

DEPARTMENT OF INFORMATICS ENGINEERING

Logic and Constraint Programming

Masters in Informatics and Computing Engineering 2022/2023 - 2nd Semester

Constraint Logic Programming using SICStus Prolog

SICStus Prolog User's Manual (Release 4.8)

Section 10.10: Constraint Logic Programming over Finite Domains—library(clpfd)

Content

- 1. Available Constraint Domains
- 2. CLP(FD) Solver Interface
 - Structure of a Program in CLP
 - Domain Declaration
 - Posting Constraints
 - Reified Constraints
- 3. Available Constraints
- 4. Enumeration Predicates
 - Search and Optimization
- 5. Statistics Predicates

CLP in SICStus Prolog

1. AVAILABLE CONSTRAINT DOMAINS

Boolean and Real Domains

- Booleans:
 - clp(B) Scheme
 - use module(library(clpb)).
 - Section 10.9 of SICStus manual
- Reals and Rationals
 - clp(Q,R) Scheme
 - use_module(library(clpq)). use_module(library(clpr)).
 - Section 10.11 of SICStus manual
- Not seen in detail in this curricular unit

Finite Domains

- clp(FD) solver is an instance of the general CLP schema introduced in [Jafar & Michaylov 87].
- Useful to model discrete optimization problems
 - Scheduling, planning, resource allocation, packing, timetabling, ...
- *clp(FD)* solver characteristics:
 - Two constraint classes: primitive and global
 - Very efficient propagators for global constraints
 - The logical value of a primitive constraint can be reflected into a binary variable (0/1) materialization (or reification)
 - New primitive constraints can be added, by writing indexicals
 - New global constraints can be written in Prolog

CLP in SICStus Prolog

2. CLP(FD) SOLVER INTERFACE

CLP(FD) Solver Interface

The clp(FD) solver is available as a library

```
:- use module(library(clpfd)).
```

- It contains predicates to test the consistency and entailment of constraints over finite domains, as well as to obtain solutions by attributing values to the variables
- A finite domain is a subset of small integers and a constraint over finite domains is a relation between a tuple of small integers
- Only small integers and non-instantiated variables are allowed in constraints over finite domains
 - Small integer: [-2²⁸, 2²⁸-1] in 32-bit platforms, or [-2⁶⁰, 2⁶⁰-1] in 64-bit platforms
 - You can use the *prolog_flag/2* predicate to obtain these values

CLP(FD) Solver Interface

- All domain variables have an associated finite domain, explicitly declared in the program or implicitly imposed by the solver
 - Temporarily, a variable's domain can be infinite, if it doesn't have a finite lower or upper bound
 - The domain of a variable reduces as constraints are added
- If a domain becomes empty, the constraints are not satisfiable as a whole, and the current computing branch fails
- At the end of the computation, it is usual for each variable to have its domain constrained to a single value (singleton)
 - Typically, some search is required for this to happen
- Each constraint is implemented by a (set of) propagator(s)
 - Indexicals
 - Global propagators

Structure of a CLP Program

- A CLP program is structured in the following three steps:
 - Declaration of variables and respective domains
 - Declaration of constraints over the variables
 - Search for a solution

```
:- use_module(library(clpfd)).
example:-
    A in 1..7,
    domain([B, C], 1, 10),
    A + B + C #= A * B * C,
    A #> B,
    labeling([], [A, B, C]).
} constraints
search for solution
```

```
| ?- example.
A = 2,
B = 1,
C = 3 ?
```

Structure of a CLP Program

- The order of these steps is important
 - If we invert the order, by placing the search for a solution first and only then the constraints, we end up with the much less efficient traditional Generate & Test mechanism

```
:- use_module(library(clpfd)).
badExample:-
          A in 1..7,
          domain([B, C], 1, 10),
          labeling([], [A, B, C]),
          write('.'),
          A + B + C #= A * B * C,
          A #> B.
```

```
| ?- badExample.

A = 2,
B = 1,
C = 3 ?
```

Variable Domains

- A variable may have its domain declared using in/2 and a range of values, given by a ConstantRange:
 - GradePLR in 16..20

 The definition of ConstantRange allows for the declaration of much more complex domains

```
ConstantSet

ConstantRange

::= {integer,...,integer}

::= ConstantSet

| Constant ... Constant
| Constant ... Constant
| ConstantRange / ConstantRange
| ConstantRange / ConstantRange
| ConstantRange / ConstantRange
| ConstantRange
```

Variable Domains

- You may also use in_set/2 to declare the domain of a variable
 - The second argument of in_set/2 is a Finite Domain Set, which can be obtained from a list by using the list_to_fdset(+List, -FD_Set) predicate
 - See section 10.10.9.3 for operations over *FD Sets*

```
Numbers = [4, 8, 15, 16, 23, 42],
list_to_fdset(Numbers, FDS_Numbers),
Var in_set FDS_Numbers.
```

Variable Domains

- The *domain(+List_of_Variables, +Min, +Max)* predicate can be used to declare a simple domain for a list of variables:
 - domain([A, B, C], 5, 12)

- Other constraints limit the domains of the variables involved
 - A #> 8
 - B + C #< 12
 - A + B + C #= 20

Posting Constraints

A constraint is called just like any other Prolog predicate

```
?- X in 1..5,
                                               ?- X in 1..5,
                      | ?- X in 1...5,
    Y in 1..3,
                        Y in 2..8,
                                               T in 3..13,
    X * Y #= 20.
                         X+Y #= T.
                                                X+Y #= T.
                                             X in 1..5,
                      X in 1..5,
no
                      Y in 2..8,
                                             T in 3..13,
                      T in 3..13
                                             Y in -2..12
```

- The existence of an answer is an indication that, after filtering and propagation, all variables have valid domains (at least 1 possible value)
 - This, by itself, is no guarantee that a solution to the problem exists
- The constraints associated to each variable are not shown

Posting Constraints

- By posting a constraint, the propagation mechanism is called, which limits variable domains
 - This mechanism can be computationally heavy in some cases
- It is possible to post a set of constraints at once (in batch), suspending the propagation mechanism until all these constraints have been placed, using the fd_batch(+Constraints) predicate
 - Constraints is a list with the constraints to be placed

```
| ?- domain([A,B,C], 5, 12),

A #> 8,

B+C #< 12,

A+B+C #= 20.

A in 9..10,

B in 5..6,

C in 5..6 ?
```

Forgetting Constraints

- Variables and associated constraints can also be 'forgotten' using the fd_purge(+Variable) predicate
 - Note that this erases the variable and all associated constraints

Materialized (Reified) Constraints

- Sometimes it is useful to reflect the truth value of a constraint into a boolean variable B (0/1) such that:
 - The constraint is placed if B has value 1
 - The negation of the constraint is placed if B has value 0
 - B is set to 1 if the constraint is entailed
 - B is set to 0 if the constraint is disentailed
- This mechanism is known as materialization or reification
- A materialized constraint is written in the form

Constraint #<=> B.

