

Answer Set Programming

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Contact

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- Research associate in the AI group
- Interests: Knowledge Representation and Reasoning
 - ▶ Formalisation knowledge, belief, actions, sensing
 - ▶ Tractable reasoning for highly expressive languages

ASP at a Glance

- ASP = Answer Set Programming
 - ▶ ASP \neq Microsoft's Active Server Pages

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 - ▶ Unlike Prolog: no procedural control
 - ▶ Order has no impact on semantics

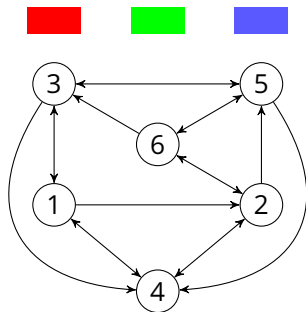
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- Declarative programming
 - ▶ Unlike Prolog: no procedural control
 - ▶ Order has no impact on semantics
- ASP programs compute *models*
 - ▶ Unlike Prolog: not query-oriented, no resolution
 - ▶ Unlike Prolog: not Turing-complete
 - ▶ Tool for problems in NP and NP^{NP} (common belief: $NP \subsetneq NP^{NP}$)

Example: Graph Colouring

Definition: graph colouring problem

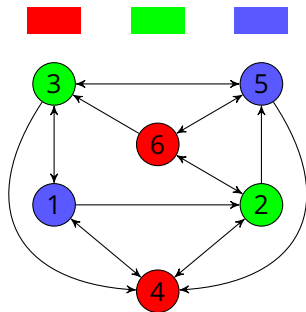
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Is there a mapping $m : V \rightarrow C$ with $m(x) \neq m(y)$ for all $(x,y) \in E$?



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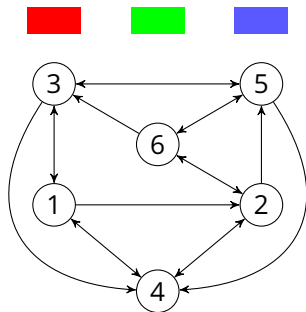
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- Graph Colouring is NP-complete
 - ▶ NP: guess solution, verify in polynomial time
 - ▶ NP-complete: among hardest in NP
- Many applications:
 - ▶ Mapping (neighbouring countries to different colors)
 - ▶ Scheduling (e.g., conflicting jobs to different time slots)
 - ▶ Allocation problems, Sudoku, ...

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$v(1)$ $v(6)$.

$e(1,2)$. $e(1,3)$. $e(1,4)$.

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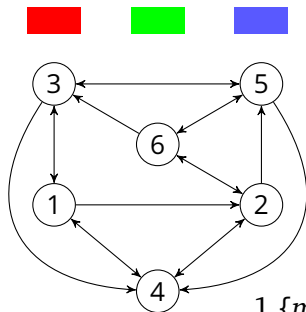
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$e(3, 1) . e(3, 4) . e(3, 5) .$

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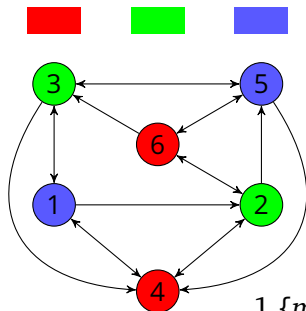
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:- $e(X, Y), m(X, C), m(Y, C)$. verify $m(X) \neq m(Y)$

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Applications of ASP

- Automated production configuration
 - Decision-support system for space shuttle
 - Bioinformatics (diagnosis, inconsistency detection)
 - General game playing
-
- Several implementations are available
 - For this lecture: **Clingo** www.potassco.org

Overview of the Lecture

- **Semantics of ASP programs**
- Extensions of ASP programs
- Handling of variables in ASP
- ASP as modelling language

Motivation

Consider the following logic program:

■ $a.$	$a.$
$c \leftarrow a, b.$	$c :- a, b.$
$d \leftarrow a, \text{not } b.$	$d :- a, \text{not } b.$

Motivation

Consider the following logic program:

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- What is the *semantics* of this logic program?

