

Fall 2021 | Lecture 5

## Constraint Satisfaction Problems

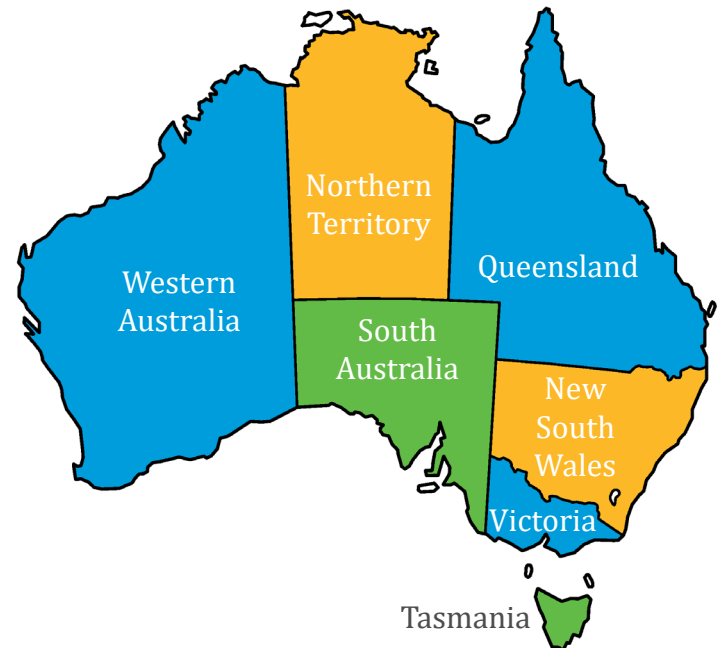
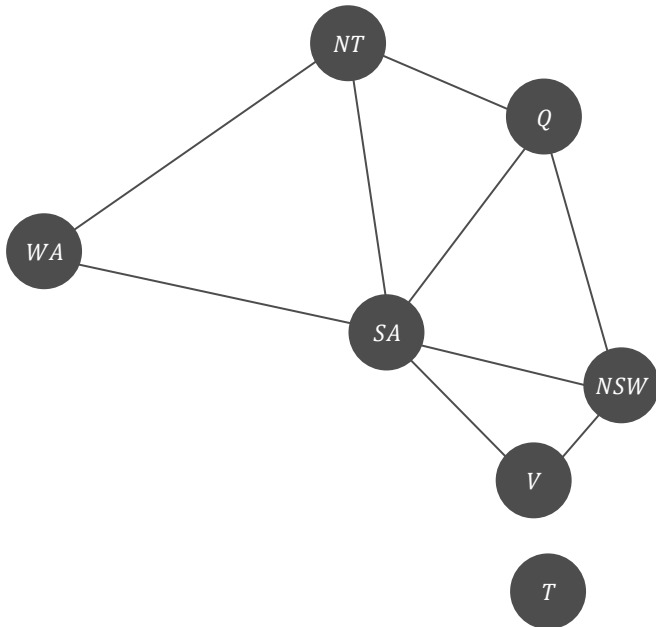
Ariel Procaccia | Harvard University

# WHAT ARE CSPS?

- A **constraint satisfaction problem** (CSP):
  - Variables  $\{X_1, \dots, X_n\}$
  - Domains  $\{D_1, \dots, D_n\}$
  - A set of constraints: defined on subsets of variables, give allowable tuples of values
- Consider (possibly partial) assignments of values to variables
- A **solution** is a **complete** and **consistent** assignment
- CSP are search problems but their structure allows general purpose algorithms that don't require domain knowledge

# EXAMPLE: MAP COLORING

- Variables = {WA, NT, SA, Q, NSW, V, T}
- $D_i = \{\text{blue, orange, green}\}$
- Constraints: adjacent regions have different colors

