2. Answer Sets of Traditional Logic Programs

Beyond the study of proofs in mathematics, logic has been applied to designing and reasoning about computer hardware and software. One of the prominent applications of logic in computer science is the use of logic as a programming language. *Declarative programming* specifies what is to be computed, but not necessarily how it is computed.

A logic program is a set of axioms, or rules, defining relations between objects. A computation of a logic program is a deduction of consequences of the program. A program defines a set of consequences, which is its meaning. The art of logic programming is constructing concise and elegant programs that have the desired meaning. (From *The Art of Prolog*, page 9.)

We will study "Answer Set Programming (ASP)," a form of declarative programming oriented towards difficult combinatorial search problems. It has been applied, for instance, to knowledge representation, to plan generation, and to product configuration problems in artificial intelligence and to graph-theoretic problems arising in VLSI design, in historical linguistics, and in bioinformatics.

The idea of ASP is to represent the search problem we are interested in as the problem of finding the answer sets of a formula, and then find the solutions using *answer set solvers*, such as the systems SMODELS and DLV.

Horn Formulas and their Answer Sets

The definition of an answer set was originally proposed as a semantics for Prolog programs with negation, and later extended to more general logic programs. We begin by considering the case of "Horn" formulas.

Syntax

A Horn formula is a conjunction of several (0 or more) implications of the form $F \to A$, where F is a conjunction of several (0 or more) atoms, and A

is an atom. Traditionally, it is written as a set of implications (written backward), which are often called "positive rules", that have the form

$$A_0 \leftarrow A_1 \wedge \dots \wedge A_m \tag{1}$$

where $m \geq 0$ and A_0, \ldots, A_m are propositional atoms. The consequent A_0 of implication (1) is often called the *head* of the rule, and the antecedent

$$A_1 \wedge \cdots \wedge A_m$$

is often called the *body* of the rule. If the body is the empty conjunction (m=0) then the rule is called a *fact* and identified with its head A_0 .

A positive program is a finite set of positive rules. For instance,

$$\begin{array}{l}
p \\
r \leftarrow p \land q
\end{array} \tag{2}$$

is a positive program. We identify a positive program with the corresponding Horn formula.

Answer Sets of a Positive Program

Below we identify an interpretation with the set of atoms that are true in it. For instance, an interpretation I on signature $\{p,q\}$ such that $I(p) = \mathbf{f}$, and $I(q) = \mathbf{t}$ is identified with $\{q\}$.

2.1 For any positive program Π , the intersection of all sets satisfying Π satisfies Π also.

This assertion allows us to talk about the *smallest* set of atoms that satisfies Π [van Emden and Kowalski, 1976]. This set is called the answer set of Π .

For instance, the sets of atoms satisfying program (2) are

$$\{p\}, \{p,r\}, \{p,q,r\},$$

and its answer set is $\{p\}$.

Intuitively, we can think of (1) as a rule for generating atoms: once you have generated A_1, \ldots, A_m , you are allowed to generate A_0 . The answer set is the set of all atoms that can be generated by applying rules of the program in any order. For instance, the first rule of (2) allows us to include p in the answer set. The second rule says that we can add r to the answer set if we have already included p and q. Given these two rules only, we can generate no atoms besides p. If we extend program (2) by adding the rule $q \leftarrow p$ then the answer set will become $\{p, q, r\}$.

Answer Sets of a Program with Negation

A traditional rule extends the syntax of a positive rule as in the following:

$$A_0 \leftarrow A_1 \wedge \dots \wedge A_m \wedge \neg A_{m+1} \wedge \dots \wedge \neg A_n \tag{3}$$

where $n \geq m \geq 0$ and A_0, \ldots, A_n are propositional atoms. Similarly as before, we call A_0 the head, and

$$A_1 \wedge \cdots \wedge A_m \wedge \neg A_{m+1} \wedge \cdots \wedge \neg A_n$$

the body of the rule.

