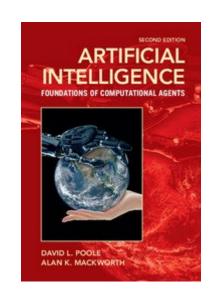
Chapter 5 **Propositions and Inference**

Textbook: Artificial Intelligence Foundations of Computational Agents, David L. Poole and Alan K Mackworth, Cambridge University Press.



Introduction

- This chapter presents several reasoning formalisms that use propositions.
- Here, a simple form of a knowledge base (KB) is presented, composed of facts and rules related to a problem world.
 - An agent can use such a KB, together with its observations, to determine the *truth* in the world.
- When queried about "what must be true given a KB," it answers the query without enumerating all the possible worlds.

Propositions

- Statements about the world provide constraints about what could be true
 - Constraints could be specified extensionally as tables of legal assignments to *variables* or *formulas*.
- There are several reasons for using <u>propositions</u> for specifying <u>constraints</u> and <u>queries</u>:
 - It gives a more concise and readable logical statement about the relationship among variables.
 - The form of knowledge can be exploited to make reasoning more efficient.
 - They are modular, so small changes to the problem result in minor changes to the KB.

Syntax of Propositional Calculus

- A proposition is a sentence written in a language with a truth value (either true or false) in a world.
 - A proposition is built from atomic propositions using logical connectives.
 - We use the convention that propositions consist of letters, digits, and the underscore ('_') and start with a lowercase letter.
 - For example, ai_is_fun, live_outside, and sunny are all propositional atoms.
- Propositions can be built from more straightforward propositions using logical connectives
 - A proposition or logical formula is either an atomic proposition or a compound proposition

Syntax of Propositional Calculus

• a **compound proposition** of the form

```
\neg p "not p"

p \land q "p \text{ and } q"

p \lor q "p \text{ or } q"

p \rightarrow q "p \text{ implies } q"

p \leftarrow q "p \text{ if } q"

p \leftrightarrow q "p \text{ if and only if } q"
```

negation of *p* **conjunction** of *p* and *q* **disjunction** of *p* and *q* **implication** of *q* from *p* **implication** of *p* from *q* **equivalence** of *p* and *q*

where p and q are propositions.

The operators \neg , \land , \lor , \rightarrow , \leftarrow and \leftrightarrow are **logical connectives**.

Syntax of Propositional Calculus

p	q	$\neg p$	$p \wedge q$	$p \vee q$	$p \leftarrow q$	$p \rightarrow q$	$p \leftrightarrow q$
true	true	false	true	true	true	true	true
true	false	false	false	true	true	false	false
false	true	true	false	true	false	true	false
false	false	true	false	false	true	true	true

Figure 5.1: Truth table defining \neg , \land , \lor , \leftarrow , \rightarrow , and \leftrightarrow

Semantics of the Propositional Calculus

- Semantics defines the meaning of the sentences of a language.
- When the sentences are about a world, semantics specifies how to put symbols of the language into correspondence with the world.
- The **truth of atoms** gives the **truth** of other propositions in interpretations (see **Figure 5.1**).
- An interpretation consists of a function π that maps atoms to {true, false}.
 - If $\pi(a) = true$, atom a is true
 - If $\pi(a)$ = false, atom a is false
 - Suppose there are three atoms: ai_is_fun , happy, and $light_on$; $\pi(ai_is_fun) = true$, $\pi(happy) = false$, $\pi(light_on) = true$

Semantics of the Propositional Calculus

Example 5.20 Consider the knowledge base KB_2 :

false
$$\leftarrow$$
 $a \land b$.
 $a \leftarrow c$.
 $b \leftarrow d$.
 $b \leftarrow e$.

Either c is false or d is false in every model of KB_2 . If they were both true in some model I of KB_2 , both a and b would be true in I, so the first clause would be false in I, a contradiction to I being a model of KB_2 . Similarly, either c is false or e is false in every model of KB_2 . Thus,

$$KB_2 \models \neg c \lor \neg d$$

 $KB_2 \models \neg c \lor \neg e$.