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Consider the following logic program:

- $a.$ a
 $c \leftarrow a, b.$ $a \wedge b \rightarrow c$
 $d \leftarrow a, \text{not } b.$ $a \wedge \neg b \rightarrow d$

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- ▶ Models: $M_1 =$

a	b	c	d
1	0	0	1

 $M_2 =$

a	b	c	d
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 ...

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- ▶ Models: $M_1 = \begin{array}{|c|c|c|c|} \hline a & b & c & d \\ \hline 1 & 0 & 0 & 1 \\ \hline \end{array}$ $M_2 = \begin{array}{|c|c|c|c|} \hline a & b & c & d \\ \hline 1 & 1 & 1 & 0 \\ \hline \end{array}$...
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- ▶ M_1 corresponds to Prolog, what is special about M_1 ?

- ▶ M_1 is a **stable model** a.k.a. **answer set**:

M_1 only satisfies *justified* propositions

ASP gives **semantics** to **logic programming**

Intuition

The motivating guidelines behind stable model semantics are:

- A stable model satisfies all the rules of a logic program
- The reasoner shall not believe anything they are not forced to believe — the **rationality principle**

The rationality principle is related to *non-monotonic reasoning*:

- Closed-world assumption
- Autoepistemic logic
- Default logic

For now: only ground programs, i.e., no variables

Definition: normal logic program (NLP)

A **normal logic program** P is a set of (normal) rules of the form

$$A \leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n.$$

where A, B_i, C_j are atomic propositions.

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For such a rule r , we define:

- $\text{Head}(r) = \{A\}$
- $\text{Body}(r) = \{B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n\}$

In code, r is written as $A \text{ :- } B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n.$

Semantics: Interpretation

Definition: partial interpretation, satisfaction

A **partial interpretation** S is a set of atomic propositions.

S **satisfies** $A \leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n$ iff
 $A \in S$ or $B_i \notin S$ for some i or $C_j \in S$ for some j .

In English:

- S satisfies rule iff S satisfies the head or falsifies the body
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$S = \{a, b, c\}$ satisfies $c \leftarrow a, b$ as well as $d \leftarrow a, \text{not } b$.

Semantics without Negation

Definition: stable model for programs without negation

For P without negated literals:

S is a **stable model** of P iff

S is a minimal set (w.r.t. \subseteq) that satisfies all $r \in P$.

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Theorem: unique-model property

If P is negation-free (i.e., contains no $(\text{not } C)$), then there is exactly one stable model, which can be computed in linear time.

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$$\underline{\text{Ex.:}} P_2 = \{a \leftarrow b. \quad b \leftarrow a.\}$$

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Ex.: $P_3 = \{a \leftarrow b. \quad b \leftarrow a. \quad a.\}$

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Definition: reduct

The **reduct** P^S of P relative to S is the least set such that
if $A \leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n \in P$ and $C_1, \dots, C_n \notin S$
then $A \leftarrow B_1, \dots, B_m \in P^S$.

In English: for each rule r from P ,

- if $(\text{not } C) \in \text{Body}(r)$ for some $C \in S$: drop it
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X

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if $A \leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n \in P$ and $C_1, \dots, C_n \notin S$
then $A \leftarrow B_1, \dots, B_m \in P^S$.

In English: for each rule r from P ,

- if $(\text{not } C) \in \text{Body}(r)$ for some $C \in S$: drop it
- else: remove all negated literals and add to P^S

Ex.: $P = \{a. \quad c \leftarrow a, b. \quad d \leftarrow a, \text{not } b.\}$

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Definition: stable model for programs with negation

For P with negated literals:

S is a **stable model** of P iff S is a stable model of P^S .

Semantics with Negation

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No stable model!

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then $A \leftarrow B_1, \dots, B_m \in P^S$.

Definition: stable model

If P contains no (not C):

S is a **stable model** of P iff

S is a minimal set (w.r.t. \subseteq) that satisfies all $r \in P$.

If P contains (not C):

S is a **stable model** of P iff S is a stable model of P^S .

