

Constraint Satisfaction Problems

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- Constraint satisfaction problems (CSPs) are mathematical problems defined as a set of objects whose state must satisfy a number of constraints or limitations.

Constraint satisfaction problems

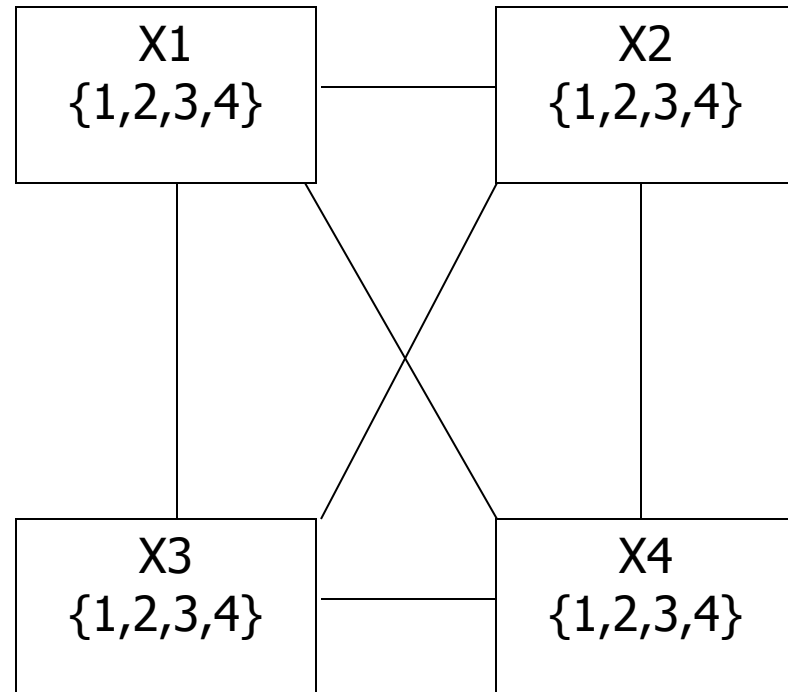
- What is a CSP?
 - **Finite set of variables** X_1, X_2, \dots, X_n
 - Where each variable X_i is a Nonempty domain of possible values D_1, D_2, \dots, D_d
 - **Finite set of constraints** C_1, C_2, \dots, C_m
 - Each constraint C_i limits the values that variables can take, e.g., $X_1 \neq X_2$
- A **state** is defined as an **assignment** of values to some or all variables.
- **Consistent assignment**: assignment does not violate the constraints.

Constraint satisfaction problems

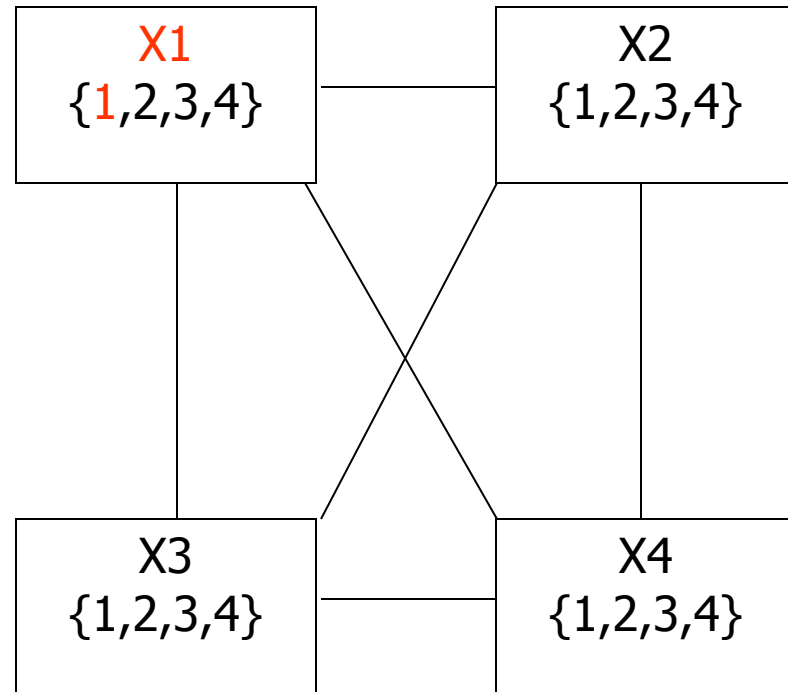
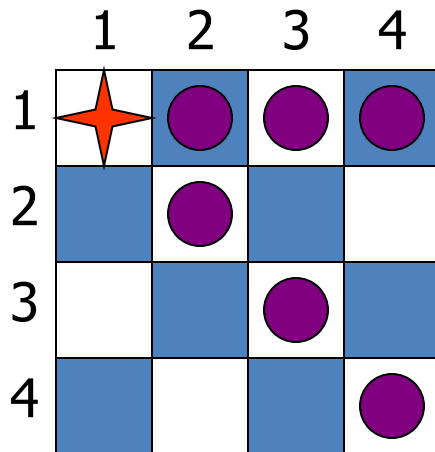
- An assignment is **complete** when every variable is assigned a value.
- A **solution** to a CSP is a complete assignment that satisfies all constraints.
- Applications:
 - Eight queens puzzle
 - Map coloring problem
 - Sudoku etc.....

