

$$\begin{aligned} \text{head}(r) &= \{a_1, \dots, a_m\} \\ \text{body}(r) &= \{a_{m+1}, \dots, a_n, \text{not } a_{n+1}, \dots, \text{not } a_o\} \\ \text{body}(r)^+ &= \{a_{m+1}, \dots, a_n\} \\ \text{body}(r)^- &= \{a_{n+1}, \dots, a_o\} \\ \text{atom}(P) &= \bigcup_{r \in P} (\text{head}(r) \cup \text{body}(r)^+ \cup \text{body}(r)^-) \\ \text{body}(P) &= \{\text{body}(r) \mid r \in P\} \end{aligned}$$

- 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)



# 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)
- **Disjunctive programs**
  - 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  $r \in P$ ,  $head(r) \cap X \neq \emptyset$  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)
- **Disjunctive programs**
  - The **reduct**,  $P^X$ , of a disjunctive program  $P$  relative to a set  $X$  of atoms is defined by

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

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

# A “positive” example

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

# A “positive” example

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

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

# A “positive” example

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

- 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)).
```

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

# 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

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

# More Examples

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

# More Examples

- $P_1 = \{a ; b ; c \leftarrow\}$ 
  - stable models  $\{a\}$ ,  $\{b\}$ , and  $\{c\}$
- $P_2 = \{a ; b ; c \leftarrow , \leftarrow a\}$ 
  - stable models  $\{b\}$  and  $\{c\}$
- $P_3 = \{a ; b ; c \leftarrow , \leftarrow a , b \leftarrow c , c \leftarrow b\}$ 
  - 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  $X \not\subseteq Y$

# 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  $X \not\subseteq Y$
- 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

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

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

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

# An example with variables

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

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

# An example with variables

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

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



# An example with variables

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

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

# An example with variables

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

- 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

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

# An example with variables

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

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

# An example with variables

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

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

# An example with variables

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

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

# An example with variables

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

- 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  $0 \leq i \leq p$



# 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  $0 \leq i \leq p$

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

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

# 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  $0 \leq i \leq p$

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

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

- 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

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

# An example

- The program

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

yields

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

# 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  $\{\tilde{a}\}$

# 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

## 4 Complexity

# Outline

## 4 Complexity



# Complexity

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

# 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

# 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

# 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  $\Delta_2^P$ -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
  - Deciding whether  $a$  is in a stable model of  $P$  is  $\text{NP}^{\text{NP}}$ -complete
- 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  $\text{NP}^{\text{NP}}$ -complete
- For a disjunctive logic program  $P$  with optimization statements:
  - Deciding whether  $X$  is an optimal stable model of  $P$  is co- $\text{NP}^{\text{NP}}$ -complete
  - Deciding whether  $a$  is in an optimal stable model of  $P$  is  $\Delta_3^{\text{P}}$ -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
  - Deciding whether  $a$  is in a stable model of  $P$  is  $\text{NP}^{\text{NP}}$ -complete
- 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  $\text{NP}^{\text{NP}}$ -complete
- For a disjunctive logic program  $P$  with optimization statements:
  - Deciding whether  $X$  is an optimal stable model of  $P$  is co- $\text{NP}^{\text{NP}}$ -complete
  - Deciding whether  $a$  is in an optimal stable model of  $P$  is  $\Delta_3^{\text{P}}$ -complete
- 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

# 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>