Foundations of Artificial Intelligence 5. Constraint Satisfaction Problems CSPs as Search Problems, Solving CSPs, Problem Structure

Wolfram Burgard, Bernhard Nebel and Martin Riedmiller



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Constraint Satisfaction Problems

- A Constraint Satisfaction Problems (CSP) consists of
 - a set of variables $\{X_1, X_2, \dots, X_n\}$ to which
 - values $\{d_1, d_2, \dots, d_k\}$ can be assigned
 - such that a set of constraints over the variables is respected
- A CSP is solved by a variable assignment that satisfies all given constraints.
- In CSPs, states are explicitly represented as variable assignments. CSP search algorithms take advantage of this structure.
- The main idea is to exploit the constraints to eliminate large portions of search space.
- Formal representation language with associated general inference algorithms

Example: Map-Coloring



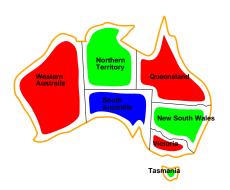
- Variables: WA, NT, SA, Q, NSW, V, T
- Values: $\{red, green, blue\}$
- Constraints: adjacent regions must have different colors, e.g., $NSW \neq V$

Australian Capital Territory (ACT) and Canberra (inside NSW)



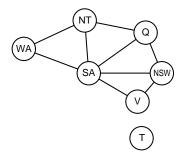
View of the Australian National University and Telstra Tower

One Solution



- Solution assignment:
 - $\{WA = red, NT = green, Q = red, NSW = green, V = red, SA = blue, T = green\}$
 - Perhaps in addition ACT = blue

Constraint Graph



- a constraint graph can be used to visualize binary constraints
- for higher order constraints, hyper-graph representations might be used
- Nodes = variables, arcs = constraints

Note: Our problem is three-colorability for a planar graph

Variations

- Binary, ternary, or even higher arity (e.g., ALL_DIFFERENT)
- Finite domains $(d \text{ values}) \rightarrow d^n$ possible variable assignments
- Infinite domains (reals, integers)
 - linear constraints (each variable occurs only in linear form): solvable (in P if real)
 - nonlinear constraints: unsolvable

Applications

- Timetabling (classes, rooms, times)
- Configuration (hardware, cars, ...)
- Spreadsheets
- Scheduling
- Floor planning
- Frequency assignments
- Sudoku
- . . .

Backtracking Search over Assignments

- Assign values to variables step by step (order does not matter)
- Consider only one variable per search node!
- DFS with single-variable assignments is called backtracking search
- ullet Can solve n-queens for n pprox 25

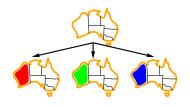
Algorithm

```
function BACKTRACKING-SEARCH(csp) returns a solution, or failure
  return BACKTRACK(\{\}, csp)
function BACKTRACK(assignment, csp) returns a solution, or failure
  if assignment is complete then return assignment
  var \leftarrow Select-Unassigned-Variable(csp)
  for each value in Order-Domain-Values(var, assignment, csp) do
      if value is consistent with assignment then
         add \{var = value\} to assignment
         inferences \leftarrow Inference(csp, var, value)
         if inferences \neq failure then
            add inferences to assignment
            result \leftarrow BACKTRACK(assignment, csp)
            if result \neq failure then
              return result
      remove \{var = value\} and inferences from assignment
  return failure
```

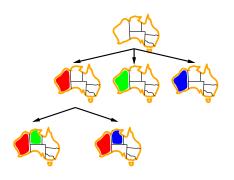
Example (1)



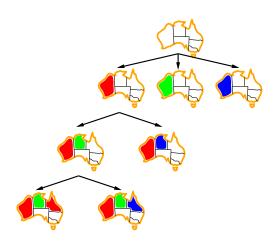
Example (2)



Example (3)



Example (4)



Improving Efficiency: CSP Heuristics & Pruning Techniques

- Variable ordering: Which one to assign first?
- Value ordering: Which value to try first?
- Try to detect failures early on
- Try to exploit problem structure
- → Note: all this is not problem-specific!

