Logic for Computer Scientists

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1 Datalog

[no textbook reference]

1.1 Informal Examples

1.1.1 Example

We want to formalize the following statements.

- Marian is the mother of Michelle.
- Craig is the brother of Michelle.
- Ann is the mother of Barack.
- Barack is the father of Malia.
- Michelle is the mother of Malia.
- Barack is the father of Natasha.
- Michelle is the mother of Natasha.
- Craig is male.
- Natasha is female.

We can write these as so-called *Datalog facts*:

\[
\begin{align*}
\text{motherOf} & (\text{marian}, \text{michelle}) & (1) \\
\text{brotherOf} & (\text{craig}, \text{michelle}) & (2) \\
\text{motherOf} & (\text{ann}, \text{barack}) & (3) \\
\text{fatherOf} & (\text{barack}, \text{malia}) & (4) \\
\text{motherOf} & (\text{michelle}, \text{malia}) & (5) \\
\text{fatherOf} & (\text{barack}, \text{natasha}) & (6) \\
\text{motherOf} & (\text{michelle}, \text{natasha}) & (7) \\
\text{male} & (\text{craig}) & (8) \\
\text{female} & (\text{natasha}) & (9)
\end{align*}
\]

Say, we also want to formalize the following.

- Every father of a person is also a parent of that person.
- Every mother of a person is also a parent of that person.
- If somebody is the mother of another person, who in turn is the parent of a third person, then this first person is the grandmother of this third person.
- If a person is the brother of another person, and this other person is the parent of a third person, then this first person is the uncle of this third person.
- Every father is male.
We can write these as so-called *Datalog rules*:

\[
\begin{align*}
\text{fatherOf}(x, y) & \rightarrow \text{parentOf}(x, y) & (10) \\
\text{motherOf}(x, y) & \rightarrow \text{parentOf}(x, y) & (11) \\
\text{motherOf}(x, y) \land \text{parentOf}(y, z) & \rightarrow \text{grandmotherOf}(x, z) & (12) \\
\text{brotherOf}(x, y) \land \text{parentOf}(y, z) & \rightarrow \text{uncleOf}(x, z) & (13) \\
\text{fatherOf}(x, y) & \rightarrow \text{male}(x) & (14)
\end{align*}
\]

If we take all statements (1) to (14) together, then we can derive new knowledge, which is *implicit* in these statements, e.g. the following.

- from (4) and (10): \( \text{parentOf(barack, malia)} \) (15)
- from (2), (7), (11) and (13): \( \text{uncleOf(craig, natasha)} \) (16)
- from (3), (15) and (12): \( \text{grandmotherOf(ann, malia)} \) (17)

Note that we reused (15) to derive (17). Derived knowledge can be used to derive even further knowledge.

---

**1.1.2 Example**

Consider the following sentences.

- Every human is mortal.
- Socrates is a human.

We can write these in Datalog as follows.

- \( \text{human}(x) \rightarrow \text{mortal}(x) \)
- \( \text{human(socrates)} \)

From these two rules we can derive

- \( \text{mortal(socrates)} \).

**1.1.3 Example**

Consider the following facts.

- \( \text{newsFrom(Merkel is Chancellor, berlin)} \)
- \( \text{newsFrom(Obamacare is constitutional, dc)} \)

And furthermore assume there is a set of facts about locations of cities.

- \( \text{locatedIn(berlin, germany)} \)
- \( \text{locatedIn(dc, usa)} \)
We can also state the following Datalog rule.

\[ \text{newsFrom}(x, y) \land \text{locatedIn}(y, z) \rightarrow \text{newsFrom}(x, z) \]

Derived knowledge is then, e.g., the following.

\[ \text{newsFrom}(<\text{Merkel is Chancellor}, \text{germany}>), \text{newsFrom}(<\text{Obamacare is constitutional}, \text{usa}>) \]

**1.1.4 Example**

In Datalog, we can state, e.g., that locatedIn is *transitive*:

\[ \text{locatedIn}(x, y) \land \text{locatedIn}(y, z) \rightarrow \text{locatedIn}(x, z) \]

**1.1.5 Example**

We can write directed graphs as Datalog facts, e.g., as follows. If \( V = \{a, b, c, d\} \) is the set of vertices of the graph, and \( E = \{(a, b), (b, b), (b, c), (d, d), (d, a), (d, b)\} \) is the set of edges of the graph (see Figure 1), then we can write it as follows.

\[ \begin{align*}
\text{edge}(a, b) \\
\text{edge}(b, b) \\
\text{edge}(b, c) \\
\text{edge}(d, d) \\
\text{edge}(d, a) \\
\text{edge}(d, b)
\end{align*} \]

We can now formally define what it means that there is a path from a vertex to another:

\[ \begin{align*}
\text{edge}(x, y) \rightarrow \text{path}(x, y) \\
\text{path}(x, y) \land \text{path}(y, z) \rightarrow \text{path}(x, z)
\end{align*} \]

Then we can derive, e.g., the following.

\[ \begin{align*}
\text{path}(a, b) \\
\text{path}(b, c) \\
\text{path}(a, c)
\end{align*} \]
We can also specify that two vertices are connected if there is a path in either direction.

\[
\begin{align*}
\text{path}(x, y) & \rightarrow \text{connected}(x, y) \\
\text{connected}(x, y) & \rightarrow \text{connected}(y, x)
\end{align*}
\]

Then we can derive, e.g., the following.

\[
\begin{align*}
\text{connected}(a, b) \\
\text{connected}(b, a) \\
\text{connected}(a, c) \\
\text{connected}(c, a)
\end{align*}
\]

### 1.2 Syntax and Formal Semantics

#### 1.2.1 Definition

A *Datalog language* \( L = (V, C, R) \) consists of the following.

- A finite set \( V \) of *variables*: \( x_1, x_2, \ldots, x_n \) (also \( y, z, \ldots \)).
- A finite non-empty set \( C \) of *constants*: \( a, b, c, \ldots \)
- A finite non-empty set \( R \) of *predicate symbols*: \( p_1, p_2, \ldots \) (also \( q, r, \ldots \)), each with an *arity* \((\in \mathbb{N})\) (number of parameters).

An *atom* (or *atomic formula*) is of the form

\[ p(v_1, \ldots, v_n), \]

where \( p \) is a predicate symbol of arity \( n \) and each of the \( v_i \) is either a constant or a variable. An atom is called a *ground atom* if all the \( v_i \) are constants.

#### 1.2.2 Example

Let \( L \) consist of constants \( a, b \), of variables \( x, y \), and of predicate symbols \( p \) with arity 1 and \( q \) with arity 2.

Then the following are examples for atomic formulas over \( L \).

\[
\begin{align*}
p(a), & \quad p(y), & \quad q(a, b), & \quad q(b, b), & \quad q(b, x), & \quad q(y, y)
\end{align*}
\]

Of these, \( p(a) \), \( q(a, b) \) and \( q(b, b) \) are ground atoms.

The following are *not* atomic formulas over \( L \):

\[
\begin{align*}
p(a, b), & \quad q(x), & \quad p(c), & \quad a(x)
\end{align*}
\]

#### 1.2.3 Definition

A *Datalog rule* is a statement of the form

\[ B_1 \land \cdots \land B_n \rightarrow A, \]

where the \( B_i \) and \( A \) are atoms. \( B_1 \land \cdots \land B_n \) is called the *body* of the rule, each \( B_i \) is called a *body atom* of the rule, and \( A \) is called the *head* of the rule.

A rule with \( n = 0 \), i.e. with no body, is called a *fact*, and the arrow is omitted in this case.

A *Datalog program* is a set of Datalog rules.
1.2.4 Example
The following are examples of Datalog rules.

\[
\text{newsFrom}(x, y) \land \text{locatedIn}(y, z) \rightarrow \text{newsFrom}(x, z)
\]
\[
p(a, x) \land q(x, y) \rightarrow r(a, x, y)
\]
\[
p_2(a) \land p_3 \rightarrow q_2(a)
\]

Note that in this example, \( p_3 \) is a predicate symbol of arity 0.

1.2.5 Example
The statements (1) to (14) from Example 1.1.1 constitute a Datalog program. Statements (1) to (9) are facts.

1.2.6 Definition
Given a Datalog language \( L \) and a Datalog program \( P \) over \( L \), a Herbrand interpretation for \( P \) is a set of ground atoms over \( L \).

1.2.7 Example
Consider \( P \) to consist only of the following statements from Example 1.1.1 (with abbreviated notation, \( c, m, n \) are constants).

\[
\text{mOf}(x, y) \rightarrow \text{pOf}(x, y)
\]
\[
\text{bOf}(x, y) \land \text{pOf}(y, z) \rightarrow \text{uOf}(x, z)
\]
\[
\text{bOf}(c, m)
\]
\[
\text{mOf}(m, n)
\]

Then the following are examples for Herbrand interpretations.

\[
I_1 = \{ \text{bOf}(c, m), \text{mOf}(m, n), \text{pOf}(m, n), \text{uOf}(c, n) \}
\]
\[
I_2 = \{ \text{bOf}(m, c), \text{mOf}(c, n), \text{pOf}(m, n), \text{uOf}(n, c) \}
\]

1.2.8 Example
Some examples for interpretations of the following Datalog program, where \( a, b, c, d \) are constants.

\[
p(a, b)
\]
\[
p(b, c)
\]
\[
p(c, a)
\]
\[
p(d, d)
\]
\[
p(x, y) \rightarrow q(x, y)
\]
\[
q(x, y) \land q(y, z) \rightarrow q(x, z)
\]
\[
q(x, y) \rightarrow r(x, y)
\]
\[
r(x, y) \rightarrow r(y, x)
\]
\[
r(x, x) \rightarrow t(x)
\]
are the following.

\[ I_1 = \{ p(c, a), p(c, b), t(a) \} \]
\[ I_2 = \emptyset \]
\[ I_3 = \{ p(a, b), p(b, c), p(c, a), p(d, d), q(a, b), r(a, b), r(b, a), t(a), t(b) \} \]

1.2.9 Definition
A substitution \([x_1/c_1, \ldots, x_n/c_n]\), where the \(x_i\) are variables and the \(c_i\) are constants, is a mapping which maps each Datalog rule \(R\) to the rule \(R[x_1/c_1, \ldots, x_n/c_n]\), which is obtained from \(R\) by replacing all occurrences of \(x_i\) by \(c_i\), for all \(i = 1, \ldots, n\).

