

# Appendix A

## Logic and Resolution

One of the earliest formalisms for the representation of knowledge is *logic*. The formalism is characterized by a well-defined syntax and semantics, and provides a number of *inference rules* to manipulate logical formulas on the basis of their form in order to derive new knowledge. Logic has a very long and rich tradition, going back to the ancient Greeks: its roots can be traced to Aristotle. However, it took until the present century before the mathematical foundations of modern logic were laid, amongst others by T. Skolem, J. Herbrand, K. Gödel, and G. Gentzen. The work of these great and influential mathematicians rendered logic firmly established before the area of computer science came into being.

Already from the early 1950s, as soon as the first digital computers became available, research was initiated on using logic for problem solving by means of the computer. This research was undertaken from different points of view. Several researchers were primarily interested in the mechanization of mathematical proofs: the efficient automated generation of such proofs was their main objective. One of them was M. Davis who, already in 1954, developed a computer program which was capable of proving several theorems from number theory. The greatest triumph of the program was its proof that the sum of two even numbers is even. Other researchers, however, were more interested in the study of human problem solving, more in particular in heuristics. For these researchers, mathematical reasoning served as a point of departure for the study of heuristics, and logic seemed to capture the essence of mathematics; they used logic merely as a convenient language for the formal representation of human reasoning. The classical example of this approach to the area of theorem proving is a program developed by A. Newell, J.C. Shaw and H.A. Simon in 1955, called the *Logic Theory Machine*. This program was capable of proving several theorems from the Principia Mathematica of A.N. Whitehead and B. Russell. As early as 1961, J. McCarthy, amongst others, pointed out that theorem proving could also be used for solving non-mathematical problems. This idea was elaborated by many authors. Well known is the early work on so-called *question-answering systems* by J.R. Slagle and the later work in this field by C.C. Green and B. Raphael.

After some initial success, it soon became apparent that the inference rules known at that time were not as suitable for application in digital computers as hoped for. Many AI researchers lost interest in applying logic, and shifted their attention towards the development of other formalisms for a more efficient representation and manipulation of information. The breakthrough came thanks to the development of an efficient and flexible inference rule in 1965, named *resolution*, that allowed applying logic for automated problem solving by the

computer, and theorem proving finally gained an established position in artificial intelligence and, more recently, in the computer science as a whole as well.

Logic can directly be used as a knowledge-representation formalism for building knowledge systems; currently however, this is done only on a small scale. But then, the clear semantics of logic makes the formalism eminently suitable as a point of departure for understanding what the other knowledge-representation formalisms are all about. In this chapter, we first discuss the subject of how knowledge can be represented in logic, departing from propositional logic, which although having a rather limited expressiveness, is very useful for introducing several important notions. First-order predicate logic, which offers a much richer language for knowledge representation, is treated in Section A.2. The major part of this chapter however will be devoted to the algorithmic aspects of applying logic in an automated reasoning system, and resolution in particular will be the subject of study.

## A.1 Propositional logic

Propositional logic may be viewed as a representation language which allows us to express and reason with statements that are either *true* or *false*. Examples of such statements are:

‘A full-adder is a logical circuit’  
 ‘10 is greater than 90’

Clearly, such statement need not be true. Statements like these are called *propositions* and are usually denoted in propositional logic by uppercase letters. Simple propositions such as  $P$  and  $Q$  are called *atomic propositions* or *atoms* for short. Atoms can be combined with so-called *logical connectives* to yield *composite propositions*. In the language of propositional logic, we have the following five connectives at our disposal:

negation:	$\neg$	(not)
conjunction:	$\wedge$	(and)
disjunction:	$\vee$	(or)
implication:	$\rightarrow$	(if then)
bi-implication:	$\leftrightarrow$	(if and only if)

For example, when we assume that the propositions  $G$  and  $D$  have the following meaning

$G =$  ‘A Bugatti is a car’  
 $D =$  ‘A Bugatti has 5 wheels’

then the composite proposition

$G \wedge D$

has the meaning:

‘A Bugatti is a car *and* a Bugatti has 5 wheels’

However, not all formulas consisting of atoms and connectives are (composite) propositions. In order to distinguish syntactically correct formulas that do represent propositions from those that do not, the notion of a well-formed formula is introduced in the following definition.

**Definition A.1** A well-formed formula in propositional logic is an expression having one of the following forms:

- (1) An atom is a well-formed formula.
- (2) If  $F$  is a well-formed formula, then  $(\neg F)$  is a well-formed formula.
- (3) If  $F$  and  $G$  are well-formed formulas, then  $(F \wedge G)$ ,  $(F \vee G)$ ,  $(F \rightarrow G)$  and  $(F \leftrightarrow G)$  are well-formed formulas.
- (4) No other formula is well-formed.

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**EXAMPLE A.1**

Both formulas  $(F \wedge (G \rightarrow H))$  and  $(F \vee (\neg G))$  are well-formed according to the previous definition, but the formula  $(\rightarrow H)$  is not.

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In well-formed formulas, parentheses may be omitted as long as no ambiguity can occur; the adopted priority of the connectives is, in decreasing order, as follows:

$$\neg \wedge \vee \rightarrow \leftrightarrow$$

In the following, the term formula is used as an abbreviation when a well-formed formula is meant.

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**EXAMPLE A.2**

The formula  $P \rightarrow Q \wedge R$  is the same as the formula  $(P \rightarrow (Q \wedge R))$ .

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The notion of well-formedness of formulas only concerns the syntax of formulas in propositional logic: it does not express the formulas to be either *true* or *false*. In other words, it tells us nothing with respect to the semantics or meaning of formulas in propositional logic. The truth or falsity of a formula is called its *truth value*. The meaning of a formula in propositional logic is defined by means of a function  $w : \text{PROP} \rightarrow \{\text{true}, \text{false}\}$  which assigns to each proposition in the set of propositions PROP either the truth value *true* or *false*. Consequently, the information that the atom  $P$  has the truth value *true*, is now denoted by  $w(P) = \text{true}$ , and the information that the atom  $P$  has the truth value *false*, is denoted by  $w(P) = \text{false}$ . Such a function  $w$  is called an interpretation function, or an *interpretation* for short, if it satisfies the following properties (we assume  $F$  and  $G$  to be arbitrary well-formed formulas):

- (1)  $w(\neg F) = \text{true}$  if  $w(F) = \text{false}$ , and  $w(\neg F) = \text{false}$  if  $w(F) = \text{true}$ .
- (2)  $w(F \wedge G) = \text{true}$  if  $w(F) = \text{true}$  and  $w(G) = \text{true}$ ; otherwise  $w(F \wedge G) = \text{false}$ .
- (3)  $w(F \vee G) = \text{false}$  if  $w(F) = \text{false}$  and  $w(G) = \text{false}$ ; in all other cases, that is, if at least one of the function values  $w(F)$  and  $w(G)$  equals *true*, we have  $w(F \vee G) = \text{true}$ .
- (4)  $w(F \rightarrow G) = \text{false}$  if  $w(F) = \text{true}$  and  $w(G) = \text{false}$ ; in all other cases we have  $w(F \rightarrow G) = \text{true}$ .
- (5)  $w(F \leftrightarrow G) = \text{true}$  if  $w(F) = w(G)$ ; otherwise  $w(F \leftrightarrow G) = \text{false}$ .

Table A.1: The meanings of the connectives.

$F$	$G$	$\neg F$	$F \wedge G$	$F \vee G$	$F \rightarrow G$	$F \leftrightarrow G$
<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>
<i>true</i>	<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>
<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>
<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>true</i>

Table A.2: Truth table for  $P \rightarrow (\neg Q \wedge R)$ .

$P$	$Q$	$R$	$\neg Q$	$\neg Q \wedge R$	$P \rightarrow (\neg Q \wedge R)$
<i>true</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>false</i>
<i>true</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>false</i>	<i>false</i>
<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>
<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>
<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>
<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>
<i>false</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>
<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>true</i>

These rules are summarized in Table A.1. The first two columns in this table list all possible combinations of truth values for the atomic propositions  $F$  and  $G$ ; the remaining columns define the meanings of the respective connectives. If  $w$  is an interpretation which assigns to a given formula  $F$  the truth value *true*, then  $w$  is called a *model* for  $F$ .

By repeated applications of the rules listed in table 2.1, it is possible to express the truth value of an arbitrary formula in terms of the truth values of the atoms the formula is composed of. In a formula containing  $n$  different atoms, there are  $2^n$  possible ways of assigning truth values to the atoms in the formula.

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**EXAMPLE A.3**

Table A.2 lists all possible combinations of truth values for the atoms in the formula  $P \rightarrow (\neg Q \wedge R)$ ; for each combination, the resulting truth value for this formula is determined. Such a table where all possible truth values for the atoms in a formula  $F$  are entered together with the corresponding truth value for the whole formula  $F$ , is called a *truth table*.

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**Definition A.2** A formula is called a *valid formula* if it is true under all interpretations. A valid formula is often called a *tautology*. A formula is called *invalid* if it is not valid.

So, a valid formula is true regardless of the truth or falsity of its constituent atoms.

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**EXAMPLE A.4**

The formula  $((P \rightarrow Q) \wedge P) \rightarrow Q$  is an example of a valid formula. In the previous example we dealt with an invalid formula.

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**Definition A.3** A formula is called *unsatisfiable* or *inconsistent* if the formula is false under all interpretations. An *unsatisfiable* formula is also called a *contradiction*. A formula is called *satisfiable* or *consistent* if it is not unsatisfiable.

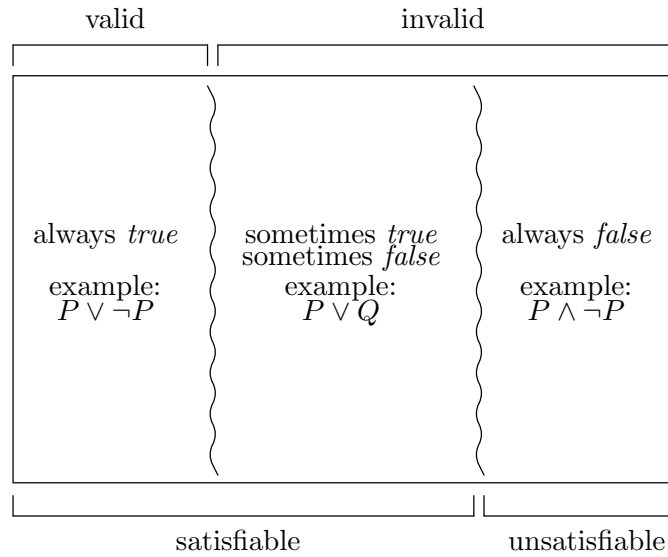


Figure A.1: Relationship between validity and satisfiability.

Table A.3: Truth table of  $\neg(P \wedge Q)$  and  $\neg P \vee \neg Q$ .

$P$	$Q$	$\neg(P \wedge Q)$	$\neg P \vee \neg Q$
<i>true</i>	<i>true</i>	<i>false</i>	<i>false</i>
<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>
<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>
<i>false</i>	<i>false</i>	<i>true</i>	<i>true</i>

Note that a formula is valid precisely when its negation is unsatisfiable and vice versa.

**EXAMPLE A.5**

The formulas  $P \wedge \neg P$  and  $(P \rightarrow Q) \wedge (P \wedge \neg Q)$  are both unsatisfiable.

Figure A.1 depicts the relationships between the notions of valid, invalid, and satisfiable, and unsatisfiable formulas.

**Definition A.4** Two formulas  $F$  and  $G$  are called equivalent, written as  $F \equiv G$ , if the truth values of  $F$  and  $G$  are the same under all possible interpretations.

