# Lecture 9

Constraint Logic Programming ITI0021 J.Vain 2018

### **Definitions**

- Constraint programming (CP) is a declarative formalism that lets you describe conditions a solution must satisfy.
- CP can be used to model and solve various combinatorial problems such as
  - planning,
  - scheduling
  - allocation of tasks.

### CLP in SWI-Prolog

- library(clpfd): Constraint Logic Programming over Finite Domains
- library(clpr): Constraint Logic Programming over Rationals and Reals<sup>1</sup>

<sup>1</sup>- library must be loaded explicitly before using it:

```
:- use_module(library(clpq)).
```

# Constraint Logic Programming over Finite Domains (clpfd)

- Predicates of clpfd are
  - finite domain constraints, which are relations over integers.
  - generalise arithmetic evaluation of integer expressions in that propagation can proceed in all directions.
- Enumeration predicates let systematically search for solutions on variables whose domains are finite.

# Finite domain expressions

```
an integer
a variable
-Expr
Expr + Expr
Expr * Expr
Expr - Expr
min(Expr,Expr)
max(Expr,Expr)
Expr mod Expr
abs(Expr)
Expr / Expr
```

- Given value
- Unknown value
- Unary minus
- Addition
- Multiplication
- Subtraction
- Minimum of two expressions
- Maximum of two expressions
- Remainder of integer division
- Absolute value
- Integer division

### Finite domain constraints

The constraints in/2, #=/2, #=/2, #</2, #>/2, #=</2, and #>=/2 can be *reified*, which means reflecting their truth values by integers 0 and 1.

# Reifiable constraints and Boolean variables

Let P and Q denote reifiable constraints, then			
#\Q	true	iff	Q is false
P #\/ Q	true	iff	either <i>P</i> or <i>Q</i>
P #/\ Q	true	iff	both <i>P</i> and <i>Q</i>
P #<==> Q	true	iff	P and Q are equivalent
P #==> Q	true	iff	P implies Q
P #<== Q	true	iff	Q implies P

# Example

- ?- [library(clpfd)].
- ?- X #> 3.
- X in 4..sup.
- ?- X #\= 20.
- X in inf..19 \/ 21..sup.
- ?- 2\*X #= 10.
- X = 5.
- ?- X\*X #= 144.
- X in  $-12 \/12$ .

# Example

```
?- 4*X + 2*Y #= 24, X + Y #= 9, [X,Y] ins 0..sup.
X = 3,
Y = 6.
?- Vs = [X,Y,Z], Vs ins 1...3, all_different(Vs), X = 1, Y \# = 2.
Vs = [1, 3, 2],
X = 1,
Y = 3,
Z = 2.
?- X #= Y #<==> B, X in 0..3, Y in 4..5.
B = 0,
X in 0..3,
Y in 4..5.
```

# Usage of CLP

- Common scenario:
  - 1. Post the desired constraints among the variables of a model
  - 2. use enumeration predicates to search for solutions.

<u>Example</u> of constraint satisfaction problem: cryptoarithmetic puzzle SEND + MORE = MONEY,

where different letters denote distinct integers between 0 and 9.

# Example (continues)

Modeling <u>SEND + MORE = MONEY</u> in CLP(FD):

```
:- use_module(library(clpfd)).
puzzle([S,E,N,D] + [M,O,R,E] = [M,O,N,E,Y]) :-
   Vars = [S,E,N,D,M,O,R,Y],
   Vars ins 0..9,
    all different(Vars),
       S*1000 + E*100 + N*10 + D +
       M*1000 + O*100 + R*10 + E
        #=
        M*10000 + O*1000 + N*100 + E*10 + Y
   M \# = 0, S \# = 0.
                                            % largest decimal places cannot
                                               be 0-s
```

### Example (continues)

Sample query and its result:

```
?- puzzle(As+Bs=Cs).
As = [9, G10107, G10110, G10113],
Bs = [1, 0, G10128, G10107],
Cs = [1, 0, G10110, G10107, G10152],
G10107 in 4..7,
1000*9+91* G10107+ -90* G10110+ G10113+ -9000*1+ -
  900*0+10<sup>*</sup> G10128+ -1* G10152#=0,
all_different([_G10107, _G10110, _G10113, _G10128, _G10152, 0, 1, 9]),
G10110 in 5..8,
G10113 in 2..8,
G10128 in 2..8,
G10152 in 2..8.
```

# Example (continues)

- Constraint solver deduces bounds for all variables.
- Keeping the modeling part separate from the search allows more easily experiment with different search strategies.
- Labeling can then be used to search for solutions:

### Example

?- puzzle(As+Bs=Cs), label(As).

% label(As) – is trying out explicit values for the finite domain variables

### Variable domain constraints

#### ?Var in +Domain

Var is an element of Domain where the Domain is one of:

- Integer
   Singleton set consisting only of Integer.
- All integers I such that Lower =< I =< Upper. Lower must be an integer or the atom inf, which denotes negative infinity.

  Upper must be an integer or the atom sup, which denotes positive infinity.
- Domain1 \/ Domain2
   The union of Domain1 and Domain2.

### Variable domain constraints

#### +Vars ins +Domain

The variables in the list Vars are elements of Domain.

#### indomain(?Var)

- Bind Var to all feasible values of its domain on backtracking.
- The domain of Var must be finite.

