



Meta-interpreters in Prolog

The Power of Prolog

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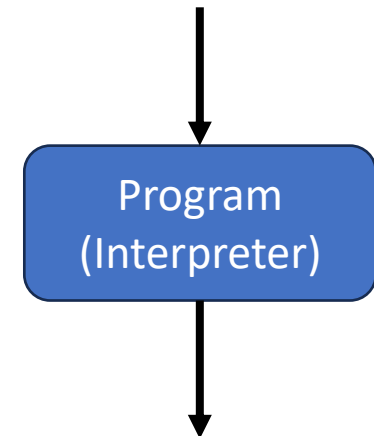
Interpretation is pervasive

Most programs are interpreters for specific languages.

Examples:

- a web browser interprets HTML, JavaScript, HTTP...
- Python interpreter reads and executes Python code.
- many programs interpret command line arguments
- an editor interprets user input, regular expressions etc.
- configuration files and settings need to be interpreted
- documents and graphs are described in PostScript, Tex etc.
- ...

Specific Language Input



Meaningful Behavior
or Execution Result

Meta-interpreters

A *meta-interpreter (MI)* interprets a language similar or identical to its own implementation language.

Why write *MI* in Prolog? — — Create **special-purpose** interpreters

The writing of meta-programs, is particularly easy in Prolog, because:

- The equivalence of programs and data: both are Prolog terms
- Prolog's implicit mechanisms can be used in interpreters
- Prolog is a simple language. Only construct “Head:-Body”

The Coarsest Meta-interpreters

```
prove(A) :- call(A).
```

- This is the **coarsest form** of meta-programming:
By using `call/1`, we **absorb** backtracking, unification, handling of conjunctions, the call stack etc.

Absorption: *Implicitly* using features of the underlying engine

Reification: Making features *explicit*

- We can make these features **explicit** and subsequently adjust and extend them at will
- Differences in meta-interpreters can be characterized in their **granularity**


The Best Known Meta-interpreters

```
prove(true).  
prove((A, B)) :- prove(A), prove(B).  
prove(G) :- clause(G, Body), prove(Body).
```

Need cut or extra
conditions like

$G \neq \text{true},$
 $G \neq (_, _),$

Preliminary: clause/2

`clause(Head, Body)`  There is a clause “Head :- Body.” (by unification)

Example: `h(X,Y) :- f(X), g(Y).
h(u,v).`

`?- clause(h(A,B), Body).`

Body = (f(A),g(B)) % **Unification** is implicit
; Body = true, A = u, B = v % The body of facts is true

Example: `complicated(A) :- g1(A), g2(A), g3(A).`

`?- clause(complicated(Z), Body).`

Body = (g1(Z), g2(Z), g3(Z)).

The Best Known/Vanilla Meta-interpreters

```
prove(true).  
prove((A, B)) :- prove(A), prove(B).  
prove(G) :- clause(G, Body), prove(Body).
```

Need cut or extra
conditions like

$G \neq \text{true},$
 $G \neq (_, _),$

- Explanation:**
- The empty goal represented by the constant **true**
 - A conjunction **(A, B)** is true if **A** is true and **B** is true
(guarantees that the leftmost goal in the conjunction is solved first)
 - A goal **G** is true if there is a clause **G: -Body** in program such that **Body** is true (responsible for giving different solutions on backtracking)

Example: Membership of A List

Program x Program +

```
1 % base
2 prove(true).
3
4 % 合取
5 prove((A,B)):-prove(A), prove(B).
6
7 % goal不是base也不是合取
8 % 递归: 尝试Goal:-Body且prove(Body)
9 prove(G):-G \= true,
10           G \= (_,_),
11           clause(G,Body),
12           prove(Body).
13
14 % member/2定义
15 %如果 X 是列表的第一个元素, 则 X 在列表中。
16 %如果 X 不是第一个元素, 就递归检查剩余部分 Ys。
17 member(X, [X|_]).
18 member(X, [_|Ys]):-member(X, Ys).
```

<https://www.swi-prolog.org/>

member(X, [a,b,c]).	
X	
a	1
b	2
c	3
false	



prove(member(X, [a,b,c])).	
X	
a	1
b	2
c	3
false	

Meta-interpreters Explicit Trace

```
% base
prove(true):-
    writeln('成功: true').

% 递归
prove(G):-G \= true,
    G \= (_,_),
    clause(G,Body),
    format('尝试匹配规则: ~w :- ~w~n', [G, Body]),
    prove(Body).

member(X, [X|_]).
member(X, [_|Ys]) :- member(X, Ys).
```

