



# **Artificial Intelligence**

# 6. First Order Logic Inference

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#### **1. Unification**

- 2. Forward Chaining
- 3. Backward Chaining
- 4. Resolution

#### "Compound Expressions"

# Formulas and function terms sometimes are described as **compound expressions**.

For a compound expression, its operator and its arguments is defined:

 $\begin{array}{ll} \operatorname{op}(P(t_1,\ldots,t_n)) \coloneqq P & \operatorname{args}(P(t_1,\ldots,t_n)) \coloneqq (t_1,\ldots,t_n) \\ \operatorname{op}(f(t_1,\ldots,t_n)) \coloneqq f & \operatorname{args}(f(t_1,\ldots,t_n)) \coloneqq (t_1,\ldots,t_n) \\ \operatorname{op}(\neg \phi) \coloneqq \neg & \operatorname{args}(\neg \phi) \coloneqq (\phi) \\ \operatorname{op}(\phi \oplus \psi) \coloneqq \oplus & \operatorname{args}(\phi \oplus \psi) \coloneqq (\phi,\psi), \quad \oplus \in \{\wedge,\vee,\rightarrow,\leftrightarrow\} \\ \operatorname{op}(\forall x \phi) \coloneqq \forall & \operatorname{args}(\forall x \phi) \coloneqq (x,\phi) \\ \operatorname{op}(\exists x \phi) \coloneqq \exists & \operatorname{args}(\exists x \phi) \coloneqq (x,\phi) \end{array}$ 

Atomic terms, i.e., constants and variables, are not considered compound expressions.



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#### Unification / Algorithm



*i* unify $(x, y, \theta)$ : 2 **if**  $\theta$  = failure return failure 3 4 elsif x = yreturn  $\theta$ 5 6 elsif is-variable(x)<u>**return**</u> unify-var $(x, y, \theta)$ 7 8 **<u>elsif</u>** is-variable(y)<u>**return**</u> unify-var $(y, x, \theta)$ 9 elsif is-compound(x) and is-compound(y) 10 return  $unify(args(x), args(y), unify(op(x), op(y), \theta))$ 11 12 **<u>elsif</u>** is-list(x) and is-list(y) <u>**return**</u> unify $((x_2,\ldots,x_n),(y_2,\ldots,y_n),$  unify $(x_1,y_1,\theta))$ 13 14 else return failure 15 16 **fi** 17 18 unify-var(var,  $x, \theta$ ) : 19 if  $\theta(var) \neq \emptyset$ <u>**return**</u> unify $(\theta(var), x, \theta)$ 20 21 elsif  $\theta(x) \neq \emptyset$ **return** *unify*(*var*,  $\theta(x), \theta$ ) 22 elsif occurs(var, x)23 return failure 24 25 else return  $\theta \cup \{ var \mapsto x \}$ 26

#### Unification / Example



 $\mathsf{unify}(\mathsf{Knows}(\mathsf{John}, x), \mathsf{Knows}(y, \mathsf{Mother}(y)), \emptyset)$ 

- $= \mathsf{unify}((\mathsf{John}, x), (y, \mathsf{Mother}(y)), \mathsf{unify}(\mathsf{Knows}, \mathsf{Knows}, \emptyset))$
- $= \mathsf{unify}((\mathsf{John}, x), (y, \mathsf{Mother}(y)), \emptyset)$
- $= \mathsf{unify}((x), (\mathsf{Mother}(y)), \mathsf{unify}(\mathsf{John}, y, \emptyset))$
- $= \mathsf{unify}((x), (\mathsf{Mother}(y)), \{y/\mathsf{John}\})$
- = unify-var $(x, Mother(y), \{y/John\})$
- $= \{y/\mathsf{John}, x/\mathsf{Mother}(y)\}$



1. Unification

**2. Forward Chaining** 

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#### **Generalized Modus Ponens**



premise	conclusion	name
$\overline{\mathcal{F} \vdash F, \mathcal{F} \vdash F \to G}$	$\mathcal{F} \vdash G$	ightarrow -elimination / modus ponens
$\mathcal{F} \vdash F$	$\mathcal{F} \vdash F\theta$	universial instantiation
$\mathcal{F} \vdash F, \mathcal{F} \vdash F' \to G, F\theta = F'\theta$	$\theta \ \mathcal{F} \vdash G \theta$	generalized modus ponens

#### Lemma 1. Generalized modus ponens is sound.

Proof.

