

Title: Dual Viewpoint Heuristics for
Binary Constraint Satisfaction Problems

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Foundations of Constraint Satisfaction
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Contributions

I- New heuristics for variable & value selection

II- Double-viewpoint strategy

(common in scheduling: job vs. resource-centered perspective)

[Sadeh '91]

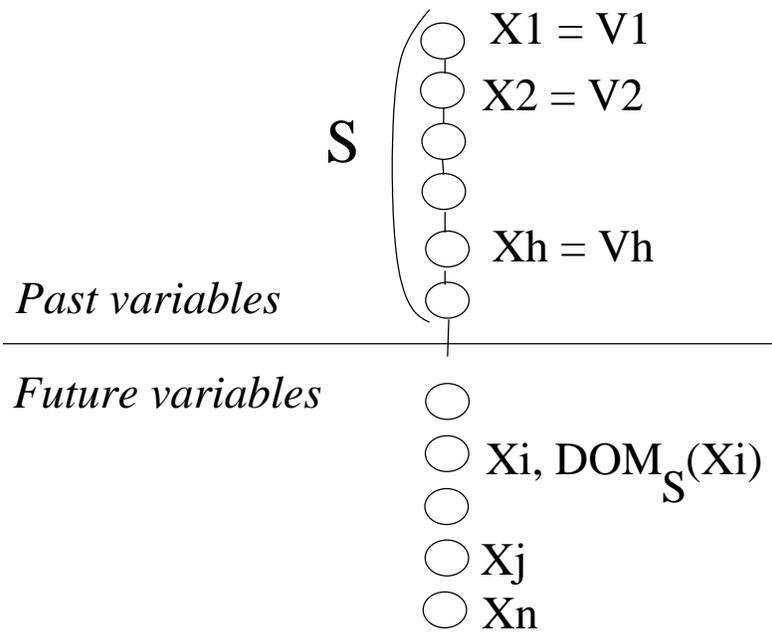
III- Validation on the n -Queen problem

Assumptions

- Binary constraints, finite domains
- Seeking **one** solution (relevant for value ordering)
- Using backtracking (BT) and forward-checking (FC)

I- Goal of Var/Val orderings in BT

- **avoid constraint violation**
 - select values that do not cause constraint violation
 - most promising value first
- **discover constraint violation quickly**
 - select variables that do not delay constraint violation
 - most constrained variable first (fail-first principle)



After forward-checking given S
 $\left\{ \begin{array}{l} \text{DOM}_S(X_i) \\ \vdots \\ \text{DOM}_S(X_j) \\ \text{DOM}_S(X_n) \end{array} \right.$

For $X_i = V_i$, we can separate $\text{DOM}_S(X_j)$ into two sets in $\mathcal{O}(a)$:

- values consistent with $X_i = V_i$
 \rightarrow of size $\text{LEFT}_S(X_j \mid X_i = V_i)$
- values inconsistent with $X_i = V_i$
 \rightarrow of size $\text{LOST}_S(X_j \mid X_i = V_i)$

Value selection for X_i

Choose the least constraining value V_i

1. Minimize X_j future variables

$$\text{Cost}_S(X_i = V_i) = \sum_{X_j \neq i} \text{LOST}_S(X_j \mid X_i = V_i)$$

2. Minimize X_j future variables

$$\text{Cruciality}_S(X_i = V_i) = \sum_{X_j \neq i} \frac{\text{LOST}_S(X_j \mid X_i = V_i)}{|\text{DOM}_S(X_j)|}$$

3. Maximize X_j future variables

$$\text{Promise}_S(X_i = V_i) = \prod_{X_j \neq i} \text{LEFT}_S(X_j \mid X_i = V_i)$$

→ number of assignments that $X_i = V_i$ and can be done such that no constraint on X_i is violated

(3) is more discriminating ($\text{LEFT}_S(X_j \mid X_i = V_i) = 0$)

Value selection: example

- Minimize X_j future variables

$$\text{Cost}_S(X_i = V_i) = \sum_{X_j \neq i} \text{LOST}_S(X_j | X_i = V_i)$$

| | | | | |
|----|---|---|---|---|
| X1 | 6 | 6 | 6 | 6 |
| X2 | 6 | 8 | 8 | 6 |
| X3 | 6 | 8 | 8 | 6 |
| X4 | 6 | 6 | 6 | 6 |

- Maximize X_j future variables

$$\text{Promise}_S(X_i = V_i) = \prod_{X_j \neq i} \text{LEFT}_S(X_j | X_i = V_i)$$

| | | | | |
|----|---|---|---|---|
| X1 | 8 | 6 | 6 | 8 |
| X2 | 8 | 2 | 2 | 8 |
| X3 | 8 | 2 | 2 | 8 |
| X4 | 8 | 6 | 6 | 8 |

Variable selection

Choose the most constrained variable (FFP) X_i

1. Least domain (LD)

2. Maximize $V_i \in \text{DOM}_S(X_i)$

Criticality $_S(X_i) =$

$$\prod_{V_i} \frac{1}{(1+|\text{DOM}_S(X_i)| \times \text{Cruciality}_S(X_i=V_i))}$$

3. Minimize $V_i \in \text{DOM}_S(X_i)$

$$\text{Promise}_S(X_i) = \sum_{V_i} \text{Promise}_S(X_i = V_i)$$

→ number of assignments that can be done such that no constraint on X_i is violated