Example: 4-Queens Problem








	1	2	3	4
1				
2				
3				
4				

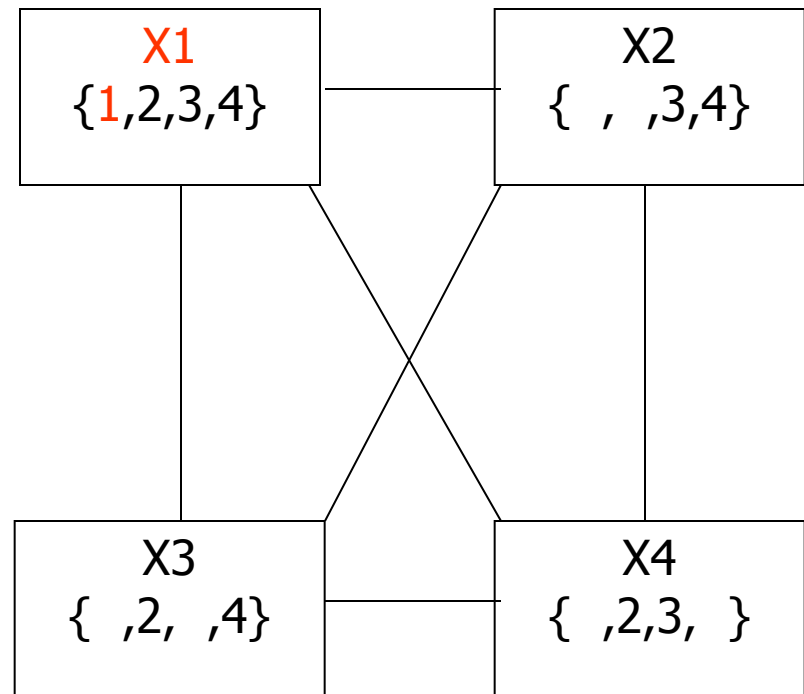


Example: 4-Queens Problem



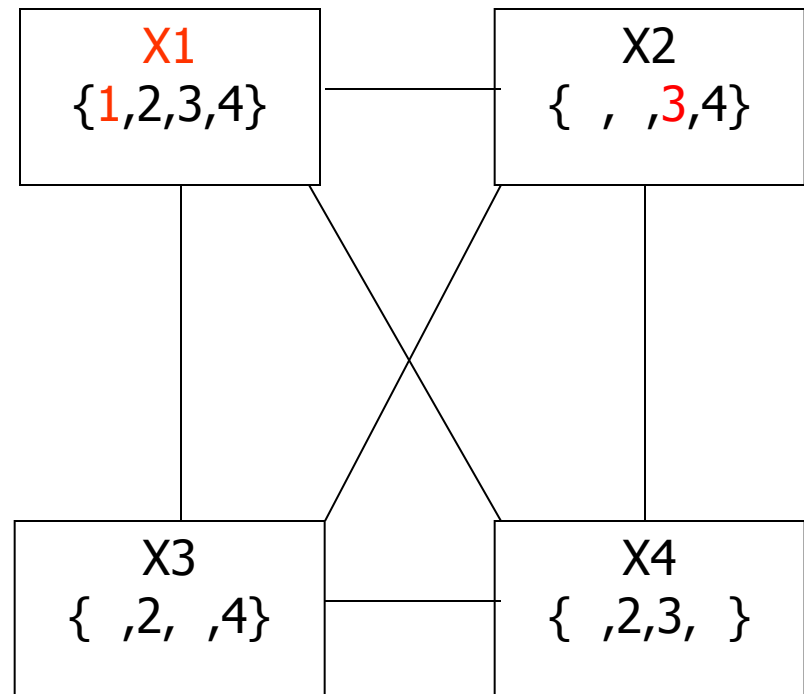
Example: 4-Queens Problem

	1	2	3	4
1				
2				
3				
4				



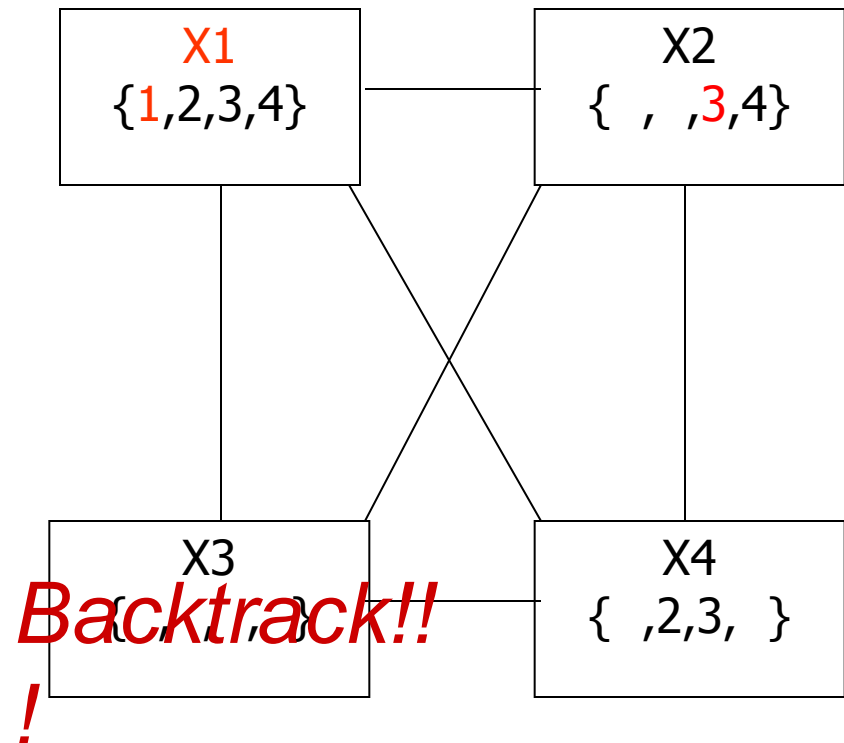
Example: 4-Queens Problem

	1	2	3	4
1	★	●	●	●
2		●	●	
3		★	●	●
4			●	●



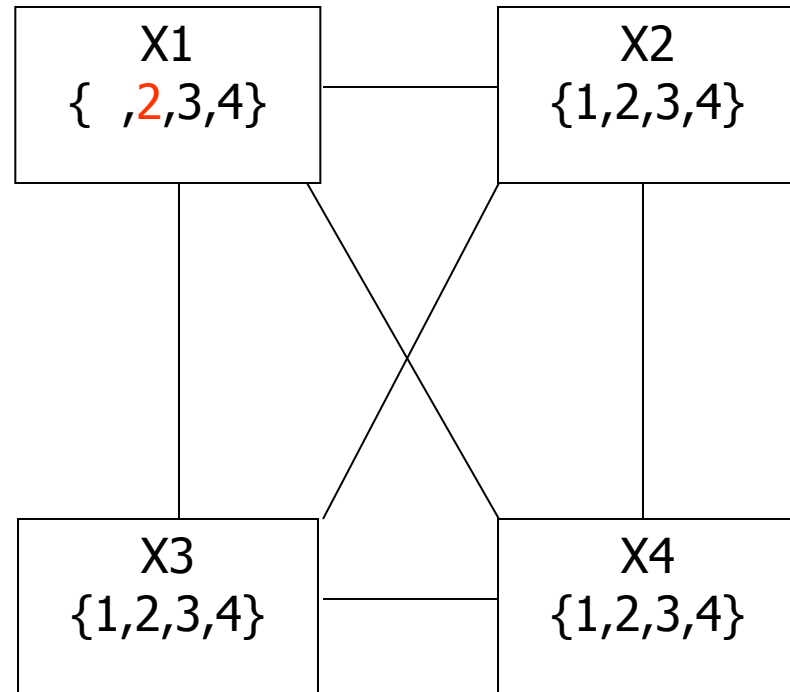
Example: 4-Queens Problem

	1	2	3	4
1	★	●	●	●
2	■	●	●	□
3	□	★	●	●
4	■	□	●	●



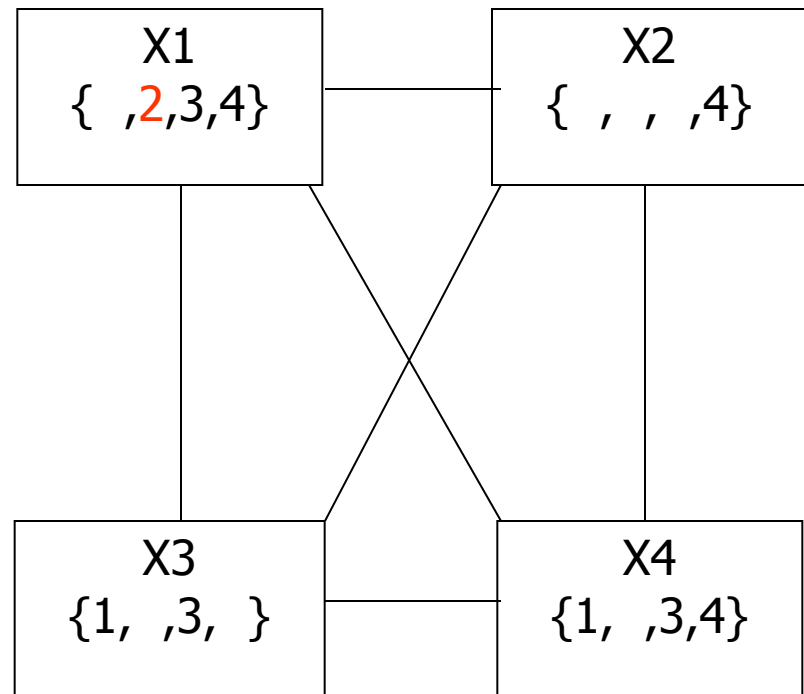
Example: 4-Queens Problem

	1	2	3	4
1		●		
2	★	●	●	●
3		●		
4			●	



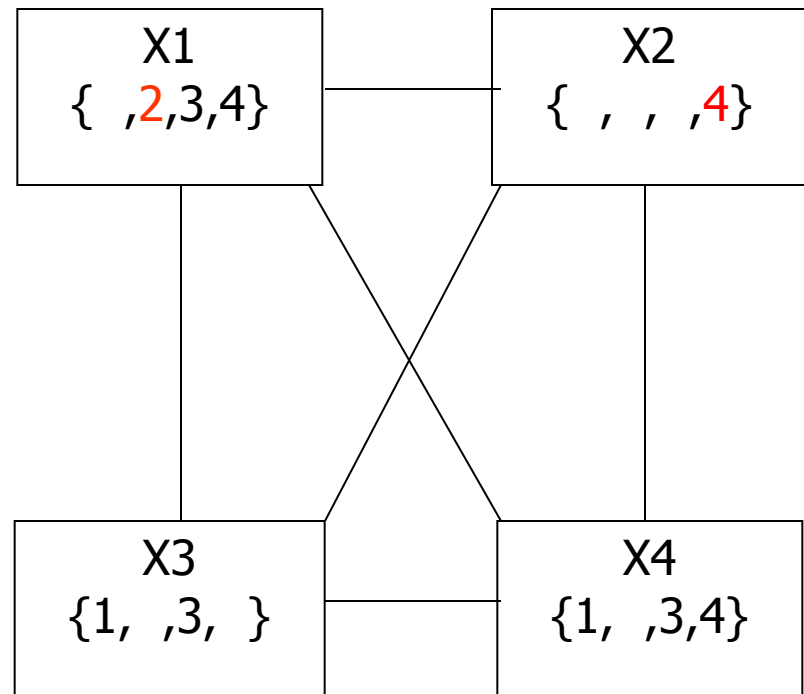
Example: 4-Queens Problem

	1	2	3	4
1		●		
2	★	●	●	●
3		●		
4			●	



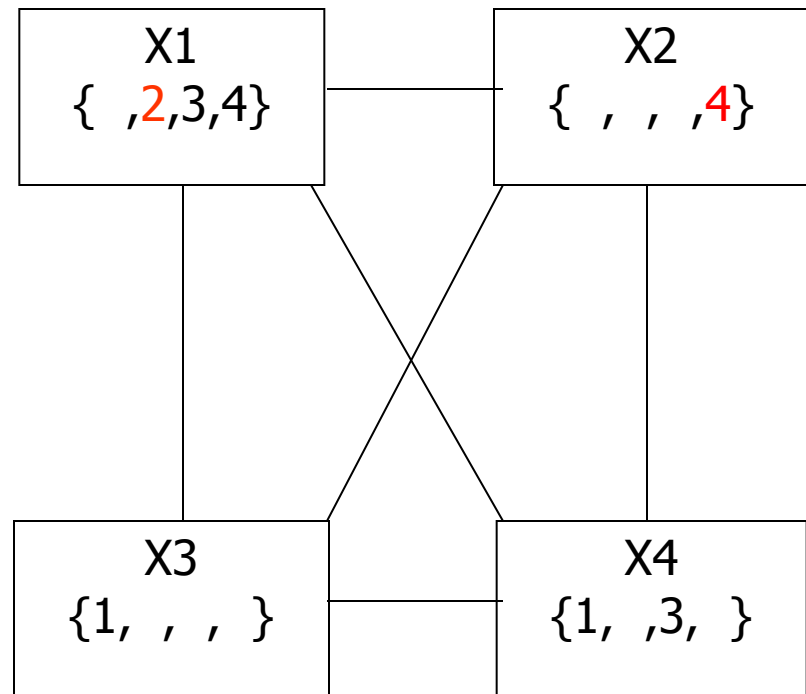
Example: 4-Queens Problem

	1	2	3	4
1		●		
2	★	●	●	●
3		●	●	
4		★	●	●



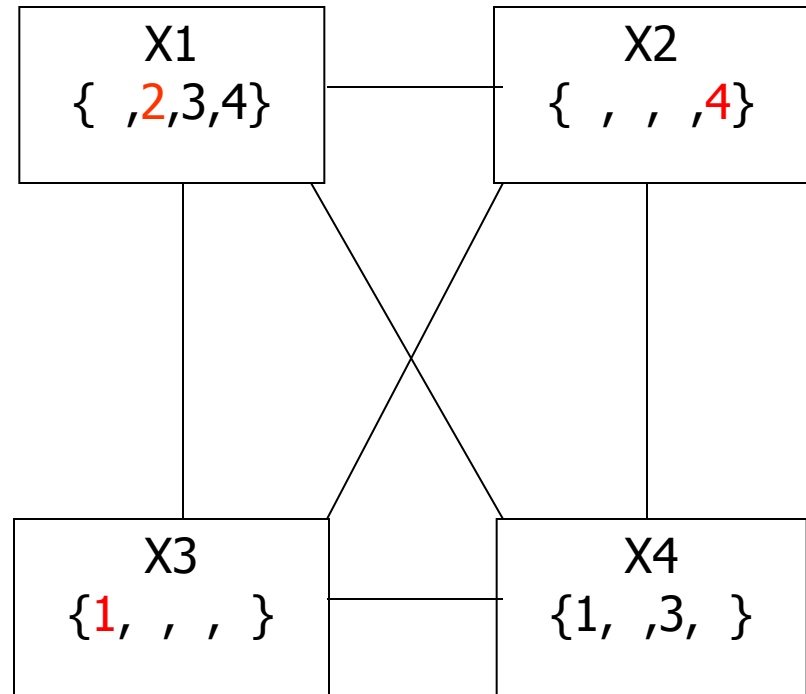
Example: 4-Queens Problem

	1	2	3	4
1		●		
2	★	●	●	●
3		●	●	
4		★	●	●



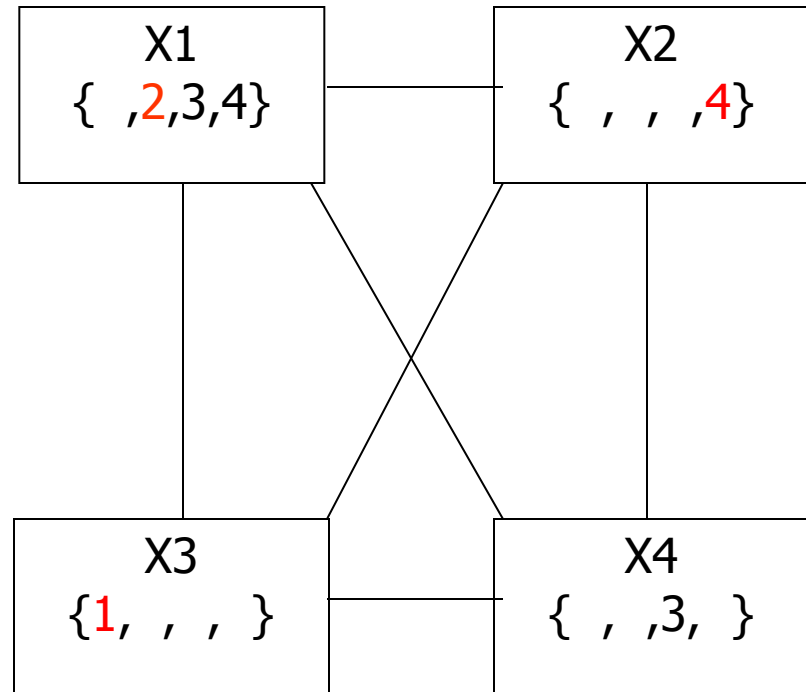
Example: 4-Queens Problem

	1	2	3	4
1		●	★	●
2	★	●	●	●
3		●	●	
4		★	●	●

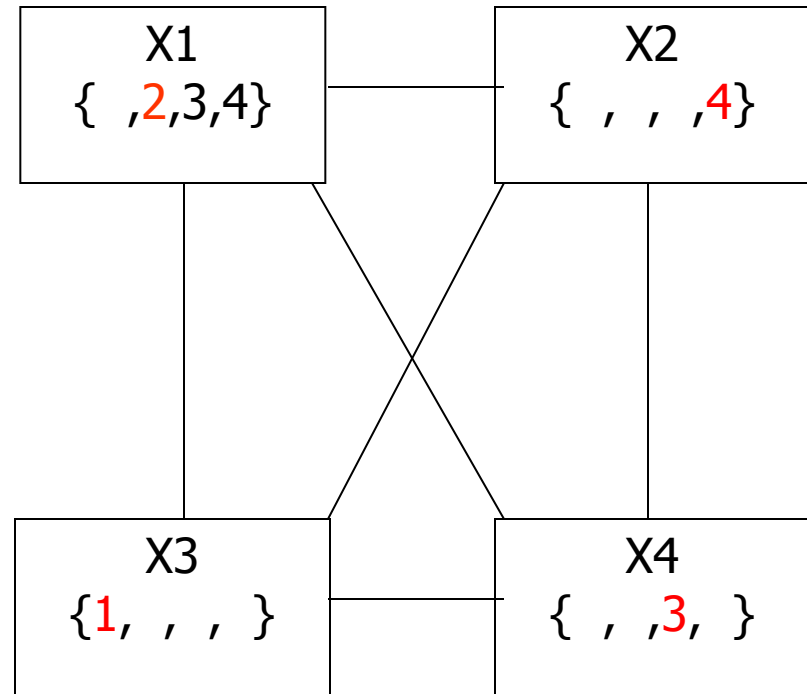
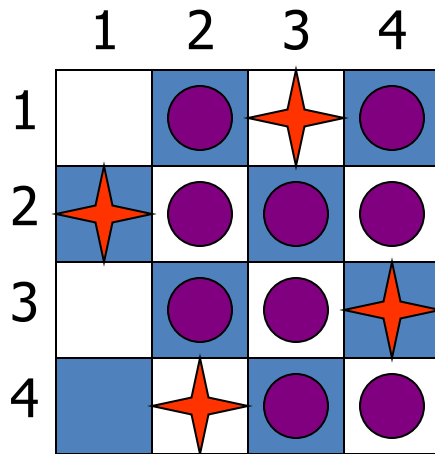


Example: 4-Queens Problem

	1	2	3	4
1		●	★	●
2	★	●	●	●
3		●	●	
4		★	●	●



Example: 4-Queens Problem



