

# Logic Programming

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Outline

## Outline of the Lecture

### Monotone Logic Programs

introduction, basic constructs, logic foundations, unification, semantics, database and recursive programming, termination, complexity

### Incomplete Data Structures and Constraints

incomplete data structures, definite clause grammars, **constraint logic programming**, **answer set programming**

### Full Prolog

semantics (revisted), correctness proofs, meta-logical predicates, cuts non-deterministic programming, efficient programs, complexity

## Summary of Last Lecture

### Definitions (CLP on finite domains)

- `use_module(library(clpfd))` loads the clpfd library
- `Xs ins N .. M` specifies that all values in `Xs` must be in the given range
- `all_different(Xs)` specifies that all values in `Xs` are different
- `label(Xs)` all variables in `Xs` are evaluated to become values
- `#=`, `#\=`, `#>`, ... like the arithmetic comparison operators, but may contain (constraint) variables

### standard approach

- load the library
- specify all constraints
- call `label` to start efficient computation of solutions

Outline

## Performance

*"Neng-Fa Zhou, the author of B-Prolog, has kindly integrated our constraint solver in his benchmarks, available from <http://www.probp.com/performance.htm>. The results show that our solver is on average two orders of magnitude slower on these benchmarks than the fastest system (B-Prolog itself), and about 30 times slower than the constraint solver of SICStus Prolog."*<sup>1</sup>

<sup>1</sup>Markus Triska: The Finite Domain Constraint Solver of SWI-Prolog. FLOPS 2012: 307-316

## Example

a  $n \times n$  square is **magic** if cells contain  $\{1, \dots, n^2\}$  and the row sums, the column sums and the sums of both diagonals are all equal

## Example

```

magicsquare3(Xs) :-
    magicsquare3_(Xs,Ys),
    labeling([],Ys).

magicsquare3_([[X1,X2,X3],[X4,X5,X6],[X7,X8,X9]],Ys) :-
    Ys = [X1,X2,X3,X4,X5,X6,X7,X8,X9],
    Ys ins 1..9,
    all_different(Ys),
    X1 + X2 + X3 #= N, X4 + X5 + X6 #= N,
    X7 + X8 + X9 #= N,
    X1 + X4 + X7 #= N, X2 + X5 + X8 #= N,
    X3 + X6 + X9 #= N,
    X1 + X5 + X9 #= N, X7 + X5 + X3 #= N.

```

## Example

/\* magic square + bit math:  $N = n \cdot (n^2 + 1) / 2$  \*/

```

magicsquare3_2(Xs) :-
    magicsquare3_(Xs,Ys),
    labeling([],Ys).

magicsquare3_2([[X1,X2,X3],[X4,X5,X6],[X7,X8,X9]],Ys) :-
    Ys = [X1,X2,X3,X4,X5,X6,X7,X8,X9],
    Ys ins 1..9,
    all_different(Ys),
    N #= 15,
    X1 + X2 + X3 #= N, X4 + X5 + X6 #= N,
    X7 + X8 + X9 #= N,
    X1 + X4 + X7 #= N, X2 + X5 + X8 #= N,
    X3 + X6 + X9 #= N,
    X1 + X5 + X9 #= N, X7 + X5 + X3 #= N.

```

## Example

remove symmetric solutions, due to rotations and mirroring

## Example

```

magicsquare3nred(Xs) :-
    magicsquare3nred_(Xs,Ys),
    labeling([],Ys).

magicsquare3nred_(Xs,Ys) :-
    magicsquare3_(Xs,Ys),
    Ys = [X1,X2,X3,X4,_X5,X6,_X7,X8,X9],
    X1 #> X3,
    X6 #> X9,
    X2 #> X4,
    X6 #> X8.

:- time(magicsquare3nred(_Xs)).
% 177,052 inferences, 0.060 CPU in 0.060 seconds
% proper testing shows even speed up over clever variant

```

## Efficient Constraint Logic Programming

## Strategies for Solutions

- **take termination seriously**  
non-termination is a sign of inefficiency
- **choose suitable labeling strategies**
- **use system predicates**

```

:- Zs = [A,B,C], Zs ins 1..2,
   A #\= B, B #\= C, A #\= C.
:- Zs = [A,B,C], Zs ins 1..2,
   all_different(Zs).

```

- **make use of redundant constraints**  
recall the magic square example, where the sums equal  $n \cdot (n^2 + 1) / 2$ ; using this redundant constraint, the search may be quicker; however, such constraints are difficult to find

## Labeling Strategies

### Strategies for Solutions (cont'd)

- **minimise the solution space**  
consider the exclusion of rotations and symmetries for magic square
- **improve representation of solutions**  
inefficient/redundant representations increase the solution space unnecessarily

### Definition

`labeling (+Options,+Vars)` assign a value to each variable in *Vars*; three categories of options exist

- variable selection strategy
- value order strategy
- branching strategy

### Definition (variable selection strategy)

- **leftmost**, select the variables in the order they occur in *Vars* (default)
- **min**, select the leftmost variable with lowest lower bound next

```
:- X in 1..2, Y in 3..4, labeling([min],[X,Y]).
X = 1, Y = 3;
X = 1, Y = 4;
X = 2, Y = 3;
X = 2, Y = 4
```