1. $\mathcal{F} \vdash F$	[assumption]
2. $\mathcal{F} \vdash F\theta$	[universal instantiation applied to 1]
3. $\mathcal{F} \vdash F' \to G$	[assumption]
4. $\mathcal{F} \vdash F'\theta \to G\theta$	[universal instantiation applied to 3]
5. $\mathcal{F} \vdash G\theta$	[ $ ightarrow$ -elemination applied to 2,4]

Artificial Intelligence / 2. Forward Chaining

Generalized Modus Ponens / Example

Let the knowledge base  ${\mathcal F}$  be

$$\begin{split} \operatorname{King}(x) \wedge \operatorname{Greedy}(x) & \to \operatorname{Evil}(x) \\ \operatorname{King}(\operatorname{John}) \\ \operatorname{Greedy}(y) \end{split}$$

#### Now use

$$F := \mathsf{King}(\mathsf{John}) \land \mathsf{Greedy}(y)$$
$$F' := \mathsf{King}(x) \land \mathsf{Greedy}(x)$$
$$G := \mathsf{Evil}(x)$$

then for

$$\theta := \{x / \mathsf{John}, y / \mathsf{John}\}$$

we have

$$F\theta = \text{King(John)} \land \text{Greedy(John)} = F'\theta$$

and thus we can derive

 $G\theta = \text{Evil}(\text{John})$ 

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## Forward Chaining



Definitions for **conjunctive normal forms** (CNF), **Horn clauses** and **Horn formulas** are the same as in propositional logic. Here, atoms are formulas

$$P(t_1, t_2, \ldots, t_n)$$

where P is a predicate symbol and  $t_i$  are any terms (including variables).

A Horn clause *C* is called **definite** it it contains exactly one positive literal, i.e., implications of type

$$(\bigvee_{i=1}^n \neg L_i) \equiv (\bigwedge_{i=1}^n L_i \to \mathsf{false})$$

are not possible.

If the knowledge base consists of **Horn clauses** only, then generalized modus ponens can be used just like modus ponens to infer statements iteratively by forward chaining.

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#### Example



The law says that it is a crime for an American to sell weapons to hostile nations. The country Nono, an enemy of America, has some missiles, and all of its missiles were sold to it by Colonel West, who is American.

Prove that Col. West is a criminal.



Example (2/4)

The law says that it is a crime for an American to sell weapons to hostile nations.

 $\forall \mathsf{American}(x) \land \mathsf{Weapon}(y) \land \mathsf{Hostile}(z) \land \mathsf{Sell}(x,y,z) \rightarrow \mathsf{Criminal}(x)$ 

The country Nono,

Country(Nono)

```
an enemy of America,
```

Enemy(Nono, America)

has some missiles,

 $\exists x \mathsf{Missile}(x) \land \mathsf{Owns}(\mathsf{Nono}, x)$ 

and all of its missiles were sold to it by Colonel West,

 $\forall x \mathsf{Missile}(x) \land \mathsf{Owns}(\mathsf{Nono}, x) \rightarrow \mathsf{Sell}(\mathsf{West}, x, \mathsf{Nono})$ 

who is American.

American(West)

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Example (3/4)



Additional background knowledge: Missiles are weapons.

```
\forall x \mathsf{Missile}(x) \rightarrow \mathsf{Weapon}(x)
```

Enemies of America are hostile.

 $\forall x \texttt{Enemy}(x, \texttt{America}) \rightarrow \texttt{Hostile}(x)$ 

Prove that Col. West is a criminal

Criminal(West)?



#### Example (4/4)

The knowledge base can be simplified by

- existential instantiation and
- omitting universal quantifiers

   (as all free variables are universally quantified anyway)

```
\begin{array}{l} \mathsf{American}(x) \land \mathsf{Weapon}(y) \land \mathsf{Hostile}(z) \land \mathsf{Sell}(x,y,z) \rightarrow \mathsf{Criminal}(x) \\ \mathsf{Country}(\mathsf{Nono}) \\ \mathsf{Enemy}(\mathsf{Nono}, \mathsf{America}) \\ \mathsf{Missile}(M_1) \land \mathsf{Owns}(\mathsf{Nono}, M_1) \\ \mathsf{Missile}(x) \land \mathsf{Owns}(\mathsf{Nono}, x) \rightarrow \mathsf{Sell}(\mathsf{West}, x, \mathsf{Nono}) \\ \mathsf{American}(\mathsf{West}) \\ \mathsf{Missile}(x) \rightarrow \mathsf{Weapon}(x) \\ \mathsf{Enemy}(x, \mathsf{America}) \rightarrow \mathsf{Hostile}(x) \end{array}
```