Semantics of the Propositional Calculus

- A KB is a set of true propositions.
- An element of the KB is an axiom.
- A model of a KB is an interpretation in which all the propositions in KB are true.
- If a knowledge base is KB, and g is a proposition, KB |= g is g logically follows from KB, or KB entails g.
- The logical entailment "**KB** |= g" is a semantic relation between a set of propositions in **KB** and an external proposition, g.
- Both KB and g are symbolic so that they can be represented in a computer.

Propositional Constraints

- The class of propositional satisfiability problems has:
 - Boolean variables: If X is a Boolean variable, X = false as ¬x.
 Thus, given a Boolean variable Happy, the proposition happy means Happy = true, and ¬happy means Happy = false.
 - <u>Clausal constraints</u>: a clause is disjoints of atoms and is expressed as $(I_1 \lor I_2 \lor \ldots \lor I_k)$, where each I_i is literal.
 - A literal is an atom or the negation of an atom
 - A clause is satisfied in a possible world if at least one of the literals that make up the clause is true in that possible world.

Propositional Constraints

Example 5.4 The clause $happy \lor sad \lor \neg living$ is a constraint among the variables Happy, Sad, and Living, which is true if Happy has value true, Sad has value true, or Living has value false. The atoms happy and sad appear positively in the clause, and living appears negatively in the clause.

The assignment $\neg happy$, $\neg sad$, living violates the constraint of clause $happy \lor sad \lor \neg living$. It is the only assignment of these three variables that violates this clause.

Propositional Constraints

- It is possible to convert any finite CSP into a Propositional Satisfiable Problem (PSP):
 - A CSP variable Y with domain $\{v_1, \ldots, v_k\}$ can be converted into k Boolean variables $\{Y_1, \ldots, Y_k\}$, where Y_i is true when Y has value v_i and is false otherwise. Each Y_i is called an indicator variable. Thus k atoms, y_1, \ldots, y_k , are used to represent the CSP variable.
 - There are **constraints** that specify that y_i and y_j cannot both be **true** when $(i \neq j)$. There is a **constraint** that one of the y_i must be **true**. Thus, the **KB** contains the **clauses**: $\neg y_i \lor \neg y_j$ for i < j and $y_1 \lor \neg \neg v_j$.
 - There is a clause for each false assignment in each constraint; for example, the clauses (a ∨ b ∨ c) and (a ∨ b ∨ ¬c) can be combined with (a ∨ b).

Clausal Form for Consistency Algorithms

- Consistency algorithms can be made more efficient for propositional satisfiability problems (PSP)
 - When there are only two values, pruning a value from the domain is equivalent to assigning the opposite value
 - Thus, if X has domain {true, false}, pruning true from the domain of X is the same as assigning X to have the value false
- Arc consistency can be used to prune the set of values and the set of constraints.

Clausal Form for Consistency Algorithms

- Assigning a value to a Boolean variable can simplify the set of constraints:
 - If X is assigned true, all of the clauses with X = true become redundant; they are automatically satisfied. These clauses can be removed. Similarly, if X is assigned false, clauses containing X = false can be removed.
 - If X is assigned true, any clause with X = false can be simplified by removing X = false from the clause. Similarly, if X is assigned false, then X = true can be removed from any clause it appears in. This step is called unit resolution.
 - After pruning the clauses, there is a clause that contains just one assignment, Y = v, the other value can be removed from the domain of Y. This is a form of arc consistency
 - If all of the assignments are removed from a clause, the constraints are unsatisfiable.

Clausal Form for Consistency Algorithms

- If a variable has the same value in all remaining clauses, and the algorithm must only find one model, it can assign that value to that variable
 - For example, if variable Y only appears as Y = true (i.e., ¬y is not in any clause), then Y can be assigned the value true.
 - That assignment does not remove all of the models;
 - A variable that has only one value in all of the clauses is called a pure literal.