Theorem: necessary satisfaction condition

If S is a stable model and $A \in S$, then S satisfies some $r \in P$ with
 $A \in \text{Head}(r)$.

Semantics – Examples

Ex.: $P = \{a \leftarrow a. \quad b \leftarrow \text{not } a.\}$

S

P^S

Stable model?

Ex.: $P = \{a \leftarrow \text{not } b. \quad b \leftarrow \text{not } c.\}$

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Stable model?

Example on paper

Entailment

Definition: entailment, cautious monotonicity

P **entails** a rule r iff every stable model of P satisfies r .

P is **cautiously monotonic** iff

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Good news: some classes of programs are cautiously monotonic.

Overview of the Lecture

- Semantics of ASP programs
- **Extensions of ASP programs**
- Handling of variables in ASP
- ASP as modelling language

Integrity Constraints

Definition: integrity constraint

An **integrity constraint** is a rule r of the form

$$\leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n$$

S **satisfies** r iff $B_i \notin S$ for some i or $C_j \in S$ for some j .

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Theorem: reduction to normal rules

Let P' be like P except that every integrity constraint

$$\leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n$$

is replaced with

$$\text{dummy} \leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n, \text{not } \text{dummy}$$

for some new atom *dummy*.

Then P and P' have the same stable models.

Proof on paper

Choice Rules

Definition: choice rule

A **choice rule** is a rule the form

$$\{A_1, \dots, A_k\} \leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n$$

which allows any subset of $\{A_1, \dots, A_k\}$ in a stable model.

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Further extensions:

- Conditional literals: $\{A : B\}$
Ex.: $\{m(v, C) : c(C)\}$ expands to $\{m(v, r), m(v, g), m(v, b)\}$
- Cardinality constraints: $\min \{A_1, \dots, A_k\} \max$
Ex.: $1 \{m(v, r), m(v, g), m(v, b)\} 1$

Negation in the Rule Head

Definition: rules with negated head

A rule with **negated head** is of the form

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and

$$\textit{dummy} \leftarrow \text{not } A$$

for some new atom \textit{dummy} .

Then P and P' have the same stable models (modulo dummy propositions).

Complexity

Theorem: complexity of NLPs without negations

Is S a stable model of a negation-free P ? – **Linear time**

Does a negation-free P have a stable model? – **Yes**

Theorem: complexity of NLPs with negations

Is S a stable model of P ? – **Linear time**

Does P have a stable model? – **NP-complete**

Note: integrity constraints, choice rules, conditional literals, cardinality constraints, negation in heads **preserve complexity**

Disjunctive Logic Programs

Definition: disjunctive rule

A **disjunctive rule** is of the form

$$A_1; \dots; A_k \leftarrow B_1, \dots, B_m, \text{not } C_1, \dots, \text{not } C_n$$

and means that A_1 or A_2 or ... or A_k is true if the body is true.

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Theorem: complexity of disjunctive logic programs

Is S a stable model of P ? – **co-NP-complete**

Does P have a stable model? – **NP^{NP}-complete**

Reason: P^S may have multiple minimal models!

We won't consider disjunctive logic problems any further

Overview of the Lecture

- Semantics of ASP programs
- Extensions of ASP programs
- **Handling of variables in ASP**
- ASP as modelling language

Programs with Variables

- Atomic propositions may now contain variables, e.g.,
 $p(X, Z) \leftarrow e(X, Y), p(Y, Z)$.

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- Herbrand universe
 - ▶ U contains all constants from P
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Programs with Variables

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 - ▶ U contains all constants from P
 - ▶ U contains all $f(t_1, \dots, t_k)$ from P if f is a k -ary function in P and U contains t_1, \dots, t_k
- ASP **grounds** variables with Herbrand universe
 - ▶ Unlike Prolog: instantiation instead of unification
 - ▶ Caution: the ground program may grow exponentially
 - ▶ Caution: function symbols make grounding Turing-complete
 - ▶ If P is finite and mentions only constants, grounding is finite