#### $\rightsquigarrow$ This knowledge base consists of definite Horn clauses only !

#### Forward Chaining



```
i entails-fc(FOL definite horn formula F, query atom Q) :
 2 \mathcal{C} := \emptyset
 \mathcal{C}' := \operatorname{clauses}(F)
 4 while C' \neq \emptyset do
                \mathcal{C} := \mathcal{C} \cup \mathcal{C}'
 5
                \mathcal{C}' := \emptyset
 6
               for C \in \mathcal{C} do
 7
                      C' := \text{standardize-apart}(C)
 8
                      <u>for</u> atoms A_1, A_2, \ldots, A_n \in \mathcal{C} and \theta with body(C')\theta = (A_1 \land A_2 \land \ldots \land A_n)\theta <u>do</u>
 9
                             H := \text{head}(C')\theta
10
                             \underline{\mathbf{if}} \ H \notin \mathcal{C} \text{ and } H \notin \mathcal{C}'
11
                                 \mathcal{C}' := \mathcal{C}' \cup \{H\}
12
                                 \underline{if} unify(H, Q) <u>return</u> true <u>fi</u>
13
                             <u>fi</u>
14
                       od
15
                od
16
17 od
18 return false
```

#### Forward Chaining / Example



American(West)

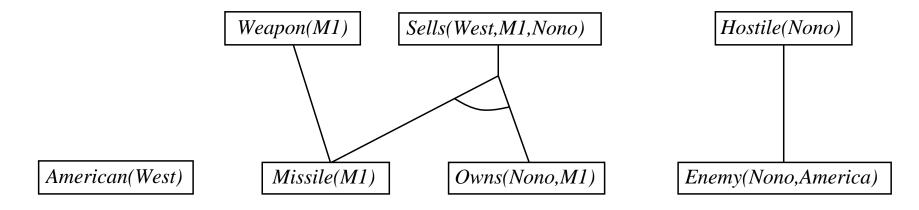
Missile(M1)

Owns(Nono,M1)

Enemy(Nono,America)

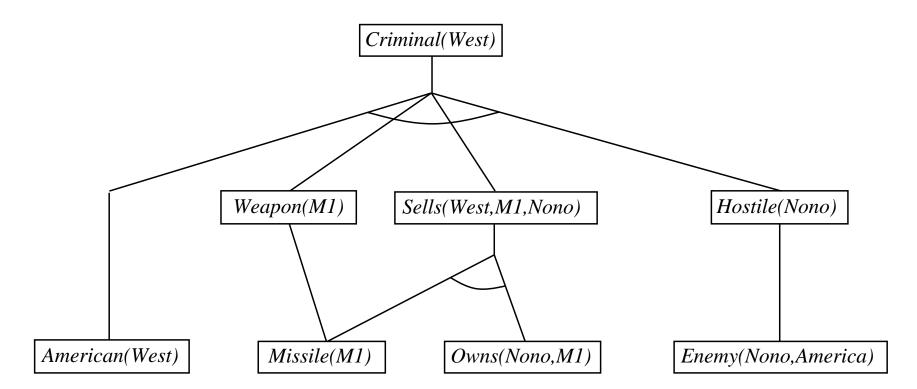
#### Forward Chaining / Example





#### Forward Chaining / Example







1. Unification

2. Forward Chaining

**3. Backward Chaining** 

4. Resolution

#### **Backward Chaining**



Backward chaining works the other way around:

- keep a list of yet unsatisfied atoms Q
  starting with the query atom.
- try to find rules whichs head match atoms in Q (after unification) and replace the atom from Q by the atoms of the body of the matching rule.
- proceed recursively until no more atoms have to be satisfied.

Backward chaining keeps track of the substitution needed during the proof.