- The syntax of definite propositional clauses is defined as follows:
 - An atomic proposition or atom is the same as in propositional calculus.
 - A <u>definite clause</u> is of the form: $a \leftarrow b_1 \land \ldots \land b_m$, where a is the head of the clause and is always positive, and a and each b_i are called atoms. It can be read as "a if b_1 and b_2 ... and b_m "
 - $(a \leftarrow (b_1 \land b_2 \land \dots \land b_m)) = (a \lor \neg b_1 \lor \neg b_2 \lor \dots \lor \neg b_m)$
 - If m > 0, the clause is called a rule, where $b_1 \wedge \ldots \wedge b_m$ is the **body** of the clause ("if" part of the rule).
 - If m = 0, the arrow can be omitted, and the clause is called an atomic clause or a <u>fact</u> (it is the clause with an empty body).
 - A knowledge base (KB) is a set of definite clauses.

Example 5.6 The elements of the knowledge base in Example 5.20 are all definite clauses.

The following are *not* definite clauses:

```
\neg apple\_is\_eaten.

apple\_is\_eaten \land bird\_eats\_apple.

sam\_is\_in\_room \land night\_time \leftarrow switch\_1\_is\_up.

Apple\_is\_eaten \leftarrow Bird\_eats\_apple.

happy \lor sad \lor \neg alive.
```

The fourth statement is not a definite clause because an atom must start with a lower-case letter.

 Note that a definite clause is a restricted form of a clause. For example, the definite clause

```
(a \leftarrow b \land c \land d) is equivalent to the clause (a \lor \neg b \lor \neg c \lor \neg d). That is, (a \leftarrow b \land c \land d) = (a \lor \neg b \lor \neg c \lor \neg d)
```

 In general, a definite clause is equivalent to a clause with precisely one positive literal.

• **Example 5.7** Consider how to axiomatize the electrical environment of **Figure 5.2** following the methodology for the user's view of semantics.

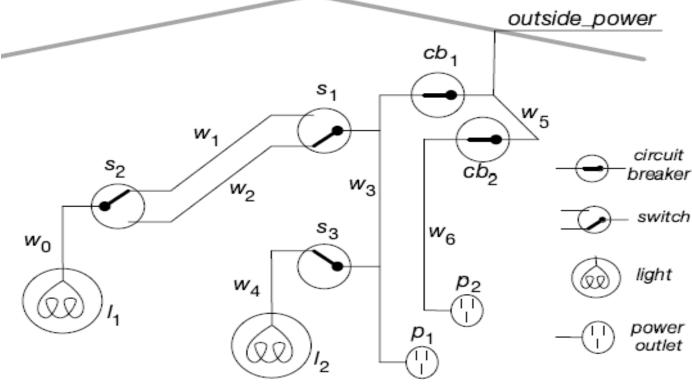


Figure 5.2: An electrical environment with components named

- The designer may look at part of the domain and know that light I₁ is live if wire w₀ is live because they are connected but may not know whether w₀ is live. Such knowledge is expressible in terms of rules.
- The KB consists of all of the <u>definite clauses</u>, whether specified as background knowledge or as observations.

```
live_{-1} \leftarrow live_{-w_0}.
  live\_w_0 \leftarrow live\_w_1 \land up\_s_2.
  live\_w_0 \leftarrow live\_w_2 \wedge down\_s_2.
  live\_w_1 \leftarrow live\_w_3 \wedge up\_s_1.
  live\_w_2 \leftarrow live\_w_3 \land down\_s_1.
 live 1_2 \leftarrow live w_4.
 live\_w_4 \leftarrow live\_w_3 \wedge up\_s_3.
 live\_p_1 \leftarrow live\_w_3.
 live\_w_3 \leftarrow live\_w_5 \land ok\_cb_1.
 live\_p_2 \leftarrow live\_w_6.
 live\_w_6 \leftarrow live\_w_5 \land ok\_cb_2.
live\_w_5 \leftarrow live\_outside.
lit J_1 \leftarrow light J_1 \wedge live J_1 \wedge ok J_1.
lit J_2 \leftarrow light J_2 \wedge live J_2 \wedge ok J_2.
```

Questions and Answers

- A query is a way of asking whether a proposition is a logical consequence of a KB.
- A query is a question with the answer "yes" if the body is a logical consequence of the KB or
- the answer "no" if the body is not a consequence of the KB.