where Constraint is a reifiable constraint and B a binary variable

Materialized (Reified) Constraints

- Example: exactly(X, L, N)
 - True if X occurs exactly N times in list L
 - It can be recursively defined

```
exactly( , [], 0).
exactly(X, [Y|L], N):-
    X #= Y #<=> B,
    N #= M + B,
    exactly(X, L, M).
```

Reifiable constraints can be used as terms in arithmetic expressions:

```
| ?- x \#= 10,
B = 3,
X = 10
```

```
?- X \#= 10,

B \#= (X\#>=2) + (X\#>=4) + (X\#>=8).

| ?- X in 1..3,

B \#= (X\#>=1) + (X\#>=2) + (X\#>=3),
                                                  labeling([], [X]).
                                             X = 1, B = 1 ? ;
                                             X = 2, B = 2 ?;
                                             X = 3, B = 3?;
                                             no
```

CLP in SICStus Prolog

3. AVAILABLE CONSTRAINTS

Available Constraints

- Arithmetic Constraints
- Membership Constraints
- Propositional Constraints
- Combinatorial Constraints
 - Arithmetic-Logical
 - Scheduling
 - Placement
 - Graph
 - Sequence
 - Extensional

Arithmetic Constraints

Expr RelOp Expr

```
- RelOp: #= | #\= | #< | #=< | #> | #>=
```

- The expressions can be linear or non-linear
- Linear expressions lead to a better propagation
 - For instance, X/Y and X mod Y block until Y is ground
- Linear arithmetic constraints maintain interval consistency
- Arithmetic constraints can be materialized
- Example:

Sum

sum(Xs, RelOp, Value)

- Xs is a list of integers and/or domain variables, RelOp is a relational operator and Value is an integer or domain variable
- The constraint holds if sum(Xs) RelOp Value (the sum of all elements in Xs has relation RelOp to Value)
- It approximates sumlist/2 from the lists library (when RelOp is #=)
- Cannot be materialized
- Examples:

```
| ?- domain([X,Y], 1, 10),

sum([X,Y], #<, 10).

X in 1..8,

Y in 1..8

| ?- domain([X,Y], 1, 10),

sum([X,Y], #=, Z).

X in 1..10,

Y in 1..10,

Z in 2..20
```

Scalar Product

- scalar_product(Coeffs, Xs, RelOp, Value)
- scalar_product(Coeffs, Xs, RelOp, Value, Options)
 - Coeffs is a list of length n of integers, Xs a list of length n of integers and/or domain variables, RelOp a relational operator and Value an integer or domain variable
 - The constraint holds if sum(Coeffs*Xs) RelOp Value
 - This constraint is materializable
 - Options is a list of options
 - among(Least, Most, Range) imposes that at least Least and at most Most elements from Xs must have values within Range (a ConstantRange)
 - consistency(Cons) denotes the level of consistency to be used by the constraint. Cons can take the
 values domain, bounds or value (by default, bounds consistency is maintained).

Scalar Product

- scalar_product_reif(Coeffs, Xs, RelOp, Value, Reif)
- scalar_product_reif(Coeffs, Xs, RelOp, Value, Reif, Options)
 - Reified version of scalar_product/[4,5].
 - Equivalent to materializing the previous constraint
 - Examples:

Minimum / Maximum

- minimum(Value, Xs)
- maximum(Value, Xs)
 - Xs is a list of integers and/or domain variables
 - Value is an integer or domain variable
 - The constraint holds if Value is the minimum (or maximum) value of Xs
 - Corresponds to min_member/2 (max_member/2) from library(lists)
 - Cannot be materialized
 - Examples:

Minimum_arg / Maximum_arg

- minimum_arg(Xs, Index)
- maximum_arg(Xs, Index)
 - Xs is a list of integers and/or domain variables
 - Index is an integer or domain variable
 - The constraint holds if *Index* is the index of the minimum (maximum) value of *Xs*
 - If the value appears more than once in Xs, Index points to the first occurrence
 - Cannot be materialized
 - Examples:

```
| ?- minimum_arg([3,1,2,5], A).

A = 2

| ?- maximum_arg([3,1,2,5], B).

B = 4
```

If Then Else

- if_then_else(If, Then, Else, Value)
 - If is an integer or domain variable (can only take values 0 or 1)
 - Then, Else, and Value are integers or domain variables
 - The constraint holds when $\mathbf{If} = 1$ and $\mathbf{Value} = \mathbf{Then}$, or $\mathbf{If} = 0$ and $\mathbf{Value} = \mathbf{Else}$
 - Corresponds to a typical if-then-else construct
 - Example:

```
| ?- domain([A,B,C], 1, 2),
    A #\= B,
    A #> B #<=> D,
    if_then_else(D, A, B, C),
    labeling([], [A,B,C]).

A = 1,
B = 2,
C = 2,
D = 0 ?;
```

Membership Constraints

- Predicates used to define variable domains
 - domain(+Vars, +Min, +Max)
 - True if all elements in *Vars* are in the interval *Min..Max*
 - ?X in +Range
 - True if X is an element of the ConstantRange Range
 - ?X in_set +FDSet
 - True if X is an element of the set FDSet
 - in/2 and in_set/2 maintain domain consistency and are materializable
 - Examples:

Propositional Constraints

- Allow for the definition of propositional formulas over materializable constraints
- Example: x #= 2 #\/ y #= 4
 - Expresses the disjunction between two equality constraints
- The leafs of the propositional formulas can be materializable constraints, constants 0 and 1, or boolean variables (domain 0/1)
- These constraints maintain domain consistency (even though domains are not affected by posting propositional constraints)
- New primitive materializable constraints can be defined using indexicals
- Example:

Propositional Constraints

#\ :Q	true if constraint Q is false (NOT)
:P #/\ :Q	true if constraints P and Q are both true (AND)
:P #\ :Q	true if exactly one of the constraints P and Q is true (XOR)
:P # \ ∕ :Q	true if at least one of the constraints P and Q is true (OR)
:P #=> :Q :Q #<= :P	true if constraint Q is true or if constraint P is false (implication)
:P #<=> :Q	true if P and Q are both true or both false (equivalence)

Note that materialization is a particular case of the equivalence propositional constraint (when Q is a variable with domain 0/1)

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Combinatorial Constraints

- Combinatorial Constraints are also called symbolic constraints
- Usually they maintain interval consistency on the variables involved

Arithmetic-Logical

- smt/1 (deprecated, see case)
- count/4 (deprecated)
- global_cardinality/[2,3]
- nvalue/2
- all_equal/1, all_equal_reif/2
- all_different/[1,2]
- all_distinct/[1,2]
- all_different_except_0/1
- all_distinct_except_0/1
- symmetric_all_different/1
- symmetric_all_distinct/1
- assignment/[2,3]
- sorting/3
- keysorting/[2,3]

- lex_chain/[1,2]
- bool_[and,or,xor]/2
- bool_channel/4

Scheduling

- cumulative/[1,2]
- cumulatives/[2,3]
- multi_cumulative/[2,3]

Placement

- bin_packing/2
- disjoint1/[1,2]
- disjoint2/[1,2]
- diffn/[1,2]
- geost/[2,3,4]

Graph

- *circuit/[1,2]*
- subcircuit/[1,2]

