
On the Existence of Answer Sets in Normal Extended Logic Programs

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Answer Set Programming: ASP

In ASP a problem is represented by an **extended logic program (ELP)** whose declarative meaning is given by the **answer set semantics**.

Program: $r \leftarrow p,$
 $\neg r \leftarrow q,$
 $p \leftarrow \text{not } q,$
 $q \leftarrow \text{not } p,$
Answer Sets: $\{ p, r \}, \{ q, r \}$

The problem of ASP

- ⌘ Answer set semantics is **FRAGILE!** Relatively **small pieces** of information may cause a **total absence** of answer sets.
 - ⌘ A notorious example is the existence of a “**negative loop**” such as $p \leftarrow \text{not } p$. A program including such **anomalous information** is **unusual** anyway.
 - ⌘ **More serious problem** is: programs that are seemingly natural and do not include any such anomalous information often **fail to have** answer sets.
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Example: Married John

Consider the following information:

- (1) John wears something that looks like a wedding ring.
 - (2) John parties with his friends until late.
 - (3) Someone wearing a wedding ring is usually married.
 - (4) A party-animal is usually a bachelor.
 - (5) A married person, by definition, has a spouse.
 - (6) A bachelor, by definition, does not have a spouse.
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It is naturally represented by the program.

$r \leftarrow,$ $p \leftarrow,$
 $m \leftarrow r, \text{ not } \neg m,$ $b \leftarrow p, \text{ not } \neg b,$
 $hs \leftarrow m,$ $\neg hs \leftarrow b.$

(r :ring, p :party, m :married, b :bachelor, hs : has spouse)

! This program has no answer set, even though it contains no negative loops.

Justification: $S = \{ r, p, m, hs \}$ or $\{ r, p, b, \neg hs \}$ produces the reduct

$r \leftarrow,$ $p \leftarrow,$ $m \leftarrow r,$ $b \leftarrow p,$ $hs \leftarrow m,$ $\neg hs \leftarrow b,$

which has the minimal closed set Lit. And $S = \text{Lit}$ produces the reduct

$r \leftarrow,$ $p \leftarrow,$ $hs \leftarrow m,$ $\neg hs \leftarrow b,$

which has the minimal closed set $\{ r, p \}$.

The problem does not arise if we encode the same knowledge in Reiter's default logic:

$r, \quad p, \quad m \supset hs, \quad b \supset \neg hs,$

$r : m / m, \quad p : b / b.$

The above default theory has two extensions:

$\{ r, p, b, \neg hs, \neg m \}$ and $\{ r, p, m, hs, \neg b \}.$

* This is a normal default theory that always has an extension.

Why such difference happens?

- ⌘ In extended LPs, every rule is interpreted as a one way “**inference rule**”.
 - ⌘ A rule does not have a **contrapositive meaning** even if it is a definite rule without default negation.
 - ⌘ In default theories, definite information is represented by **first-order formulas** having contrapositive meaning.
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To bridge the gap between ELPs and default theories, [Gelfond and Lifschitz] characterize the program in terms of non-normal default theories:

$$\begin{array}{l} r, \quad p, \quad m: / hs, \quad b: / \neg hs, \\ r : m / m, \quad p : b / b, \end{array}$$

which has no extension.

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- ⌘ The fact that ELPs with **normal default rules** may not have any answer set is **discouraging**. Because normal default theories are “a very large and natural class of default theories” [Reiter].
 - ⌘ A knowledge engineer can encode problems like “Married-John” in a normal default theory in a straightforward manner, while he/she may **fail to encode the same problem in a logic program**.
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Questions

- ⌘ How can one provide a **natural meaning** of an **ELP with normal default rules**?
- ⌘ Is there any condition to **guarantee the existence of answer sets** of ELPs with **normal default rules**?

Our solution:

- ⌘ We introduce the class of **normal ELPs** whose default rules are in the form of **normal default rules**.
 - ⌘ We provide **syntactic conditions to guarantee the existence of answer sets in normal ELPs**.
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Terms and Notations

⌘ A **strict rule** is of the form:

$$c \leftarrow a_1, \dots, a_n \quad (a_i: \text{literal}, c: (\text{nonempty}) \text{ literal})$$

⌘ A **defeasible rule** is of the form:

$$c \leftarrow a_1, \dots, a_n, \text{not } b_1, \dots, \text{not } b_m \quad (b_j: \text{literal}, m \neq 0)$$

A defeasible rule is **normal** if it is of the form:

$$c \leftarrow a_1, \dots, a_n, \text{not } \neg c$$

⌘ A **rule** is either strict or defeasible.