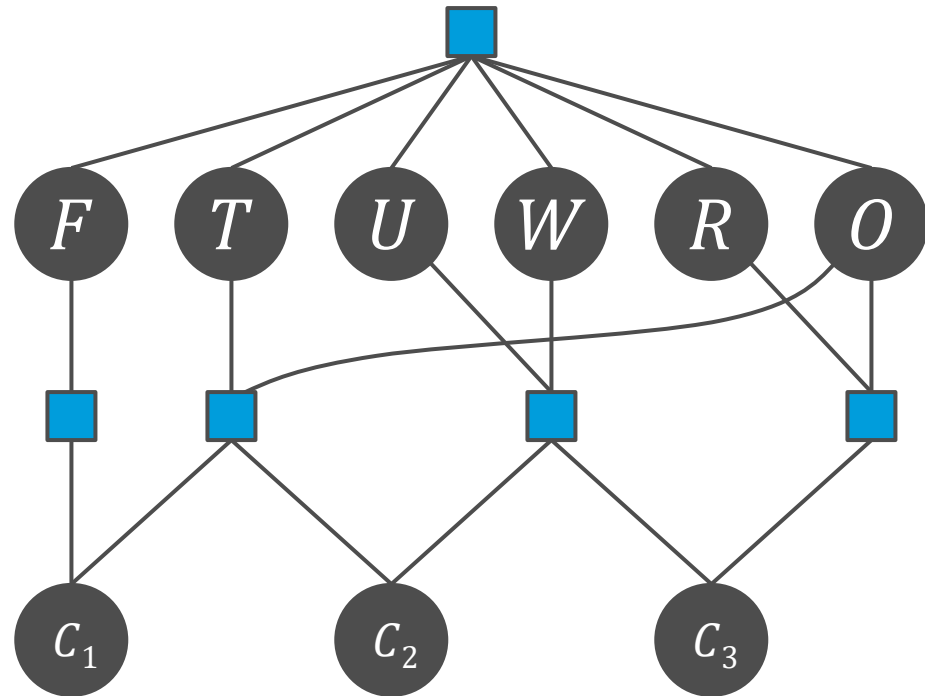


# TYPES OF CONSTRAINTS

- A **unary constraint** restricts the value of a single variable
- A **binary constraint** relates two variables
- A CSP with only unary and binary constraints (such as map coloring) is called a **binary CSP** and can be represented using a constraint graph
- Constraints involving more than two variables can be represented using a hypergraph

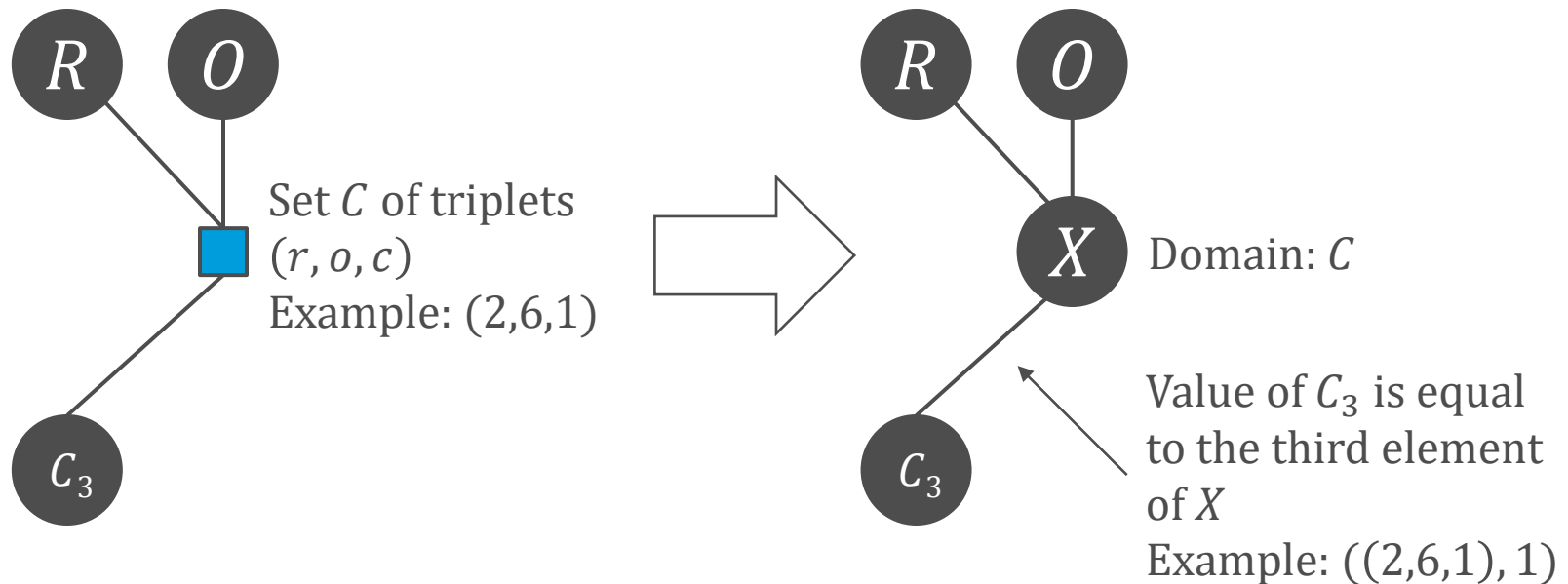
# EXAMPLE: CRYPTARITHMETIC

$$\begin{array}{r} \text{T W O} \\ + \text{T W O} \\ \hline \text{F O U R} \end{array}$$



# REDUCTION TO BINARY

- Any constraint can be reduced to a set of binary constraints by introducing additional variables
- So we can assume that the given CSP is binary



