

CSE 473: Artificial Intelligence

Constraint Satisfaction Examples

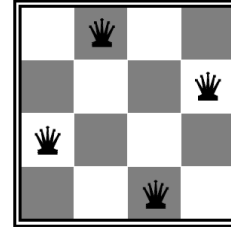
Cynthia Matuszek

Multiple slides adapted from Dan Klein, Stuart Russell, Andrew Moore,
Paula Matuszek

Why do we care about CSPs?

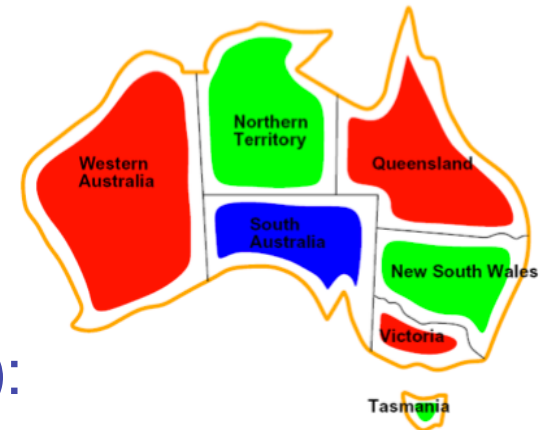
- Standard search problems:

- State is a “black box”
- **Any function** can be goal, successor function can be anything



- Constraint satisfaction problems (CSPs):

- Search problems that vary in the goal test.
- State is defined by **variables** X_i with values from a **domain** D
- Goal test is a **set of constraints**



- Why do we care?

- **Allows for informed search**
- **Using structure of problems to search wisely**

CSP heuristics
& methods

Revisiting and Reviewing

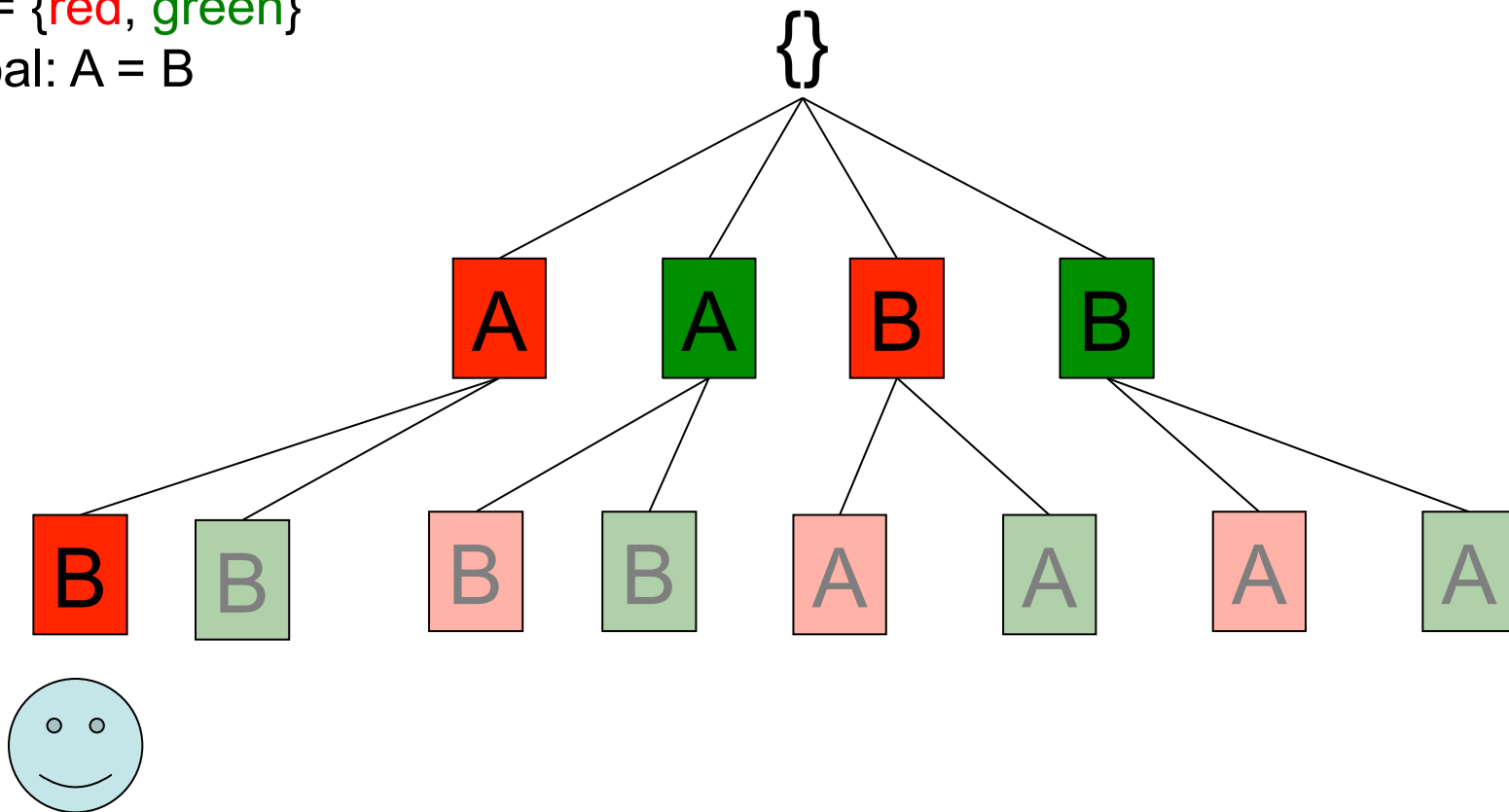
- Uninformed Search for Constraint Satisfaction Problems
- Backtracking Search
- Forward Checking
- *k*-Consistency
- Ordering Heuristics
 - Minimum Remaining Values Ordering
 - Least Constraining Values
- Tree- and almost-tree CSPs

