## Artificial Intelligence

A Course About Foundations



#### Horn Clauses and SLD Resolution

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Artificial Intelligence 2024–2025 SLD Resolution [1]

### Back to Propositional Logic

Artificial Intelligence 2024–2025 SLD Resolution [2]

### Horn Clauses (in $L_P$ )

Definition

A *Horn Clause* is a wff in CF that contains at most <u>one</u> literal in positive form

Three types of Horn Clauses:

Rule: two or more literals, one positive

Examples:  $\{B, \neg D, \neg A, \neg C\}, \{A, \neg B\}$  (equivalent to:  $(D \land A \land C) \rightarrow B, B \rightarrow A$ )

Facts: just one positive literal

Examples:  $\{B\}$ ,  $\{A\}$ 

*Goal*: one or more literals, all negative

Examples:  $\{\neg B\}$ ,  $\{\neg A, \neg B\}$ 

#### More terminology:

Rules and facts are also called definite clauses

Goals are allo called *negative clauses* 

### Lost in Translation...

#### Many wffs can be translated into Horn clauses:

```
(A \land B) \rightarrow C
                                                                     (rewriting \rightarrow)
    \neg (A \land B) \lor C
     \neg A \lor \neg B \lor C
                                                                     (De Morgan - CF – it is a rule)
A \rightarrow (B \land C)
     \neg A \lor (B \land C)
                                                                     (rewriting \rightarrow)
    (\neg A \lor B) \land (\neg A \lor C)
                                                                      (distributing V)
    (\neg A \lor B), (\neg A \lor C)
                                                                     (CF – two rules)
(A \lor B) \to C
     \neg (A \lor B) \lor C
                                                                     (rewriting \rightarrow)
    (\neg A \land \neg B) \lor C
                                                                     (De Morgan)
     (\neg A \lor C) \land (\neg B \lor C)
                                                                     (distributing V)
     (\neg A \lor C), (\neg B \lor C)
                                                                     (CF – two rules)
```

#### But not all of them:

$$(A \land \neg B) \rightarrow C$$
  
 $\neg (A \land \neg B) \lor C$   
 $\neg A \lor B \lor C$   
 $A \rightarrow (B \lor C)$   
 $\neg A \lor B \lor C$   
 $(rewriting \rightarrow)$ 

### SLD Resolution

Linear resolution with Selection function for Definite clauses

#### Algorithm

Starts from a set of definite clauses (also the program) + a goal

- 1) At each step, the selection function identifies a literal in the goal (i.e. subgoal)
- 2) All definite clause applicable to the subgoal are selected, in the given order
- 3) The resolution rule is applied generating the resolvent

Termination: either the empty clause { } is obtained or step 2) fails.

#### Example:

Selection function: leftmost subgoal first Definite clauses:  $\{C\}$ ,  $\{D\}$ ,  $\{B, \neg D\}$ ,  $\{A, \neg B, \neg C\}$  Goal:  $\{\neg A\}$ 

$$\{ \neg A \}$$

$$\{ \neg B, \neg C \}$$

$$\{ \neg B, \neg C \}$$

$$\{ B, \neg D \}$$

$$\{ \neg D, \neg C \}$$

$$\{ D \}$$

$$\{ \neg C \}$$

$$\{ C \}$$

$$\{ \}$$

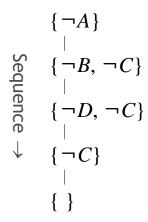
Artificial Intelligence 2024–2025 SLD Resolution [5]

#### SLD trees

#### **SLD** derivations

Example:  $\{C\}$ ,  $\{D\}$ ,  $\{B, \neg D\}$ ,  $\{A, \neg B, \neg C\}$  goal  $\{\neg A\}$ 

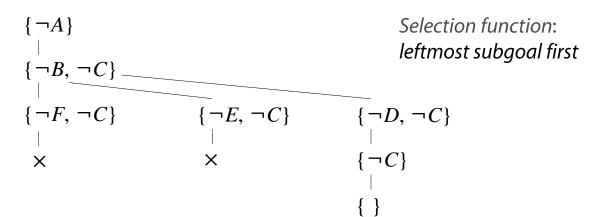
In this example each subgoal can be resolved in one mode only This is not true in general