# INFERENCE

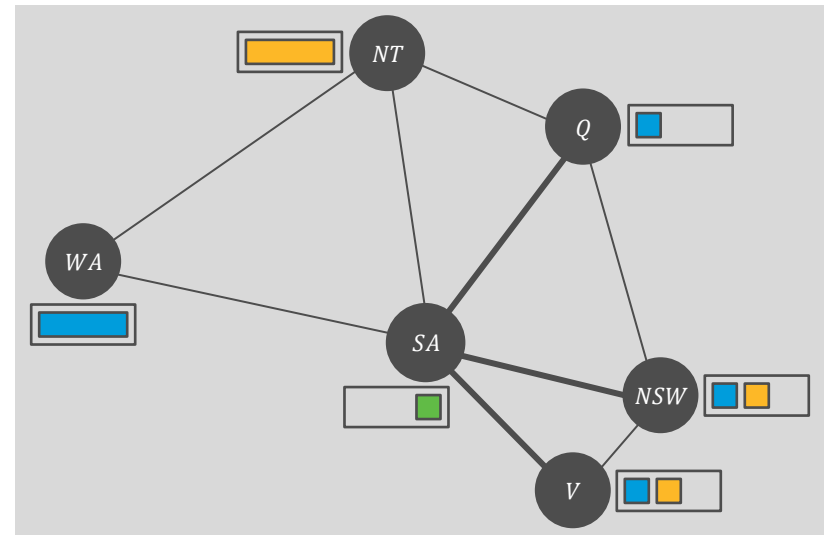
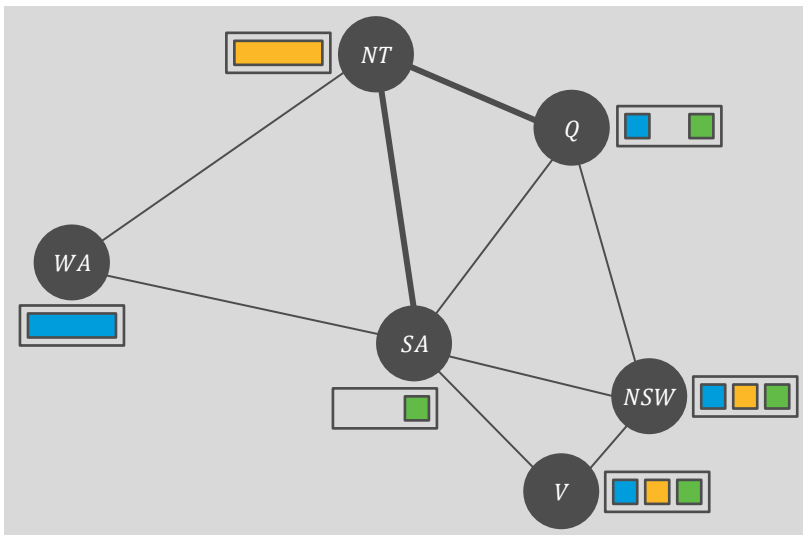
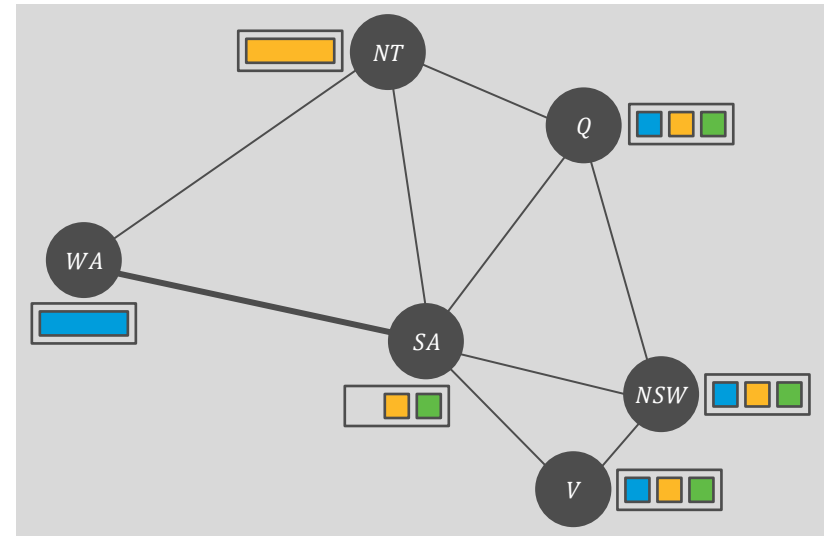
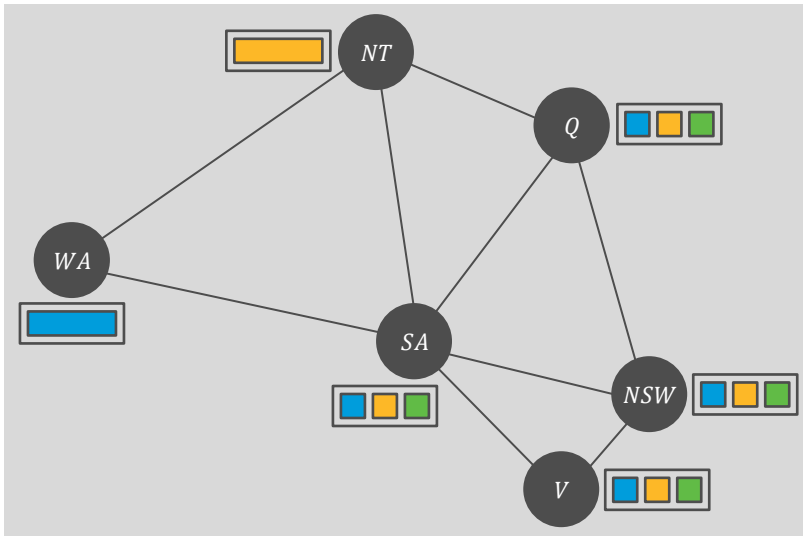
- We can use the constraints to directly eliminate potential assignments of variables
- This can be done as a preprocessing step before search, or it can be interleaved with search
- Sometimes search isn't needed at all!

