

Computational Logic

A “Hands-on” Introduction to Pure Logic Programming

Syntax: Terms (Variables, Constants, and Structures)

(using Prolog notation conventions)

- **Variables:** start with uppercase character (or “_”), may include “_” and digits:

Examples: X, Im4u, A_little_garden, _, _x, _22

- **Constants:** lowercase first character, may include “_” and digits. Also, numbers and some special characters. Quoted, any character:

Examples: a, dog, a_big_cat, 23, 'Hungry man', []

- **Structures:** a **functor** (the structure name, is like a constant name) followed by a fixed number of arguments between parentheses:

Example: date(monday, Month, 1994)

Arguments can in turn be variables, constants and structures.

- ◇ **Arity:** is the number of arguments of a structure. Functors are represented as *name/arity*. A constant can be seen as a structure with arity zero.

Variables, constants, and structures as a whole are called **terms** (they are the terms of a “first-order language”): the *data structures* of a logic program.

Syntax: Terms

(using Prolog notation conventions)

- Examples of terms:

<i>Term</i>	<i>Type</i>	<i>Main functor:</i>
dad	constant	dad/0
time(min, sec)	structure	time/2
pair(Calvin, tiger(Hobbes))	structure	pair/2
Tee(Alf, rob)	illegal	—
A_good_time	variable	—

- *Functors* can be defined as *prefix*, *postfix*, or *infix* operators (just syntax!):

a + b	is the term	'+' (a, b)	if +/2 declared infix
- b	is the term	'-' (b)	if -/1 declared prefix
a < b	is the term	'<' (a, b)	if </2 declared infix
john father mary	is the term	father(john, mary)	if father/2 declared infix

We assume that some such operator definitions are always preloaded.

Syntax: Rules and Facts (Clauses)

- **Rule:** an expression of the form:

$$\begin{array}{l} p_0(t_1, t_2, \dots, t_{n_0}) \leftarrow \\ \quad p_1(t_1^1, t_2^1, \dots, t_{n_1}^1), \\ \quad \dots \\ \quad p_m(t_1^m, t_2^m, \dots, t_{n_m}^m). \end{array}$$

- ◇ $p_0(\dots)$ to $p_m(\dots)$ are *syntactically* like *terms*.
- ◇ $p_0(\dots)$ is called the **head** of the rule.
- ◇ The p_i to the right of the arrow are called *literals* and form the **body** of the rule. They are also called **procedure calls**.
- **Fact:** an expression of the form $p(t_1, t_2, \dots, t_n) \leftarrow .$ (i.e., a rule with empty body).

Example: meal(soup, beef, coffee) <- .
meal(First, Second, Third) <-
 appetizer(First),
 main_dish(Second),
 dessert(Third) .

- Rules and facts are both called **clauses**.

Syntax: Predicates, Programs, and Queries

- **Predicate** (or *procedure definition*): a set of clauses whose heads have the same name and arity (called the **predicate name**).

Examples:

```
pet(spot) <- .                animal(spot) <- .
pet(X) <- animal(X), barks(X). animal(barry) <- .
pet(X) <- animal(X), meows(X). animal(hobbes) <- .
```

Predicate `pet/1` has three clauses. Of those, one is a fact and two are rules.
Predicate `animal/1` has three clauses, all facts.

- **Logic Program:** a set of predicates.
- **Query:** an expression of the form:
(i.e., a clause without a head).

$$\leftarrow p_1(t_1^1, \dots, t_{n_1}^1), \dots, p_n(t_1^n, \dots, t_{n_m}^n).$$

A query represents a *question to the program*.

Example: `<- pet(X).`

“Declarative” Meaning of Facts and Rules

The declarative meaning is the corresponding one in first order logic, according to certain conventions:

- **Facts:** state things that are true.

(Note that a fact “ $p \leftarrow .$ ” can be seen as the rule “ $p \leftarrow \text{true}.$ ”)

Example: the fact `animal(spot) <-.`

can be read as “spot is an animal”.

- **Rules:**

- ◇ Commas in rule bodies represent conjunction, i.e.,

$p \leftarrow p_1, \dots, p_m.$ represents $p \leftarrow p_1 \wedge \dots \wedge p_m.$

- ◇ “ \leftarrow ” represents as usual logical implication.

Thus, a rule $p \leftarrow p_1, \dots, p_m.$ means “if p_1 and ... and p_m are true, then p is true”

Example: the rule `pet(X) <- animal(X), barks(X).`

can be read as “X is a pet if it is an animal and it barks”.

“Declarative” Meaning of Predicates and Queries

- **Predicates:** clauses in the same predicate

$p \leftarrow p_1, \dots, p_n$

$p \leftarrow q_1, \dots, q_m$

...

provide different *alternatives* (for p).

Example: the rules

$\text{pet}(X) \leftarrow \text{animal}(X), \text{barks}(X).$

$\text{pet}(X) \leftarrow \text{animal}(X), \text{meows}(X).$

express two ways for X to be a pet.

- **Note** (variable *scope*): the X vars. in the two clauses above are different, despite the same name. Vars. are *local to clauses* (and are *renamed* any time a clause is used –as with vars. local to a procedure in conventional languages).
- A **query** represents a *question to the program*.

Examples:

$\leftarrow \text{pet}(\text{spot}).$

asks whether spot is a pet.

$\leftarrow \text{pet}(X).$

asks: “Is there an X which is a pet?”

“Execution” and Semantics

- Example of a **logic program**:

```
pet(X) <- animal(X), barks(X).  
pet(X) <- animal(X), meows(X).  
  
animal(spot) <- .           barks(spot) <- .  
animal(barry) <- .         meows(barry) <- .  
animal(hobbes) <- .       roars(hobbes) <- .
```

- **Execution**: given a program and a query, *executing* the logic program is *attempting to find an answer to the query*.

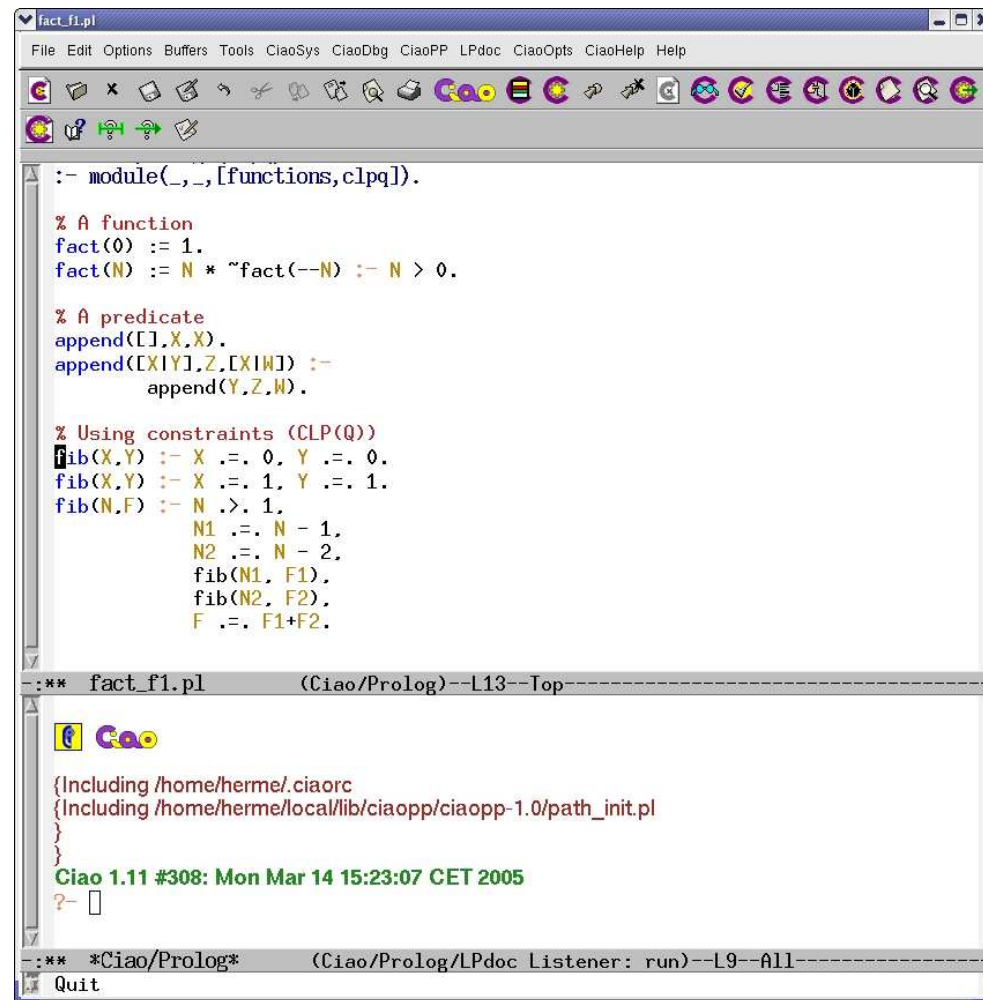
Example: given the program above and the query `<- pet(X)`, the system will try to find a “substitution” for X which makes `pet(X)` true.