Sequence

- automaton/[3,8,9]
- regular/2
- value_precede_chain/[2,3]
- seq_precede_chain/[1,2]

Extensional

- element/[2,3]
- relation/3 (deprecated)
- table/[2,3]
- case/[3,4]

Count

(deprecated, see global_cardinality)

count(+Val, +List, +RelOp, ?Count)

- Constrains the number of occurrences of a given value within a list
- Val is an integer, List a list of integers and/or domain variables, Count an integer or domain variable, and RelOp a relational operator
- The constraint holds if the number of occurrences of *Val* within *List* has relation *RelOp* with *Count*
- Maintains domain consistency, but global_cardinality/2 is a better alternative
- Examples:

Global Cardinality

- global_cardinality(+Xs, +Vals)
- global_cardinality(+Xs, +Vals, +Options)
 - Constrains the number of occurrences of each value within a list of variables
 - Xs is a list of integers and/or domain variables; Vals is a list of K-V terms, where K is a unique integer and V is an integer or a domain variable
 - True if each element in Xs equals one of K and for each pair K-V exactly V elements of Xs equal K
 - If Xs or Vals are ground, and in other special cases, it maintains domain consistency; interval
 consistency may not be assured
 - Options is a list of options (to control propagation) (see documentation)

Nvalue

nvalue(?N, +Variables)

- Constrains the list of variables Variables in such a way that there are exactly N
 distinct values
- Variables is a list of integers and/or domain variables with finite limits and N is an integer or domain variable
- Can be seen as a relaxed version of all_distinct/2
- Examples:

All Equal

- all_equal(+Variables)
- all_equal_reif(+Variables, B)
 - Variables is list of integers and/or domain variables, B a boolean variable/integer
 - The constrain holds if all values in the Variables list are the same
 - Equivalent to a #= constraint for each pair of variables
 - This constraint is materializable, and has a reified version
 - Examples:

All Different / All Distinct

- all_different(+Variables) / all_different(+Variables, +Options)
- all_distinct(+Variables) / all_distinct(+Variables, +Options)
 - True when all values in the *Variables* list are distinct
 - Equivalent to a #\= constraint for each pair of variables
 - Variables is a list of integers and/or domain variables
 - Options is a list with zero or more options (see documentation), controlling propagation, or adding additional side constraints

All Different / Distinct Except 0

- all_different_except_0(+Variables)
- all_distinct_except_0(+Variables)
 - Variables is a list of integers and/or domain variables
 - The constraint holds if the Variables list contains distinct values, with the exception of variables with the value 0
 - Examples:

Symmetric All Different / Distinct

- symmetric_all_different (+Variables)
- symmetric_all_distinct (+Variables)
 - Variables is a list of integers and/or domain variables
 - The constraint holds if all variables in the Variables list have distinct values, and for all variables Xi=j iff Xj=i
 - Examples:

Assignment

- assignment(+Xs, +Ys)
- assignment(+Xs, +Ys, +Options)
 - Xs = [X1,...,Xn] and Ys = [Y1,...,Yn] are lists of length n of integers and/or domain variables
 - True if all Xi, Yi are in [1,n], are unique within their list and Xi=j iff Yj=i (the lists are dual)
 - Options is a list that may contain, among others (see documentation), the options:
 - circuit(Bool), subcircuit(Bool): if true, then circuit(Xs, Ys) / subcircuit(Xs, Ys) must hold

– Examples:

```
| ?- assignment([4,1,5,2,3], Ys).

Ys = [2,4,5,1,3]
```

Sorting

sorting(+Xs, +Ps, +Ys)

- Captures the relation between a list of values, a list of values ordered increasingly and the positions of the values in the original list
- Xs, Ps and Ys are lists of equal length n of integers and/or domain variables
- The constraint holds if:
 - Ys is ordered increasingly; Ps is a permutation of [1..n]; For each i in [1..n], Xs[i] = Ys[Ps[i]]
- Examples:

```
| ?- length(Ys, 5), length(Ps, 5), | ?- length(Ys, 5), sorting([2,7,9,1,3], Ps, Ys). | Sorting([2,7,9,1,3], Ps = [1,2,3,7,9], Ps = [2,4,5,1,3] ? | Ps = [2,5,_A _ A in 3..4,
```

Keysorting

- keysorting(+Xs, +Ys)
- *keysorting(+Xs, +Ys, +Options)*
 - Generalization of sorting/3 but ordering tuples of variables
 - Tuples are separated into key and value, being ordered only by the key (maintains the order of tuples with the same key)
 - Xs and Ys are lists of the same size n of tuples of variables; all tuples (lists of variables) have the same size m
 - Options is a list of options:
 - keys(Keys) Keys is the size of the key (positive integer; the default value is 1)
 - permutation(Ps) Ps is a list of variables (permutation of [1..n], such that for each i in [1..n], j in [1..m] : Ys[i,j] = Xs[Ps[i],j].) | ?- Lst = [[1,5],[6,5],[4,3],[7,9],[4,5],[7,8],[3,3]],
 - Example:

```
length( Lst, Len), length(Srt, Len), maplist(ln2, Srt),
                         length(P, Len), keysorting(_Lst, Srt, [permutation(P)]).
ln2(X):-
                    Srt = [[1,5],[3,3],[4,3],[4,5],[6,5],[7,9],[7,8]]
    length (X, 2).
                    P = [1,7,3,5,2,4,6] ? ;
```

Lex Chain

- lex_chain(+Vectors)
- lex_chain(+Vectors, +Options)
 - Vectors is a list of vectors (lists) of integers and/or domain variables
 - The constraint holds if *Vectors* is in an ascending lexicographic order (actually, non-decreasing by default)
 - Options is a list of options:
 - op(Op) Op is #=< (default) or #< (strictly ascending)
 - *increasing* internal lists ordered in a strictly ascending manner
 - among(Least, Most, Values) between Least and Most values of each Vector belong to the Values list
 - Example:

Bool And / Or / Xor

- bool_and(+List, +Lit)
- bool_or(+List, +Lit)
- bool_xor(+List, +Lit)
 - List is a list of literals, and Lit is a literal (here, a literal is a Boolean variable or its negation)
 - The constraint holds if the conjunction / disjunction / exclusive or of the values in List equals Lit
 - It is usually more efficient than using the corresponding propositional constraints
 - Examples:

```
| ?- A in 1..2, B in 1..2,
        A #> 1 #<=> S, B #> 1 #<=> T,
        bool_xor([S, T], 1),
        labeling([], [A,B]).

A = 1, B = 2, S = 0, T = 1 ?;
A = 2, B = 1, S = 1, T = 0 ?;
no
```

```
| ?- A in 1..2, B in 1..2,
        A #> 1 #<=> S, B #> 1 #<=> T,
        bool_and([S, T], T),
        labeling([], [A,B]).

A = 1, B = 1, S = 0, T = 0 ?;
A = 2, B = 1, S = 1, T = 0 ?;
A = 2, B = 2, S = 1, T = 1 ?;
no
```