# ARC CONSISTENCY

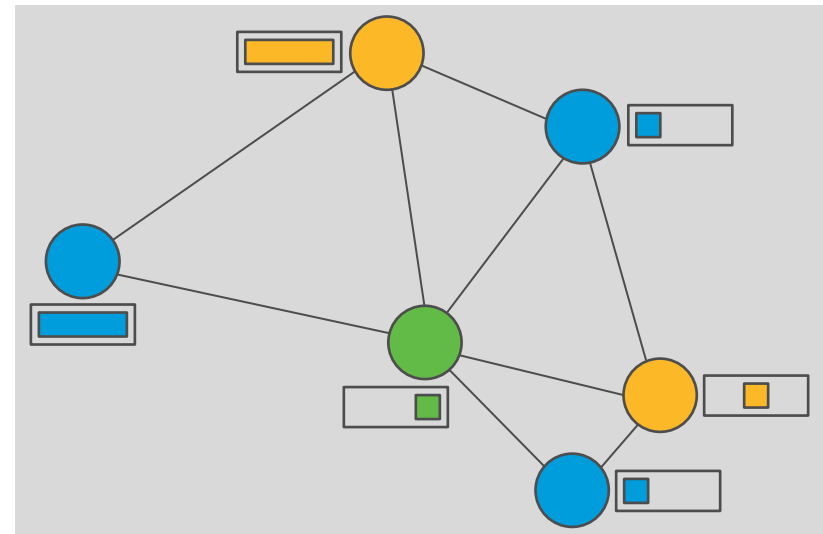
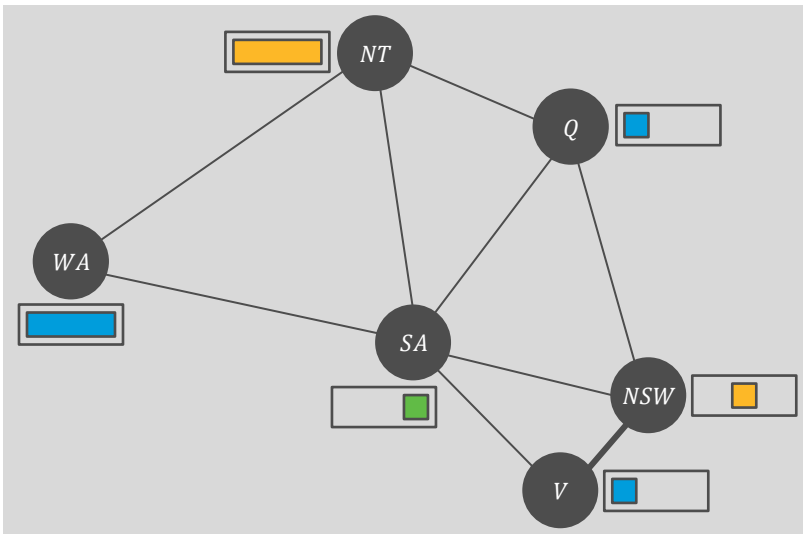
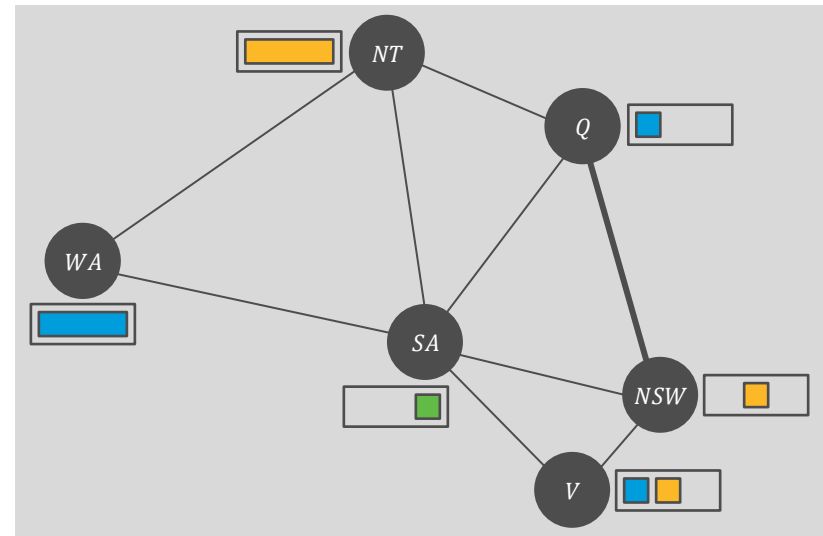
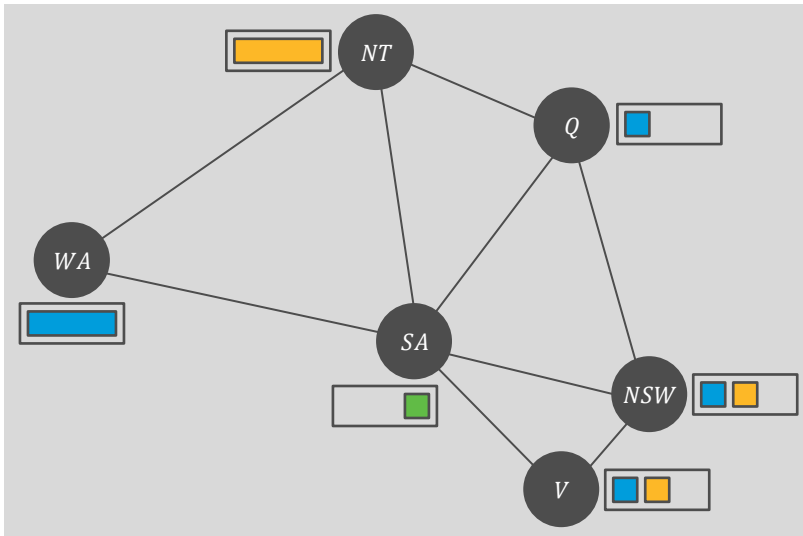
- A variable  $X_i$  is **arc-consistent** with respect to variable  $X_j$  if for every value in  $D_i$  there is a value in  $D_j$  that satisfies the binary constraint on  $(X_i, X_j)$
- The CSP is arc-consistent if every variable is arc-consistent with respect to every other variable