- ◇ The **declarative semantics** specifies *what* should be computed (all possible answers).
⇒ Intuitively, we have two possible answers: `X = spot` and `X = barry`.
- ◇ The **operational semantics** specifies *how* answers are computed (which allows us to determine *how many steps* it will take).

Running Pure Logic Programs: the Ciao System's bf/af Packages

- We will be using *Ciao*, a multiparadigm programming system which includes (as one of its “paradigms”) a *pure logic programming* subsystem:
 - ◇ A number of *fair* search rules are available (breadth-first, iterative deepening, ...): we will use “breadth-first” (bf or af).
 - ◇ Also, a module can be set to *pure* mode so that impure built-ins are not accessible to the code in that module.
 - ◇ This provides a reasonable first approximation of “Greene’s dream” (of course, at a cost in memory and execution time).
- Writing programs to execute in bf mode:
 - ◇ All files should start with the following line:
`:- module(_,_, [bf]).` (or `:- module(_,_, ['bf/af']).`)
or, for “user” files, i.e., files that are not modules: `:- use_package(bf).`
 - ◇ The *neck* (arrow) of rules must be `<-` .
 - ◇ Facts must end with `<- .` .

Ciao Programming Environment: file being edited and top-level



The screenshot shows the Ciao Programming Environment interface. The top window, titled 'fact_f1.pl', contains the following Prolog code:

```
:- module(_,_,[functions,clpq]).

% A function
fact(0) := 1.
fact(N) := N * ~fact(--N) :- N > 0.

% A predicate
append([],X,X).
append([X1|Y],Z,[X1|W]) :-
    append(Y,Z,W).

% Using constraints (CLP(Q))
fib(X,Y) :- X .=. 0, Y .=. 0.
fib(X,Y) :- X .=. 1, Y .=. 1.
fib(N,F) :- N .>. 1,
            N1 .=. N - 1,
            N2 .=. N - 2,
            fib(N1, F1),
            fib(N2, F2),
            F .=. F1+F2.
```

Below the code editor, the top-level shell is visible, showing the Ciao logo and the following text:

```
{Including /home/herme/.ciaorc
{Including /home/herme/local/lib/ciaopp/ciaopp-1.0/path_init.pl
}
Ciao 1.11 #308: Mon Mar 14 15:23:07 CET 2005
?-
```

The bottom status bar indicates the environment is running: `--** *Ciao/Prolog* (Ciao/Prolog/LPdoc Listener: run)--L9--All`. A 'Quit' button is visible at the bottom left.

Top Level Interaction Example

- File `pets.pl` contains:
:- `module(_,_,[bf]).`
+ *the pet example code as in previous slides.*
- Interaction with the system query evaluator (the “top level”):

```
Ciao 1.13 #0: Mon Nov 7 09:48:51 MST 2005
?- use_module(pets).
yes
?- pet(spot).
yes
?- pet(X).
X = spot ? ;
X = barry ? ;
no
?-
```

Simple (Top-Down) Operational Meaning of Programs

- A logic program is operationally a set of *procedure definitions* (the predicates).
- A query $\leftarrow p$ is an initial *procedure call*.
- A procedure definition with one *clause* $p \leftarrow p_1, \dots, p_m$ means:
“to execute a call to p you have to *call* p_1 and \dots and p_m ”
 - ◇ In principle, the order in which p_1, \dots, p_n are called does not matter, but, in practical systems it is fixed.
- If several clauses (definitions) $p \leftarrow p_1, \dots, p_n$ means:
 $p \leftarrow q_1, \dots, q_m$
“to execute a call to p , call $p_1 \wedge \dots \wedge p_n$, or, alternatively, $q_1 \wedge \dots \wedge q_n$, or \dots ”
 - ◇ Unique to logic programming –it is like having several alternative procedure definitions.
 - ◇ Means that several possible paths may exist to a solution and they *should be explored*.
 - ◇ System usually stops when the first solution found, user can ask for more.
 - ◇ Again, in principle, the order in which these paths are explored does not matter (*if certain conditions are met*), but, for a given system, this is typically also fixed.

In the following we define a more precise operational semantics.

Unification: uses

- **Unification** is the mechanism used in procedure calls to:
 - ◇ Pass parameters.
 - ◇ “Return” values.
- It is also used to:
 - ◇ Access parts of structures.
 - ◇ Give values to variables.

Unification

- **Unifying two terms (or literals) A and B:** is asking if they can be made syntactically identical by giving (minimal) values to their variables.
 - ◇ I.e., find a **variable substitution** θ such that $A\theta = B\theta$ (or, if impossible, *fail*).
 - ◇ Only variables can be given values!
 - ◇ Two structures can be made identical only by making their arguments identical.

E.g.:

A	B	θ	$A\theta$	$B\theta$
dog	dog	\emptyset	dog	dog
X	a	$\{X = a\}$	a	a
X	Y	$\{X = Y\}$	Y	Y
$f(X, g(t))$	$f(m(h), g(M))$	$\{X=m(h), M=t\}$	$f(m(h), g(t))$	$f(m(h), g(t))$
$f(X, g(t))$	$f(m(h), t(M))$	Impossible (1)		
$f(X, X)$	$f(Y, l(Y))$	Impossible (2)		

- (1) Structures with different name and/or arity cannot be unified.
- (2) A variable cannot be given as value a term which contains that variable, because it would create an infinite term. This is known as the **occurs check**.

Unification

- Often several solutions exist, e.g.:

A	B	θ_1	$A\theta_1$ and $B\theta_1$
$f(X, g(T))$	$f(m(H), g(M))$	$\{ X=m(a), H=a, M=b, T=b \}$	$f(m(a), g(b))$
"	"	$\{ X=m(H), M=f(A), T=f(A) \}$	$f(m(H), g(f(A)))$

These are correct, but a simpler (“more general”) solution exists:

A	B	θ_1	$A\theta_1$ and $B\theta_1$
$f(X, g(T))$	$f(m(H), g(M))$	$\{ X=m(H), T=M \}$	$f(m(H), g(M))$

- Always a unique (modulo variable renaming) *most general* solution exists (unless unification fails).
- This is the one that we are interested in.
- The *unification algorithm* finds this solution.

Unification Algorithm

- Let A and B be two terms:
 - 1 $\theta = \emptyset, E = \{A = B\}$
 - 2 while not $E = \emptyset$:
 - 2.1 delete an equation $T = S$ from E
 - 2.2 case T or S (or both) are (distinct) variables. Assuming T variable:
 - * (occur check) if T occurs in the term $S \rightarrow$ halt with failure
 - * substitute variable T by term S in all terms in θ
 - * substitute variable T by term S in all terms in E
 - * add $T = S$ to θ
 - 2.3 case T and S are non-variable terms:
 - * if their names or arities are different \rightarrow halt with failure
 - * obtain the arguments $\{T_1, \dots, T_n\}$ of T and $\{S_1, \dots, S_n\}$ of S
 - * add $\{T_1 = S_1, \dots, T_n = S_n\}$ to E
 - 3 halt with θ being the m.g.u of A and B

Unification Algorithm Examples (I)

- Unify: $A = p(X, X)$ and $B = p(f(Z), f(W))$

θ	E	T	S
$\{ \}$	$\{ p(X, X) = p(f(Z), f(W)) \}$	$p(X, X)$	$p(f(Z), f(W))$
$\{ \}$	$\{ X = f(Z), X = f(W) \}$	X	$f(Z)$
$\{ X = f(Z) \}$	$\{ f(Z) = f(W) \}$	$f(Z)$	$f(W)$
$\{ X = f(Z) \}$	$\{ Z = W \}$	Z	W
$\{ X = f(W), Z = W \}$	$\{ \}$		

