Chapter 3 Solving problems by searching



Search

- We will consider the problem of designing goal-based agents in observable, deterministic, discrete, known environments
 - > The solution is a fixed sequence of actions
 - Search is the process of looking for the sequence of actions that reaches the goal
 - Once the agent begins executing the search solution, it can ignore its percepts (open-loop system)

Search problem components

- Initial state
- Actions
- Transition model
 - What is the result of performing a given action in a given state?
- Goal state
- Path cost
 - Assume that it is a sum of nonnegative step costs



• The **optimal solution** is the sequence of actions that gives the lowest path cost for reaching the goal

Example: Romania

- On vacation in Romania; currently in Arad
- Flight leaves tomorrow from Bucharest
- Initial state
 - Arad
- Actions
 - Go from one city to another
- Transition model
 - If you go from city A to city B, you end up in city B
- Goal state
 - Bucharest
- Path cost
 - Sum of edge costs





State space

- The initial state, actions, and transition model define the state space of the problem
 - The set of all states reachable from initial state by any sequence of actions
 - Can be represented as a directed graph where the nodes are states and links between nodes are actions
- What is the state space for the Romania problem?

Example: Vacuum world



- States
 - Agent location and dirt location
 - How many possible states?
 - What if there are *n* possible locations?
- Actions
 - Left, right, suck
- Transition model

Vacuum world state space graph



7

Example: The 8-puzzle

States

- Locations of tiles
 - 8-puzzle: 181,440 states
 - 15-puzzle: 1.3 trillion states
 - 24-puzzle: 10²⁵ states
- Actions

– Move blank left, right, up, down

- Path cost
 - -1 per move



Start State



Goal State

Optimal solution of n-Puzzle is NP-hard

Example: Robot motion planning



- States
 - Real-valued coordinates of robot joint angles
- Actions
 - Continuous motions of robot joints
- Goal state
 - Desired final configuration (e.g., object is grasped)
- Path cost
 - Time to execute, smoothness of path, etc.

Other Real-World Examples

- Routing
- Touring
- VLSI layout
- Assembly sequencing
- Protein design

Tree Search

- Let's begin at the start node and expand it by making a list of all possible successor states
- > Maintain a **fringe** or a list of unexpanded states
- > At each step, pick a state from the fringe to expand
- Keep going until you reach the goal state
- > Try to expand as few states as possible

Search tree

- "What if" tree of possible actions and outcomes
- The root node corresponds to the starting state
- The children of a node correspond to the successor states of that node's state
- A path through the tree corresponds to a sequence of actions
 - A solution is a path ending in the goal state



Tree Search Algorithm Outline

- Initialize the **fringe** using the **starting state**
- While the fringe is not empty
 - Choose a fringe node to expand according to search strategy
 - If the node contains the **goal state**, return solution
 - Else expand the node and add its children to the fringe

Tree search example





Tree search example





Tree search example





Search strategies

- A search 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?
 - Optimality: does it always find a least-cost solution?
 - Time complexity: number of nodes generated
 - **Space complexity:** maximum number of nodes in memory
- 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 length of any path in the state space (may be infinite)

Uninformed search strategies

Uninformed search strategies use only the information available in the problem definition

- Breadth-first search
- Uniform-cost search
- Depth-first search
- Iterative deepening search

- Expand shallowest unexpanded node
- Implementation:
 - *fringe* is a FIFO queue, i.e., new successors go at end



- Expand shallowest unexpanded node
- Implementation:
 - *fringe* is a FIFO queue, i.e., new successors go at end



- Expand shallowest unexpanded node
- Implementation:
 - *fringe* is a FIFO queue, i.e., new successors go at end



- Expand shallowest unexpanded node
- Implementation:
 - *fringe* is a FIFO queue, i.e., new successors go at end



- Expand shallowest unexpanded node
- Implementation:
 - *fringe* is a FIFO queue, i.e., new successors go at end



Properties of breadth-first search

Complete?

Yes (if branching factor *b* is finite)

• Optimal?

Yes – if cost = 1 per step

• Time?

Number of nodes in a *b*-ary tree of depth d: $O(b^d)$ (*d* is the depth of the optimal solution)

• **Space?** *O*(*b^d*)

• Space is the bigger problem (more than time)

Uniform-cost search

- Expand least-cost unexpanded node
- Implementation: *fringe* is a queue ordered by path cost (priority queue)
- Equivalent to breadth-first if step costs all equal
- Complete?

