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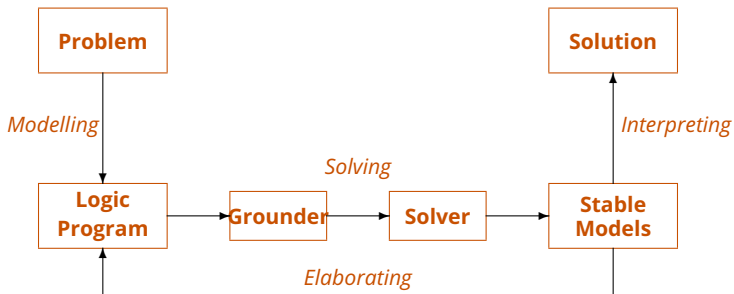
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ASP: Computation and Characterisation

Lecture 12, 12th Jan 2026 // Foundations of Logic Programming, WS 2025/26

Previously ...

- The language of normal logic programs can be extended by constructs:
 - **Integrity constraints** for eliminating unwanted solution candidates
 - **Choice rules** for choosing subsets of atoms
 - **Cardinality rules** for counting certain present/absent atoms
 - **Conditional literals** for improving conciseness
- All of them can be translated back into normal logic program rules.
- The modelling methodology of ASP is **generate and test**:
Generate solution candidates & Eliminate infeasible ones



Overview

Computation

- Consequence Operator

- Computation from First Principles

- Complexity

Axiomatic Characterisation

- Completion

- Tightness

- Loops and Loop Formulas

Computation

Consequence Operator

Recall

Let P be a positive program and X a set of atoms.
The **consequence operator** T_P assigns as follows:

$$T_P(X) = \{head(r) \mid r \in P \text{ and } body(r) \subseteq X\}$$

Iterated applications of T_P are written as T_P^j for $j \geq 0$, where

- $T_P^0(X) = X$ and
- $T_P^i(X) = T_P(T_P^{i-1}(X))$ for $i \geq 1$

For any positive program P , we have

- $Cn(P) = \bigcup_{i \geq 0} T_P^i(\emptyset)$
- $X \subseteq Y$ implies $T_P(X) \subseteq T_P(Y)$
- $Cn(P)$ is the \subseteq -least fixpoint of T_P

Approximating Stable Models

First Idea

Approximate a stable model X by two atom sets L and U such that $L \subseteq X \subseteq U$

- L and U constitute lower and upper bounds on X
- L and $(\mathcal{A} \setminus U)$ describe a three-valued model of the program

Observation

$$L \subseteq U \text{ implies } P^U \subseteq P^L \text{ implies } Cn(P^U) \subseteq Cn(P^L)$$

Properties

Let X be a stable model of normal logic program P .

- If $L \subseteq X$, then $X \subseteq Cn(P^L)$
- If $X \subseteq U$, then $Cn(P^U) \subseteq X$
- If $L \subseteq X \subseteq U$, then $L \cup Cn(P^U) \subseteq X \subseteq U \cap Cn(P^L)$

Approximating Stable Models

Second Idea

repeat

replace L **by** $L \cup Cn(P^U)$

replace U **by** $U \cap Cn(P^L)$

until L and U do not change anymore

Observations

- At each iteration step
 - L becomes larger (or equal)
 - U becomes smaller (or equal)
- $L \subseteq X \subseteq U$ is invariant for every stable model X of P
- If $L \not\subseteq U$, then P has no stable model
- If $L = U$, then L is a stable model of P

The Simplistic expand Algorithm

expand_p(L, U)

repeat

$L' \leftarrow L$

$U' \leftarrow U$

$L \leftarrow L' \cup \text{Cn}(P^{U'})$

$U \leftarrow U' \cap \text{Cn}(P^{L'})$

if $L \not\subseteq U$ **then return**

until $L = L'$ and $U = U'$

The algorithm:

- tightens the approximation on stable models
- is stable model preserving

An Example

Consider $P = \left\{ \begin{array}{l} a \leftarrow \\ b \leftarrow a, \sim c \\ d \leftarrow b, \sim e \\ e \leftarrow \sim d \end{array} \right\}$ over atoms $\mathcal{A} = \{a, b, c, d, e\}$.