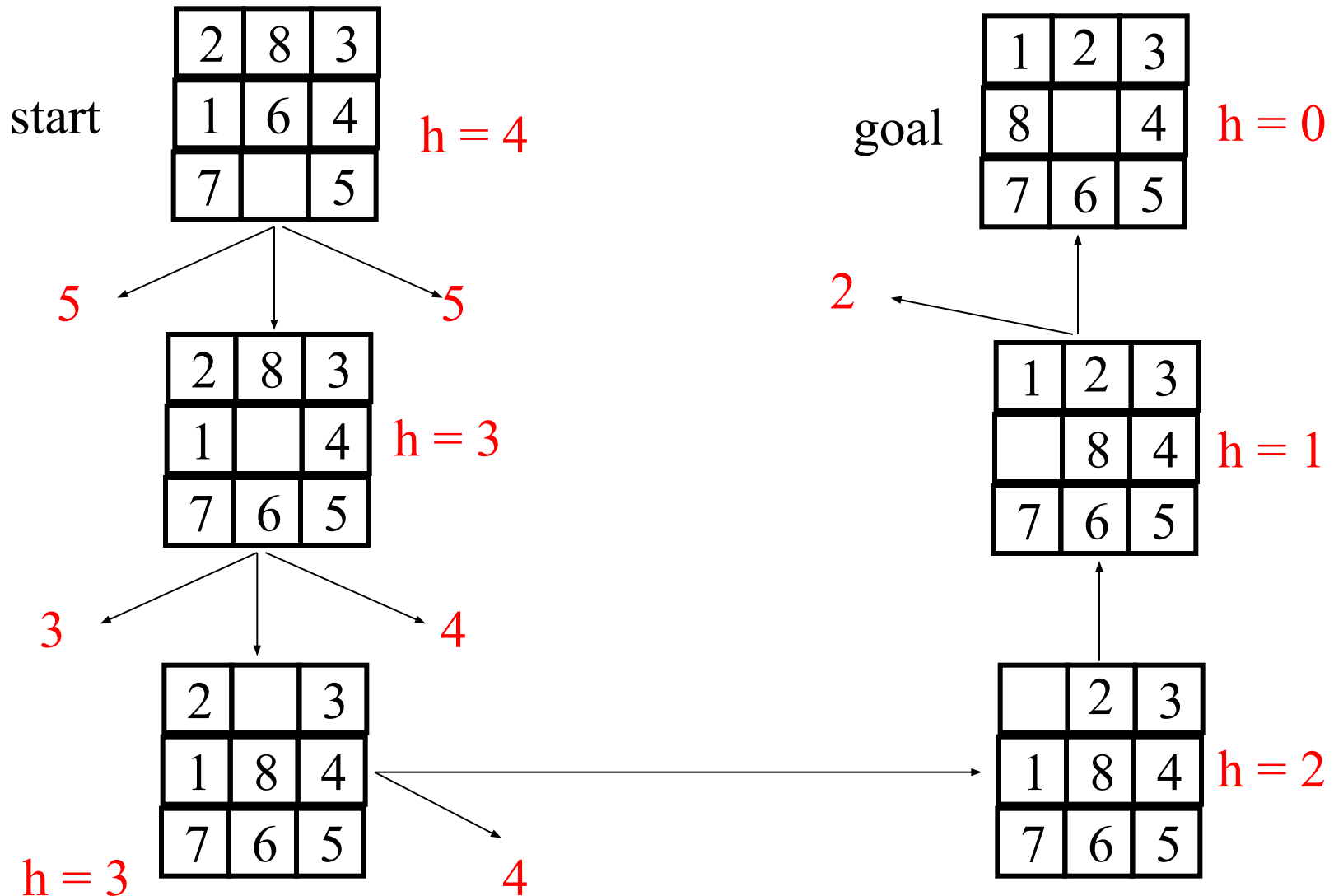
The Silver Bullet?

Is there an “intelligence algorithm”?

1957 GPS (General Problem Solver)

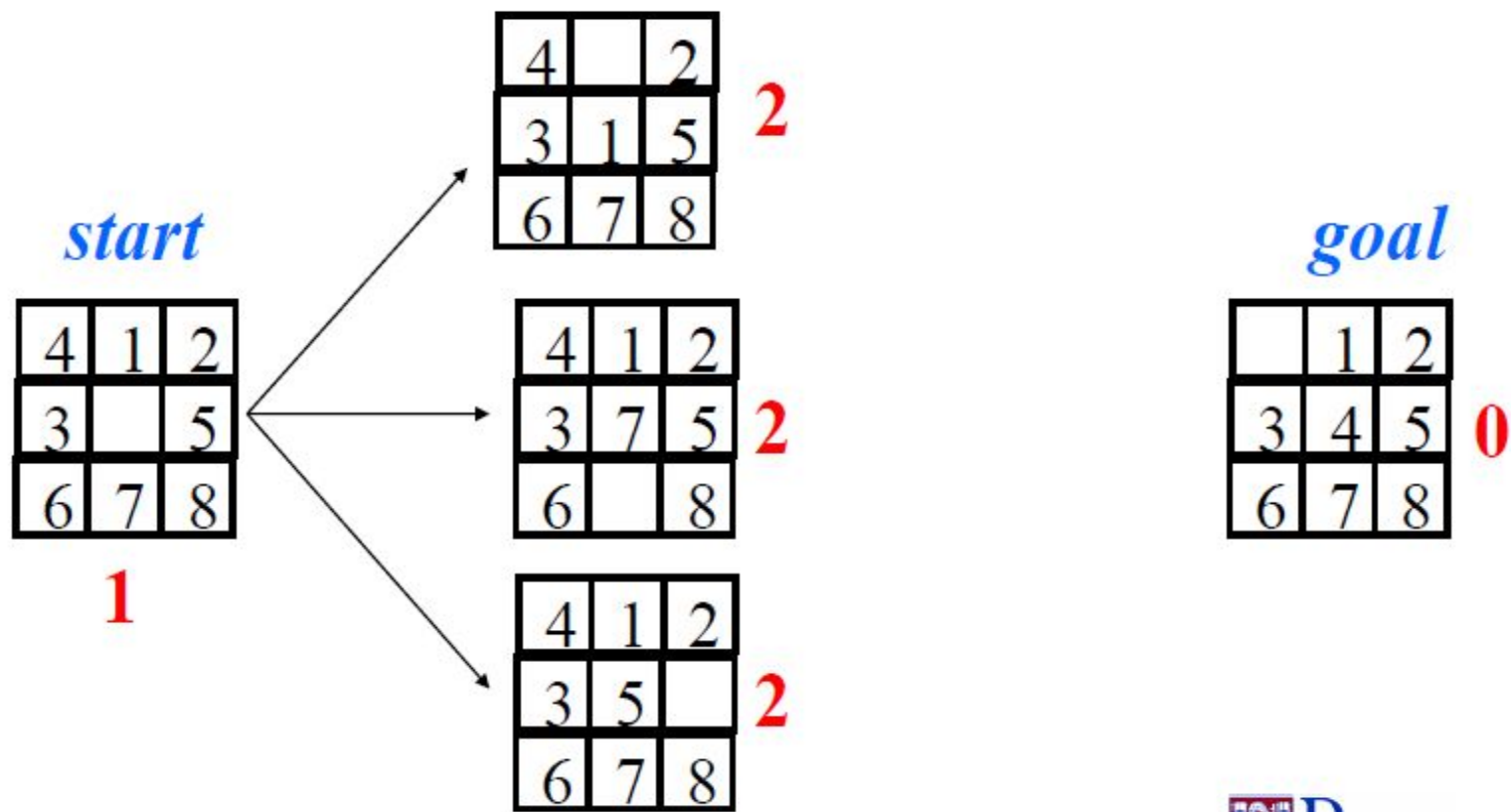


8-Puzzle example



$f(n) = (\text{number of tiles out of place})$

Toy Example of a local "maximum"



Varieties of CSPs

Discrete variables

finite domains:

n variables, domain size $d \Rightarrow O(d^n)$ complete assignments

e.g., Boolean CSPs, includes Boolean satisfiability (NP-complete) *Line Drawing Interpretation*

infinite domains:

integers, strings, etc.

e.g., job scheduling, variables are start/end days for each job need a constraint language, e.g., $StartJob_1 + 5 \leq StartJob_3$

Continuous variables

e.g., start/end times for Hubble Space Telescope observations linear constraints solvable in polynomial time by linear programming

Varieties of constraints

Unary constraints involve a single variable,
e.g., $SA \neq \text{green}$

Binary constraints involve pairs of variables,
e.g., $SA \neq WA$

Higher-order constraints involve 3 or more variables
e.g., crypt-arithmetic column constraints

Preference (soft constraints) e.g. *red is better than green* can be represented by a cost for each variable assignment

Constrained optimization problems.

