Informed Search

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Search as a Problem-Solving Technique







Branching Factor *b*=3

Types of Blind Search

- Breadth-First Search
- Depth-First Search
- Depth Limited Search
- Iterative Deepening Search
- Bi-directional Search

Search as a Problem-Solving Technique







Branching Factor *b*=3

Improving Blind Search: Avoiding Repeated States



- Simple caching could be used to store the expected values of sub-trees.
 - Must maintain a table of all visited states and the result
- Change the rules for generating the tree
 - Do not generate repeated states
 - Do not generate paths with cycles

Heuristic Functions

- These techniques are all still brute-force
- Can we do anything more intelligent?
- If we could identify an *evaluation function*, which described how valuable each state was in obtaining the goal, then we could simply always choose to expand the leaf node with the best value.
- A *heuristic function* is an inexact estimate of the evaluation function.



Greedy Best-First Search



- Rely on a heuristic function to determine which node to expand
- Better name is "bestguess-first" search
- Airline example
 - Find the shortest path from Boston to Phoenix

Greedy Best-First-Search



	Distance to Phoenix	
Boston	2299	
Chicago	1447	
Nashville	1444	
Key West	1927	
Austin	870	
San Francisco	658	

 Minimize estimated cost to reach a goal (in this case, the distance to Phoenix)



Greedy Best-First-Search

- Optimal?
 - No, as the previous example demonstrated
- Complete?
 - No, just as depth first search
- Worst-case time complexity?
 O(b^m) where b=branch factor, m=max. depth
- Worst-case space complexity?
 - Same as time complexity... entire tree kept in memory
- Actual time/space complexity
 - Depends on the quality of the heuristic function

A* Search



f=3377

3377

1927

870

658

Key West

San Francisco

Austin

- Combine Greedy search with Uniform
- Minimize the total path cost (f) =actual path so far (g) + estimate of future path to goal (h)



Total Distance Flown

How does A* Search Work?



- A^{*} expands nodes in order of increasing *f* value
- Gradually adds "f-contours" of nodes
- Requires that the heuristic function h must be admissible

- It must never over-estimate the cost to reach the goal

Proving the Optimality of A*

- Assume that G_2 has been chosen for expansion over n
- Because *h* is admissible

 $f^* \geq f(n)$

- If *n* is not chosen for expansion over G_2 , we must have $f(n) \ge f(G_2)$
- Combining these, we get $f^* \ge f(G_2)$
- However, this violates our assertion that G₂ is sub-optimal
- Therefore, A* never selects a sub-optimal goal for expansion

Completeness of A*

- A* expands nodes in order of increasing *f*
- When would a solution not be found?
 - Node with an infinite branching factor
 - A path with a finite path cost but an infinite number of nodes
- A* is complete when
 - There is a finite branching factor
 - Every operator costs at least some positive $\boldsymbol{\epsilon}$

Complexity of A*

- Computation time is limited by the quality of the heuristic function (but is still exponential)
 - Issue #1 : Choosing the right heuristic function can have a large impact
- More serious problem is that all generated nodes need to be kept in memory
 - Issue #2 : Can we limit the memory requirements?

Issue #1: Choosing a Heuristic Functions





Goal State

6

2

3

4

5

- Must be admissible (never over-estimate)
- Heuristics for the 8-Puzzle

-h1 = number of tiles in the wrong position (h1=7)

h2 = sum of the distances of the tiles from their goal positions (*city block / Manhattan distance*)

h2 = 2+3+3+2+4+2+0+2 = 18

Effect of Heuristic Accuracy on Performance in the 8-puzzle

	Search Cost		
d	IDS	$A^*(h_1)$	$A^*(h_2)$
2	10	6	6
4	112	13	12
6	680	20	18
8	6384	39	25
10	47127	93	39
12	364404	227	73
14	3473941	539	113
16	-	1301	211
18	-	3056	363
20	-	7276	676
22	-	18094	1219
24	_	39135	1641

- Compare iterativedeepening search (IDS) with A* using
 - h1 (# misplaced tiles)
 - h2 (city block distance)
- Always better to use a heuristic with higher values, so long as it does not over-estimate

Issue #2

Limiting Memory Utilization

- If we can maintain a bound on the memory, we might be willing to wait for a solution
- Two techniques for Memory Bounded Search:
 - Iterative deepening A* (IDA*)
 - Recursive Best-First-Search (RBFS)

Iterative Deepening A* Search (IDA*)

- Each iteration is a depth-first search with a limit based on *f* rather than on depth
- Complete and optimal (with same caveats as A*)
- Requires space proportional to the longest path that it explores
- Can have competitive time complexity, since the overhead of maintaining the nodes in memory is greatly reduced

Problems with IDA*



- In the TSP, different heuristic function value for each state
- Each contour contains only one additional node
- If A* expands N nodes, the IDA* will expand 1+2+3+4+...+N = O(N²) nodes
- If N is too large for memory, N² is too long to wait
- Runs into problems because it recalculates every node



- total path cost (f) = actual path so far (g) + heuristic estimate of future path to goal (h)
- Red values best f-value in an alternate branch

Recursive Best-First Search (RBFS)

RBFS will

- be complete given sufficient memory to store the shallowest solution path
- be optimal if the heuristic function is admissible (and you have enough memory to store the solution)
- Both RBFS and IDA* use not enough memory.
 - Require at most linear space with the depth of the tree

Coming Up Next...

- Can you search without building a tree?
- What happens when you don't get to make the choice at each level of the search space?
- Game Playing
 - Minimax
 - Alpha-beta pruning







Administrivia

- Problem Set 1 is now out
- Next week:
 - Monday: Game Playing
 - Wednesday: No class
 - Friday: Guest lecture Dragomir Radev