Two formulas can be shown to be equivalent by demonstrating that their truth tables are identical.

**EXAMPLE A.6**

Table A.3 shows that  $\neg(P \wedge Q) \equiv \neg P \vee \neg Q$ .

Using truth tables the logical equivalences listed in Table 2.4 can easily be proven. These equivalences are called *laws of equivalence*. Law (a) is called the *law of double negation*; the laws (b) and (c) are called the *commutative laws*; (d) and (e) are the so-called *associative*

Table A.4: Laws of equivalence.

$\neg(\neg F) \equiv F$	(a)
$F \vee G \equiv G \vee F$	(b)
$F \wedge G \equiv G \wedge F$	(c)
$(F \wedge G) \wedge H \equiv F \wedge (G \wedge H)$	(d)
$(F \vee G) \vee H \equiv F \vee (G \vee H)$	(e)
$F \vee (G \wedge H) \equiv (F \vee G) \wedge (F \vee H)$	(f)
$F \wedge (G \vee H) \equiv (F \wedge G) \vee (F \wedge H)$	(g)
$F \leftrightarrow G \equiv (F \rightarrow G) \wedge (G \rightarrow F)$	(h)
$F \rightarrow G \equiv \neg F \vee G$	(i)
$\neg(F \wedge G) \equiv \neg F \vee \neg G$	(j)
$\neg(F \vee G) \equiv \neg F \wedge \neg G$	(k)

laws, and (f) and (g) are the *distributive laws*. The laws (j) and (k) are known as the *laws of De Morgan*. These laws often are used to transform a given well-formed formula into a logically equivalent but syntactically different formula.

In the following, a conjunction of formulas is often written as a set of formulas, where the elements of the set are taken as the conjunctive subformulas of the given formula.

**EXAMPLE A.7**

The set  $S = \{F \vee G, H\}$  represents the following formula:  $(F \vee G) \wedge H$ .

Truth tables can be applied to determine whether or not a given formula follows logically from a given set of formulas. Informally speaking, a formula logically follows from a set of formulas if it is satisfied by all interpretations satisfying the given set of formulas; we say that the formula is a logical consequence of the formulas in the given set. The following is a formal definition of this notion.

**Definition A.5** A formula  $G$  is said to be a logical consequence of the set of formulas  $F = \{F_1, \dots, F_n\}$ ,  $n \geq 1$ , denoted by  $F \models G$ , if for each interpretation  $w$  for which  $w(F_1 \wedge \dots \wedge F_n) = \text{true}$ , we have  $w(G) = \text{true}$ .

**EXAMPLE A.8**

The formula  $R$  is a logical consequence of the set of formulas  $\{P \wedge \neg Q, P \rightarrow R\}$ . Thus we can write  $\{P \wedge \neg Q, P \rightarrow R\} \models R$ .

Note that another way of stating that two formulas  $F$  and  $G$  are logically equivalent, that is,  $F \equiv G$ , is to say that both  $\{F\} \models G$  and  $\{G\} \models F$  hold. This tells us that the truth value of  $F$  and  $G$  are explicitly related to each other, which can also be expressed as  $\models (F \leftrightarrow G)$ .

Satisfiability, validity, equivalence and logical consequence are *semantic* notions; these properties are generally established using truth tables. However, for deriving logical consequences from of a set of formulas for example, propositional logic provides other techniques than using truth tables as well. It is possible to derive logical consequences by *syntactic* operations only. A formula which is derived from a given set of formulas then is guaranteed

to be a logical consequence of that set if the syntactic operations employed meet certain conditions. Systems in which such syntactic operations are defined, are called (*formal*) *deduction systems*. Various sorts of deduction systems are known. An example of a deduction system is an *axiomatic system*, consisting of a formal language, such as the language of propositional logic described above, a set of *inference rules* (the syntactic operations) and a set of *axioms*. In Section 2.4 we shall return to the subject of logical deduction.

## A.2 First-order predicate logic

In propositional logic, atoms are the basic constituents of formulas which are either *true* or *false*. A limitation of propositional logic is the impossibility to express general statements concerning similar cases. *First-order predicate logic* is more expressive than propositional logic, and such general statements can be specified in its language. Let us first introduce the language of first-order predicate logic. The following symbols are used:

- *Predicate symbols*, usually denoted by uppercase letters. Each predicate symbol has associated a natural number  $n$ ,  $n \geq 0$ , indicating the number of arguments the predicate symbol has; the predicate symbol is called an *n-place* predicate symbol. 0-place or *nullary* predicate symbols are also called (*atomic*) *propositions*. One-place, two-place and three-place predicate symbols are also called *unary*, *binary* and *ternary* predicate symbols, respectively.
- *Variables*, usually denoted by lowercase letters from the end of the alphabet, such as  $x$ ,  $y$ ,  $z$ , possibly indexed with a natural number.
- *Function symbols*, usually denoted by lowercase letters halfway the alphabet. Each function symbol has associated a natural number  $n$ ,  $n \geq 0$ , indicating its number of arguments; the function symbol is called *n-place*. Nullary function symbols are usually called *constants*.
- The *logical connectives* which have already been discussed in the previous section.
- Two *quantifiers*: the *universal quantifier*  $\forall$ , and the *existential quantifier*  $\exists$ . The quantifiers should be read as follows: if  $x$  is a variable, then  $\forall x$  means ‘for each  $x$ ’ or ‘for all  $x$ ’, and  $\exists x$  means ‘there exists an  $x$ ’.
- A number of *auxiliary symbols* such as parentheses and commas.

Variables and functions in logic are more or less similar to variables and functions in for instance algebra or calculus.

Before we define the notion of an atomic formula in predicate logic, we first introduce the notion of a term.

**Definition A.6** *A term is defined as follows:*

- (1) *A constant is a term.*
- (2) *A variable is a term.*
- (3) *If  $f$  is an  $n$ -place function symbol,  $n \geq 1$ , and  $t_1, \dots, t_n$  are terms, then  $f(t_1, \dots, t_n)$  is a term.*

(4) *Nothing else is a term.*

So, a term is either a constant, a variable or a function of terms. Recall that a constant may also be viewed as a nullary function symbol. An atomic formula now consists of a predicate symbol and a number of terms to be taken as the arguments of the predicate symbol.

**Definition A.7** *An atomic formula, or atom for short, is an expression of the form  $P(t_1, \dots, t_n)$ , where  $P$  is an  $n$ -place predicate symbol,  $n \geq 0$ , and  $t_1, \dots, t_n$  are terms.*

**EXAMPLE A.9**

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If  $P$  is a unary predicate symbol and  $x$  is a variable, then  $P(x)$  is an atom.  $Q(f(y), c, g(f(x), z))$  is an atom if  $Q$  is a ternary predicate symbol,  $c$  is a constant,  $f$  a unary function symbol,  $g$  a binary function symbol, and  $x, y$  and  $z$  are variables. For the same predicate symbols  $P$  and  $Q$ ,  $P(Q)$  is not an atom, because  $Q$  is not a term but a predicate symbol.

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Composite formulas can be formed using the five connectives given in Section 2.1, together with the two quantifiers  $\forall$  and  $\exists$  just introduced. As was done for propositional logic, we now define the notion of a well-formed formula in predicate logic. The following definition also introduces the additional notions of free and bound variables.

**Definition A.8** *A well-formed formula in predicate logic, and the set of free variables of a well-formed formula are defined as follows:*

- (1) *An atom is a well-formed formula. The set of free variables of an atomic formula consists of all the variables occurring in the terms in the atom.*
- (2) *Let  $F$  be a well-formed formula with an associated set of free variables. Then,  $(\neg F)$  is a well-formed formula. The set of free variables of  $(\neg F)$  equals the set of free variables of  $F$ .*
- (3) *Let  $F$  and  $G$  be well-formed formulas and let for each of these formulas a set of free variables be given. Then,  $(F \vee G)$ ,  $(F \wedge G)$ ,  $(F \rightarrow G)$  and  $(F \leftrightarrow G)$  are well-formed formulas. The set of free variables of each of these last mentioned formulas is equal to the union of the sets of free variables of  $F$  and  $G$ .*
- (4) *If  $F$  is well-formed formula and  $x$  is an element of the set of free variables of  $F$ , then both  $(\forall xF)$  and  $(\exists xF)$  are well-formed formulas. The set of free variables of each of these formulas is equal to the set of free variables of  $F$  from which the variable  $x$  has been removed. The variable  $x$  is called bound by the quantifier  $\forall$  or  $\exists$ .*
- (5) *Nothing else is a well-formed formula.*

Note that we have introduced the notion of a formula in the preceding definition only from a purely syntactical point of view: nothing has been said about the meaning of such a formula.

Parentheses will be omitted from well-formed formulas as long as ambiguity cannot occur; the quantifiers then have a higher priority than the connectives.

**Definition A.9** *A well-formed formula is called a closed formula, or a sentence, if its set of free variables is empty; otherwise it is called an open formula.*



**EXAMPLE A.10** 

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The set of free variables of the formula  $\forall x \exists y (P(x) \rightarrow Q(y, z))$  is equal to  $\{z\}$ . So, only one of the three variables in the formula is a free variable. The formula  $\forall x (P(x) \vee R(x))$  has no free variables at all, and thus is an example of a sentence.

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In what follows, we shall primarily be concerned with closed formulas; the term formula will be used to mean a closed formula, unless explicitly stated otherwise.

In the formula  $\forall x (A(x) \rightarrow G(x))$  all occurrences of the variable  $x$  in  $A(x) \rightarrow G(x)$  are governed by the associated universal quantifier;  $A(x) \rightarrow G(x)$  is called the *scope* of this quantifier.

**EXAMPLE A.11** 

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The scope of the universal quantifier in the formula

$$\forall x (P(x) \rightarrow \exists y R(x, y))$$

is  $P(x) \rightarrow \exists y R(x, y)$ ; the scope of the existential quantifier is the subformula  $R(x, y)$ .

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In propositional logic, the truth value of a formula under a given interpretation is obtained by assigning either the truth value *true* or *false* to each of its constituent atoms according to this specific interpretation. Defining the semantics of first-order predicate logic is somewhat more involved than in propositional logic. In predicate logic, a structure representing the ‘reality’ is associated with the *meaningless* set of symbolic formulas: in a structure the objects or elements of the domain of discourse, or domain for short, are enlisted, together with functions and relations defined on the domain.

**Definition A.10** A structure  $S$  is a tuple

$$S = (D, \{\bar{f}_i^n : D^n \rightarrow D, n \geq 1\}, \{\bar{P}_i^m : D^m \rightarrow \{true, false\}, m \geq 0\})$$

having the following components:

- (1) A non-empty set of elements  $D$ , called the domain of  $S$ ;
- (2) A set of functions defined on  $D^n$ ,  $\{\bar{f}_i^n : D^n \rightarrow D, n \geq 1\}$ ;
- (3) A non-empty set of mappings, called predicates, from  $D^m$  to the set of truth values  $\{true, false\}$ ,  $\{\bar{P}_i^m : D^m \rightarrow \{true, false\}, m \geq 0\}$ .

The basic idea underlying the definition of a structure is that we associate functions  $\bar{f}_i^n$  to function symbols  $f_i$  and predicates  $\bar{P}_i^m$  to predicate symbols  $P_i$ . Hence, we have to express how a given meaningless formula should be interpreted in a given structure: it is not possible to state anything about the truth value of a formula as long as it has not been prescribed which elements from the structure are to be associated with the elements in the formula.

**EXAMPLE A.12** 

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Consider the formula  $A(c)$ . We associate the predicate having the intended meaning ‘is a car’ with the predicate symbol  $A$ . The formula should be *true* if the constant representing a Bugatti is associated with  $c$ ; on the other hand, the same formula should be *false* if the constant representing a Volvo truck is associated with  $c$ . However, if we associate the predicate ‘Truck’ with  $A$ , the truth values of  $A(c)$  for the two constants should be opposite to the ones mentioned before.