Bread-first search & CSPs

$X = \{A, B\}$

$D = \{\text{red}, \text{green}\}$

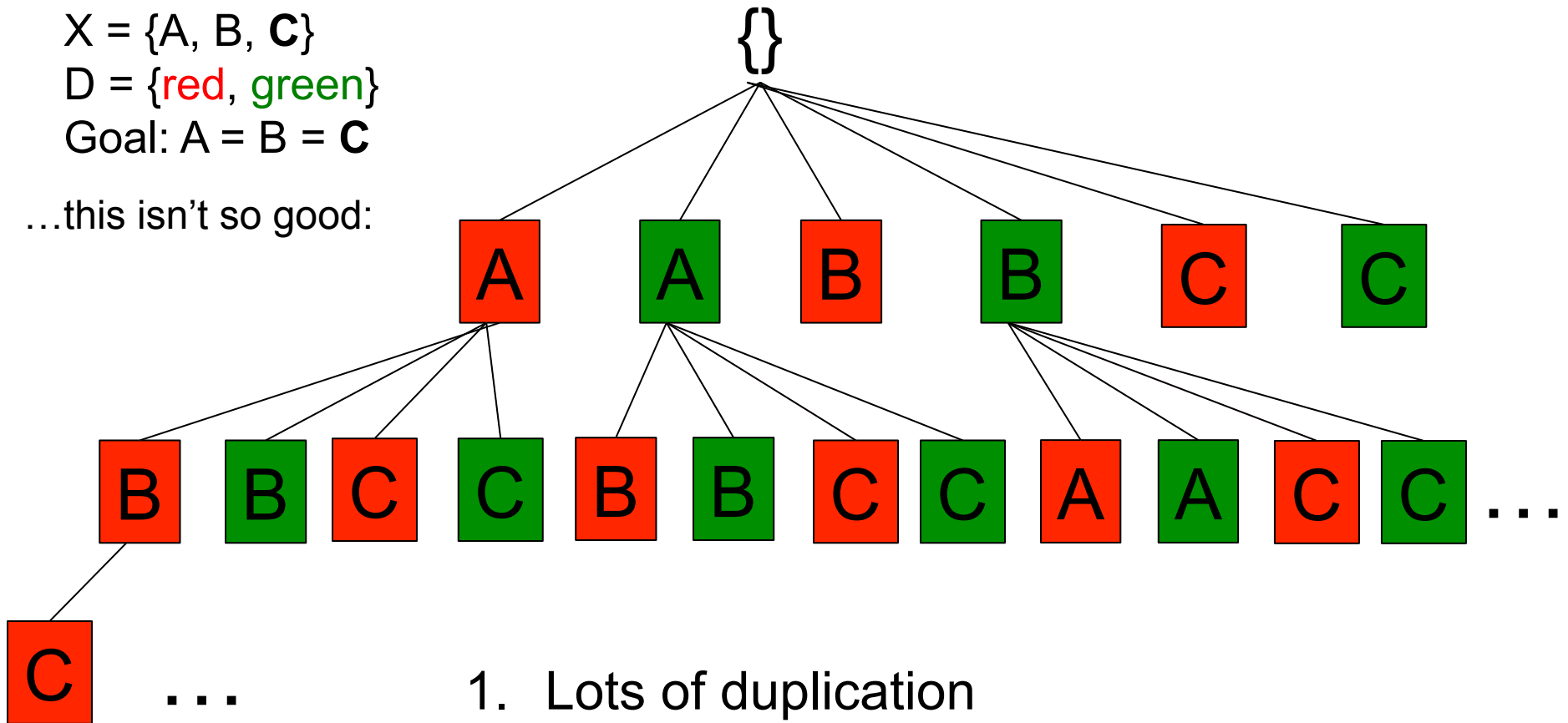
Goal: $A = B$



Bread-first search & CSPs

$X = \{A, B, C\}$
 $D = \{\text{red}, \text{green}\}$
Goal: $A = B = C$

...this isn't so good:



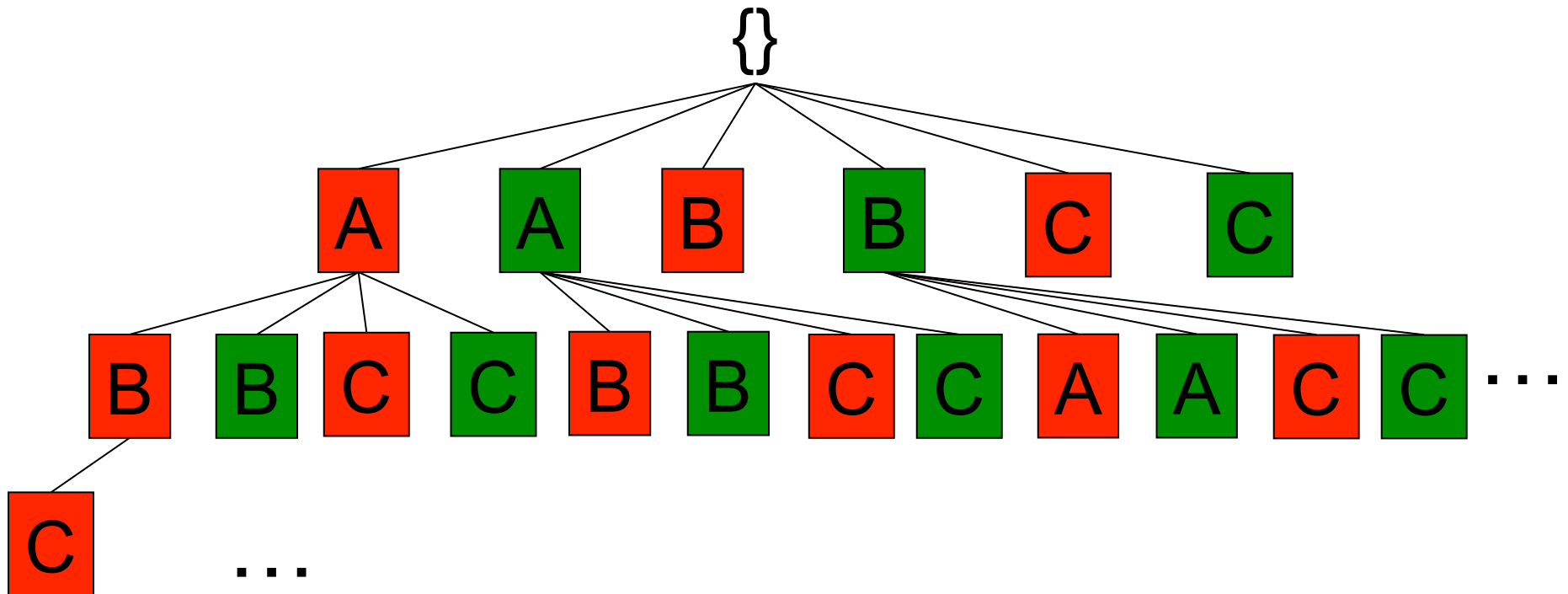
1. Lots of duplication
2. BFS always fills out the top of the search tree, when the solutions are at the bottom

Can We Do Better?

