

Topic 13: Consistency ¹

(Version of 20th November 2020)

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Course 1DL441:
Combinatorial Optimisation and Constraint Programming,
whose part 1 is Course 1DL451:
Modelling for Combinatorial Optimisation

¹Based partly on material by Christian Schulte and Yves Deville



Outline

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

1. Definitions

2. Value Consistency

3. Domain Consistency

4. Bounds Consistency

5. Consistency and Backtracking



Outline

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

1. Definitions

2. Value Consistency

3. Domain Consistency

4. Bounds Consistency

5. Consistency and Backtracking



Constraint Problems

Definitions

A **constraint satisfaction problem (CSP)** is $\langle V, D, C \rangle$ where:

- $V = [v_1, \dots, v_m]$ is a finite sequence of variables, which are often called decision variables.
- $D = [D_1, \dots, D_m]$ is a finite sequence of **domains**: the set of possible values for v_i is D_i , for all $i \in 1..m$.
- $C = \{c_1, \dots, c_p\}$ is a finite set of constraints on the variables. A **constraint** $\gamma(v_{i_1}, \dots, v_{i_q})$ is a relation, of **arity** q . We assume $i_j = j$, without loss of generality.

A **constrained optimisation problem (COP)** is $\langle V, D, C, f \rangle$:

- The triple $\langle V, D, C \rangle$ is a CSP, as above.
- $f(v_1, \dots, v_m)$ is a function from $D_1 \times \dots \times D_m$ to \mathbb{R} or \mathbb{Z} , called the **objective function**, the **objective** here being to minimise it, without loss of generality.

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking



More on problems:

Definitions

Value Consistency

Domain Consistency

Bounds Consistency

Consistency and Backtracking

- Without loss of generality, we often simplify notation by here requiring that all variables initially have the same domain U , called the **universe**: $D_1 = \dots = D_m = U$. We then refer to a triple $\langle V, U, C \rangle$ as a CSP, and to a quadruple $\langle V, U, C, f \rangle$ as a COP.
- We here focus on **discrete finite** domains, and thus also refer to a CSP or COP as a **combinatorial problem**.
- We distinguish a problem from its **instances**, defined by **instance data**. **Example:** n -Queens vs 8-Queens.



Stores and Solutions

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

Definition

The **store** of a CP solver with systematic search maps each decision variable v of a CSP or COP to its **current** domain, which is initialised to the domain of v in the CSP or COP.

Example

The function $\{x \mapsto \{1, 7\}, y \mapsto \{2, 5\}\}$ is a store, where the current domain $\{1, 7\}$ of x is $\{x \mapsto \{1, 7\}, y \mapsto \{2, 5\}\}(x)$.

Definitions

A variable v is **fixed** in store s iff its domain is a singleton: $|s(v)| = 1$. A store s is **fixed** iff all its variables are fixed in s .

Notation:

If the name, say s , of the store is irrelevant, then we denote the current domain $s(v)$ of a decision variable v by $\text{dom}(v)$.



Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

Definition

A store s is a **solution store to a constraint** $c = \gamma(v_1, \dots, v_q)$ if and only if s is fixed and denotes a solution to c : we have $s(v_i) = \{d_i\}$ for all $i \in 1..q$, and $\langle d_1, \dots, d_q \rangle \in c$.

Example

The store $\{x \mapsto \{3\}, y \mapsto \{4\}\}$ is a solution store to $x \leq y$.

Definition

A **solution** $\langle d_1, \dots, d_q \rangle$ to a constraint $\gamma(v_1, \dots, v_q)$ **is in** (also denoted \in) **a store** s iff every value belongs to the domain of the corresponding variable: $d_i \in s(v_i)$ for all $i \in 1..q$.

Example

The solution $\langle 3, 4 \rangle$ to the constraint $x \leq y$ is in the store $\{x \mapsto \{1, 3\}, y \mapsto \{2, 4\}, z \mapsto \{5, 6\}\}$.