Questions and Answers

Example 5.8 Once the computer has been told the knowledge base of Example 5.7 (page 183), it can answer queries such as

ask $light J_1$.

for which the answer is yes. The query

ask $light 1_6$.

has answer no. The computer does not have enough information to know whether or not l_6 is a light. The query

ask $lit \perp 1_2$.

has answer *yes*. This atom is true in all models.

The user can interpret this answer with respect to the intended interpretation.

Proofs

- A theorem is a provable proposition.
 - A proof is a mechanically derivable demonstration that a proposition logically follows from a KB.
- A proof procedure is a possibly <u>non-deterministic</u> <u>algorithm</u> for deriving consequences of a KB
 - A proof procedure is <u>sound</u> concerning semantics if everything that can be derived from a KB is a <u>logical consequence</u> of the KB.
 - A proof procedure is <u>complete</u> concerning semantics if there is proof of each logical consequence of the KB.
 - Two ways to **construct proofs** for definite propositional clauses:
 - a bottom-up procedure and
 - a top-down procedure.

- A bottom-up proof procedure can be used to derive all logical consequences of a KB
 - The bottom-up proof procedure builds on atoms that have already been established.
 - We say that a bottom-up procedure is <u>forward chaining</u> on the definite clauses, in the sense of going forward from what is known rather than going backward from the query.
 - The general idea is based on one rule of derivation is a generalized form of the rule of inference called modus ponens:
 - If " $h \leftarrow a_1 \land \ldots \land a_m$ " is a **definite clause** in the **KB**, and each a_i has been derived, then h can be derived.
- Figure 5.3 gives a procedure for computing the consequence set C
 of a set KB of definite clauses.

 Modus Ponens: Create a conclusion from a rule and a fact. For example, consider the following rule and a fact:

Rule: $(a \leftarrow b)$

Fact: **b**

Conclusion: a

```
1: procedure Prove_DC_BU(KB)
       Inputs
2:
           KB: a set of definite clauses
3:
       Output
4:
           Set of all atoms that are logical consequences of KB
5:
       Local
6:
           C is a set of atoms
7:
       C := \{\}
8:
       repeat
9:
           select "h \leftarrow a_1 \land \ldots \land a_m" in KB where a_i \in C for all i, and h \notin C
10:
           C := C \cup \{h\}
11:
       until no more definite clauses can be selected
12:
       return C
13:
```

Figure 5.3: Bottom-up proof procedure for computing consequences of *KB*

Example 5.9 Suppose the system is given the knowledge base *KB*:

$$a \leftarrow b \land c$$
.
 $b \leftarrow d \land e$.
 $b \leftarrow g \land e$.
 $c \leftarrow e$.
 d .
 e .
 $f \leftarrow a \land g$.

One trace of the value assigned to *C* in the bottom-up procedure is

```
{}
{d}
{e,d}
{c,e,d}
{b,c,e,d}
{a,b,c,e,d}.
```

The algorithm terminates with $C = \{a, b, c, e, d\}$. Thus, $KB \vdash a$, $KB \vdash b$, and so on.

The last rule in KB is never used. **The bottom-up proof procedure** cannot derive f or g. This is as it should be because there is a model of the KB in which f and g are both f are g are g are g are g and g are g are g and g are g are g are g and g are g are g are g are g and g are g are g and g are g

- An alternative proof method is to search backward or top-down from a query to determine whether it is a logical consequence of the given definite clauses.
- Top-down procedure is also called backward chaining
- The top-down proof procedure can be understood in the <u>answer</u> clause (it is a definite close with the head "yes").
 - An **answer clause** is of the form: $yes \leftarrow a_1 \land a_2 \land \ldots \land a_m$ where yes is a special atom.
 - Intuitively, yes is going to be true exactly when the answer to the query is "yes."
- If the query is $(ask q_1 \land ... \land q_m)$, then the initial answer clause is $yes \leftarrow q_1 \land ... \land q_m$

- Given an answer clause, the top-down algorithm selects an atom in the body of the answer clause.
- Suppose it selects a_1 . The atom selected is called a **subgoal**.
- The algorithm proceeds by doing steps of resolution
- In one step of resolution, it chooses a definite clause in KB with a₁ as the head. If there is no such clause, the algorithm fails.
- The **resolvent** of the above **answer clause** on the selection a_1 with the **definite clause** $(a_1 \leftarrow b_1 \land \ldots \land b_p)$ is the **answer clause**: $(yes \leftarrow b_1 \land \ldots \land b_p \land a_2 \land \ldots \land a_m)$
- That is, the subgoal in the answer clause is replaced by the body of the chosen definite clause.
- An answer is an answer clause with an empty body (m = 0), it is the answer clause (yes ←).