- It's actually hard to understand why uninformed search does so badly. Why?
- Because you would never implement these problems that way.
 - Better successor functions, internal checks, ...
- Hence, “uninformed”

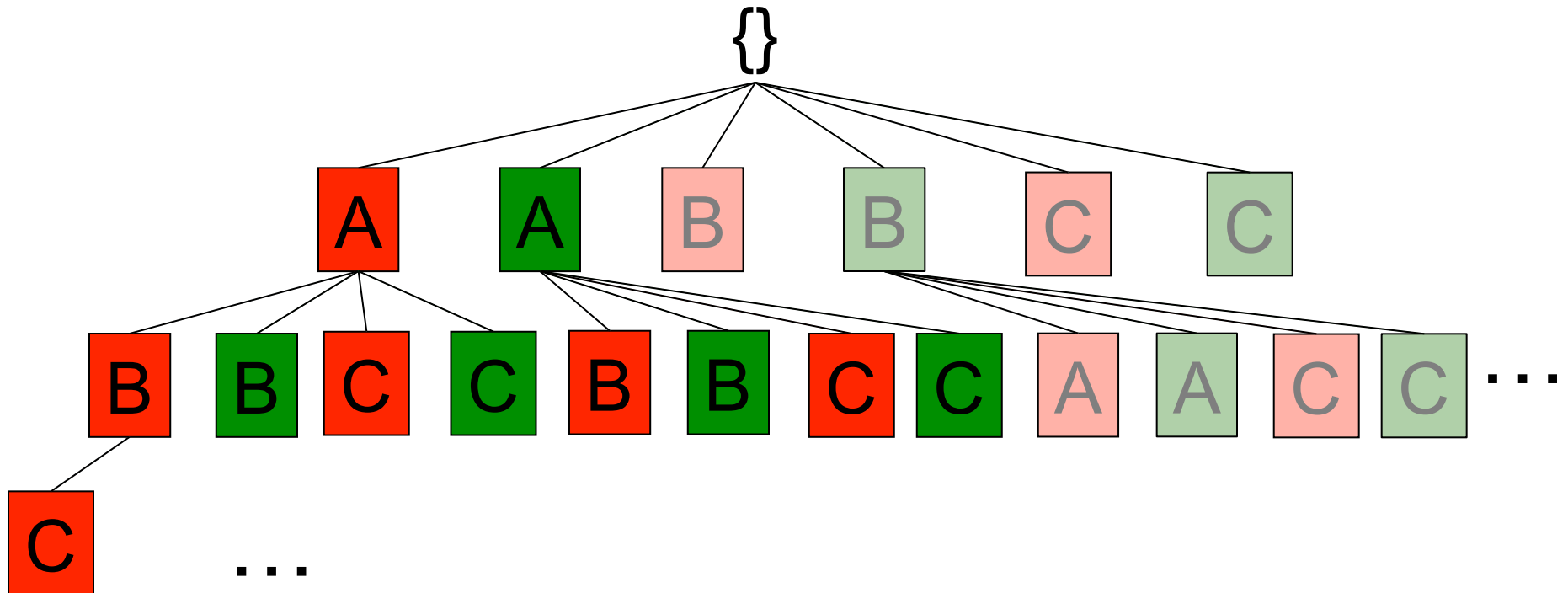
Improvement 1: Commutativity

- Idea 1: Only consider a single variable at each point
 - Variable assignments are commutative, so fix ordering
 - I.e., [A = red then B = green] same as [B = green then A = red]
 - Only need to consider assignments to a single variable at each step



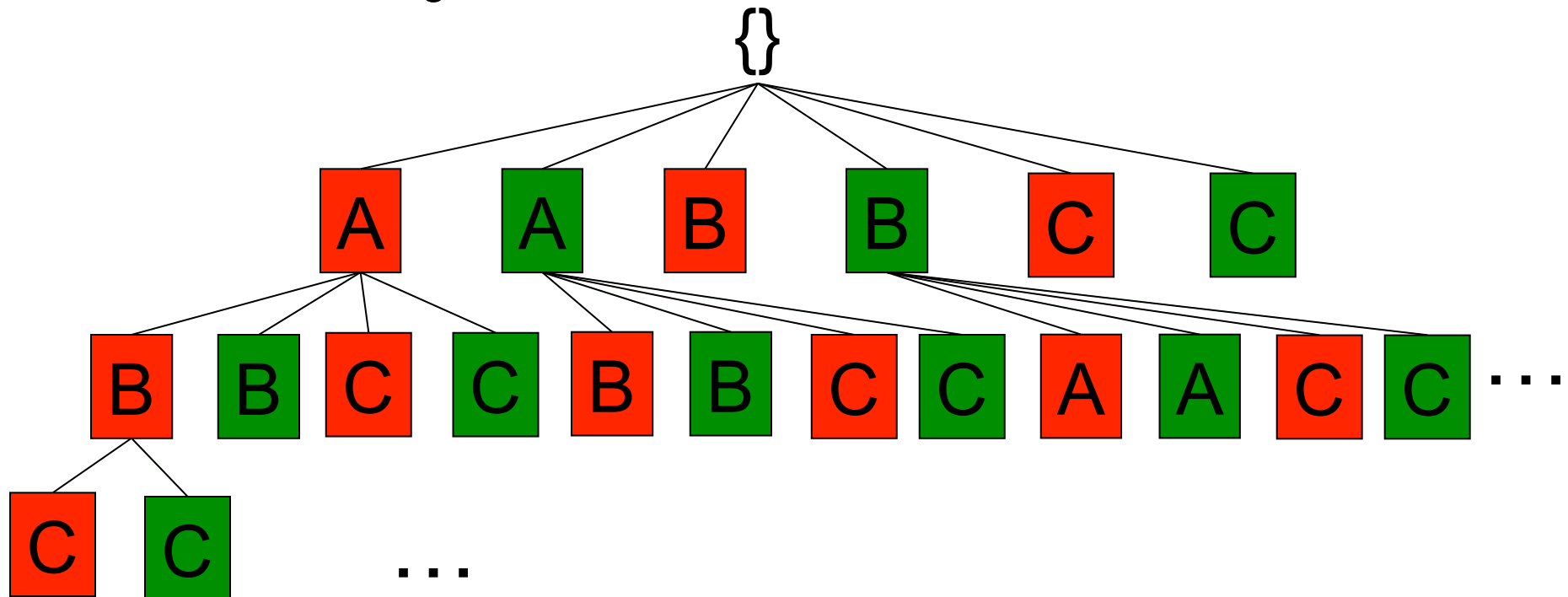
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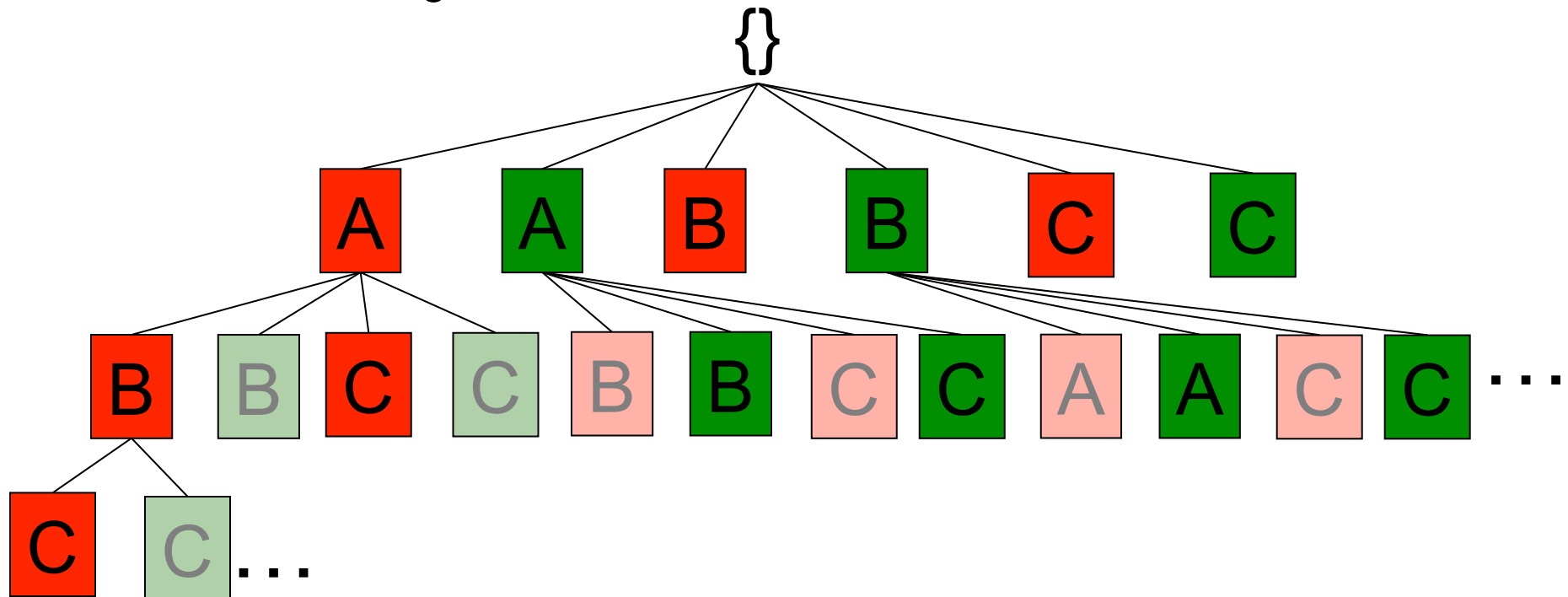
Improvement 2: Legal Assignments