1.2.10 Example
In the following, \(a, b, c, d\) are constants, while \(x, y, z\) are variables.

\[
(p(x, y) \land q(y, z) \rightarrow r(x, z))[x/a, z/b] = p(a, y) \land q(y, b) \rightarrow r(a, b) \tag{18}
\]
\[
(q(x) \land r(x, y) \rightarrow p(y))[x/b, y/c] = q(b) \land r(b, c) \rightarrow p(c) \tag{19}
\]
\[
q(x, z, y)[x/b, z/b, y/a] = q(b, b, a) \tag{20}
\]
\[
(p(x, y) \land q(y, z, z) \rightarrow r(x, y))[x/b][x/a][y/c] = (p(b, y) \land q(y, z, z) \rightarrow r(b, y))[x/a][y/c]
= (p(b, y) \land q(y, z) \rightarrow r(b, y))[y/c]
= (p(b, c) \land q(c, z, z) \rightarrow r(b, c)) \tag{21}
\]

1.2.11 Definition
A ground rule is a Datalog rule which contains no variables. A substitution \(\varphi\) for a Datalog rule \(R\) is called a ground substitution for \(R\) if \(R\varphi\) is a ground rule.

1.2.12 Example
In Example 1.2.10, the substitutions in (19) and (20) are ground substitutions for these rules, while those in (18) and (21) are not.

1.2.13 Definition
Given a Datalog rule \(R\), we define \(\text{ground}(R)\) to be the Datalog program which consists of all ground rules \(R\varphi\) which can be obtained from \(R\) via a ground substitution \(\varphi\) for \(R\). In other words,

\[ \text{ground}(R) = \{ R\varphi \mid \varphi \text{ is a ground substitution for } R \} \]

Each \(S \in \text{ground}(R)\) is called a grounding of \(R\).
Given a Datalog program \(P\), we define the grounding of \(P\) as

\[ \text{ground}(P) = \bigcup_{R \in P} \text{ground}(R). \]

1.2.14 Remark
\(\text{ground}(P)\) is always a finite set if \(P\) is finite, since the underlying language \(L\) by definition has only a finite number of constants.
If \(L\) is not explicitly given, then we assume that the set \(C\) of constants in \(L\) contains exactly the constants occurring in \(P\).
1.2.15 Example
For the program $P$ in Example 1.2.7, ground($P$) consists of the following rules.

\[
\begin{align*}
  \text{mOf}(c, c) &\rightarrow \text{pOf}(c, c) \\
  \text{mOf}(c, m) &\rightarrow \text{pOf}(c, m) \\
  \text{mOf}(c, n) &\rightarrow \text{pOf}(c, n) \\
  \text{mOf}(m, c) &\rightarrow \text{pOf}(m, c) \\
  \text{mOf}(m, m) &\rightarrow \text{pOf}(m, m) \\
  \text{mOf}(m, n) &\rightarrow \text{pOf}(m, n) \\
  \text{mOf}(n, c) &\rightarrow \text{pOf}(n, c) \\
  \text{mOf}(n, m) &\rightarrow \text{pOf}(n, m) \\
  \text{mOf}(n, n) &\rightarrow \text{pOf}(n, n) \\
  \text{bOf}(c, c) \land \text{pOf}(c, c) &\rightarrow \text{uOf}(c, c) \\
  \text{bOf}(c, m) \land \text{pOf}(m, n) &\rightarrow \text{uOf}(c, m) \\
  \text{bOf}(c, n) \land \text{pOf}(n, m) &\rightarrow \text{uOf}(c, n) \\
\end{align*}
\]

\vdots \quad \text{overall 27 groundings of this rule}

\begin{align*}
  \text{bOf}(c, m) \\
  \text{mOf}(m, n)
\end{align*}

1.2.16 Definition
A Herbrand interpretation $I$ of $P$ is called a Herbrand model of $P$ if the following condition holds: For every rule

\[ B_1 \land \cdots \land B_n \rightarrow A \]

in ground($P$) with $\{B_1, \ldots, B_n\} \subseteq I$, we also have $A \in I$.

1.2.17 Example
In Example 1.2.7, $I_1$ is a Herbrand model of $P$, while $I_2$ is not a model of $P$.

1.2.18 Example
For the Datalog program $P$ consisting of the rules

\[
\begin{align*}
p(a) \\
q(a, b) \\
q(b, c) \\
p(x) \rightarrow r(x) \\
r(x) \land q(x, y) \rightarrow r(y) \\
r(x) \land q(y, x) \rightarrow q(x, y)
\end{align*}
\]
the following are Herbrand models:
\[ \{p(a), q(a, b), q(b, c), r(a), r(c), q(c, b), r(b), q(b, a)\} \]
\[ \{p(a), q(a, b), q(b, c), r(a), r(c), q(c, b), r(b), q(b, a), q(c, c)\} \]

However,
\[ \{p(a), q(a, b), q(b, c), r(a), r(c), q(c, b), r(b), q(b, a), q(a, c)\} \]
is not a Herbrand model.

1.2.19 Example
The Datalog program \( P \) consisting of the rules
\[
\begin{align*}
p(a) \\
q(b) \\
q(x) \rightarrow q(x)
\end{align*}
\]
has the following Herbrand models:
\[
\begin{align*}
\{p(a), q(b)\} \\
\{p(a), q(b), q(a)\} \\
\{p(a), q(b), q(a), p(b)\}
\end{align*}
\]

1.3 Fixed-point Semantics

1.3.1 Example
Consider the program \( P \) from Example 1.2.18. There is a systematic way of obtaining a Herbrand model, as follows. In order to do this, consider \( \text{ground}(P) \).

First, collect all facts from \( \text{ground}(P) \):
\[ I_1 = \{p(a), q(a, b), q(b, c)\}. \]

Next, collect all heads of rules in \( \text{ground}(P) \) for which all body atoms are in \( I_1 \), and add them to \( I_1 \):
\[ I_2 = I_1 \cup \{r(a)\}. \]

Next, collect all heads of rules in \( \text{ground}(P) \) for which all body atoms are in \( I_2 \), and add them to \( I_2 \):
\[ I_3 = I_2 \cup \{r(b)\}. \]

Iteratively continue this process until nothing is added any more:
\[ I_4 = I_3 \cup \{r(c), q(b, a)\} \]
\[ I_5 = I_4 \cup \{q(c, b)\} \]
\[ I_6 = I_5 \cup \emptyset = I_5 \]

We have already seen in Example 1.2.18 that \( I_5 \) is indeed a Herbrand model.
We now develop this idea systematically.

1.3.2 Definition
Given a Datalog program $P$ with underlying language $L$, let $B_P$ be the set of all ground atoms over $L$, called the Herbrand base of $P$. Furthermore, let $I_P = 2^{B_P}$ be the power set of $B_P$, i.e., the set of all subsets of $B_P$.

1.3.3 Example
For $P$ as in Example 1.2.18, we have

$$B_P = \{ p(a), p(b), p(c), r(a), r(b), r(c), q(a, a), q(a, b), q(a, c), q(b, a), q(b, b), q(b, c), q(c, a), q(c, b), q(c, c) \}$$

which consists of 15 atoms. Correspondingly, $I_P$ has $2^{15} = 32,768$ elements.

1.3.4 Remark
$I_P$ is in fact the set of all Herbrand interpretations for $P$.

1.3.5 Definition
Given a Datalog program $P$, define a function $T_P : I_P \rightarrow I_P$ by

$$T_P(I) = \{ A \in B_P \mid (B_1 \land \cdots \land B_n \rightarrow A) \in \text{ground}(P) \text{ and } B_i \in I \text{ for all } i = 1, \ldots, n \}.$$ 

This function is called the single-step operator, or immediate consequence operator, or simply the $T_P$-operator for $P$.

1.3.6 Example
Considering Example 1.3.1, we have

$$T_P(\emptyset) = I_1$$
$$T_P(I_1) = I_2$$
$$T_P(I_2) = I_3$$
$$T_P(I_3) = I_4$$
$$T_P(I_4) = I_5$$
$$T_P(I_5) = I_5$$

To give some further examples, we also have

$$T_P(\{r(c), q(c, c)\}) = \{ p(a), q(a, b), q(b, c), r(c), q(c, c) \}$$
$$T_P(\{p(b)\}) = \{ p(a), q(a, b), q(b, c), r(b) \}$$

1.3.7 Example
For $P$ as in Example 1.2.19, we have

$$T_P(\{q(c)\}) = \{p(a), q(b), q(c)\}$$
$$T_P(\{p(a), q(b), q(c)\}) = \{p(a), q(b), q(c)\}$$
1.3.8 Definition
Given a Datalog program \( P \) and \( I \in I_P \), we call \( I \) a pre-fixed point of \( T_P \) if \( T_P(I) \subseteq I \). We call \( I \) a fixed point of \( T_P \) if \( T_P(I) = I \).

1.3.9 Example
In Example 1.3.1, we have \( T_P(I_5) = I_5 \), hence \( I_5 \) is a fixed point of \( T_P \).

1.3.10 Example
For \( P \) as in Example 1.2.19, we have \( T_P(\{ p(a), q(b) \}) = \{ p(a), q(b) \} \), which therefore is a fixed point of \( T_P \).

We also have \( T_P(\{ p(a), q(b), p(b) \}) = \{ p(a), q(b) \} \subseteq \{ p(a), q(b), p(b) \} \), therefore \( \{ p(a), q(b), p(b) \} \) is a pre-fixed point of \( T_P \).

1.3.11 Theorem
Given any Datalog program \( P \), the pre-fixed points of \( T_P \) are exactly the Herbrand models of \( P \).

Proof: Let \( I \) be a pre-fixed point of \( T_P \), i.e., \( T_P(I) \subseteq I \). Now let \( B_1 \land \cdots \land B_n \rightarrow A \) be any rule in \( \text{ground}(P) \). If \( \{ B_1, \ldots, B_n \} \subseteq I \), then \( A \in T_P(I) \) by definition of \( T_P \), and hence \( A \in I \) by the assumption that \( T_P(I) \subseteq I \). This shows that \( I \) is a Herbrand model of \( P \).