The purpose of a solver with systematic search is to find eventually and provably each (provably optimal) solution:

Definitions

For variables $V = [v_1, \dots, v_m]$, domains $D = [D_1, \dots, D_m]$, and a store s such that $s(v_i) = \{d_i\}$ for all $i \in 1..m$:

- The store s is a **solution store to a CSP** $\langle V, D, C \rangle$ if and only if s is a solution store to each constraint in C and $d_i \in D_i$ for all $i \in 1..m$.
- The store s is a **solution store to a COP** $\langle V, D, C, f \rangle$ if and only if s is a solution store to the CSP $\langle V, D, C \rangle$ and the **objective value** of s , namely $f(d_1, \dots, d_m)$, is minimal, that is less than or equal to the objective value of every other solution store to $\langle V, D, C \rangle$.

Definition

A CP solver with systematic search is **correct** if and only if it finds each solution store to any CSP or COP exactly once.



Overview

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

Definition

A **consistency** is the targeted characterisation of the domain values kept in the store by a **propagator** for a constraint, but correctness of the solver must not depend on enforcing it.

We distinguish:

- Value consistency (VC) is quite weak.
- Domain consistency (DC) is very strong.
- Bounds consistency (BC) is between VC and DC.

We now discuss VC, DC, and various flavours of BC.



Outline

1. Definitions

2. Value Consistency

3. Domain Consistency

4. Bounds Consistency

5. Consistency and Backtracking



Value Consistency

Definitions

Value Consistency

Domain Consistency

Bounds Consistency

Consistency and Backtracking

Example (Value consistency for `distinct`)

If a variable becomes fixed, then its value does not appear in the domains of all the other variables of the constraint.

Consider `distinct`($[x, y, z]$):

- Store $s = \{x, y \mapsto \{1, 2\}, z \mapsto \{5\}\}$ **is** value consistent.
- Store $s = \{x, y, z \mapsto \{1, 2\}\}$ **is** value consistent, hence search **is** needed to show that there is no solution in s .
- Store $s = \{x, y \mapsto \{1, 2\}, z \mapsto \{1, 2, 5\}\}$ **is** value consistent, hence search **is** needed to show that there are two solutions in s , both with $z = 5$.

Enforcing value consistency for `distinct`($[v_1, \dots, v_q]$) is known as **naïve distinct**, and takes $\mathcal{O}(q)$ time:

- Store $\{w, x, y, z \mapsto \{1, 2, 5\}\}$ is contracted upon $w = 5$ to the store $\{w \mapsto \{5\}, x, y, z \mapsto \{1, 2\}\}$.



Enforcing value consistency, in general now:

To enforce value consistency for a constraint $\gamma(\dots)$:
whenever a decision variable becomes fixed,
any impossible values according to the semantics of $\gamma(\dots)$
are deleted from the domains of its other decision variables.

More about value consistency:

In the literature, value consistency (denoted below by VC)
is also known as forward-checking consistency (FCC).



Outline

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

1. Definitions

2. Value Consistency

3. Domain Consistency

4. Bounds Consistency

5. Consistency and Backtracking



Domain Consistency

Definition

A store s is **domain consistent** for a constraint $\gamma(\dots)$ iff for each decision variable v and each value in its domain $s(v)$, there exist values in the domains of the other variables such that all these values form a solution to $\gamma(\dots)$.

Example (Domain consistency & `distinct([x, y, z])`)


- Store $s = \{x, y, z \mapsto \{1, 2\}\}$ is domain **inconsistent**.
Store $s' = \{x, y, z \mapsto \emptyset\}$ **is** domain consistent, hence **no** search is needed to show that there is no solution in s' .
- $\{x, y \mapsto \{1, 2\}, z \mapsto \{1, 2, 5\}\}$ is domain **inconsistent**.
 $\{x, y \mapsto \{1, 2\}, z \mapsto \{5\}\}$ **is** domain consistent, so **no** search is needed to show that $z = 5$ in all solutions.

☞ See `distinct` propagator in Topic 16: Propagators.