A traditional program is a finite set of traditional rules. For instance, the following is a traditional program that is not positive.

$$\begin{array}{l}
p \leftarrow \neg q \\
q \leftarrow \neg r
\end{array} \tag{4}$$

To extend the definition of an answer set to traditional programs, we will introduce an auxiliary definition. The $reduct \Pi^X$ of a traditional program Π relative to a set X of atoms is obtained from Π by replacing every occurrence of the form $\neg A$ by \bot if $X \models A$ and \top otherwise. Note that Π^X can be equivalently rewritten as a positive program by first dropping all rules that contains \bot and then removing all occurrences of \top from the remaining rules. Thus Π^X is essentially a transformation that turns a traditional program into a positive program.

We say that X is an answer set of Π if X is the answer set of Π^X (that is, the smallest set of atoms satisfying Π^X).

This definition was proposed in the paper [Gelfond and Lifschitz, 1988], where answer sets were called "stable models." Its idea came from early work on nonmonotonic reasoning [McDermott and Doyle, 1980] and [Reiter, 1980], which was related to logic programming in [Gelfond, 1987].

If Π is positive then, for any X, $\Pi^X = \Pi$. It follows that the new definition of an answer set is a generalization of the definition from the previous section: for any positive traditional program Π , X is the smallest set of atoms satisfying Π^X iff X is the smallest set of atoms satisfying Π .

Example: if Π is (4) and $X = \{q\}$ then the reduct Π^X is

$$\begin{array}{l} p \leftarrow \bot \\ q \leftarrow \top, \end{array}$$

which is equivalent to the single fact q. The answer set of Π^X is $\{q\}$. Consequently, $\{q\}$ is an answer set of Π .

Intuitively, rule (3) allows us to generate A_0 as soon as we generated the atoms A_1, \ldots, A_m provided that none of the atoms A_{m+1}, \ldots, A_n can be generated using the rules of the program. There is a vicious circle in this sentence: to decide whether a rule of Π can be used to generate a new atom, we need to know which atoms can be generated using the rules of Π . The definition of an answer set overcomes this difficulty by employing a "fixpoint construction." Take a set X that you suspect may be exactly the set of atoms that can be generated using the rules of Π . Under this assumption, Π has the same meaning as the positive program Π^X . Consider the answer set of Π^X . If this set is exactly identical to the set X that you started with then X was a "good guess"; it is indeed an answer set of Π .

2.2 Check that $\{p\}$ is not an answer set of program (4).

Does program (4) have answer sets other than $\{q\}$? We can find this out by checking each of the remaining 6 subsets of $\{p,q,r\}$. We can do a little better by using the following general properties of answer sets of traditional programs.

- **2.3** If X is an answer set of a traditional program Π then every element of X is the head of one of the rules of Π .
- **2.4** (a) For any traditional program Π and any sets X, Y of atoms, if $X \subseteq Y$, then any set Z of atoms that satisfies Π^X satisfies Π^Y as well. (b) If X is an answer set for a traditional program Π then no proper subset of X can be an answer set of Π .

In application to program (4), the assertion of Problem 2.3 tells us that its answer sets do not contain r, so that we only need to check \emptyset and $\{p,q\}$. By the assertion of Problem 2.4(b), \emptyset cannot be an answer set because it is a proper subset of the answer set $\{q\}$, and $\{p,q\}$ cannot be an answer set because the answer set $\{q\}$ is its proper subset. Consequently, $\{q\}$ is the only answer set of (4).

2.5 Find an answer set of the program Π_n consisting of n rules

$$p_i \leftarrow \neg p_{i+1} \qquad (1 \le i \le n)$$

where n is a positive integer. (Program (4) is essentially Π_2 .)

Each of the programs Π_n has actually a unique answer set. On the other hand, the program

$$\begin{array}{l}
p \leftarrow \neg q \\
q \leftarrow \neg p
\end{array} \tag{5}$$

has two answer sets: $\{p\}$ and $\{q\}$. The one-rule program

$$r \leftarrow \neg r$$
 (6)

has no answer sets.

