

Constraint Satisfaction Problems/Programming

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State space search → CSP

- until now – no assumptions on the states
- CSP:
 - state → array of variables
 - context → domains for variables, set of constraints among the variables
 - goal → all variables have an assigned value, no constraint is violated
- alternative goals
 - optimization variant

Advantages/Disadvantages

- general enough to model many problems
- more efficient algorithms that exploit the structure
 - we can do better than DFS/BFS
 - advanced search techniques can be possibly reused in non-CSP problems
 - generic purpose CSP solvers
- not all problems can be modeled as CSP
- some formulations can be inefficient

Baseline Algorithm

```
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
```

Example

- map coloring, satisfiability, ...
- sudoku, crypto-arithmetic
- scheduling requests to hotel rooms
 - list of requests (# of people, from, to)
 - set of rooms (# of beds)

Binary CSP

- we can do better than passively checking constraints
- unified structure of CSP instances
 - all constraints are binary
 - can we formalize all CSP problems with binary variables?
- constraints can be represented as a graph
 - nodes – variables
 - edges – constraints between variables

Consistency

- nodes
- edges / arcs

Consistency

- nodes
 - unary constraints are reflected in the domain of the variable
 - every value in the domain satisfies all unary constraints
- edges / arcs
 - directed $v_i \rightarrow v_j$
 - for every x from v_i domain there exists a value y from v_j domain that satisfies all constraints between v_i and v_j

Arc Consistency

- make each edge in the graph of constraints consistent

Arc consistency algorithm

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**

$(X_i, X_j) \leftarrow \text{REMOVE-FIRST}(\textit{queue})$

if REMOVE-INCONSISTENT-VALUES(X_i, X_j) **then**

for each X_k **in** NEIGHBORS[X_i] **do**

 add (X_k, X_i) to *queue*

function REMOVE-INCONSISTENT-VALUES(X_i, X_j) **returns** true iff succeeds

removed \leftarrow false

for each x **in** DOMAIN[X_i] **do**

if no value y in DOMAIN[X_j] allows (x, y) to satisfy the constraint $X_i \leftrightarrow X_j$

then delete x from DOMAIN[X_i]; *removed* \leftarrow true

return *removed*