Example (Domain consistency for $x \neq y, y \neq z, z \neq x$)

- $\{x, y, z \mapsto \{1, 2\}\}$ **is** domain consistent for all three constraints, hence search **is** needed to show that there is no solution in this store.
- $\{x, y \mapsto \{1, 2\}, z \mapsto \{1, 2, 5\}\}$ **is** domain consistent, hence search **is** needed to show $z = 5$ in all solutions.

Decomposing constraint $\text{distinct}([v_1, \dots, v_q])$ into $\frac{q \cdot (q-1)}{2}$ constraints $v_i \neq v_j$ ($1 \leq i < j \leq q$) yields **VC** for distinct and requires $\mathcal{O}(q^2)$ space:  see Topic 16: Propagators.

Example (Domain consistency for $x = 3 \cdot y + 5 \cdot z$)

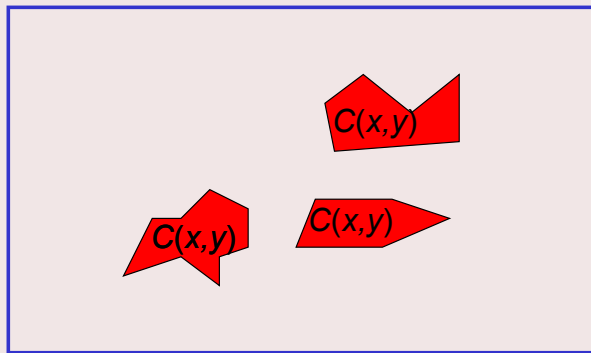
- Only the solutions $\langle 3, 1, 0 \rangle$, $\langle 5, 0, 1 \rangle$, and $\langle 6, 2, 0 \rangle$ are in $\{x \mapsto \{2, \dots, 7\}, y \mapsto \{0, 1, 2\}, z \mapsto \{-1, \dots, 2\}\}$.
- Hence $\{x \mapsto \{3, 5, 6\}, y \mapsto \{0, 1, 2\}, z \mapsto \{0, 1\}\}$ is domain consistent, but has $3 \cdot 3 \cdot 2 - 3$ non-solutions!

 CP = reasoning with sets of (at least all) **possible** values!



Geometric intuition (pictures: © Yves Deville)

$\text{dom}(y)$

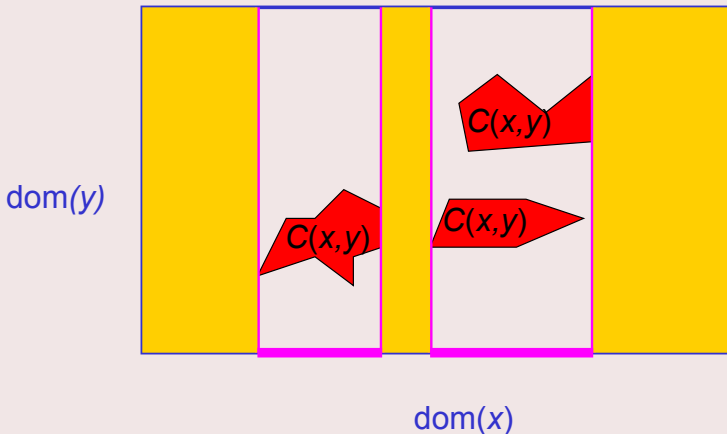


$\text{dom}(x)$

In general, a domain is a union of intervals.



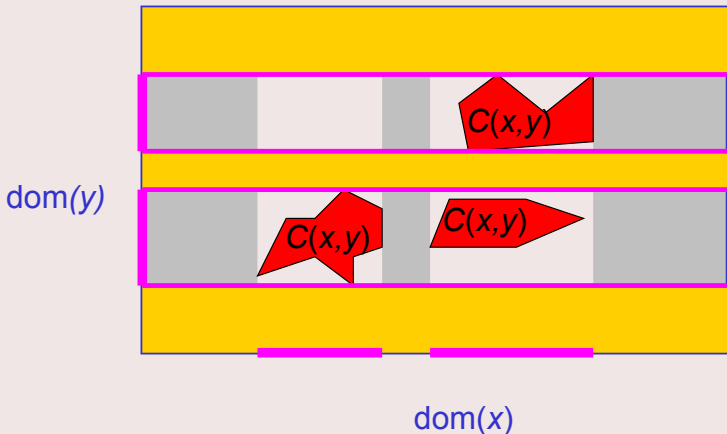
Geometric intuition (pictures: © Yves Deville)



