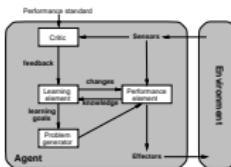


Clinical Decision Support Systems, 23/24

Inês Dutra and Pedro Rodrigues
DCC-FCUP & MEDCIDS-FMUP

ines@dcc.fc.up.pt, pprodrigues@med.up.pt

March 1st, 2024



(Multi) Relational Data Mining

- Inductive Logic Programming (ILP)
- Probabilistic Reasoning – PR
- ILP + PR = SRL
- Statistical Relational Learning

Inductive Logic Programming

Deductive Reasoning

T

U

B

⊧

E

parent(X,Y) :- mother(X,Y)
parent(X,Y) :- father(X,Y)

U

mother(penelope,victoria)
mother(penelope,artur)
father(christopher,victoria)
father(christopher,artur)

⊧

parent(penelope,victoria)
parent(penelope,artur)
parent(christopher,victoria)
parent(christopher,artur)

Inductive Reasoning

E

U

B

⊧

T

parent(penelope,victoria)
parent(penelope,artur)
parent(christopher,victoria)
parent(christopher,artur)

U

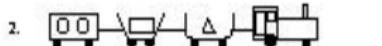
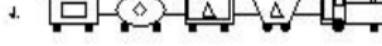
mother(penelope,victoria)
mother(penelope,artur)
father(christopher,victoria)
father(christopher,artur)

⊧

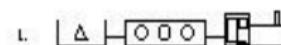
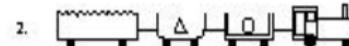
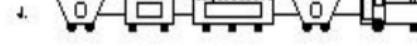
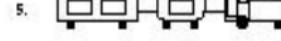
parent(X,Y) :- mother(X,Y)
parent(X,Y) :- father(X,Y)

Inductive Logic Programming

TRAINS GOING EAST

1. 
2. 
3. 
4. 
5. 

TRAINS GOING WEST

1. 
2. 
3. 
4. 
5. 

Inductive Logic Programming

Inductive Logic Programming: example



short(car_12).

closed(car_12).

long(car_11).

long(car_13).

short(car_14).

open_car(car_11).

open_car(car_13).

open_car(car_14).

shape(car_11,rectangle).

shape(car_12,rectangle).

shape(car_13,rectangle).

shape(car_14,rectangle).

load(car_11,rectangle,3).

load(car_12,triangle,1).

load(car_13,hexagon,1).

load(car_14,circle,1).

wheels(car_11,2).

wheels(car_12,2).

wheels(car_13,3).

wheels(car_14,2).

has_car(east1,car_11).

has_car(east1,car_12).

has_car(east1,car_13).

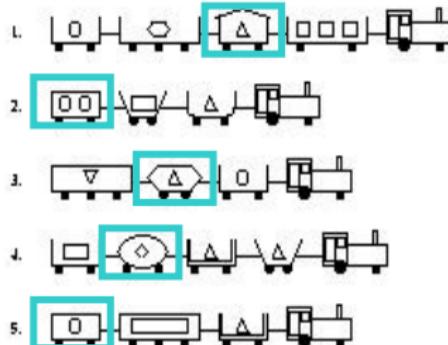
has_car(east1,car_14).