- Unify: $A = p(X, f(Y))$ and $B = p(Z, X)$

θ	E	T	S
$\{ \}$	$\{ p(X, f(Y)) = p(Z, X) \}$	$p(X, f(Y))$	$p(Z, X)$
$\{ \}$	$\{ X = Z, f(Y) = X \}$	X	Z
$\{ X = Z \}$	$\{ f(Y) = Z \}$	$f(Y)$	Z
$\{ X = f(Y), Z = f(Y) \}$	$\{ \}$		

Unification Algorithm Examples (II)

- Unify: $A = p(X, f(Y))$ and $B = p(a, g(b))$

θ	E	T	S
$\{ \}$	$\{ p(X, f(Y)) = p(a, g(b)) \}$	$p(X, f(Y))$	$p(a, g(b))$
$\{ \}$	$\{ X = a, f(Y) = g(b) \}$	X	a
$\{ X = a \}$	$\{ f(Y) = g(b) \}$	$f(Y)$	$g(b)$
<i>fail</i>			

- Unify: $A = p(X, f(X))$ and $B = p(Z, Z)$

θ	E	T	S
$\{ \}$	$\{ p(X, f(X)) = p(Z, Z) \}$	$p(X, f(X))$	$p(Z, Z)$
$\{ \}$	$\{ X = Z, f(X) = Z \}$	X	Z
$\{ X = Z \}$	$\{ f(Z) = Z \}$	$f(Z)$	Z
<i>fail</i>			

A (Schematic) Interpreter for Logic Programs (SLD-resolution)

Input: A logic program P , a query Q

Output: $Q\mu$ (answer substitution) if Q is provable from P , *failure* otherwise

Algorithm:

1. Initialize the “resolvent” R to be $\{Q\}$
 2. While R is nonempty do:
 - 2.1. Take the leftmost literal A in R
 - 2.2. Choose a (*renamed*) clause $A' \leftarrow B_1, \dots, B_n$ from P , such that A and A' *unify* with unifier θ
(if no such clause can be found, branch is *failure*; explore another branch)
 - 2.3. Remove A from R , add B_1, \dots, B_n to R
 - 2.4. Apply θ to R and Q
 3. If R is empty, output Q (a solution). Explore another branch for more sol's.
- Step 2.2 defines *alternative paths* to be explored to find answer(s); execution explores this tree (for example, breadth-first).

A (Schematic) Interpreter for Logic Programs (Contd.)

- Since step 2.2 is left open, a given logic *programming* system must specify how it deals with this by providing one (or more)
 - ◊ **Search rule(s)**: “how are clauses/branches selected in 2.2.”
- If the search rule is not specified execution is *nondeterministic*, since choosing a different clause (in step 2.2) can lead to different solutions (finding solutions in a different order).

Example (two valid executions):

```
?- pet(X).
```

```
X = spot ? ;
```

```
X = barry ? ;
```

```
no
```

```
?-
```

```
?- pet(X).
```

```
X = barry ? ;
```

```
X = spot ? ;
```

```
no
```

```
?-
```

- In fact, there is also some freedom in step 2.1, i.e., a system may also specify:
 - ◊ **Computation rule(s)**: “how are literals selected in 2.1.”

Running programs

C₁: pet(X) ← animal(X), barks(X).

C₂: pet(X) ← animal(X), meows(X).

C₃: animal(spot) ←.

C₄: animal(barry) ←.

C₅: animal(hobbes) ←.

C₆: barks(spot) ←.

C₇: meows(barry) ←.

C₈: roars(hobbes) ←.

- `<- pet(P).`

Q	R	Clause	θ
pet(P)	<u>pet(P)</u>	C ₂ *	{P = X ₁ }
pet(X ₁)	<u>animal(X₁), meows(X₁)</u>	C ₄ *	{X ₁ = barry}
pet(barry)	<u>meows(barry)</u>	C ₇	{}
pet(barry)	—	—	—

* means there is a *choice-point*, i.e., there are other clauses whose head unifies.

- System response: `P = barry ?`
- If we type “;” after the ? prompt (i.e., we ask for another solution) the system can go and execute a different branch (i.e., a different choice in C₂* or C₄*).

Running programs (different strategy)

C₁: pet(X) ← animal(X), barks(X).

C₂: pet(X) ← animal(X), meows(X).

C₃: animal(spot) ←.

C₄: animal(barry) ←.

C₅: animal(hobbes) ←.

C₆: barks(spot) ←.

C₇: meows(barry) ←.

C₈: roars(hobbes) ←.

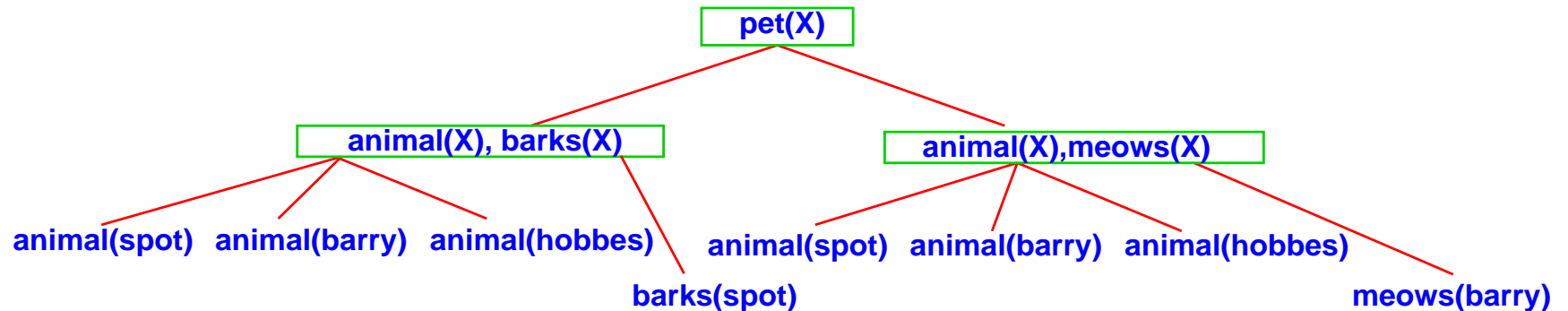
- ← pet(P). (different strategy)

Q	R	Clause	θ
pet(P)	<u>pet(P)</u>	C ₁ [*]	{P = X ₁ }
pet(X ₁)	<u>animal(X₁)</u> , barks(X ₁)	C ₅ [*]	{X ₁ = hobbes}
pet(hobbes)	<u>barks(hobbes)</u>	???	failure
→ explore another branch (different choice in C ₁ [*] or C ₅ [*]) to find a solution. We take C ₃ instead of C ₅ :			
pet(P)	<u>pet(P)</u>	C ₁ [*]	{P = X ₁ }
pet(X ₁)	<u>animal(X₁)</u> , barks(X ₁)	C ₃ [*]	{X ₁ = spot}
pet(spot)	<u>barks(spot)</u>	C ₆	{}
pet(spot)	—	—	—

The Search Tree

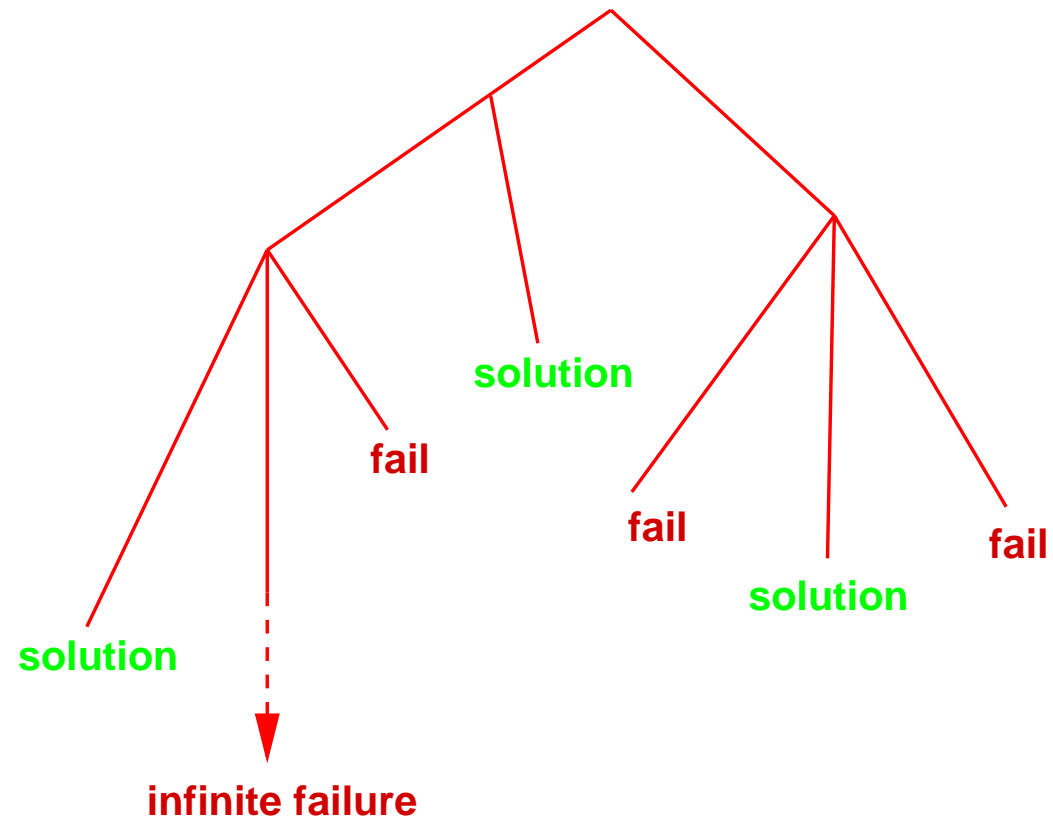
- A query + a logic program together specify a *search tree*.