- Figure 5.4 specifies a non-deterministic procedure for solving a query. It follows the definition of a derivation.
- In this procedure, **G** is the set of **atoms** in the body of the **answer** clause.
- The procedure is nondeterministic: line 12 has to choose a definite clause to resolve against.
- If there are choices that result in *G* being the *empty set*, the algorithm returns *yes*; otherwise, it *fails*, and the answer is *no*.
- This algorithm treats the body of a clause as a set of atoms, and G is also a set of atoms.
- An alternative is to have **G** as an ordered list of atoms, perhaps with an atom appearing more than once.

```
1: non-deterministic procedure Prove_DC_TD(KB, Query)
       Inputs
 2:
           KB: a set of definite clauses
 3:
           Query: a set of atoms to prove
 4:
       Output
 5:
           yes if KB \models Query and the procedure fails otherwise
 6:
       Local
 7:
           G is a set of atoms
 8:
       G := Query
9:
       repeat
10:
           select an atom a in G
11:
           choose definite clause "a \leftarrow B" in KB with a as head
12:
           G := B \cup (G \setminus \{a\})
13:
       until G = \{\}
14:
       return yes
15:
```

Figure 5.4: Top-down definite clause proof procedure

Example 5.10 Suppose the system is given the knowledge base:

```
a \leftarrow b \land c.
b \leftarrow d \land e.
b \leftarrow g \land e.
c \leftarrow e.
d.
e.
f \leftarrow a \land g.
```

It is asked the query: ask a.

The following shows a derivation that corresponds to a sequence of assignments to *G* in the repeat loop of Figure 5.4. Here we have written *G* in the form of an <u>answer clause</u>, and always selected the leftmost atom in the body:

$$yes \leftarrow a$$
 $yes \leftarrow b \land c$
 $yes \leftarrow d \land e \land c$
 $yes \leftarrow e \land c$
 $yes \leftarrow c$
 $yes \leftarrow e$
 $yes \leftarrow e$
 $yes \leftarrow e$

The following shows a sequence of choices, where the second definite clause for *b* was chosen. This choice does not lead to a proof.

$$yes \leftarrow a$$
$$yes \leftarrow b \land c$$
$$yes \leftarrow g \land e \land c$$

If *g* is selected, there are no rules that can be chosen. This proof attempt is said to *fail*.

- The non-deterministic top-down algorithm of Figure 5.4, together with a selection strategy, induces a search graph.
- Each node in the search graph represents an answer clause.
- The neighbors of a **node** ($yes \leftarrow a_1 \land \ldots \land a_m$), where a_1 is the **selected atom**, represent all of the possible answer clauses obtained by resolving on a_1 .
- There is a neighbor for each **definite clause** whose head is a_1 .
- The goal nodes of the search are of the form yes ← .

