

State space search A* Algorithm and way to it via Breath-first search and Dijkstra algorithms

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Motivation

- Many analytical tasks can be solved by searching through a space of possible states.
- Starting from an initial state, we try reaching a goal state.
- Sequence of actions leading from initial to goal state is the solution to the problem.
- The issues: large number of states a many choices to make in each state.
- Search has to be performed in a systematic manner.





Typical search tasks





State space search, the basic idea

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- State space search amounts to a search through a directed graph.
 - graph nodes = states
 - arcs (directed edges) = transitions between states.
- Graph may be defined explicitly or implicitly.
- Graph may contain cycles.

If we also need the transition costs, we work with a weighted directed graph.



Size of the search space

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- The state space can be HUGE! (Combinatorial explosion)
- Right representation helps.
 - Eight puzzle: 181,440
 - Draughts / Checkers / in Czech dáma: 10⁴⁰
 - Chess: 10¹²⁰ (in an average length game)
 - Theorem Proving: Infinite!
- Control strategy helps choose which operators to apply:
 - Small # of operators: general, but bushy tree.
 - Large #: perhaps overly specific, but less bushy trees.

Search tree

- By searching through a directed graph, we gradually construct a search tree.
- We do this by expanding one node after the other: we use the successor function to generate the descendants of each node.
- Open nodes or "the frontier": nodes that have been generated, but have not yet been expanded.
- Closed nodes: already expanded nodes.
- Search strategy is defined by the order in which the nodes are expanded. Different orders yield different strategies.





State space vs. search tree



- Search tree is created while searching through the state space.
- Search tree can be infinite even if the state space is finite. E.g. if the state space contains cycles → search tree is infinite.



Open nodes, pictorial illustration

V. Hlaváč, State space search



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The basic search algorithm

Initialize: put the start node into OPEN while OPEN is not empty take a node N from OPEN if N is a goal node, report success put the children of N onto OPEN Report failure

- If OPEN is a stack, this is a depth-first search.
- If OPEN is a queue, this is a breadth-first search.
- If OPEN is a priority queue, sorted according to most promising first, we have a best-first search (Dijkstra algorithm).



Breadth-first search



(abbrev. BFS)

Implementation:

- Pick and remove a location from the OPEN (frontier).
- Mark the location as visited so that we know not to process it again.
- Expand it by looking at its neighbors. Any neighbors we haven't seen yet we add to the frontier.

Breadth-first search (2)







Breadth-first search (3)

- visits all reachable places
- efficiency:
 - time: $O(b^d)$
 - space: $O(b^d)$
 - *b*=branching factor, *d*=depth of goal
- no priority
- possible improvements:
 - early exit = search stops when the goal is reached
 - movement cost \rightarrow Dijkstra algorithm



Dijkstra algorithm

- Adding movement cost to Breath-first search algorithm, expands in all directions
- Using priority queue
 - Choosing move with the lowest cost
- Time efficiency: O(/E/+/V/ log/V/), V=number of nodes, E=number of edges





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V. Hlaváč, State space search

Dijkstra algorithm vs. BFS





Breadth First Search

Dijkstra's Algorithm



Greedy best first search



- better for finding path to one exact location in Prague
- use of heuristics:
 - distance to the goal
 - e.g.: def heuristics(a,b): return abs(a.x - b.x) + abs(a.y + b.y)
- time/space efficiency: O(b^m)
 - good heuristics can give huge improvements
- priority queue
 - priority = distance to goal

Greedy best-first search - examples





Greedy Best-First Search

Greedy best-first search - examples

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- Problem with obstacles.
- May not find the shortest path.



A* algorithm (read "A star")

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- Using the best of both Djikstra and Greedy algorithms, worst time/space: O(b^d)
- Expanding based on:
 - distance from start
 - distance to goal (=heuristics)









A* algorithm





Map of Manhattan



How would you find a path from S to G?



Best-First Search



- The Manhattan distance $(\Delta x + \Delta y)$ is an estimate of the distance to the goal
 - It is a heuristic function
- Best-First Search
 - Order nodes in priority queue to minimize estimated distance to the goal h(n)
- Compare: Dijkstra
 - Order nodes in priority queue to minimize distance from the start



How would you find a path from S to G?



Problem 1: Led astray



Eventually will expand vertex to get back on the right track



Problem 2: Optimality



- With Best-first search, are you guaranteed a shortest path is found when
 - goal is first seen?
 - when goal is removed from priority queue (as with Dijkstra?)

Sub-optimal solution



No! Goal is by definition at distance 0: will be removed from priority queue immediately, even if a shorter path exists!



Synergy?



- Dijkstra / Breadth First guaranteed to find optimal solution
- Best First often visits *far fewer* vertices, but may not provide optimal solution
 - Can we get the best of both?

A*, heuristics



Order vertices in priority queue to minimize (distance from start) + (estimated distance to goal)

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

f(n) = priority of a node g(n) = true distance from start h(n) = heuristic distance to goal

Optimality



- Suppose the estimated distance (h) is always less than or equal to the true distance to the goal
 - heuristic is a *lower bound on true distance*
 - heuristic is admissible

Then: when the goal is removed from the priority queue, we are guaranteed to have found a shortest path!















f(n)

7

5

7



















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Importance of Heuristics

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- h1 = number of tiles in the wrong place
- h2 = sum of distances of tiles from correct location

IDS	A*(h1)	A*(h2)
10	6	6
112	13	12
680	20	18
6384	39	25
47127	93	39
364404	227	73
3473941	539	113
	3056	363
	39135	1641
	IDS 10 112 680 6384 47127 364404 3473941	$\begin{array}{cccc} \text{IDS} & A^*(\text{h1}) \\ 10 & 6 \\ 112 & 13 \\ 680 & 20 \\ 6384 & 39 \\ 47127 & 93 \\ 364404 & 227 \\ 3473941 & 539 \\ 3056 \\ 39135 \end{array}$

Summary



Finding path to ALL locations:

Finding path to ONE location:

- Same cost → Breadth-first search
- Costs vary → Dijkstra algorithm

 Preferably use A* algorithm