- Idea 2: Only allow legal assignments at each point
 - Only assign values which don't *eventually* doom the search
 - Might have to do some extra computation
 - “Incremental goal test”



Improvement 2: Legal Assignments

- Idea 2: Only allow legal assignments at each point
 - Only assign values which do not conflict with existing assignments
 - Might have to do some extra computation
 - “Incremental goal test”



Idea 1 + Idea 2 = Backtracking

- Depth-first search for CSPs with these fixes is *backtracking search*
 - Backtrack when there's no legal assignment for the next variable
- Backtracking search is the basic uninformed algorithm for CSPs

```
function BACKTRACKING-SEARCH(csp) returns solution/failure
  return RECURSIVE-BACKTRACKING({ }, csp)

function RECURSIVE-BACKTRACKING(assignment, csp) returns soln/failure
  if assignment is complete then return assignment
  var ← SELECT-UNASSIGNED-VARIABLE(VARIABLES[csp], assignment, csp)
  for each value in ORDER-DOMAIN-VALUES(var, assignment, csp) do
    if value is consistent with assignment given CONSTRAINTS[csp] then
      add {var = value} to assignment
      result ← RECURSIVE-BACKTRACKING(assignment, csp)
      if result ≠ failure then return result
      remove {var = value} from assignment
  return failure
```

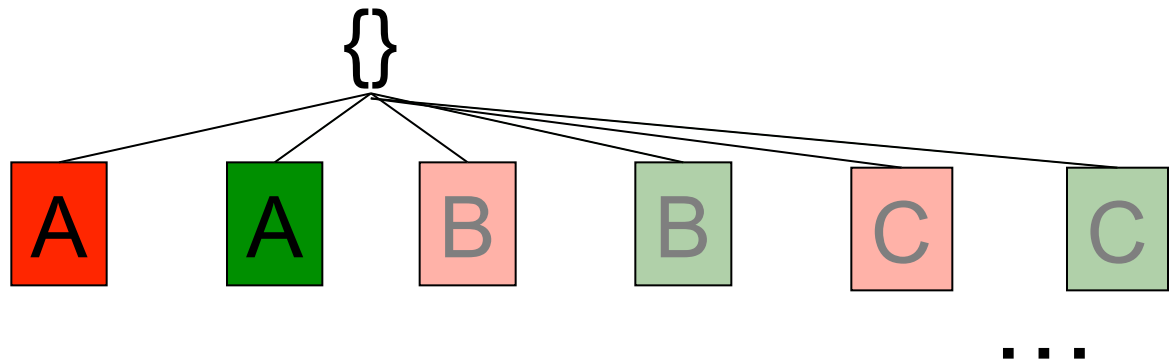
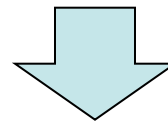
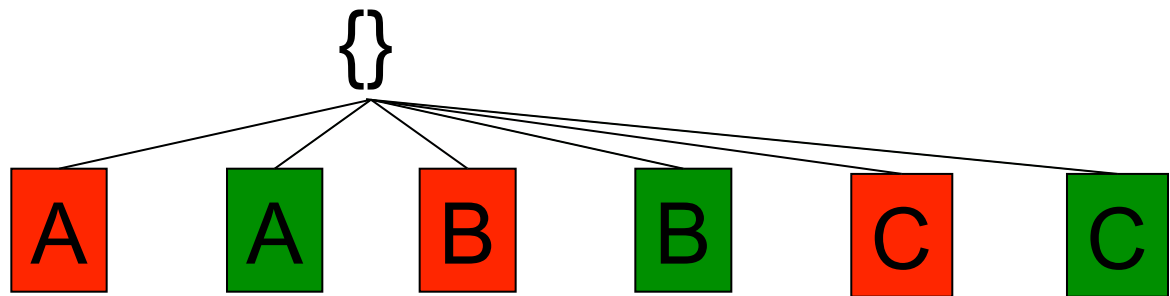
Backtracking