Real-world CSPs

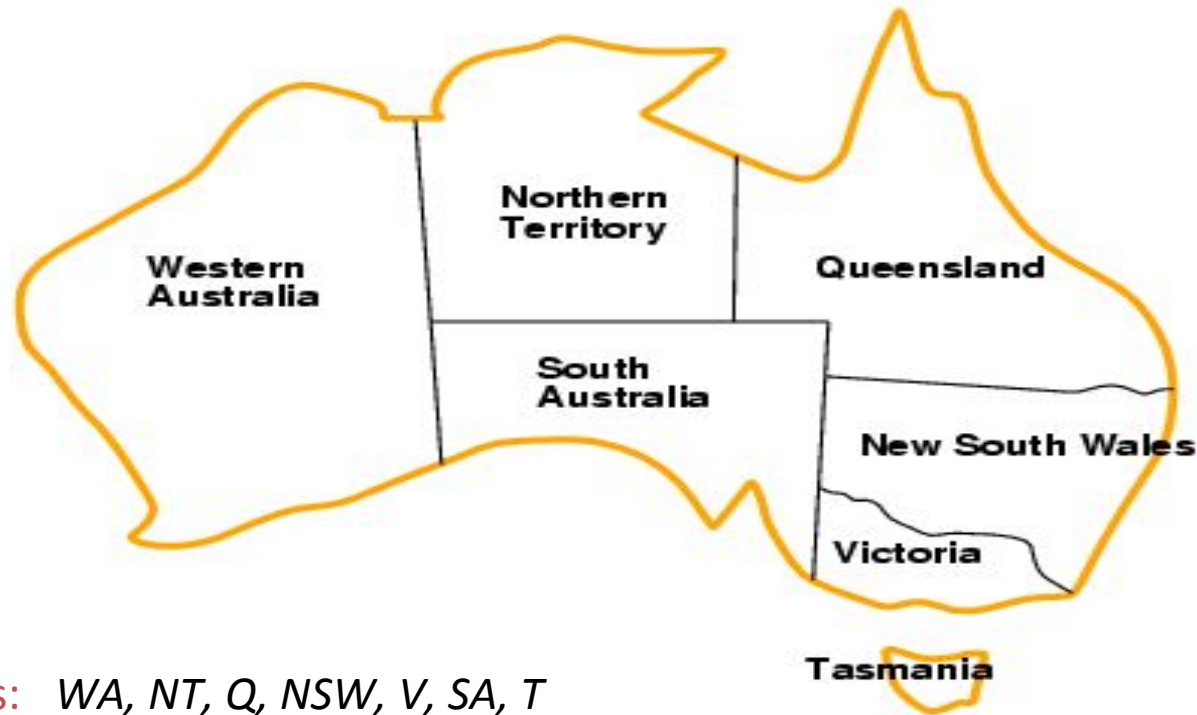
- Assignment problems
 - e.g., who teaches what class
- Timetabling problems
 - e.g., which class is offered when and where?
- Transportation scheduling
- Factory scheduling
- Notice that many real-world problems involve real-valued variables

Comparison of Methods

- **Backtracking** tree search is a blind search.
- **Forward checking** checks constraints between the current variable and all future ones.
- **Arc consistency** then checks constraints between all pairs of future (unassigned) variables.

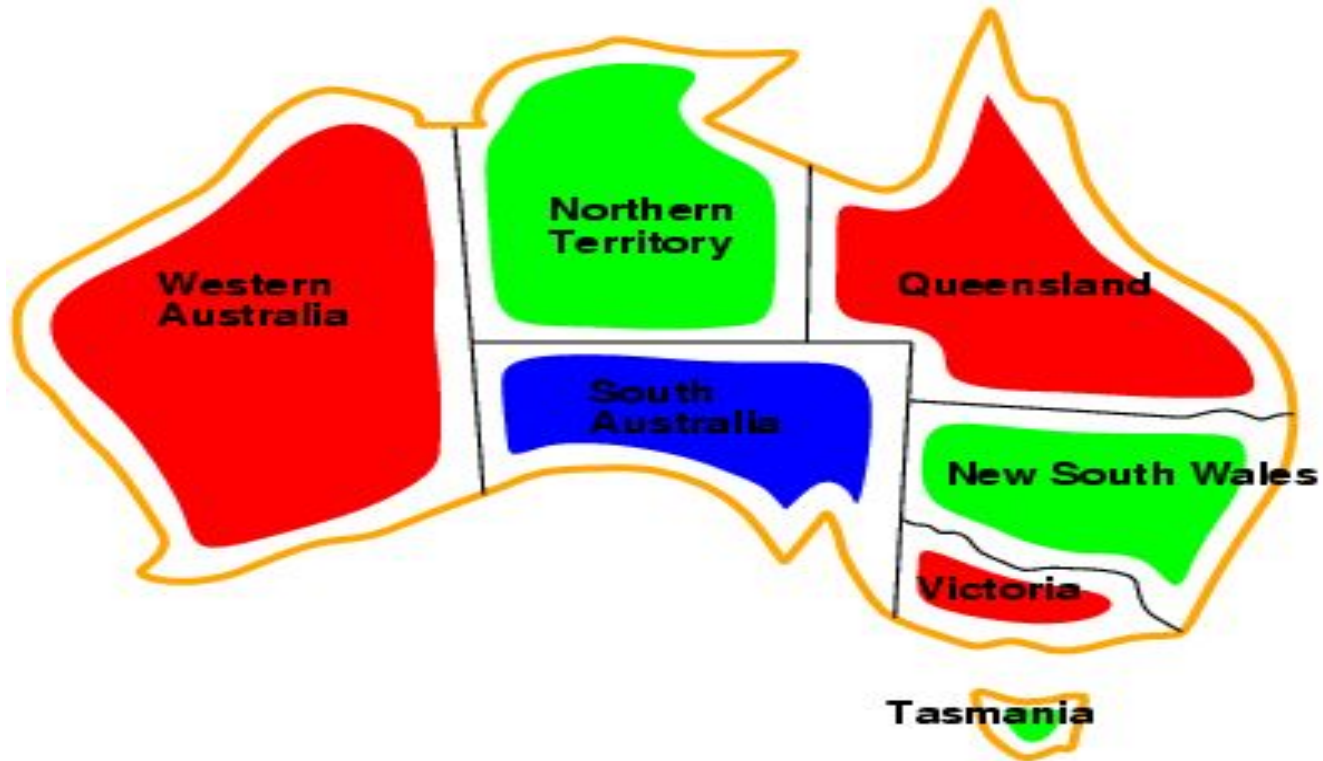
- What is the complexity of a backtracking tree search?
- How do forward checking and arc consistency affect that?

Example: Map-Coloring



- **Variables:** WA, NT, Q, NSW, V, SA, T
- **Domains:** $D_i = \{\text{red, green, blue}\}$
- **Constraints:** adjacent regions must have different colors
 - e.g., $WA \neq NT$ So (WA, NT) must be in $\{(\text{red, green}), (\text{red, blue}), (\text{green, red}), \dots\}$

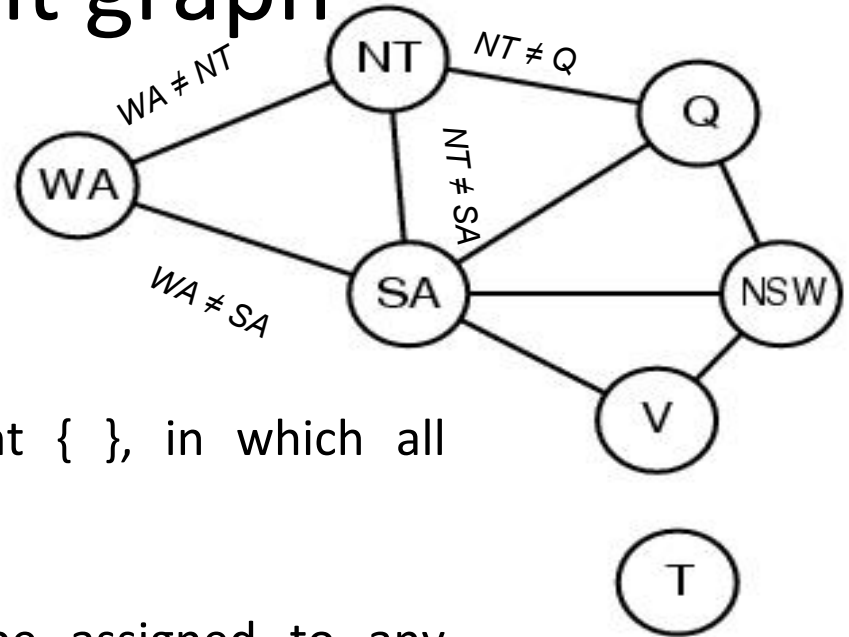
Example: Map-Coloring



Solutions are **complete** and **consistent** assignments,

- e.g., WA = red, NT = green, Q = red, NSW = green, V = red, SA = blue, T = green

Constraint graph



- **Constraint graph:**
 - **nodes** are variables
 - **arcs** are constraints
- **Initial state:** the empty assignment { }, in which all variables are unassigned.
- **Successor function:** a value can be assigned to any unassigned variable, provided that it does not conflict with previously assigned variables.
- **Goal test:** the current assignment is complete.
- **Path cost:** a constant cost (e.g., 1) for every step.

Backtracking search for CSPs

- Crucial property to all CSPs is Commutativity .(eg: $AB=BA$)
- The term backtracking search is used for a depth-first search that chooses values for BACKTRACKING SEARCH **one variable at a time** and backtracks when a variable has no legal values left to assign.
- The DFS algorithm is a recursive algorithm that uses the idea of backtracking. It involves exhaustive searches of all the nodes by going ahead, if possible, else by backtracking.