Contracting the domain of x from 1 interval into 2 intervals



Geometric intuition (pictures: © Yves Deville)



Contracting the domain of y from 1 interval into 2 intervals



More about domain consistency:

- In the literature, domain consistency (denoted by **DC**) is also known as **generalised arc consistency (GAC)** or **hyper-arc consistency (HAC)**, and **arc consistency (AC)** in the case of binary constraints (of arity 2).
- DC is the strongest consistency, and thus implies VC for instance, but enforcing it is sometimes prohibitively expensive, for instance on linear equality constraints.
- Naïve ways to enforce DC for a constraint c are:
 - Execute the DC definition as an algorithm with 3 nested loops to find values staying in the variable domains of c .
 - Compute all solutions to c and lose them by projection onto each variable domain: see example at slide 15.

Both are impractical! 🙅 It is often possible to exploit the combinatorial structure of a constraint to enforce DC much faster: examples are in Topic 16: Propagators.



Outline

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

1. Definitions

2. Value Consistency

3. Domain Consistency

4. Bounds Consistency

5. Consistency and Backtracking



Bounds Consistency

Example (Consistency for $2 \cdot x = y$)

Consider the store $s = \{x \mapsto \{1, 2, 6\}, y \mapsto \{0, 2, 3, 4, 5\}\}$:

- Enforcing DC contracts s to $\{x \mapsto \{1, 2\}, y \mapsto \{2, 4\}\}$.
- But Gecode contracts s to $\{x \mapsto \{1, 2\}, y \mapsto \{2, \textcolor{red}{3}, 4\}\}$!

Definitions

A store s is **bounds(\mathbb{Z}) consistent** for a constraint $\gamma(\dots)$ iff for each decision variable v and the lower & upper **bounds** of its domain $s(v)$, there exist values **between the bounds** inclusive of the domains of the other variables such that all these values form an **integer** solution to $\gamma(\dots)$.

Similarly for a store being **bounds(\mathbb{R}) consistent**.



Definition

A store s is **bounds(D) consistent** for a constraint $\gamma(\dots)$ iff for each decision variable v and the lower & upper **bounds** of its domain $s(v)$, there exist values **in the domains** of the other variables such that all values form a solution to $\gamma(\dots)$.

Note that bounds(D) is *not* a misspelling of $\text{bounds}(\mathbb{D})$.

Example (Bounds consistencies for $\max(x, y) = z$)

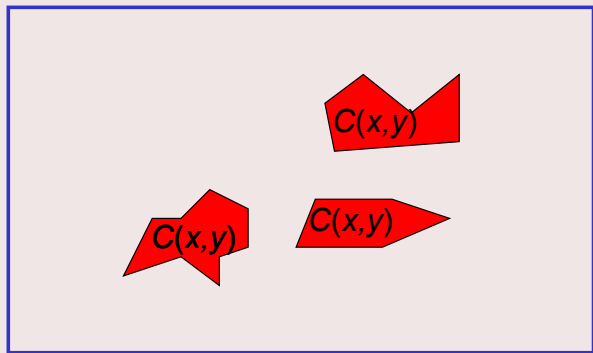
Consider $s = \{x \mapsto \{2, 3, 5\}, y \mapsto \{\mathbf{3}, 4, 6\}, z \mapsto \{4, 6\}\}$:

- Enforcing $\text{bounds}(\mathbb{Z})$ or $\text{bounds}(\mathbb{R})$ consistency leaves s unchanged.
- Enforcing bounds(D) consistency contracts s to $\{x \mapsto \{2, 3, 5\}, y, z \mapsto \{4, 6\}\}$.



