

Constraint Satisfaction Problem

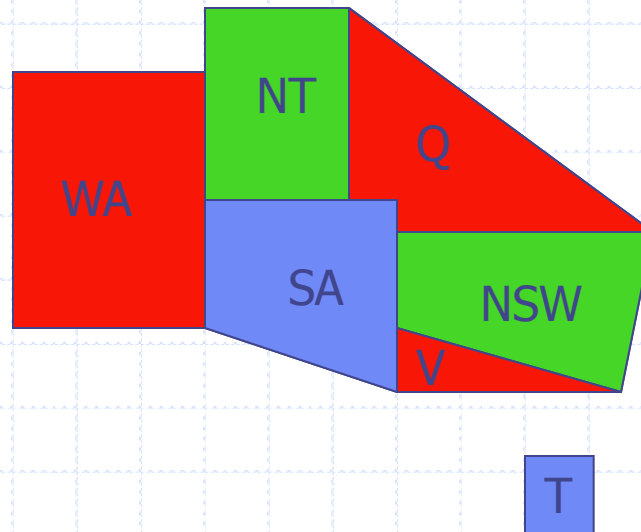
Constraint Satisfaction Problem

- ◆ Set of variables $\{X_1, X_2, \dots, X_n\}$
- ◆ Each variable X_i has a domain D_i (also denoted D_{X_i}) of possible values
 - Usually D_i is discrete and finite
- ◆ Set of constraints $\{C_1, C_2, \dots, C_p\}$
 - Each constraint C_k involves a subset of variables, e.g. we may write $C_k(X_i, \dots, X_j)$ to indicate the variables involved.
 - and specifies the allowable combinations of values of these variables
- ◆ Assign a value to every variable such that all constraints are satisfied

Example: 8-Queens Problem

- ◆ 8 variables X_i , $i = 1$ to 8
- ◆ Domain for each variable $\{1, 2, \dots, 8\}$
- ◆ Constraints are of the forms:
 - $X_i = k \rightarrow X_j \neq k$ for all $j = 1$ to 8 , $j \neq i$
 - $X_i = k_i, X_j = k_j \rightarrow |i - j| \neq |k_i - k_j|$
 - ◆ for all $j = 1$ to 8 , $j \neq i$

Example: Map Coloring



- 7 variables {WA, NT, SA, Q, NSW, V, T}
- Each variable has the same domain {red, green, blue}
- No two adjacent variables have the same value:

WA \neq NT, WA \neq SA, NT \neq SA, NT \neq Q, SA \neq Q, SA \neq NSW, SA \neq V, Q \neq NSW, NSW \neq V

CSP as a Search Problem

- ◆ **Initial state:** empty assignment
- ◆ **Successor function:** a value is assigned to any unassigned variable, which does not conflict with the currently assigned variables
- ◆ **Goal test:** the assignment is complete

Remark

- ◆ Finite CSP include 3SAT as a special case
- ◆ 3SAT is known to be NP-complete
- ◆ So, in the worst-case, we cannot expect to solve a finite CSP in less than exponential time