# ENFORCING ARC CONSISTENCY



# ENFORCING ARC CONSISTENCY



# THE AC-3 ALGORITHM

**function** AC-3 (*csp*)

*queue*  $\leftarrow$  all arcs in *csp*

**while** *queue* is not empty **do**

$(X_i, X_j) \leftarrow \text{POP}(\textit{queue})$

**if** REVISE(*csp*,  $X_i, X_j$ ) **then**

**if**  $|D_i| = 0$  **then return** false

**for** each neighbor  $X_k \neq X_j$  of  $X_i$

                add  $(X_k, X_i)$  to *queue*

**return** true

**function** REVISE(*csp*,  $X_i, X_j$ )

*revised*  $\leftarrow$  false

**for** each  $x$  in  $D_i$  **do**

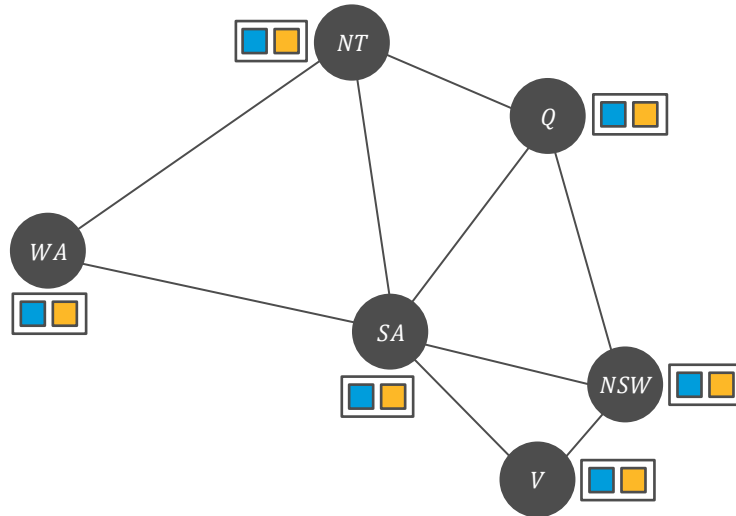
**if** no value  $y$  in  $D_j$  allows  $(x, y)$  to satisfy the  
        constraint between  $X_i$  and  $X_j$  **then**

            delete  $x$  from  $D_i$

*revised*  $\leftarrow$  true

**return** *revised*

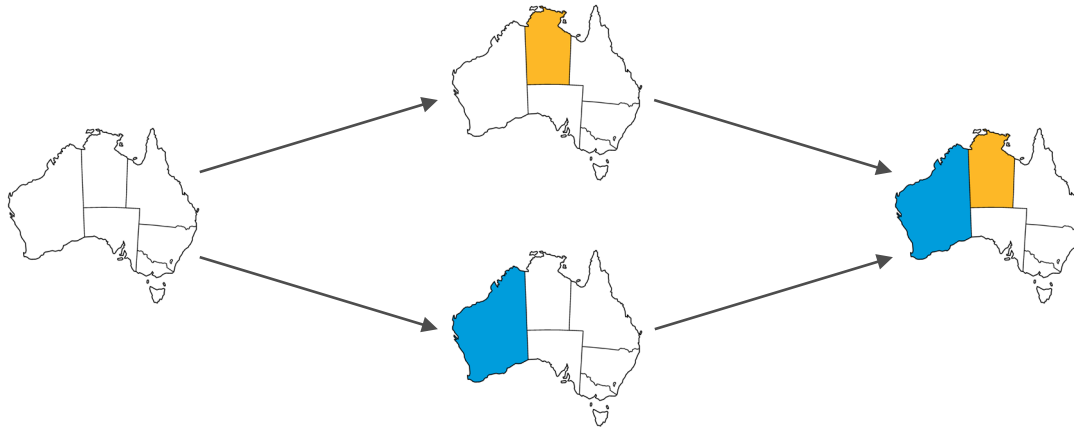
# SHORTCOMING OF ARC CONSISTENCY



- There is no solution but AC-3 does nothing!
- Enforcing consistency for subsets of variables of size  $k \geq 3$  is sufficient here