Geometric intuition (pictures: © Yves Deville)

$\text{dom}(y)$

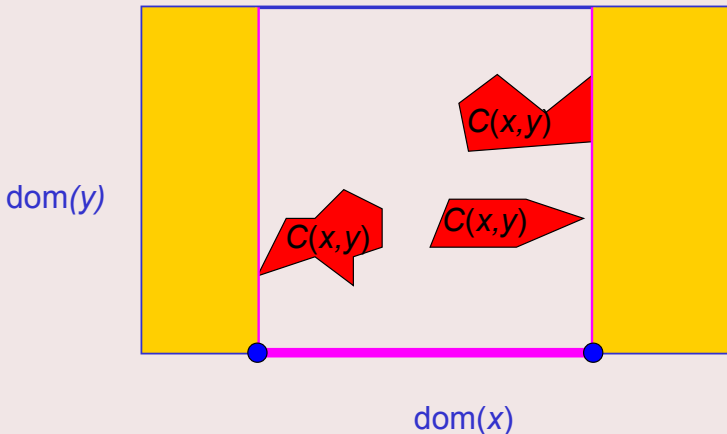


$\text{dom}(x)$

In general, a domain is a union of intervals.



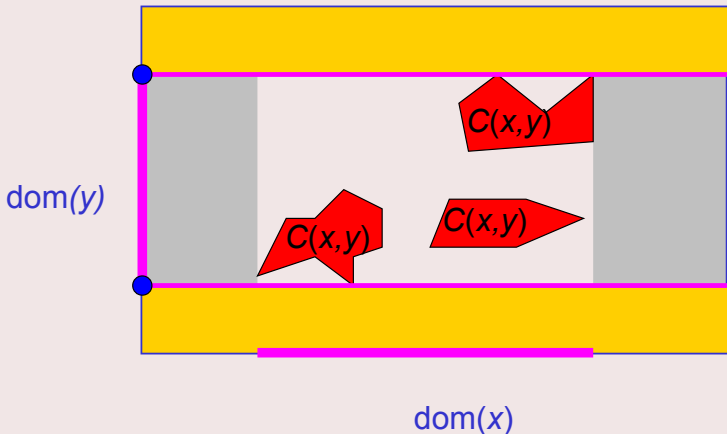
Geometric intuition (pictures: © Yves Deville)



Contracting the domain of x into a tighter interval



Geometric intuition (pictures: © Yves Deville)



Contracting the domain of y into a tighter interval



More about bounds consistencies:

In the literature, bounds(\mathbb{R}) consistency, denoted below by $BC(\mathbb{R})$, is also known as *interval consistency*. By default, Gecode enforces $BC(\mathbb{R})$ for arithmetic constraints. Note:

$$DC \Rightarrow BC(D) \Rightarrow VC$$

$$BC(D) \Rightarrow BC(\mathbb{Z}) \Rightarrow BC(\mathbb{R})$$

Example (Consistency for SEND + MORE = MONEY)

Enforcing DC for both `distinct(...)` and the linear equality suffices to solve the problem, **without** search! However, this is **not** faster than search interleaved with enforcing DC for `distinct(...)` and $BC(\mathbb{R})$ for the linear equality, as enforcing DC for linear equality is prohibitively expensive (see Topic 16: Propagators).



Outline

Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking

1. Definitions

2. Value Consistency

3. Domain Consistency

4. Bounds Consistency

5. Consistency and Backtracking



More about Consistency

Definitions

- The **existentially** quantified values in the definitions of DC and $BC(\cdot)$ are called **supports** (or **witnesses**).
- If at least one support exists for a considered value d of a **universally** quantified decision variable v in those definitions, then d is **supported**, else d is **unsupported**.

Definitions

The enforcing of some consistency for some constraint is called **propagation** and is performed by an algorithm called a **propagator**, which targets to delete the unsupported domain values according to that consistency.