⌘ A **program** P is a finite set $\text{strict}(P) \cup \text{defeasible}(P)$ where $\text{strict}(P)$ (resp. $\text{defeasible}(P)$) is the set of all strict (resp. defeasible) rules in P .

⌘ We handle **finite propositional programs**.

Let s_1, s_2, s_3 be strict rules.

⌘ s_2 is a **transpositive** version of s_1 iff

$$s_1 = c \leftarrow a_1, \dots, a_n \quad \text{and}$$

$$s_2 = \neg a_i \leftarrow a_1, \dots, a_{i-1}, \neg c, a_{i+1}, a_n \quad \text{for some } 1 \leq i \leq n$$

⌘ s_3 is a **transitive** version of s_1 and s_2 iff

$$s_1 = c \leftarrow a_1, \dots, a_n$$

$$s_2 = a_i \leftarrow b_1, \dots, b_m \quad \text{for some } 1 \leq i \leq n, \text{ and}$$

$$s_3 = c \leftarrow a_1, \dots, a_{i-1}, b_1, \dots, b_m, a_{i+1}, \dots, a_n$$

⌘ s_2 is an **antecedent cleaned** version of s_1 iff

$$s_1 = \neg a_i \leftarrow a_1, \dots, a_i, \dots, a_n$$

$$s_2 = \neg a_i \leftarrow a_1, \dots, a_{i-1}, a_{i+1}, \dots, a_n$$

Let S be a set of strict rules. Then,

- ⌘ S is **closed under transposition** iff for each rule $s_1 \in S$, if s_2 is a transpositive version of s_1 then $s_2 \in S$.
 - ⌘ S is **closed under transitivity** iff for each rule $s_1, s_2 \in S$, if s_3 is a transitive version of s_1 and s_2 then $s_3 \in S$.
 - ⌘ S is **closed under antecedent cleaning** iff for each rule $s_1 \in S$, if s_2 is an antecedent cleaned version of s_1 then $s_2 \in S$.
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Normal Extended Logic Programs

⌘ A program P is called **normal** iff

1. $\text{strict}(P)$ is **closed under transposition, transitivity, and antecedent cleaning**, and
2. $\text{defeasible}(P)$ consists of **normal rules** only.

⌘ **Proposition**

Given an ELP P such that $\text{defeasible}(P)$ consists of normal rules only, there is a unique normal ELP which is obtained by closing $\text{strict}(P)$ under three operations.

⌘ Theorem

Every normal ELP has at least one answer set.

⌘ Example (Married-John)

$r \leftarrow,$ $p \leftarrow,$
 $m \leftarrow r, \text{ not } \neg m,$ $b \leftarrow p, \text{ not } \neg b,$
 $hs \leftarrow m,$ $\neg hs \leftarrow b,$
 $\neg m \leftarrow \neg hs,$ $\neg b \leftarrow hs,$
 $\neg m \leftarrow b,$ $\neg b \leftarrow m.$

The above program has two answer sets:

$\{ r, p, m, \neg b, hs \}$ and $\{ r, p, b, \neg m, \neg hs \}.$

Connection to Default Logic

- ⌘ Given a normal ELP P , its **associated default theory** $\Delta = (W, D)$ is defined as:

$$W = \{ c \vee \neg a_1 \vee \dots \vee \neg a_n \mid \\ c \leftarrow a_1, \dots, a_n \in \text{strict}(P) \}.$$

$$D = \{ a_1 \wedge \dots \wedge a_n : c / c \mid \\ c \leftarrow a_1, \dots, a_n, \text{not } \neg c \in \text{defeasible}(P) \}.$$

- ⌘ **Theorem** (not in the paper)

There is a 1-1 correspondence between answer sets of P and extensions of Δ .