Programs with Variables

- $f(X) \leftarrow b(X)$, not $a(X)$.
 $a(X) \leftarrow p(X)$.
 $b(\text{sam})$.
 $b(\text{tweety.})$
 $p(\text{tweety.})$
- $f(\text{sam}) \leftarrow b(\text{sam})$, not $a(\text{sam})$.
 $f(\text{tweety}) \leftarrow b(\text{tweety})$, not $a(\text{tweety})$.
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ASP Modeling

$c(r). c(g). c(b).$
 $v(1). \dots v(6).$
 $e(1,2). e(1,3). e(1,4).$
 $e(2,4). e(2,5). e(2,6).$
 $e(3,1). e(3,4). e(3,5).$
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 $e(6,2). e(6,3). e(6,5).$

Typical ASP structure:

- Problem **instance**: a set of facts
- Problem **class**: a set of rules

- ▶ Generator rules: often choice rules $\{m(X,C) : c(C)\} 1 :- v(X).$
- ▶ Test rules: often integrity constraints $:- e(X,Y), m(X,C), m(Y,C).$

Ideal modeling is **uniform**: problem class encoding fits all instances

Semantically equivalent encodings may differ immensely in performance!

Example: Non-monotonic Reasoning

Tweety the penguin:

- Normal birds fly.
- Penguins are abnormal.
- Tweety is a bird. So Tweety flies.
- Tweety is a penguin. So Tweety doesn't fly.

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$P_1 = \{f(t) \leftarrow b(t), \text{not } a(t). \quad a(t) \leftarrow p(t). \quad b(t).\}$

$S_1 = \{b(t), f(t)\} \Rightarrow P_1^{S_1} = \{f(t) \leftarrow b(t). \quad a(t) \leftarrow p(t). \quad b(t).\}$ ✓

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Ex.: $U_2 = U_1 \cup \{p(t).\}$

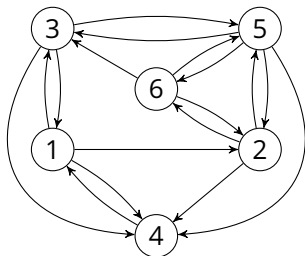
$P_2 = P_1 \cup \{p(t).\}$

$S_2 = \{a(t), b(t), p(t)\} \Rightarrow P_2^{S_2} = \{a(t) \leftarrow p(t). \quad b(t). \quad p(t).\} \quad \checkmark$

Example: Hamilton Cycle

Definition: Hamilton cycle problem

Input: graph with vertex set V and edges $E \subseteq V \times V$.
Is there a cycle that visits every vertex exactly once?

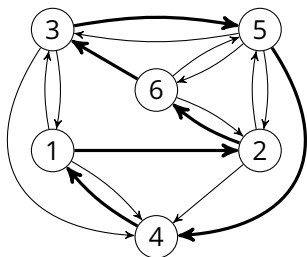


$$\begin{aligned} \{p(X, Y)\} &\leftarrow e(X, Y). \\ r(X) &\leftarrow p(1, X). \\ r(Y) &\leftarrow r(X), p(X, Y). \\ &\leftarrow 2 \{p(X, Y)\}, v(X). \\ &\leftarrow 2 \{p(X, Y)\}, v(Y). \\ &\leftarrow \text{not } r(X), v(X). \end{aligned}$$

Example: Hamilton Cycle

Definition: Hamilton cycle problem

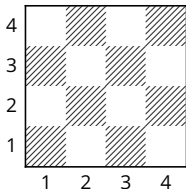
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$$\{p(X, Y)\} \leftarrow e(X, Y).$$
$$r(X) \leftarrow p(1, X).$$
$$r(Y) \leftarrow r(X), p(X, Y).$$
$$\leftarrow 2 \{p(X, Y)\}, v(X).$$
$$\leftarrow 2 \{p(X, Y)\}, v(Y).$$
$$\leftarrow \text{not } r(X), v(X).$$

Example: N -Queens

Definition: N -queens problem

Place N queens on a $N \times N$ chessboard so that they do not attack each other, i.e., share no row, column, or diagonal.

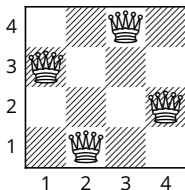


Program on paper

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Program on paper