Variable Ordering: Most constrained first

- Most constrained variable:
 - choose the variable with the fewest remaining legal values
 - → reduces branching factor!



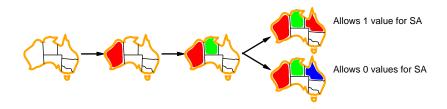
Variable Ordering: Most Constraining Variable First

- Break ties among variables with the same number of remaining legal values:
 - choose variable with the most constraints on remaining unassigned variables
 - → reduces branching factor in the next steps



Value Ordering: Least Constraining Value First

- Given a variable,
 - choose first a value that rules out the fewest values in the remaining unassigned variables
 - → We want to find an assignment that satisfies the constraints (of course, does not help if unsat.)

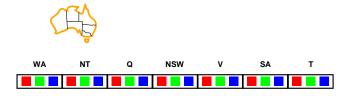


Rule out Failures early on: Forward Checking

- Whenever a value is assigned to a variable, values that are now illegal for other variables are removed
- Implements what the ordering heuristics implicitly compute
- WA = red, then NT cannot become red
- If all values are removed for one variable, we can stop!

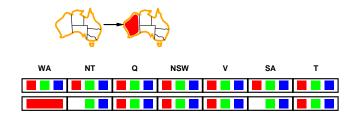
Forward Checking (1)

- Keep track of remaining values
- Stop if all have been removed



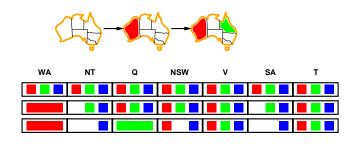
Forward Checking (2)

- Keep track of remaining values
- Stop if all have been removed



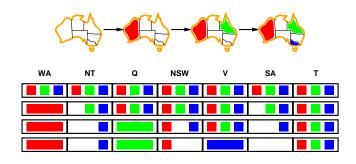
Forward Checking (3)

- Keep track of remaining values
- Stop if all have been removed



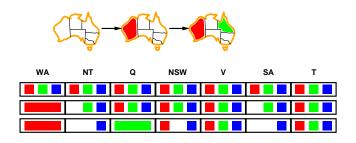
Forward Checking (4)

- Keep track of remaining values
- Stop if all have been removed



Forward Checking: Sometimes it Misses Something

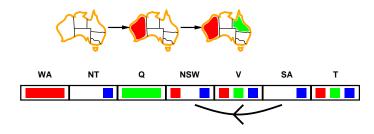
- Forward Checking propagates information from assigned to unassigned variables
- However, there is no propagation between unassigned variables



Arc Consistency

- A directed arc $X \to Y$ is "consistent" iff
 - for every value x of X, there exists a value y of Y, such that (x,y) satisfies the constraint between X and Y
- Remove values from the domain of X to enforce arc-consistency
- Arc consistency detects failures earlier
- Can be used as preprocessing technique or as a propagation step during backtracking

Arc Consistency Example



AC-3 Algorithm

```
(X_i, X_i) \leftarrow \text{REMOVE-FIRST}(queue)
     if REVISE(csp, X_i, X_i) then
       if size of D_i = 0 then return false
       for each X_k in X_i. NEIGHBORS - \{X_i\} do
          add (X_k, X_i) to queue
  return true
function REVISE(csp, X_i, X_j) returns true iff we revise the domain of X_i
  revised \leftarrow false
  for each x in D_i do
     if no value y in D_i allows (x,y) to satisfy the constraint between X_i and X_j then
       delete x from D_i
       revised \leftarrow true
  return revised
```

function AC-3(csp) **returns** false if an inconsistency is found and true otherwise

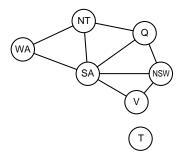
inputs: csp, a binary CSP with components (X, D, C) **local variables**: queue, a queue of arcs, initially all the arcs in csp

while queue is not empty do

Properties of AC-3

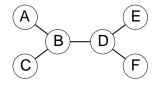
- AC-3 runs in $O(d^3n^2)$ time, with n being the number of nodes and d being the maximal number of elements in a domain
- Of course, AC-3 does not detect all inconsistencies (which is an NP-hard problem)

Problem Structure (1)