Bool Channel

- bool_channel(+List, ?Y, +RelOp, +Offset)
 - List is a list of literals, Y is a domain variable, RelOp a relational operator, and Offset an integer
 - The constraint holds if Li #<=> (Y RelOp i+Offset) for all Li in List
 - It is usually more efficient than using several reifications
 - Examples:

```
| ?- length(L, 5),
        bool_channel(L, 3, #=, 1).
L = [0,0,1,0,0] ?;
no
| ?- bool_channel([0,0,0,1,0], X, #=, 1).
X = 4 ?;
no
```

Cumulative

- cumulative(+Tasks)
- cumulative(+Tasks, +Options)
 - Constrains n task to be scheduled in such a way that the instantaneous resource consumption never exceeds a given limit at any time
 - Tasks is a list of n terms in the form task(Oi, Di, Ei, Hi, Ti)
 - Oi = start time, Di = duration (non-negative), Ei = end time, Hi = resource consumption (non-negative),
 Ti = task identified (all fields can be integers or domain variables)
 - The constraint holds if, for all task i, Oi+Di=Ei and at any given time H1+H2+...+Hn =< L (resource limit, 1 by default)
 - Hi is only accounted for during the instants between Oi and Ei; otherwise it's 0
 - Options is a list of options:
 - *limit(L)*: L is the resource limit to use
 - precedences(Ps): precedence between tasks; Ps is a list of terms in the form Ti-Tj #= Dij, with Oi-Oj = Dij
 - global(Boolean): if true, uses a more costly algorithm to obtain better interval pruning

Cumulative

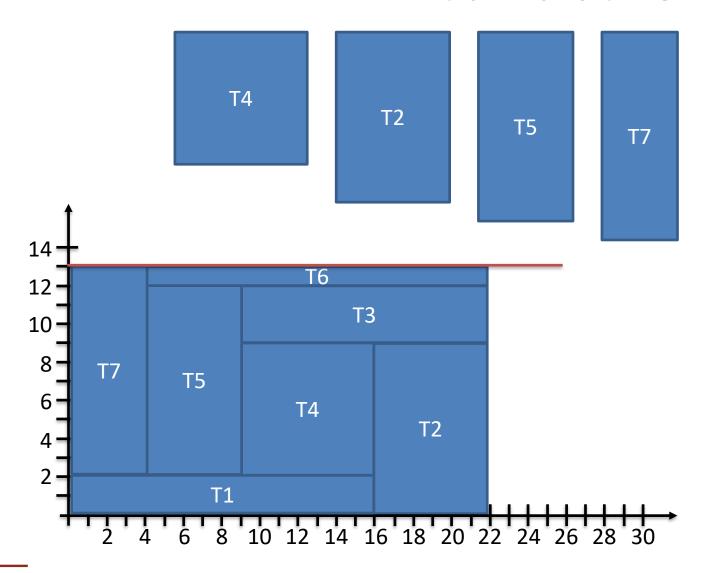
- Example:
 - Task scheduling:

Task	Duration	Resources
T1	16	2
T2	6	9
T3	13	3
T4	7	7
T5	5	10
T6	18	1
T7	4	11

• Resource limit = 13

```
schedule(Ss, End) :-
   Ss = [S1, S2, S3, S4, S5, S6, S7],
   Es = [E1, E2, E3, E4, E5, E6, E7],
   Tasks = [
        task(S1, 16, E1, 2, 1),
        task(S2, 6, E2, 9, 2),
        task(S3, 13, E3, 3, 3),
        task(S4, 7, E4, 7, 4),
        task(S5, 5, E5, 10, 5),
        task (S6, 18, E6, 1, 6),
        task(S7, 4, E7, 11, 7) ],
   domain(Ss, 1, 30),
   maximum(End, Es),
   cumulative(Tasks, [limit(13)]),
   labeling([minimize(End)], Ss).
```

Cumulative



Duration	Resources
16	2
6	9
13	3
7	7
5	10
18	1
4	11
	16 6 13 7 5 18

Resource limit = 13

T6
T1
T3

Cumulatives

- cumulatives(+Tasks, +Machines)
- cumulatives(+Tasks, +Machines, +Options)
 - Constrains n tasks to be scheduled in m machines, where each machine has a (minimum or maximum)
 resource limit
 - Tasks is a list of terms in the form task(Oi, Di, Ei, Hi, Mi)
 - **Oi** = start time, **Di** = duration (non-negative), **Ei** = end time, **Hi** = resource consumption (if positive) or resource production (if negative), **Mi** = machine identifier (all fields can be integers or domain variables)
 - Machines is a list of terms in the form machine(Mj, Lj)
 - Mj = machine identifier (unique integer), Lj = machine resource limit (integer or domain variable with defined limits)
 - The constraint holds if for all tasks Oi+Di=Ei and in all machines and at all times H1m+H2m+...+Hnm >= Lm (if lower bound), or H1m+H2m+...+Hnm =< Lm (if upper bound)</p>
 - Options is a list of options:
 - **bound(B)** type of limit: **lower** (default) or **upper**
 - prune(P) all (default value) or next: points to the pruning level to use
 - generalization(Boolean), task_intervals(Boolean) if true some extra processing is performed

Cumulatives

• Example:

– Task scheduling:

Task	Duration	Resources	Machine
T1	16	2	1
T2	6	9	2
Т3	13	3	1
T4	7	7	2
T5	5	10	1
T6	18	1	2
T7	4	11	1

- Resource limit for M1 = 12
- Resource limit for M2 = 10

```
schedule(Ss, End) :-
   Ss = [S1, S2, S3, S4, S5, S6, S7],
   Es = [E1, E2, E3, E4, E5, E6, E7],
    Tasks = [
        task(S1, 16, E1, 2, 1),
        task(S2, 6, E2, 9, 2),
        task(S3, 13, E3, 3, 1),
        task(S4, 7, E4, 7, 2),
        task(S5, 5, E5, 10, 1),
        task(S6, 18, E6, 1, 2),
        task(S7, 4, E7, 11, 1) ],
   Machines = [machine(1, 12),
        machine (2, 10)],
    domain(Ss, 1, 30),
    maximum (End, Es),
    cumulatives (Tasks, Machines,
        [bound(upper)]),
   labeling([minimize(End)], Ss).
```

Multi_Cumulative

- multi_cumulative(+Tasks, +Capacities)
- multi_cumulative(+Tasks, +Capacities, +Options)
 - Generalization of the *cumulative* constraint, allowing tasks to consume multiple resources simultaneously;
 these can be of two types:
 - cumulative resources as used in the cumulative / cumulatives constraints
 - colored each task specifies a color (coded as an integer); the number of colors in use at each moment must not exceed a given limit; the color 0 means that the task does not use any color
 - Tasks is a list of terms in the form task(Oi, Di, Ei, Hsi, Ti)
 - Oi = start time, Di = duration (non-negative), Ei = end time, Hsi = list of resource consumption/used color, Ti = task id
 - Oi and Ei are domain variables; the remaining fields must be integers
 - Capacities is a list of terms in the format cumulative(Limit) or colored(Limit)
 - The size of the *Capacities* list must be equal to the size of all *Hsi* lists
 - The constraint holds if no resource exceeds its limit at any given time
 - Options is a list of options:
 - greedy(Flag): Flag is a variable with domain 0..1 denoting whether the greedy mode should be used
 - precedences(Ps): task precedence; Ps is a list of terms in the form Ti-Tj (Ti and Tj are task identifiers) denoting that Ti must finish before Tj starts