- Idea 1

$X = \{A, B, C\}$

$D = \{\text{red}, \text{green}\}$

Goal: $A = B = C$



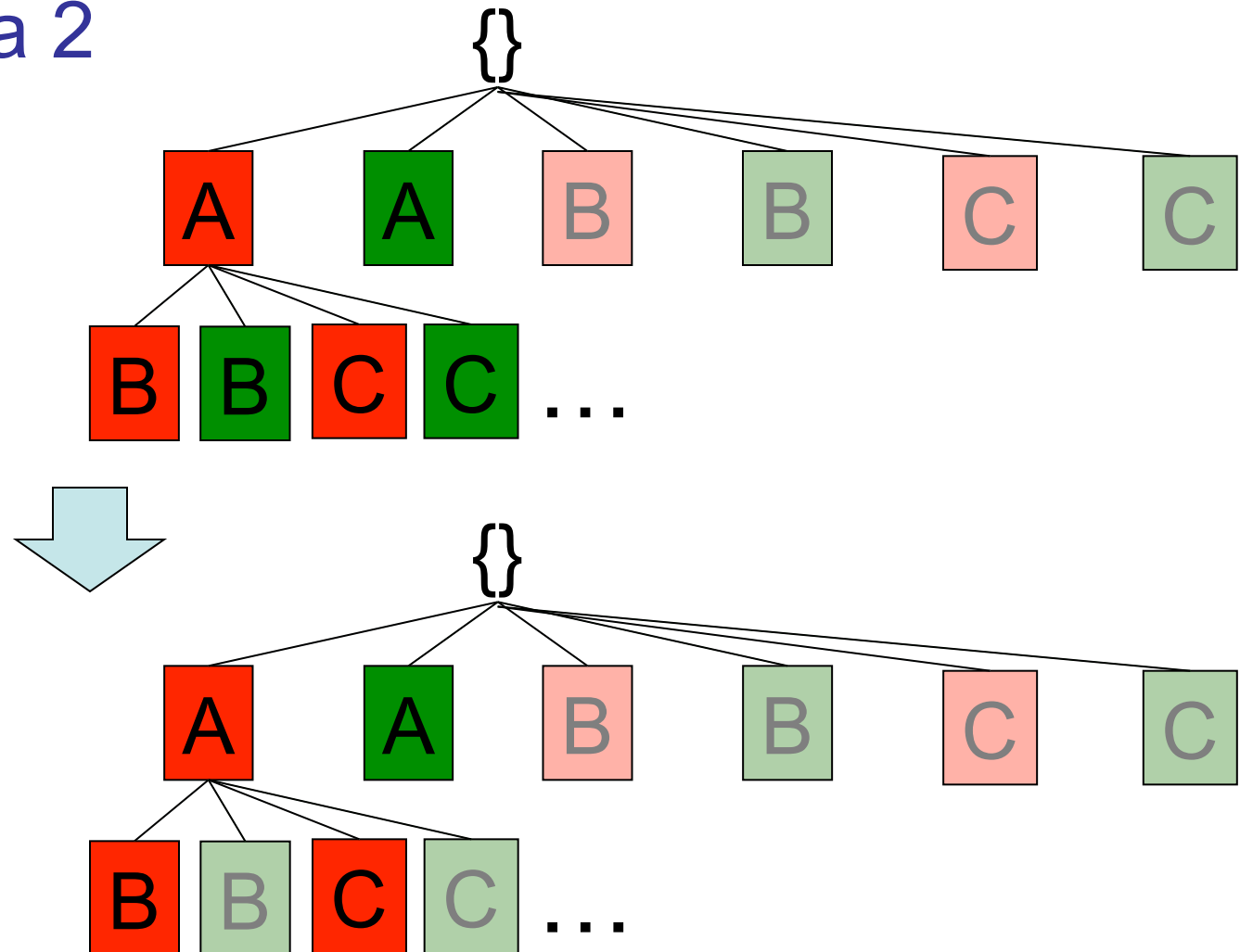
Backtracking

■ Plus Idea 2

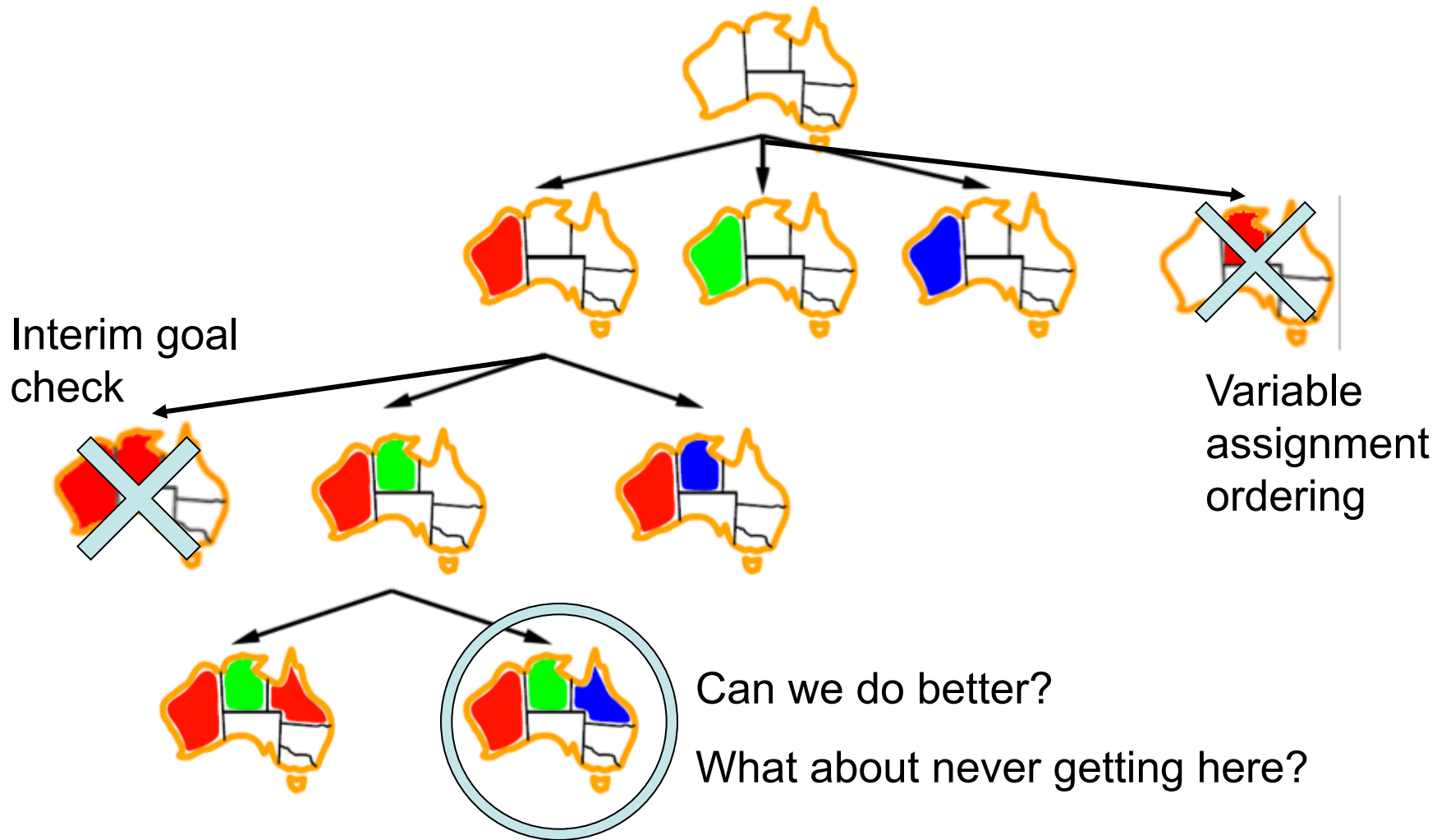
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Backtracking Example