The **expand** algorithm – started on the trivial pair (\emptyset, \mathcal{A}) – yields:

	L'	$Cn(P^{U'})$	L	U'	$Cn(P^{L'})$	U
1	\emptyset	$\{a\}$	$\{a\}$	$\{a, b, c, d, e\}$	$\{a, b, d, e\}$	$\{a, b, d, e\}$
2	$\{a\}$	$\{a, b\}$	$\{a, b\}$	$\{a, b, d, e\}$	$\{a, b, d, e\}$	$\{a, b, d, e\}$
3	$\{a, b\}$	$\{a, b\}$	$\{a, b\}$	$\{a, b, d, e\}$	$\{a, b, d, e\}$	$\{a, b, d, e\}$

Note

We have $\{a, b\} \subseteq X$ and $(\mathcal{A} \setminus \{a, b, d, e\}) \cap X = (\{c\} \cap X) = \emptyset$ for every stable model X of P .

Let us expand with $d \dots$

$$P = \left\{ \begin{array}{l} a \leftarrow \\ b \leftarrow a, \sim c \\ d \leftarrow b, \sim e \\ e \leftarrow \sim d \end{array} \right\}$$

	L'	$Cn(P^{U'})$	L	U'	$Cn(P^{L'})$	U
1	$\{d\}$	$\{a\}$	$\{a, d\}$	$\{a, b, c, d, e\}$	$\{a, b, d\}$	$\{a, b, d\}$
2	$\{a, d\}$	$\{a, b, d\}$	$\{a, b, d\}$	$\{a, b, d\}$	$\{a, b, d\}$	$\{a, b, d\}$
3	$\{a, b, d\}$	$\{a, b, d\}$	$\{a, b, d\}$	$\{a, b, d\}$	$\{a, b, d\}$	$\{a, b, d\}$

Note

$\{a, b, d\}$ is a stable model of P .

Let us expand with $\sim d$...

$$P = \left\{ \begin{array}{l} a \leftarrow \\ b \leftarrow a, \sim c \\ d \leftarrow b, \sim e \\ e \leftarrow \sim d \end{array} \right\}$$

	L'	$Cn(P^{U'})$	L	U'	$Cn(P^{L'})$	U
1	\emptyset	$\{a, e\}$	$\{a, e\}$	$\{a, b, c, e\}$	$\{a, b, d, e\}$	$\{a, b, e\}$
2	$\{a, e\}$	$\{a, b, e\}$	$\{a, b, e\}$	$\{a, b, e\}$	$\{a, b, e\}$	$\{a, b, e\}$
3	$\{a, b, e\}$	$\{a, b, e\}$	$\{a, b, e\}$	$\{a, b, e\}$	$\{a, b, e\}$	$\{a, b, e\}$

Note

$\{a, b, e\}$ is a stable model of P .

A Simplistic Solving Algorithm

$solve_P(L, U)$

$(L, U) \leftarrow expand_P(L, U)$

// propagation

if $L \not\subseteq U$ **then failure**

// failure

if $L = U$ **then output** L

// success

else choose $a \in U \setminus L$

// choice

$solve_P(L \cup \{a\}, U)$

$solve_P(L, U \setminus \{a\})$

A Simplistic Solving Algorithm

Close to the approach taken by the ASP solver `smodels`, inspired by the Davis-Putman-Logemann-Loveland (DPLL) procedure for SAT solving:

- Backtracking search building a binary search tree
- A node in the search tree corresponds to a three-valued interpretation
- The search space is pruned by
 - deriving deterministic consequences and detecting conflicts (**expand**)
 - making one choice at a time by appeal to a heuristic (**choose**)
- Heuristic choices are made on atoms

Quiz: Solving

```
solveP(L, U)
  (L, U) ← expandP(L, U)
  if  $L \not\subseteq U$  then failure
  if  $L = U$  then output L
  else choose  $a \in U \setminus L$ 
    solveP( $L \cup \{a\}$ , U)
    solveP(L,  $U \setminus \{a\}$ )
```

```
expandP(L, U)
  repeat
     $L' \leftarrow L; U' \leftarrow U$ 
     $L \leftarrow L' \cup \text{Cn}(P^{U'})$ 
     $U \leftarrow U' \cap \text{Cn}(P^{L'})$ 
    if  $L \not\subseteq U$  then return
  until  $L = L'$  and  $U = U'$ 
```

Quiz

...

Complexity

Problem: Stable-model-existence

Given: A propositional normal logic program P .

Question: Does P have a stable model?

Theorem

Stable-model-existence is NP-complete.

Proof.