 `prove(member(X, [a,b,c])).`   

尝试匹配规则: `member(a,[a,b,c]) :- true`

成功: true

X = a

尝试匹配规则: `member(_736,[a,b,c]) :- member(_736,[b,c])`

尝试匹配规则: `member(b,[b,c]) :- true`

成功: true

X = b

尝试匹配规则: `member(_730,[b,c]) :- member(_730,[c])`

尝试匹配规则: `member(c,[c]) :- true`

成功: true

X = c

尝试匹配规则: `member(_730,[c]) :- member(_730,[])`

false

- backtracking
- unification

Meta-interpreters Explicit Trace

?-prove(member(X, [a, b, c])).

clause(member(X, [a, b, c]), B)	{X=a, B=true}
prove(true)	
true Output: X = a	
clause(member(X, [a, b, c]), B)	{B=member(X, [b,c])}
prove(member(X, [b, c]))	
→ clause(member(X, [b, c]), B1)	{X=b, B1=true}
prove(true)	
true Output: X = b	
→ clause(member(X, [b, c]), B1)	{B1=member(X, [c])}
prove(member(X, [c]))	
→ clause(member(X, [c]), B2)	{X=c, B2=true}
prove(true)	
true Output: X = c	
→ clause(member(X, [c]), B2)	{B2=member(X, [])}
prove(member(X, []))	
→ clause(member(X, []), B3)	false
no (more) solutions	

```
prove(true).
prove(G):-G \= true,
          G \= (_,_),
          clause(G, Body),
          prove(Body).

member(X, [X|_]).
member(X, [_|Ys]):-member(X, Ys).
```

Inductive Logic Programming (ILP)

- A form of machine learning
- Given training examples, induce a hypothesis (a set of logical rules) that generalizes

Background knowledge:

reverse(X,Y): − ...
upper(X,Y): − ...
lower(X,Y): − ...

Positive examples:

target(['a','l','i','c','e'], ['E','C','I','L','A']),
target(['a','l','i','c','e'], ['E','C','I','L','A'])

Negative examples:

target(['a','l','i','c','e'], ['a','l','i','c','e'])

Solution:

target(X,Y) : − *reverse*(X,Z), *upper*(Z,Y).

Metarules

- P, Q, R: second-order variables (i.e. can unify with predicate symbols)
- A, B, C: first-order variables (i.e. can unify with constant symbols)

```
metarule([P,Q], [P,A,B], [[Q,A,B]]). % identity
metarule([P,Q], [P,A,B], [[Q,B,A]]). % inverse
metarule([P,Q,R], [P,A,B], [[Q,A],[R,A,B]]). % precon
metarule([P,Q,R], [P,A,B], [[Q,A,B],[R,B]]). % postcon
metarule([P,Q,R], [P,A,B], [[Q,A,C],[R,C,B]]). % chain
```

```
P(A,B):-Q(A,B).
P(A,B):-Q(B,A).
P(A,B):-Q(A),R(A,B).
P(A,B):-Q(A,B),R(B).
P(A,B):-Q(A,C),R(C,B).
```

- A way to define the hypothesis space
- Restrict infinite hypothesis space to make the search feasible
- A hypothesis is an instance of the given metarules

Example

META-RULE	含义	示例
<code>P(A,B) :- Q(A,B)</code>	直接继承	<code>parent(X,Y) :- father(X,Y)</code>
<code>P(A,B) :- Q(B,A)</code>	反向继承	<code>sibling(X,Y) :- sibling(Y,X)</code>
<code>P(A,B) :- Q(A), R(A,B)</code>	前提 + 关系	<code>can_teach(T,C) :- qualified(T), teaches(T,C)</code>
<code>P(A,B) :- Q(A,B), R(B)</code>	关系 + 约束	<code>can_drive(P,C) :- has_license(P,C), registered(C)</code>
<code>P(A,B) :- Q(A,C), R(C,B)</code>	间接关系	<code>grandparent(G,C) :- parent(G,P), parent(P,C)</code>

Meta-Interpretive Learning

<https://github.com/metagol/metagol>

Professor Stephen H. Muggleton

FREng

Professor of Machine Learning

[School of Artificial Intelligence](#)
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- Metagol is an inductive logic programming (ILP) system based on meta-interpretive learning (MIL).
- Metagol is written in Prolog and runs with SWI-Prolog.
- The key idea of MIL is to **use metarules to restrict the form of hypothesis space**