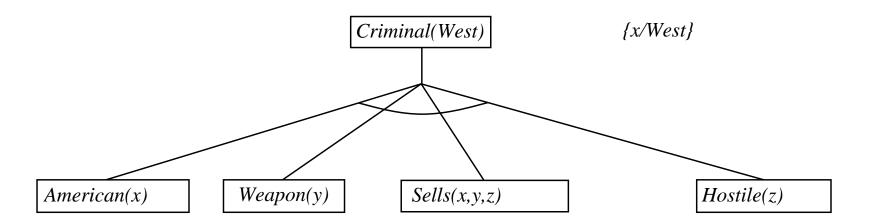
#### Backward Chaining / Algorithm



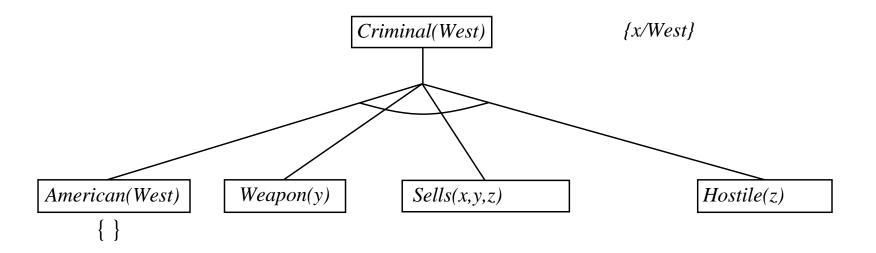
```
i entails-bc(FOL definite horn formula F, query atom Q) :
 2 <u>return</u> entails-bc-goals(clauses(F), {Q}, \emptyset) \neq \emptyset
 3
4 entails-bc-goals(set of FOL definite Horn clauses C, set of FOL atoms Q, \theta) :
 5 \operatorname{if} Q = \emptyset \operatorname{return} \{\theta\} \operatorname{fi}
 \Theta := \emptyset
 7 for C \in \mathcal{C} do
         C' := \text{standardize-apart}(C)
 8
         \theta' := unify(head(C'), Q[1]\theta)
 9
         <u>if</u> \theta' \neq failure
10
            \Theta := \Theta \cup \text{entails-bc-goals}(\mathcal{C}, \operatorname{atoms}(\operatorname{body}(C')) \cup (Q \setminus \{C\}), \theta \cup \theta')
11
         fi
12
13 od
14 return \Theta
```



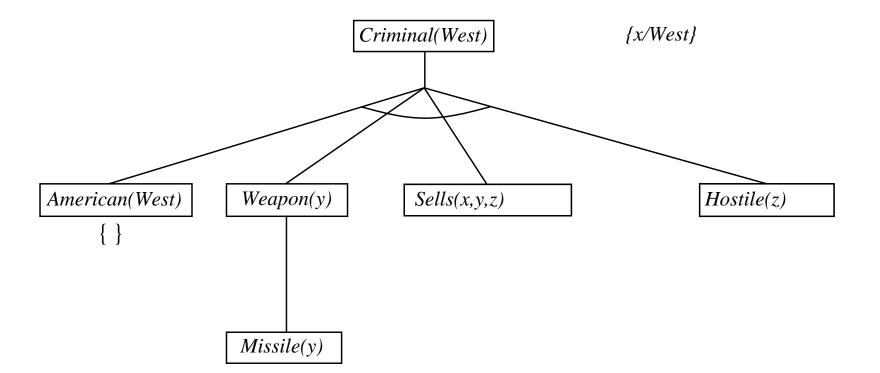
Criminal(West)