In the following definition, we introduce the notion of an assignment, which is a function that assigns elements from the domain of a structure to the variables in a formula.

**Definition A.11** An assignment (valuation)  $v$  to a set of formulas  $F$  in a given structure  $S$  with domain  $D$  is a mapping from the set of variables in  $F$  to  $D$ .

The interpretation of (terms and) formulas in a structure  $S$  under an assignment  $v$  now consists of the following steps. First, the constants in the formulas are assigned elements from  $D$ . Secondly, the variables are replaced by the particular elements from  $D$  that have been assigned to them by  $v$ . Then, the predicate and function symbols occurring in the formulas are assigned predicates and functions from  $S$ . Finally, the truth values of the formulas are determined.

Before the notion of an interpretation is defined more formally, a simple example in which no function symbols occur, is given. For the reader who is not interested in the formal aspects of logic, it suffices to merely study this example.

**EXAMPLE A.13**

The open formula

$$F = A(x) \rightarrow O(x)$$

contains the unary predicate symbols  $A$  and  $O$ , and the free variable  $x$ . Consider the structure  $S$  consisting of the domain  $D = \{\text{bugatti}, \text{volvo-truck}, \text{alfa-romeo}\}$  and the set of predicates comprising of the following elements:

- a unary predicate  $Car$ , with the intended meaning ‘is a car’, defined by  $Car(\text{bugatti}) = \text{true}$ ,  $Car(\text{alfa-romeo}) = \text{true}$  and  $Car(\text{volvo-truck}) = \text{false}$ , and
- the unary predicate  $FourWheels$  with the intended meaning ‘has four wheels’, defined by  $FourWheels(\text{bugatti}) = \text{false}$ ,  $FourWheels(\text{volvo-truck}) = \text{false}$  and  $FourWheels(\text{alfa-romeo}) = \text{true}$ .

Let us take for the predicate symbol  $A$  the predicate  $Car$ , and for the predicate symbol  $O$  the predicate  $FourWheels$ . It will be obvious that the atom  $A(x)$  is *true* in  $S$  under any assignment  $v$  for which  $Car(v(x)) = \text{true}$ ; so, for example for the assignment  $v(x) = \text{alfa-romeo}$ , we have that  $A(x)$  is *true* in  $S$  under  $v$ . Furthermore,  $F$  is *true* in the structure  $S$  under the assignment  $v$  with  $v(x) = \text{alfa-romeo}$ , since  $A(x)$  and  $O(x)$  are both *true* in  $S$  under  $v$ . On the other hand,  $F$  is *false* in the structure  $S$  under the assignment  $v'$  with  $v'(x) = \text{bugatti}$ , because  $Car(\text{bugatti}) = \text{true}$  and  $FourWheels(\text{bugatti}) = \text{false}$  in  $S$ . Now, consider the closed formula

$$F' = \forall x(A(x) \rightarrow O(x))$$

and again the structure  $S$ . It should be obvious that  $F'$  is *false* in  $S$ .

Table A.5: Laws of equivalence for quantifiers.

$\neg\exists xP(x) \equiv \forall x\neg P(x)$	(a)
$\neg\forall xP(x) \equiv \exists x\neg P(x)$	(b)
$\forall x(P(x) \wedge Q(x)) \equiv \forall xP(x) \wedge \forall xQ(x)$	(c)
$\exists x(P(x) \vee Q(x)) \equiv \exists xP(x) \vee \exists xQ(x)$	(d)
$\forall xP(x) \equiv \forall yP(y)$	(e)
$\exists xP(x) \equiv \exists yP(y)$	(f)

**Definition A.12** An interpretation of terms in a structure  $S = (D, \{\bar{f}_i^n\}, \{\bar{P}_i^m\})$  under an assignment  $v$ , denoted by  $I_v^S$ , is defined as follows:

- (1)  $I_v^S(c_i) = d_i$ ,  $d_i \in D$ , where  $c_i$  is a constant.
- (2)  $I_v^S(x_i) = v(x_i)$ , where  $x_i$  is a variable.
- (3)  $I_v^S(f_i^n(t_1, \dots, t_n)) = \bar{f}_i^n(I_v^S(t_1), \dots, I_v^S(t_n))$ , where  $\bar{f}_i^n$  is a function from  $S$  associated with the function symbol  $f_i^n$ .

The truth value of a formula in a structure  $S$  under an assignment  $v$  for a given interpretation  $I_v^S$  is obtained as follows:

- (1)  $I_v^S(P_i^m(t_1, \dots, t_m)) = \bar{P}_i^m(I_v^S(t_1), \dots, I_v^S(t_m))$ , meaning that an atom  $P_i^m(t_1, \dots, t_m)$  is true in the structure  $S$  under the assignment  $v$  for the interpretation  $I_v^S$  if  $\bar{P}_i^m(I_v^S(t_1), \dots, I_v^S(t_m))$  is true, where  $\bar{P}_i^m$  is the predicate from  $S$  associated with  $P_i^m$ .
- (2) If the truth values of the formulas  $F$  and  $G$  have been determined, then the truth values of  $\neg F$ ,  $F \wedge G$ ,  $F \vee G$ ,  $F \rightarrow G$  and  $F \leftrightarrow G$  are defined by the meanings of the connectives as listed in Table A.1.
- (3)  $\exists xF$  is true under  $v$  if there exists an assignment  $v'$  differing from  $v$  at most with regard to  $x$ , such that  $F$  is true under  $v'$ .
- (4)  $\forall xF$  is true under  $v$  if for each  $v'$  differing from  $v$  at most with regard to  $x$ ,  $F$  is true under  $v'$ .

The notions valid, invalid, satisfiable, unsatisfiable, logical consequence, equivalence and model have meanings in predicate logic similar to their meanings in propositional logic. In addition to the equivalences listed in Table A.4, predicate logic also has some laws of equivalence for quantifiers, which are given in Table A.5. Note that the properties  $\forall x(P(x) \vee Q(x)) \equiv \forall xP(x) \vee \forall xQ(x)$  and  $\exists x(P(x) \wedge Q(x)) \equiv \exists xP(x) \wedge \exists xQ(x)$  do *not* hold.

We conclude this subsection with another example.

#### EXAMPLE A.14

We take the unary (meaningless) predicate symbols  $C$ ,  $F$ ,  $V$ ,  $W$  and  $E$ , and the constants  $a$  and  $b$  from a given first-order language. Now, consider the following formulas:

- (1)  $\forall x(C(x) \rightarrow V(x))$

- (2)  $F(a)$
- (3)  $\forall x(F(x) \rightarrow C(x))$
- (4)  $\neg E(a)$
- (5)  $\forall x((C(x) \wedge \neg E(x)) \rightarrow W(x))$
- (6)  $F(b)$
- (7)  $\neg W(b)$
- (8)  $E(b)$

Consider the structure  $S$  in the reality with a domain consisting of the elements  $a$  and  $b$ , which are assigned to the constants  $a$  and  $b$ , respectively. The set of predicates in  $S$  comprises the unary predicates  $Car$ ,  $Fast$ ,  $Vehicle$ ,  $FourWheels$ , and  $Exception$ , which are taken for the predicate symbols  $C$ ,  $F$ ,  $V$ ,  $W$ , and  $E$ , respectively. The structure  $S$  and the mentioned interpretation have been carefully chosen so as to satisfy the above-given closed formulas, for instance by giving the following intended meaning to the predicates:

$Car$	=	‘is a car’
$Fast$	=	‘is a fast car’
$Vehicle$	=	‘is a vehicle’
$FourWheels$	=	‘has four wheels’
$Exception$	=	‘is an exception’

In the given structure  $S$ , the formula numbered 1 expresses the knowledge that every car is a vehicle. The fact that an alfa-romeo is a fast car, has been stated in formula 2. Formula 3 expresses that every fast car is a car, and formula 4 states that an alfa-romeo is not an exception to the rule that cars have four wheels, which has been formalized in logic by means of formula 5. A Bugatti is a fast car (formula 6), but contrary to an alfa-romeo it does not have 4 wheels (formula 7), and therefore is an exception to the last mentioned rule; the fact that Bugattis are exceptions is expressed by means of formula 8.

It should be noted that in another structure with another domain and other predicates, the formulas given above might have completely different meanings.

### A.3 Clausal form of logic

Before turning our attention to reasoning in logic, we introduce in this section a syntactically restricted form of predicate logic, called the *clausal form of logic*, which will play an important role in the remainder of this chapter. This restricted form however, can be shown to be as expressive as full first-order predicate logic. The clausal form of logic is often employed, in particular in the fields of theorem proving and logic programming.

We start with the definition of some new notions.

**Definition A.13** A literal is an atom, called a positive literal, or a negation of an atom, called a negative literal.

**Definition A.14** A clause is a closed formula of the form

$$\forall x_1 \cdots \forall x_s (L_1 \vee \cdots \vee L_m)$$

where each  $L_i$ ,  $i = 1, \dots, m$ ,  $m \geq 0$ , is a literal, with  $L_i \neq L_j$  for each  $i \neq j$ , and  $x_1, \dots, x_s$ ,  $s \geq 0$ , are variables occurring in  $L_1 \vee \cdots \vee L_m$ . If  $m = 0$ , the clause is said to be the empty clause, denoted by  $\square$ .

The empty clause  $\square$  is interpreted as a formula which is always *false*, in other words,  $\square$  is an unsatisfiable formula.

A clause

$$\forall x_1 \cdots \forall x_s (A_1 \vee \cdots \vee A_k \vee \neg B_1 \vee \cdots \vee \neg B_n)$$

where  $A_1, \dots, A_k, B_1, \dots, B_n$  are atoms and  $x_1, \dots, x_s$  are variables, is equivalent to

$$\forall x_1 \cdots \forall x_s (B_1 \wedge \cdots \wedge B_n \rightarrow A_1 \vee \cdots \vee A_k)$$

as a consequence of the laws  $\neg F \vee G \equiv F \rightarrow G$  and  $\neg F \vee \neg G \equiv \neg(F \wedge G)$ , and is often written as

$$A_1, \dots, A_k \leftarrow B_1, \dots, B_n$$

The last notation is the more conventional one in logic programming. The commas in  $A_1, \dots, A_k$  each stand for a disjunction, and the commas in  $B_1, \dots, B_n$  indicate a conjunction.  $A_1, \dots, A_k$  are called the *conclusions* of the clause, and  $B_1, \dots, B_n$  the *conditions*.

Each well-formed formula in first-order predicate logic can be translated into a set of clauses, which is viewed as the conjunction of its elements. As we will see, this translation process may slightly alter the meaning of the formulas. We shall illustrate the translation process by means of an example. Before proceeding, we define two normal forms which are required for the translation process.

**Definition A.15** A formula  $F$  is in prenex normal form if  $F$  is of the form

$$Q_1 x_1 \cdots Q_n x_n M$$

where each  $Q_i$ ,  $i = 1, \dots, n$ ,  $n \geq 1$ , equals one of the two quantifiers  $\forall$  and  $\exists$ , and where  $M$  is a formula in which no quantifiers occur.  $Q_1 x_1 \dots Q_n x_n$  is called the prefix and  $M$  is called the matrix of the formula  $F$ .

**Definition A.16** A formula  $F$  in prenex normal form is in conjunctive normal form if the matrix of  $F$  is of the form

$$F_1 \wedge \cdots \wedge F_n$$

where each  $F_i$ ,  $i = 1, \dots, n$ ,  $n \geq 1$ , is a disjunction of literals.

