Recent Results in Symmetry Breaking

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Joint work (only one name appears on all papers) between:

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- and Magnus Ågren (Uppsala)

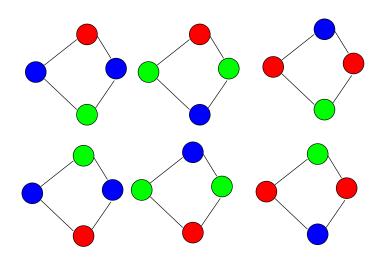
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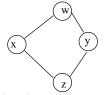
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- Specialised search procedures to break symmetry,
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- Dealing with conditional symmetry.

What is a Symmetry?



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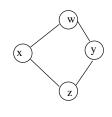
 The most general idea of a symmetry is a bijection that takes candidate solutions to candidate solutions such that solutions are preserved.

A solution is a map from $V = \{w, x, y, z\}$ to

 $D = \{R(=0), G(=1), B(=2)\}$. A symmetry is a bijection on the set of maps that preserves solutions.

Example: $\sigma(f:V\to D)(v)=(f(v)+1)\mod 3$

What is a Symmetry?



• Given any permutation π on $\{R, G, B\}$ then:

$$\sigma_{\pi}(f:V\to D)=\lambda v(\pi(f(v))$$

• Also we can swap around x and y. So

$$\sigma_{x \leftrightarrow y} f(v) = egin{cases} f(y) & \text{if } v = x \\ f(x) & \text{if } v = y \\ f(v) & \text{otherwise} \end{cases}$$

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- You can make it smaller by requiring good behaviour of the group action w.r.t. partial maps (getting essentially variable value pairs map to variable value pairs)
- Or simply declare which subgroups you are interested in.

How to Break Symmetry?

Modify the CSP or the search procedure so that the following two conditions are met:

- If there is a solution then the modified CSP has, or the search procedure produces, a symmetrically equivalent solution.
- We only pick out one representative from each orbit.

Sometimes this is too hard so the second condition is relaxed.

What's on the Market?

Adding constraints before search.

Sometimes it is obvious from the problem which constraints to add.

A list of n variables $[x_1, \ldots, x_n]$ representing a set, so all the variables are interchangeable. Then simply assert that:

$$x_1 < x_2 < \cdots < x_n$$

Other symmetry breaking predicates include lex, lex-chain, multi-set ordering.

Lex Constraints - Crawford and Company

When there is a group of variable permutation symmetries, for each variable permutation add a constraint that lexicographically picks out one of the solutions.

For each permutation π of the variables of the problem add the constraint:

$$[x_1,\ldots,x_n] \leq_{\text{lex}} [x_{\pi(1)},\ldots,x_{\pi(n)}]$$

Example

Consider the following matrix problem

$$\left(\begin{array}{cc} x_1 & x_2 \\ x_3 & x_4 \end{array}\right)$$

where rows and columns can be swapped, that is each solution has 4 (that is 3) symmetric solutions:

$$\left(\begin{array}{cc} x_2 & x_1 \\ x_4 & x_3 \end{array}\right) \quad \left(\begin{array}{cc} x_3 & x_4 \\ x_1 & x_2 \end{array}\right) \quad \left(\begin{array}{cc} x_4 & x_3 \\ x_2 & x_1 \end{array}\right)$$

Then add the constraints: $[x_1, x_2, x_3, x_4] \leq [x_2, x_1, x_4, x_3]$, $[x_1, x_2, x_3, x_4] < [x_3, x_4, x_1, x_2], [x_1, x_2, x_3, x_4] < [x_4, x_3, x_2, x_1].$

Problems with Lex-leader

- Big groups have many lex-constraints.
- Lots of constraints, large overhead.

Hard Problem: find the smallest set of constraints breaking all symmetry or even small sets breaking a large amount of the symmetry.

(NP-complete in general)

Big groups often have simple symmetry breaking predicates.

Symmetry Breaking During Search

Two main methods on the market.

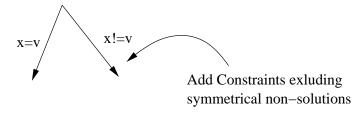
- Add constraints during search that break the symmetry.
- Dominance checking (Global Cuts)

and also

Modifying the search procedure directly for a specific symmetry group.

Symmetry Breaking During Search

Gent & Smith and Backofen & Will.



The idea is that we take the current no-good and add the symmetric set of no-goods, one for each symmetry in the symmetry group.

Symmetry Breaking by Dominance Checking

Fahle, Schamberger & Sellmann and Focacci & Milano.

- In some sense dual to the SBDS.
- At each level in the search tree check if there is a symmetrically equivalent non-solution.

Of course it is more complicated ...

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 - A value symmetry is a bijection $\pi:D\to D$ such that $f:V\to D$ is a solution implies that $\pi\circ f$ is a solution.
- Combinations f is a solution implies $\pi \circ f \circ \sigma$ is a solution.

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- the symmetric group, consider CSPs where either or both π and σ come from the full symmetric group S_V or S_D . (sometimes called full symmetries)
- Partial symmetries. Given a partition $P = P_1 \cup \cdots \cup P_n$ study symmetries the preserve the partition $x \in P_i$ implies $x.\pi$ in P_i .
- Various subgroups of wreath products. In the social golfer you can either permute whole weeks or within each week you can permute the groups.

The idea is that these symmetries are easy to understand and capture some interesting problems.

Intractability

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The dominance checking problem for such CSPs is NP-hard. The proof is not as easy as it first might seem. You have to show the NP-hardness for dominance detecting problems that arise during search.

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- With full value symmetry the dominance check is very simple.
 - At each node if you have used the values d_1, \ldots, d_k then the equivalence classes are:

$$\{d_1\},\ldots,\{d_k\},\{d_{k+1},\ldots,\}$$

• Pick an previously used value or one new value.

Labelling Algorithm

```
bool fValllabel(\mathcal{P}) {
     return fValllabelA(\mathcal{P}, \epsilon);
bool fValllabelA(\langle V, D, C \rangle, \theta) {
     if scope(\theta) = V then
          return C(\theta);
     select v in V \setminus scope(\theta);
     A := image(\theta);
     if A \neq D then
          select f in D \setminus A; A := A \cup \{f\};
     forall(d \in A)
          \theta' := \theta \ \& \ v = d:
          if \neg Failure(\langle V, D, C \rangle, \theta') then
               if fValllabelA(\langle V, D, C \rangle, \theta') then
                     return true:
```

Generalises to wreath problems

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- Generalised into a much more general context in GE-trees.

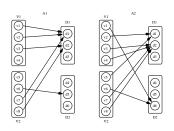
- Generalises to wreath problems
- Generalised into a much more general context in GE-trees.
- Although specialised search procedures often have better complexity bounds.

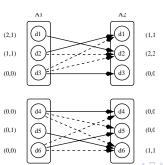
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- Key idea build signatures the count how many times values appears in variable partitions.
- Define $sig(d_i)$ to be a vector of values of how many times d_i appears in each variable partition.
- Build a bipartite graph the has a perfect matching iff one assignments dominates the other.





Conditional Symmetry

- A conditional symmetry is symmetry with a precondition.
- For example

$$(v_1=1)
ightarrow V_2$$
 and v_3 are symmetric

- A set of conditional symmetries forms a groupoid.
- The dominance checking problem with groupoids is much harder than for groups.

References

All of the papers can possibly be found at http://www.it.uu.se/research/astra or more likely by following the links from http://user.it.uu.se/~justin.