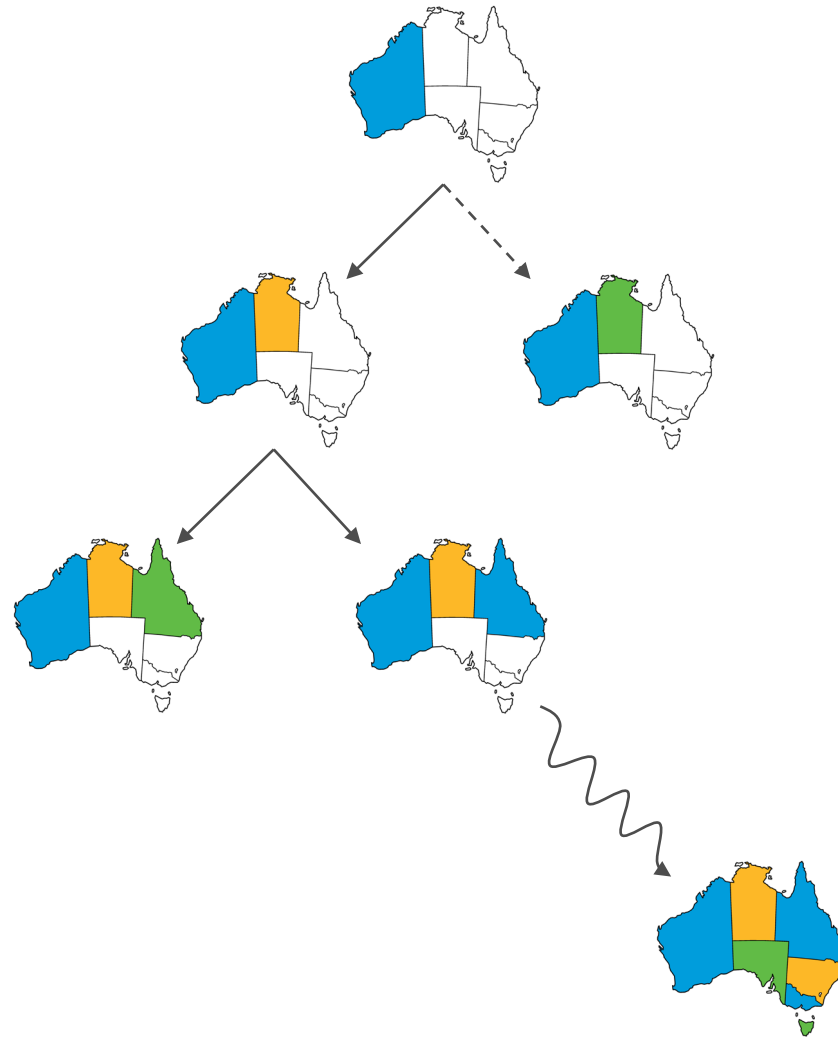
# BACKTRACKING SEARCH

- Typically, inference would leave variables with multiple possible values; we need to search!
- A useful property of CSPs is commutativity: the order in which we assign variables doesn't matter



- Backtracking search assigns the variables one by one, and backtracks if needed

# BACKTRACKING SEARCH



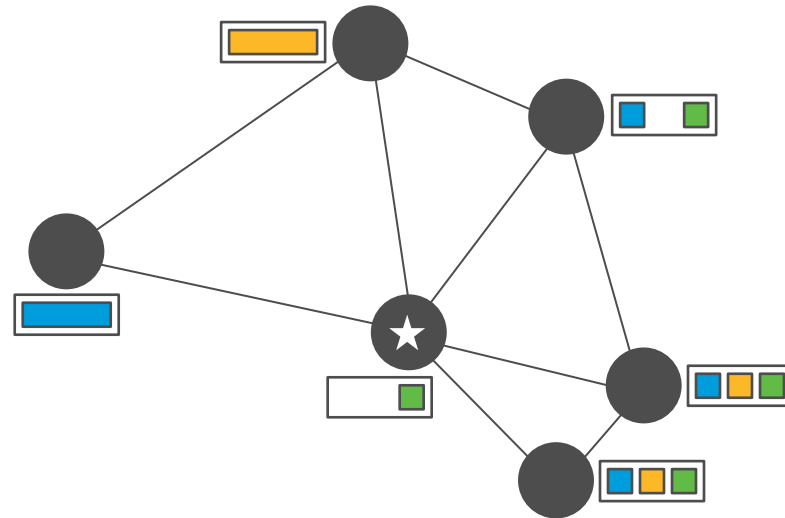
# BACKTRACKING SEARCH

```
function BACKTRACK(csp, assign)
  if assign is complete then return assign
  var ← SELECT-UNASSIGNED-VARIABLE(csp, assign)
  for each value in ORDER-DOMAIN-VALUES(csp, var, assign) do
    if value is consistent with assign then
      add {var = value} to assign
      inferences ← INFERENCE(csp, var, assign)
      if inferences ≠ failure then
        add inferences to csp
        result ← BACKTRACK(csp, assign)
        if result ≠ failure then return result
      remove inferences from csp
      remove {var = value} from assign
  return failure
```