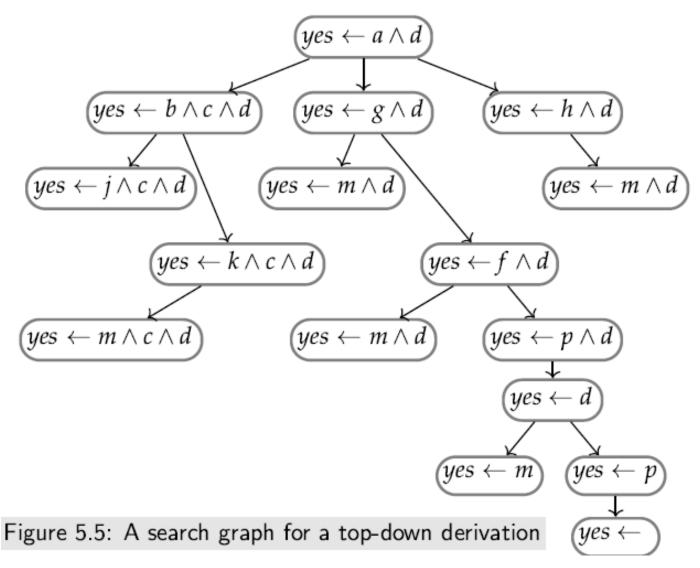
Example 5.11 Given the knowledge base

$$a \leftarrow b \land c$$
. $a \leftarrow g$. $a \leftarrow h$. $b \leftarrow j$. $b \leftarrow k$. $d \leftarrow m$. $d \leftarrow p$. $f \leftarrow m$. $f \leftarrow p$. $g \leftarrow m$. $g \leftarrow f$. $k \leftarrow m$. $h \leftarrow m$. p .

and the query

ask $a \wedge d$.

the search graph for an SLD derivation, assuming the leftmost atom is selected in each answer clause, is shown in Figure 5.5.



Top-Down Proof Procedure

 It is possible that the proof procedure can get into an infinite loop, as in the following example (without cycle pruning).

Example 5.12 Consider the knowledge base and query:

```
g \leftarrow a.

a \leftarrow b.

b \leftarrow a.

g \leftarrow c.

c.

ask g.
```

Atoms g and c are the only atomic logical consequences of this knowledge base, and the bottom-up proof procedure will halt with fixed point $\{c,g\}$. However, the top-down proof procedure with a depth-first search will go on indefinitely, and not halt if the first clause for g is chosen, and there is no cycle pruning.

Background Knowledge and Observations

- Assume an <u>observation</u> is a set of atomic propositions which are implicitly conjoined.
- Observations do not provide rules directly.
- Observation is like background knowledge in a KB, allowing the agent to do something useful.
- At design time or offline, the information arrives online as observations from users, sensors, and external knowledge sources
 - For example, a medical diagnosis program may have knowledge represented as definite clauses about the possible diseases and symptoms, but it would not know the actual symptoms manifested by a particular patient
 - A simple way to acquire information from a user is to incorporate an ask-the-user mechanism into the top-down proof procedure.

Knowledge-Level Explanation

- The explicit use of semantics allows explanation and debugging at the knowledge level
- To make a system usable by people, the system cannot just give an answer and expect the user to believe it
 - Consider the case of a system advising doctors who are legally responsible for the treatment they carry out based on the diagnosis.
 - The doctors must be convinced that the diagnosis is appropriate.
 - The system must be able to justify that its answer is correct.
 - The exact mechanism can explain how the system found a result and debugged the KB.

Knowledge-Level Explanation

- Three complementary means of interrogation are used to explain the relevant knowledge in a KB:
 - (1) how the question is used to describe how an answer was proved,
 - (2) why the question is used to ask the system why it is asking the user a question, and
 - (3) why not question is used to ask why an atom was not proven.
- To explain how an answer was proved, a "how" question can be asked by a user when the system has returned the answer
 - The system provides the definite clause used to deduce the answer.
 - The user can ask "why" when asked a question.
 - The system replies by giving the rule that produced the question.
 - The user can then ask why the head of that rule was being proved.

Knowledge-Level Explanation

- How a System can Prove an Atom?
- The first explanation procedure allows the user to ask "how" an atom was proved.
- If there is a proof for **g**, either **g** must be an **atomic clause** or there must be a rule:

 $g \leftarrow (a_1 \land \ldots \land a_k)$ such that each a_i has been proved.

- Definite clauses can be used in a proof by contradiction by allowing rules that give contradictions.
 - The definite clause language does not allow a contradiction to be stated.
- An <u>integrity constraint</u> is a clause of the form: false ← a₁ Λ . . . Λ
 a_k, where the a_i are atoms and false is a particular atom that is false in all interpretations.
- A Horn clause is either a definite clause or an integrity constraint.
- That is, a Horn clause has either false or a normal atom as its head.

