

STATES & MANAGEMENT

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Introduction to Artificial Intelligence IFT3335 Lecture 3: Uninformed Search Bang Liu, Jian-Yun Nie

THE PARTY





- Slides from UC Berkeley CS188, Dan Klein and Pieter Abbeel
 - Uninformed Search: <u>https://inst.eecs.berkeley.edu/~cs188/su21/</u>



- Agents to solve problems
 - Types of problem
 - Problem formulation
- Sample search problems
- Problem solving by searching
 - Breadth-First Search
 - Depth-First Search
 - Uniform-Cost Search

Agents to Solve Problems

5 Problem-Solving Agents

- Output Description of the second state of t
- The computational process it undertakes is called **search**.
- **Problem-solving agents** use **atomic** representations: states of the world are considered as wholes
 - Planning agents use factored or structured representations of states
- Informed v.s. Uninformed algorithm: whether the agent can estimate how far it is from the goal

• Problem-Solving Agent is a Goal-Based Agent that decides what to do by finding sequences of

Example: Travel in Romania 6

- On vacation in Romania; currently in Arad.
- Formulate the goal:
 - Be in Bucharest
- Formulate the problem:
 - **States**: be in different cities
 - Actions: driving between cities
- Find a solution:
 - Sequence of cities, e.g., Arad, Sibiu, Fagaras, Bucharest



7 Problem-Solving Agents





A search problem is defined by:

1. **State space** A set of possible states the environment can be in.

Initial state 2.

Where the agent starts in

Goal test 3.

> Usually a condition, sometimes the description of a state

Successor function 4.

Include action and cost: describe how state changes with an action, and the cost of applying action a to transit from state s to s'

A solution is a sequence of actions (a plan) which transforms the start state to a goal state







9 What is in a State Space?





A search state keeps only the details needed for planning (abstraction)

- Problem: Eat-All-Dots
 - States: {(x,y), dot booleans}
 - Actions: NSEW
 - Successor: update location and possibly a dot boolean
 - Goal test: dots all false



- Our formulation of the problem (e.g., getting to Bucharest) is a **model**—an abstract mathematical description—and not the real thing.
- The process of removing detail from a representation is called **abstraction**.



Go around the city taking each bridge once and only once? In 1735, Euler showed that this is impossible using graph theory and topology. Solution in graph theory: for this to be possible, each node must have an even degree.

Example: Route finding 11



Problem: find path from Arad to Bucharest

12 Example: Route finding



13 Example: Route finding

Formul	late	σο	•
	all	guai	L •

• Be in Bucharest

Formulate problem:	ormulate problem:	
• States : various cities		

- Actions: go to adjacent city
- **Goal test**: whether in Bucharest
- Path cost: distance

• Find solution:

- Sequence of cities
- e.g., Arad, Sibiu, Fagaras, Bucharest



Formulate tasks as **Searching**

Where is the Map? State Space Graphs and Search Trees

State space graph 16

- State space graph: A mathematical representation of a search problem
 - Nodes are (abstracted) world configurations
 - Arcs represent successors (action results)
 - The goal test is a set of goal nodes (maybe only one)
- In a state space graph, each state occurs only once!
- We can rarely build this full graph in memory (it's too big), but it's a useful idea





State space graph for Pacman: eat all the dots



Tiny state space graph for a tiny search problem





• A search tree:

- A "what if" tree of plans and their outcomes
- The start state is the root node
- Children correspond to successors
- Nodes show states, but correspond to PLANS that achieve those states
- For most problems, we can never actually build the whole tree



This is now / start

Possible futures

19 State Space Graphs vs. Search Trees



Each NODE in in the search tree is an entire PATH in the state space graph.

We construct both on demand – and we construct as little as possible.





Consider this 4-state graph:



How big is its search tree (from S)?





Consider this 4-state graph:



How big is its search tree (from S)?





Consider this 4-state graph:



How big is its search tree (from S)?

Important: Lots of repeated structure in the search tree!

Sample Search Problems

24 Example: Vacuum World

- Formulate problem:
 - **States**: ?
 - Actions: ?
 - Goal test: ?
 - Path cost: ?