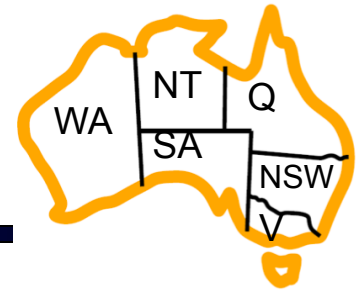


Forward Checking

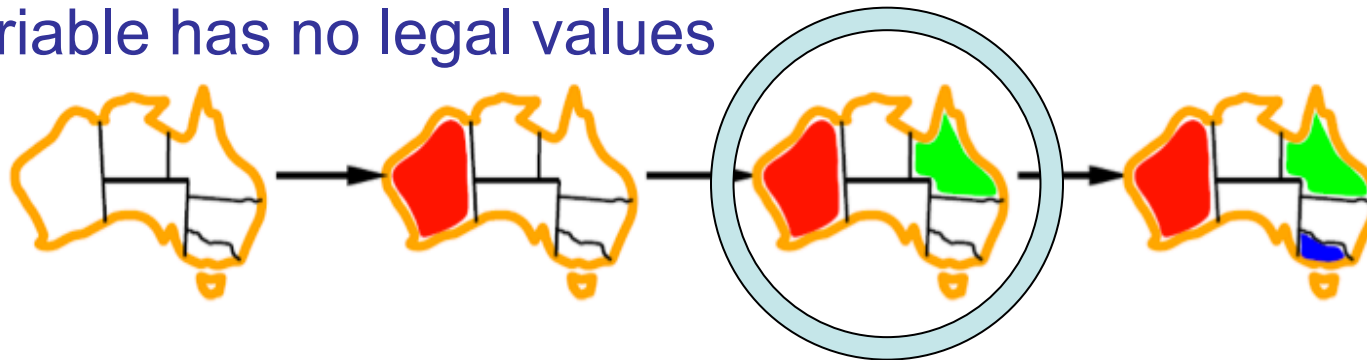
- Idea: Keep track of remaining legal values for unassigned variables (using immediate constraints)
- Terminate/prune when any *as-yet-unassigned* variable has no legal values



Forward Checking



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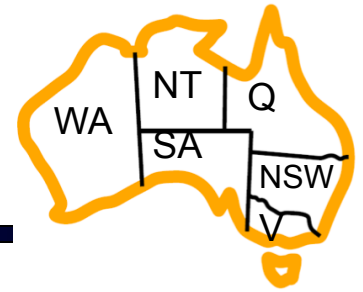
WA	NT	Q	NSW	V	SA	T
Red	Green	Blue	Red	Green	Blue	Red
Red	Green	Blue	Red	Green	Blue	Red
Red	Blue	Green	Red	Green	Blue	Red
Red	Blue	Green	Red	Blue	!	Red

Improving Forward Checking

- Why does forward checking allow this?
- Forward checking *propagates* information from assigned to adjacent unassigned variables, but doesn't detect more distant failures



Constraint Propagation



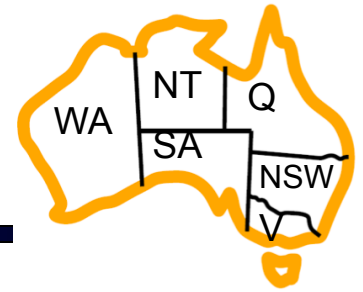
- Why does forward checking allow this?



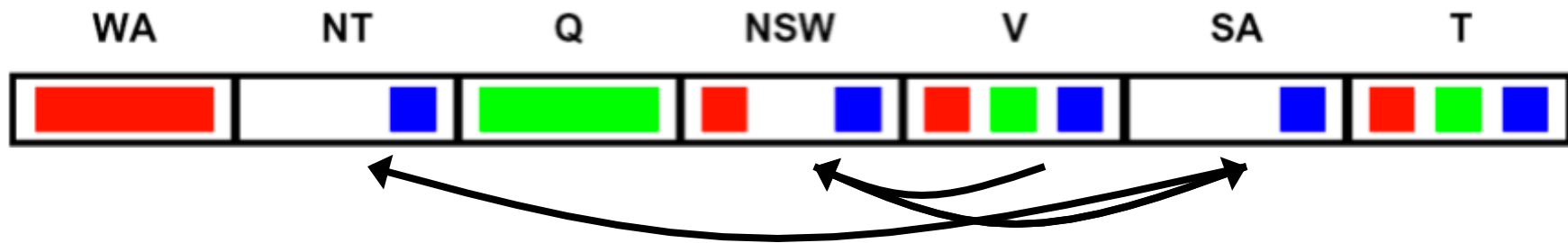
WA	NT	Q	NSW	V	SA	T
Red Green Blue	Red Green Blue	Red Green Blue	Red Green Blue	Red Green Blue	Red Green Blue	Red Green Blue
Red	Green Blue	Red Green Blue	Red Green Blue	Red Green Blue	Green Blue	Red Green Blue
Red	Blue	Green	Red Blue	Red Green Blue	Blue	Red Green Blue

- Neither SA nor NT have *no* possible assignments.
- How do we fix it?

Arc Consistency



- Simplest form of propagation makes each arc consistent
- Every pair of variables that affect each other share an arc
 - $X \rightarrow Y$ is consistent iff for every value x there is some allowed
- If X loses a value, neighbors of X need to be rechecked!