Bin Packing

- bin_packing(+Items, +Bins)
 - Assigns 'items' of certain sizes to 'containers' with given capacities
 - Items is a list of terms in the format item(Bin, Size)
 - *Bin* is the container to which the item is assigned (integer or domain variable); *Size* is the item size (integer >=0)
 - Bins is a list of terms in the format bin(ID, Cap)
 - *ID* is the identifier of each container (integer, all distinct); *Cap* is the container capacity (integer or domain variable)
 - The constraint holds if all items are assigned to an existing container and the sum of the sizes of all items assigned to each container equals its capacity
 - It may be required to use either 'ghost objects' or bins with variable capacities

Bin Packing

• Example:

6 objects, 3 compartments

Item	Size
А	5
В	6
С	3
D	7
E	9
F	4

Bin	Сар
1	9
2	14
3	11

```
place(Vars) :-
    Vars = [A, B, C, D, E, F],
    Items = [ item(A, 5), item(B, 6),
        item(C, 3), item(D, 7),
        item(E, 9), item(F, 4) ],
    Bins = [ bin(1, 9),
        bin(2, 14),
        bin(3, 11) ],
    bin_packing(Items, Bins),
    labeling([], Vars).
```

```
| ?- place(Vars).
Vars = [2,1,1,3,2,3] ?;
Vars = [2,2,2,3,1,3] ?;
Vars = [3,3,2,2,1,2] ?;
no
```

- disjoint1(+Lines)
- disjoint1(+Lines, +Options)
 - Constrains a set of lines in such a way that they do not overlap
 - 1D view of space (all lines are aligned)
 - Lines is a list of terms in the format F(Sj,Dj) or F(Sj, Dj, Tj)
 - Sj and Dj represent origin and size of line j (integer or domain variable); F is any functor
 - Tj is an optional atomic term (0 by default) denoting the type of line
 - Options is a list of options
 - global(Boolean) if true a redundant algorithm is used to attain a more complete pruning
 - wrap(Min, Max) space is seen as a circle, where values Min and Max (integers) coincide; this option forces values back to interval [Min, Max-1]
 - margin(T1, T2, D) imposes a minimum distance D between the end of any line of type T1 and the beginning of any line of type T2; D must be a positive integer or sup: all lines of type T2 end before any line of type T1

– Example:

```
place(Starts) :-
    Starts = [A, B, C],
    domain(Starts, 1, 10),
    Lines = [
        line(A, 5),
        line(B, 7),
        line(C, 3)
    ],
    A #< C,
    disjoint1(Lines),
    labeling([], Starts).</pre>
```

```
Starts = [1,9,6] ?;
Starts = [1,10,6] ?;
Starts = [1,10,7] ?;
Starts = [2,10,7] ?;
no
```

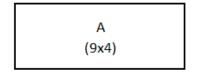
```
place(Starts) :-
    Starts = [A, B, C, D],
    domain (Starts, 1, 12),
    Lines = [
        line (A, 4, r),
        line(B, 2, g),
        line(C, 3, r),
        line(D, 2, q)
    A #< B,
    disjoint1(Lines, [margin(r, g, 3)]),
    labeling([], Starts).
```

```
Starts = [1,8,12,10] ?;
Starts = [1,10,12,8] ?;
Starts = [3,10,12,1] ?;
no
```

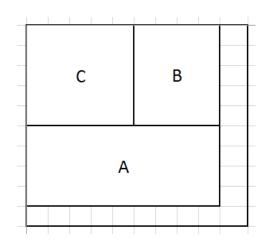
- disjoint2(+Rectangles)
- disjoint2(+Rectangles, +Options)
 - Constrains a set of rectangles in such a way that they do not overlap
 - Rectangles is a list of terms in the format F(Xj, Lj, Yj, Hj) or F(Xj, Lj, Yj, Hj, Tj)
 - **Xj** and **Yj** represent the origin of rectangle j, while **Lj** and **Hj** represent its dimensions (integers or domain variables); **F** is any functor; **Tj** is an atomic term (0 by default) denoting the type of rectangle
 - Options is a list of options
 - wrap(Min1, Max1, Min2, Max2) Min1 and Max1 refer to the X dimension, while Min2 and Max2 refer to the Y dimension; if all values are integers, the space is seen as a toroid; the values inf and sup can be used (for Min and Max in one of the dimensions) to obtain a cylindrical space
 - margin(T1, T2, D1, D2) imposes a minimum distance D1 in X and D2 in Y between the end of any rectangle of type T1 and the start of any rectangle of type T2; D1 and D2 must be positive integers or sup: all rectangles of type T2 end before any rectangle of type T1 in the relevant dimension

Example:

Position three rectangles in a 10x10 grid







```
place(StartsX, StartsY) :-
    StartsX = [Ax, Bx, Cx],
    StartsY = [Ay, By, Cy],
    domain(StartsX, 1, 10),
    domain(StartsY, 1, 10),
    Rectangles = [
        rect(Ax, 9, Ay, 4),
        rect(Bx, 4, By, 5),
        rect(Cx, 5, Cy, 5)],
    Ax + 9 \# = < 10, Ay + 4 \# = < 10,
    Bx + 4 \# = < 10, By + 5 \# = < 10,
    Cx + 5 \# = < 10, Cy + 5 \# = < 10,
    disjoint2 (Rectangles),
    append(StartsX, StartsY, Vars),
    labeling([], Vars).
```

```
StartsX = [1, 6, 1],

StartsY = [6, 1, 1]?
```

Diffn

- diffn(+Boxes)
- diffn(+Boxes, +Options)
 - Constrains the location in space of multidimensional boxes (Boxes) to not overlap
 - diffn/[1,2] should be used instead of disjoint1/[1,2] and disjoint2/[1,2]
 - Boxes is a list of boxes, each represented by a list of terms in the format Origin-Length
 - Origin and Length represent the origin and size of the box in each dimension
 - All boxes must have the same dimensionality (i.e., the lists must have the same size)
 - Options is a list of options
 - **strict(Boolean)** if **false**, the disjunction admits boxes with no length in some dimension(s); if **true**, the disjunction is more strict

```
| ?- diffn([ [1-3, 1-3], [2-3, 4-3] ]).

yes
| ?- diffn([ [1-3, 1-0], [2-3, 0-3] ],

[strict(false)]).

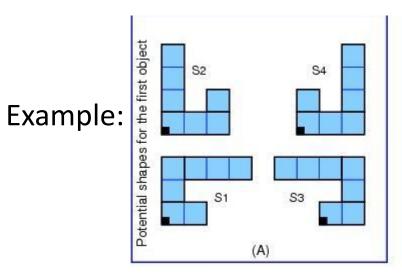
yes
| ?- diffn([ [1-3, 1-0], [2-3, 0-3] ],

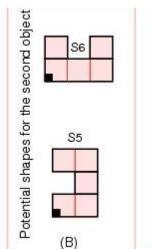
[strict(true)]).

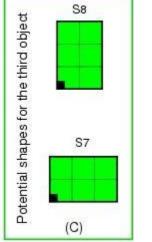
no
```