SLD trees (= trace of all SLD derivations from a goal)

Example: 
$$\{C\}$$
,  $\{D\}$ ,  $\{B, \neg F\}$ ,  $\{B, \neg E\}$ ,  $\{B, \neg D\}$ ,  $\{A, \neg B, \neg C\}$  goal  $\{\neg A\}$ 

A few new rules have been added: there are now different possibilities



Each branch correspond to a possible resolution for a *subgoal* 

Artificial Intelligence 2024–2025 SLD Resolution [6]

### SLD Resolution

• A resolution method for Horn clauses in  $L_P$ 

It always terminates

It is *correct*:  $\Gamma \vdash \varphi \Rightarrow \Gamma \models \varphi$ 

It is complete:  $\Gamma \models \varphi \Rightarrow \Gamma \vdash \varphi$ 

Computationally efficient

It has polynomial time complexity (w.r.t the # of propositional symbols occurring in  $\Gamma$  and  $\varphi$ )

Artificial Intelligence 2024–2025 SLD Resolution [7]

# SLD resolution in First-Order Logic

### Horn Clauses in $L_{FO}$

The definition is very similar to the propositional case

Horn Clauses (of the skolemization of a set sentences)

Each clause contains at most one literal in positive form

```
Facts, rules and goals
```

**Fact**: a clause with just an individual *atom* 

```
\{Greek(socrates)\}, \{Pyramid(x)\}, \{Sister(sally, motherOf(paul))\}
```

Rule: a clause with at least two literals, exactly one in positive form

```
\{Human(x), \neg Greek(x)\},\

\forall x (Greek(x) \rightarrow Human(x))

\{\neg Female(x), \neg Parent(k(x),x), \neg Parent(k(y),y), Sister(x,y)\}

\forall x \forall y ((Female(x) \land \exists z (Parent(z,x) \land Parent(z,y))) \rightarrow Sister(x,y))

\{\neg Above(x,y), On(x,k(x))\}, \{\neg Above(x,y), On(j(y),y)\}

\forall x \forall y (Above(x,y) \rightarrow (\exists z On(x,z) \land \exists v On(v,y)))
```

**Goal**: a clause containing negative literals only

```
\{\neg Mortal(socrates)\}\
\{\neg Sister(sally,x), \neg Sister(x,paul)\}\
Negation of \exists x (Sister(sally,x) \land Sister(x,paul))
```

Artificial Intelligence 2024-2025

### SLD Resolution in $L_{FO}$

#### • Input: a program $\Pi$ and a goal $\phi$

Program  $\Pi$  (i.e. a set of *definite clauses:* rules + facts) in some predefined linear order:

 $\gamma_1, \gamma_2, \dots, \gamma_n$  (each  $\gamma_i$  is a definite clause)

Goal  $\phi$  (i.e. a conjunction of facts in negated form), which becomes the current goal  $\psi$ 

Note: the *selection function* for the *current goal* and *subgoal* will be discussed in the next slide

#### Procedure:

- 1) Select a negative literal  $\neg \alpha$  (i.e. the subgoal) in the current goal  $\psi$
- 2) Scan the program (in the predefined order) to identify a clause candidate literal  $\gamma_i$
- 3) Try unifying  $\neg \alpha$  and  $std(\gamma_i)$  (i.e. apply the standardization of variables to  $\alpha'$ )
- 4) If there is a *unifier*  $\sigma$  of  $\neg \alpha$  and  $std(\gamma_i)$ , replace the current goal with the *resolvent* of  $std(\gamma_i)[\sigma]$  and  $\psi[\sigma]$  (i.e. apply  $\sigma$  to both  $\psi$  and  $std(\gamma_i)$  then generate the resolvent)
- 5) Then, if the *resolvent* is the empty clause, terminate with <u>success</u>, otherwise select a new *current goal* and resume from step 1)
- 6) Else, if the unification fails , scan the program and select a new candidate literal  $\gamma_i$  and resume from step 3)
- 7) Else, if there are no further clauses in the program, select a new current goal and resume from step 1)
- 8) If all the goals in the tree have been fully explored, terminate with failure