25 Example: Vacuum World

• Formulate problem:

- **States**: dirt and robot location
- Actions: left (L), right (R), suck (S)
- Goal test: no dirt at all locations
- **Path cost**: 1 per action



26 Example: The 8-puzzle

• Formulate problem:

- **States**: location of tiles
- Actions: move blank left, right, up, down
- Goal test: given goal state
- Path cost: 1 per move

8	2	
3	4	7
5	1	6

Initial state









Size of the state space = 181,440

15-puzzle \rightarrow .65 x 10¹²

24-puzzle $\rightarrow .5 \times 10^{25}$ 12 billio



Example: Robotic Assembly 28

• Formulate problem:

- **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
- **Path cost**: time to execute

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Example: Robotic Assembly 29











30 Example: The 8-Queens

- Formulate problem:
 - **States**: ?
 - Actions: ?
 - Goal test: ?
 - Path cost: ?

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Place 8 queens in a chessboard so that no two queens are in the same row, column, or diagonal.

31 Example: The 8-Queens

• Formulation 1:

- **States**: Any arrangement of 0 to 8 queens on the board.
- Actions: Add a queen in any square.
- **Goal test**: 8 queens on the board, none attacked.
- **Path cost**: 1 per move.

\rightarrow 64⁸ states with 8 queens



Place 8 queens in a chessboard so that no two queens are in the same row, column, or diagonal.

32 Example: The 8-Queens

• Formulation 2:

- States: Any arrangement of k = 0 to 8 queens in the k leftmost columns with none attacked.
- Actions: Add a queen to any square in the leftmost empty column such that it is not attacked by any other queen.
- Goal test: 8 queens on the board, none attacked
- Path cost: 1 per move

 \rightarrow 2067 states



Place 8 queens in a chessboard so that no two queens are in the same row, column, or diagonal.





A Riddle: A Wolf, a sheep and a cabbage need to cross the river. How can you bring them across, one by one, without the sheep eating the cabbage, nor the wolf eating the sheep?

34 Answer: River Crossing

• State space S: all valid configurations

- Initial state: I = {(CSDF,)}
- **Goal state**: $G = \{(,CSDF)\}$
- Actions: states reachable in one step from S
 - $Succs((CSDF,)) = \{(CD, SF)\}$
 - Succs((CD, SF)) = {(CD, FS), (D, CFS), (C, DFS)}
- **Path cost**: 1 for each transition

Other example problems 36

• Travelling Salesperson Problem (TSP)

• Each city must be visited exactly once – find the shortest tour

• VLSI (Very Large Scale Integration) Layout

• Given schematic diagram comprising components (chips, resistors, capacitors, etc) and interconnections (wires), find optimal way to place components on a printed circuit board, under the constraint that only a small number of wire layers are available (and wires on a given layer cannot cross !)

Robot navigation

• A generalization fo the route-finding problem.

THE TRAVELLING SALESMAN PROBLEM

JHAT'S THE SHORTEST ROUTE TO VISIT ALL LOCATIONS AND RETURN?

Problem Solving by Searching

Searching as problem solving technique 38

Problem solving by searching:

- Searching is the process of looking for the solution of a problem through a set of possibilities (state space).
- Search conditions include :
 - Current state where one is;
 - **Goal state** the solution reached; check whether it has been reached;
 - **Cost** of obtaining the solution.
- The **Solution** is a path from the current state to the goal state.

Searching as problem solving technique 39

Process of Searching:

- Searching proceeds as follows:
 - Check the current state;
 - Execute **allowable actions** to move to the next state;
 - exhausted.

• Check if the new state is the solution state; if it is not, then the new state becomes the current state and the process is repeated until a solution is found or the state space is



Assumptions in basic search:

- The environment is **Static**
- The environment is **Discretizable**
- The environment is **Observable**
- The actions are **Deterministic**



function TREE-SEARCH(problem, strategy) returns a solution, or failure
initialize the search tree using the initial state of problem
loop do
 if there are no candidates for expansion then return failure
 choose a leaf node for expansion according to strategy
 if the node contains a goal state then return the corresponding solution
 else expand the node and add the resulting nodes to the search tree
end

- Basic idea:
 - offline, simulated exploration of state spa states (a.k.a.~**expanding** states)

• Main question: which fringe (or frontier) nodes to explore?