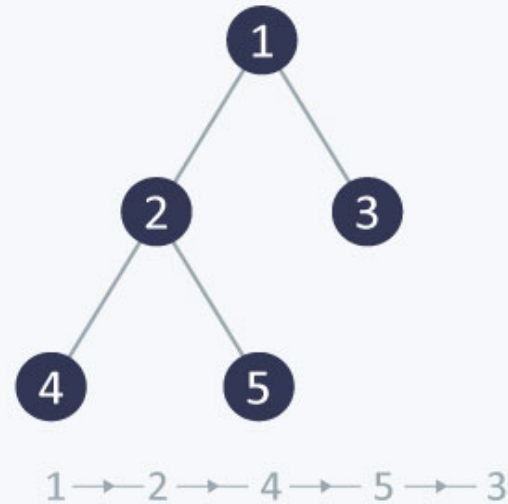
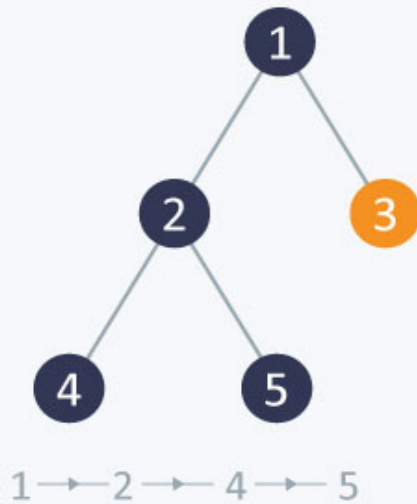
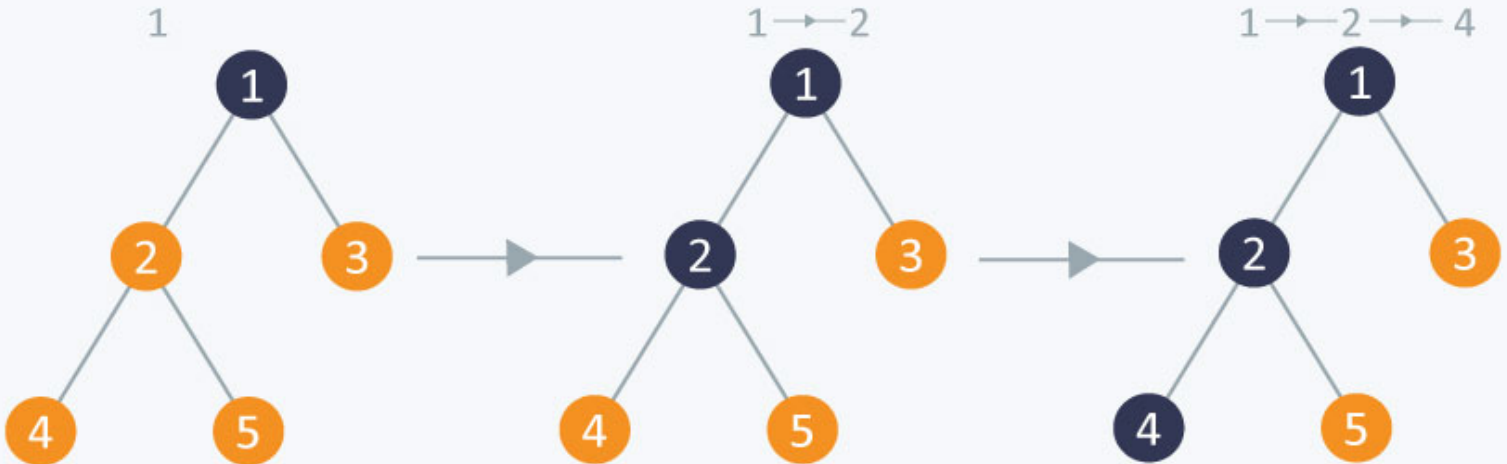
Backtracking search algorithm

function BACKTRACKING-SEARCH(*csp*) **returns** a solution, or failure
 return RECURSIVE-BACKTRACKING($\{ \}$, *csp*)

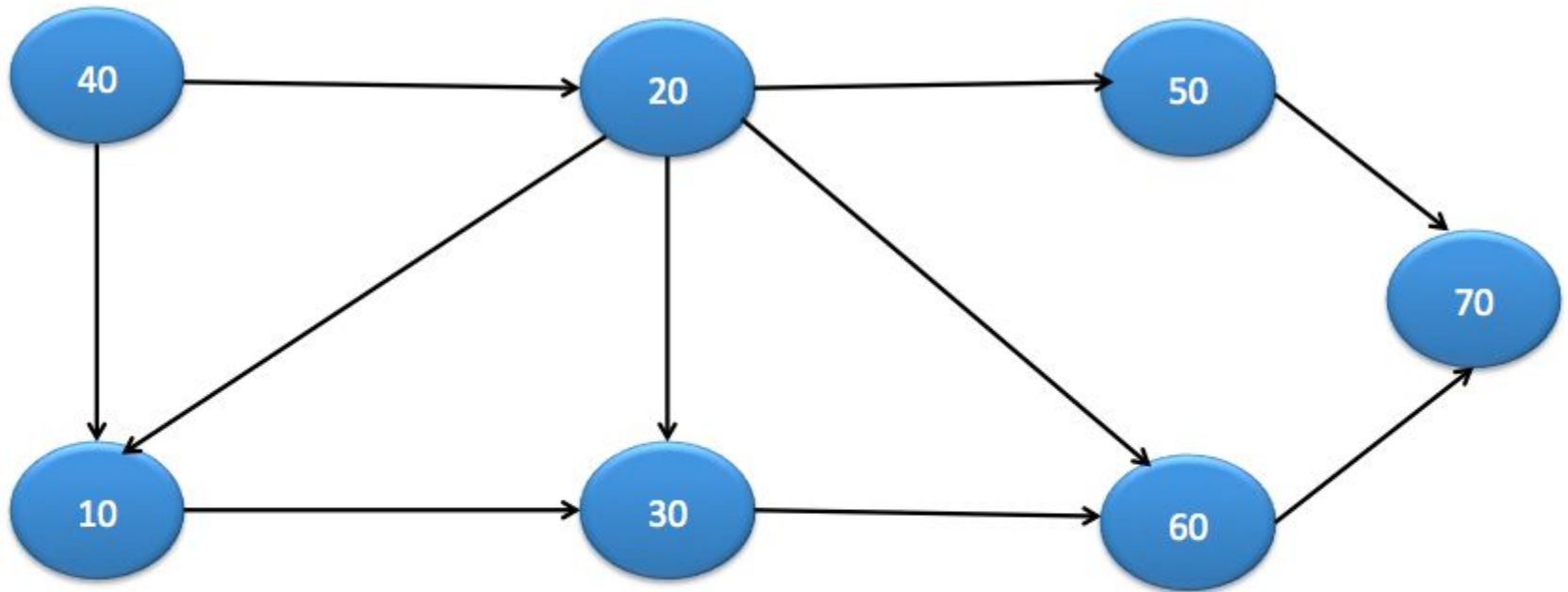
function RECURSIVE-BACKTRACKING(*assignment*, *csp*) **returns** a solution, or failure
 if *assignment* is complete **then return** *assignment*
 var \leftarrow SELECT-UNASSIGNED-VARIABLE(VARIABLES[*csp*], *assignment*, *csp*)
 for each *value* **in** ORDER-DOMAIN-VALUES(*var*, *assignment*, *csp*) **do**
 if *value* is consistent with *assignment* according to CONSTRAINTS[*csp*] **then**
 add $\{var = value\}$ to *assignment*
 result \leftarrow RECURSIVE-BACKTRACKING(*assignment*, *csp*)
 if *result* \neq failure **then return** *result*
 remove $\{var = value\}$ from *assignment*
 return failure