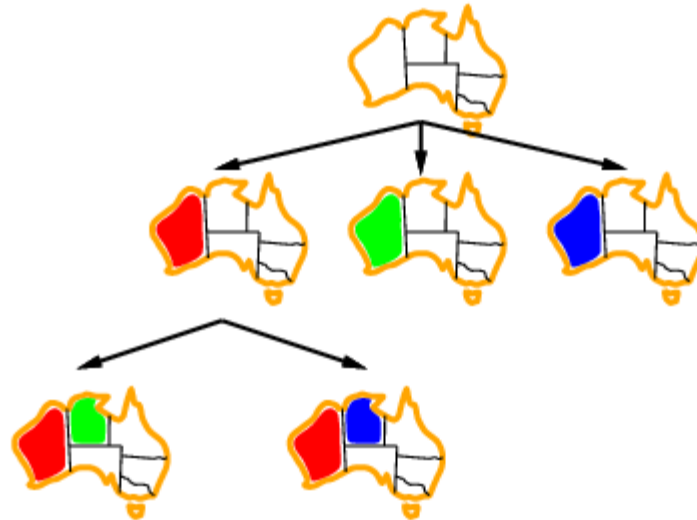
Backtracking example



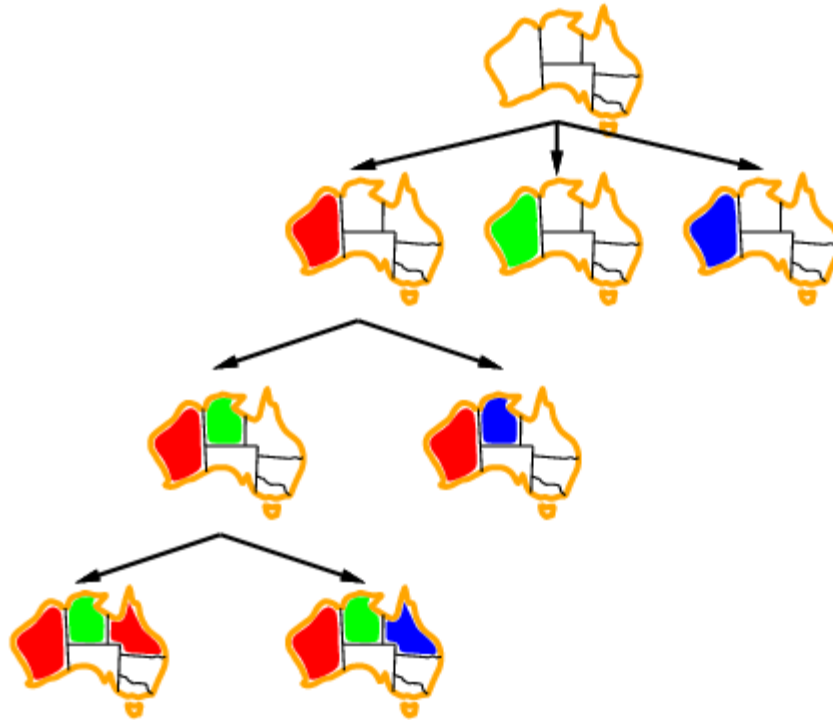
Backtracking example



Backtracking example



Backtracking illustration



Backtracking Algorithm

CSP-BACKTRACKING(PartialAssignment a)

- If a is complete then return a
- $X \leftarrow$ select an unassigned variable
- $D \leftarrow$ select an ordering for the domain of X
- For each value v in D do
 - ◆ If v is consistent with a then
 - Add ($X = v$) to a
 - $result \leftarrow$ CSP-BACKTRACKING(a)
 - If $result \neq failure$ then return $result$
 - Remove ($X = v$) from a
- Return *failure*

Start with CSP-BACKTRACKING($\{\}$)

Improving backtracking efficiency

- Which variable should be assigned next?
- In what order should its values be tried?
- Can we detect inevitable failure early?

Most constrained variable

- ◆ Most constrained variable:
choose the variable with the fewest legal values
- ◆ a.k.a. minimum remaining values (MRV) heuristic

Most constraining variable

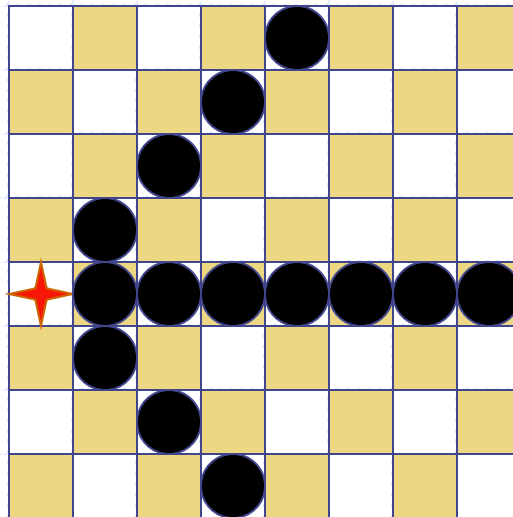
- ◆ Tie-breaker among most constrained variables
- ◆ Most constraining variable:
 - choose the variable involved in largest # of constraints on remaining variables

Least constraining value

- ◆ Given a variable, choose the least constraining value:
 - the one that rules out the fewest values in the remaining variables
- ◆ Combining these heuristics makes 1000 queens feasible

Forward Checking

After a variable **X** is assigned a value **v**, look at each unassigned variable **Y** that is connected to **X** by a constraint and deletes from **Y**'s domain any value that is inconsistent with **v**



Constraint propagation

- ◆ Forward checking propagates information from assigned to unassigned variables, but doesn't provide early detection for all failures:

Arc-consistency

A constraint $C(x,y)$ is said to be arc-consistent w.r.t. x if for each value v of x , there is an allowed value of y .

Similarly, we define that $C(x,y)$ is arc-consistent w.r.t. y .

A binary CSP is arc-consistent iff every constraint $C(x,y)$ is arc-consistent w.r.t. x as well as w.r.t. y .

Enforcing/Maintaining arc-consistency

Enforcing arc-consistency:

When a CSP is not arc-consistent, we can make it arc-consistent using an arc-consistency algorithm.

Maintaining arc-consistency:

During backtrack search, arc-consistency is maintained:

- every time when a value is assigned to a variable;
- every time a value is rejected (e.g. empty domain is generated by enforcing arc-consistency);

Of course, we can also perform arc-consistency as a pre-process.

Example

◆ Consider constraints:

$$X > Y, Y > Z$$

Domains:

$$D_x = D_y = D_z = \{1,2,3\}$$

1 in D_x is removed by enforcing arc consistency, w.r.t. the constraint $X < Y$.

You can work out the rest. The resulting domains are

$$D'_x = \{1\}, D'_y = \{2\}, D'_z = \{3\}$$

No search is needed in this case.

Solving a CSP

◆ Search:

- can find solutions, but must examine non-solutions along the way

◆ Constraint Propagation:

- Prune search space by domain reduction.

◆ Interleave **constraint propagation** and **backtrack search**

- Perform constraint propagation at each search step.

Summary

- ◆ Constraint Satisfaction Problems (CSP)
- ◆ CSP as a search problem
 - Backtracking algorithm
 - General heuristics
- ◆ Forward checking
- ◆ Constraint propagation: AC
- ◆ Interweaving CP and backtracking