• offline, simulated exploration of state space by generating successors of already-explored



Problem: find path from Arad to Bucharest







46 Implementation: general tree search

function TREE-SEARCH(problem, fringe) returns a solution, or failure
fringe ← INSERT(MAKE-NODE(INITIAL-STATE[problem]), fringe)
loop do

if fringe is empty then return failure $node \leftarrow \text{REMOVE-FRONT}(fringe)$ if GOAL-TEST[problem](STATE[node]) then return SOLUTION(node) fringe $\leftarrow \text{INSERTALL}(\text{EXPAND}(node, problem), fringe)$

function EXPAND(node, problem) returns a set of nodes $successors \leftarrow$ the empty set for each action, result in SUCCESSOR-FN[problem](STATE[node]) do $s \leftarrow$ a new NODE PARENT-NODE[s] \leftarrow node; ACTION[s] \leftarrow action; STATE[s] \leftarrow result PATH-COST[s] \leftarrow PATH-COST[node] + STEP-COST(node, action, s) DEPTH[s] \leftarrow DEPTH[node] + 1 add s to successors return successors



- A state is a representation of a physical configuration
- cost g(x), depth



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

• A **node** is a data structure constituting part of a search tree, includes state, parent node, action, path



- paths is called the *fringe*, *frontier* or *border*.
- Implemented as a queue FRINGE.
 - INSERT (node, FRINGE)
 - REMOVE (FRINGE)
- The ordering of the nodes in FRINGE defines the search strategy.



• If you're performing a tree (or graph) search, then the set of all nodes at the end of all visited





Search strategies 50

- A search strategy is defined by picking the order of node expansion
- Strategies are evaluated along the following dimensions:
 - correctly report failure when there is not?
 - **Cost optimality**: Does it find a solution with the lowest path cost of all solutions?
 - or more abstractly by the number of states and actions considered.
 - **Space complexity**: How much memory is needed to perform the search?
- 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 ∞)

• **Completeness**: Is the algorithm guaranteed to find a solution when there is one, and to

• Time complexity: How long does it take to find a solution? This can be measured in seconds,

Uninformed vs. Informed Strategies 51

- Uninformed (or blind) strategies have no clue about how close a state is to the goal(s).
- promising" than another.

Uninformed search

- look for solutions by systematically generating new states and checking each of them against the goal.
- 1. It is very **inefficient** in most cases.
- Most successor states are "obviously" a bad choice.
- 3. Such strategies do not use problem-specific knowledge

• Informed (or heuristic) strategies exploits such information to assess that one node is "more

		Informed search
/	1.	They are almost always <u>more</u> <u>efficient</u> than uninformed strategies.
	2.	May reduce time and space complexities.
	3.	Evaluation function f(n) measures distance to the goal.
	4.	Order nodes in Frontier according to f(n) and decide which node to expand next.



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

- Breadth-First Search
- Opth-First Search
- Uniform-Cost Search
- Depth-Limited Search
- Iterative Deepening Search

Breadth-First Search





depth (strategy: expand a shallowest node first)



• BFS first visit all the nodes at the same depth first and then proceed visiting nodes at a deeper

Breadth-First Search 55

- A node can be reached from different nodes using different paths
- But we need to visit each node only once.
- So, we mark each node differently into 3 categories **unvisited**, **discovered** and **complete**.
 - Initially, all nodes are unvisited. After visiting a node for the first time, it becomes discovered.
 - A node is complete if all of its adjacent nodes have been visited.
 - Thus, all the adjacent nodes of a complete node are either discovered or complete.



https://www.codesdope.com/course/algorithms-bfs/



Breadth-First Search 56

- Thus after visiting a node, we first visit all its sibling and then their children.
- We can use a first in first out (FIFO) queue to achieve this (Fringe is a FIFO queue).
- Starting from the source, we can put all its adjacent nodes in a queue























Breadth-First Search (BFS) Properties 58

- Is it complete?
 - Yes. *d* must be finite if a solution exists.
- Is it optimal?
 - Only if costs are all 1.
- Search time?