Inductive Logic Programming

Inductive Logic Programming: example

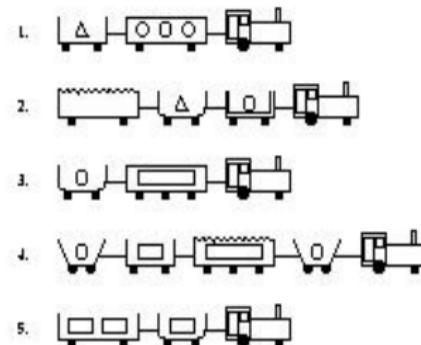
TRAINS GOING EAST

1. TRAINS GOING EAST



TRAINS GOING WEST

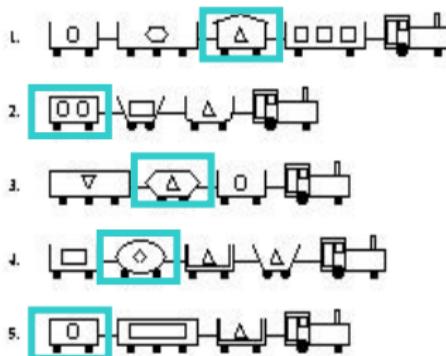
2. TRAINS GOING WEST



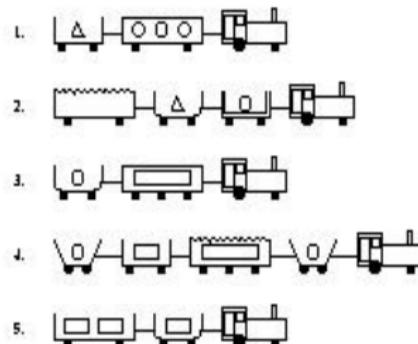
Inductive Logic Programming

Inductive Logic Programming: example

TRAINS GOING EAST



TRAINS GOING WEST



eastbound(T) IF has_car(T,C) AND short(C) AND closed(C)

Inductive Logic Programming

Another example: extracting knowledge from mammogram annotations

```
is_malignant(A) if
  'BIRADS category'(A,b5), 'MassPAO'(A,present),
  'Age'(A,age6570),
  previous_finding(A,B), 'MassesShape'(B,none),
  'Calc_Punctate'(B,notPresent),
  previous_finding(A,C), 'BIRADS_category'(C,b3).
```

The learned rule above says:

A is classified as BI-RADS 5 AND had a mass present ←
in a patient who:
was between the ages of 65 and 70
had two prior mammograms (B, C)
AND prior mammogram (B):
had no mass shape described
had no punctate calcifications
AND prior mammogram (C) was classified as BI-RADS 3 ←

BI-RADS: Breast Imaging Reporting And Data System

Inductive Logic Programming

- More formally:
- Given:
 - A set of examples \mathbf{e} (observations, cases, instances) labelled as positive or negative (class \mathbf{c})
 - A language
 - Possibly, a set of constraints
- Find:
 - A hypothesis \mathbf{h} , such that $\mathbf{h}(\mathbf{e}_i) = \mathbf{c}_i$
 - For most examples

Inductive Logic Programming

- Advantages:
 - Utilization of a language that is easy to interpret
 - More concise classifiers
 - More powerful representation: relations
- Disadvantages:
 - Very large search space
 - Non-probabilistic classification

Properties

- Prior satisfiability

$$B \wedge E^- \not\models \square$$

(H is not a consequence of B and E-)

- Posterior sufficiency

$$B \wedge H \models E^+$$

(H allows to explain E+ relative to B)

- Posterior satisfiability

$$B \wedge H \wedge E^- \not\models \square$$

(B and H are consistent with E-)

- Prior necessity

$$B \not\models E^+$$

(some e+ must be false relative to the model found for B)

ILP: A Common Approach

- Use a greedy covering algorithm.
 - Repeat while some positive examples remain uncovered (not entailed):
 - Find a *good clause* (one that covers as many positive examples as possible but no/few negatives).
 - Add that clause to the current theory, and remove the positive examples that it covers.
 - ILP algorithms use this approach but vary in their method for finding a *good clause*.



Some ILP Systems

- PROGOL, ALEPH (top-down): **saturates** first **uncovered** positive example, and then performs **top-down admissible search** of the **lattice** above this saturated example.
- GOLEM (**bottom-up**), FOIL (top-down), LINUS/DINUS.
- Tilde, Claudien, IndLog, ...

ILP Saturation

- Consists of building a *bottom clause* (seed)
- Incorporates background knowledge to an atomic formula
- Example: (gene that codes for a protein responsible for metabolism)

```
metabolism(A) :-  
    essential(A,'Non-Essential'), motif(A,'PS00510'), chromosome(A,'14'),  
    interaction(A,B,C,E),  
    essential(B,'Non-Essential'), motif(B,'PS00188'), chromosome(B,'2'),  
    interaction(A,F,D,G),  
    intertype(C,'Genetic'), intertype(D,?),  
    interaction(B,A,C,E),  
    interaction(B,H,C,I),  
    interaction(F,A,D,G),  
    interaction(H,B,C,I), interaction(H,_,_,_).
```

ILP: Aleph

- Procedure to extract theories from examples
- Complete (branch-and-bound) search for best clause in the **whole** space
- Search subject to several user control settings
 - Max clause length
 - Max chaining length
 - Minacc
 - Max nodes
 - Search strategy, etc.