**Poll 1:** What are some potentially good heuristics for SELECT-UNASSIGNED-VARIABLE?

# VARIABLE ORDERING

- Backtracking search includes the function SELECT-UNASSIGNED-VARIABLE
- **Minimum remaining values heuristic:** choose the variable with the fewest legal values

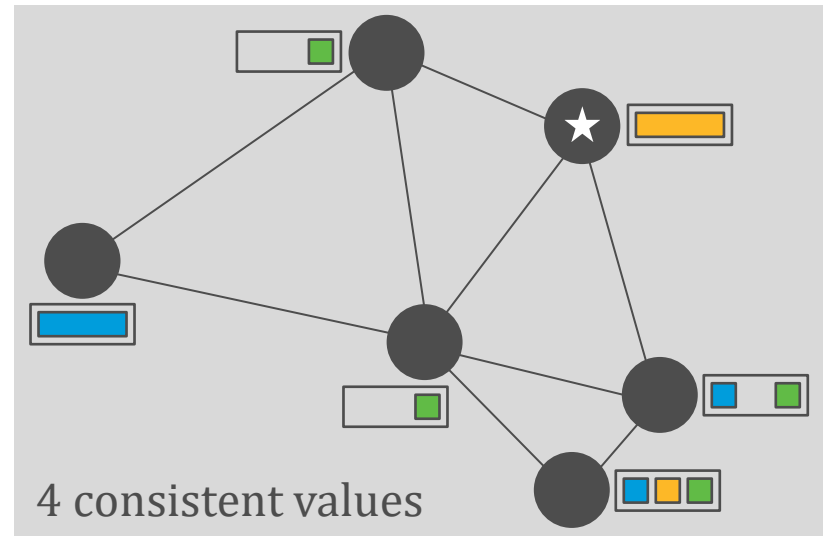
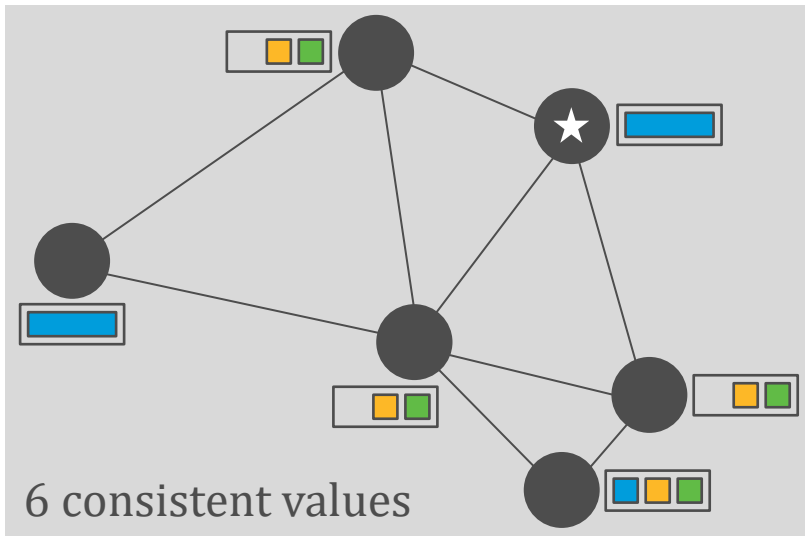


- Logic: prune as early as possible



# VALUE ORDERING

- Backtrack search includes the function ORDER-DOMAIN-VALUES
- **Least constraining value heuristic:** choose value that rules out fewest choices for other variables



- Logic: consider value that's most likely to succeed

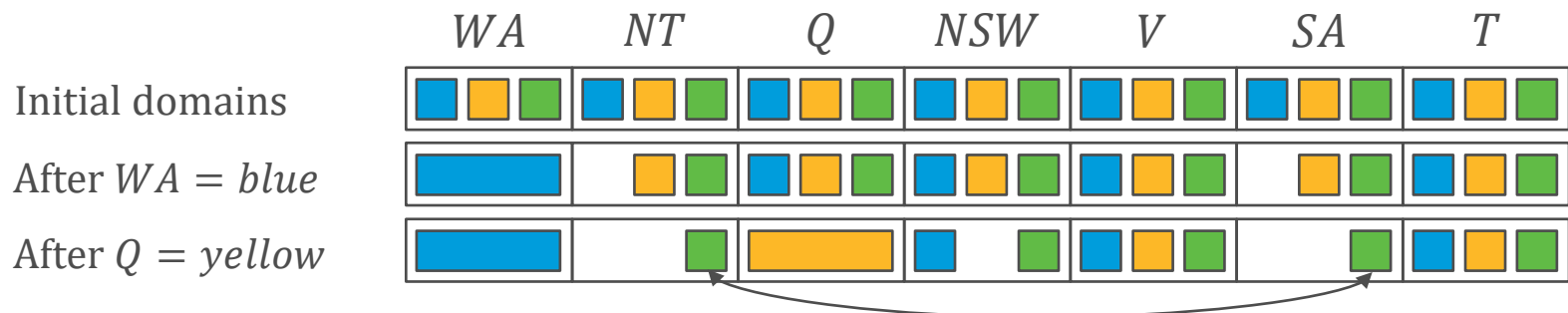
# FORWARD CHECKING

- The function INFERENCE gives us an opportunity to interleave inference and search
- The simplest inference is **forward checking**, which enforces arc consistency for each variable that's assigned a value