**EXAMPLE A.15** \_\_\_\_\_

Consider the following three formulas:

$$\begin{aligned} & \forall x (P(x) \vee \exists y Q(x, y)) \\ & \forall x \exists y \forall z ((P(x) \wedge Q(x, y)) \vee \neg R(z)) \\ & \forall x \exists y ((\neg P(x) \vee Q(x, y)) \wedge (P(y) \vee \neg R(x))) \end{aligned}$$

The first formula is not in prenex normal form because of the occurrence of an existential quantifier in the ‘inside’ of the formula. The other two formulas are both in prenex normal form; moreover, the last formula is also in conjunctive normal form.

---

The next example illustrates the translation of a well-formed formula into a set of clauses. The translation scheme presented in the example however is general and can be applied to any well-formed formula in first-order predicate logic.

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**EXAMPLE A.16**

Consider the following formula:

$$\forall x(\exists yP(x, y) \vee \neg\exists y(\neg Q(x, y) \rightarrow R(f(x, y))))$$

This formula is transformed in eight steps, first into prenex normal form, subsequently into conjunctive normal form, amongst others by applying the laws of equivalence listed in the tables A.4 and A.5, and finally into a set of clauses.

*Step 1.* Eliminate all implication symbols using the equivalences  $F \rightarrow G \equiv \neg F \vee G$  and  $\neg(\neg F) \equiv F$ :

$$\forall x(\exists yP(x, y) \vee \neg\exists y(Q(x, y) \vee R(f(x, y))))$$

If a formula contains bi-implication symbols, these can be removed by applying the equivalence

$$F \leftrightarrow G \equiv (F \rightarrow G) \wedge (G \rightarrow F)$$

*Step 2.* Diminish the scope of the negation symbols in such a way that each negation symbol only governs a single atom. This can be accomplished by using the equivalences  $\neg\forall xF(x) \equiv \exists x\neg F(x)$ ,  $\neg\exists xF(x) \equiv \forall x\neg F(x)$ ,  $\neg(\neg F) \equiv F$ , together with the laws of De Morgan:

$$\forall x(\exists yP(x, y) \vee \forall y(\neg Q(x, y) \wedge \neg R(f(x, y))))$$

*Step 3.* Rename the variables in the formula using the equivalences  $\forall xF(x) \equiv \forall yF(y)$  and  $\exists xF(x) \equiv \exists yF(y)$ , so that each quantifier has its own uniquely named variable:

$$\forall x(\exists yP(x, y) \vee \forall z(\neg Q(x, z) \wedge \neg R(f(x, z))))$$

Formulas only differing in the names of their bound variables are called *variants*.

*Step 4.* Eliminate all existential quantifiers. For any existentially quantified variable  $x$  not lying within the scope of a universal quantifier, all occurrences of  $x$  in the formula within the scope of the existential quantifier can be replaced by a new, that is, not previously used, constant symbol  $c$ . The particular existential quantifier may then be removed. For instance, the elimination of the existential quantifier in the formula  $\exists xP(x)$  yields a formula  $P(c)$ . However, if an existentially quantified variable  $y$  lies within the scope of one or more universal quantifiers with the variables  $x_1, \dots, x_n$ ,  $n \geq 1$ , then the variable  $y$  may be functionally dependent upon  $x_1, \dots, x_n$ . Let this dependency be represented explicitly by means of a new  $n$ -place function symbol  $g$  such that  $g(x_1, \dots, x_n) = y$ . All occurrences of  $y$  within the scope of the existential quantifier then are replaced by the function term  $g(x_1, \dots, x_n)$ , after which the existential

quantifier may be removed. The constants and functions introduced in order to allow for the elimination of existential quantifiers are called *Skolem functions*.

The existentially quantified variable  $y$  in the example lies within the scope of the universal quantifier with the variable  $x$ , and is replaced by  $g(x)$ :

$$\forall x(P(x, g(x)) \vee \forall z(\neg Q(x, z) \wedge \neg R(f(x, z))))$$

Note that by replacing the existentially quantified variables by Skolem functions, we lose logical equivalence. Fortunately, it can be shown that a formula  $F$  is satisfiable if and only if the formula  $F'$ , obtained from  $F$  by replacing existentially quantified variables in  $F$  by Skolem functions, is satisfiable as well. In general, the satisfiability of  $F$  and  $F'$  will not be based on the same model, since  $F'$  contains function symbols not occurring in  $F$ . In the following, it will become evident that this property is sufficient for our purposes.

*Step 5.* Transform the formula into prenex normal form, by placing all the universal quantifiers in front of the formula:

$$\forall x \forall z (P(x, g(x)) \vee (\neg Q(x, z) \wedge \neg R(f(x, z))))$$

Note that this is allowed because by step 3 each quantifier applies to a uniquely named variable; this means that the scope of all quantifiers is the entire formula.

*Step 6.* Bring the matrix in conjunctive normal form using the distributive laws:

$$\forall x \forall z ((P(x, g(x)) \vee \neg Q(x, z)) \wedge (P(x, g(x)) \vee \neg R(f(x, z))))$$

*Step 7.* Select the matrix by disregarding the prefix:

$$(P(x, g(x)) \vee \neg Q(x, z)) \wedge (P(x, g(x)) \vee \neg R(f(x, z)))$$

All variables in the matrix are now implicitly considered to be universally quantified.

*Step 8.* Translate the matrix into a set of clauses, by replacing formulas of the form  $F \wedge G$  by a set of clauses  $\{F', G'\}$ , where  $F'$  and  $G'$  indicate that  $F$  and  $G$  are now represented using the notational convention of logic programming:

$$\{P(x, g(x)) \leftarrow Q(x, z), P(x, g(x)) \leftarrow R(f(x, z))\}$$

or the notation used in automated theorem proving:

$$\{P(x, g(x)) \vee \neg Q(x, z), P(x, g(x)) \vee \neg R(f(x, z))\}$$

We conclude this subsection with the definition of a special type of clause, a so-called Horn clause, which is a clause containing at most one positive literal.

**Definition A.17** *A Horn clause is a clause having one of the following forms:*

(1)  $A \leftarrow$

(2)  $\leftarrow B_1, \dots, B_n, n \geq 1$

(3)  $A \leftarrow B_1, \dots, B_n, n \geq 1$

*A clause of the form 1 is called a unit clause; a clause of form 2 is called a goal clause.*

Horn clauses are employed in the programming language Prolog. We will return to this observation in Section 2.7.2.

## A.4 Reasoning in logic: inference rules

In the Sections 2.1 and 2.2 we described how a meaning could be attached to a meaningless set of logical formulas. This is sometimes called the *declarative semantics* of logic. The declarative semantics offers a means for investigating for example whether or not a given formula is a logical consequence of a set of formulas. However, it is also possible to answer this question without examining the semantic contents of the formulas concerned, by applying so-called *inference rules*. Contrary to truth tables, inference rules are purely syntactic operations which only are capable of modifying the form of the elements of a given set of formulas. Inference rules either add, replace or remove formulas; most inference rules discussed in this book however add new formulas to a given set of formulas. In general, an inference rule is given as a schema in which a kind of meta-variables occur that may be substituted by arbitrary formulas. An example of such a schema is shown below:

$$\frac{A, A \rightarrow B}{B}$$

The formulas above the line are called the *premises*, and the formula below the line is called the *conclusion* of the inference rule. The above-given inference rule is known as *modus ponens*, and when applied, removes an implication from a formula. Another example of an inference rule, in this case for introducing a logical connective, is the following schema:

$$\frac{A, B}{A \wedge B}$$

Repeated applications of inference rules give rise to what is called a *derivation* or *deduction*. For instance, modus ponens can be applied to draw the conclusion  $S$  from the two formulas  $P \wedge (Q \vee R)$  and  $P \wedge (Q \vee R) \rightarrow S$ . It is said that there exists a derivation of the formula  $S$  from the set of clauses  $\{P \wedge (Q \vee R), P \wedge (Q \vee R) \rightarrow S\}$ . This is denoted by:

$$\{P \wedge (Q \vee R), P \wedge (Q \vee R) \rightarrow S\} \vdash S$$

The symbol  $\vdash$  is known as the *turnstile*.

---

### EXAMPLE A.17

Consider the set of formulas  $\{P, Q, P \wedge Q \rightarrow S\}$ . If the inference rule

$$\frac{A, B}{A \wedge B}$$

is applied to the formulas  $P$  and  $Q$ , the formula  $P \wedge Q$  is derived; the subsequent application of modus ponens to  $P \wedge Q$  and  $P \wedge Q \rightarrow S$  yields  $S$ . So,

$$\{P, Q, P \wedge Q \rightarrow S\} \vdash S$$


---

Now that we have introduced inference rules, it is relevant to investigate how the declarative semantics of a particular class of formulas and its *procedural semantics*, described by means of inference rules, are interrelated: if these two notions are related to each other, we are in the desirable circumstance of being able to assign a meaning to formulas which have been derived using inference rules, simply by our knowledge of the declarative meaning of the original set



of formulas. On the other hand, when starting with the known meaning of a set of formulas, it will then be possible to derive only formulas which can be related to that meaning. These two properties are known as the soundness and the completeness, respectively, of a collection of inference rules.

More formally, a collection of inference rules is said to be *sound* if and only if for each formula  $F$  derived by applying these inference rules on a given set of well-formed formulas  $S$  of a particular class (for example clauses), we have that  $F$  is a logical consequence of  $S$ . This property can be expressed more tersely as follows, using the notations introduced before:

$$\text{if } S \vdash F \text{ then } S \models F.$$

In other words, a collection of inference rules is sound if it preserves truth under the operations of a derivation. This property is of great importance, because only by applying sound inference rules it is possible to assign a meaning to the result of a derivation.

---

**EXAMPLE A.18**

The previously discussed inference rule modus ponens is an example of a sound inference rule. From the given formulas  $F$  and  $F \rightarrow G$ , the formula  $G$  can be derived by applying modus ponens, that is, we have  $\{F, F \rightarrow G\} \vdash G$ . On the other hand, if  $F \rightarrow G$  and  $F$  are both *true* under a particular interpretation  $w$ , then from the truth Table 2.1 we have that  $G$  is *true* under  $w$  as well. So,  $G$  is a logical consequence of the two given formulas:  $\{F, F \rightarrow G\} \models G$ .

---

The reverse property that by applying a particular collection of inference rules, each logical consequence  $F$  of a given set of formulas  $S$  can be derived, is called the *completeness* of the collection of inference rules:

$$\text{if } S \models F \text{ then } S \vdash F.$$

---

**EXAMPLE A.19**

The collection of inference rules only consisting of modus ponens is not complete for all well-formed formulas in propositional logic. For example, it is not possible to derive the formula  $P$  from  $\neg Q$  and  $P \vee Q$ , although  $P$  is a logical consequence of the two formulas. However, by combining modus ponens with other inference rules, it is possible to obtain a complete collection of inference rules.

---

The important question now arises if there exists a mechanical proof procedure, employing a particular sound and complete collection of inference rules, which is capable of determining whether or not a given formula  $F$  can be derived from a given set of formulas  $S$ . In 1936, A. Church and A.M. Turing showed, independently, that such a general proof procedure does not exist for first-order predicate logic. This property is called the *undecidability* of first-order predicate logic. All known proof procedures are only capable of deriving  $F$  from  $S$  (that is, are able to prove  $S \vdash F$ ) if  $F$  is a logical consequence of  $S$  (that is, if  $S \models F$ ); if  $F$  is not a logical consequence of  $S$ , then the proof procedure is not guaranteed to terminate.

However, for propositional logic there do exist proof procedures which always terminate and yield the right answer: for checking whether a given formula is a logical consequence of a certain set of formulas, we can simply apply truth tables. So, propositional logic is decidable.