- in NP: Given a candidate X , we can compute the reduct P^X and then $Cn(P^X)$, then check $X = Cn(P^X)$, all in deterministic polynomial time.
- NP-hard: We reduce from SAT. Let $\Phi = \varphi_1 \wedge \dots \wedge \varphi_m$ be a CNF over A . Set

$$P_\Phi := \{\{a\} \leftarrow \mid a \in A\} \cup \{\leftarrow \overline{\varphi_i} \mid 1 \leq i \leq m\}$$

where $\overline{\ell_1 \vee \dots \vee \ell_k} := \overline{\ell_1}, \dots, \overline{\ell_k}$ with $\neg \overline{a} := a$ and $\overline{a} := \sim a$.



Axiomatic Characterisation

Motivation

- There exist sophisticated algorithms and efficient implementations for SATisfiability testing in propositional logic
- Can we harness these systems for answer set programming?

Question

Is there a propositional formula/theory $F(P)$ such that the models of $F(P)$ correspond one-to-one to the **stable** models of P ?

Recall

- For every normal program P , there is a propositional theory $\text{comp}(P)$ such that its models correspond one-to-one to the **supported** models of P .
- Every **stable** model is a **supported** model, but not vice versa.

\rightsquigarrow Can we add a second theory $T(P)$ such that the models of $\text{comp}(P) \cup T(P)$ correspond one-to-one to the **stable** models of P ?

Program Completion: A Closer Look

The theory $comp(P)$ is logically equivalent to $\overleftarrow{comp}(P) \cup \overrightarrow{comp}(P)$, where

$$\overleftarrow{comp}(P) = \left\{ a \leftarrow \bigvee_{B \in body_P(a)} BF(B) \mid a \in atom(P) \right\}$$

$$\overrightarrow{comp}(P) = \left\{ a \rightarrow \bigvee_{B \in body_P(a)} BF(B) \mid a \in atom(P) \right\}$$

$$body_P(a) = \{ body(r) \mid r \in P \text{ and } head(r) = a \}$$

$$BF(body(r)) = \bigwedge_{a \in body(r)^+} a \wedge \bigwedge_{a \in body(r)^-} \neg a$$

- $\overleftarrow{comp}(P)$ characterises the classical models of P .
- $\overrightarrow{comp}(P)$ characterises that all **true** atoms must be **supported**.
- \rightsquigarrow How to axiomatise that all **true** atoms must be **well-supported**?

Stable vs. Supported Models: An Example

Example

$$P = \left\{ \begin{array}{lll} a \leftarrow & c \leftarrow a, \sim d & e \leftarrow b, \sim f \\ b \leftarrow \sim a & d \leftarrow \sim c, \sim e & e \leftarrow e \end{array} \right\}$$

- P has 21 models, including $\{a, c\}$, $\{a, d\}$, but also $\{a, b, c, d, e, f\}$.
- P has 3 supported models, namely $\{a, c\}$, $\{a, d\}$, and $\{a, c, e\}$.
- P has 2 stable models, namely $\{a, c\}$ and $\{a, d\}$.
- The model $\{a, c, e\}$ is not well-supported (stable) because e supports itself.

Observation

Atoms in a strictly positive cycle (not being “supported from outside the cycle”) cannot be “derived” from a program in a finite number of steps.

Positive Atom Dependency Graph

Definition

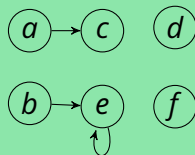
The **positive atom dependency graph** $G(P)$ of a logic program P is given by

$$(atom(P), \{(a, b) \mid r \in P, a \in body(r)^+, head(r) = b\})$$

A logic program P is called **tight** $:\Leftrightarrow G(P)$ is acyclic.

Example

- $P = \left\{ \begin{array}{lll} a \leftarrow & c \leftarrow a, \sim d & e \leftarrow b, \sim f \\ b \leftarrow \sim a & d \leftarrow \sim c, \sim e & e \leftarrow e \end{array} \right\}$
- $G(P) = (\{a, b, c, d, e\}, \{(a, c), (b, e), (e, e)\})$
- P has supported models: $\{a, c\}$, $\{a, d\}$, and $\{a, c, e\}$
- P has stable models: $\{a, c\}$ and $\{a, d\}$



Theorem (Fages)

For tight normal logic programs, stable and supported models coincide.

Motivation

Question

Is there a propositional formula $F(P)$ such that the models of $F(P)$ correspond to the stable models of P ?

