

Constraint Relationships Step-by-Step Enhancing a MiniZinc Model



Synopsis



We will ...

- start with an n-queens model in MiniZinc
- add some additional soft constraints (e.g. one queen should be placed in the center)
- prioritize these soft constraints using constraint relationships
- switch dominance properties (SPD, TPD)
- and solve with: Branch-and-Bound (BaB), Large Neighborhood Search (LNS)

http://isse-augsburg.github.io/constraint-relationships/

Satisfaction Model



```
include "globals.mzn";
int: n; array[1..n] of var 1..n: queens; n = 8;

solve :: int_search(queens, first_fail, indomain_median, complete)
satisfy;

constraint all_different(queens);
constraint all_different([queens[i]+i | i in 1..n]);
constraint all_different([queens[i]-i | i in 1..n]);
```

```
queens = array1d(1..8,[4, 6, 1, 5, 2, 8, 3, 7]);
```

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Step 1: Add Soft Constraint Support



```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
int: n; array[1..n] of var 1..n: queens; n = 8;

solve :: int_search(queens, first_fail, indomain_median, complete)
satisfy;

constraint all_different(queens);
constraint all_different([queens[i]+i | i in 1..n]);
constraint all_different([queens[i]-i | i in 1..n]);
```

- Adds nScs (a new parameter) boolean variables, one for each soft constraint
- Adds parameter penalties (use [1 | i in 1..nScs] if not relevant)
- Intuition: The cost of violating soft constraint i should be penalties[i]

Step 1: Behind the Scenes



```
include "link set to booleans.mzn":
% MODEL-SPECIFIC parameters: nScs and penalties
int: nScs; set of int: SOFTCONSTRAINTS = 1..nScs;
% reified variables for each soft constraint
array[SOFTCONSTRAINTS] of var bool: satisfied;
array[SOFTCONSTRAINTS] of var bool: violated =
   [not satisfied[sc] | sc in SOFTCONSTRAINTS] :
array[SOFTCONSTRAINTS] of int: penalties;
var int: penSum = sum(sc in SOFTCONSTRAINTS)
  (bool2int(not satisfied[sc]) * penalties[sc]);
var set of SOFTCONSTRAINTS: violatedScs;
constraint link_set_to_booleans(violatedScs, violated);
```

- Using satisfied and violated is done for convenience in expressing soft constraints
- There is also a variant without violatedScs for solvers that do not support set variables (e.g. MIP): soft_constraints_noset.mzn

Step 1: Behind the Scenes



```
% find the sorted permutation of soft constraint instances
array[SOFTCONSTRAINTS] of SOFTCONSTRAINTS: sortPermScs =
   arg_sort(penalties);
% invert, since arg_sort use <= and we need decreasing order
array[SOFTCONSTRAINTS] of SOFTCONSTRAINTS: mostImpFirst =
   [ sortPermScs[nScs-s+1] | s in SOFTCONSTRAINTS];</pre>
```

- mostImpFirst gives an array of soft constraint indices in decreasing order (start with most important ones)
- Used later as a (generic) labeling strategy

Step 2: Add Soft Constraints



```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
int: n; array[1..n] of var 1..n: queens; n = 8;
solve :: int_search(queens, first_fail, indomain_median, complete)
minimize penSum;
constraint all_different(queens);
constraint all_different([queens[i]+i | i in 1..n]);
constraint all_different([queens[i]-i | i in 1..n]);
% soft constraint business
nScs = 3; penalties = [1 | i in SOFTCONSTRAINTS];
constraint queens[n div 2] = n div 2 <-> satisfied[1] ;
constraint queens[2] = queens[1] + 2 <-> satisfied[2];
constraint queens[3] = queens[2] + 2 <-> satisfied[3];
```

- Just tie reified soft constraint variables to actual soft constraints
- [1 | i in 1..nScs] makes it a Max-CSP when minimizing penSum

Step 2: What about Globals?



What if want to use global constraints as soft constraints? Consider (reified.mzn):

```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
array[1..5] of var 0..5: x;
constraint alldifferent(x); % hard constraint

% but it would be nice to have x increase ...
nScs = 1; penalties = [1];
constraint increasing(x) <-> satisfied[1];

solve maximize bool2int(satisfied[1]);
output["satisfied[1] = \(satisfied[1]), x = \(x)"];
```

Solved with Gecode:

MiniZinc: flattening error: 'increasing_int' is used in a reified context but no reified version is available

Step 2: What about Globals?