Conversely, let \( I \) be a Herbrand model of \( P \). Now for any \( A \in T_P(I) \) there must be a rule \( B_1 \land \cdots \land B_n \rightarrow A \) in \( \text{ground}(P) \) with \( \{ B_1, \ldots, B_n \} \subseteq I \). Since \( I \) is a Herbrand model we obtain \( A \in I \). This shows \( T_P(I) \subseteq I \). \( \blacksquare \)

1.3.12 Lemma
Given any Datalog program \( P \) and \( I_1 \subseteq I_2 \in I_P \), we have that
\[
T_P(I_1) \subseteq T_P(I_2),
\]
i.e., the \( T_P \)-operator is monotonic.

Proof: For any \( A \in T_P(I_1) \) there must be a rule \( B_1 \land \cdots \land B_n \rightarrow A \) in \( \text{ground}(P) \) with \( \{ B_1, \ldots, B_n \} \subseteq I_1 \). Since \( I_1 \subseteq I_2 \) we obtain \( \{ B_1, \ldots, B_n \} \subseteq I_2 \), and hence \( A \in T_P(I_2) \) as required. \( \blacksquare \)

09/11/14

1.3.13 Definition
Given any Datalog program \( P \), we iteratively define the following.
\[
\begin{align*}
T_P \uparrow 0 &= \emptyset \\
T_P \uparrow 1 &= T_P(T_P \uparrow 0) \\
& \quad \vdots \\
T_P \uparrow (n + 1) &= T_P(T_P \uparrow n)
\end{align*}
\]
We furthermore define
\[ T_P^\omega = \bigcup_{n \in \mathbb{N}} T_P^n. \]

The sets \( T_P^n \) are called *iterates* of the \( T_P \)-operator.

### 1.3.14 Example
Returning to Example 1.3.1, we have \( I_n = T_P^n \) for all \( n = 1, \ldots, 6 \) and \( T_P^\omega = I_5 \).

### 1.3.15 Theorem
For every Datalog program \( P \), the following hold.

(a) \( T_P^\omega \) is a fixed point of \( T_P \).

(b) \( T_P^\omega \) is a Herbrand model for \( P \).

(c) For every Herbrand model \( M \) for \( P \) we have that \( T_P^\omega \subseteq M \).

Condition (c) states that \( T_P^\omega \) is the *least* Herbrand model of \( P \) (with respect to the set inclusion ordering).

**Proof:** First note that
\[ T_P^0 = \emptyset \subseteq T_P(\emptyset) = T_P^1, \]
and due to monotonicity of \( T_P \), we obtain
\[ T_P^n \subseteq T_P^{n+1} \]
for each \( n \in \mathbb{N} \). Thus, the iterates of \( T_P \) form an increasing chain:
\[ T_P^0 \subseteq T_P^1 \subseteq T_P^2 \subseteq \ldots \]
Since furthermore \( T_P^\omega \subseteq B_P \), and \( B_P \) is a finite set, there must be an \( n_P \) such that \( T_P^{n_P} = T_P^{n_P+1} = T_P^\omega \), i.e., \( T_P^{n_P} = T_P^\omega \) must be a fixed points of \( T_P \). This shows (a).

Every fixed point of \( T_P \) is a pre-fixed point of \( T_P \). Hence \( T_P^\omega \) is a pre-fixed point of \( T_P \) and thus a Herbrand model for \( P \) by Theorem 1.3.11. This shows (b).

Now assume that \( M \) is another Herbrand model for \( P \). Clearly, \( \emptyset \subseteq M \), and thus \( T_P^1 = T_P(\emptyset) \subseteq T_P(M) \subseteq M \) by monotonicity of \( T_P \) and since \( M \) is a pre-fixed point of \( T_P \). By iteratively repeating this argument we obtain \( T_P^n \subseteq T_P(M) \subseteq M \) for all \( n \in \mathbb{N} \), and thus \( T_P^\omega \subseteq M \), which shows (c).

### 1.3.16 Definition
We say that a Datalog program \( P \) *Herbrand-entails* a ground atom \( A \), written \( P \models_H A \), if \( A \in T_P^\omega \). In this case we also call \( A \) a *logical consequence of \( P \)* under the *Herbrand semantics*. 
1.3.17 Theorem
Given a Datalog program \( P \) and a ground atom \( A \), we have \( P \models_H A \) if and only if \( A \in M \) for all Herbrand models \( M \) of \( P \).

Proof: If \( P \models_H A \), then \( A \in T_P^\omega \). For any Herbrand model \( M \) of \( P \) we know that \( T_P^\omega \subseteq M \), and we obtain \( A \in M \) as required.
If \( A \in M \) for all Herbrand models \( M \) of \( P \), then we obtain \( A \in T_P^\omega \) because \( T_P^\omega \) is a Herbrand model of \( P \). Thus, \( P \models_H A \). ■

1.3.18 Example
Consider Example 1.2.19, note that \( \{ p(a), q(b) \} \) is the intersection of all the Herbrand models, and thus contains exactly the logical consequences of \( P \).
2 Propositional Logic

2.1 Syntax

[Schöning, 1989, Chapter 1.1]

Let \{A_1, A_2, \ldots\} be an infinite set of propositional variables.

2.1.1 Definition

An atomic formula is a propositional variable. Formulas are defined by the following inductive process.

1. All atomic formulas are formulas.

2. For every formula \(F\), \(\neg F\) is a formula, called the negation of \(F\).

3. For all formulas \(F\) and \(G\), also \((F \lor G)\) and \((F \land G)\) are formulas, called the disjunction and the conjunction of \(F\) and \(G\), respectively.

The symbols \(\neg\), \(\lor\), \(\land\) are called connectives. \(\neg\) is a unary connective, while \(\lor\) and \(\land\) are binary connectives.

If a formula \(F\) occurs in another formula \(G\), then it is called a subformula of \(G\). Note that every formula is a subformula of itself.

2.1.2 Notation

We use the following abbreviations:

\(A, B, C, \ldots\) instead of \(A_1, A_2, \ldots\) and other obvious variants.

[Be careful with the use of \(F\) and \(G\)!]

We sometimes omit brackets if it can be done safely. [Be careful with this!]

\((F \rightarrow G)\) instead of \((\neg F \lor G)\)

\((F \leftrightarrow G)\) instead of \((F \rightarrow G) \land (G \rightarrow F)\)

\(\rightarrow\) and \(\leftrightarrow\) are also called connectives.

\((\bigvee_{i=1}^{n} F_i)\) instead of \((F_1 \lor F_2 \lor \cdots \lor F_n)\)

\((\bigwedge_{i=1}^{n} F_i)\) instead of \((F_1 \land F_2 \land \cdots \land F_n)\)

2.1.3 Example

\((\neg B \rightarrow F)\) is \((\neg \neg B \lor F)\).

Some Subformulas: \(\neg \neg B\), \(\neg B\).

2.1.4 Example

\(((I \lor \neg B) \rightarrow \neg F)\) is \((\neg (I \lor \neg B) \lor \neg F)\).

Some Subformulas: \(\neg (I \lor \neg B)\), \(I\), \(\neg B\).

2.1.5 Remark

Formulas can be represented in a unique way as trees. [Example 2.1.4 on whiteboard.]
2.2 Semantics

[Schöning, 1989, Chapter 1.1 cont.]

2.2.1 Definition

\(\mathbb{T} = \{0, 1\}\) – the set of truth values: false, and true, respectively.

An assignment is a function \(\mathcal{A} : D \to \mathbb{T}\), where \(D\) is a set of atomic formulas.

Given such an assignment \(\mathcal{A}\), we extend it to \(\mathcal{A}' : E \to \mathbb{T}\), where \(E\) is the set of all formulas containing only elements from \(D\) as atomic subformulas:

1. \(\mathcal{A}'(A_i) = \mathcal{A}(A_i)\) for each \(A_i \in D\)

2. \(\mathcal{A}'(F \land G) = \begin{cases} 1, & \text{if } \mathcal{A}'(F) = 1 \text{ and } \mathcal{A}'(G) = 1 \\ 0, & \text{otherwise} \end{cases}\)

3. \(\mathcal{A}'(F \lor G) = \begin{cases} 1, & \text{if } \mathcal{A}'(F) = 1 \text{ or } \mathcal{A}'(G) = 1 \\ 0, & \text{otherwise} \end{cases}\)

4. \(\mathcal{A}'(\neg F) = \begin{cases} 1, & \text{if } \mathcal{A}'(F) = 0 \\ 0, & \text{otherwise} \end{cases}\)

[From now on, drop distinction between \(\mathcal{A}\) and \(\mathcal{A}'\).]

2.2.2 Example

Let \(\mathcal{A}(B) = \mathcal{A}(F) = 1\) and \(\mathcal{A}(I) = 0\).

\[
\mathcal{A}(\neg(B \land F) \lor I) = \begin{cases} 1, & \text{if } \mathcal{A}(\neg(B \land F)) = 1 \text{ or } \mathcal{A}(I) = 1 \\ 0, & \text{otherwise} \end{cases}
\]

\(= \begin{cases} 1, & \text{if } \mathcal{A}(B \land F) = 0 \text{ or } \mathcal{A}(I) = 0 \\ 0, & \text{otherwise} \end{cases}\)

\(= \begin{cases} 1, & \text{if } \mathcal{A}(B) = 0 \text{ or } \mathcal{A}(F) = 0 \text{ or } \mathcal{A}(I) = 0 \\ 0, & \text{otherwise} \end{cases}\)

\(= 1\)

2.2.3 Remark

The same thing can be expressed via truth tables.