Artificial Intelligence 2024–2025 SLD Resolution [10]

### SLD Resolution in $L_{FO}$

#### ■ Two alternative selection functions:

**Depth-first** (which is the most common...)

- Always select the most recent goal, i.e. the resolvent which has been generated last, as the current goal  $\phi$
- Then, in the current goal  $\phi$ , select the leftmost subgoal  $\neg \alpha$  not selected yet
- When the current goal  $\phi$  is fully explored and no new *resolvent* has been generated, select the next *most recent* goal in the tree (*backtracking*)

#### **Breadth-first**

- Always select the <u>least</u> recent goal as the current goal  $\phi$
- Then, in the current goal  $\phi$ , select the leftmost subgoal  $\neg \alpha$  not selected yet
- When the current goal  $\phi$  is fully explored select the next *least recent* goal in the tree

#### Comparison

Breadth-first is a *fair* selection function, in the sense that every possible resolution will be eventually attempted (i.e. 'it leaves nothing behind').

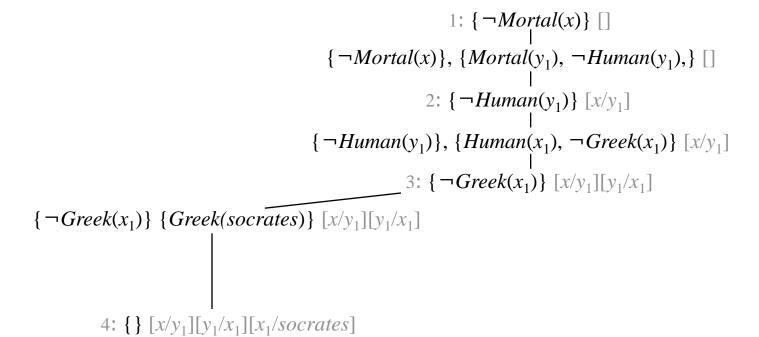
Depth-first tends to save memory and be more efficient, but it is NOT fair (more to follow)

Artificial Intelligence 2024–2025 SLD Resolution [11]

#### **SLD Trees**

Example (depth-first selection function):

```
\Pi \equiv \{\{Human(x), \neg Greek(x)\}, \{Mortal(y), \neg Human(y)\}, \\ \{Greek(socrates)\}, \{Greek(plato)\}, \{Greek(aristotle)\}\} \\ goal \equiv \{\neg Mortal(x)\} \\ \text{"Is there anyone who is mortal?"}
```

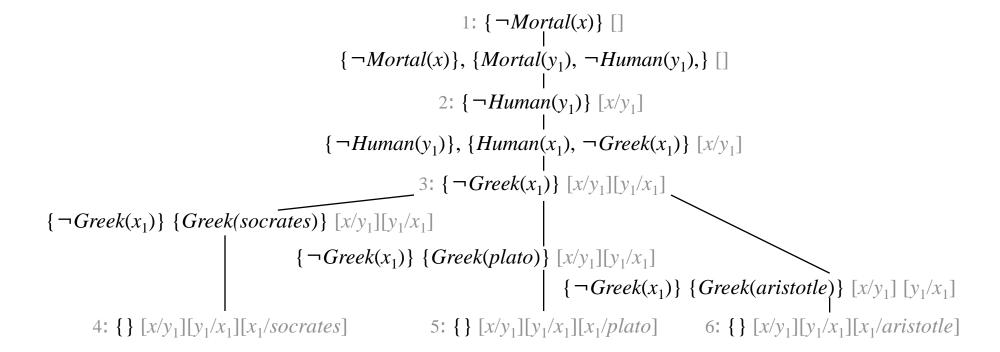


Artificial Intelligence 2024–2025 SLD Resolution [12]

#### **SLD Trees**

**Example** (depth-first selection function, forcing full exploration of SLD tree):

```
\Pi \equiv \{\{Human(x), \neg Greek(x)\}, \{Mortal(y), \neg Human(y)\}, \\ \{Greek(socrates)\}, \{Greek(plato)\}, \{Greek(aristotle)\}\} \\ goal \equiv \{\neg Mortal(x)\} \\ \text{"Is there anyone who is mortal?"}
```