DFS

DFS



DFS



Depth first traversal of above graph can be :**40,20,50,70,60,30,10**

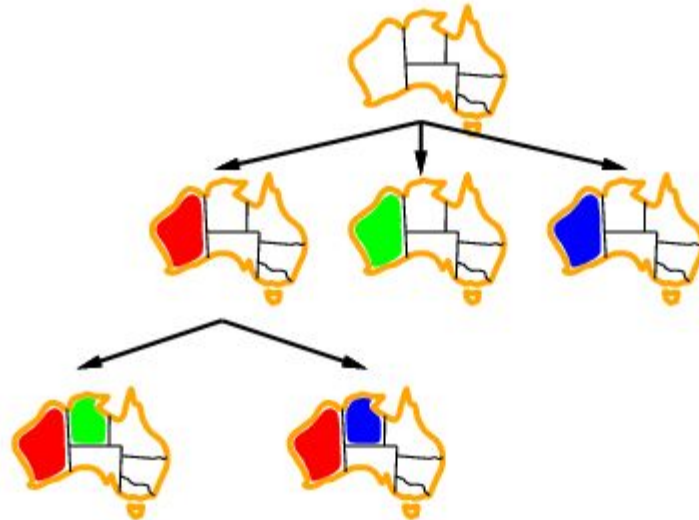
Backtracking example



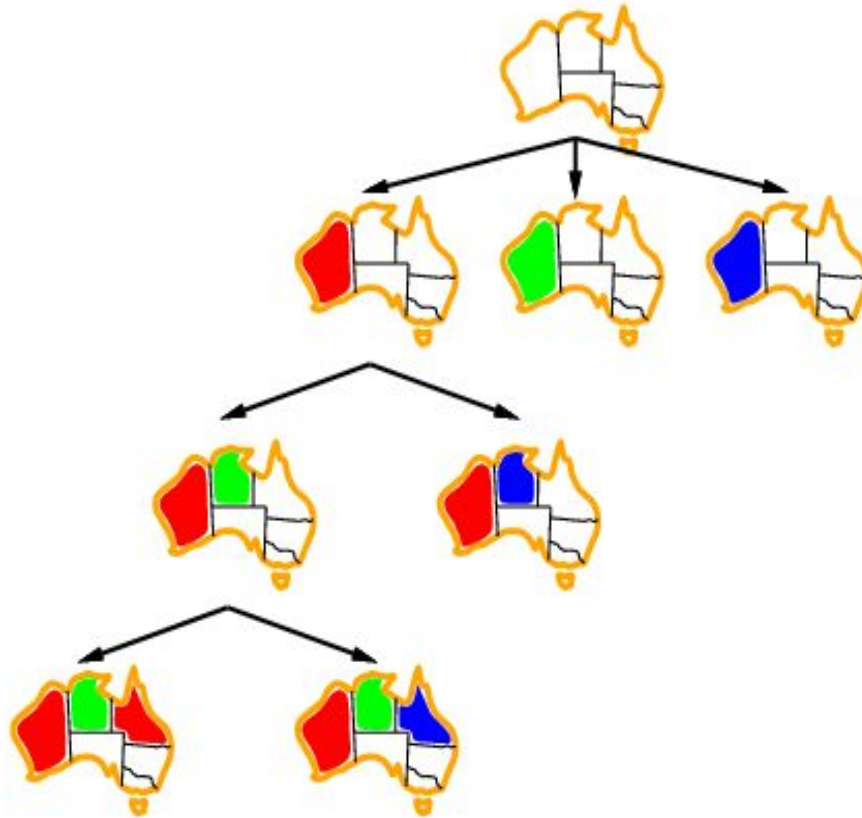
Backtracking example



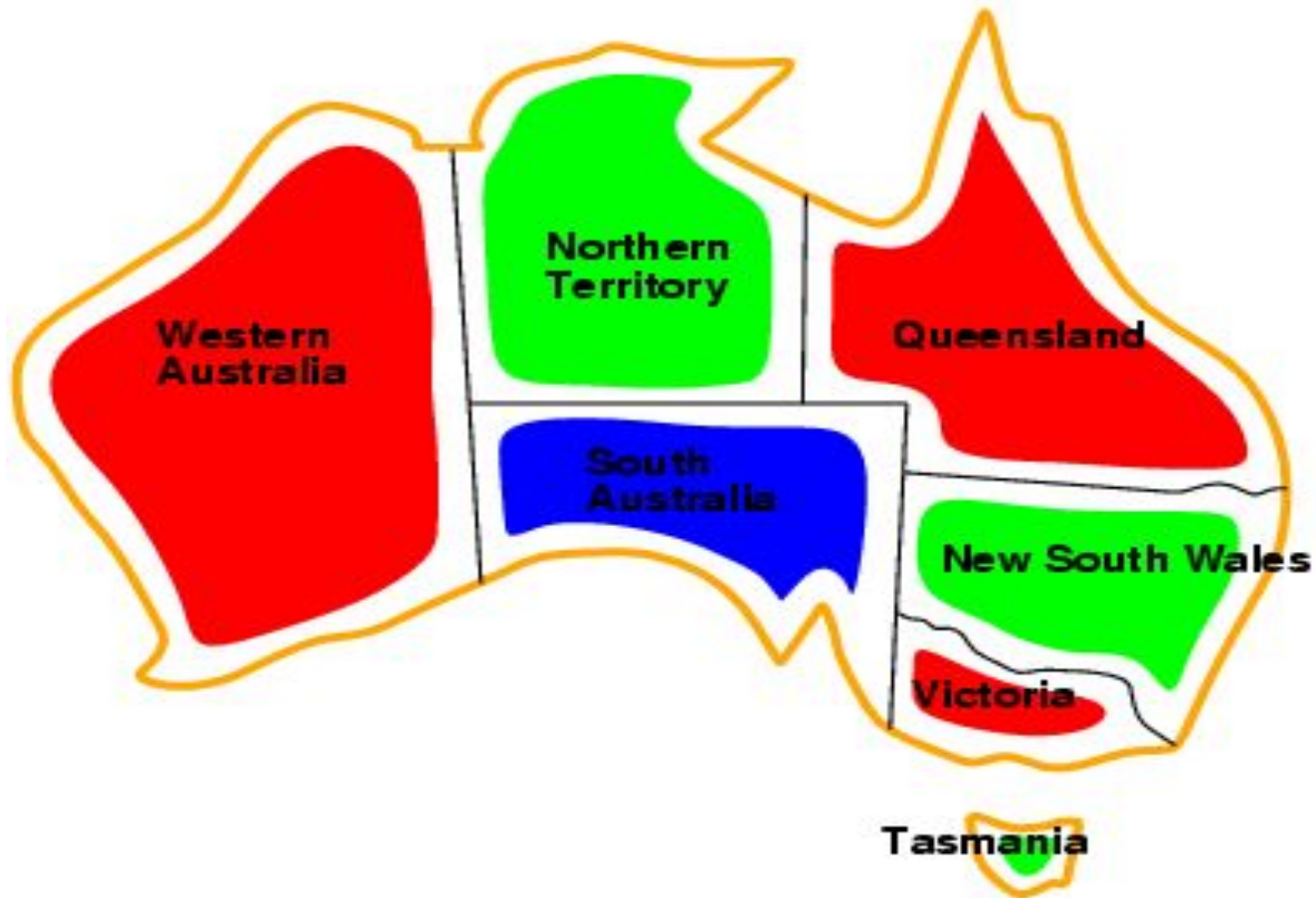
Backtracking example



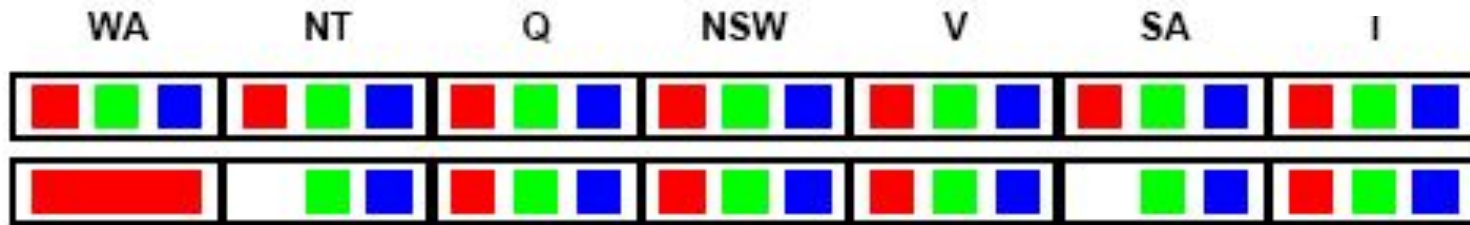
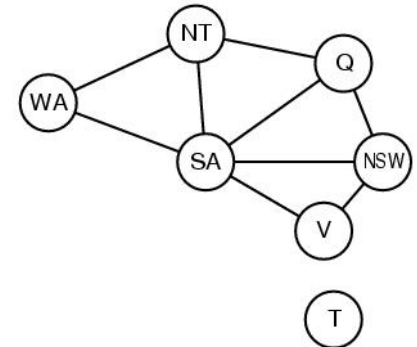
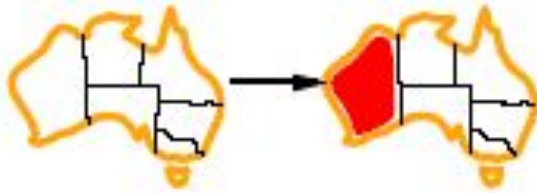
Backtracking example



Final (one of the solution)



Forward checking

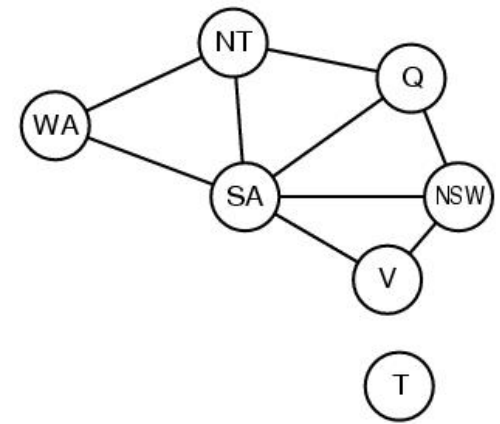


- Assign $\{WA=red\}$
- Effects on other variables connected by constraints to WA
 - *NT can no longer be red*
 - *SA can no longer be red*

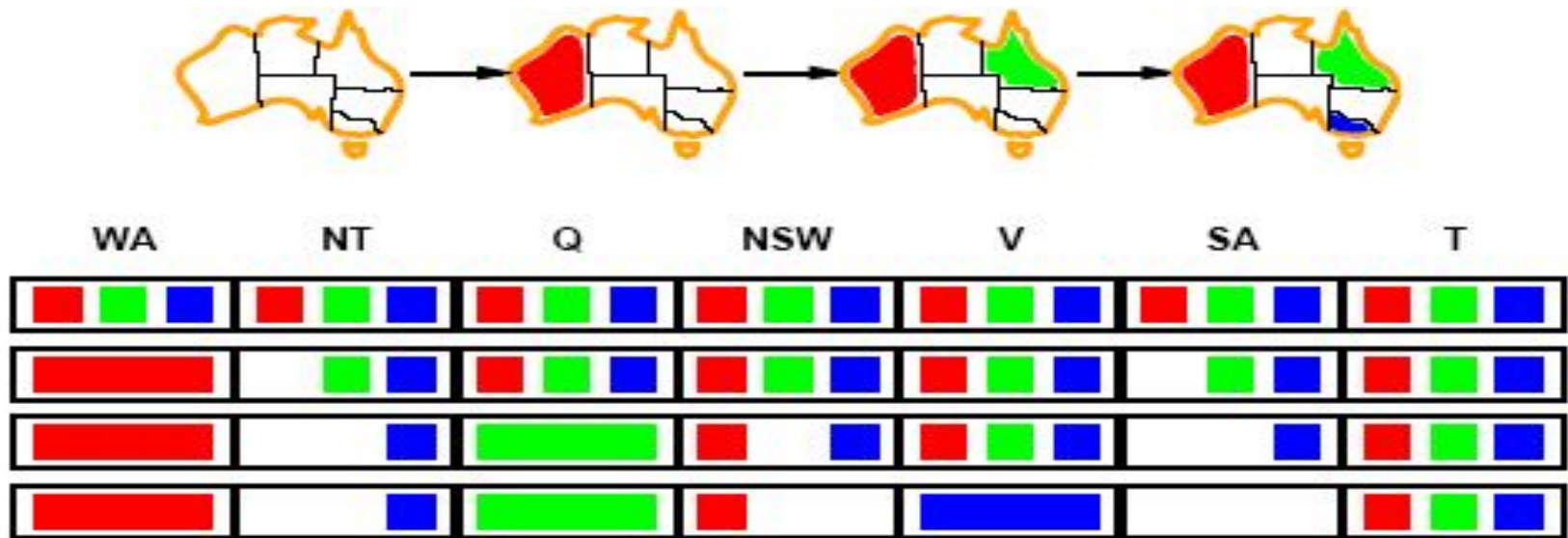
Forward checking



- Assign $\{Q=green\}$
- Effects on other variables connected by constraints with
 - *NT can no longer be green*
 - *NSW can no longer be green*
 - *SA can no longer be green*
- *MRV heuristic* would automatically select NT or SA next



Forward checking

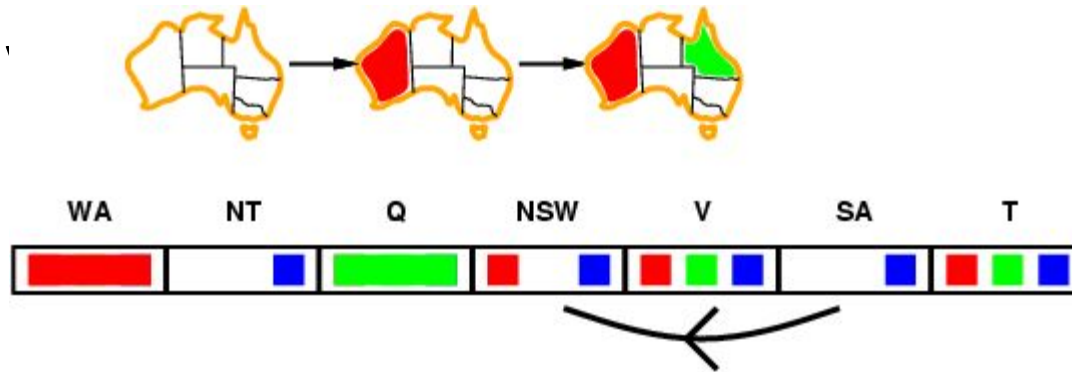


- If *V* is assigned *blue*
- Effects on other variables connected by constraints with *WA*
 - *NSW* can no longer be *blue*
 - *SA* is empty
- FC has detected that partial assignment is *inconsistent* with the constraints and backtracking can occur.
- **No more value for *SA*: backtrack**

Arc Consistency

- Simplest form of propagation makes each arc **consistent**
- $X \bowtie Y$ is consistent iff