Yes, if step cost is greater than some positive constant ε

• Optimal?

Yes – nodes expanded in increasing order of path cost

• Time?

Number of nodes with path cost \leq cost of optimal solution (C^*), $O(b^{C^*/\varepsilon})$ This can be greater than $O(b^d)$: the search can explore long paths consisting of small steps before exploring shorter paths consisting of larger steps

• Space?

 $O(b^{C^*/\varepsilon})$

- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front



Properties of depth-first search

• Complete?

Fails in infinite-depth spaces, spaces with loops Modify to avoid repeated states along path → complete in finite spaces

• Optimal?

No – returns the first solution it finds

• Time?

Could be the time to reach a solution at maximum depth $m: O(b^m)$ Terrible if m is much larger than dBut if there are lots of solutions, may be much faster than BFS

• Space?

O(bm), i.e., linear space!

- Use DFS as a subroutine
 - 1. Check the root
 - 2. Do a DFS searching for a path of length 1
 - 3. If there is no path of length 1, do a DFS searching for a path of length 2
 - 4. If there is no path of length 2, do a DFS searching for a path of length 3...

Limit = 0









Properties of iterative deepening search

Complete?

Yes

Optimal?

Yes, if step cost = 1

• Time?

 $(d+1)b^0 + db^1 + (d-1)b^2 + \dots + b^d = O(b^d)$

• Space?

O(bd)

Informed search

- Idea: give the algorithm "hints" about the desirability of different states
 - Use an *evaluation function* to rank nodes and select the most promising one for expansion
- Greedy best-first search
- A* search

Heuristic function

- Heuristic function h(n) estimates the cost of reaching goal from node n
- Example:



Heuristic for the Romania problem



Greedy best-first search

 Expand the node that has the lowest value of the heuristic function h(n)

















Properties of greedy best-first search

• Complete?

No – can get stuck in loops



Properties of greedy best-first search

• Complete?



Properties of greedy best-first search

• Complete?

No – can get stuck in loops

• Optimal?

No

• Time?

Worst case: $O(b^m)$

Best case: O(bd) - If h(n) is 100% accurate

• Space?

Worst case: $O(b^m)$

A^{*} search

- Idea: avoid expanding paths that are already expensive
- The evaluation function *f*(*n*) is the estimated total cost of the path through node *n* to the goal:

$$f(n) = g(n) + h(n)$$

g(n): cost so far to reach n (path cost)
h(n): estimated cost from n to goal (heuristic)

























Properties of A*

• Complete?

Yes – unless there are infinitely many nodes with $f(n) \leq C^*$

• Optimal?

Yes

• Time?

Number of nodes for which $f(n) \leq C^*$ (exponential)

• Space?

Exponential

Admissible heuristics

- A heuristic h(n) is admissible if for every node n, h(n)
 ≤ h*(n), where h*(n) is the true cost to reach the goal state from n
- An admissible heuristic never overestimates the cost to reach the goal, i.e., it is optimistic
- Example: straight line distance never overestimates the actual road distance
- Theorem: If *h*(*n*) is admissible, A^{*} is optimal

Designing heuristic functions

• Heuristics for the 8-puzzle

 $h_1(n)$ = number of misplaced tiles

 $h_2(n)$ = total Manhattan distance (number of squares from desired location of each tile)



Start State

 3
 4
 5

 6
 7
 8

2

Goal State

$$h_1(\text{start}) = 8$$

 $h_2(\text{start}) = 3+1+2+2+3+3+2 = 18$

• Are h_1 and h_2 admissible?

Comparison of search strategies

Algorithm	Complete?	Optimal?	Time complexity	Space complexity
BFS	Yes	If all step costs are equal	O(b ^d)	O(b ^d)
UCS	Yes	Yes	Number of nod	es with g(n) ≤ C*
DFS	No	No	O(b ^m)	O(bm)
IDS	Yes	If all step costs are equal	O(b ^d)	O(bd)
Greedy	No	No	Worst case: O(b ^m) Best case: O(bd)	
A *	Yes	Yes	Number of nodes	with g(n)+h(n) $\leq C^*$