The undecidability of first-order predicate logic has not refrained the research area of automated theorem proving from further progress. The major result of this research has been the development of an efficient and flexible inference rule, which is both sound and complete for proving inconsistency, called *resolution*. This is sometimes called *refutation completeness* (see below). However, the resolution rule is only suitable for manipulating formulas in clausal form. Hence, to use this inference rule on a set of arbitrary logical formulas in first-order predicate logic, it is required to translate each formula into the clausal form of logic by means of the procedure discussed in Section 2.3. This implies that resolution is *not* complete for *unrestricted* first-order predicate logic. The formulation of resolution as a suitable inference rule for automated theorem proving in the clausal form of logic has been mainly due to J.A. Robinson, who departed from earlier work by D. Prawitz. The final working-out of resolution in various algorithms, supplemented with specific implementation techniques, has been the work of a large number of researchers. Resolution is the subject of the remainder of this chapter.

## A.5 Resolution and propositional logic

We begin this section with a brief, informal sketch of the principles of resolution. Consider a set of formulas  $S$  in clausal form. Suppose we are given a formula  $G$ , also in clausal form, for which we have to prove that it can be derived from  $S$  by applying resolution. Proving  $S \vdash G$  is equivalent to proving that the set of clauses  $W$ , consisting of the clauses in  $S$  supplemented with the negation of the formula  $G$ , that is  $W = S \cup \{-G\}$ , is unsatisfiable. Resolution on  $W$  now proceeds as follows: first, it is checked whether or not  $W$  contains the empty clause  $\square$ ; if this is the case, then  $W$  is unsatisfiable, and  $G$  is a logical consequence of  $S$ . If the empty clause  $\square$  is not in  $W$ , then the resolution rule is applied on a suitable pair of clauses from  $W$ , yielding a new clause. Every clause derived this way is added to  $W$ , resulting in a new set of clauses on which the same resolution procedure is applied. The entire procedure is repeated until some generated set of clauses has been shown to contain the empty clause  $\square$ , indicating unsatisfiability of  $W$ , or until all possible new clauses have been derived.

The basic principles of resolution are best illustrated by means of an example from propositional logic. In Section 2.6 we turn our attention to predicate logic.

### EXAMPLE A.20

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Consider the following set of clauses:

$$\{C_1 = P \vee R, C_2 = \neg P \vee Q\}$$

These clauses contain *complementary* literals, that is, literals having opposite truth values, namely  $P$  and  $\neg P$ . Applying resolution, a new clause  $C_3$  is derived being the disjunction of the original clauses  $C_1$  and  $C_2$  in which the complementary literals have been cancelled out. So, application of resolution yields the clause

$$C_3 = R \vee Q$$

which then is added to the original set of clauses.

---

The resolution principle is described more precisely in the following definition.

**Definition A.18** Consider the two clauses  $C_1$  and  $C_2$  containing the literals  $L_1$  and  $L_2$  respectively, where  $L_1$  and  $L_2$  are complementary. The procedure of resolution proceeds as follows:

- (1) Delete  $L_1$  from  $C_1$  and  $L_2$  from  $C_2$ , yielding the clauses  $C'_1$  and  $C'_2$ ;
- (2) Form the disjunction  $C'$  of  $C'_1$  and  $C'_2$ ;
- (3) Delete (possibly) redundant literals from  $C'$ , thus obtaining the clause  $C$ .

The resulting clause  $C$  is called the resolvent of  $C_1$  and  $C_2$ . The clauses  $C_1$  and  $C_2$  are said to be the parent clauses of the resolvent.

Resolution has the important property that when two given parent clauses are *true* under a given interpretation, their resolvent is *true* under the same interpretation as well: resolution is a sound inference rule. In the following theorem we prove that resolution is sound for the case of propositional logic.

**THEOREM A.1** (soundness of resolution) Consider two clauses  $C_1$  and  $C_2$  containing complementary literals. Then, any resolvent  $C$  of  $C_1$  and  $C_2$  is a logical consequence of  $\{C_1, C_2\}$ .

**Proof:** We are given that the two clauses  $C_1$  and  $C_2$  contain complementary literals. So, it is possible to write  $C_1$  and  $C_2$  as  $C_1 = L \vee C'_1$  and  $C_2 = \neg L \vee C'_2$  respectively for some literal  $L$ . By definition, a resolvent  $C$  is equal to  $C'_1 \vee C'_2$  from which possibly redundant literals have been removed. Now, suppose that  $C_1$  and  $C_2$  are both *true* under an interpretation  $w$ . We then have to prove that  $C$  is *true* under the same interpretation  $w$  as well. Clearly, either  $L$  or  $\neg L$  is *false*. Suppose that  $L$  is *false* under  $w$ , then  $C_1$  obviously contains more than one literal, since otherwise  $C_1$  would be *false* under  $w$ . It follows that  $C'_1$  is *true* under  $w$ . Hence,  $C'_1 \vee C'_2$ , and therefore also  $C$ , is *true* under  $w$ . Similarly, it can be shown that the resolvent is *true* under  $w$  if it is assumed that  $L$  is *true*. So, if  $C_1$  and  $C_2$  are *true* under  $w$  then  $C$  is *true* under  $w$  as well. Hence,  $C$  is a logical consequence of  $C_1$  and  $C_2$ .  $\diamond$

Resolution is also a complete inference rule. Proving the completeness of resolution is beyond the scope of this book; we therefore confine ourselves to merely stating the property.

#### EXAMPLE A.21

---

In the definition of a clause in Section 2.3, it was mentioned that a clause was not allowed to contain duplicate literals. This condition appears to be a necessary requirement for the completeness of resolution. For example, consider the following set of formulas:

$$S = \{P \vee P, \neg P \vee \neg P\}$$

It will be evident that  $S$  is unsatisfiable, since  $P \vee P \equiv P$  and  $\neg P \vee \neg P \equiv \neg P$ . However, if resolution is applied to  $S$  then in every step the tautology  $P \vee \neg P$  is derived. It is not possible to derive the empty clause  $\square$ .

---

Until now we have used the notion of a derivation only in an intuitive sense. Before giving some more examples, we define the notion of a derivation in a formal way.

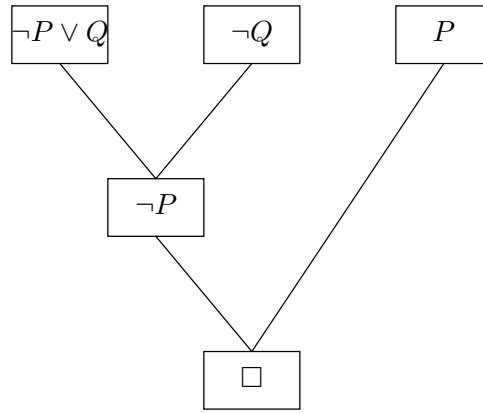


Figure A.2: A refutation tree.

**Definition A.19** Let  $S$  be a set of clauses and let  $C$  be a single clause. A derivation of  $C$  from  $S$ , denoted by  $S \vdash_{\mathcal{B}} C$ , is a finite sequence of clauses  $C_1, C_2, \dots, C_n$ ,  $n \geq 1$ , where each  $C_k$  either is a clause in  $S$  or a resolvent with parent clauses  $C_i$  and  $C_j$ ,  $i < k$ ,  $j < k$ ,  $i \neq j$ , from the sequence, and  $C = C_n$ . If  $C_n = \square$ , then the derivation is said to be a refutation of  $S$ , indicating that  $S$  is unsatisfiable.

---

**EXAMPLE A.22**

Consider the following set of clauses:

$$S = \{\neg P \vee Q, \neg Q, P\}$$

From  $C_1 = \neg P \vee Q$  and  $C_2 = \neg Q$  we obtain the resolvent  $C_3 = \neg P$ . From the clauses  $C_3$  and  $C_4 = P$  we derive  $C_5 = \square$ . So,  $S$  is unsatisfiable. The sequence of clauses  $C_1, C_2, C_3, C_4, C_5$  is a refutation of  $S$ . Note that it is not the only possible refutation of  $S$ . In general, a set  $S$  of clauses may have more than one refutation.

---

Notice that by the choice of the empty clause  $\square$  as a formula that is *false* under all interpretations, which is a *semantic* notion, the *proof-theoretical* notion of a refutation has obtained a suitable meaning. A derivation can be depicted in a graph, called a *derivation graph*. In the case of a refutation, the vertices in the derivation graph may be restricted to those clauses and resolvents which directly or indirectly contribute to the refutation. Such a derivation graph has the form of a tree and is usually called a *refutation tree*. The leaves of such a tree are clauses from the original set, and the root of the tree is the empty clause  $\square$ . The refutation tree for the derivation discussed in the previous example is shown in Figure A.2. Note that another refutation of  $S$  gives rise to another refutation tree.

## A.6 Resolution and first-order predicate logic

An important feature of resolution in first-order predicate logic, taking place in the basic resolution method, is the manipulation of terms. This has not been dealt with in the previous section, where we only had atomic propositions, connectives and auxiliary symbols as building blocks for propositional formulas. In this section, we therefore first discuss the manipulation of terms, before we provide a detailed description of resolution in first-order predicate logic.

### A.6.1 Substitution and unification

The substitution of terms for variables in formulas in order to make these formulas syntactically equal, plays a central role in a method known as *unification*. We first introduce the notion of substitution formally and then discuss its role in unification.

**Definition A.20** A substitution  $\sigma$  is a finite set of the form

$$\{t_1/x_1, \dots, t_n/x_n\}$$

where each  $x_i$  is a variable and where each  $t_i$  is a term not equal to  $x_i$ ,  $i = 1, \dots, n$ ,  $n \geq 0$ ; the variables  $x_1, \dots, x_n$  differ from each other. An element  $t_i/x_i$  of a substitution  $\sigma$  is called a binding for the variable  $x_i$ . If none of the terms  $t_i$  in a substitution contains a variable, we have a so-called ground substitution. The substitution defined by the empty set is called the empty substitution, and is denoted by  $\epsilon$ .

**Definition A.21** An expression is a term, a literal, a conjunction of literals or a disjunction of literals; a simple expression is a term or an atom.

A substitution  $\sigma$  can be applied to an expression  $E$ , yielding a new expression  $E\sigma$  which is similar to  $E$  with the difference that the variables in  $E$  occurring in  $\sigma$  have been replaced by their associated terms.

**Definition A.22** Let  $\sigma = \{t_1/x_1, \dots, t_n/x_n\}$ ,  $n \geq 0$ , be a substitution and  $E$  an expression. Then,  $E\sigma$  is an expression obtained from  $E$  by simultaneously replacing all occurrences of the variables  $x_i$  by the terms  $t_i$ .  $E\sigma$  is called an instance of  $E$ . If  $E\sigma$  does not contain any variables, then  $E\sigma$  is said to be a ground instance of  $E$ .

#### EXAMPLE A.23

---

Let  $\sigma = \{a/x, w/z\}$  be a substitution and let  $E = P(f(x, y), z)$  be an expression. Then,  $E\sigma$  is obtained by replacing each variable  $x$  in  $E$  by the constant  $a$  and each variable  $z$  by the variable  $w$ . The result of the substitution is  $E\sigma = P(f(a, y), w)$ . Note that  $E\sigma$  is not a ground instance.

---

The application of a substitution to a single expression can be extended to a set of expressions, as demonstrated in the following example.

#### EXAMPLE A.24

---

Application of the substitution  $\sigma = \{a/x, b/z\}$  to the set of expressions  $\{P(x, f(x, z)), Q(x, w)\}$  yields the following set of instances:

$$\{P(x, f(x, z)), Q(x, w)\}\sigma = \{P(a, f(a, b)), Q(a, w)\}$$

The first element of the resulting set of instances is a ground instance; the second one is not ground, since it contains the variable  $w$ .