Geost

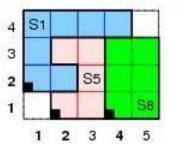
- geost(+Objects, +Shapes)
- geost(+Objects, +Shapes, +Options)
- geost(+Objects, +Shapes, +Options, +Rules)
 - Constrains the location in space of multidimensional objects (*Objects*) to not overlap, each object having a shape from a set of existing shapes (*Shapes*)
 - Objects is a list of terms in the format object(Oid, Sid, Origin)
 - *Oid* identifies the object (unique integer); *Sid* identifies the shape of the object (integer or domain variable); *Origin* denotes the coordinates of the origin of the object (list of integers and/or domain variables)
 - Shapes is a list of terms in the format sbox(Sid, Offset, Size), representing shifted boxes
 - **Sid** is the identifier of the shape (integer); **Offset** is a list of integers of size n with the displacement in each dimension of the box, relative to the origin of the object; **Size** is a list of integers of size n with the size of the box in each dimension
 - Each shape is defined by the set of **sbox/3** terms with the same **Sid**
 - Options is a list of options (see documentation)







Geost



A possible placement where object 1 is assigned shape S1 and object 2 is assigned shape S5 and object 3 is assigned shape S8

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(D)

Circuit

- circuit(+Succ)
- circuit(+Succ, +Pred)
 - Succ is a list of length n of integers and/or domain variables
 - The ith element of *Succ* (*Pred*) is the successor (predecessor) of i in the graph
 - The constraint holds if the values form a Hamiltonian circuit
 - Nodes are numbered from 1 to n, the circuit starts in node 1, visits each node and returns to the origin
 - Examples:

Subcircuit

- subcircuit(+Succ)
- subcircuit(+Succ, +Pred)
 - Succ is a list of length n of integers and/or domain variables
 - The ith element of *Succ* (*Pred*) is the successor (predecessor) of i in the graph; or i if the element is not included in the sub-circuit
 - The constraint holds if the values included form at most one Hamiltonian circuit
 - Examples:

Value Precede Chain

- value_precede_chain(+Values, +Vars)
- value_precede_chain(+Values, +Vars, +Options)
 - Provides a way of removing value symmetries
 - Values is a list of integers and Vars is a list of integers and/or domain variables
 - The constraint holds if for each pair of adjacent values X, Y in Values, Y does not exist in Vars, or, if Y exists in Vars, X appears before Y
 - Options is a list of options:
 - **global(Bool)**: if **false** (default value) a decomposition of the constraint into **automaton/3** is performed. If **true**, a custom algorithm is used. Both maintain domain consistency, but the relative performance may vary
 - Examples:

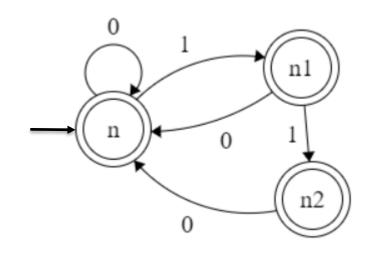
Sequence Precede Chain

- seq_precede_chain(+Vars)
- seq_precede_chain(+Vars, +Options)
 - Similar to the previous constraint, assuming *Values* = [1, 2, 3, ...]

- automaton(Signature, SourcesSinks, Arcs)
- automaton(Sequence, Template, Signature, SourcesSinks, Arcs, Counters, Initial, Final)
- automaton(Sequence, Template, Signature, SourcesSinks, Arcs, Counters, Initial, Final, Options)
 - General method of defining any constraint involving sequences that can be verified by a finite automaton, deterministic or not, extended with possible counting operations in the edges
 - If counters are not used, it maintains domain consistency
 - Signature is a sequence of integers and/or domain variables, based on which the transitions in the automaton will be performed
 - SourcesSinks is a list of elements in the form source(node) or sink(node), identifying the initial and acceptance nodes of the automaton, respectively
 - Arcs is a list of elements in the form arc(node, integer, node) or arc(node, integer, node, exprs), identifying
 the possible transitions between nodes and possible operations over variables in Counters
 - Counters, Initial and Final are lists of equal size identifying counter variables, their initial values (usually instantiated) and final values (usually non-instantiated), respectively
 - Options is a list of options (see documentation)

• Example:

```
at most two consecutive ones (Vars) :-
    automaton (Vars,
        [source(n), sink(n), sink(n1), sink(n2)],
        [arc(n, 0, n), arc(n, 1, n1),
         arc(n1, 1, n2), arc(n1, 0, n),
         %arc(n2, 1, false),
         arc(n2, 0, n) ]).
\mid ?- at most two consecutive ones([0,0,0,1,1,1]).
no
\mid ?- at most two consecutive ones([0,1,1,0,1,1]).
yes
\mid ?- at most two consecutive ones([0,1,1,0,1,0]).
yes
```

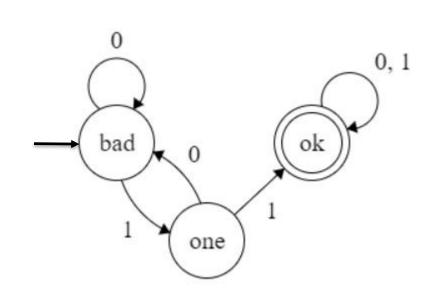


```
| ?- length(L,3),
    at_most_two_consecutive_ones(L).
L = [_A,_B,_C],
    A in 0..1, _B in 0..1, _C in 0..1

| ?- length(L,3),
    at_most_two_consecutive_ones(L),
    L = [1|_], labeling([], L).
L = [1,0,0] ?;
L = [1,0,1] ?;
L = [1,1,0] ?;
```

• Example:

```
at_least_two_consecutive_ones(Vars, N) :-
   length(Vars, N),
   %domain(Vars, 0, 1),
   automaton(Vars,
       [source(bad), sink(ok)],
       [arc(bad, 0, bad), arc(bad, 1, one),
       arc(one, 0, bad), arc(one, 1, ok),
       arc(ok, 0, ok), arc(ok, 1, ok)]),
   labeling([], Vars).
```



```
| ?- at_least_two_consecutive_ones(L, 3).

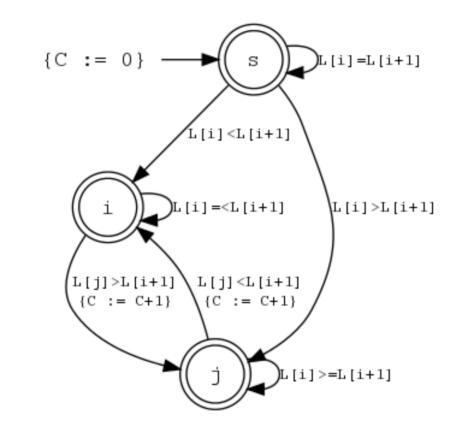
L = [0,1,1] ?;

L = [1,1,0] ?;

L = [1,1,1] ?;

no
```

```
inflexion(N, Vars) :-
    inflexion signature (Vars, Sign),
    automaton(Sign, , Sign,
        [source(s), sink(i), sink(j), sink(s)],
        [arc(s,1,s), arc(s,2,i), arc(s,0,j),
         arc(i,1,i), arc(i,2,i), arc(i,0,j,[C+1]),
         arc(j,1,j), arc(j,0,j), arc(j,2,i,[C+1])],
        [C], [O], [N]).
inflexion signature([], []).
inflexion signature([ ], []) :- !.
inflexion signature([X, Y | Ys], [S|Ss]) :-
   S in 0...2,
  X \#> Y \#<=> S \#= 0,
  X #= Y #<=> S #= 1,
  X \# < Y \# <=> S \#= 2,
   inflexion signature([Y|Ys], Ss).
```



```
| ?- inflexion(N, [1,1,4,8,8,2,7,1]).
N = 3

| ?- length(L,4), domain(L,0,1),
        inflexion(2,L), labeling([],L).
L = [0,1,0,1] ?;
L = [1,0,1,0] ?;
no
```