Example: query $\leftarrow \text{pet}(X)$ with the previous program generates this search tree (the boxes represent the “and” parts [except leaves]):



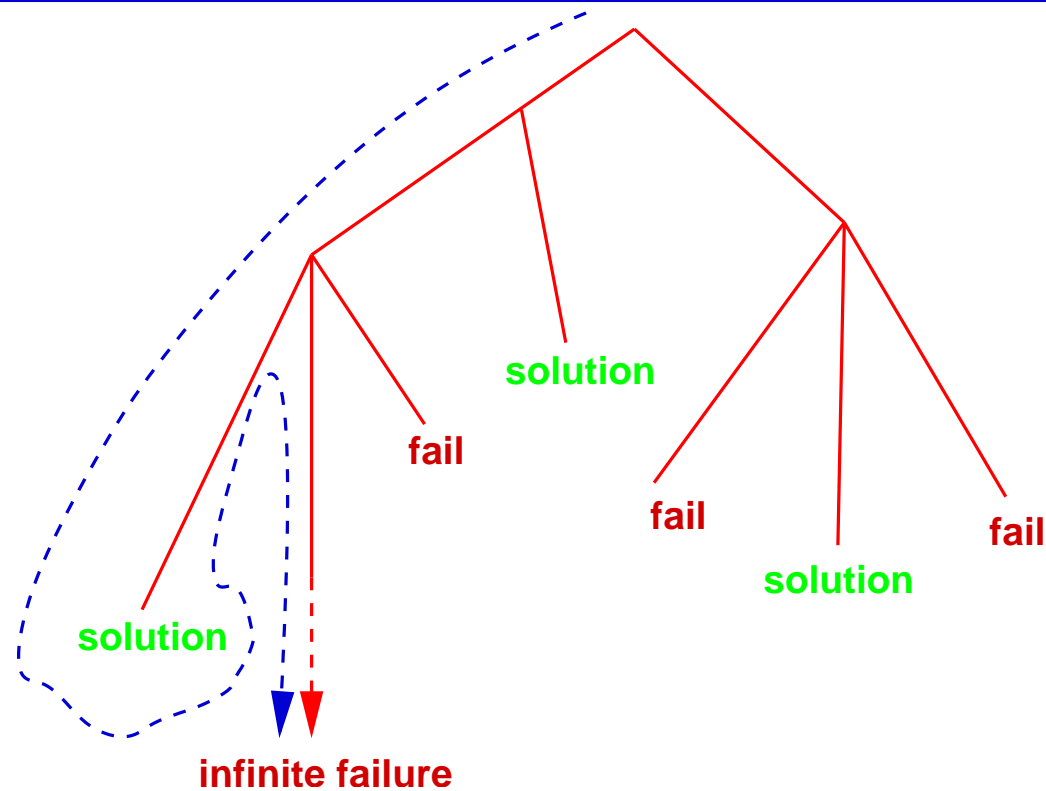
- Different query \rightarrow different tree.
- The search and computation rules explain how the search tree will be explored during execution.
- How can we achieve completeness (guarantee that all solutions will be found)?

Characterization of The Search Tree



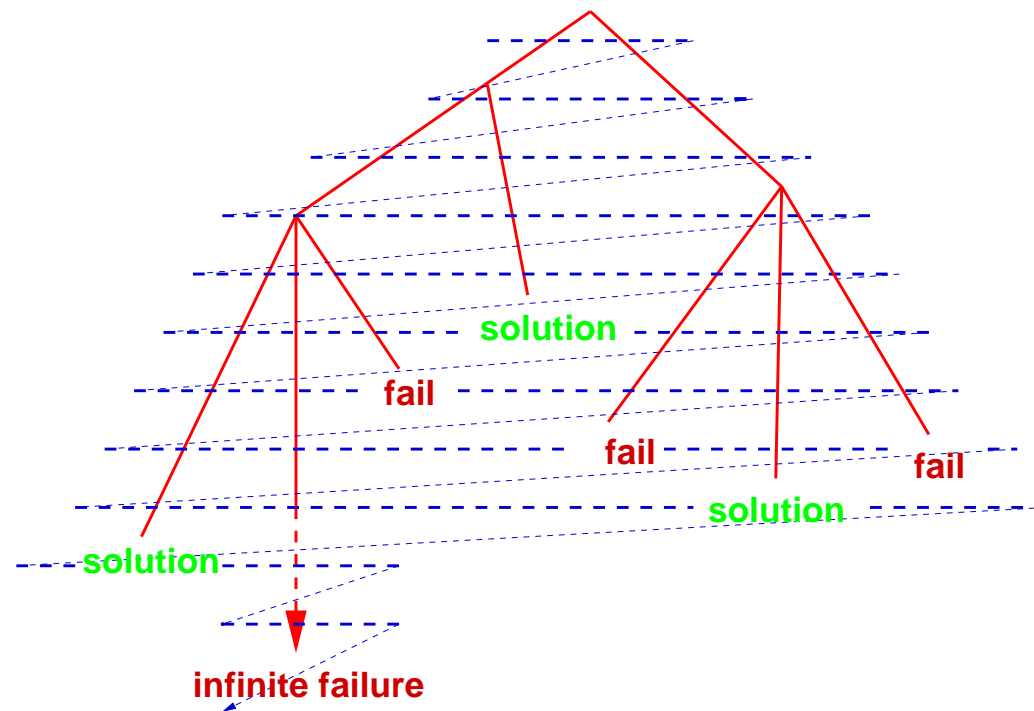
- All solutions are at *finite depth* in the tree.
- Failures can be at finite depth or, in some cases, be an infinite branch.

Depth-First Search



- Incomplete: may fall through an infinite branch before finding all solutions.
- But very efficient: it can be implemented with a call stack, very similar to a traditional programming language.

Breadth-First Search



- Will find all solutions before falling through an infinite branch.
- But costly in terms of time and memory.
- Used in all the following examples (via Ciao's bf package).

Role of Unification in Execution and Modes

- As mentioned before, unification used to *access data* and *give values to variables*.

Example: Consider query `<- animal(A), named(A,Name).` with:

`animal(dog(barry)) <- . named(dog(Name),Name) <- .`

- Also, unification is used to *pass parameters* in procedure calls and to *return values* upon procedure exit.

Q	R	Clause	θ
<code>pet(P)</code>	<code>pet(P)</code>	C_1^*	$\{ P=X_1 \}$
<code>pet(X₁)</code>	<code>animal(X₁), barks(X₁)</code>	C_3^*	$\{ X_1=spot \}$
<code>pet(spot)</code>	<code>barks(spot)</code>	C_6	$\{ \}$
<code>pet(spot)</code>	—	—	—

- In fact, argument positions are not fixed a priori to be input or output.

Example: Consider query `<- pet(spot).` vs. `<- pet(X).`

or `<- add(s(0),s(s(0)),Z).` vs. `<- add(s(0),Y,s(s(s(0))))).`

- Thus, procedures can be used in different **modes** (different sets of arguments are input or output in each mode).