Artificial Intelligence 2024–2025 SLD Resolution [13]

#### **SLD Trees**

Another example (depth-first selection function):

```
\Pi \equiv \{\{Mortal(felix), \neg Cat(felix)\}, \{Human(x), \neg Greek(x)\}, \{Mortal(y), \neg Human(y)\}, \\ \{Greek(socrates)\}, \{Greek(plato)\}, \{Greek(aristotle)\}\} \\ goal \equiv \{\neg Mortal(x)\} \\ \text{"Is there anyone who is mortal?"}
```

Artificial Intelligence 2024–2025 SLD Resolution [14]

### \*The discreet charme of functions

• Representing data structures: *lists of items* [a, b, c, ...]

```
Symbols in \Sigma
    cons/2
    it's a function that associates items (e.g. a) to a list (e.g. [b, c])
    cons(a, cons(b, cons(c, nil))) represents the list [a, b, c]
    Append/3
    it's a predicate: each pair of lists x and y is associated to their concatenation z
    nil
    it's a constant, represents the empty list.
Axioms (AL)
  \forall x Append(nil, x, x)
  \forall x \ \forall y \ \forall z \ (Append(x, y, z) \rightarrow \forall s \ Append(cons(s, x), y, cons(s, z)))
Examples of entailment
    \{AL + \exists z \ Append(cons(a, nil), cons(b, cons(c, nil), z) \}
                                \models Append(cons(a, nil), cons(b, cons(c, nil)), cons(a, cons(b, cons(c, nil))))
     \{AL + \exists x \exists y \ Append(x, y, cons(a, cons(b, nil)))\}\
                                \models Append(cons(a, nil), cons(b, nil), cons(a, cons(b, nil)))
                                \models Append(nil, cons(a, cons(b, nil)), cons(a, cons(b, nil)))
                                \models Append(cons(a, cons(b, nil)),nil, cons(a, cons(b, nil)))
```

Artificial Intelligence 2024–2025 SLD Resolution [15]

### The world of lists

• Lists of items [a, b, c, ...]

```
cons/2
     it's a function that associates items (e.g. a) to a list (e.g. [b, c])
     cons(a,cons(b,cons(c,nil))) is the list [a,b,c]
     Append/3
     it's a predicate: each pair of lists x and y is associated to their concatenation z
     nil
     it's a constant, the empty list.
  Shorthand notation (Prolog):
                                              [] \Leftrightarrow nil
                                               [a] \Leftrightarrow cons(a,nil)
                                               [a,b] \Leftrightarrow cons(a,cons(b,nil))
                                               [a/[b,c]] \Leftrightarrow cons(a,[b,c])
Axioms (AL)
  \forall x Append(nil,x,x)
  \forall x \forall y \forall z \ (Append(x,y,z) \rightarrow \forall s \ Append([s,x],y,[s,z]))
```

Artificial Intelligence 2024–2025 SLD Resolution [16]

### The world of lists

```
Problem: \forall x \ Append(nil, x, x) \models \exists y \ \forall x \ Append(nil, cons(y, x), cons(a, x))

1: \forall x \ Append(nil, x, x), \ \neg \exists y \ \forall x \ Append(nil, cons(y, x), cons(a, x)) (refutation)

2: \forall x \ Append(nil, x, x), \ \forall y \ \exists x \ \neg Append(nil, cons(y, x), cons(a, x)) (prenex normal form)

3: \{Append(nil, x, x)\}, \{\neg Append(nil, cons(y, k(y)), cons(a, k(y)))\}

(k/1 is a Skolem function, clausal form)

(N.B. there is no skolemization in Prolog: the programmer does it)
```

#### The pair of **literals**

```
Append(nil, x, x), \neg Append(nil, cons(y, k(y)), cons(a, k(y))))
```

... contains the same predicate Append/3 but the arguments are different

There is however an MGU  $\sigma = [x/cons(a, k(a)), y/a]$  that yields  $\{Append(nil, cons(a, k(a)), cons(a, k(a)))\}, \{\neg Append(nil, cons(a, k(a)), cons(a, k(a)))\}$  From this, the resolvent is the empty clause.