Observation

Starting from the completion of a program, the problem boils down to eliminating the circular support of atoms holding in the supported models.

Idea

Add formulas prohibiting **circular support** of sets of atoms.

Circular support between atoms a and b is possible if a has a path to b and b has a path to a in the program's positive atom dependency graph.

Loops

Definition

Let P be a normal logic program with positive atom dependency graph $G(P) = (atom(P), E)$.

- A non-empty set $L \subseteq atom(P)$ is a **loop** of P
: \iff it induces a non-trivial strongly connected subgraph of $G(P)$.
- We denote the set of all loops of P by $loops(P)$.

That is, each pair of atoms in a loop L is connected by a path of non-zero length in $(L, E \cap (L \times L))$.

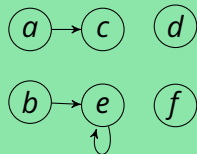
Observation

A program P is tight iff $loops(P) = \emptyset$.

Loops: Examples (1)

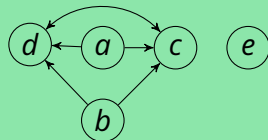
Example

- $P = \left\{ \begin{array}{lll} a \leftarrow & c \leftarrow a, \sim d & e \leftarrow b, \sim f \\ b \leftarrow \sim a & d \leftarrow \sim c, \sim e & e \leftarrow e \end{array} \right\}$
- $loops(P) = \{\{e\}\}$



Example

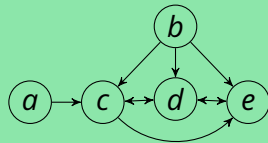
- $P = \left\{ \begin{array}{llll} a \leftarrow \sim b & c \leftarrow a, b & d \leftarrow a & e \leftarrow \sim a, \sim b \\ b \leftarrow \sim a & c \leftarrow d & d \leftarrow b, c & \end{array} \right\}$
- $loops(P) = \{\{c, d\}\}$



Loops: Examples (2)

Example

- $P = \left\{ \begin{array}{llll} a \leftarrow \sim b & c \leftarrow a & d \leftarrow b, c & e \leftarrow b, \sim a \\ b \leftarrow \sim a & c \leftarrow b, d & d \leftarrow e & e \leftarrow c, d \end{array} \right\}$
- $loops(P) = \{\{c, d\}, \{d, e\}, \{c, d, e\}\}$



Loop Formulas

Definition

Let P be a normal logic program.

- For $L \subseteq \text{atom}(P)$, define the **external supports** of L for P as

$$ES_P(L) := \{r \in P \mid \text{head}(r) \in L \text{ and } \text{body}(r)^+ \cap L = \emptyset\}$$

- Define the **external bodies** of L in P as $EB_P(L) := \text{body}(ES_P(L))$.
- The (disjunctive) **loop formula** of L for P is

$$LF_P(L) := (\bigvee_{a \in L} a) \rightarrow \left(\bigvee_{B \in EB_P(L)} BF(B) \right) \equiv \left(\bigwedge_{B \in EB_P(L)} \neg BF(B) \right) \rightarrow (\bigwedge_{a \in L} \neg a)$$

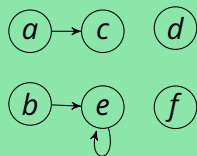
- Define $LF(P) := \{LF_P(L) \mid L \in \text{loops}(P)\}$.

The loop formula of L enforces all atoms in L to be *false* whenever L is not externally supported.

Loop Formulas: Examples (1)

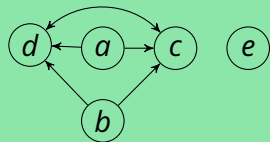
Example

- $P = \left\{ \begin{array}{lll} a \leftarrow & c \leftarrow a, \sim d & e \leftarrow b, \sim f \\ b \leftarrow \sim a & d \leftarrow \sim c, \sim e & e \leftarrow e \end{array} \right\}$
- $loops(P) = \{\{e\}\}$
- $LF(P) = \{e \rightarrow b \wedge \neg f\}$



Example

- $P = \left\{ \begin{array}{llll} a \leftarrow \sim b & c \leftarrow a, b & d \leftarrow a & e \leftarrow \sim a, \sim b \\ b \leftarrow \sim a & c \leftarrow d & d \leftarrow b, c & \end{array} \right\}$
- $loops(P) = \{\{c, d\}\}$
- $LF(P) = \{c \vee d \rightarrow (a \wedge b) \vee a\}$