# Labeling

#### labeling(+Options, +Vars)

- Labeling means systematically trying out values for the finite domain variables Vars until all of them are ground.
- The domain of each variable in Vars must be finite.
- +Options is a list of options that exhibits some control over the search process.
- Several categories of options exist

# Labeling strategy options

**leftmost** - Label the variables in the order they occur in Vars (that is default)

- **ff** first fail. Label the leftmost variable with smallest domain next, in order to detect infeasibility early. This is often a good strategy.
- **ffc** label the variables with smallest domains, the leftmost one participating in <u>most</u> constraints is labeled next.
- min label the leftmost variable next, whose lower bound is the lowest.
- max label the leftmost variable next, whose upper bound is the highest.

The value order is one of:

up - try the elements of the chosen variable's domain in ascending morder. This is default.

down - try the domain elements in descending order.

The branching strategy options:

- **step** for each variable X, a choice is made between X = V and X # V, where V is determined by the value ordering options (default).
- **enum** for each variable X, a choice is made between  $X = V_1$ ,  $X = V_2$  ..., for all values  $V_i$  of the domain of X.

The order is determined by the value ordering options.

**bisect** - for each variable X, a choice is made between X #=< M and X #> M, where M is the midpoint of the domain of X.

At most one option of each category can be specified, and an option must not occur repeatedly.

The order of solutions option:

min(Expr) - generates solutions in ascending order w.r.t. the evaluation of the arithmetic expression Expr

max(Expr) - generates solutions in descending order

- Labeling Vars must make Expr ground.
- If several options are specified, they are interpreted from left to right.

• Example:

```
?-[X,Y] ins 10...20, labeling([max(X),min(Y)],[X,Y]).
```

- This generates solutions of X in descending order,
- but for each binding of X, solutions of Y are generated in ascending order.

# Other labeling options

all\_different(+Vars) -

all variables have pairwise distinct values

sum(+Vars, +Rel, ?Expr) -

The sum of elements of the list Vars is in relation Rel to Expr.

For example:

?-[A,B,C] ins o..sup, sum([A,B,C], #=, 100).

A in o..100,

A+B+C#=100,

B in o..100,

C\_in\_o..100.

### Other labeling options

#### scalar\_product(+Cs, +Vs, +Rel, ?Expr)

- Cs is a list of integer constants,
- Vs is a list of variables and integers.
- True if the scalar product of Cs and Vs is in relation Rel to Expr.
  - Example:
  - Scalar\_product([4,5], [A,B], >, A-B).
  - solves an inequation 4\*A + 5\*B > A-B

### Sudoku

```
sudoku(Rows) :-
  length(Rows, 9), maplist(length_(9), Rows),
  append(Rows, Vs), Vs ins 1..9,
  maplist(all_distinct, Rows),
  transpose(Rows, Columns),
  maplist(all_distinct, Columns),
  Rows = [A,B,C,D,E,F,G,H,I],
  blocks(A, B, C), blocks(D, E, F), blocks(G, H, I).
```

- % maplist(:Goal, ?List) true if Goal can successfully be applied on all elements of the List.
- % maplist(:Goal, ?List1, ?List2) true if Goal can successfully be applied to all succesive pairs of elements of List1 and List2.

```
length_(L, Ls) :-
  length(Ls, L).

blocks([], [], []).
blocks([A,B,C|Bs1], [D,E,F|Bs2], [G,H,I|Bs3]) :-
  all_distinct([A,B,C,D,E,F,G,H,I]),
  blocks(Bs1, Bs2, Bs3).
```

```
problem(1,
[[_,_,_,],
[_,_,,3,_,8,5],
[_,_,1,_,2,_,_,],
[_,_,5,_,7,_,_],
[_,_,4,_,_,1,_,],
[_,9,_,_,_,],
[5,_,_,,,7,3],
[_,_,2,_,1,_,_,_],
[_,_,_,4,_,_,9]]).
```

transpose(+Matrix, ?Transpose).
 Transposes a list of lists of the same length.

• Example:

?- transpose([[1,2,3],[4,5,6],[7,8,9]], Ts). Ts = [[1, 4, 7], [2, 5, 8], [3, 6, 9]]

### Query

?- problem(1, Rows), sudoku(Rows), maplist(writeln, Rows).

```
[9, 8, 7, 6, 5, 4, 3, 2, 1]

[2, 4, 6, 1, 7, 3, 9, 8, 5]

[3, 5, 1, 9, 2, 8, 7, 4, 6]

[1, 2, 8, 5, 3, 7, 6, 9, 4]

[6, 3, 4, 8, 9, 2, 1, 5, 7]

[7, 9, 5, 4, 6, 1, 8, 3, 2]

[5, 1, 9, 2, 8, 6, 4, 7, 3]

[4, 7, 2, 3, 1, 9, 5, 6, 8]

[8, 6, 3, 7, 4, 5, 2, 1, 9]
```

Rows = [[9, 8, 7, 6, 5, 4, 3, 2|...], ..., [...|...]].

# Machine games

- <u>Draughts</u>, <u>English</u> (Checkers) 8×8 variant of <u>draughts</u>
- weakly solved on April 29, 2007 by the team of <u>Jonathan Schaeffer</u>, known for <u>Chinook</u>,
- Checkers is the largest game that has been solved to date, with a search space of 5×10<sup>20</sup>. [7]
- The number of calculations involved was 10<sup>14</sup>, which were done over a period of 18 years. The process involved 50 200 <u>desktop computers</u>