- a) Use a solver that has the reified global, e.g., here:
 - G12-FD
 - JaCoP
- b) Use a decomposition from the MiniZinc standard library (reified_fix.mzn)

```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
array[1..5] of var 0..5: x;
constraint alldifferent(x); % hard constraint
nScs = 1; penalties = [1];
% copied from share/minizinc/std/increasing_int.mzn
predicate my_increasing_int(array[int] of var int: x) =
   forall(i in index_set(x) diff { min(index_set(x)) })
      (x[i-1] \le x[i]):
constraint my_increasing_int(x) <-> satisfied[1];
solve maximize bool2int(satisfied[1]);
output["satisfied[1] = \sl(satisfied[1]), x = \sl(x)"];
```

Step 2: What about Globals?



Solved reified_fix.mzn with Gecode (same as reified.mzn solved with JaCoP or G12-FD):

We do not get the performance benefits of globals (in a decomposition) but at least their conciseness.

Step 3: Prioritize Soft Constraints (Weights)



```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
int: n; array[1..n] of var 1..n: queens; n = 8;
solve :: int_search(queens, first_fail, indomain_median, complete)
minimize penSum; output["penSum = \((penSum), queens = \((queens)")];
constraint all_different(queens);
constraint all_different([queens[i]+i | i in 1..n]);
constraint all_different([queens[i]-i | i in 1..n]);
% soft constraint business
nScs = 3; penalties = [2, 1, 1];
constraint queens[n div 2] = n div 2 <-> satisfied[1] ;
constraint queens[2] = queens[1] + 2 <-> satisfied[2];
constraint queens[3] = queens[2] + 2 <-> satisfied[3];
```

- If you manually set weights, just add them
- Switch from [1 | i in 1..nScs] to [2, 1, 1]

Step 3: Prioritize Soft Constraints (Relations)



```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
include "soft_constraints/cr_types.mzn"; % graph types
include "soft_constraints/cr_weighting.mzn"; % weighting functions
int: n; array[1..n] of var 1..n: queens; n = 8;
solve :: int_search(queens, first_fail, indomain_median, complete)
minimize penSum; output["penSum = \((penSum), queens = \((queens)")];
constraint all_different(queens);
constraint all_different([queens[i]+i | i in 1..n]);
constraint all_different([queens[i]-i | i in 1..n]);
% soft constraint business
nScs = 3:
penalties = [weighting(s, SOFTCONSTRAINTS, crEdges, true)
             | s in SOFTCONSTRAINTS];
constraint queens[n div 2] = n div 2 <-> satisfied[1] ;
constraint queens[2] = queens[1] + 2 <-> satisfied[2];
constraint queens[3] = queens[2] + 2 <-> satisfied[3];
```

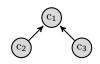
Step 3: Prioritize Soft Constraints (Relations)



```
include "soft_constraints/cr_types.mzn"; % graph types
[... inside cr_types.mzn]
% constraint-relationship-types
int: nCrEdges;
array[1..nCrEdges, 1..2] of SOFTCONSTRAINTS: crEdges;
```

Concrete instantiation

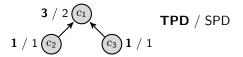
```
% data file for soft edges,
% constraint c1 is more important than c2 and c3
nCrEdges = 2;
crEdges = [| 2, 1 | 3, 1 |];
```



Step 3: Prioritize Soft Constraints (Relations)



```
% weighting(s, SOFTCONSTRAINTS, crEdges, true)
[2, 1, 1]
% weighting(s, SOFTCONSTRAINTS, crEdges, false) more than chi. together
[3, 1, 1]
```



Step 4: Switch to PVS Architecture



```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
include "soft_constraints/spd_worse.mzn"; % solution ordering
include "soft_constraints/pvs_spd.mzn"; % concrete PVS
int: n; array[1..n] of var 1..n: queens; n = 8; nScs = 3;
solve :: int_search(queens, first_fail, indomain_median, complete)
minimize penSum; output["penSum = \((penSum), queens = \((queens)")];
constraint all_different(queens);
constraint all_different([queens[i]+i | i in 1..n]);
constraint all_different([queens[i]-i | i in 1..n]);
constraint queens[n div 2] = n div 2 <-> satisfied[1] ;
constraint queens[2] = queens[1] + 2 <-> satisfied[2];
constraint queens[3] = queens[2] + 2 <-> satisfied[3];
```

- Defines single-predecessor-dominance as solution ordering with predicate spd_worse
- Hides the weight calculation in pvs_spd.mzn

Step 4: Switch to PVS Architecture



Idea Many soft constraint formalisms are generalized by partial valuation structures (Gadducci et al., 2013) that give the codomain of the objective function an *algebraic structure*.