Artificial Intelligence 2024–2025 SLD Resolution [17]

### The world of lists in Prolog

```
% Identical to built-in predicate append/3, although it uses "cons"
% as a defined predicate, thus allowing trace-ability.

append(cons(S,X),Y,cons(S,Z)) :- append(X,Y,Z).

append(nil,X,X).

% WARNING: express your queries with cons. Examples:
% ?- append(cons(a,nil), cons(b,cons(c, nil)),cons(a,cons(b,cons(c, nil)))).
% ?- append(X,Y,cons(a,cons(b,cons(c, nil)))).
```

Artificial Intelligence 2024–2025 SLD Resolution [18]

An example:

$$\Pi \equiv \{ \{ S(a,b) \}, \{ S(b,c) \}, \{ S(x,z), \neg S(x,y), \neg S(y,z) \} \}$$

$$\neg \phi \equiv \{ \neg S(a,x) \}$$

$$\text{goal: } \neg S(a,x) []$$

$$\{ \neg S(a,x) \}, \{ S(a,b) \} []$$

$$\{ \} [x/b]$$

Easy...

Artificial Intelligence 2024–2025 SLD Resolution [19]

An example:

$$\Pi = \{\{S(a,b)\}, \{S(b,c)\}, \{S(x,z), \neg S(x,y), \neg S(y,z)\}\} \\
\neg \phi = \{\neg S(a,x)\} \\
\text{goal: } \neg S(a,x) [] \\
\{\neg S(a,x)\}, \{S(a,b)\} [] \\
\{\neg S(a,x)\}, \{S(x_3,z_3), \neg S(x_3,y_3), \neg S(y_3,z_3)\} [] \\
\{\neg S(a,y_3), \neg S(y_3,z_3)\}, \{S(a,b)\} [x/z_3, x_3/a, y_3/b] \\
\{\neg S(b,z_3)\}, \{S(b,c)\} [x/z_3, x_3/a] \\
\{\neg S(b,z_3)\}, \{S(b,z)\}, \{S(b$$

Forcing to backtrack... (easy again)

Artificial Intelligence 2024–2025 SLD Resolution [20]

An example:

$$\Pi \equiv \{\{S(a,b)\}, \{S(b,c)\}, \{S(x,z), \neg S(x,y), \neg S(y,z)\}\}$$

$$\neg \phi \equiv \{\neg S(a,x)\}$$

$$= \{\neg S(a,x)\}$$

$$= \{\neg S(a,x)\}, \{S(x_3,z_3), \neg S(x_3,y_3), \neg S(y_3,z_3)\} []$$

$$= \{\neg S(a,y_3), \neg S(y_3,z_3), \neg S(y_3,z_3)\} [x_3/a, x/z_3]$$

$$= \{\neg S(b,z_3)\}, \{S(a,b)\} [x/z_3, x_3/a]$$

$$= \{\neg S(b,z_3)\}, \{S(b,c)\} [x/z_3, x_3/a]$$

$$= \{\neg S(b,z_3)\}, \{S(b,c)\} [x/z_3, x_3/a]$$

$$= \{\neg S(b,z_3)\}, \{S(x_4,z_4), \neg S(x_4,y_4), \neg S(y_4,z_4)\} [x/z_3, x_3/a]$$

$$= \{\neg S(b,y_4), \neg S(y_4,z_4)\}, \{S(x_5,z_5), \neg S(x_5,y_5), \neg S(y_5,z_5)\} [x/z_3, x_3/a, z_3/z_4, x_4/b]$$

$$= \{\neg S(b,y_4), \neg S(y_4,z_4)\}, \{S(x_5,z_5), \neg S(x_5,y_5), \neg S(y_5,z_5)\} [x/z_3, x_3/a, z_3/z_4, x_4/b]$$

$$= \{\neg S(b,y_5), \neg S(y_5,z_5), \neg S(z_5,z_4)\} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(b,y_5), \neg S(y_5,z_5), \neg S(z_5,z_4)\} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \} [x/z_3, x_3/a, z_3/z_4, x_4/b, y_4/z_5, x_5/b]$$

$$= \{\neg S(a,x)\} \{x/z_3, x_3/a, z_3/z_4, z_3/a, z_$$

Artificial Intelligence 2024–2025 SLD Resolution [21]

