NOPT042 Constraint programming: Tutorial 7 - Rostering, table constraint

What was in the lecture? Nothing, the lecture was canceled.

Exercise:

- Explain why PC is equivalent to path consistency for paths of length two
- Give an example of an instance which is AC but not PC
- · Give an example of an instance which is PC (with all domains nonempty) but not solvable

```
In [1]: %load_ext ipicat
```

Picat version 3.9

The constraint regular

regular(L, Q, S, M, Q0, F)

Given a finite automaton (DFA or NFA) of Q states numbered $1, 2, \ldots, Q$ with input from $\{1, \ldots, S\}$, transition matrix M, initial state Q_0 ($1 \leq Q_0 \leq Q$), and a list of accepting states F, this constraint is true if the list L is accepted by the automaton. The transition matrix M represents a mapping from $\{1,\ldots,Q\}\times\{1,\ldots,S\}$ to $\{0,\ldots,Q\}$, where 0 denotes the error state. For a DFA, every entry in ${\cal M}$ is an integer, and for an NFA, entries can be a list of integers.

---from the guide

Exercise: Global contiguity

Given a 0-1 sequence, express that if there are 1's, they must form a single, contiguous subsequence, e.g. accept 0000 and 0001111100 but not 00111010. (Problem from the book.)

```
*** error(failed, main/1)
```

```
In [3]: !cat global-contiguity/global_contiguity.pi
       % Adapted from Constraint Solving and Planning with Picat, Springer
       % by Neng-Fa Zhou, Hakan Kjellerstrand, and Jonathan Fruhman
       import cp.
       main([Xstr]) =>
         X = map(to_int,Xstr),
         global_contiguity(X),
         solve(X),
         println("ok").
       global_contiguity(X) =>
         N = X.length,
         InputMax = 2,
         % Translate X's 0..1 to 1..2
         RegInput = new_list(N),
         RegInput :: 1..InputMax, % 1..2
         foreach (I in 1..N)
           RegInput[I] #= X[I]+1
         end,
         % DFA for the regex "0*1*0*"
         Transition = [
           [1,2], % state 1: 0*
           [3,2], % state 2: 1*
           [3,0] % state 3: 0*
         ],
         NStates = 3,
         InitialState = 1,
         FinalStates = [1,2,3],
         regular(RegInput,NStates,InputMax,Transition,InitialState,FinalStates).
```

Exercise: Nurse roster

Schedule the shifts of NumNurses nurses over NumDays days. Each nurse is scheduled for each day as either: (d) on day shift, (n) on night shift, or (o) off. In each four day period a nurse must have at least one day off, and no nurse can be scheduled for 3 night shifts in a row.

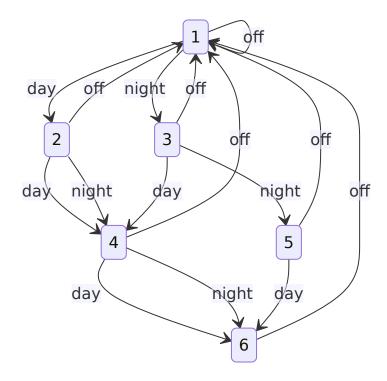
We require ReqDay nurses on day shift each day, and ReqNight nurses on night shift, and that each nurse takes at least MinNight night shifts. (Problem from the MiniZinc tutorial, a similar problem is in the book.)

In [5]: !picat nurse-roster/nurse_roster_regular instance

CPU time 0.0 seconds. Backtracks: 12

day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	day	off	day	night	night	
day	day	off	day	night	night	off	
day	day	off	day	night	night	off	
night	night	off	night	off	day	day	
night	off	night	night	off	day	day	
off	night	night	day	off	day	day	

State diagram of the DFA (in mermaid, will not render in the RISE slides), start state is 1, all states are final:



In [6]: !cat nurse-roster/nurse_roster_regular.pi

```
% Adapted from Constraint Solving and Planning with Picat, Springer
% by Neng-Fa Zhou, Hakan Kjellerstrand, and Jonathan Fruhman
import cp.
main([Filename]) =>
 cl(Filename),
 instance(NumNurses, NumDays, ReqDay, ReqNight, MinNight),
 nurse_rostering(NumNurses, NumDays, ReqDay, ReqNight, MinNight, Roster, Stat),
 Vars = Roster.vars() ++ Stat.vars(),
 time2(solve(Vars)),
 output(Roster).
nurse rostering(NumNurses, NumDays, ReqDay, ReqNight, MinNight, Roster, Stat) =>
 DayShift = 1,
 NightShift = 2,
 OffShift = 3,
 % decision variables
 Roster = new array(NumNurses, NumDays),
 Roster :: DayShift..OffShift,
 % summary of the shifts: day-night-off
 Stat = new_array(NumDays,3),
 Stat :: 0..NumNurses,
 % The DFA for the regular constraint.
 Transition = [
   % day-night-off
   [2,3,1], % state 1
    [4,4,1], % state 2
   [4,5,1], % state 3
   [6,6,1], % state 4
   [6,0,1], % state 5
   [0,0,1] % state 6
  ],
 NStates
             = Transition.length, % number of states
                      % 3 states
 InputMax = 3,
 InitialState = 1,
                                 % start at state 1
                         % all states are final
 FinalStates = 1..6,
 % constraints
 % valid schedule
 foreach (I in 1..NumNurses)
    regular([Roster[I,J] : J in 1..NumDays],
   NStates,
   InputMax,
   Transition,
   InitialState,
   FinalStates)
 end,
 % statistics for each day
 foreach (Day in 1..NumDays)
   foreach (Type in 1..3)
     Stat[Day,Type] #= sum([Roster[Nurse,Day] #= Type : Nurse in 1..NumNurses])
```

```
end,
    sum([Stat[Day,Type] : Type in 1..3]) #= NumNurses,
   % For each day the must be at least 3 nurses with
   % day shift, and 2 nurses with night shift
   Stat[Day,DayShift] #>= ReqDay,
   Stat[Day,NightShift] #>= ReqNight
 % each nurse gets MinNight shifts
 foreach (Nurse in 1..NumNurses)
    sum([Roster[Nurse, Day] #= NightShift : Day in 1..NumDays]) #>= MinNight
output(Roster) =>
 Shifts = new_map(3,[1="| day ",2="| night ",3="| off "]),
 foreach(Nurse in Roster)
   foreach(I in 1..Nurse.length)
      print(get(Shifts,Nurse[I]))
   print("|\n")
  end.
```

Constraint sliding_sum (not available in Picat)

```
sliding_sum(Low, Up, Seq, Variables) =>
  foreach(I in 1..Variables.length-Seq+1)
    Sum #= sum([Variables[J] : J in I..I+Seq-1]),
    Sum #>= Low,
    Sum #=< Up
end.</pre>
```

-- from Hakank's Picat webpage, model sliding sum.pi.

The table constraint

A table constraint, or an extensional constraint, over a tuple of variables specifies a set of tuples that are allowed (called positive) or disallowed (called negative) for the variables. A positive constraint takes the form

```
table_in(Vars,R)
```

where Vars is either a tuple of variables or a list of tuples of variables, and R is a list of tuples in which each tuple takes the form $[a_1,\ldots,a_n]$, where a_i is an integer or the don't-care symbol *. A negative constraint takes the form:

```
table_notin(Vars, R)
```