<table>
<thead>
<tr>
<th>(\mathcal{A}(F))</th>
<th>(\mathcal{A}(G))</th>
<th>(\mathcal{A}(F \land G))</th>
<th>(\mathcal{A}(F))</th>
<th>(\mathcal{A}(G))</th>
<th>(\mathcal{A}(F \lor G))</th>
<th>(\mathcal{A}(F))</th>
<th>(\mathcal{A}(\neg F))</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0</td>
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<td>1</td>
<td>0</td>
</tr>
</tbody>
</table>
2.2.4 Example
Determining the truth values of formulas using truth tables:
[Use the tree structure of formulas.]
\[
\begin{array}{cccccccc}
A(B) & A(F) & A(I) & A(B \land F) & A(\neg(B \land F)) & A(\neg I) & A(\neg(B \land F) \lor \neg I) \\
0 & 0 & 0 & 0 & 1 & 1 & 1 \\
0 & 0 & 1 & 0 & 1 & 0 & 1 \\
0 & 1 & 0 & 0 & 1 & 1 & 1 \\
0 & 1 & 1 & 0 & 1 & 0 & 1 \\
1 & 0 & 0 & 0 & 1 & 1 & 1 \\
1 & 0 & 1 & 0 & 1 & 0 & 1 \\
1 & 1 & 0 & 1 & 0 & 1 & 1 \\
1 & 1 & 1 & 1 & 0 & 0 & 0 \\
\end{array}
\]

2.2.5 Remark
The truth value of a formula is uniquely determined by the truth values of the propositional variables it contains as subformulas.

2.2.6 Remark
\[
\begin{array}{cccc}
A(F) & A(G) & A(F \rightarrow G) \\
0 & 0 & 1 \\
0 & 1 & 0 \\
1 & 0 & 0 \\
1 & 1 & 1 \\
\end{array}
\quad
\begin{array}{cccc}
A(F) & A(G) & A(F \leftrightarrow G) \\
0 & 0 & 1 \\
0 & 1 & 0 \\
1 & 0 & 0 \\
1 & 1 & 1 \\
\end{array}
\]

2.2.7 Definition
\(F\), a formula, \(\mathcal{A}\), an assignment.
\(\mathcal{A}\) is suitable if it is defined for all atomic formulas occurring in \(F\).
We write \(\mathcal{A} \models F\) if \(\mathcal{A}\) is suitable for \(F\) and \(\mathcal{A}(F) = 1\). We say \(F\) holds under \(\mathcal{A}\) or \(\mathcal{A}\) is a model for \(F\). Otherwise, we write \(\mathcal{A} \not\models F\).
\(F\) is satisfiable if \(F\) has at least one model. Otherwise, it is called unsatisfiable or contradictory.
A set \(M\) of formulas is satisfiable if there is an assignment \(\mathcal{A}\) which is a model for each formula in \(M\). In this case, \(\mathcal{A}\) is called a model of \(M\), and we write \(\mathcal{A} \models M\). [Note the overloading of notation.]
\(F\) is called valid or a tautology if every suitable assignment for \(F\) is a model for \(F\). In this case we write \(\models F\), and otherwise \(\not\models F\).

2.2.8 Example
Examples of models for \(p \lor \neg q \lor \neg r\) are the following. \(\mathcal{A}_1\) with \(\mathcal{A}_1(p) = \mathcal{A}_1(q) = \mathcal{A}_1(r) = 1\); \(\mathcal{A}_2\) with \(\mathcal{A}_2(p) = 1\) and \(\mathcal{A}_2(q) = \mathcal{A}_2(r) = 0\); \(\mathcal{A}_3\) with \(\mathcal{A}_3(p) = \mathcal{A}_3(q) = 0\) and \(\mathcal{A}_3(r) = 1\). You can find models by making the truth table for the formula: the assignments for which the truth value is 1 are models.

2.2.9 Example
For the formula \((p \land \neg q) \lor \neg p\), the assignment \(\mathcal{A}_1\) with \(\mathcal{A}_1(p) = \mathcal{A}_1(q) = 0\) is a model, because \(\mathcal{A}_1((p \land \neg q) \lor \neg p) = 1\). The assignment \(\mathcal{A}_2\) with \(\mathcal{A}_2(p) = 0\) and \(\mathcal{A}_2(q) = 1\) is also a model,
because $A_2((p \land \neg q) \lor \neg p) = 1$. This can also be seen from the truth table for $(p \land \neg q) \lor \neg p$. The assignment $A_3$ which only assigns $A_3(p) = 0$ is not a model for the formula because it is not suitable for the formula.

2.2.10 Example

$A \lor \neg A$ is a tautology.

[This is established by the following truth table:

<table>
<thead>
<tr>
<th>$A(A)$</th>
<th>$A(\neg A)$</th>
<th>$A(A \lor \neg A)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>1</td>
<td>1</td>
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<tr>
<td>1</td>
<td>0</td>
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</tbody>
</table>
]

2.2.11 Theorem

A formula $F$ is a tautology if and only if $\neg F$ is unsatisfiable.

Proof: $F$ is a tautology

iff every suitable assignment for $F$ is a model for $F$

iff every suitable assignment for $F$ (hence also for $\neg F$) is not a model for $\neg F$

iff $\neg F$ does not have a model

iff $\neg F$ is unsatisfiable \[■\]

2.2.12 Definition

A formula $G$ is a (logical) consequence of a set $M = \{F_1, \ldots, F_n\}$ of formulas if for every assignment $A$ which is suitable for $G$ and for all elements of $M$, it follows that whenever $A \models F_i$ for all $i = 1, \ldots, n$, then $A \models G$.

If $G$ is a logical consequence of $M$, we write $M \models G$ and say $M$ entails $G$. [Note the overloading of notation!]

2.2.13 Theorem

The following assertions are equivalent.

1. $G$ is a consequence of $\{F_1, \ldots, F_n\}$.

2. $((\land_{i=1}^n F_i) \to G)$ is a tautology.

3. $((\land_{i=1}^n F_i) \land \neg G)$ is unsatisfiable.

2.2.14 Example

Using Theorem 2.2.13, we can determine logical consequences using truth tables.

E.g., *modus ponens*: $\{P, P \to Q\} \models Q$.

We have to show: $(P \land (P \to Q)) \to Q$ is a tautology.

<table>
<thead>
<tr>
<th>$A(P)$</th>
<th>$A(Q)$</th>
<th>$A(P \to Q)$</th>
<th>$A(P \land (P \to Q))$</th>
<th>$A((P \land (P \to Q)) \to Q)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
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</table>
2.3 Datalog Revisited: Semantics By Grounding

We can relate Datalog and propositional logic as follows.

2.3.1 Definition
Given a Datalog language $L$, we can define a set of propositional variables as follows. For every ground atom $p(v_1, \ldots, v_n)$ over $L$, let $p_{v_1, \ldots, v_n}$ be a propositional variable. Furthermore, let $v$ be the function from ground atoms to propositional variables defined as

$$v(p(v_1, \ldots, v_n)) = p_{v_1, \ldots, v_n}.$$ 

Given a ground Datalog rule $B_1 \land \cdots \land B_k \rightarrow A$, furthermore define

$$v(B_1 \land \cdots \land B_k \rightarrow A) = v(B_1) \land \cdots \land v(B_k) \rightarrow v(A)$$

if $k \geq 1$, and

$$v(\rightarrow A) = v(A)$$

(for facts).

2.3.2 Example
For the ground Datalog rule 

$$p(a, b) \land q(c) \rightarrow p(a, c)$$

we have

$$v(p(a, b) \land q(c) \rightarrow p(a, c)) = v(p(a, b)) \land v(q(c)) \rightarrow v(p(a, c)) = p_{a,b} \land q_c \rightarrow p_{a,c}.$$ 

For the ground fact $p(b, c)$ we have $v(p(b, c)) = p_{b,c}$.

2.3.3 Definition
Given a Datalog program $P$, define the associated set $v(P)$ of propositional formulas as

$$v(P) = \{v(r) \mid r \in \text{ground}(P)\}.$$
2.3.4 Example
For the program $P$ in Examples 1.2.7 and 1.2.15, $v(P)$ consists of the following formulas.

$$
\begin{align*}
&m_{c,c} \rightarrow p_{c,c} \\
&m_{c,m} \rightarrow p_{c,m} \\
&m_{c,n} \rightarrow p_{c,n} \\
&m_{m,c} \rightarrow p_{m,c} \\
&m_{m,m} \rightarrow p_{m,m} \\
&m_{m,n} \rightarrow p_{m,n} \\
&m_{n,c} \rightarrow p_{n,c} \\
&m_{n,m} \rightarrow p_{n,m} \\
&m_{n,n} \rightarrow p_{n,n} \\
&b_{c,c} \land p_{c,c} \rightarrow u_{c,c} \\
&b_{c,m} \land p_{m,n} \rightarrow u_{c,m} \\
&b_{c,n} \land p_{n,m} \rightarrow u_{c,n}
\end{align*}
$$

\[\vdots\] overall 27 groundings of this rule

\[b_{c,m}, m_{m,n}\]

2.3.5 Theorem
Let $P$ be a Datalog program and $A \in B_P$. Then $P \models_H A$ if and only if $v(P) \models v(A)$.

Proof: For every Herbrand interpretation $I$ define an assignment $A_I$ by setting, for every ground atom $B$,

$$A_I(v(B)) = \begin{cases} 
1 & \text{if } B \in I \\
0 & \text{if } B \not\in I.
\end{cases}$$

Clearly, if $I$ is a Herbrand model for $P$, then $A_I$ is a model for $v(P)$.

Now assume $v(P) \models v(A)$, and let $M$ be a Herbrand model of $P$. Then $A_M$ is a model of $v(P)$ and hence $A_M(v(A)) = 1$. By definition of $A_M$ we obtain $A \in M$ as required.

Conversely, for every assignment $A$, define a Herbrand interpretation $I_A$ as

$$I_A = \{B \mid A(v(B)) = 1\}.$$ 

Clearly, if $A$ is a model for $v(P)$ then $I_A$ is a Herbrand model for $P$.

Now assume $P \models_H A$, and let $M$ be model for $v(P)$. Then $I_M$ is a Herbrand model for $P$ and hence $A \in I_M$. By definition of $I_M$ we obtain $M(v(A)) = 1$ as required.

Remark:
Theorem 2.3.5 shows how the problem of determining logical consequences for Datalog can be reduced (transformed) to the problem of determining logical consequences for propositional logic.
However, the theorem does not work in the other direction. E.g., in propositional logic we have \( \{\neg p, p \lor q\} \models q \) and \( \{p \to q, q \to r\} \models p \to r \), and neither of these can be transformed into a Datalog problem based on the theorem.