ILP: Aleph

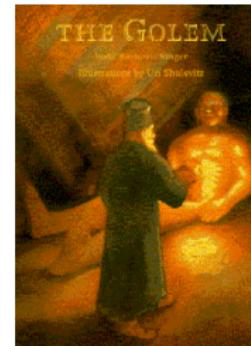
- Aleph
 - Desenvolvido na Universidade de Oxford por Ashwin Srinivasan

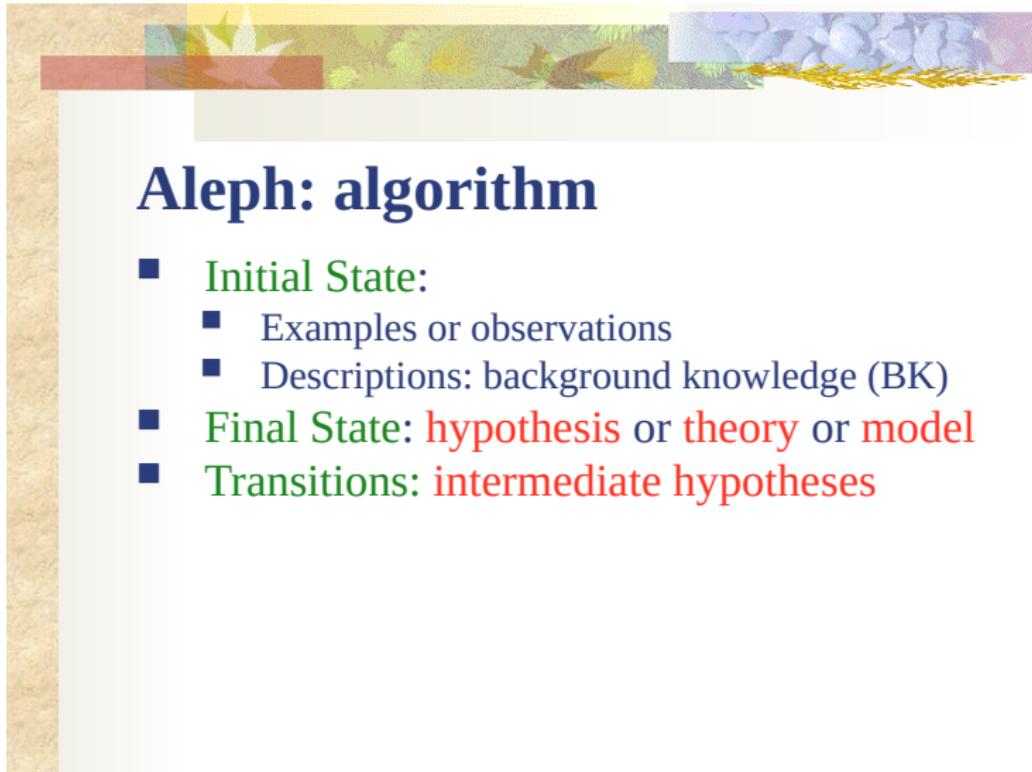
<http://www.comlab.ox.ac.uk/oucl/research/areas/machlearn/Aleph/>

ILP: Aleph

Then the Rabbi said,
*“Golem, you have not been
completely formed, but I am about to
finish you now... You will do as I will
tell you.”*

Saying these words, Rabbi Leib
finished engraving the letter **Aleph**.
Immediately the golem began to rise.





Aleph: algorithm

- Initial State:
 - Examples or observations
 - Descriptions: background knowledge (BK)
- Final State: hypothesis or theory or model
- Transitions: intermediate hypotheses

Aleph: algorithm

- Select example
- Build most-specific-clause (**bottom clause**)
- Search. Find a clause more general than the bottom clause
- Remove redundant. The clause with the best score is added to the current theory, and all examples made redundant are removed. This step is sometimes called the "**cover removal**" step. Note here that the best clause may make clauses other than the examples redundant
- Return to first step