- A partially ordered valuation structure is described by $(M, \oplus_M, \leq_M, \varepsilon_M)$ where
 - M . . . is a set of violation/satisfaction degrees, e.g., $\mathbb N$ for weights, [0,1] for probabilities etc.
 - \bigoplus_M ... is a binary *combination* operation to aggregate values from M, e.g., $3 \bigoplus_M 5 \equiv 3+5$
 - \leq_M ... is a partial *order* over M operation to rank values from M, e.g., $5 \leq_M 3 \equiv 5 \geq 3$, $m \leq_M n$ means m is worse than n
 - $m \leq_M \varepsilon_M$... for every $m \in M$, i.e. ε_M is the *best* possible solution, e.g. 0 violation
- Similar (Bistarelli et al., 1999): c-semirings and (total) valuation structures (toulbar2)

Step 4: Switch to PVS Architecture



Concrete PVS	М	⊕м	≤ _M	ε_{M}
Weighted CSP	N	+	≥	0
Fuzzy CSP	[0, 1]	min	\leq	1
Constraint Relationships ¹	2 ^{C_s}	U	⊆spd	Ø

Main Idea

Implement search strategies (BaB and LNS) for partially ordered valuation structures. Instantiate for concrete problems.

From now on we rely on MiniSearch (www.minizinc.org/minisearch/).

 $^{{}^{1}\}mathit{C}_{s}$ is the set of soft constraints, \subseteq_{SPD} is the SPD-ordering on sets.

Step 5: Use PVS-based Search



```
include "globals.mzn";
include "soft_constraints/soft_constraints.mzn";
include "soft_constraints/spd_worse.mzn"; % solution ordering
include "soft_constraints/pvs_spd.mzn"; % concrete PVS
include "minisearch.mzn"; % custom search combinators
include "soft_constraints/constraint_relationship_search.mzn";
int: n; array[1..n] of var 1..n: queens; n = 8; nScs = 3;
output["queens= \(queens), obj = \(violatedScs) "];
solve :: int_search(queens, first_fail, indomain_median, complete)
search strictlyBetterBAB(violatedScs) /\ print();
constraint all_different(queens);
constraint all_different([queens[i]+i | i in 1..n]);
constraint all_different([queens[i]-i | i in 1..n]);
constraint queens[n div 2] = n div 2 <-> satisfied[1] ;
constraint queens[2] = queens[1] + 2 <-> satisfied[2];
constraint queens[3] = queens[2] + 2 <-> satisfied[3];
```

minisearch soft-queens-5.mzn cr-soft-queens.dzn

Step 5: Use PVS-based Search



Only relies on isWorse!

```
[... inside pvs_spd.mzn]

predicate isWorse(var set of int: leftViolatedScs,

var set of int: rightViolatedScs) = (

spd_worse(leftViolatedScs, rightViolatedScs, SOFTCONSTRAINTS, crEdges)
);
```

Step 5: Use PVS-based Search



Similar for large neighborhood search

```
% adapted from lns_max an objective value
function ann: lns_cr (var set of SOFTCONSTRAINTS: violatedScs,
                    array[int] of var int: x,
                 int: iterations, float: d) =
   repeat (i in 1..iterations) (
       scope(
           post(neighbourhoodCts(x,d)) /\ next() /\ commit() /\
           print("Intermediate solution: \((sol(violatedScs))\n") /\
          print()
       ) /\
       let { set of SOFTCONSTRAINTS: lb = sol(violatedScs); } in
       ( post(
          isWorse(lb. violatedScs)
  );
```

Only relies on isWorse!

Summary



This concludes our little transition guide to enhance a MiniZinc model

Make sure to check out our other slides about:

- Language features (more insight about the search strategies and consistency helpers)
- Case Studies (for some specific examples)

http://isse-augsburg.github.io/constraint-relationships/

Quellen I



Bistarelli, S., Montanari, U., Rossi, F., Schiex, T., Verfaillie, G., and Fargier, H. (1999).

Semiring-Based CSPs and Valued CSPs: Frameworks, Properties, and Comparison.

Constraints, 4(3):199-240.

Gadducci, F., Hölzl, M., Monreale, G., and Wirsing, M. (2013). Soft constraints for lexicographic orders.

In Castro, F., Gelbukh, A., and González, M., editors, *Proc.* 12th Mexican Int. Conf. Artificial Intelligence (MICAI'2013), Lect. Notes Comp. Sci. 8265, pages 68–79. Springer.