- 👉 See in-depth discussion in Topic 14: Propagation.
- 👉 See example propagators in Topic 16: Propagators.



- *Not* all propagators enforce VC, BC(\cdot), or DC, which have simple definitions: there are many useful but unnamed consistencies that can be, and are, enforced. For example, DC can be enforced for *some* variables of the constraint, and BC(\cdot) for its *remaining* variables.
- Pragmatism often prevails in propagator design: maximally contract domains within a reasonable time and space complexity, typically *polynomial* in the number of decision variables of the constraint.
- A CP solver may offer *a few* propagators / consistencies for a constraint predicate, one being the default.
- The modeller *must* make experiments to choose for each constraint a suitable propagator / consistency, given typical instances of the problem at hand.
- A variable can be subjected to *several* consistencies in the constraints it participates in: the data structure for stores is *not* specific to a particular consistency.



Complexity of Consistencies

Preview of Topic 16: Propagators:

Example (`distinct([v_1, \dots, v_q])`)

- Value consistency: $\mathcal{O}(q)$ time
- Bounds consistency: $\mathcal{O}(q \cdot \lg q)$ time; often $\mathcal{O}(q)$ time
- Domain consistency: $\mathcal{O}(m \cdot \sqrt{q})$ time, $\mathcal{O}(m \cdot q)$ space,
for $m \geq q$ domain values

Example (Linear Arithmetic on q decision variables)

- Value consistency (useless): $\mathcal{O}(q)$ time
- Bounds consistency: $\mathcal{O}(q)$ time
- Domain consistency: exponential time (as NP-hard) for equality ($=$), but no higher time complexity than $\text{BC}(\mathbb{R})$ for disequality (\neq) and inequality ($<, \leq, \geq, >$)



n -Queens Revisited (pics: © Ch. Lecoutre)

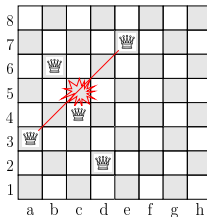
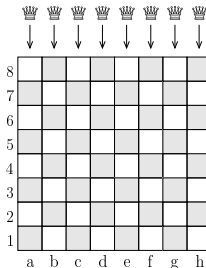
Definitions

Value
Consistency

Domain
Consistency

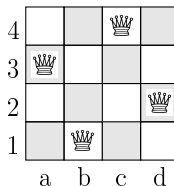
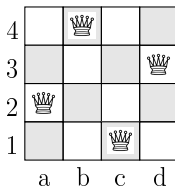
Bounds
Consistency

Consistency
and
Backtracking



$\text{distinct}([r_a, r_b, \dots, r_h]), \text{distinct}([|r_a-1|, |r_b-2|, \dots, |r_h-8|])$

The two solutions to the 4-queens instance:





4-Queens: Backtracking Search (BT)

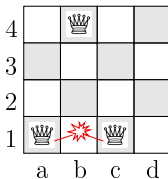
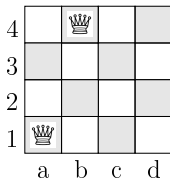
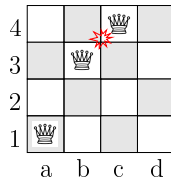
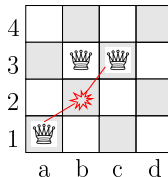
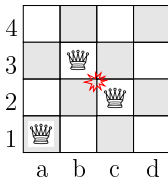
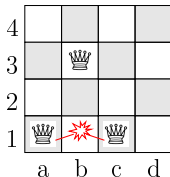
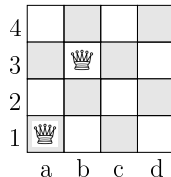
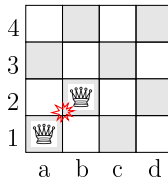
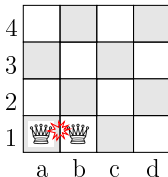
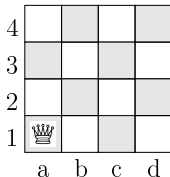
Definitions

Value
Consistency

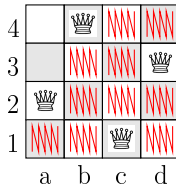
Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking



... 15 steps
omitted ...