Database Programming

- A Logic Database is a set of facts and rules (i.e., a logic program):

```
father_of(john,peter) <- .      | grandfather_of(L,M) <- father_of(L,N),
father_of(john,mary) <- .      |                             father_of(N,M).
father_of(peter,michael) <- . | grandfather_of(X,Y) <- father_of(X,Z),
                               |                             mother_of(Z,Y).
mother_of(mary, david) <- .
```

- Given such database, a logic programming system can answer questions (queries) such as:

```
<- father_of(john, peter).
```

Answer: *Yes*

```
<- father_of(john, david).
```

Answer: *No*

```
<- father_of(john, X).
```

Answer: $\{X = \textit{peter}\}$

Answer: $\{X = \textit{mary}\}$

```
<- grandfather_of(X, michael).
```

Answer: $\{X = \textit{john}\}$

```
<- grandfather_of(X, Y).
```

Answer: $\{X = \textit{john}, Y = \textit{michael}\}$

Answer: $\{X = \textit{john}, Y = \textit{david}\}$

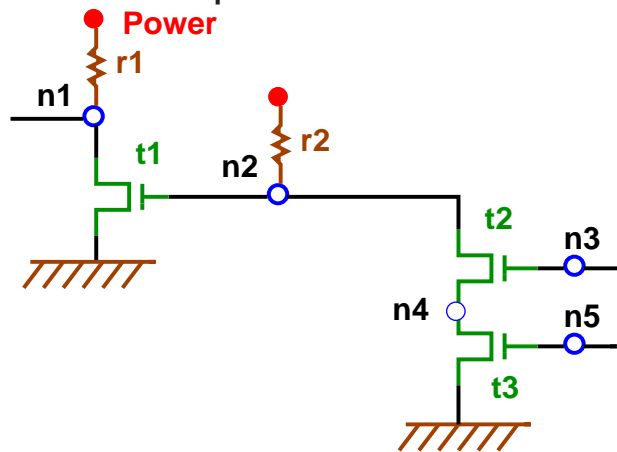
```
<- grandfather_of(X, X).
```

Answer: *No*

- Rules for grandmother_of(X, Y)?

Database Programming (Contd.)

- Another example:



```
resistor(power,n1) <- .
resistor(power,n2) <- .
```

```
transistor(n2,ground,n1) <- .
transistor(n3,n4,n2) <- .
transistor(n5,ground,n4) <- .
```

```
inverter(Input,Output) <-
  transistor(Input,ground,Output), resistor(power,Output).
nand_gate(Input1,Input2,Output) <-
  transistor(Input1,X,Output), transistor(Input2,ground,X),
  resistor(power,Output).
and_gate(Input1,Input2,Output) <-
  nand_gate(Input1,Input2,X), inverter(X, Output).
```

- Query `and_gate(In1,In2,Out)` has solution: $\{In1=n3, In2=n5, Out=n1\}$

Structured Data and Data Abstraction (and the '=' Predicate)

- *Data structures* are created using (complex) terms.
- Structuring data is important:
`course(complog,wed,19,00,20,30,'M.','Hermenegildo',new,5102) <-.`
- When is the Computational Logic course?
`<- course(complog,Day,StartH,StartM,FinishH,FinishM,C,D,E,F).`

- Structured version:

```
course(complog,Time,Lecturer, Location) <-  
    Time = t(wed,18:30,20:30),  
    Lecturer = lect('M.','Hermenegildo'),  
    Location = loc(new,5102).
```

Note: “X=Y” is equivalent to “’=’(X,Y)”

where the predicate =/2 is defined as the fact “’=’(X,X) <-.” – Plain unification!

- Equivalent to:

```
course(complog, t(wed,18:30,20:30),  
    lect('M.','Hermenegildo'), loc(new,5102)) <-.
```

Structured Data and Data Abstraction (and The Anonymous Variable)

- Given:

```
course(complog,Time,Lecturer, Location) <-  
  Time = t(wed,18:30,20:30),  
  Lecturer = lect('M.', 'Hermenegildo'),  
  Location = loc(new,5102).
```

- When is the Computational Logic course?

```
<- course(complog,Time, A, B).
```

has solution:

```
{Time=t(wed,18:30,20:30), A=lect('M.', 'Hermenegildo'), B=loc(new,5102)}
```

- Using the *anonymous variable* (“_”):

```
<- course(complog,Time, _, _).
```

has solution:

```
{Time=t(wed,18:30,20:30)}
```

Structured Data and Data Abstraction (Contd.)

- The circuit example revisited:

```
resistor(r1,power,n1) <- .      transistor(t1,n2,ground,n1) <- .
resistor(r2,power,n2) <- .      transistor(t2,n3,n4,n2) <- .
                                   transistor(t3,n5,ground,n4) <- .
```

```
inverter(inv(T,R),Input,Output) <-
  transistor(T,Input,ground,Output), resistor(R,power,Output).
```

```
nand_gate(nand(T1,T2,R),Input1,Input2,Output) <-
  transistor(T1,Input1,X,Output), transistor(T2,Input2,ground,X),
  resistor(R,power,Output).
```

```
and_gate(and(N,I),Input1,Input2,Output) <-
  nand_gate(N,Input1,Input2,X), inverter(I,X,Output).
```

- The query `<- and_gate(G,In1,In2,Out).`

has solution: `{G=and(nand(t2,t3,r2),inv(t1,r1)),In1=n3,In2=n5,Out=n1}`

Logic Programs and the Relational DB Model

Traditional → Codd's Relational Model

File	Relation	Table
Record	Tuple	Row
Field	Attribute	Column

- Example:

Name	Age	Sex
Brown	20	M
Jones	21	F
Smith	36	M

Person

Name	Town	Years
Brown	London	15
Brown	York	5
Jones	Paris	21
Smith	Brussels	15
Smith	Santander	5

Lived-in

- The order of the rows is immaterial.
- (Duplicate rows are not allowed)

Logic Programs and the Relational DB Model (Contd.)

<u>Relational Database</u>	→	<u>Logic Programming</u>
Relation Name	→	Predicate symbol
Relation	→	Procedure consisting of ground facts (facts without variables)
Tuple	→	Ground fact
Attribute	→	Argument of predicate

- Example:

```
person(brown,20,male) <-.  
person(jones,21,female) <-.  
person(smith,36,male) <-.  

```

Name	Age	Sex
Brown	20	M
Jones	21	F
Smith	36	M

- Example:

```
lived_in(brown,london,15) <-.  
lived_in(brown,york,5) <-.  
lived_in(jones,paris,21) <-.  
lived_in(smith,brussels,15) <-.  
lived_in(smith,santander,5) <-.  

```

Name	Town	Years
Brown	London	15
Brown	York	5
Jones	Paris	21
Smith	Brussels	15
Smith	Santander	5

Logic Programs and the Relational DB Model (Contd.)

- The operations of the relational model are easily implemented as rules.

- ◇ *Union:*

$r_union_s(X_1, \dots, X_n) \leftarrow r(X_1, \dots, X_n).$

$r_union_s(X_1, \dots, X_n) \leftarrow s(X_1, \dots, X_n).$

- ◇ *Set Difference:*

$r_diff_s(X_1, \dots, X_n) \leftarrow r(X_1, \dots, X_n), \text{ not } s(X_1, \dots, X_n).$

$r_diff_s(X_1, \dots, X_n) \leftarrow s(X_1, \dots, X_n), \text{ not } r(X_1, \dots, X_n).$

(we postpone the discussion on *negation* until later.)

- ◇ *Cartesian Product:*

$r_X_s(X_1, \dots, X_m, X_{m+1}, \dots, X_{m+n}) \leftarrow r(X_1, \dots, X_m), s(X_{m+1}, \dots, X_{m+n}).$

- ◇ *Projection:*

$r13(X_1, X_3) \leftarrow r(X_1, X_2, X_3).$

- ◇ *Selection:*

$r_selected(X_1, X_2, X_3) \leftarrow r(X_1, X_2, X_3), \leq(X_2, X_3).$

(see later for definition of $\leq/2$)

Logic Programs and the Relational DB Model (Contd.)

- Derived operations – some can be expressed more directly in LP:

- ◇ Intersection:

$$r_meet_s(X_1, \dots, X_n) \leftarrow r(X_1, \dots, X_n), s(X_1, \dots, X_n).$$

- ◇ Join:

$$r_joinX2_s(X_1, \dots, X_n) \leftarrow r(X_1, X_2, X_3, \dots, X_n), s(X'_1, X_2, X'_3, \dots, X'_n).$$

- Duplicates an issue: see “setof” later in Prolog.

Deductive Databases

- The subject of “deductive databases” uses these ideas to develop *logic-based databases*.
 - ◇ Often syntactic restrictions (a subset of definite programs) used (e.g. “Datalog” – no functors, no existential variables).
 - ◇ Variations of a “bottom-up” execution strategy used: Use the T_p operator (explained in the theory part) to compute the model, restrict to the query.