Example: Learning the Grandparent Relation

The following code demonstrates **learning the grandparent relation** given the mother and father relations as background knowledge:

```
:- use_module('metagol').
```

```
%% metagol settings
```

```
body_pred(mother/2).
```

```
body_pred(father/2).
```

```
%% background knowledge
```

```
mother(ann,amy).
```

```
mother(ann,andy).
```

```
mother(amy,amelia).
```

```
mother(linda,gavin).
```

```
father(steve,amy).
```

```
father(steve,andy).
```

```
father(gavin,amelia).
```

```
father(andy,spongebob).
```

```
%% metarules
```

```
metarule([P,Q],[P,A,B],[[Q,A,B]]).
```

```
metarule([P,Q,R],[P,A,B],[[Q,A,B],[R,A,B]]).
```

```
metarule([P,Q,R],[P,A,B],[[Q,A,C],[R,C,B]]).
```

```
%% learning task
```

```
:-
```

```
%% positive examples
```

```
Pos = [
```

```
    grandparent(ann,amelia),
```

```
    grandparent(steve,amelia),
```

```
    grandparent(ann,spongebob),
```

```
    grandparent(steve,spongebob),
```

```
    grandparent(linda,amelia)
```

```
],
```

```
%% negative examples
```

```
Neg = [
```

```
    grandparent(amy,amelia)
```

```
],
```

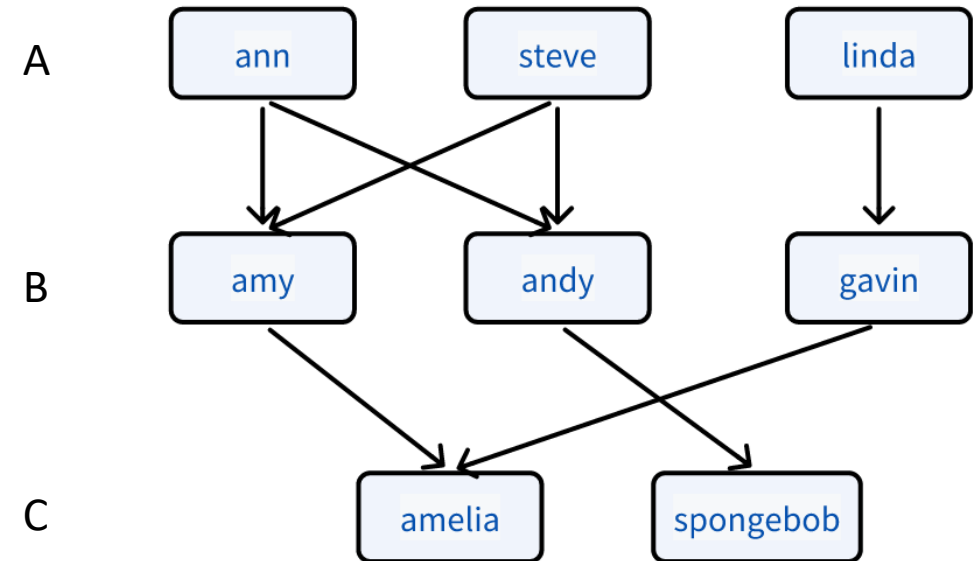
```
learn(Pos,Neg).
```

Running the above program will print the output:

next page

Output

```
% clauses: 1
% clauses: 2
% clauses: 3
grandparent(A,B):-grandparent_1(A,C),grandparent_1(C,B).
grandparent_1(A,B):-mother(A,B).
grandparent_1(A,B):-father(A,B).
```



Why metagol?

- **Predicate Invention** → grandparent_1/2
- **Recursion rule learning** →
 - 1 ancestor(A, D) :- parent(A, D).
 - 2 ancestor(A, D) :- parent(A, B), ancestor(B, D).

Hypothesising an Algorithm from One Example: the Role of Specificity (2023)



Expert Systems in Prolog

The Power of Prolog

2025-07-09

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Introduction

An expert system emulates the decision-making ability of a human expert.