Aleph: Search

Level 0

`eastbound(A)`

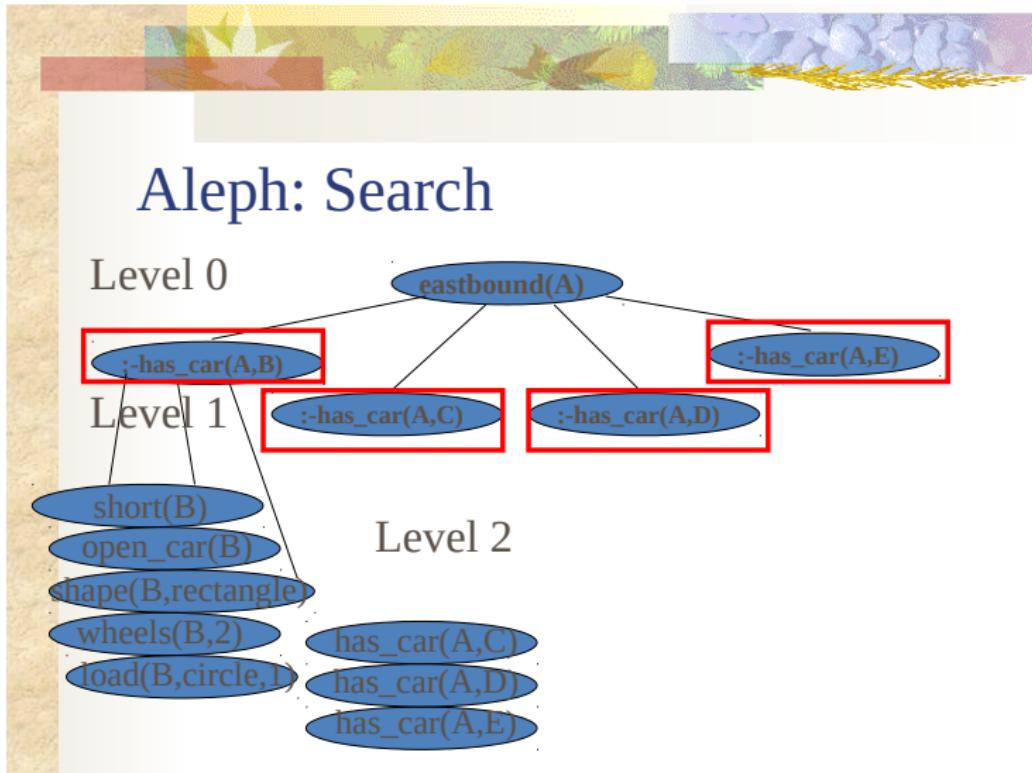
Level 1

`:-has_car(A,B)`

`:-has_car(A,C)`

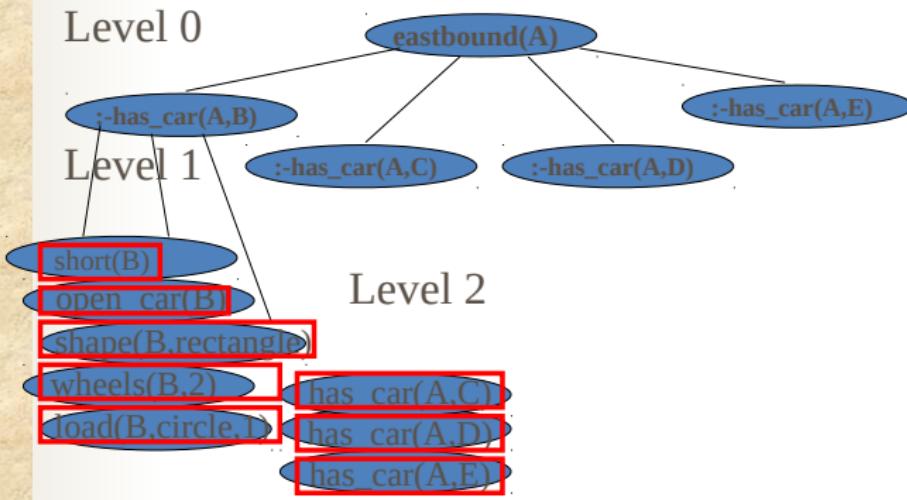
`:-has_car(A,D)`

`:-has_car(A,E)`



Aleph: Search

Level 0





Aleph: Knowledge Representation

Input Files: Prolog Syntax

dtp.b: BK

dtp.f: pos examples

dtp.n: neg examples

Representation: BK

```
chromosome('G234064','1').  
chromosome('G234065','1').  
chromosome('G234070','1').  
chromosome('G234073','1').  
chromosome('G234074','1').  
chromosome('G234076','1').  
chromosome('G234084','2').  
chromosome('G234085','2').  
chromosome('G234089','2').
```

