



Modelling for Constraint Programming

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4. Combining Viewpoints, Modelling Advice

Which Viewpoint to Choose?



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- Sometimes one viewpoint is clearly better, e.g. if we can't express the constraints easily in one
- But different perspectives often allow different expression of the constraints and different implied constraints
 - can be hard to decide which is better
- We don't need to choose one viewpoint – we can use two (or more) at once
- We need *channelling constraints* to link the variables



- Dual viewpoints of a permutation problem with variables x_1, x_2, \dots, x_n and d_1, d_2, \dots, d_n
- Combine them using the channelling constraints
 - $(x_i = j) \equiv (d_j = i)$
- The channelling constraints are sufficient to ensure that x_1, x_2, \dots, x_n are all different
 - but might be beneficial to have a specific allDifferent constraint as well and enforce GAC

Combining Viewpoints: Integer & Set Variables



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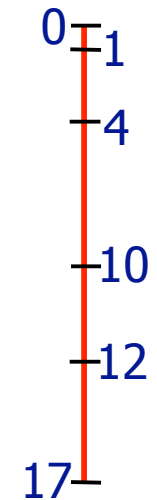
- In a nurse rostering problem, we can allocate shifts to nurses or nurses to shifts
- First viewpoint:
 - an integer variable n_{ij} for each nurse i and day j
 - its value is the shift that nurse i works on day j
- Second viewpoint:
 - a set variable S_{kj} for each shift k and day j
 - its value is the set of nurses that work shift k on day j
- Channelling constraints: $(n_{ij} = k) \equiv (i \in S_{kj})$

The Golomb Ruler Problem



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- A Golomb ruler with m marks consists of
 - a set of m integers $0 = x_1 < x_2 < \dots < x_m$
 - the $m(m-1)/2$ differences $x_j - x_i$ are all different
 - Objective: find a ruler with minimum length x_m
- First viewpoint: variables x_1, x_2, \dots, x_m
 - $x_j - x_i \neq x_l - x_k$ for all distinct pairs
 - $x_1 < x_2 < \dots < x_m$
- Second viewpoint: variables $d_{ij}, 1 \leq i < j \leq m$
 - $\text{allDifferent}(d_{11}, d_{12}, \dots, d_{m-1,m})$
 - $d_{ik} = d_{ij} + d_{jk}$ for $1 \leq i < j < k \leq m$
- Channelling constraints: $d_{ij} = x_j - x_i$





Constraints in Combined Viewpoints

- It is safe to combine two complete models of a problem, with channelling constraints
- But often unnecessary and inefficient
- If some constraints are more easily expressed in one viewpoint, we don't need them in both
 - e.g. nurse rostering
 - constraints on nurse availability are stated in the 'nurse' viewpoint
 - constraints on work requirements (e.g. no. of nurses required for each shift) are stated in the 'shift' viewpoint
- or if they propagate better in one viewpoint
 - e.g. $x_j - x_i \neq x_l - x_k$ v. `allDifferent($d_{11}, d_{12}, \dots, d_{m-1,m}$)` in the Golomb ruler problem



- We need to choose a set of variables such that an assignment to each one, satisfying the constraints, will give a complete solution to the problem
- Assume we pass the search variables to the search algorithm in a list or array
 - the order defines a static variable ordering
 - though we can still use a dynamic ordering



- When a model combines two (or more) viewpoints of a problem, which variables should drive the search?
- Assigning values to either set of variables would be sufficient to solve the problem
 - even if we did not express the problem constraints on those variables
 - the channelling constraints ensure that we can assign values to one set of variables but define the constraints in the other viewpoint, if we want



- We can use both sets of variables as search variables
 - e.g. use a dynamic variable order e.g. variable with smallest domain in either viewpoint
 - combines variable and value ordering: the dual variable with smallest domain corresponds (in the other viewpoint) to the value occurring in fewest domains



- Whether a given node is on a given ring:
 - $x_{ij} = 1$ if node i is on ring j
- Which ring(s) each node is on:
 - N_i = set of rings node i is on
- Which nodes are on each ring
 - R_j = set of nodes on ring j
- In principle, any of these viewpoints could be the basis of a complete CSP model
 - channelling constraints $(x_{ij} = 1) \equiv (i \in R_j) \equiv (j \in N_i)$
- There are also auxiliary variables
 - n_i = the number of rings each node is on ($= |N_i|$)



- Use just one set of variables, e.g. x_{ij} – the others are just for constraint propagation
- Use two (or more) sets of variables (of the same type) e.g. R_j, N_i
 - interleave them in a sensible (static) order
 - or use a dynamic ordering applied to both sets of variables
- Use an incomplete set of variables first, to reduce the search space before assigning a complete set
 - e.g. decide how many rings each node is on (search variables n_i) and then which rings each node is on (x_{ij})
 - another strategy adds assigning the objective variable first – see earlier



- There are lots of choices to make in modelling a problem as a CSP
 - difficult even with experience
- Can it be automated?
 - some initial steps so far
 - e.g. systems that propose models given a high-level specification
 - descriptions of common patterns in modelling



- *Reduce the number of variables*
 - if we only use one viewpoint:
 - a model which needs fewer variables to describe the solutions to the problem is likely to be a better model
 - e.g. an integer model is probably better than a Boolean model
 - *But* only if the variables allow the constraints to be expressed in a way that propagates well
 - artificially reducing the number of variables by inventing a single variable to replace a pair of variables will not give a better model



➤ *Reduce the number of constraints*

- One viewpoint in the magic squares problem has far more constraints than the other
- rewriting a set of constraints in a more compact form is likely to be beneficial, *if* the resulting constraints can propagate efficiently
 - e.g. combine constraints with the same scope
 - use a global constraint to replace a set of constraints
- *But* simply conjoining constraints for the sake of reducing their number will not give a better model if the new constraints cannot propagate efficiently



- *Add more variables*
 - auxiliary variables to allow constraints to be expressed
 - new viewpoints allowing a different perspective on the problem
- *Add more constraints*
 - implied constraints
 - channelling constraints to link new variables
- *Check empirically*
 - that a change does reduce run-time



- Aim for a rich model
 - multiple viewpoints
 - auxiliary variables
 - implied constraints
- Understand the problem as well as you can
 - build that insight into the model
 - the better you can understand a problem, the better you can solve it

THE END