### 2.4 Equivalence

[Schöning, 1989, Chapter 1.2]

#### 2.4.1 Definition

Formulas \( F \) and \( G \) are (semantically) equivalent (written \( F \equiv G \)) if for every assignment \( A \) that is suitable for \( F \) and \( G \), \( A(F) = A(G) \).

#### 2.4.2 Example

\( A \lor B \equiv B \lor A \). (commutativity of \( \lor \))

<table>
<thead>
<tr>
<th>( A(A) )</th>
<th>( A(B) )</th>
<th>( A(A \lor B) )</th>
<th>( A(B \lor A) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
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</tbody>
</table>

\( A \lor \neg A \equiv B \lor \neg B \). [truth table]

#### 2.4.3 Example

\( F \equiv G \) iff \( \models (F \leftrightarrow G) \). [truth table]

#### 2.4.4 Theorem

The following hold for all formulas \( F \), \( G \), and \( H \).

\begin{align*}
F \land F &\equiv F & F \lor F &\equiv F & \text{Idempotency} \\
F \land G &\equiv G \land F & F \lor G &\equiv G \lor F & \text{Commutativity} \\
(F \land G) \land H &\equiv F \land (G \land H) & (F \lor G) \lor H &\equiv F \lor (G \lor H) & \text{Associativity} \\
F \land (G \lor H) &\equiv (F \land G) \lor (F \land H) & F \lor (G \land H) &\equiv (F \lor G) \land (F \lor H) & \text{Distributivity} \\
\neg \neg F &\equiv F & \text{Double Negation} & \text{(also Involution)} \\
\neg(F \land G) &\equiv \neg F \lor \neg G & \neg(F \lor G) &\equiv \neg F \land \neg G & \text{de Morgan’s Laws}
\end{align*}

**Proof:** Straightforward using truth tables.

#### 2.4.5 Remark

Disjunction is dispensable. \([F \lor G \equiv \neg(\neg F \land \neg G)]\)

Alternatively, conjunction is dispensable. \([F \land G \equiv \neg(\neg F \lor \neg G)]\)
2.4.6 Remark
Let $F \uparrow G = \neg(F \land G)$.
\[
\neg F \equiv \neg(F \land F) \equiv F \uparrow F.
\]
\[
F \lor G \equiv \neg(\neg F \land \neg G) \equiv \neg F \uparrow \neg G \equiv (F \uparrow F) \uparrow (G \uparrow G).
\]
\[
F \land G \equiv \neg(\neg F \land G) \equiv (F \uparrow G) \equiv (F \uparrow G) \uparrow (F \uparrow G).
\]

2.4.7 Remark (The contraposition principle)
\[
\{F\} \models G \iff \{\neg G\} \models \neg F.
\]
\[
[F] \models G \iff F \rightarrow G \text{ is a tautology (Theorem 2.2.13)}.
\]
\[
F \rightarrow G \equiv \neg F \lor G \equiv \neg(\neg G) \lor (\neg F) \equiv (\neg G) \rightarrow (\neg F).
\]
\[
(\neg G) \rightarrow (\neg F) \text{ is a tautology iff } \{\neg G\} \models \neg F \text{ (Theorem 2.2.13)}
\]

2.5 Normal Forms

[Schöning, 1989, Chapter 1.2 cont.]

2.5.1 Definition
A literal is an atomic formula (a positive literal) or the negation of an atomic formula (a negative literal).
A formula $F$ is in negation normal form (NNF) if it is made up only of literals, $\lor$, and $\land$ (and brackets).

2.5.2 Theorem
For every formula $F$, there is a formula $G \equiv F$ which is in NNF.

Proof: The proof of Theorem 2.5.5 below shows this as well. ■

2.5.3 Example
\[
(\neg(I \lor \neg B) \lor \neg F) \equiv (\neg I \land B) \lor \neg F
\]

2.5.4 Definition
A formula $F$ is in conjunctive normal form (CNF) if it is a conjunction of disjunctions of literals, i.e., if
\[
F = \left(\bigwedge_{i=1}^{n} \left(\bigvee_{j=1}^{m_i} L_{i,j}\right)\right),
\]
where the $L_{i,j}$ are literals.
A formula $F$ is in disjunctive normal form (DNF) if it is a disjunction of conjunctions of literals, i.e., if
\[
F = \left(\bigvee_{i=1}^{n} \left(\bigwedge_{j=1}^{m_i} L_{i,j}\right)\right),
\]
where the $L_{i,j}$ are literals.

2.5.5 Theorem
For every formula $F$ there is a formula $F_1 \equiv F$ in CNF and a formula $F_2 \equiv F$ in DNF.
**Proof:** Proof by structural induction.

*Induction base:* If \( F \) is atomic, then it is already in CNF and in DNF.

*Induction hypothesis:* \( G \) has CNF \( G_1 \) and DNF \( G_2 \), \( H \) has CNF \( H_1 \) and DNF \( H_2 \). *Induction step:* We have 3 cases.

Case 1: \( F \) has the form \( F = \neg G \).

Then

\[
F \equiv \neg G_1 \equiv \neg \left( \bigwedge_{i=1}^{n} \left( \bigvee_{j=1}^{m_i} L_{i,j} \right) \right) \equiv \left( \bigvee_{i=1}^{n} \left( \bigwedge_{j=1}^{m_i} \neg L_{i,j} \right) \right) \equiv \left( \bigwedge_{i=1}^{n} \left( \bigvee_{j=1}^{m_i} \overline{L_{i,j}} \right) \right),
\]

where

\[
\overline{L_{i,j}} = \begin{cases} A & \text{if } L_{i,j} = \neg A \\ \neg A & \text{if } L_{i,j} = A \end{cases}
\]

and the latter formula is in DNF as required. Analogously, we can obtain from \( G_2 \) a CNF formula equivalent to \( F \).

Case 2: \( F \) has the form \( F = G \lor H \).

Then \( F \equiv G_2 \lor H_2 \), which is in DNF.

Further,

\[
F \equiv G_1 \lor H_1 \equiv \left( \bigwedge_{i=1}^{n} \left( \bigvee_{j=1}^{m_i} K_{i,j} \right) \right) \lor \left( \bigwedge_{k=1}^{o} \left( \bigvee_{l=1}^{p_k} L_{k,l} \right) \right) \equiv \left( \bigwedge_{i=1}^{n} \left( \bigvee_{k=1}^{o} \left( \bigwedge_{l=1}^{p_k} \overline{K_{i,j}} \lor \bigvee_{l=1}^{L_{k,l}} \right) \right) \right),
\]

which is in CNF.

Case 3: \( F \) has the form \( F = G \land H \).

This case is analogous to Case 2. \( \square \)

2.5.6 Remark

Structural induction is a fundamental proof technique, comparable with natural induction.

2.5.7 Remark

DNF via truth table.

If, e.g.,

<table>
<thead>
<tr>
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<th>( \mathcal{A}(B) )</th>
<th>( \mathcal{A}(C) )</th>
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</table>

then a DNF for \( F \) is \( (\neg A \land \neg B \land \neg C) \lor (A \land \neg B \land \neg C) \lor (A \land \neg B \land C) \).

2.5.8 Definition

Two formulas \( F \) and \( G \) are *equisatisfiable* if the following holds: \( F \) has a model if and only if \( G \) has a model.
2.6 Tableaux Algorithm

[Ben-Ari, 1993, Chapter 2.6, strongly modified]

Translating truth tables directly into an algorithm is very expensive. We take the following approach:

For showing $F_1, \ldots, F_n \models G$, if suffices to show that $F = F_1 \land \cdots \land F_n \land \neg G$ is unsatisfiable (Theorem 2.2.13).

We attempt to construct a model for $F$ in such a way that, if and only if the construction fails, we know that $F$ is unsatisfiable.

2.6.1 Definition

Let $F$ be a formula in NNF. A tableau branch for $F$ is a set of formulas, defined inductively as follows.

- $\{F\}$ is a tableau branch for $F$.
- If $T$ is a tableau branch for $F$ and $G \land H \in T$, then $T \cup \{G, H\}$ is a tableau branch for $F$.
- If $T$ is a tableau branch for $F$ and $G \lor H \in T$, then $T \cup \{G\}$ is a tableau branch for $F$ and $T \cup \{H\}$ is a tableau branch for $F$.

A tableau for $F$ is a set of tableau branches for $F$.

A tableau branch is **closed** if it contains an atomic formula $A$ and the literal $\neg A$. Otherwise, it is **open**.

A tableau branch $T$ is called **complete** if it satisfies the following conditions.

- $T$ is open.
- If $G \land H \in T$, then $\{G, H\} \subseteq T$.
- If $G \lor H \in T$, then $G \in T$ or $H \in T$.

A tableau $M$ for $F$ is called **complete** if it satisfies the following conditions.

- If $G \lor H \in T \in M$, and $T$ is open, then there are branches $S_1 \in M$ and $S_2 \in M$ with $\{G\} \cup T \subseteq S_1$ and $\{H\} \cup T \subseteq S_2$.
- All branches of $M$ are complete or closed.

A tableau is **closed** if it is complete and all its branches are closed.

If $F$ is not in NNF, then a tableau (resp., tableau branch) for $F$ is a tableau (resp. tableau branch) for an NNF of $F$.

2.6.2 Example

Consider $(\neg I \land B) \lor \neg F$, for which a complete (but not closed) tableau is $\{((\neg I \land B) \lor \neg F, \neg I \land B, \neg I, B), \{(\neg I \land B) \lor \neg F, \neg F\}\}$.

2.6.3 Remark

Tableaux can be represented graphically (blackboard).

2.6.4 Theorem (Soundness)

A formula $F$ is satisfiable if there is a complete tableau branch for $F$.

2.6.5 Theorem (Completeness)

If a formula $F$ is satisfiable, then there is a complete tableau branch for $F$. 
2.6.6 Theorem
A formula $F$ is
1. unsatisfiable if and only if there is a closed tableau for $F$,
2. a tautology if and only if there is a closed tableau for $\neg F$.

2.6.7 Example
Modus Ponens holds if $(P \land (P \rightarrow Q)) \rightarrow Q$ is a tautology. We construct a complete tableau (blackboard) for $\neg((P \land (P \rightarrow Q)) \rightarrow Q)$, which turns out to be closed.