Exercise: Nurse roster using table_in

Model the above nurse roster problem using the constraint table_in. The model is slower, we will need a simpler instance. And, for simplicity, assume that NumDays = 7.

```
In [7]:
       !cat nurse-roster/instance2.pi
      instance(NumNurses, NumDays, ReqDay, ReqNight, MinNight) =>
          NumNurses
                        = 8,
                        = 7,
          NumDays
                        = 2, % minimum number in day shift
          RegDay
           ReqNight
                      = 2, % minimum number in night shift
          MinNight
                       = 1. % minimum night shifts for each nurse
In [8]: !picat nurse-roster/nurse_roster_table instance2
      CPU time 0.04 seconds. Backtracks: 7796
                             | night | night |
         day
                 day
                         off
                                                off
                                                        off
         day
                 day
                         off
                             | night | night |
                                                off
                                                        off
                     | night | night |
                                        off
                 off
                                                off
                                                        day
              | off | night | night |
                                                off
                                         off
                                                        day
        night | night |
                         off
                                 off |
                                         day
                                                day
                                                        off
        night | night |
                                         day
                                                day
                                                        off
                         off
                                 off
         off
                 off
                                 day
                                         off
                                               night | night |
                         day
                 off |
                         day
                                 day |
                                        off
                                             | night | night |
In [9]: !cat nurse-roster/nurse_roster_table.pi
```

```
% Adapted from Constraint Solving and Planning with Picat, Springer
% by Neng-Fa Zhou, Hakan Kjellerstrand, and Jonathan Fruhman
import cp.
main([Filename]) =>
 cl(Filename),
 instance(NumNurses, NumDays, ReqDay, ReqNight, MinNight),
 nurse_rostering(NumNurses, NumDays, ReqDay, ReqNight, MinNight, Roster, Stat),
 Vars = Roster.vars() ++ Stat.vars(),
 time2(solve(Vars)),
 output(Roster).
% rotate valid schedules
rotate_left(L) = rotate_left(L,1).
rotate_left(L,N) = slice(L,N+1,L.length) ++ slice(L,1,N).
nurse_rostering(NumNurses, NumDays, ReqDay, ReqNight, MinNight, Roster, Stat) =>
 % Only works for 7-day rosters!
 NumDays = 7,
 DayShift = 1, D = 1,
 NightShift = 2, N = 2,
 OffShift = 3, 0 = 3,
 % Valid 7 day schedules:
 % - up to rotation:
 Valid_up_to_rotation = [
   [D,D,D,D,D,O,O],
   [N,0,N,0,D,D,0],
   [N,N,O,O,D,D,O]
 % - create all rotational variants
 Valid = [],
 foreach (V in Valid_up_to_rotation, R in 0..V.length-1)
   Rot = rotate_left(V,R).to_array(),
   Valid := Valid ++ [Rot]
 end,
 % decision variables:
 % - the roster
 Roster = new_array(NumNurses, NumDays),
 Roster :: DayShift..OffShift,
 % - summary of the shifts: day-night-off]
 Stat = new_array(NumDays,3),
 Stat :: 0..NumNurses,
 % constraints
 % - valid schedule
 foreach (Nurse in 1..NumNurses)
   table_in([Roster[Nurse,Day] : Day in 1..NumDays].to_array(), Valid)
 end,
 % - statistics for each day
```

```
foreach (Day in 1..NumDays)
 foreach (Type in 1..3)
    Stat[Day,Type] #= sum([Roster[Nurse,Day] #= Type : Nurse in 1..NumNurses])
  sum([Stat[Day,Type] : Type in 1..3]) #= NumNurses,
 % For each day the must be at least 3 nurses with
 % day shift, and 2 nurses with night shift
 Stat[Day,DayShift] #>= ReqDay,
  Stat[Day,NightShift] #>= ReqNight
end,
% - each nurse gets MinNight shifts
foreach (Nurse in 1..NumNurses)
  sum([Roster[Nurse, Day] #= NightShift : Day in 1..NumDays]) #>= MinNight
end.
output(Roster) =>
  Shifts = new_map(3,[1="| day ",2="| night ",3="| off "]),
 foreach(Nurse in Roster)
   foreach(I in 1..Nurse.length)
      print(get(Shifts,Nurse[I]))
    end,
    print("|\n")
  end.
```

Exercise: Graph homomorphism

Given a pair of graphs G,H, find all homomorphisms from G to H. A graph homomorphism is a function $f:V(G)\to V(H)$ such that

$$\{u,v\} \in E(G) \Longrightarrow \{f(u),f(v)\} \in E(H)$$

- Generalizes graph k-coloring ($c:G o K_k$)
- · Easier version: oriented graphs
- How would you model the Graph Isomorphism Problem?