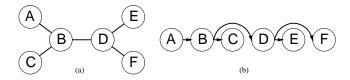
- CSP has two independent components
- Identifiable as connected components of constraint graph
- Can reduce the search space dramatically

Problem Structure (2): Tree-structured CSPs



- If the CSP graph is a tree, then it can be solved in $O(nd^2)$
 - ullet General CSPs need in the worst case $O(d^n)$
- Idea: Pick root, order nodes, apply arc consistency from leaves to root, and assign values starting at root

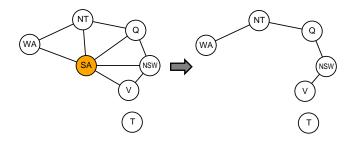
Problem Structure (2): Tree-structured CSPs



- Pick any variable as root; choose an ordering such that each variable appears after its parent in the tree.
- Apply arc-consistency to (X_i,X_k) , when X_i is the parent of X_k , for all k=n down to 2. (any tree with n nodes has n-1 arcs, per arc d^2 comparisons are needed: $O(n\,d^2)$)
- Now one can start at X_1 assigning values from the remaining domains without creating any conflict in one sweep through the tree!
- ullet Algorithm linear in n

Problem Structure (3): Almost Tree-structured

Idea: Reduce the graph structure to a tree by fixing values in a reasonably chosen subset



Instantiate a variable and prune values in neighboring variables is called Conditioning

Problem Structure (4): Almost Tree-structured

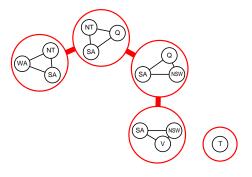
Algorithm Cutset Conditioning:

- Choose a subset S of the CSPs variables such that the constraint graph becomes a tree after removal of S. S is called a cycle cutset.
- Our Early Property of Property of States of
 - remove from the domains of the remaining variables any values that are inconsistent with the assignments for S, and
 - if the remaining CSP has a solution, return it together with the assignment for S

Note: Finding the smallest cycle cutset is NP hard, but several efficient approximation algorithms are known.

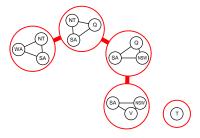
Another Method: Tree Decomposition (1)

- Decompose problem into a set of connected sub-problems, where two sub-problems are connected when they share a constraint
- Solve sub-problems independently and combine solutions



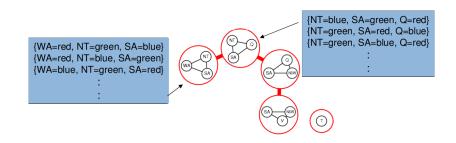
Another Method: Tree Decomposition (2)

- A tree decomposition must satisfy the following conditions:
 - Every variable of the original problem appears in at least one sub-problem
 - Every constraint appears in at least one sub-problem
 - If a variable appears in two sub-problems, it must appear in all sub-problems on the path between the two sub-problems
 - The connections form a tree



Another Method: Tree Decomposition (3)

- Consider sub-problems as new mega-variables, which have values defined by the solutions to the sub-problems
- Use technique for tree-structured CSP to find an overall solution (constraint is to have identical values for the same variable)



Tree Width

- ullet The aim is to make the subproblems as small as possible. Tree width w of a tree decomposition is the size of largest sub-problem minus 1
- Tree width of a graph is minimal tree width over all possible tree decompositions
- If a graph has tree width w and we know a tree decomposition with that width, we can solve the problem in $O(nd^{w+1})$
- Unfortunately, finding a tree decomposition with minimal tree width is NP-hard. However, there are heuristic methods that work well in practice.

Summary & Outlook

- CSPs are a special kind of search problem:
 - states are value assignments
 - goal test is defined by constraints
- Backtracking = DFS with one variable assigned per node. Other intelligent backtracking techniques possible
- Variable/value ordering heuristics can help dramatically
- Constraint propagation prunes the search space
- Path-consistency is a constraint propagation technique for triples of variables
- Tree structure of CSP graph simplifies problem significantly
- Cutset conditioning and tree decomposition are two ways to transform part of the problem into a tree
- CSPs can also be solved using local search