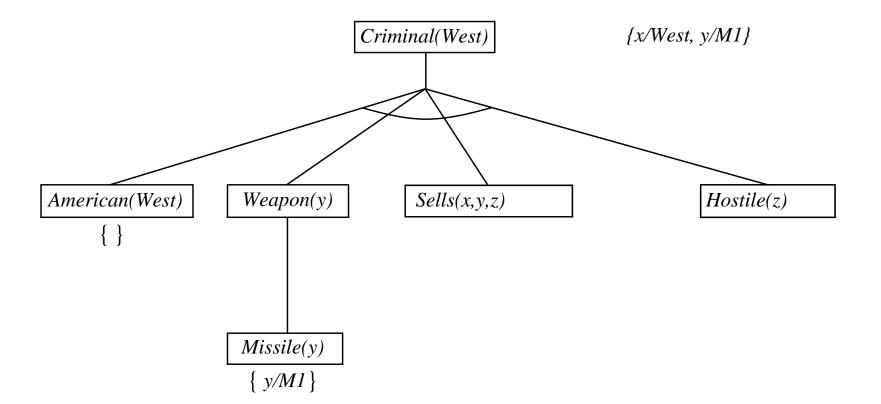






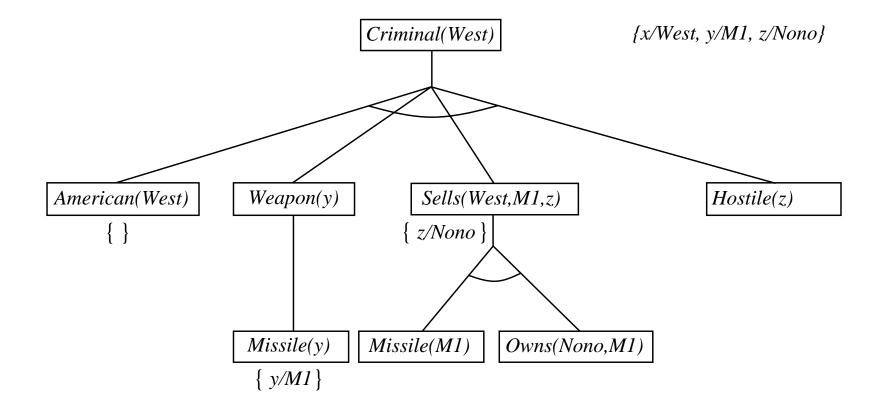




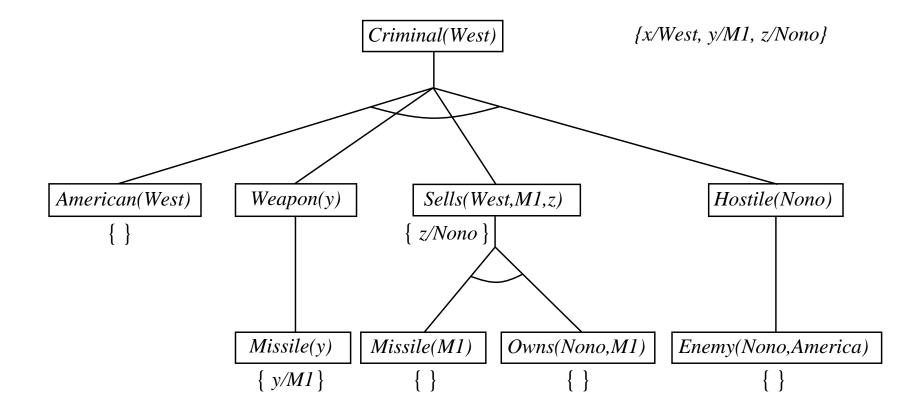












Logic Programming: Prolog



Prolog: logical programming language (PROgrammation en LOGique; Alain Colmerauer and Philippe Roussel, ca. 1972)

Allows knowlegde bases (= programs) consisting of definite Horn clauses.

Uses depth-first, left-to-right backward chaining (with several improvements).

Example:

```
evil(X) :- king(X), greedy(X).
king(john).
greedy(X).
?- evil(john)
```

## Negation as Failure



Prolog allows the usage of negated atoms in rule bodies interpreting them by **negation as failure**:

```
good(X) :- not evil(X)
```

Now the query ?- good (richard) would evaluate to true as the opposite, evil (richard) cannot be proved.

This is also called **closed world assumption**: if a fact is not encoded in the knowledge base and cannot be inferred, then it is considered not to be true.