---

**Definition A.23** Let  $\theta = \{t_1/x_1, \dots, t_m/x_m\}$  and  $\sigma = \{s_1/y_1, \dots, s_n/y_n\}$ ,  $m \geq 1$ ,  $n \geq 1$ , be substitutions. The composition of these substitutions, denoted by  $\theta\sigma$ , is obtained by removing from the set

$$\{t_1\sigma/x_1, \dots, t_m\sigma/x_m, s_1/y_1, \dots, s_n/y_n\}$$

all elements  $t_i\sigma/x_i$  for which  $x_i = t_i\sigma$ , and furthermore, all elements  $s_j/y_j$  for which  $y_j \in \{x_1, \dots, x_m\}$

The composition of substitutions is *associative*, i.e., for any expression  $E$  and substitutions  $\phi, \theta$  and  $\sigma$  we have that  $E(\phi\sigma)\theta = E\phi(\sigma\theta)$ ; the operation is not commutative. Let  $\theta = \{t_1/x_1, \dots, t_m/x_m\}$  be a substitution, and let  $V$  be the set of variables occurring in  $\{t_1, \dots, t_m\}$ , then  $\theta$  is *idempotent*, i.e.,  $E(\theta\theta) = E\theta$ , iff  $V \cap \{x_1, \dots, x_m\} = \emptyset$ .

Note that the last definition gives us a means for replacing two substitutions by a single one, being the composition of these substitutions. However, it is not always necessary to actually compute the composition of two subsequent substitutions  $\sigma$  and  $\theta$  before applying them to an expression  $E$ : it can easily be proven that  $E(\sigma\theta) = (E\sigma)\theta$ . The proof of this property is left to the reader as an exercise (see Exercise 2.11); here, we merely give an example.

---

**EXAMPLE A.25**

Consider the expression  $E = Q(x, f(y), g(z, x))$  and the two substitutions  $\sigma = \{f(y)/x, z/y\}$  and  $\theta = \{a/x, b/y, y/z\}$ . We compute the composition  $\sigma\theta$  of  $\sigma$  and  $\theta$ :  $\sigma\theta = \{f(b)/x, y/z\}$ . Application of the compound substitution  $\sigma\theta$  to  $E$  yields the instance  $E(\sigma\theta) = Q(f(b), f(y), g(y, f(b)))$ . We now compare this instance with  $(E\sigma)\theta$ . We first apply  $\sigma$  to  $E$ , resulting in  $E\sigma = Q(f(y), f(z), g(z, f(y)))$ . Subsequently, we apply  $\theta$  to  $E\sigma$  and obtain the instance  $(E\sigma)\theta = Q(f(b), f(y), g(y, f(b)))$ . So, for the given expression and substitutions, we have  $E(\sigma\theta) = (E\sigma)\theta$ .

---

In propositional logic, a resolvent of two parent clauses containing complementary literals, such as  $P$  and  $\neg P$ , was obtained by taking the disjunction of these clauses after cancelling out such a pair of complementary literals. It was easy to check for complementary literals in this case, since we only had to verify equality of the propositional atoms in the chosen literals and the presence of a negation in exactly one of them. Now, suppose that we want to compare the two literals  $\neg P(x)$  and  $P(a)$  occurring in two different clauses in first-order predicate logic. These two literals are ‘almost’ complementary. However, the first literal contains a variable as an argument of its predicate symbol, whereas the second one contains a constant. It is here where substitution comes in. Note that substitution can be applied to make expressions syntactically equal. Moreover, the substitution which is required to obtain syntactic equality of two given expressions also indicates the difference between the two. If we apply the substitution  $\{a/x\}$  to the example above, we obtain syntactic equality of the two atoms  $P(x)$  and  $P(a)$ . So, the two literals  $\neg P(x)$  and  $P(a)$  become complementary after substitution.

The *unification algorithm* is a general method for comparing expressions; the algorithm computes, if possible, the substitution that is needed to make the given expressions syntactically equal. Before we discuss the algorithm, we introduce some new notions.

**Definition A.24** A substitution  $\sigma$  is called a unifier of a given set of expressions  $\{E_1, \dots, E_m\}$  if  $E_1\sigma = \dots = E_m\sigma, m \geq 2$ . A set of expressions is called unifiable if it has a unifier.

**Definition A.25** A unifier  $\theta$  of a unifiable set of expressions  $E = \{E_1, \dots, E_m\}, m \geq 2$ , is said to be a most general unifier (mgu) if for each unifier  $\sigma$  of  $E$  there exists a substitution  $\lambda$  such that  $\sigma = \theta\lambda$ .

A set of expressions may have more than one most general unifier; however, a most general unifier is unique but for a renaming of the variables.

---

**EXAMPLE A.26**

Consider the set of expressions  $\{R(x, f(a, g(y))), R(b, f(z, w))\}$ . Some possible unifiers of this set are  $\sigma_1 = \{b/x, a/z, g(c)/w, c/y\}$ ,  $\sigma_2 = \{b/x, a/z, f(a)/y, g(f(a))/w\}$  and  $\sigma_3 = \{b/x, a/z, g(y)/w\}$ . The last unifier is also a most general unifier: by the composition of this unifier with the substitution  $\{c/y\}$  we get  $\sigma_1$ ; the second unifier is obtained by the composition of  $\sigma_3$  with  $\{f(a)/y\}$ .

---

The unification algorithm, more precisely, is a method for constructing a most general unifier of a finite, non-empty set of expressions. The algorithm considered in this book operates in the following manner. First, the left-most subexpressions in which the given expressions differ is computed. Their difference is placed in a set, called the *disagreement set*. Based on this disagreement set a ('most general') substitution is computed, which is subsequently applied to the given expressions, yielding a partial or total equality. If no such substitution exists, the algorithm terminates with the message that the expressions are not unifiable. Otherwise, the procedure proceeds until each element within each of the expressions has been processed. It can be proven that the algorithm either terminates with a failure message or with a most general unifier of the finite, unifiable set of expressions.

---

**EXAMPLE A.27**

Consider the following set of expressions:

$$S = \{Q(x, f(a), y), Q(x, z, c), Q(x, f(a), c)\}$$

The left-most subexpression in which the three expressions differ is in the second argument of the predicate symbol  $Q$ . So, the first disagreement set is  $\{f(a), z\}$ . By means of the substitution  $\{f(a)/z\}$  the subexpressions in the second argument position are made equal. The next disagreement set is  $\{y, c\}$ . By means of the substitution  $\{c/y\}$  these subexpressions are also equalized. The final result returned by the unification algorithm is the unifier  $\{f(a)/z, c/y\}$  of  $S$ . It can easily be seen that this unifier is a most general one.

---

The following section shows an implementation of the unification algorithm. In the next section we discuss the role of unification in resolution in first-order predicate logic.

### A.6.2 Resolution

Now that we have dealt with the subjects of substitution and unification, we are ready for a discussion of resolution in first-order predicate logic. We start with an informal introduction to the subject by means of an example.

---

#### EXAMPLE A.28

Consider the following set of clauses:

$$\{C_1 = P(x) \vee Q(x), C_2 = \neg P(f(y)) \vee R(y)\}$$

As can be seen, the clauses  $C_1$  and  $C_2$  do not contain complementary literals. However, the atoms  $P(x)$ , occurring in  $C_1$ , and  $P(f(y))$ , occurring in the literal  $\neg P(f(y))$  in the clause  $C_2$ , are unifiable. For example, if we apply the substitution  $\sigma = \{f(a)/x, a/y\}$  to  $\{C_1, C_2\}$ , we obtain the following set of instances:

$$\{C_1\sigma = P(f(a)) \vee Q(f(a)), C_2\sigma = \neg P(f(a)) \vee R(a)\}$$

The resulting instances  $C_1\sigma$  and  $C_2\sigma$  do contain complementary literals, namely  $P(f(a))$  and  $\neg P(f(a))$  respectively. As a consequence, we are now able to find a resolvent of  $C_1\sigma$  and  $C_2\sigma$ , being the clause

$$C'_3 = Q(f(a)) \vee R(a)$$

---

The resolution principle in first-order predicate logic makes use of the unification algorithm for constructing a most general unifier of two suitable atoms; the subsequent application of the resulting substitution to the literals containing the atoms, renders them complementary. In the preceding example, the atoms  $P(x)$  and  $P(f(y))$  have a most general unifier  $\theta = \{f(y)/x\}$ . The resolvent obtained after applying  $\theta$  to  $C_1$  and  $C_2$ , is

$$C_3 = Q(f(y)) \vee R(y)$$

The clause  $C'_3$  from the previous example is an instance of  $C_3$ , the so-called *most general clause*: if we apply the substitution  $\{a/y\}$  to  $C_3$ , we obtain the clause  $C'_3$ .

It should be noted that it is necessary to rename different variables having the same name in both parent clauses before applying resolution, since the version of the unification algorithm discussed in the previous section is not capable of distinguishing between equally named variables actually being the same variable, and equally named variables being different variables because of their occurrence in different clauses.

---

#### EXAMPLE A.29

Consider the atoms  $Q(x, y)$  and  $Q(x, f(y))$  occurring in two different clauses. In this form our unification algorithm reports failure in unifying these atoms (due to the occur check). We rename the variables  $x$  and  $y$  in  $Q(x, f(y))$  to  $u$  and  $v$  respectively, thus obtaining the atom  $Q(u, f(v))$ . Now, if we apply the unification algorithm again to compute a most general unifier of  $\{Q(u, f(v)), Q(x, y)\}$ , it will come up with the (correct) substitution  $\sigma = \{u/x, f(v)/y\}$ .



We already mentioned in Section 2.3 that the meaning of a formula is left unchanged by renaming variables. We furthermore recall that formulas only differing in the names of their (bound) variables are called variants.

From the examples presented so far, it should be clear by now that resolution in first-order predicate logic is quite similar to resolution in propositional logic: literals are cancelled out from clauses, thus generating new clauses. From now on, cancelling out a literal  $L$  from a clause  $C$  will be denoted by  $C \setminus L$ .

**Definition A.26** Consider the parent clauses  $C_1$  and  $C_2$ , respectively containing the literals  $L_1$  and  $L_2$ . If  $L_1$  and  $\neg L_2$  have a most general unifier  $\sigma$ , then the clause  $(C_1 \sigma \setminus L_1 \sigma) \vee (C_2 \sigma \setminus L_2 \sigma)$  is called a binary resolvent of  $C_1$  and  $C_2$ . Resolution in which each resolvent is a binary resolvent, is known as binary resolution.

A pair of clauses may have more than one resolvent, since they may contain more than one pair of complementary literals. Moreover, not every resolvent is necessarily a binary resolvent: there are more general ways for obtaining a resolvent. Before giving a more general definition of a resolvent, we introduce the notion of a factor.

**Definition A.27** If two or more literals in a clause  $C$  have a most general unifier  $\sigma$ , then the clause  $C\sigma$  is said to be a factor of  $C$ .

---

**EXAMPLE A.30**

Consider the following clause:

$$C = P(g(x), h(y)) \vee Q(z) \vee P(w, h(a))$$

The literals  $P(g(x), h(y))$  and  $P(w, h(a))$  in  $C$  have a most general unifier  $\sigma = \{g(x)/w, a/y\}$ . So,

$$C\sigma = P(g(x), h(a)) \vee Q(z) \vee P(g(x), h(a)) = P(g(x), h(a)) \vee Q(z)$$

is a factor of  $C$ . Note that one duplicate literal  $P(g(x), h(a))$  has been removed from  $C\sigma$ .