•
$$1+b+b^2+b^3+\cdots+b^d=O(b^d)$$

- Fringe space?
 - Has roughly the last tier, so $O(b^d)$







Depth-First Search

Depth-First Search (DFS) 60

• DFS first explores the depth of the graph before the breadth i.e., it traverses along the and then do the same with the sibling node.



increasing depth and upon reaching the end, it backtracks to the node from which it was started



- discovered and complete respectively.
- Fringe = LIFO queue, i.e., a stack that put successors at front.



• Similar to the BFS, we also mark the vertices white, gray and black to represent unvisited,



https://www.codesdope.com/course/algorithms-dfs/











Depth-First Search (DFS) Properties 63

• Is it complete?

- *m* could be infinite, so only if we prevent cycles
- Is it optimal?
 - No, it finds the "leftmost" solution, regardless of depth or cost
- Search time?
 - $O(b^m)$ Terrible if m is much larger than d, but if solutions are dense, may be much faster than breadth-first.
- Fringe space?
 - Only has siblings on path to root, so *O(bm)*, i.e., linear space!





- When will BFS outperform DFS?
- When will DFS outperform BFS?



\bullet = depth-first search with depth limit L, i.e., nodes at depth L have no successors





Iterative Deepening Search 66

• Idea: get DFS's space advantage with BFS's time / shallow-solution advantages • Run a DFS with depth limit 1. If no solution...

- Run a DFS with depth limit 2. If no solution...
- Run a DFS with depth limit 3.



Suppose, we want to find node- '2' of the given infinite undirected graph/tree. A DFS node to be searched starting from node- 0 will dive left, towards node 1 and so on. Whereas, the node 2 is just adjacent to node 1. Hence, a DFS wastes a lot of time in coming back to node 2. An Iterative Deepening Depth First Search overcomes this and quickly find the required node. • • • . .





BFS finds the shortest path in terms of number of actions. It does not find the least-cost path. We will now cover a similar algorithm which does find the least-cost path.

Uniform-Cost Search



Output Search (Dijkstra's algorithm)

Strategy: expand a cheapest node first:

Fringe is a priority queue (priority: cumulative cost)





Equivalent to breadth-first if step costs all equal.

Properties of Uniform-Cost Search 70

- <u>Complete?</u> Yes, if step cost $\geq \varepsilon > 0$ (ε is the lower bound of each action), it will never getting caught going down a single infinite path
- c ≤ 1 • Time? # of nodes with $g \le cost$ of optimal solution, c ≤ 2 O($b^{ceiling(C^*/\epsilon)}$) where C^{*} is the cost of the optimal C^*/ε "tiers" c ≤ 3 solution. • Space? # of nodes with $g \le cost$ of optimal solution, O(bceiling(C*/ ε)).
- <u>Optimal?</u> Yes nodes expanded in increasing order of g(n).

Uniform-Cost Search Issues 71

- Remember: UCS explores increasing cost contours
- The good: UCS is complete and optimal!
- The bad:
 - Explores options in every "direction"
 - No information about goal location
- We'll fix that soon!







Criterion	Breadth- First	Uniform- Cost	Depth- First	Depth- Limited	Iterative Deepening
Complete?	Yes ¹	Yes ^{1,2}	No	No	Yes ¹
Optimal cost?	Yes ³	Yes	No	No	Yes ³
Time	$O(b^d)$	$O(b^{1+\lfloor C^*/\epsilon floor})$	$O(b^m)$	$O(b^\ell)$	$O(b^d)$
Space	$O(b^d)$	$O(b^{1+\lfloor C^*/\epsilon floor})$	O(bm)	$O(b\ell)$	O(bd)


• All these search algorithms are the same except for fringe strategies

- Can even code one implementation that takes a variable queuing object

• Conceptually, all fringes are priority queues (i.e. collections of nodes with attached priorities)

Thanks! Q&A



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