2.6.8 Lemma
Let $F$ be a formula, $T$ be a complete tableau branch for $F$, and $L_1, \ldots, L_n$ be all the literals contained in $T$. Then any assignment $A$ with $A(L_1 \land \cdots \land L_n) = 1$ is a model for $F$.

**Proof:** We show by structural induction, that $A$ is a model for each formula $F'$ in $T$.
Induction Base: Let $F' = L$ be a literal. Then by definition $A(F') = 1$.
Induction Hypothesis: $A(G) = A(H) = 1$ for $G, H \in T$.
Induction Step: (1) Let $F' = G \land H \in T$. Then $G \in T$ and $H \in T$. By IH, $A(F') = A(G \land H) = 1$. (2) Let $F' = G \lor H$. Then $G \in T$ or $H \in T$. By IH, $A(G) = 1$ or $A(H) = 1$, hence $A(F') = 1$. (3) The case $F' = \neg G \in T$ cannot happen since all formulas are in NNF, and the literal case was dealt with in the induction base.

**Proof of Theorem 2.6.4:** By Lemma 2.6.8, we obtain that $F$ has a model, hence it is satisfiable.

2.6.9 Example
Is the following formula valid? satisfiable? unsatisfiable?

$((A \rightarrow B) \rightarrow A) \rightarrow A$

(done on whiteboard)

2.6.10 Example
Is the following formula valid? satisfiable? unsatisfiable?

$(A \rightarrow (B \rightarrow C)) \rightarrow ((A \land B) \rightarrow C)$

(done on whiteboard)

**Proof of Theorem 2.6.5:** First note the following, for any assignment $M$ and all formulas $G$ and $H$:
- If $M \models G \land H$, then $M \models G$ and $M \models H$.
- If $M \models G \lor H$, then $M \models G$ or $M \models H$.

Since $F$ is satisfiable, it has a model $M$. Construct a tableau branch $T$ for $F$ recursively as follows.
- If $G \land H \in T$, set $T := T \cup \{G, H\}$.
- If $G \lor H \in T$ with $M \models G$, set $T := T \cup \{G\}$, otherwise set $T := T \cup \{H\}$.

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The recursion terminates since only subformulas of $F$ are added and sets cannot contain duplicate elements. The resulting $T$ is a complete tableau branch, and $M \models T$, by definition.

Proof of Theorem 2.6.6:
We prove Statement 1. Statement 2 is shown in Exercise ??.
Let $A$ be the statement “$F$ is unsatisfiable”, and let $B$ be the statement “$F$ has a closed tableau”.
We need to show: $A \equiv B$, for which it suffices to show that $A \leftrightarrow B \equiv (A \rightarrow B) \land (B \rightarrow A)$ is valid.
By the contraposition principle, it therefore suffices to show that $(\neg B \rightarrow \neg A) \land (\neg A \rightarrow \neg B) \equiv (\neg B \leftrightarrow \neg A)$ is valid, i.e., that $\neg A \equiv \neg B$.

$\neg A$ is the statement “$F$ is not unsatisfiable”, i.e. “$F$ is satisfiable”.
$\neg B$ is the statement “$F$ does not have a closed tableau”. Since, every formula has a complete tableau, this is equivalent to the statement “$F$ has a complete tableau branch”.
It thus remains to show: $F$ is satisfiable if and only if $F$ has a complete tableau branch. This was shown in Theorems 2.6.4 and 2.6.5.

2.6.11 Remark
In short, Statement 1 of Theorem 2.6.6 holds because it expresses the contrapositions of Theorem 2.6.4 and 2.6.5.

2.7 Theoretical Aspects
[Schöning, 1989, Part of Chapter 1.4 plus some more]

2.7.1 Theorem (monotonicity of propositional logic)
Let $M, N$ be sets of formulas. If $M \subseteq N$ then $\{F \mid M \models F\} \subseteq \{F \mid N \models F\}$.

Proof: Let $F$ be such that $M \models F$.
Let $\mathcal{A}$ be a model for $N$. Then all formulas in $N$, and hence all formulas in $M$, are true under $\mathcal{A}$. Hence $\mathcal{A} \models F$. This holds for all models of $N$, and hence $N \models F$.

2.7.2 Definition
A problem with a yes/no answer (a decision problem) is decidable if there exists an algorithm which terminates on any allowed input of the problem and, upon termination, outputs the correct answer.

2.7.3 Example
“Is $n$ an even number?” is decidable (allowed input: any $n \in \mathbb{N}$).

1. If $n=1$, terminate with output ’No’.
2. If $n=0$, terminate with output ’Yes’.
4. Go to 1.
2.7.4 Theorem (decidability of finite entailment)
The problem of deciding whether a finite set $M$ of formulas entails some other formula $F$ is decidable.

Proof: $M$ contains only a finite number of propositional variables. Use truth tables to check whether all models of $M$ are models of $F$. ■

2.7.5 Theorem (decidability of Datalog entailment)
The problem of deciding whether a finite Datalog program $P$ Herbrand-entails some $A \in B_P$ is decidable.

Proof: The set of propositional formulas $v(P)$ as defined in Definition 2.3.3 is finite. Theorems 2.3.5 and 2.7.4 then complete the proof. ■

2.7.6 Definition
A decision problem is semi-decidable if there exists an algorithm which, on any allowed input of the problem, terminates if the answer is 'yes' and outputs the correct answer.

2.7.7 Theorem (semi-decidability of infinite entailment)
The problem of deciding whether a countably infinite set $M$ of formulas entails some other formula $F$ is semi-decidable.

Proof: $M \models F$ if and only if $M \cup \{\neg F\}$ is unsatisfiable. [Exercise ??] By the compactness theorem, $M \cup \{\neg F\}$ is unsatisfiable if and only if one of its finite subsets is unsatisfiable. Now use an enumeration $M_1, M_2, \ldots$ of all these finite subsets and check satisfiability of each of them in turn, using truth tables. If one of the sets is unsatisfiable, terminate and output that $M \models F$. ■

2.7.8 Theorem (compactness of propositional logic)
A set $M$ of formulas is satisfiable if and only if every finite subset of it is satisfiable.

Proof:
⇒: Every model for $M$ is also a model for each finite subset of $M$.
⇐: Assume every finite subset of $M$ is satisfiable. 
Let $\{A_1, A_2, \ldots\}$ be all propositional variables.
Define $M_n$ to be the set of all elements of $M$ which contains only the propositional variables $A_1, \ldots, A_n$.
$M_n$ contains at most $2^{2^n}$ many formulas with different truth tables.
Thus, there is a set $\mathcal{F}_n = \{F_1, \ldots, F_k\} \subseteq M_n$ ($k \leq 2^{2^n}$), such that for every $F \in M$, $F \equiv F_i$ for some $i$.
Hence, every model for $\mathcal{F}_n$ is a model for $M_n$.
By assumption, $\mathcal{F}_n$ is satisfiable, say with model $\mathcal{A}_n$.
$\mathcal{A}_n$ is also a model for $M_1, \ldots, M_{n-1}$. [$M_i \subseteq M_{i+1}$ for all $i$]
For all $k \in \mathbb{N}$, define $\mathcal{A}(A_k) = \limsup_{n \to \infty} \mathcal{A}_n(A_k)$.

Note: For each $k \in \mathbb{N}$ there exists $n_k \in \mathbb{N}$ s.t. for all $n \geq n_k$ we have $\mathcal{A}_n(A_k) = \mathcal{A}_{n+1}(A_k)$.

It remains to show: $\mathcal{A} \models M$:

Let $F \in M$. Then $F \in M_k$ for some $k$.

With $n' = \max\{n_1, \ldots, n_k\}$ we have that $\mathcal{A}$ and all $\mathcal{A}_n$ with $n \geq n'$ agree on all propositional variables in $F$.

We have $\mathcal{A}_m \models F$ for all $m \geq \max\{k, n'\}$.

Hence $\mathcal{A} \models F$ as required.

\textbf{2.7.9 Theorem (complexity of finite satisfiability)}

The problem of deciding whether a finite set of formulas is satisfiable, is NP-complete.

\textbf{Proof:} See any book or lecture on computational complexity theory.

\textbf{2.7.10 Theorem (complexity of finite entailment)}

The problem of deciding whether a finite set of formulas entails some other formula is NP-complete.

\textbf{Proof:} Because of Exercise ??, finite entailment and finite satisfiability can be reduced to each other, hence they have the same complexity.
3 First-order Predicate Logic

3.1 Syntax

[Schöning, 1989, Chapter 2.1]

3.1.1 Example
Difficult/impossible to model in propositional logic:
• For all $n \in \mathbb{N}$, $n! \geq n$.

3.1.2 Example
Difficult/impossible to model in propositional logic:
1. Healthy beings are not dead.
2. Every cat is alive or dead.
3. If somebody owns something, (s)he cares for it.
4. A happy cat owner owns a cat and all beings he cares for are healthy.
5. Schrödinger is a happy cat owner.

3.1.3 Definition
• Variables: $x_1, x_2, \ldots$ (also $y, z, \ldots$).
• Function symbols: $f_1, f_2, \ldots$ (also $g, h, \ldots$), each with an arity $(\in \mathbb{N})$ (number of parameters).
  Constants are function symbols with arity 0.
• Predicate symbols: $P_1, P_2, \ldots$ (also $Q, R, \ldots$), each with an arity $(\in \mathbb{N})$ (number of parameters).

Terms are inductively defined:
• Each variable is a term.
• If $f$ is a function symbol of arity $k$, and if $t_1, \ldots, t_k$ are terms, then $f(t_1, \ldots, t_k)$ is a term.

Formulas are inductively defined:
• If $P$ is a predicate symbol of arity $k$, and if $t_1, \ldots, t_k$ are terms, then $P(t_1, \ldots, t_k)$ is a formula (called atomic).
• For each formula $F$, $\neg F$ is a formula.
• For all formulas $F$ and $G$, $(F \land G)$ and $(F \lor G)$ are formulas.
• If $x$ is a variable and $F$ is a formula, then $\exists x F$ and $\forall x F$ are formulas.