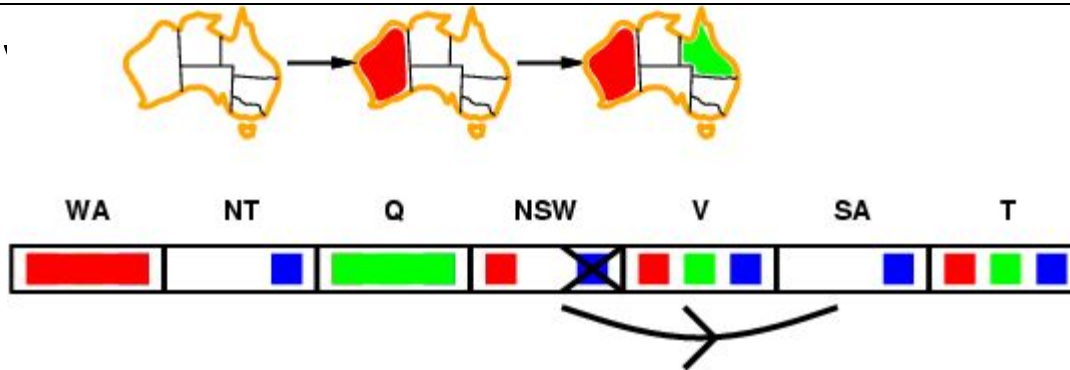
for every



Arc Consistency

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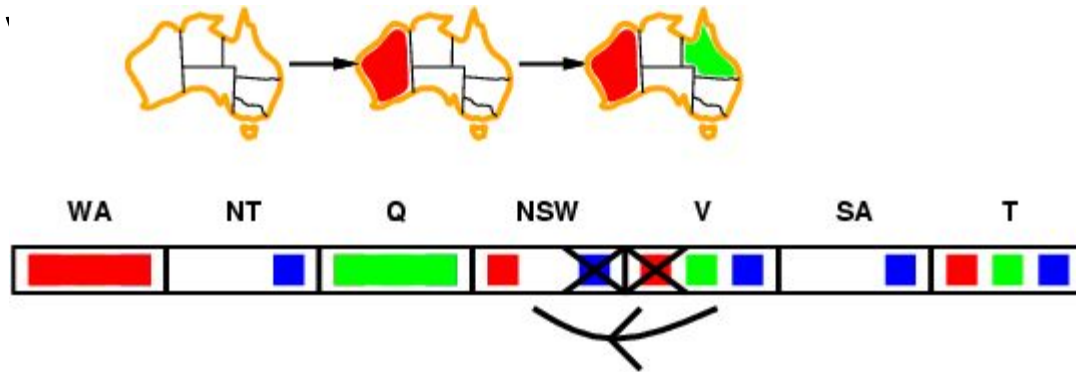
for every



Arc Consistency

- Simplest form of propagation makes each arc **consistent**
- $X \stackrel{?}{\sim} Y$ is consistent iff

for **every** y of Y



- If X loses a value, neighbors of X need to be rechecked

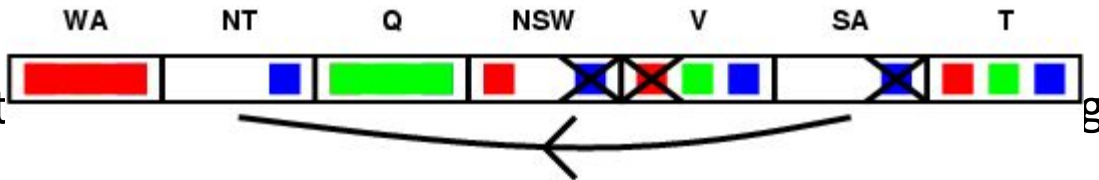
Arc consistency

- Simplest form of propagation makes each arc **consistent**
- $X \bowtie Y$ is consistent iff

for **every** value x of X there is **some** allowed value y of Y



- If X loses a
- Arc consist



- Can be run as a preprocessor or after each assignment

Arc Consistency Algorithm AC-3

Sometimes called Discrete Relaxation

```
function AC-3(csp) returns the CSP, possibly with reduced domains
  inputs: csp, a binary CSP with variables  $\{X_1, X_2, \dots, X_n\}$ 
  local variables: queue, a queue of arcs, initially all the arcs in csp

  while queue is not empty do
     $(X_i, X_j) \leftarrow \text{REMOVE-FIRST}(\text{queue})$ 
    if RM-INCONSISTENT-VALUES( $X_i, X_j$ ) then
      for each  $X_k$  in NEIGHBORS[ $X_i$ ] do
        add  $(X_k, X_i)$  to queue



---


function RM-INCONSISTENT-VALUES( $X_i, X_j$ ) returns true iff remove a value
  removed  $\leftarrow$  false
  for each  $x$  in DOMAIN[ $X_i$ ] do
    if no value  $y$  in DOMAIN[ $X_j$ ] allows  $(x, y)$  to satisfy constraint( $X_i, X_j$ )
      then delete  $x$  from DOMAIN[ $X_i$ ]; removed  $\leftarrow$  true
  return removed
```

- Time complexity: $O(n^2d^3)$

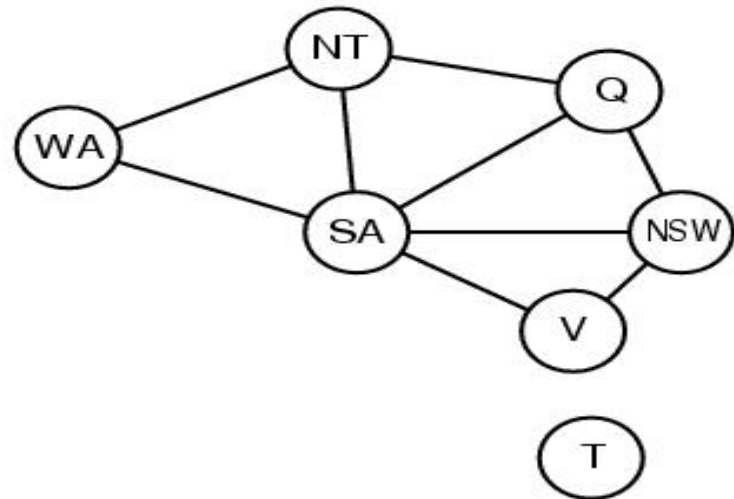
Variable and value ordering

- For Ex:
 - After assigning WA=red and NT=green there is only one possible value for SA i.e blue.
 - After SA is assigned, Values for Q,NSW, V are all forced
 - Choosing the variable with fewest legal moves is called MRV heuristic or MCV or fail-first heuristic.



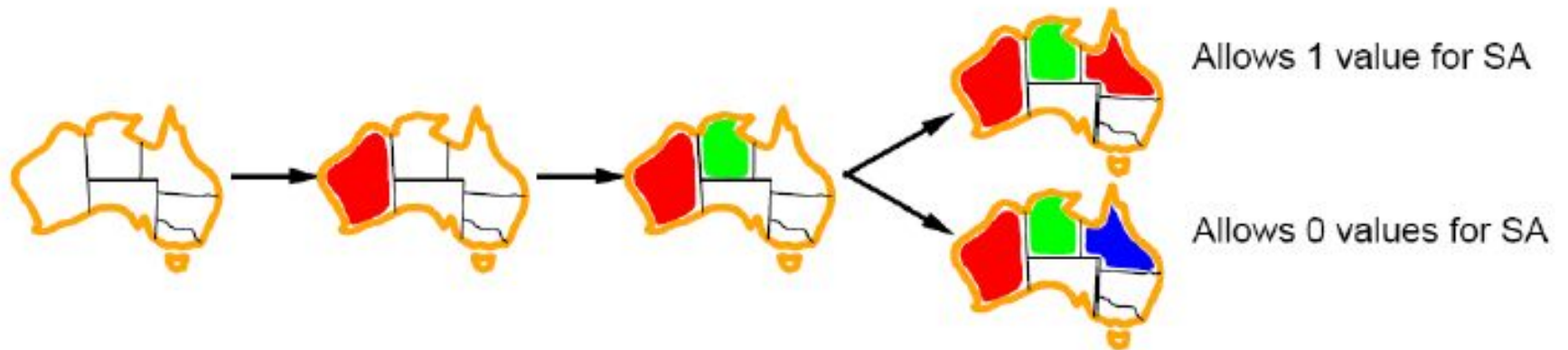
Variable and value ordering

- Degree heuristic: Degree heuristic can be useful as a tie breaker after MRV
- MRV heuristic is a more powerful guide, but the degree heuristic can be useful as a tie breaker



- LCV heuristic: The value that rules out the fewest values in the remaining variables(i.e what value will leave the most other values for other variable)

LCV: Least constraining value

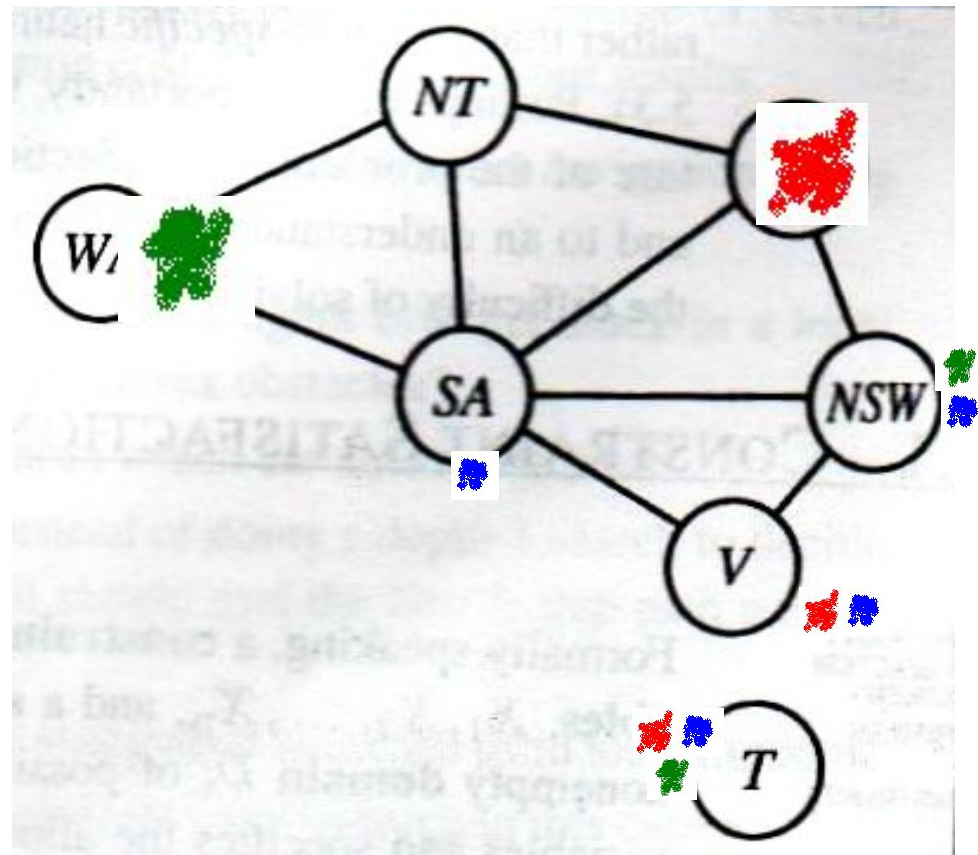


Constraint propagation

- There are some inconsistencies with forward checking
- Constraint propagation is stronger than forward checking by considering arc consistency
- the arc is consistent if, for *every* value x of SA , there is *some* value y of NSW that is consistent with x .
- For Ex: $SA = \{\text{blue}\}$ and $NSW = \{\text{red}, \text{blue}\}$