---

The generalized form of resolution makes it possible to cancel out more than one literal from one or both of the parent clauses by first computing a factor of one or both of these clauses.

---

**EXAMPLE A.31**

Consider the following set of clauses:

$$\{C_1 = P(x) \vee P(f(y)) \vee R(y), C_2 = \neg P(f(a)) \vee \neg R(g(z))\}$$

In the clause  $C_1$  the two literals  $P(x)$  and  $P(f(y))$  have a most general unifier  $\sigma = \{f(y)/x\}$ . If we apply this substitution  $\sigma$  to the clause  $C_1$ , then one of these literals can be removed:

$$\begin{aligned} (P(x) \vee P(f(y)) \vee R(y))\sigma &= P(f(y)) \vee P(f(y)) \vee R(y) \\ &= P(f(y)) \vee R(y) \end{aligned}$$

The result is a factor of  $C_1$ . The literal  $P(f(y))$  in  $C_1\sigma$  can now be unified with the atom  $P(f(a))$  in the literal  $\neg P(f(a))$  occurring in  $C_2$ , using the substitution  $\{a/y\}$ . We obtain the resolvent

$$C_3 = R(a) \vee \neg R(g(z))$$

Note that a total of three literals has been removed from  $C_1$  and  $C_2$ . The reader can easily verify that there are several other resolvents from the same parent clauses:

- By taking  $L_1 = P(x)$  and  $L_2 = \neg P(f(a))$  we get the resolvent  $P(f(y)) \vee R(y) \vee \neg R(g(z))$ ;
- Taking  $L_1 = P(f(y))$  and  $L_2 = \neg P(f(a))$  results in the resolvent  $P(x) \vee R(a) \vee \neg R(g(z))$ ;
- By taking  $L_1 = R(y)$  and  $L_2 = \neg R(g(z))$  we obtain  $P(x) \vee P(f(g(z))) \vee \neg P(f(a))$ .

We now give the generalized definition of a resolvent in which the notion of a factor is incorporated.

**Definition A.28** *A resolvent of the parent clauses  $C_1$  and  $C_2$  is one of the following binary resolvents:*

- (1) *A binary resolvent of  $C_1$  and  $C_2$ ;*
- (2) *A binary resolvent of  $C_1$  and a factor of  $C_2$ ;*
- (3) *A binary resolvent of a factor of  $C_1$  and  $C_2$ ;*
- (4) *A binary resolvent of a factor of  $C_1$  and a factor of  $C_2$ .*

The most frequent application of resolution is refutation: the derivation of the empty clause  $\square$  from a given set of clauses. The following procedure gives the general outline of this resolution algorithm.

```

procedure Resolution( $S$ )
  clauses  $\leftarrow S$ ;
  while  $\square \notin$  clauses do
     $\{c_i, c_j\} \leftarrow$  SelectResolvable(clauses);
    resolvent  $\leftarrow$  Resolve( $c_i, c_j$ );
    clauses  $\leftarrow$  clauses  $\cup$  {resolvent}
  od
end

```

This algorithm is non-deterministic. The selection of parent clauses  $c_i$  and  $c_j$  can be done in many ways; how it is to be done has not been specified in the algorithm. Several different strategies have been described in the literature, each of them prescribing an unambiguous way of choosing parent clauses from the clause set. Such strategies are called the *control strategies* of resolution or *resolution strategies*. Several of these resolution strategies offer

particularly efficient algorithms for making computer-based theorem proving feasible. Some well-known strategies are: *semantic resolution*, which was developed by J.R. Slagle in 1967, *hyperresolution* developed by J.A. Robinson in 1965, and various forms of *linear resolution*, such as *SLD resolution*, in the development of which R.A. Kowalski played an eminent role. At present, SLD resolution in particular is a strategy of major interest, because of its relation to the programming language Prolog.

## A.7 Resolution strategies

Most of the basic principles of resolution have been discussed in the previous section. However, one particular matter, namely the efficiency of the resolution algorithm, has not explicitly been dealt with as yet. It is needless to say that the subject of efficiency is an important one for automated reasoning.

Unfortunately, the general refutation procedure introduced in Section A.6.3 is quite inefficient, since in many cases it will generate a large number of redundant clauses, that is, clauses not contributing to the derivation of the empty clause.

### EXAMPLE A.32

---

Consider the following set of clauses:

$$S = \{P, \neg P \vee Q, \neg P \vee \neg Q \vee R, \neg R\}$$

To simplify referring to them, the clauses are numbered as follows:

- (1)  $P$
- (2)  $\neg P \vee Q$
- (3)  $\neg P \vee \neg Q \vee R$
- (4)  $\neg R$

If we apply the resolution principle by systematically generating all resolvents, without utilizing a more specific strategy in choosing parent clauses, the following resolvents are successively added to  $S$ :

- (5)  $Q$  (using 1 and 2)
- (6)  $\neg Q \vee R$  (using 1 and 3)
- (7)  $\neg P \vee R$  (using 2 and 3)
- (8)  $\neg P \vee \neg Q$  (using 3 and 4)
- (9)  $R$  (using 1 and 7)
- (10)  $\neg Q$  (using 1 and 8)
- (11)  $\neg P \vee R$  (using 2 and 6)
- (12)  $\neg P$  (using 2 and 8)
- (13)  $\neg P \vee R$  (using 3 and 5)
- (14)  $\neg Q$  (using 4 and 6)
- (15)  $\neg P$  (using 4 and 7)

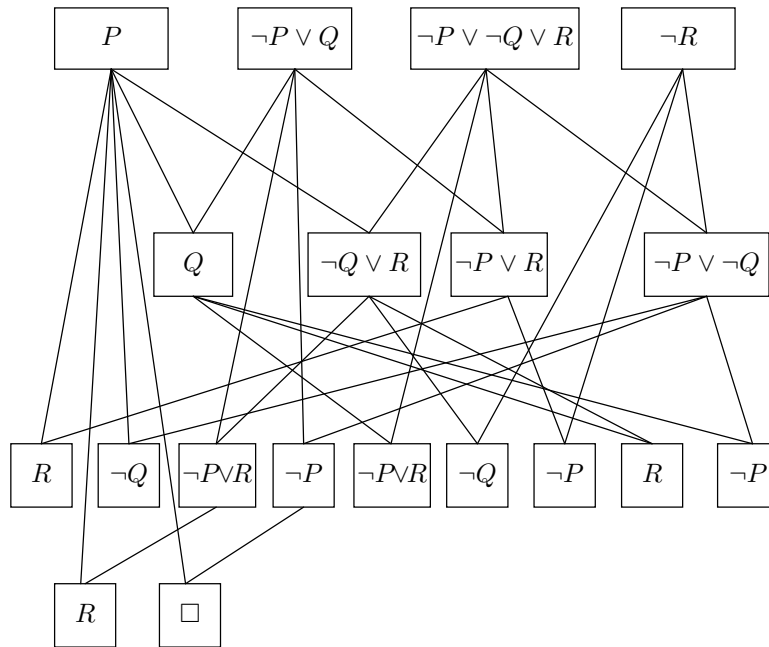


Figure A.3: Refutation of  $\{P, \neg P \vee Q, \neg P \vee \neg Q \vee R, \neg R\}$ .

- (16)  $R$  (using 5 and 6)
- (17)  $\neg P$  (using 5 and 8)
- (18)  $R$  (using 1 and 11)
- (19)  $\square$  (using 1 and 12)

This derivation of the empty clause  $\square$  from  $S$  has been depicted in Figure A.3 by means of a derivation graph. As can be seen, by systematically generating all resolvents in a straightforward manner, fifteen of them were obtained, while, for instance, taking the two resolvents

- (5')  $\neg P \vee R$  (using 2 and 3)
- (6')  $R$  (using 1 and 5')

would lead directly to the derivation of the empty clause:

- (7')  $\square$  (using 4 and 6')

In the latter refutation, significantly less resolvents were generated.

The main goal of applying a resolution strategy is to restrict the number of redundant clauses generated in the process of resolution. This improvement in efficiency is achieved by incorporating particular algorithmic refinements in the resolution principle.

## A.8 Applying logic for building intelligent systems

In the preceding sections, much space has been devoted to the many technical details of knowledge representation and automated reasoning using logic. In the present section, we

shall indicate how logic can actually be used for building a logic-based intelligent system. As logic offers suitable basis for model-based systems, more about the use of logic in the context of intelligent systems will be said in Chapter 6.

In the foregoing, we have seen that propositional logic offers rather limited expressiveness, which in fact is too limited for most real-life applications. First-order predicate logic offers much more expressive power, but that alone does not yet render the formalism suitable for building intelligent systems. There are some problems: any automated reasoning method for full first-order logic is doomed to have a worst-case time complexity at least as bad as that of checking satisfiability in propositional logic, which is known to be NP-complete (this means that no one has been able to come up with a better deterministic algorithm than an exponentially time-bounded one, although it has not been proven that better ones do not exist). Furthermore, we know that first-order predicate logic is undecidable; so, it is not even sure that an algorithm for checking satisfiability will actually terminate. Fortunately, the circumstances are not always as bad as that. A worst-case characterization seldom gives a realistic indication of the time an algorithm generally will spend on solving an arbitrary problem. Moreover, several suitable syntactic restrictions on first-order formulas have been formulated from which a substantial improvement of the time complexity of the algorithm is obtained; the Horn clause format we have paid attention to is one such restriction.

Since syntactic restrictions are only acceptable as far as permitted by a problem domain, we consider some examples, such as the logical circuit introduced in Chapter 1. However, first we discuss special standard predicates that are often used in modelling practical applications.

### A.8.1 Reasoning with equality and ordering predicates

The special binary predicate symbols  $>$  (*ordering predicate*) and  $=$  (*equality predicate*) are normally specified in infix position, since this is normal mathematical practice. Both equality and the ordering predicates have a special meaning, which is described by means of a collection of axioms. The meaning of the equality predicate is defined by means of the following four axioms:

$$E_1 \text{ (reflexivity): } \forall x(x = x)$$

$$E_2 \text{ (symmetry): } \forall x\forall y(x = y \rightarrow y = x)$$

$$E_3 \text{ (transitivity): } \forall x\forall y\forall z(x = y \wedge y = z \rightarrow x = z)$$

$$E_4 \text{ (substitutivity): } \forall x_1 \dots \forall x_n \forall y_1 \dots \forall y_n ((x_1 = y_1 \wedge \dots \wedge x_n = y_n) \rightarrow f(x_1, \dots, x_n) = f(y_1, \dots, y_n)), \text{ and } \forall x_1 \dots \forall x_n \forall y_1 \dots \forall y_n ((x_1 = y_1 \wedge \dots \wedge x_n = y_n \wedge P(x_1, \dots, x_n)) \rightarrow P(y_1, \dots, y_n))$$

Axiom  $E_1$  states that each term in the domain of discourse is equal to itself; axiom  $E_2$  expresses that the order of the arguments of the equality predicate is irrelevant. Axiom  $E_3$  furthermore states that two terms which are equal to some common term, are equal to each other. Note that axiom  $E_2$  follows from the axioms  $E_1$  and  $E_3$ ; nevertheless, it is usually mentioned explicitly. The three axioms  $E_1$ ,  $E_2$  and  $E_3$  together imply that equality is an *equivalence relation*. Addition of axiom  $E_4$  renders it a *congruence relation*. The first part of axiom  $E_4$  states that equality is preserved under the application of a function; the second part expresses that equal terms may be substituted for each other in formulas.