Representation: BK

```
interaction('G234062','G235011','Physical',?).  
interaction('G234064','G234126','Genetic-  
Physical','0.9141').  
interaction('G234064','G235065','Genetic-  
Physical','0.7515').  
interaction('G234064','G235571','Physical','0.9691').  
interaction('G234065','G234073','Physical','0.7492').  
interaction('G234065','G235042','Physical','-0.4659').
```

Representation: Examples

```
metabolism('G239098').  
metabolism('G234980').  
metabolism('G235245').  
metabolism('G234108').  
metabolism('G238387').  
metabolism('G240504').  
metabolism('G236733').
```

Example of clause learned

```
metabolism(A) :-  
    chromosome(A, '15'),  
    interaction(A, B, _, _),  
    complex(B, 'Transcription  
complexes/Transcriptosome').
```

A and B are variables that represent genes

Aleph: algorithm

- Example: Michalski's trains

Aleph: algorithm

- Saturation (saturated / bottom clause):

eastbound(A) :-

```
has_car(A,B), has_car(A,C), has_car(A,D), has_car(A,E),
short(B), short(D), closed(D), long(C),
long(E), open_car(B), open_car(C), open_car(E),
shape(B,rectangle), shape(C,rectangle), shape(D,rectangle),
shape(E,rectangle),
wheels(B,2), wheels(C,3), wheels(D,2), wheels(E,2),
load(B,circle,1), load(C,hexagon,1), load(D,triangle,1),
load(E,rectangle,3).
```

Aleph: algorithm

- Search: most general clause

eastbound(A) :-

```
has_car(A,B), has_car(A,C), has_car(A,D), has_car(A,E),
short(B), short(D), closed(D), long(C),
long(E), open_car(B), open_car(C), open_car(E),
shape(B,rectangle), shape(C,rectangle), shape(D,rectangle),
shape(E,rectangle),
wheels(B,2), wheels(C,3), wheels(D,2), wheels(E,2),
load(B,circle,1), load(C,hexagon,1), load(D,triangle,1),
load(E,rectangle,3).
```

Aleph: algorithm

- Search: add possible descendants (candidate literals of level 1)

`eastbound(A) :-`

```
has_car(A,B), has_car(A,C), has_car(A,D), has_car(A,E),
short(B), short(D), closed(D), long(C),
long(E), open_car(B), open_car(C), open_car(E),
shape(B,rectangle), shape(C,rectangle), shape(D,rectangle),
shape(E,rectangle),
wheels(B,2), wheels(C,3), wheels(D,2), wheels(E,2),
load(B,circle,1), load(C,hexagon,1), load(D,triangle,1),
load(E,rectangle,3).
```

Aleph: algorithm

- Search: add possible descendants of level 2

```
eastbound(A) :-
```

```
    has_car(A,B), has_car(A,C), has_car(A,D), has_car(A,E),  
    short(B), short(D), closed(D), long(C),  
    long(E), open_car(B), open_car(C), open_car(E),  
    shape(B,rectangle), shape(C,rectangle), shape(D,rectangle),  
    shape(E,rectangle),  
    wheels(B,2), wheels(C,3), wheels(D,2), wheels(E,2),  
    load(B,circle,1), load(C,hexagon,1), load(D,triangle,1),  
    load(E,rectangle,3).
```

Aleph: algorithm

- Search: second descendant of level 1

eastbound(A) :-

```
has_car(A,B), has_car(A,C), has_car(A,D), has_car(A,E),
short(B), short(D), closed(D), long(C),
long(E), open_car(B), open_car(C), open_car(E),
shape(B,rectangle), shape(C,rectangle), shape(D,rectangle),
shape(E,rectangle),
wheels(B,2), wheels(C,3), wheels(D,2), wheels(E,2),
load(B,circle,1), load(C,hexagon,1), load(D,triangle,1),
load(E,rectangle,3).
```

Aleph: algorithm

- Search: descendants of second descendant...

eastbound(A) :-

```
has_car(A,B), has_car(A,C), has_car(A,D), has_car(A,E),
short(B), short(D), closed(D), long(C),
long(E), open_car(B), open_car(C), open_car(E),
shape(B,rectangle), shape(C,rectangle), shape(D,rectangle),
shape(E,rectangle),
wheels(B,2), wheels(C,3), wheels(D,2), wheels(E,2),
load(B,circle,1), load(C,hexagon,1), load(D,triangle,1),
load(E,rectangle,3).
```

Aleph: example of run

• aleph_trains

Aleph: how to run?