Regular

regular(Signature, RegExpr)

- Alternative (and more compact) manner of defining an automaton, by using a regular expression
- Signature is a sequence of integers and/or domain variables to validate against the regular expression
- RegExpr is a ground Prolog term representing the regular expression (see documentation)
- The constraint holds if Signature matches RegExpr

```
| ?- regular([3,2,5],
+{1,2}+(+{3,4})+(+{5})).
no

| ?- regular([1,3,5],
+{1,2}+(+{3,4})+(+{5})).
yes
```

Element

- element(?X, +List, ?Y)
- element(+List, ?Y)
 - X and Y are integers and/or domain variables; List is a list of integers and/or domain variables
 - True if the Xth element of List is Y / true if Y exists in List
 - Operationally, the domains of X and Y are constrained in such a way that, for each element in the domain of X, there is a compatible element in the domain of Y, and vice-versa
 - Maintains domain consistency on X and interval consistency in List and Y
 - Corresponds to nth1/3 from library(lists).

Relation

(*deprecated*, see *table*)

relation(?X, +MapList, ?Y)

- X and Y are integers or domain variables and MapList is a list of pairs Integer-ConstantRange,
 where each integer key occurs only once
- The constraint holds if MapList contains a pair X-R and Y is in the interval stated by R
- Examples:

Table

- table(+Tuples, +Extension)
- table(+Tuples, +Extension, +Options)
 - Defines an n-ary constraint by extension
 - Tuples is a list of lists of integers and/or domain variables, each of length n; Extension is a list of lists of integers, each of length n; Options is a list of options that allow controlling the order of the variables used internally and the data structure and algorithm (see documentation)
 - The constraint holds if each *Tuple* in *Tuples* occurs in *Extension*
 - Examples:

```
| ?- table([[A,B]],[[1,1],[1,2],[2,10],[2,20]]).
A in 1..2,
B in (1..2)\/{10}\/{20}

| ?- table([[A,B],[B,C]],[[1,1],[1,2],[2,10],[2,20]]).
A = 1,
B in 1..2,
C in (1..2)\/{10}\/{20}
```

Case

- case(+Template, +Tuples, +Dag)
- case(+Template, +Tuples, +Dag, +Options)
 - Codes an n-ary constraint, defined by extension and/or linear inequalities
 - Uses a DAG: the nodes correspond to variables, each arc is labeled by an admissible interval to the variable in the node of origin, or by linear inequalities
 - Variable order is fixed: each path from the root to a leaf must visit each variable once, in the order by which they appear in *Template*
 - Template is an arbitrary non-ground term
 - Tuples is a list of terms in the same form as in Template (they should not share variables)
 - Dag is a list of terms in the form node(ID, X, Children), where X is a variable from template and ID an integer identifying the node; the first node in the list is the root
 - Internal node: *Children* is a lista of terms (*Min..Max*)-*ID2* (or (*Min..Max*)-*SideConstraints-ID2*), where *ID2* identifies a child node
 - Leaf node: *Children* is a list of terms *(Min..Max)* (or *(Min..Max)-SideConstraints)*

Case

– Example:

```
element(X, [1,1,1,1,2,2,2,2], Y),
                                                                C = 20
element(X, [10,10,20,20,10,10,30,30], Z)
   elts(X, Y, Z) :=
       case(f(A,B,C), [f(X,Y,Z)],
             [node(0, A, [(1..2)-1, (3..4)-2, (5..6)-3, (7..8)-4]),
             node(1, B, [(1..1)-5]),
             node(2, B, [(1..1)-6]),
              node(3, B, [(2...2)-5]),
              node(4, B, [(2...2)-7]),
              node(5, C, [(10...10)]),
              node(6, C, [(20...20)]),
              node(7, C, [(30..30)]))
```

```
3..4 1..2 5..6 7..8

B B B B

1..1 1..1 2..2 2..2

C=20 C=10 C=30
```

```
| ?- elts(X, Y, Z).
X in 1..8,
Y in 1..2,
Z in {10}\/{20}\/{30}

| ?- elts(X, Y, Z), Z #>= 15.
X in(3..4)\/(7..8),
Y in 1..2,
Z in {20}\/{30}

| ?- elts(X, Y, Z), Y = 1.
Y = 1,
X in 1..4,
Z in {10}\/{20}
```

CLP in SICStus Prolog

4. ENUMERATION PREDICATES

Search

- Usually constraint solvers for finite domains are not complete, i.e., they
 do not guarantee that the set of constraints has a solution
- Search (enumeration) is necessary to verify the satisfiability and obtain concrete solutions
- Predicates to perform search:
 - indomain(?X)
 - X is an integer or domain variable
 - Assigns, via backtracking, admissible values to X, in increasing order
 - labeling(:Options, +Variables)
 - solve(:Options, :Searches)

Search

- labeling(:Options, +Variables)
 - Options is a list of search options
 - Variables is a list of integers and/or domain variables
 - The predicate succeeds if it can find [at least] one attribution of values to the variables that satisfies all constraints, failing if there is no solution / if no solution is found within the time limits
 - Examples:

```
| ?- declareVariables(Vars),
    postConstraints(Vars),
    labeling([], Vars).
```

```
?- declareVariables(Vars),
    postConstraints(Vars),
    objectiveFunction(Vars, Profit),
    labeling( [maximize(Profit), ffc,
        bisect, time out(5000,Flag)], Vars).
```

Search Options

- The Options argument of labeling/2 (also used in solve/2) controls several
 parameters of the search
 - Variable ordering
 - Value selection
 - Value ordering
 - Solutions to find
 - Search time limit
 - Search scheme (useful in optimization problems):
 - bab (uses branch-and-bound; value by default), restart
 - Assumptions:
 - assumptions(K): K is unified with the number of choices made
 - Discrepancy:
 - discrepancy(D): in the path to the solution there are at most D choice points in which there
 was backtracking