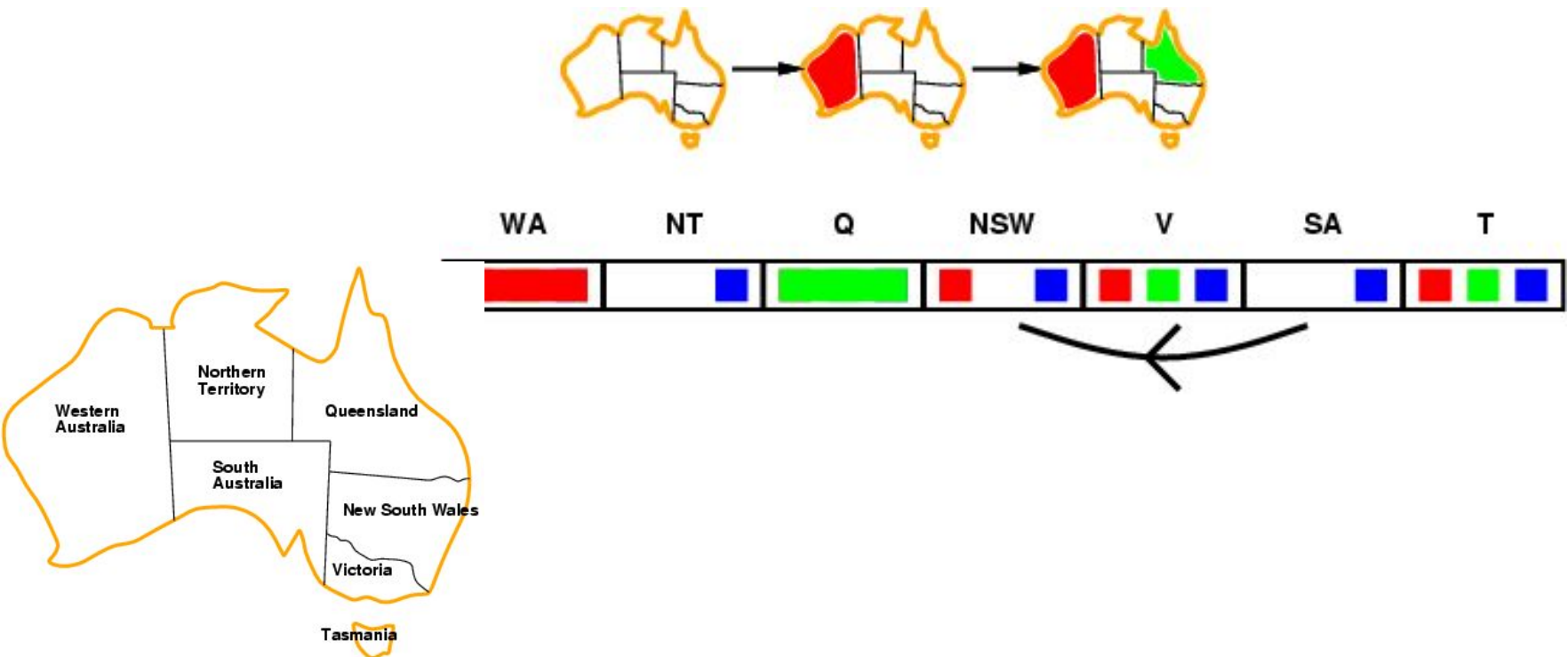
Arc Consistency

- $A \rightarrow B$ is consistent if for each remaining value in domain of A , there may be a consistent value in domain of B .
 - Consistent:
 - $SA \rightarrow NSW, NSW \rightarrow V, \dots$
 - Not Consistent:
 - $NSW \rightarrow SA, NT \rightarrow SA, \dots$



Arc consistency

- Simplest form of propagation makes each arc **consistent**
- $X \rightarrow Y$ is consistent iff
for **every** value x of X there is **some** allowed y



Special Constraints...

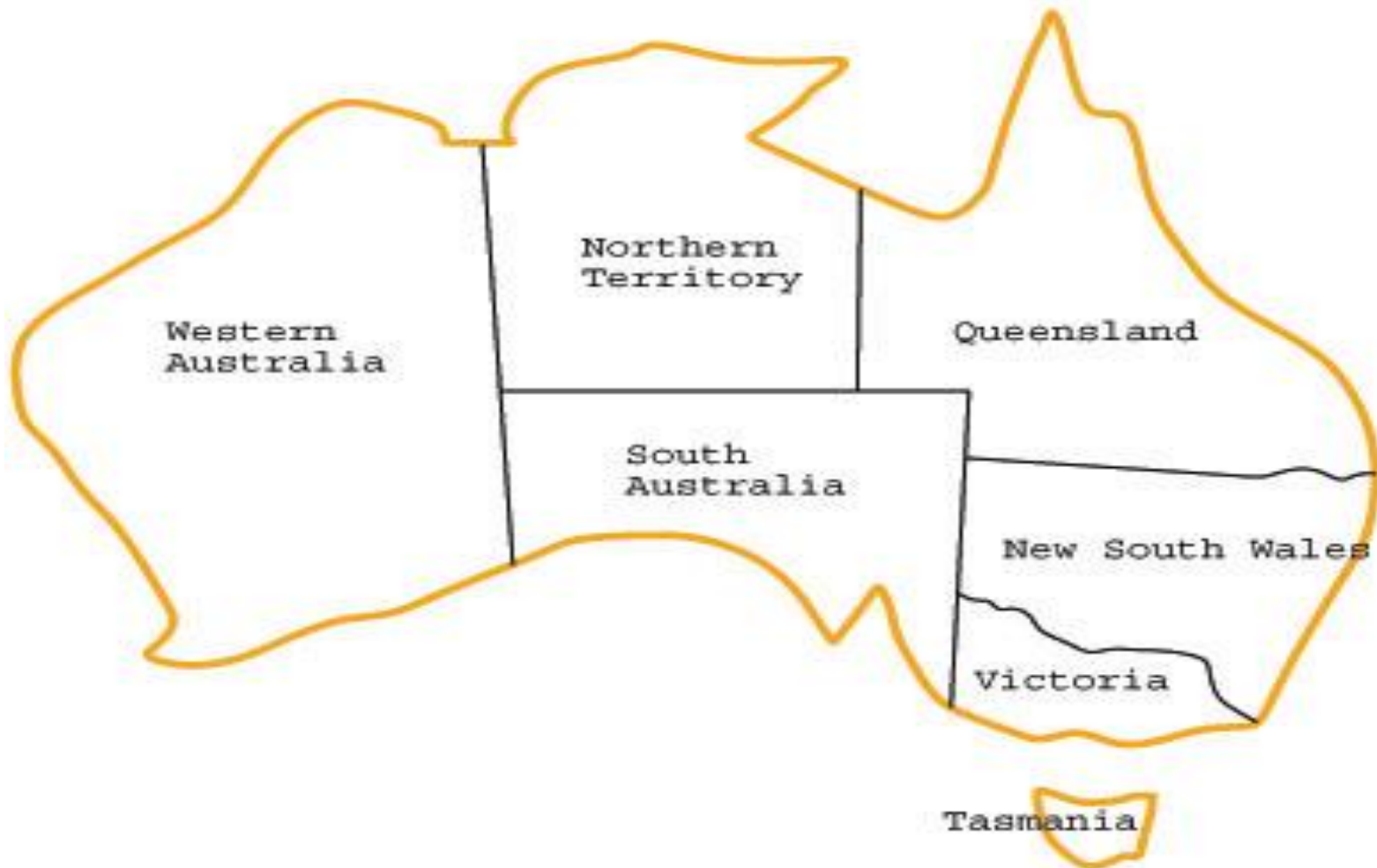
- All Different
 - Sort remaining variables based on their number of choices...
- Resource Constraints
 - Checking the sums, ...
 - $\text{Flight1} \in [0,165]$, $\text{Flight2} \in [0,385]$
 - $\text{Flight1} + \text{Flight2} = 420$
 - $\text{Flight1} \in [35,165]$, $\text{Flight2} \in [255,385]$
 - Bounds Propagation

Intelligent Backtracking

- The BACKTRACKING-SEARCH algorithm has a very simple policy for what to do when a branch of the search fails: back up to the preceding variable and try a different value for it.
- This is called chronological backtracking, because the *most recent decision* point is revisited.
- There are much better ways
- One way to get rid of the problem is using **intelligent backtracking** algorithms
- **Backjumping (BJ)** is different from BT in the following:

BJ vs. BT

We want to color each area in the map with a different color



BJ vs. BT

- Let's consider what BT does in the map coloring problem
 - Assume that variables are assigned in the order Q, NSW, V, T, SA, WA, NT
 - Assume that we have reached the partial assignment
 $Q = red, NSW = green, V = blue, T = red$
 - When we try to give a value to the next variable SA , we find out that all possible values violate constraints
 - Dead end!
 - BT will backtrack to try a new value for variable T !
 - Not a good idea!

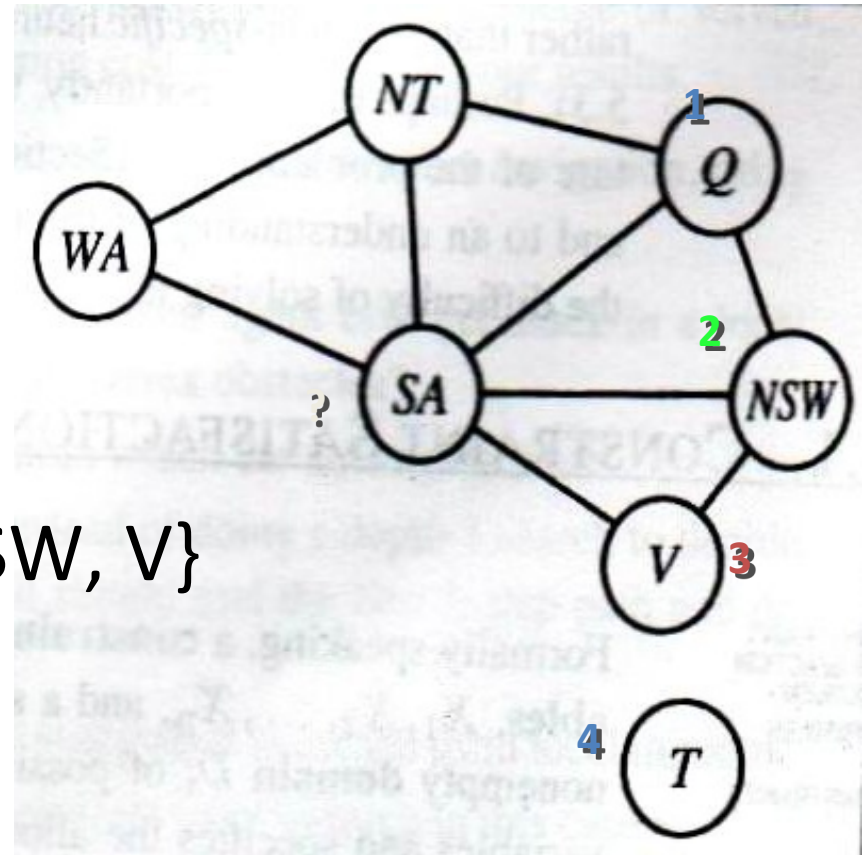
BJ vs. BT

- BJ has a smarter approach to backtracking
 - A more intelligent approach to backtracking is to go all the way back to one of the set of variables that *caused the failure*
 - The set of these variables is called a **conflict set**
 - The conflict set for SA is {Q, NSW, V}

Back Jumping

1. Q ? Red
2. NSW ? Green
3. V ? Blue
4. T ? Red
5. SA ? ?

- Conflict Set: {Q, NSW, V}



The end of CSP