3.1.4 Definition
$F \rightarrow G$ (respectively, $F \leftrightarrow G$) is shorthand for $\neg F \lor G$ (respectively, $(F \rightarrow G) \land (G \rightarrow F)$).
We also use other notational variants from propositional logic freely.

3.1.5 Example
The following are formulas ($s$ is a constant).
1. $\forall x (H(x) \rightarrow \neg D(x))$
2. $\forall x (C(x) \rightarrow (A(x) \lor D(x)))$
3. $\forall x \forall y (O(x, y) \rightarrow R(x, y))$
4. \( \forall x (P(x) \rightarrow (\exists y (O(x,y) \land C(y))) \land (\forall y (R(x,y) \rightarrow H(y)))) \)

5. \( P(s) \)

In 1, predicate symbols are \( D \) and \( H \), and \( x \) is a term.

3.1.6 Example

Example 3.1.1 could be written as

\[
\forall n (n \in \mathbb{N} \rightarrow n! \geq n),
\]

where (with abuse of our introduced formal notation), “\( \in \)” is a unary predicate symbol, “\( \geq \)” is a binary predicate symbol, and “\( ! \)” is a unary function symbol, written postfix.

3.1.7 Definition

If a formula \( F \) is part of a formula \( G \), then it is called a subformula of \( G \).

An occurrence of a variable \( x \) in a formula \( F \) is bound if it occurs within a subformula of \( F \) of the form \( \exists x G \) or \( \forall x G \). Otherwise it is free.

A formula without free variables is closed. A formula with free variables is open. \( \exists, \forall \) are quantifiers, \( \lor, \land, \neg, \rightarrow, \leftrightarrow \) are connectives.

3.1.8 Example

All subformulas of \( \forall x (C(x) \rightarrow (A(x) \lor D(x))) \):

\( C(x), A(x), D(x), A(x) \lor D(x), C(x) \rightarrow (A(x) \lor D(x)), \forall x (C(x) \rightarrow (A(x) \lor D(x))) \).

3.1.9 Example

In the formula \( P(x) \land \forall x (P(x) \rightarrow Q(f(x))) \), the first occurrence of \( x \) is free, the others are bound.

3.2 Semantics

[Schöning, 1989, Chapter 2.1 cont.]

3.2.1 Definition

A structure is a pair \( \mathcal{A} = (U_{\mathcal{A}}, I_{\mathcal{A}}) \), with \( U_{\mathcal{A}} \neq \emptyset \) a set (ground set or universe) and \( I_{\mathcal{A}} \) a mapping which maps

- each \( k \)-ary predicate symbol \( P \) to a \( k \)-ary predicate (relation) on \( U_{\mathcal{A}} \) (if \( I_{\mathcal{A}} \) is defined for \( P \))
- each \( k \)-ary function symbol \( f \) to a \( k \)-ary function on \( U_{\mathcal{A}} \) (if \( I_{\mathcal{A}} \) is defined for \( f \))
- each variable \( x \) to an alement of \( U_{\mathcal{A}} \) (if \( I_{\mathcal{A}} \) is defined for \( x \)).

Write \( P^\mathcal{A} \) for \( I_{\mathcal{A}}(P) \) etc. \( \mathcal{A} \) is suitable for a formula \( F \) if \( I_{\mathcal{A}} \) is defined for all predicate and function symbols in \( F \) and for all free variables in \( F \).

3.2.2 Example

\[
F = \forall x \forall y (P(a) \land (P(x) \rightarrow (P(s(x)) \land Q(x,x) \land ((P(y) \land Q(x,y)) \rightarrow Q(x,s(y)))))
\]
Structure $(U_A, I_A)$:

\[
U_A = \mathbb{N} \\
\alpha^A = 0(\in \mathbb{N}) \\
s^A : n \mapsto n + 1 \\
P^A = \mathbb{N} \quad (= U_A) \\
Q^A = \{ (n, k) \mid n \leq k \}
\]

Another structure $(U_B, I_B)$:

\[
U_B = \{ \varnothing, \varnothing \} \\
\alpha^B = \varnothing \\
s^B : \varnothing \mapsto \varnothing; \varnothing \mapsto \varnothing \\
P^B = U_B \\
Q^B = \{ (\varnothing, \varnothing) \}
\]

3.2.3 Definition

A formula $A = (U_A, I_A)$ a suitable structure for $F$. Define for each term $t$ in $F$ its value $t^A$:

1. If $t = x$ is a variable, $t^A = x^A$.
2. If $t = f(t_1, \ldots, t_k)$, then $t^A = f^A(t_1^A, \ldots, t_k^A)$.

Define for $F$ its truth value $A(F)$ as follows, where $A_{[x/u]}$ is identical to $A$ except $x^A_{[x/u]} = u$.

1. $A(P(t_1, \ldots, t_k)) = \begin{cases} 1, & \text{if } (A(t_1), \ldots, A(t_k)) \in P^A \\ 0, & \text{otherwise} \end{cases}$
2. $A(H \land G) = \begin{cases} 1, & \text{if } A(H) = 1 \text{ and } A(G) = 1 \\ 0, & \text{otherwise} \end{cases}$
3. $A(H \lor G) = \begin{cases} 1, & \text{if } A(H) = 1 \text{ or } A(G) = 1 \\ 0, & \text{otherwise} \end{cases}$
4. $A(\neg G) = \begin{cases} 1, & \text{if } A(G) = 0 \\ 0, & \text{otherwise} \end{cases}$
5. $A(\forall xG) = \begin{cases} 1, & \text{if for all } u \in U_A, A_{[x/u]}(G) = 1 \\ 0, & \text{otherwise} \end{cases}$
6. $A(\exists xG) = \begin{cases} 1, & \text{if there exists some } u \in U_A \text{ s.t. } A_{[x/u]}(G) = 1 \\ 0, & \text{otherwise} \end{cases}$
If \( A(F) = 1 \), we write \( A \models F \) and say \( F \) is true in \( A \) or \( A \) is a model for \( F \).  
\( F \) is valid (or a tautology, written \( \models F \)) if \( A \models F \) for every suitable structure \( A \) for \( F \).  \( F \) is satisfiable if there is \( A \) with \( A \models F \), and otherwise it is unsatisfiable.

### 3.2.4 Remark

Many notions and results carry over directly from propositional logic: logical consequence, equivalence of formulas, Theorem 2.2.13, Theorem 2.4.4, etc. See Remark 3.2.10.

### 3.2.5 Example

Consider the sentences

- James Potter is the parent of Harry Potter.
- Harry Potter is an orphan.
- Any parent of any orphan is dead.

They can be represented formally as follows.

\[
\begin{align*}
\text{parentOf}(\text{jamespotter}, \text{harrypotter}) \land \text{orphan}(\text{harrypotter}) & \quad (22) \\
\land \forall x \forall y (\text{orphan}(x) \land \text{parentOf}(y, x) \rightarrow \text{dead}(y)) & \quad (24)
\end{align*}
\]

This has

\[
\text{dead}(\text{jamespotter})
\]
as logical consequence.

Proof sketch: From lines (22) and (23) we can conclude by the rule in (24) with \( x = \text{harrypotter} \) and \( y = \text{jamespotter} \) that \( \text{dead}(\text{harrypotter}) \).

Before we go for a formal proof, let's first give some examples for signatures—see Table 1.

Now for a formal proof: Let \( A \) be any model for the formula in (22)-(24). From (22) we then obtain

\[
(\text{jamespotter}^A, \text{harrypotter}^A) \in \text{parentOf}^A.
\]

From (23) we obtain

\[
\text{harrypotter}^A \in \text{orphan}^A.
\]

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<td>1</td>
<td>( a )</td>
<td>( h )</td>
<td>( h )</td>
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<tr>
<td>jamespotter(^A)</td>
<td>( j )</td>
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<td>2</td>
<td>( a )</td>
<td>( j )</td>
<td>( j )</td>
</tr>
<tr>
<td>orphan(^A)</td>
<td>( {h} )</td>
<td>( {1, 3, 4} )</td>
<td>( {3, 4, 5} )</td>
<td>( {a} )</td>
<td>( {h} )</td>
<td>( {h} )</td>
</tr>
<tr>
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<td>( {(2, 1)} )</td>
<td>( {(1, 2), (3, 1)} )</td>
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<td>( {(h, j)} )</td>
<td>( {(j, h)} )</td>
</tr>
<tr>
<td>dead(^A)</td>
<td>( {j} )</td>
<td>( {1, 2} )</td>
<td>( {1, 3, 4} )</td>
<td>( {a} )</td>
<td>( \emptyset )</td>
<td>( {h} )</td>
</tr>
</tbody>
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Table 1: Signatures for Example 3.2.5.
From (24) we obtain that, whenever

\[ u \in \text{orphan}^A \quad \text{and} \quad (v, u) \in \text{parentOf}^A, \]

then

\[ v \in \text{dead}^A. \]

So consequently

\[ \text{jamespotter}^A \in \text{dead}^A. \]

Since this argument holds for all models \( A \), we have that

\[ \text{dead(harrypotter)} \]

is indeed a logical consequence.

### 3.2.6 Example

\[
\text{parentOf(fatherOf(harrypotter), harrypotter)}
\]

\[ \land \text{orphan(harrypotter)} \]

\[ \land \forall x \forall y (\text{orphan}(x) \land \text{parentOf}(y, x) \rightarrow \text{dead}(y)) \]

has

\[ \text{dead(fatherOf(harrypotter))} \]

as logical consequence.

### 3.2.7 Example

\[
\text{human(harrypotter)} \land \text{orphan(harrypotter)}
\]

\[ \land \forall x (\text{human}(x) \rightarrow \text{parentOf}(\text{fatherOf}(x), x)) \]

\[ \land \forall x \forall y (\text{orphan}(x) \land \text{parentOf}(y, x) \rightarrow \text{dead}(y)) \]

has

\[ \text{dead(fatherOf(harrypotter))} \]

as logical consequence.