Recursive Programming

- Example: ancestors.

```
parent(X,Y) <- father(X,Y).  
parent(X,Y) <- mother(X,Y).
```

```
ancestor(X,Y) <- parent(X,Y).  
ancestor(X,Y) <- parent(X,Z), parent(Z,Y).  
ancestor(X,Y) <- parent(X,Z), parent(Z,W), parent(W,Y).  
ancestor(X,Y) <- parent(X,Z), parent(Z,W), parent(W,K), parent(K,Y).  
...
```

- Defining ancestor recursively:

```
parent(X,Y) <- father(X,Y).  
parent(X,Y) <- mother(X,Y).
```

```
ancestor(X,Y) <- parent(X,Y).  
ancestor(X,Y) <- parent(X,Z), ancestor(Z,Y).
```

- Exercise: define “related”, “cousin”, “same generation”, etc.

Types

- *Type*: a (possibly infinite) set of terms.
- *Type definition*: A program defining a type.
- Example: Weekday:
 - ◇ Set of terms to represent: Monday, Tuesday, Wednesday, ...
 - ◇ Type definition:

```
is_weekday('Monday') <- .  
is_weekday('Tuesday') <- . ...
```
- Example: Date (weekday * day in the month):
 - ◇ Set of terms to represent: date('Monday',23), date(Tuesday,24), ...
 - ◇ Type definition:

```
is_date(date(W,D)) <- is_weekday(W), is_day_of_month(D).  
is_day_of_month(1) <- .  
is_day_of_month(2) <- .  
...  
is_day_of_month(31) <- .
```

Recursive Programming: Recursive Types

- *Recursive types*: defined by recursive logic programs.
- Example: natural numbers (simplest recursive data type):
 - ◇ Set of terms to represent: $0, s(0), s(s(0)), \dots$
 - ◇ Type definition:
`nat(0) <- .`
`nat(s(X)) <- nat(X) .`

A minimal recursive predicate:

one unit clause and one recursive clause (with a single body literal).

- We can reason about *complexity*, for a given *class of queries* (“mode”).
E.g., for mode `nat(ground)` complexity is *linear* in size of number.
- Example: integers:
 - ◇ Set of terms to represent: $0, s(0), -s(0), \dots$
 - ◇ Type definition:
`integer(X) <- nat(X) .`
`integer(-X) <- nat(X) .`

Recursive Programming: Arithmetic

- Defining the natural order (\leq) of natural numbers:

```
less_or_equal(0,X) <- nat(X).
```

```
less_or_equal(s(X),s(Y)) <- less_or_equal(X,Y).
```

- Multiple uses: `less_or_equal(s(0),s(s(0)))`, `less_or_equal(X,0)`, ...
- Multiple solutions: `less_or_equal(X,s(0))`, `less_or_equal(s(s(0)),Y)`, etc.

- Addition:

```
plus(0,X,X) <- nat(X).
```

```
plus(s(X),Y,s(Z)) <- plus(X,Y,Z).
```

- Multiple uses: `plus(s(s(0)),s(0),Z)`, `plus(s(s(0)),Y,s(0))`
- Multiple solutions: `plus(X,Y,s(s(s(0))))`, etc.

Recursive Programming: Arithmetic

- Another possible definition of addition:

```
plus(X,0,X) <- nat(X).
```

```
plus(X,s(Y),s(Z)) <- plus(X,Y,Z).
```

- The meaning of `plus` is the same if both definitions are combined.
- Not recommended: several proof trees for the same query \rightarrow not efficient, not concise. We look for minimal axiomatizations.
- The art of logic programming: finding compact and computationally efficient formulations!

- Try to define: `times(X,Y,Z)` ($Z = X*Y$), `exp(N,X,Y)` ($Y = X^N$), `factorial(N,F)` ($F = N!$), `minimum(N1,N2,Min)`, ...

Recursive Programming: Arithmetic

- Definition of $\text{mod}(X, Y, Z)$

“Z is the remainder from dividing X by Y”

$(\exists Q \text{ s.t. } X = Y * Q + Z \text{ and } Z < Y)$:

$\text{mod}(X, Y, Z) \leftarrow \text{less}(Z, Y), \text{times}(Y, Q, W), \text{plus}(W, Z, X).$

$\text{less}(0, s(X)) \leftarrow \text{nat}(X).$

$\text{less}(s(X), s(Y)) \leftarrow \text{less}(X, Y).$

- Another possible definition:

$\text{mod}(X, Y, X) \leftarrow \text{less}(X, Y).$

$\text{mod}(X, Y, Z) \leftarrow \text{plus}(X1, Y, X), \text{mod}(X1, Y, Z).$

- The second is much more efficient than the first one (compare the size of the proof trees).

Recursive Programming: Arithmetic/Functions

- The Ackermann function:

$$\text{ackermann}(0, N) = N+1$$

$$\text{ackermann}(M, 0) = \text{ackermann}(M-1, 1)$$

$$\text{ackermann}(M, N) = \text{ackermann}(M-1, \text{ackermann}(M, N-1))$$

- In Peano arithmetic:

$$\text{ackermann}(0, N) = s(N)$$

$$\text{ackermann}(s(M), 0) = \text{ackermann}(M, s(0))$$

$$\text{ackermann}(s(M), s(N)) = \text{ackermann}(M, \text{ackermann}(s(M), N))$$

- Can be defined as:

$$\text{ackermann}(0, N, s(N)) \leftarrow .$$

$$\text{ackermann}(s(M), 0, Val) \leftarrow \text{ackermann}(M, s(0), Val).$$

$$\text{ackermann}(s(M), s(N), Val) \leftarrow \text{ackermann}(s(M), N, Val1), \\ \text{ackermann}(M, Val1, Val).$$

- In general, *functions* can be coded as a predicate with one more argument, which represents the output (and additional syntactic sugar often available).
- Syntactic support available (see, e.g., the Ciao *functions* package).

Recursive Programming: Lists

- Binary structure: first argument is *element*, second argument is *rest* of the list.
- We need:
 - ◊ a constant symbol: the empty list denoted by the *constant* `[]`
 - ◊ a functor of arity 2: traditionally the dot “.” (which is overloaded).
- Syntactic sugar: the term `.(X,Y)` is denoted by `[X|Y]` (*X* is the *head*, *Y* is the *tail*).

<u>Formal object</u>	<u>Cons pair syntax</u>	<u>Element syntax</u>
<code>.(a,[])</code>	<code>[a []]</code>	<code>[a]</code>
<code>.(a,.(b,[]))</code>	<code>[a [b []]]</code>	<code>[a, b]</code>
<code>.(a,.(b,.(c,[])))</code>	<code>[a [b [c []]]]</code>	<code>[a, b, c]</code>
<code>.(a,X)</code>	<code>[a X]</code>	<code>[a X]</code>
<code>.(a,.(b,X))</code>	<code>[a [b X]]</code>	<code>[a, b X]</code>

- Note that:

`[a, b]` and `[a|X]` unify with $\{X = [b]\}$

`[a]` and `[a, b|X]` do not unify

`[a]` and `[a|X]` unify with $\{X = []\}$

`[]` and `[X]` do not unify

Recursive Programming: Lists

- Type definition (no syntactic sugar):
`list([]) <- .`
`list.(X,Y) <- list(Y).`
- Type definition (with syntactic sugar):
`list([]) <- .`
`list([X|Y]) <- list(Y).`

Recursive Programming: Lists (Contd.)

- X is a *member* of the list Y:

`member(a, [a]) <-.` `member(b, [b]) <-.` *etc.* \Rightarrow `member(X, [X]) <-.`
`member(a, [a,c]) <-.` `member(b, [b,d]) <-.` *etc.* \Rightarrow `member(X, [X,Y]) <-.`
`member(a, [a,c,d]) <-.` `member(b, [b,d,l]) <-.` *etc.* \Rightarrow `member(X, [X,Y,Z]) <-.`
 \Rightarrow `member(X, [X|Y]) <- list(Y).`

`member(a, [c,a]), member(b, [d,b]).` *etc.* \Rightarrow `member(X, [Y,X]).`
`member(a, [c,d,a]). member(b, [s,t,b]).` *etc.* \Rightarrow `member(X, [Y,Z,X]).`
 \Rightarrow `member(X, [Y|Z]) <- member(X,Z).`

- Resulting definition:

`member(X, [X|Y]) <- list(Y).`
`member(X, [_|T]) <- member(X,T).`

Recursive Programming: Lists (Contd.)