2.6 Find all answer sets of the following program, which extends (5) by two additional rules:

$$\begin{aligned} p &\leftarrow \neg q \\ q &\leftarrow \neg p \\ r &\leftarrow p \\ r &\leftarrow q. \end{aligned}$$

2.7 Find all answer sets of the following combination of programs (5) and (6) plus one more rule:

$$\begin{aligned} p &\leftarrow \neg q \\ q &\leftarrow \neg p \\ r &\leftarrow \neg r \\ r &\leftarrow p. \end{aligned}$$

LPARSE and SMODELS

The most widely used among the answer set solvers available today is SMODELS, with the preprocessor LPARSE. It has been designed and implemented at the Helsinki University of Technology, and it can be downloaded from its web site, http://www.tcs.hut.fi/Software/smodels/.

In the input language of LPARSE, as in Prolog, :- stands for \leftarrow , not stands for \neg , , stands for \wedge , ; stands for \vee , and each rule is followed by a period. If we want to find, for instance, the answer sets of the program from Problem 2.6, we create the file

```
p :- not q.
q :- not p.
r :- p.
r :- q.
```

called, say, p2.6. Then we invoke LPARSE and SMODELS as follows:

The zero at the end indicates that we want to compute all answer sets; a positive number k would tell SMODELS to terminate after computing k answer sets. The default value of k is 1. The main part of the output generated in response to this command line is the list of the program's answer sets:

Answer: 1

Stable Model: r p

Answer: 2

Stable Model: r q

("stable model" means "answer set").

A group of rules that follow a pattern can be often described concisely in the input language of LPARSE using schematic variables. As in Prolog, variables must be capitalized. Consider, for instance, the programs Π_n from Problem 2.5. To describe Π_7 , we don't have to write out each of its 7 rules. Instead, let's agree to use, for instance, the symbol index to represent numbers between 1 and 7. We can write Π_7 as

```
index(1..7).
#domain index(I).
p(I) :- not p(I+1).
```

The first two lines declare I to be a variable ranging over $\{1, ..., 7\}$. The auxiliary symbols used to describe the ranges of variables, such as index, are called domain predicates. Grounding—translating schematic expressions, such as p(I) := not p(I+1), into sets of rules—is the main computational task performed by LPARSE.

Instead of declaring a variable, we can specify its range in the body of the rule in which it is used:

```
index(1..7).
p(I) :- not p(I+1), index(I).
```

The family of programs Π_n (n = 1, 2, ...) can be described by a program schema with the parameter n:

```
index(1..n).
#domain index(I).
p(I) :- not p(I+1).
```

Let's name this file p2.5. When this schema is given to LPARSE as input, the value of the constant n can be specified in the command line using the option -c, as follows:

```
% lparse -c n=7 p2.5 | smodels 0
```

When the input file uses domain predicates, the output of SMODELS lists the objects satisfying each domain predicate along with the elements of the answer set. Information on the extents of domain predicates in the output can be suppressed using the LPARSE option -d none:

% lparse -c n=7 -d none p2.5 | smodels 0

- **2.7** Use SMODELS to verify that Π_n has a unique answer set for n = 10 and n = 100
- **2.8** Consider the program obtained from Π_n by adding the rule

$$p_{n+1} \leftarrow \neg p_1$$
.

How many answer sets does this program have, in your opinion? Check your conjecture for n = 7 and n = 8 using SMODELS.

Later we will discuss several other useful features of the input language of LPARSE. A complete manual can be downloaded from the SMODELS web site.

References

- [Gelfond and Lifschitz, 1988] Michael Gelfond and Vladimir Lifschitz. The stable model semantics for logic programming. In Robert Kowalski and Kenneth Bowen, editors, *Proceedings of International Logic Programming Conference and Symposium*, pages 1070–1080. MIT Press, 1988.
- [Gelfond, 1987] Michael Gelfond. On stratified autoepistemic theories. In *Proceedings of National Conference on Artificial Intelligence (AAAI)*, pages 207–211, 1987.
- [McDermott and Doyle, 1980] Drew McDermott and Jon Doyle. Nonmonotonic logic I. Artificial Intelligence, 13:41–72, 1980.
- [Reiter, 1980] Raymond Reiter. A logic for default reasoning. Artificial Intelligence, 13:81–132, 1980.
- [van Emden and Kowalski, 1976] Maarten van Emden and Robert Kowalski. The semantics of predicate logic as a programming language. *Journal of ACM*, 23(4):733–742, 1976.