Variable selection: example

Minimize: $V_i \in \text{DOM}_S(X_i)$

$$\text{Promise}_S(X_i) = \sum_{V_i} \text{Promise}_S(X_i = V_i)$$

| | | | | | |
|----|---|---|---|---|----|
| X1 | 8 | 6 | 6 | 8 | 28 |
| X2 | 8 | 2 | 2 | 8 | 20 |
| X3 | 8 | 2 | 2 | 8 | 20 |
| X4 | 8 | 6 | 6 | 8 | 28 |

X_1, X_4 promise 28 solutions

X_2, X_3 promise 20 solutions, more constraining

Start with X_2 or X_3 (more constraining) and,
choose columns 1 or 4 (more promising)

Summary

Most promising value:

- (1) Minimum cost
- (2) Minimum cruciality
- (3) Maximum promise

[Keng & Yun, 89]

[Geelen'92]

Most constrained variable:

- Least domain (LD)
- (4) Maximum criticality
 - (5) Minimum (false) promise

[Keng & Yun, 89]

[Geelen'92]

→ Dynamic variable/value orderings

Algorithms

| Identifier | Choice of Var | Choice of val |
|------------|---------------|-----------------|
| LD+1 | least domain | Minimum cost |
| LD+2 | – | Min. cruciality |
| LD+3 | – | Max promise |
| FE+2/4 | Max. critical | Min. cruciality |
| FE+3/5 | Min. promise | Max. promise |

LD: time $\mathcal{O}(na^2)$, space $\mathcal{O}(na)$

FE: time $\mathcal{O}(n^2a^2)$, space $\mathcal{O}(na)$

Implementation hack: domino effect, saves computations.

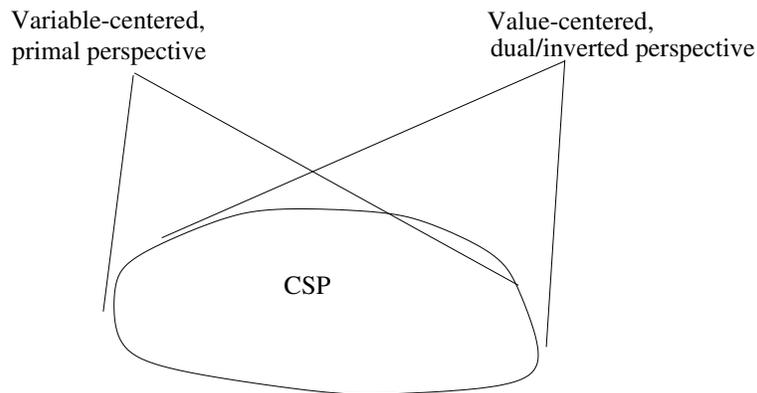
II- Permutation problems

→ $\left\{ \begin{array}{l} n = a \\ \text{constraint graph is complete} \\ \text{constraints are MUTEX, All-diffs} \end{array} \right.$

→ matching, marriage problem

find one value for every variable and
exactly one variable for every value

Viewpoints: variable vs. values



→ Which viewpoint to take?

→ How to combine computations in viewpoints?

Permutation problems (cont'd)

At any point in BT,

$$\# \text{ future variables} = \# \text{ future values}$$

→ choose the most constrained variable in the var-viewpoint, except when the most constrained value in the val-viewpoint is more constrained

PP: Least domain

| | | | | |
|----|---|---|---|---|
| X1 | | | 2 | 1 |
| X2 | ● | | | |
| X3 | | | 0 | 1 |
| X4 | | 2 | | 0 |

PP: Full evaluation

Introduce: $\begin{cases} \text{LEFT}^{inv}(V_i | X_i = V_i) \\ \text{Promise}^{inv}(V_i = X_i) \end{cases}$

Combine viewpoints: $\text{CPromise}(X_i = V_i) \min. \begin{cases} \text{Promise}(X_i = V_i) \\ \text{Promise}^{inv}(V_i = X_i) \end{cases}$

Evaluate vars/vals using: $\begin{cases} \text{CPromise}(X_i) = \sum_k \text{CPromise}(X_i = V_k) \\ \text{CPromise}^{inv}(V_i) = \sum_k \text{CPromise}(X_k = V_i) \end{cases}$

Partial permutation problems ($n \leq a$)

- fake a real PP with bogus variables
- extend (3) for PPP

III- Experiments

- 100 N -queen problems: $4 \leq N \leq 103$
- Comparison criteria
 - average number of backtrack
 - number of backtrack-free solutions
 - maximum number of backtracks
 - number of constraint checks

Algorithms: $\left\{ \begin{array}{l} \text{LD-1, LD-2, LD-3, LD-1+Dual} \\ \text{FE-2-4, FE-3-5, FE-3-5+Dual} \\ \text{FP-2-4, FP-3-5} \end{array} \right.$

| Algorithm | Average #backtrack | #backtrack-free solutions | Max. #backtrack |
|-----------------------|-----------------------|------------------------------|--------------------|
| LD+formula 1 | >45000 | 20 | >2500000 |
| LD+formula 2 | >33000 | 15 | >2500000 |
| LD+formula 3 | >1205 | 3 | 92379 |
| LD+formula 1, dual | 21.6 | 26 | 548 |
| FE+formulae 2&4 | 9.5 | 71 | 812 |
| FE+formulae 3&5 | 5.1 | 68 | 266 |
| FP+formulae 2&4 | 5.6 | 81 | 496 |
| FP+formulae 3&5 | 4.2 | 68 | 224 |
| FE+formulae 3&5, dual | 0.38 | 90 | 12 |

Dual perspective exhibits dramatic improvements

CPU time and #CC: plot hard to read, not interpreted:

- Number of #CC is prohibitive in practice
- Computations do FC implicitly.
- Can be exploited by bookkeeping.