4-Queens: BT + Value Consistency (VC)

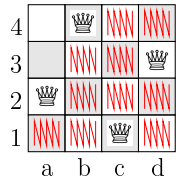
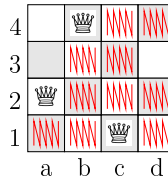
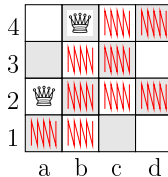
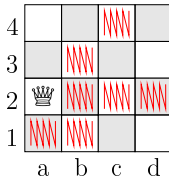
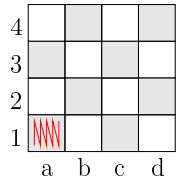
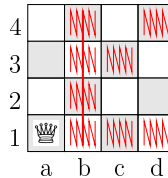
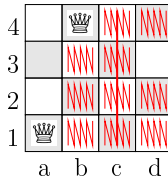
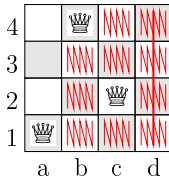
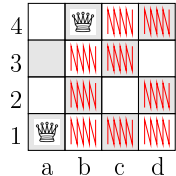
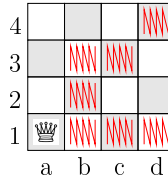
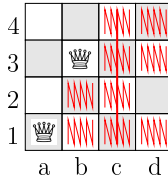
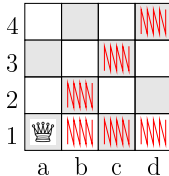
Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking





4-Queens: BT + Domain Consistency (DC)

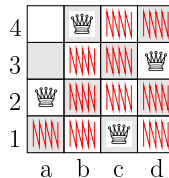
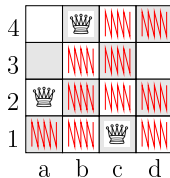
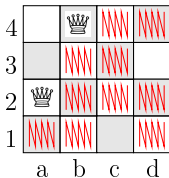
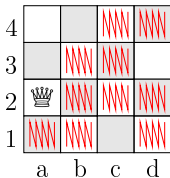
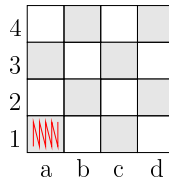
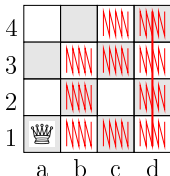
Definitions

Value
Consistency

Domain
Consistency

Bounds
Consistency

Consistency
and
Backtracking





4-Queens: BT + DC (versus BT + VC)

Why

| | | | | |
|---|---|---|---|---|
| 4 | | | | |
| 3 | | | | |
| 2 | | | | |
| 1 | | | | |
| | a | b | c | d |

under DC, versus

| | | | | |
|---|---|---|---|---|
| 4 | | | | |
| 3 | | | | |
| 2 | | | | |
| 1 | | | | |
| | a | b | c | d |

under VC?

Assume the **search guess** $r_a = 1$ is tried:

- 1 The $\text{distinct}([r_a, r_b, r_c, r_d])$ row constraint propagates to $\{r_a \mapsto \{1\}, r_b, r_c, r_d \mapsto \{2, 3, 4\}\}$, like under VC.
- 2 The $\text{distinct}([|r_a - 1|, |r_b - 2|, |r_c - 3|, |r_d - 4|])$ diagonal constraint **first** propagates, like under VC, to $\{r_a \mapsto \{1\}, r_b \mapsto \{3, 4\}, r_c \mapsto \{2, 4\}, r_d \mapsto \{2, 3\}\}$.
- 3 The previous propagator **also** notices that r_b cannot be 3 as the domain of r_c would then be wiped out; etc.
This would **not** happen with **two** diagonal constraints!

VC **only** detects the conflicts between the just fixed variable and the remaining variables, but DC **also** detects the conflicts **between** the remaining variables.

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