- **max**, select the leftmost variable with highest upper bound next

```
:- X in 1..2, Y in 3..4, labeling([min],[X,Y]).
X = 1, Y = 3;
X = 2, Y = 3;
X = 1, Y = 4;
X = 2, Y = 4
```

- **ff**, first fail, select the leftmost variable with smallest domain next, in order to detect infeasibility early

### Definition (variable selection strategy (cont'd))

- **ffc**, from the variables with smallest domain, select the one occurring most often in constraints

### Definition (value order strategy)

- **up**, try the elements of the domain in ascending order
- **down**, in descending order

### Definition (branching strategy)

- **step**, for each variable *X*, the choice is between  $X = V$  and  $X \neq V$  (*V* determined by value order)
- **enum**, enumerate the domain of *X* according to the value order
- **bisect**, choice is between  $X \leq M$  and  $X > M$  (*M* the midpoint of the domain)

## The New Kid on the Block

### Answer Set Programming

- novel approach to modelling and solving search and optimisation problems
- $\neg$  programming, but a specification language
- $\neg$  Turing complete
- purely declarative
- restricted to finite models

### Success Stories

- team building for cargo at Gioia Tauro Seaport
- expert system in space shuttle
- natural language processing
- ...

## Propositional Setting

### Definitions

- atoms, facts, rules are defined as before
- only constants (= propositions) are allowed as atoms
- negation is negation as failure
- disjunctions may appear in the head
- an **answer set** is a set of atoms corresponding to the minimal model of the program

### Example (Negation as Failure)

```
light_on :- power_on, not broken.
power_on.
```

answer set:  $\{power\_on, light\_on\}$

### Example (Disjunctive Heads)

```
open | closed :- door.
```

answer sets:  $\{open\}, \{closed\}$

### Example

```
a | b.
a | c.
```

answer sets:  $\{a\}$  and  $\{b, c\}$

```
a | b.
a :- b.
```

answer set:  $\{a\}$ , but not  $\{b\}$  nor  $\{a, b\}$

### Definition

**constraints** are negative assertions, representing fact that must not occur in any model of the program

### Example

```
a :- not a, b.
```

any answer set must not contain  $b$  and constraint simplifies to

```
:- b.
```

same notation, but different use than an assertion

### Additional Features

- finite choice functions:  $\{fact_1, fact_2, fact_3\}$ .
- choice and counting:  $1\{fact_1, fact_2, fact_3\}2$ .  
“1” or “2” may be missing

## First-Order Setting

### Definition

- extension of first-order language
- no function symbols

### Example (3-colouring)

```
red(X) | green(X) | blue(X).
:- red(X), red(Y), edge(X,Y).
:- green(X), green(Y), edge(X,Y).
:- blue(X), blue(Y), edge(X,Y).
```

### Example ((part of) 8-queens problem)

```
:- row(X), not (1 = count(Y : queen(X,Y)))
```

expresses that exactly one queen appears in every row and column

## Grounders and Solvers



### Grounders

- DLV (DLV Systems, Calabria)
- Gringo (University of Potsdam)
- Iparse (University of Helsinki)

### Solvers

- clasp (University of Potsdam)
- cmodels (University of Austin)
- smodels (University of Helsinki)

## Prolog and Answer Set Programming

- |  |   |
|--|---|
| <ul style="list-style-type: none"> <li>• proof search</li> <li>• Turing complete</li> <li>• control</li> <li>• efficiency</li> </ul> | <ul style="list-style-type: none"> <li>• model search</li> <li>• finite domain</li> <li>• specification language</li> <li>• generality</li> </ul> |
|--|---|

### Example

```

hanoi(0,_,_,_,[]).
hanoi(N,X,Y,Z,Ls) :-
    N > 0, M is N - 1,
    hanoi(M,X,Z,Y,Ls0),
    append(Ls0,[move(N,X,Z)],Ls1),
    hanoi(M,Y,X,Z,Ls2),
    append(Ls1,Ls2,Ls).
  
```

### Example

```

disk(1..n).                peg(a;b;c).
transition(0..pathlength-1). situation(0..pathlength).
location(Peg) :- peg(Peg).   location(Disk) :- disk(Disk).
#domain disk(X;Y).          #domain peg(P;P1;P2).
#domain transition(T).      #domain situation(I).
#domain location(L;L1).

on(X,L,T+1) :- on(X,L,T), not otherloc(X,L,T+1).
otherloc(X,L,I) :- on(X,L1,I), L1!=L.
:- on(X,L,I), on(X,L1,I), L1!=L.
inpeg(X,P,I) :- on(X,L,I), inpeg(L,P,I).   inpeg(P,P,I).
top(P,L,I) :- inpeg(L,P,I), not covered(L,I).
covered(L,I) :- on(X,L,I).
:- on(X,Y,I), X>Y.
on(X,L,T+1) :- move(P1,P2,T), top(P1,X,T), top(P2,L,T).
:- move(P1,P2,T), top(P1,P1,T). movement(P1,P2) :- P1 != P2.
1 {move(A,B,T) : movement(A,B)} 1.
on(n,a,0).                on(X,X+1,0) :- X<n.
onewrong :- not inpeg(X,c,pathlength).
:- onewrong.
  
```