### 3.2.8 Example

\[ \forall x (\text{human}(x) \rightarrow \text{parentOf}(\text{fatherOf}(x), x)) \]

\[ \land \forall x \forall y (\text{orphan}(x) \land \text{parentOf}(y, x) \rightarrow \text{dead}(y)) \]

has

\[ \forall x (\text{human}(x) \land \text{orphan}(x) \rightarrow \text{dead(\text{fatherOf}(x))}) \]

as logical consequence.
3.2.9 Example
Consider the formula $F = \exists x \forall y Q(x, y)$ under the structure $A = (U_A, I_A)$ from Example 3.2.2. We show $A(F) = 1$.
First note that $0 \leq n$ for all $n \in \mathbb{N}$, i.e. $A_{[x/0][y/n]}(Q(x, y)) = 1$ for all $n \in \mathbb{N} = U_A$. Thus, $A_{[x/0]}(\forall y Q(x, y)) = 1$ and therefore $A(\exists x \forall y Q(x, y)) = 1$ as desired.

3.2.10 Remark
Predicate logic “degenerates” to propositional logic if either all predicate symbols have arity 0, or if no variables are used. For the latter, a formula like $(Q(a) \land \neg R(f(b), c)) \land P(a, b)$ can be written as the propositional formula $(A \land \neg B) \land C$ with $A$ for $Q(a)$, $B$ for $R(f(b), c)$, and $C$ for $P(a, b)$.

3.2.11 Remark
We deal with first-order predicate logic. Second-order predicate logic also allows to quantify over predicate symbols.

3.3 Datalog Revisited
3.3.1 Definition
Given a Datalog rule $r = B_1 \land \cdots \land B_k \rightarrow A$, let $\pi(r)$ be the formula
$$\forall x_1 \ldots \forall x_n (B_1 \land \cdots \land B_k \rightarrow A)$$
where $x_1, \ldots, x_n$ are (all) the variables occurring in $r$.
Given a Datalog program $P$, let $\pi(P) = \{ \pi(r) \mid r \in P \}$.

3.3.2 Theorem
Let $P$ be any Datalog program and let $A \in B_P$. Then $P \models_H A$ if and only if $\pi(P) \models A$.

3.4 Equivalence
[Schöning, 1989, Chapter 2.2]
3.4.1 Theorem
The following hold for arbitrary formulas $F$ and $G$.

$$\neg \forall x F \equiv \exists x \neg F \quad \quad \neg \exists x F \equiv \forall x \neg F$$
$$\forall x F \land \forall x G \equiv \forall x (F \land G) \quad \quad \exists x F \lor \exists x G \equiv \exists x (F \lor G)$$
$$\forall x \forall y F \equiv \forall y \forall x F \quad \quad \exists x \exists y F \equiv \exists y \exists x F$$

If $x$ does not occur free in $G$, then

$$\forall x F \land G \equiv \forall x (F \land G) \quad \quad \forall x F \lor G \equiv \forall x (F \lor G)$$
$$\exists x F \land G \equiv \exists x (F \land G) \quad \quad \exists x F \lor G \equiv \exists x (F \lor G)$$
Proof: We show only $\forall x F \land \forall x G \equiv \forall x (F \land G)$:

\[ A(\forall x F \land \forall x G) = 1 \]

iff $A(\forall x F) = 1$ and $A(\forall x G) = 1$

iff for all $u \in U_A$, $A_{[x/u]}(F) = 1$ and for all $v \in U_A$, $A_{[x/u]}(G) = 1$

iff $A(\forall x (F \land G)) = 1$ \[ \blacksquare \]

3.4.2 Definition
A substitution $[x/t]$, where $x$ is a variable and $t$ a term, is a mapping which maps each formula $G$ to the formula $G[x/t]$, which is obtained from $G$ by replacing all free occurrences of $x$ by $t$.

3.4.3 Example
$(P(x, y) \land \forall y Q(x, y))[x/a][y/f(x)] = P(a, f(x)) \land \forall y Q(a, y)$

3.5 Normal Forms

[Schöning, 1989, Chapter 2.2 cont.]

3.5.1 Definition
A literal is an atomic formula (a positive literal) or the negation of an atomic formula (a negative literal).

A formula $F$ is in negation normal form (NNF) if the negation symbol $\neg$ occurs only in literals (and $\to$, $\leftrightarrow$ don’t appear in it).

3.5.2 Theorem
For every formula $F$, there is a formula $G \equiv F$ which is in NNF.

Proof: Apply de Morgan, double negation, and $\neg \forall x F \equiv \exists x \neg F$ and $\neg \exists x F \equiv \forall x \neg F$ exhaustively. \[ \blacksquare \]

3.5.3 Example
$\neg(\exists x P(x, y) \lor \forall z Q(z)) \land \neg \exists w P(f(a, w))$
\[ \equiv (\neg \exists x P(x, y) \land \neg \forall z Q(z)) \land \forall w \neg P(f(a, w)) \]
\[ \equiv (\forall x \neg P(x, y) \land \exists z \neg Q(z)) \land \forall w \neg P(f(a, w)) \]

3.6 Tableaux Algorithm

[Ben-Ari, 1993, Chapter 5.5, strongly modified]

3.6.1 Definition
Let $F$ be a formula in NNF. A tableau branch for $F$ is a set of formulas, defined inductively as follows.

- $\{F\}$ is a tableau branch for $F$. 

• If $T$ is a tableau branch for $F$ and $G \land H \in T$, then $T \cup \{G, H\}$ is a tableau branch for $F$.
• If $T$ is a tableau branch for $F$ and $G \lor H \in T$, then $T \cup \{G\}$ is a tableau branch for $F$ and $T \cup \{H\}$ is a tableau branch for $F$.
• If $T$ is a tableau branch for $F$ and $\forall x G \in T$, then $T \cup \{G[x/t]\}$ is a tableau branch for $F$, where $t$ is any term.
• If $T$ is a tableau branch for $F$ and $\exists x G \in T$, then $T \cup \{G[x/a]\}$ is a tableau branch for $F$, where $a$ is a constant symbol which does not occur in $T$ (or in the tableau currently constructed).

A tableau for $F$ is a set of tableau branches for $F$.
A tableau branch is closed if it contains an atomic formula $A$ and its negation $\neg A$. Otherwise, it is open.
A tableau $M$ for $F$ is called closed if for each $T \in M$ there is a closed $T' \in M$ with $T \subseteq T'$.

If $F$ is not in NNF, then a tableau (resp., tableau branch) for $F$ is a tableau (resp. tableau branch) for an NNF of $F$.

3.6.2 Theorem (Soundness)
If a closed formula $F$ has a closed tableau, then $F$ is unsatisfiable.

3.6.3 Theorem (Completeness)
If a closed formula $F$ is unsatisfiable, then there is a closed tableau for $F$.

3.6.4 Example
We show $\exists u \forall v (P(v, u) \models \forall x \exists y P(x, y))$. I.e. we make a tableau for

$$\exists u \forall v P(v, u) \land \exists x \forall y \neg P(x, y),$$

see Figure 2 (left).

3.6.5 Example
We show, that

$$\exists x \exists y (P(x) \lor Q(y)) \models \exists x (P(x) \lor Q(x)).$$

[done on whiteboard]

3.6.6 Example
We show, that

$$\forall x \exists y (P(x) \land Q(y)) \equiv \exists y \forall x (P(x) \land Q(y)).$$

[done on whiteboard]

3.6.7 Example
We show, that

$$\forall x (P(x) \rightarrow (\exists y (O(x, y) \land C(y)) \land (\forall z (R(x, z) \rightarrow H(z))))).$$
Figure 2: Tableaux for Example 3.6.4 (left) and Remark 3.6.8 (right).

The (predicate logic) tableaux algorithm does not in general provide a means to find out if a formula is satisfiable or falsifiable.

Consider $\forall x \exists y P(x, y) \models \exists u \forall v P(v, u)$. If we attempt to make a tableau for

$$\forall x \exists y P(x, y) \land \forall u \exists v \neg P(v, u),$$

see for example Figure 2, then the search for closing the tableau does not stop. The reason for this is that the tableau cannot close, but the occurrence of the quantifiers in the formula prompts the algorithm to ever explore new terms for the bound variables.

3.6.9 Remark
While the propositional tableaux algorithm always terminates, this is not the case for the predicate logic tableaux algorithm.
3.7 Theoretical Aspects

[Schöning, 1989, Part of Chapter 2.3 plus additional material]

3.7.1 Theorem (monotonicity of predicate logic)
Let $M, N$ be sets of formulas. If $M \subseteq N$ then $\{ F \mid M \models F \} \subseteq \{ F \mid N \models F \}$.

Proof: Similar as for propositional logic. ■

3.7.2 Theorem (compactness of predicate logic)
A set $M$ of formulas is satisfiable if and only if every finite subset of it is satisfiable.

3.7.3 Theorem (undecidability of predicate logic)
The problem “Given a formula $F$, is $F$ valid?” is undecidable.

Proof: Out of scope for this lecture. Usually by reduction of the Halting Problem. ■

3.7.4 Theorem (semi-decidability of predicate logic)
The problem “Given a formula $F$, is $F$ valid?” is semi-decidable.

Proof: We have, e.g., the tableaux calculus for this. ■

3.7.5 Remark
The formula
$$F = \forall x \forall y \forall u \forall v \forall w (P(x, f(x)) \land \neg P(y, y) \land ((P(u, v) \land P(v, w)) \rightarrow P(u, w))$$

is satisfiable but has no finite model (with $U_A$ finite).

$A = (U_A, I_A)$ is a model, where

$$U_A = \mathbb{N}$$
$$P^A = \{(m, n) \mid m < n\}$$
$$f^A(n) = n + 1$$

Assume $B = (U_B, I_B)$ is a finite model for $F$. Let $u_0 \in U_B$ and consider the sequence $(u_i)_{i \in \mathbb{N}}$ with $u_{i+1} = f^B(u_i)$. Since $U_B$ is finite, there exist $i < j$ with $u_i = u_j$. $F$ enforces transitivity of $F$, hence $(u_i, u_j) \in P^B$. But since $u_i = u_j$ this contradicts $\forall y \neg P(y, y)$.

3.7.6 Theorem (Löwenheim-Skolem)
If a (finite or) countable set of formulas is satisfiable, then it is satisfiable in a countable domain.

3.7.7 Remark
According to Theorem 3.7.6, it is impossible to axiomatize the real numbers in first-order predicate logic.
4 Application: Knowledge Representation for the World Wide Web

[See Hitzler et al., 2009 for further reading.]

[Slideset 2]