	<i>WA</i>	<i>NT</i>	<i>Q</i>	<i>NSW</i>	<i>V</i>	<i>SA</i>	<i>T</i>
Initial domains							
After <i>WA</i> = <i>blue</i>							
After <i>Q</i> = <i>yellow</i>							
After <i>V</i> = <i>green</i>							

# MAINTAINING ARC CONSISTENCY

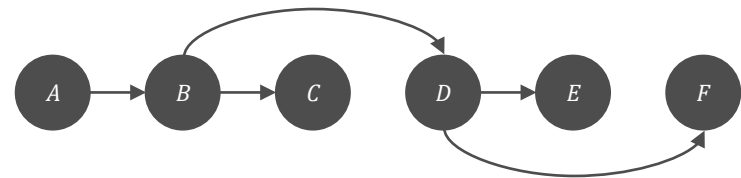
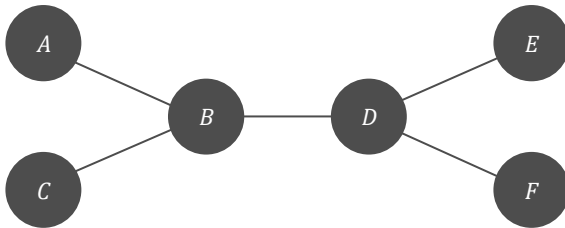
- In our forward checking example, more powerful arc consistency enforcement would have terminated earlier



- The INFERENCE function can implement AC-3, but only arcs  $(X_j, X_i)$  between the newly assigned variable  $X_i$  and unassigned neighbors  $X_j$  are initially added to the queue

# THE STRUCTURE OF PROBLEMS

- Solving a CSP whose constraint graph is structured as a tree is very easy
- We say that the CSP is **directionally arc-consistent** under an ordering of the variables  $X_1, \dots, X_n$  if and only if for every  $i < j$ ,  $X_i$  is arc-consistent with  $X_j$
- The idea is to topologically sort the tree and make the it directionally arc-consistent with respect to the  $n - 1$  edges



- Then we go down the list of variables and choose any remaining value

# TREE CSP SOLVER

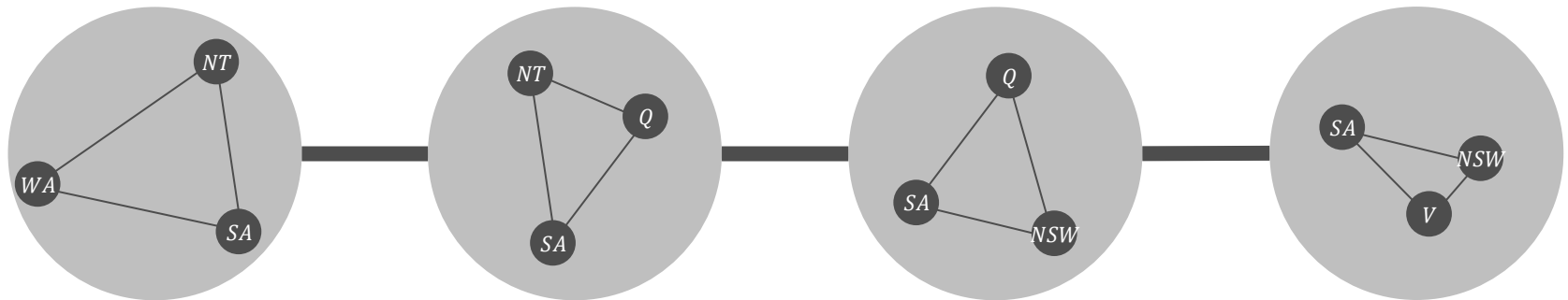
```
function TREE-CSP-SOLVER(csp)  
   $n \leftarrow$  number of variables in  $X$   
   $assign \leftarrow \emptyset$   
   $root \leftarrow$  any variable in  $X$   
   $X \leftarrow$  TOPOLOGICAL-SORT( $X, root$ )  
  for  $j = n$  down to 2 do  
    MAKE-ARC-CONSISTENT(PARENT( $X_j$ ),  $X_j$ )  
    if consistency fails then return failure  
  for  $i = 1$  to  $n$  do  
    if  $D_i$  has no consistent values then return failure  
     $assign[X_i] \leftarrow$  any consistent value from  $D_i$   
  return  $assign$ 
```

Running time  $O(nd^2)$  for  $|D_i| \leq d$

# TREE DECOMPOSITION

Any CSP can be converted into a tree:

- Every variable appears in at least one node
- If two variables are connected by a constraint, they must appear together in at least one node
- If one variable appears in two nodes, there must be a path between them and it must appear in every node on the path



# TREE DECOMPOSITION

- **Poll 2:** For a CSP with  $n$  variables and domains of size  $d$ , what is the minimum number of nodes in a tree decomposition of the CSP?
  1.  $\Theta(1)$  ✓
  2.  $\Theta(d)$
  3.  $\Theta(n/d)$
  4.  $\Theta(n)$
- If a tree decomposition has  $m$  nodes and the largest node has  $k$  variables, the running time of TREE-CSP-SOLVER is  $O(md^{2k})$