- You need to have a Prolog system
 - Yap: <http://yap.sourceforge.net> OU
 - SWI: <http://www.swi-prolog.org>
- Aleph:
<http://www.comlab.ox.ac.uk/oucl/research/areas/machlearn/Aleph/>
- Files: .b, .f, .n
- To make things easier: everything in the same directory!

Aleph: Basic Commands

- read_all
- reduce
- induce

Aleph: Parameters

```
:- set(clauselength,5).  
:- set(depth, 200).  
:- set(i,3).  
:- set(noise,0).  
:- set(minacc,0.7).  
:- set(nodes,1000000).  
:- set(m,20).  
:- set(evalfn,mestimate).  
:- set(test_pos,'/u/dutra/Protein/prot_test_set.f').  
:- set(test_neg,'/u/dutra/Protein/prot_test_set.n').  
:- set(optimise_clauses,true).  
  
:- set(record,true).  
:- set(recordfile,'prot_train_set.out').  
:- set(samplesize,0).
```

$$\text{Strength estimate} = (\text{support} + m * \text{prior}) / (\text{coverage} + m)$$

$M \rightarrow 0$, strength \rightarrow precision

Support = True positives

Coverage = True positives + false negatives

Aleph: Modes and Types

```
:- modeh(1,eastbound(+train)).  
:- modeb(1,short(+car)).  
:- modeb(1,closed(+car)).  
:- modeb(1,long(+car)).  
:- modeb(1,open_car(+car)).  
:- modeb(1,double(+car)).  
:- modeb(1,jagged(+car)).  
:- modeb(1,shape(+car,#shape)).  
:- modeb(1,load(+car,#shape,#int)).  
:- modeb(1,wheels(+car,#int)).  
:- modeb(*,has_car(+train,-car)).  
  
:- determination(eastbound/1,short/1).  
:- determination(eastbound/1,closed/1).  
:- determination(eastbound/1,long/1).  
:- determination(eastbound/1,open_car/1).  
:- determination(eastbound/1,double/1).  
:- determination(eastbound/1,jagged/1).  
:- determination(eastbound/1,shape/2).  
:- determination(eastbound/1,wheels/2).  
:- determination(eastbound/1,has_car/2).  
:- determination(eastbound/1,load/3).
```

Aleph: Modes and Types

```
:- modeh(1,metabolism(+gene)).  
:- modeb(1,essential(+gene,#essential)).  
:- modeb(1,class(+gene,#class)).  
:- modeb(1,complex(+gene,#complex)).  
:- modeb(1,phenotype(+gene,#phenotype)).  
:- modeb(1,motif(+gene,#motif)).  
:- modeb(1,chromosome(+gene,#chromosome)).  
:- modeb(*,gte(+number,#number)).  
:- modeb(*,interaction(+gene,-gene,-intertype,-number)).  
:- modeb(1,intertype(+intertype,#intertype)).
```

Example: drug discovery using Aleph refinement operators

- Given:
 - Molecules active and inactive for dtp
 - Their description in terms of coordinates and bonds
- Find small structures that model active molecules

Examples: drug discovery

- Examples of dtp groups:
`hydrophobic(m752,
 hyphob([a2, a3, a5, a8, a7, a4, a2],
 2.16452, -0.833917, 3.6379)).`
`hacc(m9706,
 hacc(a10, -6.2969, -1.3684, -0.4631)).`

Example: drug discovery

- Utilisation of refinement operator

```
refine(false,Clause):-
```

```
    member(Point1, [hydrophobic(M,P1), hdonor(M,P1),halogen(M,P1),hacc(M,P1)]),  
    member(Point2,[hydrophobic(M,P2),hdonor(M,P2),halogen(M,P2),hacc(M,P2)]),  
    Clause = (active(M) :- Point1, Point2, dist(M,P1,P2,D1,E)).
```

```
refine(Clause1,Clause2):-
```

```
    Clause1 = (active(M) :- Point1,Point2, dist(M,P1,P2,D1,E)), member(Point3,  
        [hydrophobic(M,P3),hdonor(M,P3),halogen(M,P3),hacc(M,P3)]),  
    Clause2 = (active(M) :- Point1, Point2, dist(M,P1,P2,D1,E),  
        Point3, dist(M,P1,P3,D2,E), dist(M,P2,P3,D3,E)).
```

- Reduce search space!!!