Application (1): Constraint Satisfaction Problems

- ⌘ Put N pigeons into M holes so that there is at most one pigeon in a hole. The program is coded as [Niemela]:
- $\text{pos}(P,H) \leftarrow \text{pigeon}(P), \text{hole}(H), \text{not } \neg \text{pos}(P,H),$
 - $\neg \text{pos}(P,H) \leftarrow \text{pigeon}(P), \text{hole}(H), \text{not } \text{pos}(P,H),$
 - $\leftarrow \text{pigeon}(P), \text{hole}(H), \text{hole}(K), \text{pos}(P,H), \text{pos}(P,K), H \neq K,$
 - $\leftarrow \text{pigeon}(P), \text{not } \text{hashole}(P),$
 - $\text{hashole}(P) \leftarrow \text{pigeon}(P), \text{hole}(H), \text{pos}(P,H),$
 - $\leftarrow \text{pigeon}(P), \text{pigeon}(Q), \text{hole}(H), \text{pos}(P,H), \text{pos}(Q,H), P \neq Q.$
- where $\text{pos}(P,H)$ represents a legal position of a pigeon P in a hole H.
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- ⌘ The above program is **beyond** the class of normal ELPs even after closing the strict rules under three operations, due to the existence of **integrity constraints**.
 - ⌘ By definition, normal ELPs do **not** contain integrity constraints – rules with empty heads.
 - ⌘ In fact, $\leftarrow F$ is **semantically equivalent** to $c \leftarrow F$, **not** c for any literal c under the answer set semantics, which is **not** a normal default rule.
 - ⌘ In ASP, however, **integrity constraints play an important role**; imposing conditions on the solutions or pruning unwanted answer sets.
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Handling Constraints in Normal ELPs

⌘ An **integrity constraint** is of the form:

$$\leftarrow a_1, \dots, a_n, \text{not } b_1, \dots, \text{not } b_m$$

⌘ **Proposition** Let P be a normal ELP and IC a set of integrity constraints. If S is an answer set of $P \cup IC$, then S is an answer set of P .

⌘ $P \cup IC$ eliminates answer sets of P which do not satisfy the constraints.

⌘ **Normal rules + integrity constraints** can naturally encode many CSP problems and combinatorial problems such as the n -queen problem, the k -colorability problem, etc.

Application (2): Multiagent Systems

- ⌘ Suppose a **multiagent system** which consists of k agents.
 - ⌘ An agent i has a knowledge base P_i in ASP, where definite knowledge is shared by every agent $\text{strict}(P_1) = \dots = \text{strict}(P_k)$ (e.g., **a shared ontology**), while individual agents have their own default knowledge as $\text{defeasible}(P_i)$.
 - ⌘ In this situation, **combining knowledge bases** $\bigcup_{i=1}^k P_i$ **may produce no answer set**, even if each individual program P_i is consistent and meaningful.
 - ⌘ When every P_i is coded in **normal ELPs**, $\bigcup_{i=1}^k P_i$ is also a normal ELP and **always has a consistent answer set** as far as the shared definite knowledge is consistent.
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Application (3): Program Development

- ⌘ In ASP, a piece of information introduced to a program may lead to a **total collapse** of all entailment (no answer set).
 - ⌘ This is a serious problem when dealing with knowledge bases that **evolve with time and change dynamically**.
 - ⌘ In **normal ELPs**, on the other hand, such a “total collapse” problem **never happens**. A knowledge engineer feels free to update a part of a program without spoiling the global meaning as far as defeasible information is encoded in the form of normal rules.
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Discussion: ASP vs. Normal ELP

⌘ Normal ELPs generally **increase the entailment power of ASP.**

⌘ Given the information,

mammal \leftarrow human, \neg mammal \leftarrow

ASP does **not** entail \neg human, while it is entailed in the normal ELP after closing the program by 3 operations.

⌘ **Strict rules naturally have their contrapositive meaning in many programs.** ASP does not read a strict rule in a contrapositive way, which may result in missing information **implicitly represented** in a program.

Discussion: ASP vs. Classical Logic

- ⌘ Concealing contrapositive meaning makes the theory of LP **different** from first-order logic.
 - ⌘ For example, suppose an agent having the information $P = \{ p \leftarrow q \}$. When new information $Q = \{ \neg p \}$ has arrived, what conclusion should be derived by the agent?
 - ⌘ If P is read as a **Horn logic program**, $\neg q$ is derived from $P \cup Q$ because it is logically equivalent to $\{ p \vee \neg q, \neg p \}$.
If P is read as an **ELP**, $\neg q$ is **not** derived.
 - ⌘ The same program may have **different** meaning depending on its reading, **even in the absence of defeasible rules**.
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Discussion: ASP vs. Default Logic

- ⌘ In **default logic**, defeasible rules are represented by defaults and strict rules are represented by first-order formulas.
 - ⌘ Comparing ASP and DL, default logic appears relatively simple, intuitive and straightforward to understand the principle of knowledge representation.
 - ⌘ **Normal ELPs** have the same spirit as default logic in this sense.
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Discussion: Computational Cost

- ⌘ Closure computation on strict rules introduces **additional costs**. In particular, transitivity would **exponentially increase** the number of strict rules in a program.
 - ⌘ In practice, the closure computation could be automatically done as a compilation process (**partial evaluation**). Alternatively, in runtime environments, the closure computation could be done only for the part of strict rules that are needed to solve a particular problem.
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