A second example:

$$\Pi \equiv \{\{S(x,z), \neg S(x,y), \neg S(y,z)\}, \{S(a,b)\}, \{S(b,c)\}\}$$
 
$$\neg \phi \equiv \{\neg S(a,x)\}$$
 Notice the change in clause ordering..... goal:  $\neg S(a,x)$  []

goal: 
$$\neg S(a,x)$$
 []
$$\{\neg S(a,x)\}, \{S(x_1,z_1), \neg S(x_1,y_1), \neg S(y_1,z_1)\} []$$

$$\{\neg S(a,y_1), \neg S(y_1,z_1)\} [x_1/a, x/z_1]$$

$$\{\neg S(a,y_1), \neg S(y_1,z_1)\}, \{S(x_2,z_2), \neg S(x_2,y_2), \neg S(y_2,z_2)\} [x_1/a, x/z_1]$$

$$\{\neg S(z_2,z_1), \neg S(x_2,y_2), \neg S(y_2,z_2)\} [x_1/a, x/z_1, x_2/a, y_1/z_2]$$

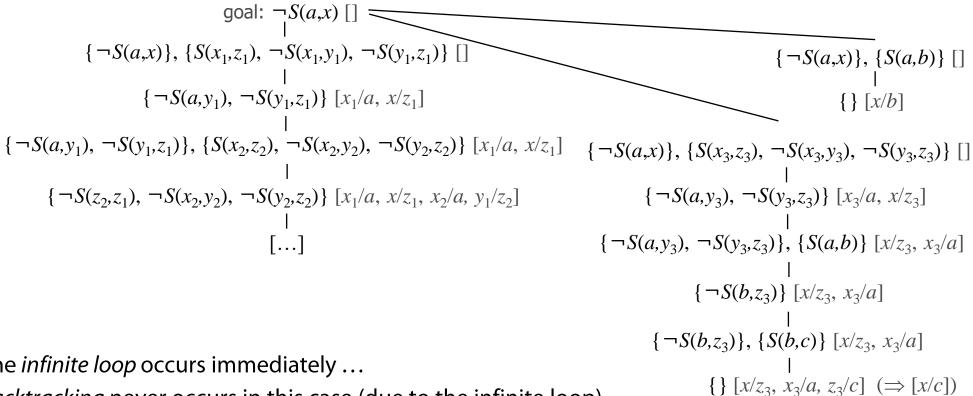
$$[\dots]$$

The *infinite loop* occurs immediately ...

Artificial Intelligence 2024–2025 SLD Resolution [22]

A second example:

$$\Pi \equiv \{\{S(x,z), \neg S(x,y), \neg S(y,z)\}, \{S(a,b)\}, \{S(b,c)\}\}$$
 
$$\neg \phi \equiv \{\neg S(a,x)\}$$
 Notice the change in clause ordering.....



The *infinite loop* occurs immediately ...

Backtracking never occurs in this case (due to the infinite loop), yet, if it occurred it would have produced the two correct results

Artificial Intelligence 2024-2025 SLD Resolution [23]

#### Moral

- In both previous examples the infinite loop (i.e. divergence) is unavoidable
- Yet in the first one, the method first produces the right results and then diverges
- So in the first case the result is *complete* (i.e. all entailed formulae are derived) while in the second case the method is not

A *fair* selection function is such that no possible resolution will be postponed indefinitely: that is, <u>any</u> possible resolution will be performed, eventually.

In the two previous examples, we used a *depth-first* exploration method of the SLD tree: which is <u>not</u> complete (in the above sense)

A breadth-first exploration method is fair hence it is complete (in the above sense)

In actual programming systems (e.g. Prolog) the depth-first is preferred for memory efficiency since the breadth-first method forces to keep (most of) the whole SLD tree in memory