Solving Problems by Searching

Solving Problems by Searching AIMA Sections 3.1–3.3

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Outline

Solving Problems by Searching

- \diamond Problem-solving agents
- \diamond Problem types
- \diamond Problem formulation
- \diamond Example problems
- \diamond General search algorithm

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Problem-solving agents

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```
function Simple-Problem-Solving-Agent(percept) returns an
action
   static: seq, an action sequence, initially empty
            state, some description of the current world state
            goal, a goal, initially null
            problem, a problem formulation
   state \leftarrow Update-State(state, percept)
   if seq is empty then
         goal \leftarrow Formulate-Goal(state)
         problem \leftarrow Formulate-Problem(state, goal)
         seq \leftarrow Search(problem)
   action \leftarrow First(seq)
   seq \leftarrow \text{Rest}(seq)
   return action
```

Problem-solving agents

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Restricted form of general agent: Goal based agents

- formulate a goal and a problem given the current state
- search for a solution
- execute the solution ignoring perceptions

Note: this is offline problem solving; solution executed "eyes closed."

Online problem solving involves acting without complete knowledge.

An example: Traveling in Romania

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Example (Holidays in Romania)

On holiday in Romania; currently in Arad. Flight leaves tomorrow from Bucharest Formulate goal: be in Bucharest Formulate problem: states: various cities actions: drive between cities Find solution:

sequence of cities, e.g., Arad, Sibiu, Fagaras, Bucharest

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An example: Traveling in Romania

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Problem types

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Deterministic, fully observable \implies single-state problem

Agent knows exactly which state it will be in; solution is a sequence

$\mathsf{Non-observable} \Longrightarrow \mathsf{conformant} \ \mathsf{problem}$

Agent may have no idea where it is; solution (if any) is a sequence

Nondeterministic and/or partially observable \implies contingency problem

percepts provide **new** information about current state solution is a contingent plan or a policy

often interleave search, execution

Unknown state space \implies exploration problem ("online")



Example: vacuum world

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Single-state, start in #5. <u>Solution</u>?? [*Right*, *Suck*]

Conformant start in $\{1, 2, 3, 4, 5, 6, 7, 8\}$ e.g., *Right* goes to $\{2, 4, 6, 8\}$. Solution??



Example: vacuum world

Solving Problems by Searching

Single-state, start in #5. <u>Solution</u>?? [*Right*, *Suck*]

Conformant start in $\{1, 2, 3, 4, 5, 6, 7, 8\}$ e.g., *Right* goes to $\{2, 4, 6, 8\}$. <u>Solution</u>?? [*Right*, *Suck*, *Left*, *Suck*]

Contingency, start in #5 Murphy's Law: *Suck* can dirty a clean carpet Local sensing: dirt, location only. <u>Solution</u>??



Example: vacuum world

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Single-state, start in #5. <u>Solution</u>?? [*Right*, *Suck*]

Conformant

start in {1,2,3,4,5,6,7,8}
e.g., Right goes to {2,4,6,8}.
Solution??
[Right, Suck, Left, Suck]

Contingency, start in #5 Murphy's Law: *Suck* can dirty a clean carpet Local sensing: dirt, location only. <u>Solution</u>?? [*Right*, if *dirt* then *Suck*]



Single-state problem formulation

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A problem is defined by four items: initial state e.g., "at Arad" successor function S(x) = set of action-state pairse.g., $S(A) = \{ < Arad \rightarrow Zerind, Zerind >, \ldots \}$ goal test, can be explicit, e.g., x = "at Bucharest" implicit, e.g., NoDirt(x)path cost (additive) e.g., sum of distances, number of actions executed, etc. c(x, a, y) is the step cost, assumed to be > 0A solution is a sequence of actions leading from the initial state to a goal state

Selecting a state space

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Real world is absurdly complex

 \Rightarrow state space must be **abstracted** for problem solving (Abstract) state = set of real states

(Abstract) action = complex combination of real actions

e.g., "Arad \rightarrow Zerind" represents a complex set

of possible routes, detours, rest stops, etc.

For guaranteed realizability, **any** real state "in Arad" must get to some real state "in Zerind" (Abstract) solution =

set of real paths that are solutions in the real world Each abstract action should be "easier" than the original problem!





states??: discrete dirt and robot locations (ignore dirt amounts)
actions??:
goal test??:
path cost??:



states??: discrete dirt and robot locations (ignore dirt amounts)
actions??: Left, Right, Suck, NoOp
goal test??:
path cost??:

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states??: discrete dirt and robot locations (ignore dirt amounts)
actions??: Left, Right, Suck, NoOp
goal test??: no dirt
path cost??:



states??: discrete dirt and robot locations (ignore dirt amounts)
actions??: Left, Right, Suck, NoOp
goal test??: no dirt
path cost??: 1 per action (0 for NoOp)

path cost??:



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Start State

Goal State

states??: integer locations of tiles (ignore intermediate
positions)
actions??:
goal test??:
path cost??:







Start State

Goal State

states??: integer locations of tiles (ignore intermediate positions) actions??: move blank left, right, up, down goal test??: path cost??:







Start State

Goal State

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states??: integer locations of tiles (ignore intermediate positions) actions??: move blank left, right, up, down goal test??: given goal state path cost??:







Start State

Goal State

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states??: integer locations of tiles (ignore intermediate positions) actions??: move blank left, right, up, down goal test??: given goal state path cost??: 1 per move







Start State

Goal State

states??: integer locations of tiles (ignore intermediate positions) actions??: move blank left, right, up, down goal test??: given goal state path cost??: 1 per move

[Note: optimal solution of *n*-Puzzle family is NP-hard]

Example: robotic assembly

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states??: real-valued coordinates of robot joint angles
parts of the object to be assembled
actions??: continuous motions of robot joints
goal test??: complete assembly with no robot included!
path cost??: time to execute

Tree search algorithm

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Basic idea:

offline, simulated exploration of state space by generating successors of already-explored states (a.k.a. expanding states)

function Tree-Search(*problem*, *strategy*) **returns** a solution, or failure

initialize the search tree using the initial state of *problem* **loop do**

if no candidates for expansion then return failure choose a leaf node for expansion according to *strategy* if node contains a goal state then return the solution else add successor nodes to the search tree (expansion) end

Tree search example



Tree search example



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Tree search example



Implementation: states vs. nodes

Solving Problems by Searching A state is a (representation of) a physical configuration
A node is a data structure constituting part of a search tree includes parent, action, children, depth, path cost (i.e., g(x))
States do not have parents, actions, children, depth, or path cost!

Implementation: states vs. nodes

Solving Problems by Searching A state is a (representation of) a physical configuration
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States do not have parents, actions, children, depth, or path cost!



The Expand function creates new nodes, filling in the various fields and using the SuccessorFn of the problem to create the corresponding states.

Implementation: general tree search

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> function Tree-Search(problem, frontier) returns a solution, or failure frontier ← Insert(Make-Node(problem.Initial-State)) loop do if frontier is empty then return failure node ← Pop(frontier) if problem.Goal-Test(node.State) then return node frontier ← InsertAll(Expand(node, problem)) end loop

Implementation: expand nodes

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function Expand(node, problem) returns a set of nodes $successors \leftarrow$ the empty set for each action, result in Successor-Fn(problem, node.State) do $s \leftarrow a$ new Node s.Parent-Node \leftarrow *node*: s.Action \leftarrow action; s. State \leftarrow result s Path-Cost \leftarrow node Path-Cost +Step-Cost(node.State, action, result) $s.Depth \leftarrow node.Depth + 1$ add s to successors return successors

Search strategies

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> A strategy is defined by picking the order of node expansion Strategies are evaluated along the following dimensions: completeness—does it always find a solution if one exists? time complexity—number of nodes generated/expanded space complexity—maximum number of nodes in memory optimality—does it always find a least-cost solution? Time and space complexity are measured in terms of b—maximum branching factor of the search tree d—depth of the least-cost solution m—maximum depth of the state space (may be ∞)

Repeated states

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Failure to detect repeated states can turn a linear problem into an exponential one!

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Graph search

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function Graph-Search(problem, frontier) returns a solution, or failure *explored* \leftarrow an empty set *frontier* ← Insert(Make-Node(*problem*.Initial-State)) loop do if frontier is empty then return failure $node \leftarrow \mathsf{Pop}(frontier)$ if problem.Goal-Test(node.State) then return node if node.State is not in explored then add node.State to explored $frontier \leftarrow \text{InsertAll}(\text{Expand}(node, problem))$ end if end loop

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