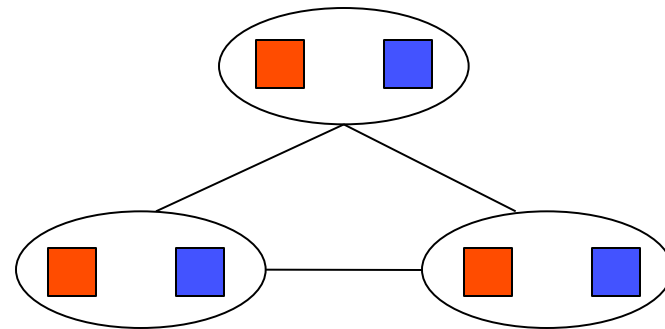
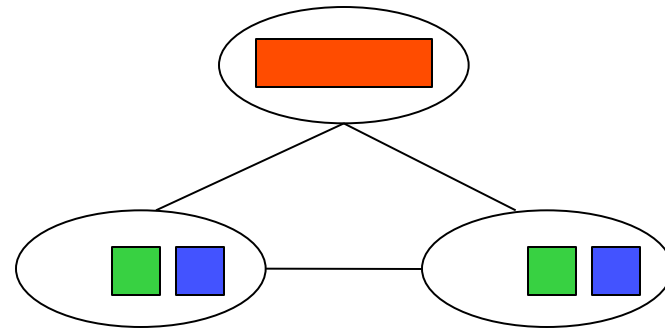


Revisiting and Reviewing

- Uninformed Search for Constraint Satisfaction Problems
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- k -Consistency
- Ordering Heuristics
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- Tree- and almost-tree CSPs

Limitations of Arc Consistency

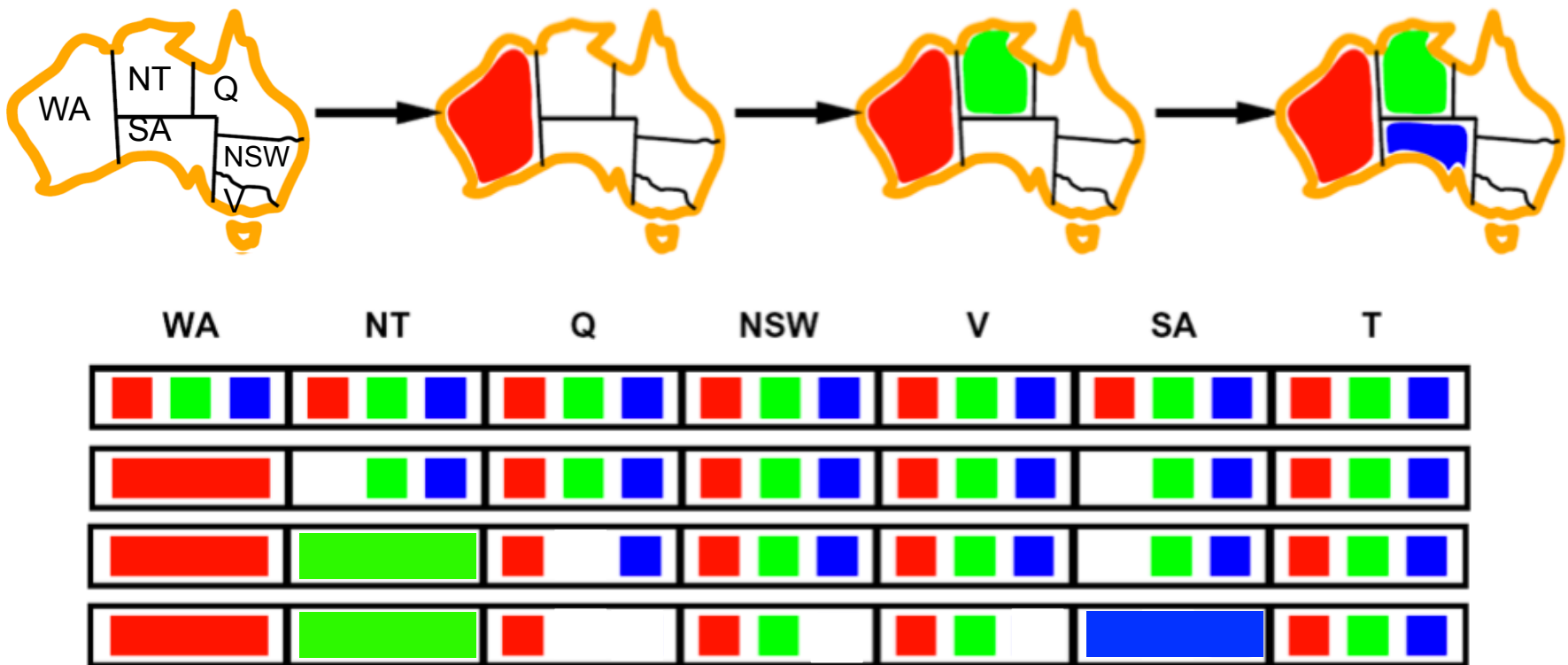
- After running arc consistency:
 - Can have one solution left
 - Can have multiple solutions left
 - Can have no solutions left (and not know it)
- How can we fix it?



What went wrong here?

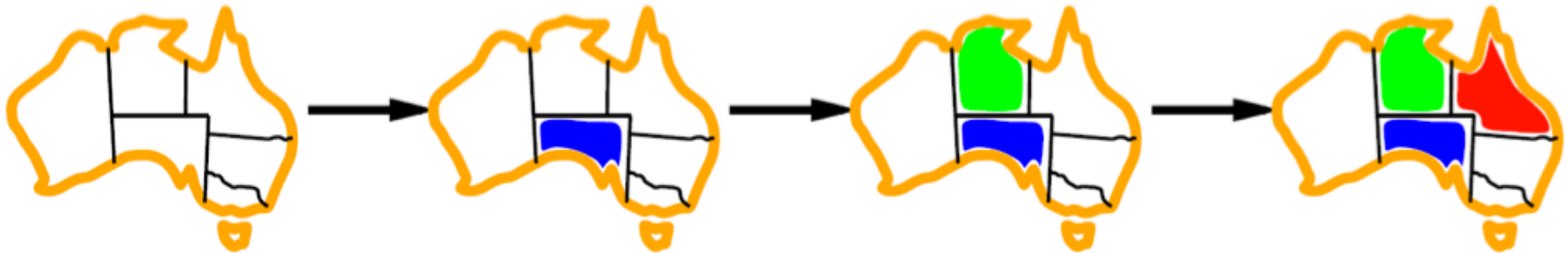
Variable Choice: Minimum Remaining Values

- Minimum remaining values (MRV):
 - Choose the variable with the fewest legal values



Ordering: Degree Heuristic

- Tie-breaker among MRV variables
- Degree heuristic:
 - Choose the variable participating in the most constraints on remaining variables (has the most arcs)