- Resulting definition:

```
member(X, [X|Y]) <- list(Y).
```

```
member(X, [_|T]) <- member(X,T).
```

- Uses of member(X,Y):

- ◇ checking whether an element is in a list (`member(b, [a,b,c])`)
- ◇ finding an element in a list (`member(X, [a,b,c])`)
- ◇ finding a list containing an element (`member(a, Y)`)

- Define: `prefix(X,Y)` (the list X is a prefix of the list Y), e.g.
`prefix([a, b], [a, b, c, d])`
- Define: `suffix(X,Y)`, `sublist(X,Y)`, ...
- Define `length(Xs,N)` (N is the length of the list Xs)

Recursive Programming: Lists (Contd.)

- Concatenation of lists:

- ◇ Base case:

$\text{append}([], [a], [a]) \leftarrow .$ $\text{append}([], [a,b], [a,b]) \leftarrow .$ *etc.*
 $\Rightarrow \text{append}([], Ys, Ys) \leftarrow \text{list}(Ys).$

- ◇ Rest of cases (first step):

$\text{append}([a], [b], [a,b]) \leftarrow .$
 $\text{append}([a], [b,c], [a,b,c]) \leftarrow .$ *etc.*
 $\Rightarrow \text{append}([X], Ys, [X|Ys]) \leftarrow \text{list}(Ys).$

$\text{append}([a,b], [c], [a,b,c]) \leftarrow .$
 $\text{append}([a,b], [c,d], [a,b,c,d]) \leftarrow .$ *etc.*
 $\Rightarrow \text{append}([X,Z], Ys, [X,Z|Ys]) \leftarrow \text{list}(Ys).$

This is still infinite \rightarrow we need to generalize more.

Recursive Programming: Lists (Contd.)

- We note that:

$\text{append}([a,b], Ys, [a|[b|Ys]]) \equiv \text{append}([a,b], Ys, [a|Zs])$

with $Zs = [b|Ys]$

$\text{append}([a,b,c], Ys, [a|[b|Ys]]) \equiv \text{append}([a,b,c], Ys, [a|Zs])$

with $Zs = [b|Ws]$, $Ws = [c|Ys]$.

$\Rightarrow \text{append}([X|Xs], Ys, [X|Zs]) \leftarrow \text{append}(Xs, Ys, Zs)$.

- So, we have:

$\text{append}([], Ys, Ys) \leftarrow \text{list}(Ys)$.

$\text{append}([X|Xs], Ys, [X|Zs]) \leftarrow \text{append}(Xs, Ys, Zs)$.

- Uses of append:

◇ concatenate two given lists: $\leftarrow \text{append}([a,b], [c], Z)$

◇ find differences between lists: $\leftarrow \text{append}(X, [c], [a,b,c])$

◇ split a list: $\leftarrow \text{append}(X, Y, [a,b,c])$

Recursive Programming: Lists (Contd.)

- `reverse(Xs, Ys)`: `Ys` is the list obtained by reversing the elements in the list `Xs`

It is clear that we will need to traverse the list `Xs`

For each element `X` of `Xs`, we must put `X` at the end of the rest of the `Xs` list already reversed:

```
reverse([X|Xs], Ys) <-  
    reverse(Xs, Zs),  
    append(Zs, [X], Ys).
```

How can we stop?

```
reverse([], []) <-.
```

- As defined, `reverse(Xs, Ys)` is very inefficient. Another possible definition:

```
reverse(Xs, Ys) <- reverse(Xs, [], Ys).
```

```
reverse([], Ys, Ys) <-.
```

```
reverse([X|Xs], Acc, Ys) <- reverse(Xs, [X|Acc], Ys).
```

- Find the differences in terms of efficiency between the two definitions.

Recursive Programming: Binary Trees

- Represented by a ternary functor `tree(Element,Left,Right)`.
- Empty tree represented by `void`.
- Definition:

```
binary_tree(void) <- .  
binary_tree(tree(Element,Left,Right)) <-  
  binary_tree(Left),  
  binary_tree(Right).
```

- Defining `tree_member(Element,Tree)`:

```
tree_member(X,tree(X,Left,Right)) <-  
  binary_tree(Left),  
  binary_tree(Right).  
tree_member(X,tree(Y,Left,Right)) <- tree_member(X,Left).  
tree_member(X,tree(Y,Left,Right)) <- tree_member(X,Right).
```

Recursive Programming: Binary Trees

- Defining `pre_order(Tree, Order)`:

```
pre_order(void, []) <- .  
pre_order(tree(X, Left, Right), Order) <-  
    pre_order(Left, OrderLeft),  
    pre_order(Right, OrderRight),  
    append([X|OrderLeft], OrderRight, Order).
```

- Define `in_order(Tree, Order)`, `post_order(Tree, Order)`.

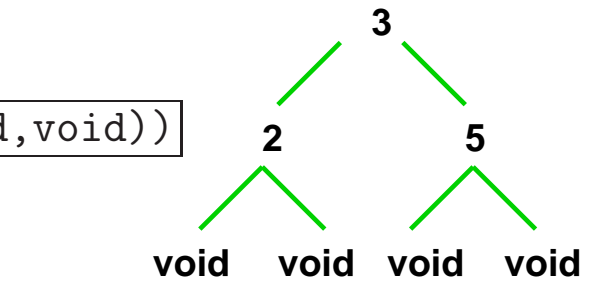
Creating a Binary Tree in Pascal and LP

- In Prolog:

```
T = tree(3, tree(2,void,void), tree(5,void,void))
```

- In Pascal:

```
type tree = ^treerec;
      treerec = record
                data : integer;
                left : tree;
                right: tree;
            end;
var t : tree;
```



```
...
new(t);
new(t^left);
new(t^right);
t^left^left := nil;
t^left^right := nil;
t^right^left := nil;
t^right^right := nil;
t^data := 3;
t^left^data := 2;
t^right^data := 5;
...
```

Polymorphism

- Note that the two definitions of `member/2` can be used *simultaneously*:

```
lt_member(X, [X|Y]) <- list(Y).  
lt_member(X, [_|T]) <- lt_member(X,T).
```

```
lt_member(X, tree(X,L,R)) <- binary_tree(L), binary_tree(R).  
lt_member(X, tree(Y,L,R)) <- lt_member(X,L).  
lt_member(X, tree(Y,L,R)) <- lt_member(X,R).
```

Lists only unify with the first two clauses, trees with clauses 3–5!

- `<- lt_member(X, [b,a,c])`.
`X = b ; X = a ; X = c`
- `<- lt_member(X, tree(b, tree(a, void, void), tree(c, void, void)))`.
`X = b ; X = a ; X = c`
- Also, try (somewhat surprising): `<- lt_member(M,T)`.

Recursive Programming: Manipulating Symbolic Expressions

- Recognizing polynomials in some term X:
 - ◇ X is a polynomial in X
 - ◇ a constant is a polynomial in X
 - ◇ sums, differences and products of polynomials in X are polynomials
 - ◇ also polynomials raised to the power of a natural number and the quotient of a polynomial by a constant

```
polynomial(X,X) <- .
polynomial(Term,X)      <- pconstant(Term).
polynomial(Term1+Term2,X) <- polynomial(Term1,X), polynomial(Term2,X).
polynomial(Term1-Term2,X) <- polynomial(Term1,X), polynomial(Term2,X).
polynomial(Term1*Term2,X) <- polynomial(Term1,X), polynomial(Term2,X).
polynomial(Term1/Term2,X) <- polynomial(Term1,X), pconstant(Term2).
polynomial(Term1^N,X)     <- polynomial(Term1,X), nat(N).
```

Recursive Programming: Manipulating Symb. Expressions (Contd.)