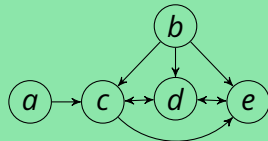
Loops: Examples (2)

Example

- $P = \left\{ \begin{array}{llll} a \leftarrow \sim b & c \leftarrow a & d \leftarrow b, c & e \leftarrow b, \sim a \\ b \leftarrow \sim a & c \leftarrow b, d & d \leftarrow e & e \leftarrow c, d \end{array} \right\}$

- $loops(P) = \{\{c, d\}, \{d, e\}, \{c, d, e\}\}$

- $LF(P) = \left\{ \begin{array}{l} c \vee d \rightarrow a \vee e \\ d \vee e \rightarrow (b \wedge c) \vee (b \wedge \neg a) \\ c \vee d \vee e \rightarrow a \vee (b \wedge \neg a) \end{array} \right\}$



Lin-Zhao Theorem and Properties

Theorem (Lin and Zhao, 2004)

Let P be a normal logic program and $X \subseteq \text{atom}(P)$. Then:

X is a stable model of P iff $X \models \text{comp}(P) \cup \text{LF}(P)$.

Properties of Loop Formulas

Let X be a supported model of normal LP P . Then, X is a stable model of P iff

- $X \models \{\text{LF}_P(U) \mid U \subseteq \text{atom}(P)\}$;
 - $X \models \{\text{LF}_P(U) \mid U \subseteq X\}$;
 - $X \models \{\text{LF}_P(L) \mid L \in \text{loops}(P)\}$, that is, $X \models \text{LF}(P)$;
 - $X \models \{\text{LF}_P(L) \mid L \in \text{loops}(P) \text{ and } L \subseteq X\}$.
-
- If supported X is not stable for P , there is a loop $L \subseteq X \setminus \text{Cn}(P^X)$ with $X \not\models \text{LF}_P(L)$.
 - There might be exponentially many loop formulas.
 - Blowup seems to be unavoidable in general [Lifschitz and Razborov, 2006].

Lin-Zhao Algorithm

Input: A ground normal logic program P .

Output: A stable model of P , if one exists, otherwise “unsatisfiable”.

Uses: A SAT solver for determining satisfiability of propositional theories.

ASSAT

```
 $T := \text{CNF}(\text{comp}(P))$  // obtain conjunctive normal form of comp(P)  
repeat  
   $M := \text{sat}(T)$  // uses SAT solver  
  if  $M = \text{“unsat”}$  return “unsatisfiable”  
   $M^- := M \setminus \text{Cn}(P^M)$  // obtain unjustified true atoms  
  if  $M^- = \emptyset$  return  $M$  // stable model found  
   $K := \text{mloops}(M^-, P)$  // obtain  $\subseteq$ -maximal loops  $L \subseteq M^-$   
  for each  $L \in K$ :  $T := T \cup \{LF_P(L)\}$ 
```

Conclusion

Summary

- The stable models of P can be approximated using the operator T_P :

$$(L, U) \rightsquigarrow (L \cup \bigcup_{i \geq 0} T_{PU}^i(\emptyset), U \cap \bigcup_{i \geq 0} T_{PL}^i(\emptyset))$$

- Solving may use non-deterministic choice, propagation, and backtracking.
- Stable-model-existence is NP-complete.
- Supported non-stable models are caused by loops in the program.
- A **loop** is a non-empty set of atoms that mutually depend on each other.
- The **loop formulas** $LF(P)$ of P enforce that every support is well-founded.
- The stable models of P can be characterised by $comp(P) \cup LF(P)$.

Suggested action points:

- Try the algorithm on Slide 12 for some example programs.

Course Summary

- LPs are a declarative language for knowledge representation and reasoning.
- PROLOG-based logic programming focuses on theorem proving.
- PROLOG is also a programming language (via non-logical side effects).
- For definite LPs, SLD resolution is a sound and complete proof theory.
- For normal LPs, SLDNF resolution is sound and (sometimes) complete.
- Stable models are recognised as the “standard” semantics for normal LPs.
- ASP-based logic programming focuses on model generation.
- ASP is a modelling language for (combinatorial) problem solving.
- Its modelling methodology is based on the generate-and-test paradigm.
- ASP solvers can make use of technology from propositional satisfiability.