Prolog is very well suited for implementing expert systems due to several reasons:

- Prolog itself can be regarded as a simple ***inference engine*** or ***theorem prover*** that derives conclusions from known rules. Very simple expert systems can be implemented by relying on Prolog's built-in search and backtracking mechanisms.
- Prolog data structures let us flexibly and conveniently represent ***rule-based systems*** that need additional functionality such as probabilistic reasoning.
- We can easily write ***meta-interpreters*** in Prolog to implement custom evaluation strategies of rules.

Example: Animal Identification

我们的目标是编写一个专家系统，用来帮助识别动物。假设我们已经掌握了以下关于动物的知识，也就是推理规则：

- 如果一个动物有毛皮 (*fur*) 并且会叫“汪” (*woof*)，那么它是狗 (**dog**)。
- 如果一个动物有毛皮 (*fur*) 并且会叫“喵” (*meow*)，那么它是猫 (**cat**)。
- 如果一个动物有羽毛 (*feathers*) 并且会叫“嘎” (*quack*)，那么它是鸭子 (**duck**)。

这些规则并不完整，但它们作为一个示例用例，可以很好地说明专家系统的一些核心思想。

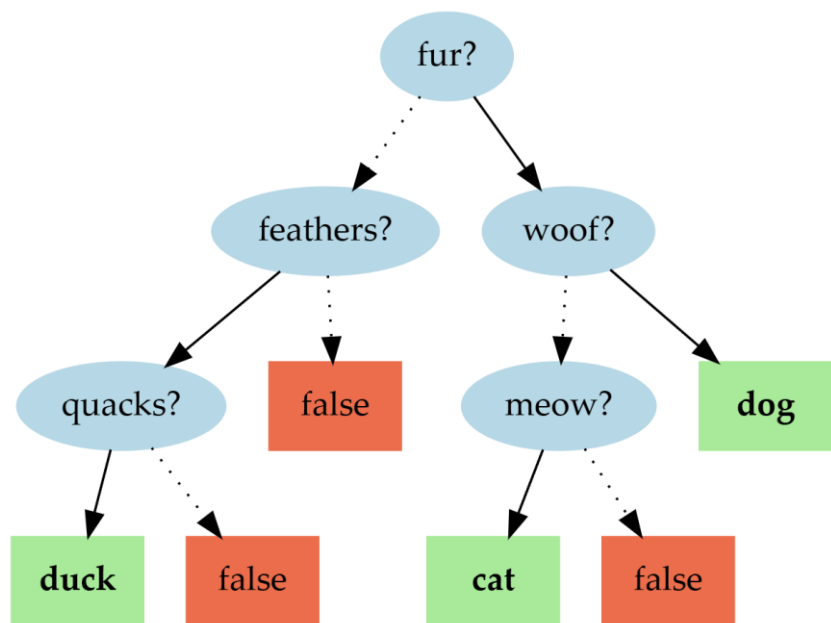
```
animal(dog) :- is_true("has fur"), is_true("says woof").
animal(cat)  :- is_true("has fur"), is_true("says meow").
animal(duck) :- is_true("has feathers"), is_true("says quack").

is_true(Q) :-
    format("~s?\n", [Q]),
    read(yes).
```

```
⚙ animal(A).
has fur?
no.
has fur?
no.
has feathers?
no.
false
```

```
⚙ animal(A).
has fur?
yes.
says woof?
yes.
A = dog
has fur?
yes.
says meow?
no.
has feathers?
no.
false
```

Using a Domain-Specific Language (DSL)



```
animals([animal(dog, [is_true("has fur"), is_true("says woof")]),  
        animal(cat, [is_true("has fur"), is_true("says meow")]),  
        animal(duck, [is_true("has feathers"), is_true("says quack")])]).
```

```
tree(if_then_else("has fur",  
    if_then_else("says woof",  
        animal(dog),  
        if_then_else("says meow",  
            animal(cat),  
            false)),  
    if_then_else("has feathers",  
        if_then_else("says quack",  
            animal(duck),  
            false),  
        false))).
```

Using a Domain-Specific Language (DSL)

```
animal(A) :-  
    tree(T),  
    tree_animal(T, A).  
  
tree_animal(animal(A), A).  
tree_animal(if_then_else(Cond, Then, Else), A) :-  
    ( is_true(Cond) ->  
      tree_animal(Then, A)  
    ; tree_animal(Else, A)  
    ).
```

Such trees can be interpreted in a straight-forward way

 *animal(A).*

has fur?

has feathers?

false

 *animal(A).*

has fur?

says woof?

says meow?

A = cat