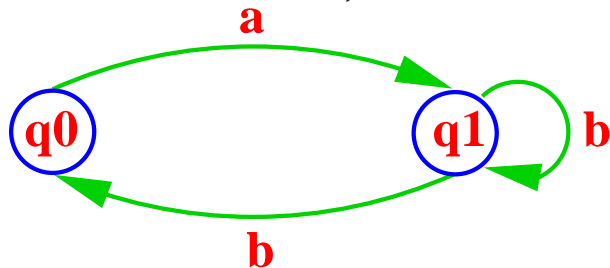
- Symbolic differentiation: `deriv(Expression, X, DifferentiatedExpression)`

```
deriv(X,X,s(0)) <- .
deriv(C,X,0) <- pconstant(C).
deriv(U+V,X,DU+DV) <- deriv(U,X,DU), deriv(V,X,DV).
deriv(U-V,X,DU-DV) <- deriv(U,X,DU), deriv(V,X,DV).
deriv(U*V,X,DU*V+U*DV) <- deriv(U,X,DU), deriv(V,X,DV).
deriv(U/V,X,(DU*V-U*DV)/V^s(s(0))) <- deriv(U,X,DU), deriv(V,X,DV).
deriv(U^s(N),X,s(N)*U^N*DU) <- deriv(U,X,DU), nat(N).
deriv(log(U),X,DU/U) <- deriv(U,X,DU).
...
```

- `<- deriv(s(s(s(0)))*x+s(s(0)),x,Y).`
- A simplification step can be added.

Recursive Programming: Automata (Graphs)

- Recognizing the sequence of characters accepted by the following *non-deterministic, finite automaton* (NFA):



where **q0** is both the *initial* and the *final* state.

- Strings are represented as lists of constants (e.g., [a,b,b]).
- Program:

```
initial(q0) <- .      delta(q0,a,q1) <- .
                        delta(q1,b,q0) <- .
final(q0) <- .      delta(q1,b,q1) <- .

accept(S)           <- initial(Q), accept_from(S,Q).

accept_from([],Q)   <- final(Q).
accept_from([X|Xs],Q) <- delta(Q,X,NewQ), accept_from(Xs,NewQ).
```

Recursive Programming: Automata (Graphs) (Contd.)

- A *nondeterministic, stack, finite automaton* (NDSFA):

```
accept(S) <- initial(Q), accept_from(S,Q, []).
```

```
accept_from([],Q,[]) <- final(Q).
```

```
accept_from([X|Xs],Q,S) <- delta(Q,X,S,NewQ,NewS),  
                           accept_from(Xs,NewQ,NewS).
```

```
initial(q0) <-.
```

```
final(q1) <-.
```

```
delta(q0,X,Xs,q0,[X|Xs]) <-.
```

```
delta(q0,X,Xs,q1,[X|Xs]) <-.
```

```
delta(q0,X,Xs,q1,Xs) <-.
```

```
delta(q1,X,[X|Xs],q1,Xs) <-.
```

- What sequence does it recognize?

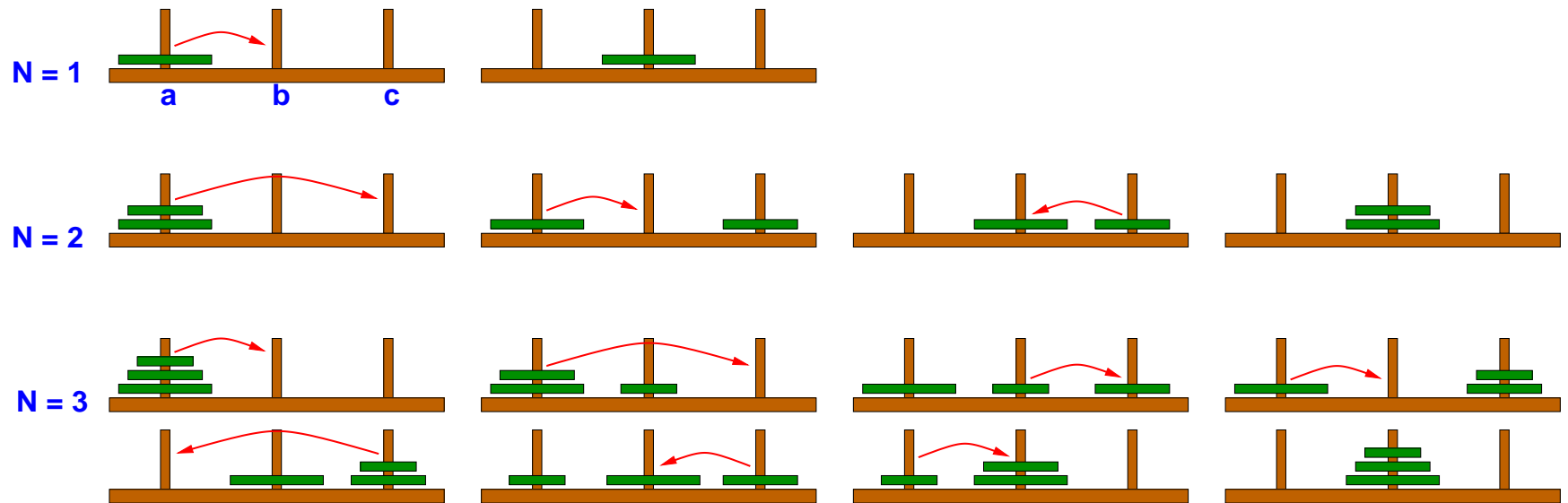
Recursive Programming: Towers of Hanoi

- Objective:

- ◇ Move tower of N disks from peg a to peg b, with the help of peg c.

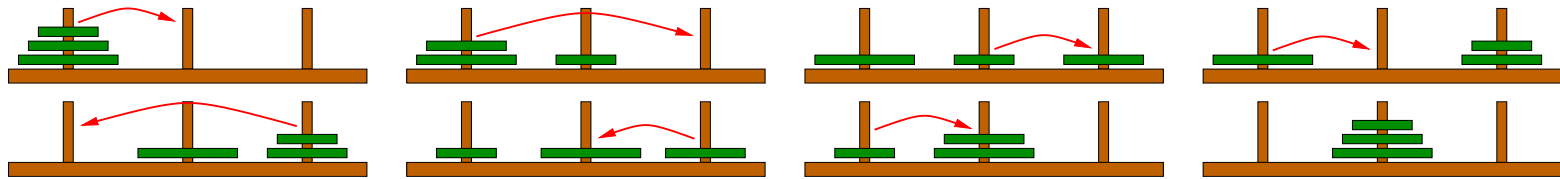
- Rules:

- ◇ Only one disk can be moved at a time.
- ◇ A larger disk can never be placed on top of a smaller disk.



Recursive Programming: Towers of Hanoi (Contd.)

- We will call the main predicate `hanoi_moves(N, Moves)`
- `N` is the number of disks and `Moves` the corresponding list of “moves”.
- Each move `move(A, B)` represents that the top disk in `A` should be moved to `B`.
- *Example:*

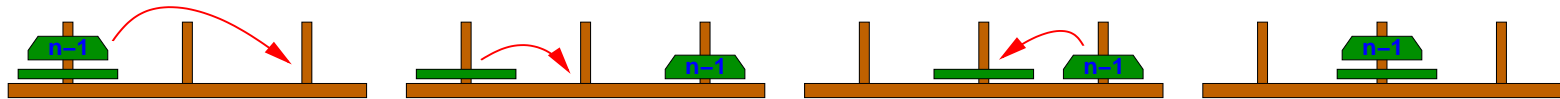


is represented by:

```
hanoi_moves( s(s(s(0))),  
             [ move(a,b), move(a,c), move(b,c), move(a,b),  
               move(c,a), move(c,b), move(a,b) ] )
```

Recursive Programming: Towers of Hanoi (Contd.)

- A general rule:



- We capture this in a predicate `hanoi(N,Orig,Dest,Help,Moves)` where “Moves contains the moves needed to move a tower of N disks from peg Orig to peg Dest, with the help of peg Help.”

```
hanoi(s(0),Orig,Dest,_Help,[move(Orig, Dest)]) <- .  
hanoi(s(N),Orig,Dest,Help,Moves) <-  
    hanoi(N,Orig,Help,Dest,Moves1),  
    hanoi(N,Help,Dest,Orig,Moves2),  
    append(Moves1,[move(Orig, Dest)|Moves2],Moves).
```

- And we simply call this predicate:

```
hanoi_moves(N,Moves) <-  
    hanoi(N,a,b,c,Moves).
```

Learning to Compose Recursive Programs

- To some extent it is a simple question of practice.
- By induction (as in the previous examples): elegant, but generally difficult – not the way most people do it.
- State first the base case(s), and then think about the general recursive case(s).
- Sometimes it may help to compose programs with a given use in mind (e.g., “forwards execution”), making sure it is declaratively correct. Consider also if alternative uses make declarative sense.
- Sometimes it helps to look at well-written examples and use the same “schemas”.
- Global top-down design approach:
 - ◇ state the general problem
 - ◇ break it down into subproblems
 - ◇ solve the pieces
- Again, best approach: practice.