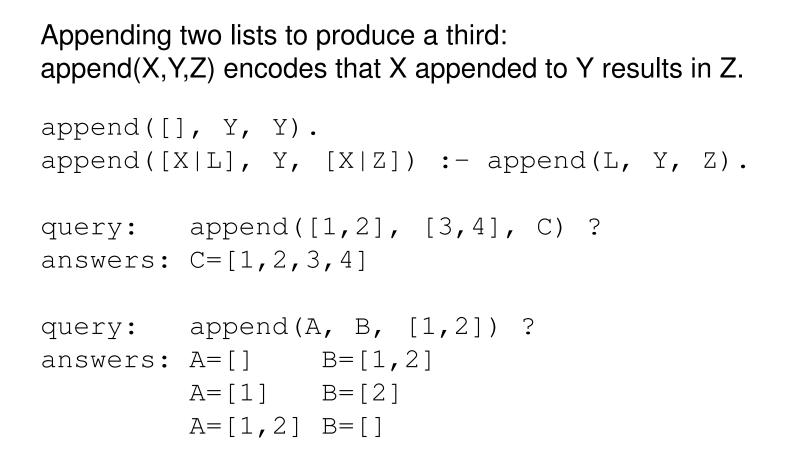
Negation as failure renders Prolog **non-monotonic**: if one adds formulas to the knowledge base, inferences may become untrue.

Example: add evil (richard) to the knowledge base, now the query ?- good (richard) evaluates to false.

In first order logics we could not derive any conclusions about good (richard).

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#### Prolog / Examples







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#### FOL Resolvents



Full first-order version:

$$\frac{\ell_1 \vee \cdots \vee \ell_k, \quad m_1 \vee \cdots \vee m_n}{(\ell_1 \vee \cdots \vee \ell_{i-1} \vee \ell_{i+1} \vee \cdots \vee \ell_k \vee m_1 \vee \cdots \vee m_{j-1} \vee m_{j+1} \vee \cdots \vee m_n)\theta}$$
  
where Unify $(\ell_i, \neg m_j) = \theta$ .

For example,

$$\frac{\neg \mathsf{Rich}(x) \lor \mathsf{Unhappy}(x), \quad \mathsf{Rich}(\mathsf{Ken})}{\mathsf{Unhappy}(\mathsf{Ken})}$$

with  $\ell_i = \neg \operatorname{Rich}(x)$ ,  $m_j = \operatorname{Rich}(\operatorname{Ken})$  and  $\theta = \{x/\operatorname{Ken}\}$ 

Apply resolution steps to  $CNF(KB \land \neg query)$ ; complete for FOL.

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#### Conversion to CNF

2003

Everyone who loves all animals is loved by someone: $\forall x [\forall y \text{Animal}(y) \implies \text{Loves}(x, y)] \implies [\exists y \text{Loves}(y, x)]$ 

1. Eliminate biconditionals and implications

 $\forall x [\neg \forall y \neg \mathsf{Animal}(y) \lor \mathsf{Loves}(x, y)] \lor [\exists y \mathsf{Loves}(y, x)]$ 

**2.** Move  $\neg$  inwards:  $\neg \forall x, p \equiv \exists x \neg p$ ,  $\neg \exists x, p \equiv \forall x \neg p$ :

$$\begin{aligned} \forall x [\exists y \neg (\neg \mathsf{Animal}(y) \lor \mathsf{Loves}(x, y))] \lor [\exists y \mathsf{Loves}(y, x)] \\ \forall x [\exists y \neg \neg \mathsf{Animal}(y) \land \neg \mathsf{Loves}(x, y)] \lor [\exists y \mathsf{Loves}(y, x)] \\ \forall x [\exists y \mathsf{Animal}(y) \land \neg \mathsf{Loves}(x, y)] \lor [\exists y \mathsf{Loves}(y, x)] \end{aligned}$$

#### Conversion to CNF

Youny 2003

3. Standardize variables: each quantifier should use a different one

 $\forall x [\exists y \mathsf{Animal}(y) \land \neg \mathsf{Loves}(x,y)] \lor [\exists z \mathsf{Loves}(z,x)]$ 

4. Skolemize: a more general form of existential instantiation. Each existential variable is replaced by a **Skolem function** of the enclosing universally quantified variables:

 $\forall x [\mathsf{Animal}(F(x)) \land \neg \mathsf{Loves}(x,F(x))] \lor \mathsf{Loves}(G(x),x)$ 

5. Drop universal quantifiers:

 $[\mathsf{Animal}(F(x)) \land \neg \mathsf{Loves}(x,F(x))] \lor \mathsf{Loves}(G(x),x)$ 

6. Distribute  $\land$  over  $\lor$ :

 $[\mathsf{Animal}(F(x)) \lor \mathsf{Loves}(G(x), x)] \land [\neg \mathsf{Loves}(x, F(x)) \lor \mathsf{Loves}(G(x), x)]$ 

#### Resolution / Example