#### EXAMPLE A.33

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Consider the following set of clauses  $S$ :

$$S = \{\neg P(f(x), y) \vee Q(x, x), P(f(a), a), a = b\}$$

Suppose that, in addition, we have the equality axioms. If we add the clause  $\neg Q(b, b)$  to  $S$ , the resulting set of clauses will be unsatisfiable. This can easily be seen informally as follows: we have  $P(f(a), a) \equiv P(f(b), a)$  using the given clause  $a = b$  and the equality axiom  $E_4$ . Now, we replace the atom  $P(f(a), a)$  by the equivalent atom  $P(f(b), a)$  and apply binary resolution.

The explicit addition of the equality axioms to the other formulas in a knowledge base suffices for rendering equality available for use in an intelligent system. However, it is well known that proving theorems in the presence of the equality axioms can be very inefficient, since many redundant clauses may be generated using resolution. Again, several refinements of the (extended) resolution principle have been developed to overcome the inefficiency problem. For dealing with equality, the resolution principle has for example been extended with an extra inference rule: *paramodulation*. Informally speaking, the principle of paramodulation is the following: if clause  $C$  contains a term  $t$  and if we have a clause  $t = s$ , then derive a clause by substituting  $s$  for a single occurrence of  $t$  in  $C$ . Therefore, in practical realizations equality is often only present implicitly in the knowledge base, that is, it is used as a ‘built-in’ predicate.

Another, more restrictive way to deal with equality is *demodulation*, which adds directionality to equality, meaning that one side of the equality may be replaced (rewritten) by the other side, but not the other way around.

**Definition A.29** A demodulator is a positive unit clause with equality predicate of the form  $(l = r)$ , where  $l$  and  $r$  are terms. Let  $C \vee L^t$  a clause; where  $L^t$  indicates that the literal  $L$  contains the term  $t$ . Let  $\sigma$  be a substitution such that  $l\sigma = t$ , then demodulation is defined as the inference rule:

$$\frac{C \vee L^t, (l = r)}{C \vee L^{t \rightarrow r\sigma}}$$

where  $L^{t \rightarrow r\sigma}$  indicates that term  $t$  is rewritten to  $r\sigma$ .

Thus a demodulator  $(l = r)$  can be interpreted as a rewrite rule  $l \rightarrow r$  with a particular orientation (here from left to right).

#### EXAMPLE A.34

Consider the demodulator

$$(brother(father(x)) = uncle(x))$$

and the clause

$$(age(brother(father(John))) = 55)$$

Application of the demodulator yields

$$(age(uncle(John)) = 55)$$

The applied substitution was  $\sigma = \{Jan/x\}$ .

In many real-life applications, a universally quantified variable ranges over a finite domain  $D = \{c_i \mid i = 1, \dots, n, n \geq 0\}$ . The following property usually is satisfied:  $\forall x(x = c_1 \vee x = c_2 \vee \dots \vee x = c_n)$ , with  $c_i \neq c_j$  if  $i \neq j$ . This property is known as the *unique names assumption*; from this assumption we have that objects with different names are different.

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**EXAMPLE A.35**


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Consider the following set of clauses  $S$ :

$$S = \{\neg P(x) \vee x = a\}$$

We suppose that the equality axioms as well as the unique name assumption hold. Now, if we add the clause  $P(b)$  to  $S$ , we obtain an inconsistency, since the derivable clause  $b = a$  contradicts with the unique name assumption.

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The ordering predicates  $<$  and  $>$  define a *total order* on the set of real numbers. They express the usual, mathematical ‘less than’ and ‘greater than’ binary relations between real numbers. Their meaning is defined by means of the following axioms:

$$O_1 \text{ (irreflexivity): } \forall x \neg(x < x)$$

$$O_2 \text{ (antisymmetry): } \forall x \forall y (x < y \rightarrow \neg(y < x))$$

$$O_3 \text{ (transitivity): } \forall x \forall y \forall z ((x < y \wedge y < z) \rightarrow x < z)$$

$$O_4 \text{ (trichotomy law): } \forall x \forall y ((x < y \vee x = y \vee x > y))$$

Axiom  $O_1$  states that no term is less than itself; axiom  $O_2$  expresses that reversing the order of the arguments of the predicate  $<$  reverses the meaning. Axiom  $O_3$  furthermore states that if a term is less than some other term, and this term is less than a third term, then the first term is less than the third one as well. Note that axiom  $O_2$  follows from  $O_1$  and  $O_3$ . The axioms  $O_1$ ,  $O_2$  and  $O_3$  concern the ordering predicate  $<$ . The axioms for the ordering predicate  $>$  are similar to these: we may just substitute  $>$  for  $<$  to obtain them. Axiom  $O_4$  states that a given term is either less than, equal to or greater than another given term. Again, in practical realizations, these axioms usually are not added explicitly to the knowledge base, but are assumed to be present implicitly as ‘built-in’ predicates or as evaluable predicates, that after it has been checked that all variables are instantiated to constants, are evaluated as an expression, returning true or false.

## Exercises

- (A.1) Consider the interpretation  $v : PROP \rightarrow \{true, false\}$  in propositional logic, which is defined by  $v(P) = false$ ,  $v(Q) = true$  and  $v(R) = true$ . What is the truth value of the formula  $((\neg P) \wedge Q) \vee (P \rightarrow (Q \vee R))$  given this interpretation  $v$ ?
- (A.2) For each of the following formulas in propositional logic determine whether it is valid, invalid, satisfiable, unsatisfiable or a combination of these, using truth tables:

Table A.6: Meaning of Sheffer stroke.

$F$	$G$	$F G$
<i>true</i>	<i>true</i>	<i>false</i>
<i>true</i>	<i>false</i>	<i>true</i>
<i>false</i>	<i>true</i>	<i>true</i>
<i>false</i>	<i>false</i>	<i>true</i>

- (a)  $P \vee (Q \rightarrow \neg P)$
- (b)  $P \vee (\neg P \wedge Q \wedge R)$
- (c)  $P \rightarrow \neg P$
- (d)  $(P \wedge \neg Q) \wedge (\neg P \vee Q)$
- (e)  $(P \rightarrow Q) \rightarrow (Q \rightarrow P)$
- (A.3) Suppose that  $F_1, \dots, F_n$ ,  $n \geq 1$ , and  $G$  are formulas in propositional logic, such that the formula  $G$  is a logical consequence of  $\{F_1, \dots, F_n\}$ . Construct the truth table of the implication  $F_1 \wedge \dots \wedge F_n \rightarrow G$ . What do you call such a formula?
- (A.4) Prove the following statements using the laws of equivalence for propositional logic:
- (a)  $P \rightarrow Q \equiv \neg P \rightarrow \neg Q$
- (b)  $P \rightarrow (Q \rightarrow R) \equiv (P \wedge Q) \rightarrow R$
- (c)  $(P \wedge \neg Q) \rightarrow R \equiv (P \wedge \neg R) \rightarrow Q$
- (d)  $P \vee (\neg Q \vee R) \equiv (\neg P \wedge Q) \rightarrow R$
- (A.5) Prove that the proposition  $((P \rightarrow Q) \rightarrow P) \rightarrow P$ , known as Peirce's law, is a tautology, using the laws of equivalence in propositional logic and the property that for any propositions  $\pi$  and  $\phi$ , the formula  $\pi \vee \neg\phi \vee \phi$  is a tautology.
- (A.6) In each of the following cases, we restrict ourselves to a form of propositional logic only offering a limited set of logical connectives. Prove by means of the laws of equivalence that every formula in full propositional logic can be translated into a formula only containing the given connectives:
- (a) the connectives  $\neg$  and  $\vee$ .
- (b) the connective  $|$  which is known as the Sheffer stroke; its meaning is defined by the truth table given in Table A.6.
- (A.7) Consider the following formula in first-order predicate logic:  $\forall x(P(x) \vee Q(y))$ . Suppose that the following structure

$$S = (\{2, 3\}, \emptyset, \{A : \{2, 3\} \rightarrow \{true, false\}, B : \{2, 3\} \rightarrow \{true, false\}\})$$

is given. The predicates  $A$  and  $B$  are associated with the predicate symbols  $P$  and  $Q$ , respectively. Now, define the predicates  $A$  and  $B$ , and a valuation  $v$  in such a way that the given formula is satisfied in the given structure  $S$  and valuation  $v$ .



(A.8) Consider the following statements. If a statement is correct, then prove its correctness using the laws of equivalence; if it is not correct, then give a counterexample.

- (a)  $\forall xP(x) \equiv \neg\exists x\neg P(x)$
- (b)  $\forall x\exists yP(x, y) \equiv \forall y\exists xP(x, y)$
- (c)  $\exists x(P(x) \rightarrow Q(x)) \equiv \forall xP(x) \rightarrow \exists xQ(x)$
- (d)  $\forall x(P(x) \vee Q(x)) \equiv \forall xP(x) \vee \forall xQ(x)$

(A.9) Transform the following formulas into the clausal form of logic:

- (a)  $\forall x\forall y\exists z(P(z, y) \wedge (\neg P(x, z) \rightarrow Q(x, y)))$
- (b)  $\exists x(P(x) \rightarrow Q(x)) \wedge \forall x(Q(x) \rightarrow R(x)) \wedge P(a)$
- (c)  $\forall x(\exists y(P(y) \wedge R(x, y)) \rightarrow \exists y(Q(y) \wedge R(x, y)))$

(A.10) For each of the following sets of clauses, determine whether or not it is satisfiable. If a given set is unsatisfiable, then give a refutation of the set using binary resolution; otherwise give an interpretation satisfying it:

- (a)  $\{\neg P \vee Q, P \vee \neg R, \neg Q, \neg R\}$
- (b)  $\{\neg P \vee Q \vee R, \neg Q \vee S, P \vee S, \neg R, \neg S\}$
- (c)  $\{P \vee Q, \neg P \vee Q, P \vee \neg Q, \neg P \vee \neg Q\}$
- (d)  $\{P \vee \neg Q, Q \vee R \vee \neg P, Q \vee P, \neg P\}$

(A.11) Let  $E$  be an expression and let  $\sigma$  and  $\theta$  be substitutions. Prove that  $E(\sigma\theta) = (E\sigma)\theta$ .

(A.12) For each of the following sets of expressions, determine whether or not it is unifiable. If a given set is unifiable, then compute a most general unifier:

- (a)  $\{P(a, x, f(x)), P(x, y, x)\}$
- (b)  $\{P(x, f(y), y), P(w, z, g(a, b))\}$
- (c)  $\{P(x, z, y), P(x, z, x), P(a, x, x)\}$
- (d)  $\{P(z, f(x), b), P(x, f(a), b), P(g(x), f(a), y)\}$

(A.13) Use binary resolution to show that each one of the following sets of clauses is unsatisfiable:

- (a)  $\{P(x, y) \vee Q(a, f(y)) \vee P(a, g(z)), \neg P(a, g(x)) \vee Q(a, f(g(b))), \neg Q(x, y)\}$
- (b)  $\{\text{append}(\text{nil}, x, x), \text{append}(\text{cons}(x, y), z, \text{cons}(x, u)) \vee \neg \text{append}(y, z, u), \neg \text{append}(\text{cons}(1, \text{cons}(2, \text{nil})), \text{cons}(3, \text{nil}), x)\}$
- (c)  $\{R(x, x), R(x, y) \vee \neg R(y, x), R(x, y) \vee \neg R(x, z) \vee \neg R(z, y), R(a, b), \neg R(b, a)\}$

*Remark.* The first three clauses in exercise (c) define an equivalence relation.

(A.14) Consider the set of clauses  $\{\neg P, P \vee Q, \neg Q, R\}$ . We employ the set-of-support resolution strategy. Why do we not achieve a refutation if we set the set of support initially to the clause  $R$ ?

(A.15) Develop a logic knowledge base for a problem domain you are familiar with.