- Integrity constraints allow the system to prove that some conjunction of atoms is false in all models of a KB.
- ¬p is the negation of p, which is true in an interpretation when p is false in that interpretation,
- and p v q is the disjunction of p and q, which is true in an interpretation if p is true or q is true, or both are true in the interpretation.
- The integrity constraint: $false \leftarrow a_1 \land \ldots \land a_k$ is logically equivalent to $\neg a_1 \lor \ldots \lor \neg a_k$.
- A Horn clause KB can imply negations of atoms, as shown in Example 5.19.

Example 5.19 Consider the knowledge base KB_1 :

$$false \leftarrow a \land b.$$

$$a \leftarrow c.$$

$$b \leftarrow c.$$

The atom c is false in all models of KB_1 . To see this, suppose instead that c is true in model I of KB_1 . Then a and b would both be true in I (otherwise I would not be a model of KB_1). Because false is false in I and a and b are true in I, the first clause is false in I, a contradiction to I being a model of KB_1 . Thus $\neg c$ is true in all models of KB_1 , which can be written as

$$KB_1 \models \neg c$$

 Although the language of Horn clauses does not allow disjunctions and negations to be input, disjunctions of negations of atoms can be derived, as the following example shows.

Example 5.20 Consider the knowledge base KB_2 :

$$false \leftarrow a \land b.$$

$$a \leftarrow c.$$

$$b \leftarrow d.$$

$$b \leftarrow e.$$

Either c is false or d is false in every model of KB_2 . If they were both true in some model I of KB_2 , both a and b would be true in I, so the first clause would be false in I, a contradiction to I being a model of KB_2 . Similarly, either c is false or e is false in every model of KB_2 . Thus,

$$KB_2 \models \neg c \lor \neg d$$

 $KB_2 \models \neg c \lor \neg e$.

- A set of clauses is unsatisfiable if it has no models.
- A set of clauses is provably inconsistent concerning a proof procedure.
- If a proof procedure is sound and complete, a set of clauses is provably inconsistent if and only if it is unsatisfiable.
- It is always possible to find a model for a set of definite clauses.
- The interpretation with all atoms true is a model of any set of definite clauses.
- Thus, <u>a definite-clause KB is always satisfiable</u>.
- However, a set of Horn clauses can be unsatisfiable.

- Abduction is a form of reasoning where assumptions are made to explain observations.
 - For example, if an agent observes that some light is not working, it hypothesizes what is happening in the world to explain why the light is not working.
- In abduction, an agent hypothesizes what may be true about an observed case.
- An agent determines what implies its observations what could be true to make them true.
- To formalize abduction, we use Horn clauses and assumable. The system is given:
 - a KB, a set of Horn clauses, and a set A of atoms, called the assumable, are the building blocks of hypotheses (see Example 5.33).

Example 5.33 Consider the following simplistic knowledge base and assumables for a diagnostic assistant:

```
bronchitis \leftarrow influenza.
bronchitis \leftarrow smokes.
coughing \leftarrow bronchitis.
wheezing \leftarrow bronchitis.
fever \leftarrow influenza.
fever \leftarrow infection.
soreThroat \leftarrow influenza.
false \leftarrow smokes \land nonsmoker.
assumable smokes, nonsmoker, influenza, infection.
```

- If the agent observes wheezing, there are two minimal explanations: {influenza} and {smokes}
- These explanations imply bronchitis and coughing.
- If (wheezing ∧ fever) is observed, the minimal explanations are
 {influenza} and {smokes, infection}.
- If (wheezing ∧ nonsmoker) was observed, there is one minimal explanation: {influenza, nonsmoker}.
- The other explanation of wheezing is inconsistent with being a non-smoker.

Example 5.34 Consider the knowledge base:

```
alarm \leftarrow tampering.
alarm \leftarrow fire.
smoke \leftarrow fire.
```

If *alarm* is observed, there are two minimal explanations:

{tampering} and {fire}.

If $alarm \land smoke$ is observed, there is one minimal explanation:

{fire}.

Notice how, when *smoke* is observed, there is no need to hypothesize *tampering* to explain *alarm*; it has been **explained away** by *fire*.