Variable Ordering

- How to select the next variable?
 - leftmost (default option): leftmost variable from the variable list
 - min: variable with the smallest value in its domain (smallest lower bound)
 - max: variable with the greatest value in its domain (greatest upper bound)
 - f: first-fail principle variable with the smallest domain (fewer possible values)
 - anti_first_fail: variable with the largest domain (more possible values)
 - occurrence: variable with more suspended constraints
 - ffc: variable with the smallest domain, breaking ties by choosing the one with more suspended constraints
 - max_regret: variable with the largest difference between the first two values in its domain

Variable Ordering

- How to select the next variable? (continued)
 - impact: variable that has been involved in the most failures
 - dom_w_deg: variable with the largest (failure count / domain size)
 - variable(Sel):
 - **Sel** is a predicate used to select the next variable, with signature Sel(Vars, Selected, Rest)
 - Must succeed determinately, unifying *Selected* with the selected variable, and *Rest* with a list containing the remaining variables
 - Example:

```
labeling([variable(selRandom)], Vars).

% selects a variable randomly
selRandom(ListOfVars, Var, Rest):-
    random_select(Var, ListOfVars, Rest). % from library(random)
```

Value Selection

- How to select the next value for the current variable?
 - step (default option): binary choice between X #= B and X #\= B, where B is the lower or upper bound of the domain of X
 - enum: multiple choice for X corresponding to the values of its domain
 - bisect: binary choice between X #=< M and X #> M, where M is the middle point of the domain of X (mean between the minimum and maximum values of the domain of X, rounded down)
 - median / middle: binary choice between X #= M and X #\= M, where M is the median / average of the domain of X

Value Selection

- How to select the next value for the current variable?
 - value(Enum):
 - Enum is a predicate that must reduce the domain of X with signature Enum(X, Rest, BBO, BB)
 - Rest is the list of variables that need labeling with the exception of X
 - **Enum** must succeed in a non-determinate manner, providing other means of domain reduction by *backtracking*
 - Must call first bound(BBO, BB) in its first solution and later bound(BBO, BB) in alternate solutions
 - Example:

```
labeling( [ value(selRandom) ], Vars).

selRandom(Var, Rest, BB0, BB1):- % selects value randomly
  fd_set(Var, Set), fdset_to_list(Set, List),
  random_member(Value, List), % da library(random)
  ( first_bound(BB0, BB1), Var #= Value ;
  later_bound(BB0, BB1), Var #\= Value ).
```

Value Ordering

- In which order should the next value for the current variable be selected?
 - (not useful with the value(Enum) option)
 - up (default value): the domain is explored in ascending order
 - down: the domain is explored in descending order

Solutions to Find

- These options indicate if the problem is a constraints satisfaction problem (any solution is valid) or an optimization problem (only the best solution matters):
 - satisfy (default value): all solutions are enumerated by backtracking
 - minimize(X) / maximize(X): the goal is to obtain a solution that minimizes / maximizes the domain variable X
 - The *labeling* mechanism must constrain X to a value for every variable value attributions
 - It is useful to combine this option with time_out/2, best or all
- Options that only make sense with optimization problems:
 - best (default option): obtains the optimal solution, after proving optimality
 - all: obtains, via backtracking, solutions that improve on the previous one

Search Time Limit

- The time_out(Time, Flag) flag defines a time limit for the search
 - Time is the maximum execution time (in milliseconds)
 - If the solver proves that there is no solution to the problem within *Time* ms, the predicate fails
 - If the time limit is reached, or the optimal solution is found, the predicate succeeds and *Flag* is unified with one of the following values:
 - **optimality** the optimal solution has been found (the **best** option was selected) within the time limit; the variables are unified with the values corresponding to the best solution
 - **success** at least one solution was found (but not proof of optimality) within the time limit; the variables are unified with the values corresponding to the best solution found until then
 - time_out the time limit was reached, with no solution found; the variables remain uninstantiated

Search

- solve(:Options, :Searches)
 - Options is a list of search options (similar to the ones used in labeling/2)
 - Searches is a list of one or more labeling/2 or indomain/1 goals
 - Used primarily in optimization problems, allowing the definition of different search heuristics for distinct [sets of] variables
 - Some options are global, while the majority are local
 - Global options in *Options* override the options in the individual *labeling/2* goals in *Searches*
 - Local options in *Options* define default values for search options not appearing in the individual *labeling/2* goals in *Searches*

Optimization

- The optimization predicates allow the search for an optimal solution (minimization / maximization of cost / profit):
 - minimize(:Goal, ?X) / minimize(:Goal, ?X, +Options)
 - maximize(:Goal, ?X) / maximize(:Goal, ?X, +Options)
 - Use a branch-and-bound algorithm to search for the assignment that minimizes/maximizes domain variable $\textbf{\textit{X}}$
 - Goal must be a goal that constrains X to a value, typically a labeling/2 goal
 - The algorithm calls *Goal* repeatedly with an *upper* (*lower*) *bound* of *X* progressively more constrained until a proof of optimality is attained (which, sometimes, can take too long...)
 - *Options* is a list containing one of:
 - best (default option): returns the optimal solution after proof of optimality
 - all: enumerates improving solutions until proof of optimality

Examples

• Enumerate solutions with static variable ordering:

- Minimize a cost function, obtaining only the best solution, dynamic variable ordering using the first-fail principle, and dividing the domain by exploring the upper part of the domain first:
 - | ?- constraints(Variables, Cost),
 labeling([ff,bisect,down,minimize(Cost)], Variables).

Examples

 Minimize the cost, using two different search strategies for two variable subsets:

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5. STATISTICS PREDICATES

Statistics Predicates

- Execution statistics specific to the *clp(fd)* solver:
 - fd_statistics(?Key, ?Value): for each possible key Key, Value is unified with the current value of a counter:
 - resumptions: number of times a constraint was resumed
 - entailments: number of times a (dis)entailment was detected
 - prunings: number of times a domain has been reduced
 - **backtracks**: number of times a contradiction was found (domain depletion or constraint failing)
 - *constraints*: number of created constraints
 - fd_statistics/0: shows a summary of the statistics above (values since the last call to the predicate)

Statistics Predicates

- Other statistics relative to CPU time, memory use and others can be obtained with the predicates:
 - statistics(?Keyword, ?List)): for each possible key Keyword, List is unified with the current value of a counter. Examples:
 - runtime / total_runtime / walltime: execution time (in ms) excluding memory management and system calls / total execution time / absolute time. The first element of the list refers to the time since the beginning of the session, while the second refers to the time since the last call to the statistics predicate.
 - memory_used: used memory (in bytes)
 - Several other options are described in section 4.10.1.2 of SICStus' manual
 - statistics/0 shows a summary of statistics related to execution time, memory, garbage collection, ...

Example

```
testStats(Vars):-
   declareVars(Vars),
   reset_timer,
   postConstraints(Vars),
   print_time('Posting Constraints: '),
   labeling([], Vars),
   print_time('Labeling Time: '),
   fd_statistics,
   statistics.
```

```
reset_timer:-
    statistics(total_runtime, _).

print_time(Msg):-
    statistics(total_runtime,[_,T]),
    TS is ((T//10)*10)/1000, nl,
    write(Msg),
    write(TS),
    write('s'), nl, nl.
```

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Q&A