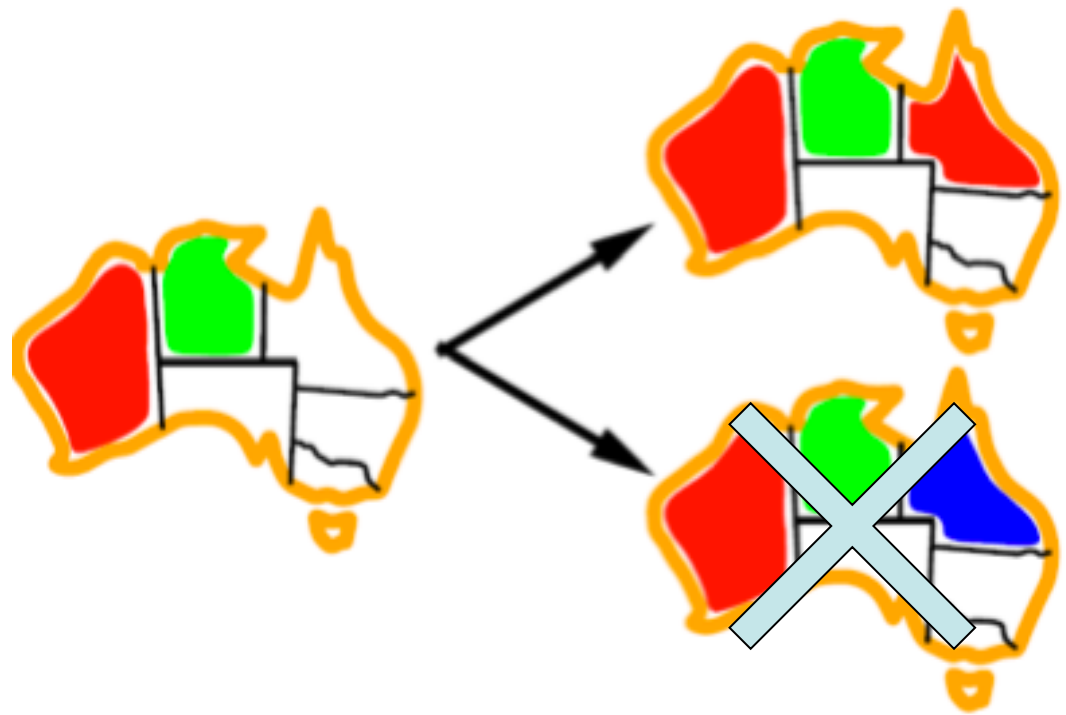
- Why most rather than fewest constraints?

Ordering: Least Constraining Value

- Given a choice of variable:
 - Choose the one that rules out the fewest values in the remaining variables

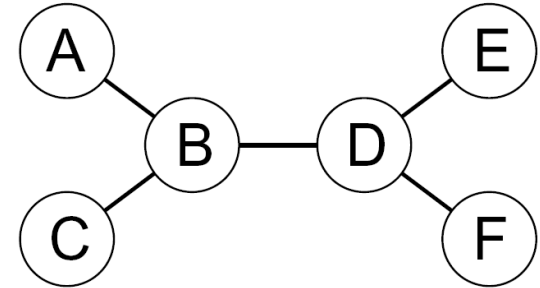
- Why?

- Computationally expensive (sometimes)



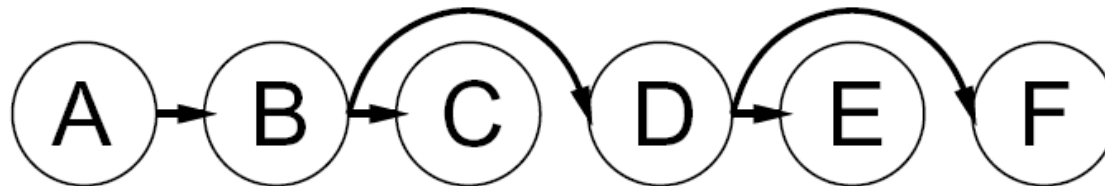
Tree-Structured CSPs

- Choose a variable as root, order variables from root to leaves such that every node's parent precedes it



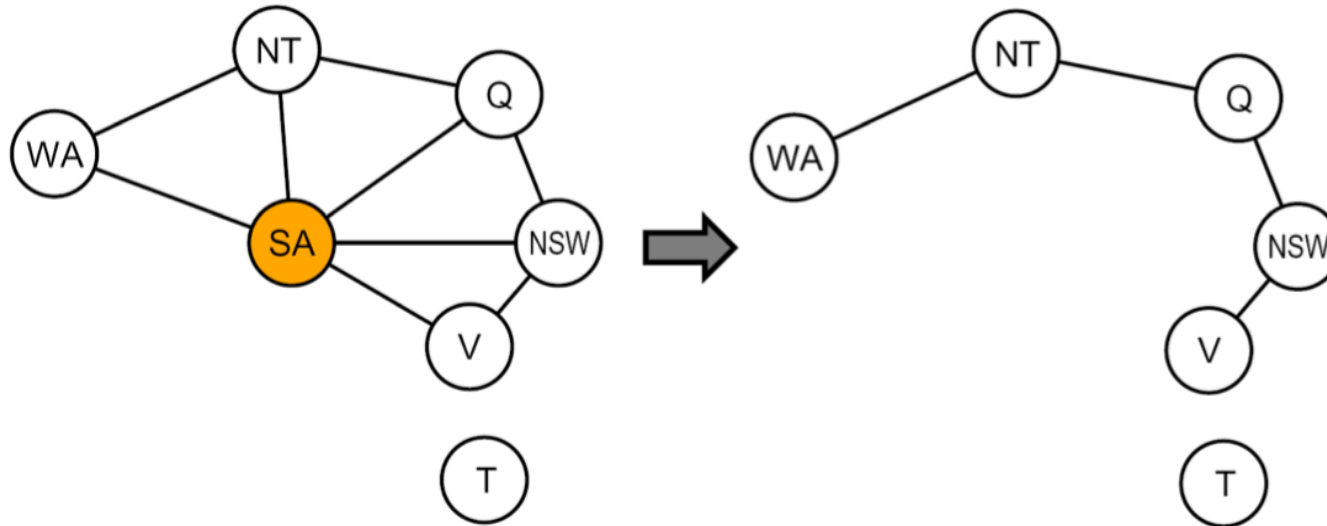
For $i = n : 2$, apply $\text{RemoveInconsistent}(\text{Parent}(X_i), X_i)$

For $i = 1 : n$, assign X_i consistently with $\text{Parent}(X_i)$



- Runtime: $O(n d^2)$
- Takeaway: tree-structured CSPs can be solved very efficiently

Nearly Tree-Structured CSPs



- **Cutset conditioning:**
 - Choose variable to instantiate that makes everything *left* into a tree
- **Instantiate a variable every possible way**
 - Here, you now have 3 tree-search problems
- **Takeaway: you can turn some CSPs into trees (which can still be solved very efficiently)**