

## ICS 271 - Solutions Homework 2

- Trace the operation of A\* search applied to the problem of getting to Bucharest from Lugoj using the straight-line distance heuristic.

**Answer:**

Using the single-letter labels from the map on page 100 and omitting simple back-tracking state expansions due to space limitations.

	Action	Open Nodes	$f(n)$	$g(n)$	$h(n)$
1	Initial State	(L, 244)	244	0	244
2	Expand Lugoj				
	Add Timisoara	(T, 440)	440	111	329
	Add Mehadia	(T, 440), (M, 301)	301	70	241
3	Expand Mehadia				
	Add Drobeta	(T, 440), (D, 387)	387	145	242
4	Expand Drobeta				
	Add Craiova	(T, 440), (C, 425)	425	265	160
5	Expand Craiova				
	Add Rimnicu Vilcea	(T, 440), (R, 604)	604	411	193
	Add Pitesti	(T, 440), (R, 604), (P, 503)	503	403	100
6	Expand Timisoara				
	Add Arad	(R, 604), (P, 503), (A, 595)	595	229	366
7	Expand Pitesti				
	Add Bucharest	(R, 604), (A, 595), (B, 504)	504	504	0
8	Expand Bucharest				
	At Goal State	(R, 604), (A, 595)	-	-	-

- The heuristic path algorithm is a best-first search in which the objective function is  $f(n) = (2 - w)g(n) + wh(n)$ . For what values of  $w$  is this algorithm guaranteed to be optimal? What kind of search does this perform when  $w = 0$ ? When  $w = 1$ ? When  $w = 2$ ?

**Answer:**

The algorithm is guaranteed to be optimal for  $0 \leq w \leq 1$ , since scaling  $g(n)$  by a constant has no effect on the relative ordering of the chosen paths, but, if  $w > 1$  then it is possible the  $wh(n)$  will overestimate the distance to the goal, making the heuristic inadmissible. If  $w \leq 1$ , then it will reduce the estimate, but it is still guaranteed to underestimate the distance to the goal state.

$w$	$f(n)$	Algorithm
$w = 0$	$f(n) = 2g(n)$	Uninformed best-first search
$w = 1$	$f(n) = g(n) + h(n)$	A* search
$w = 2$	$f(n) = 2h(n)$	Greedy best-first search

- Propose an admissible  $h$  function for this problem that is better than  $h \equiv 0$ .

**Answer:**

One of the possible admissible heuristics is:

$$h(state) = 3 - |(number\ of\ discs\ on\ peg\ B) - (number\ of\ discs\ on\ peg\ C)|$$

Obviously  $h(goal) = 0$  for both goals. Moreover  $h(stat) \leq number\ of\ steps\ to\ achieve\ goal$ .

4. Algorithms A\* does not terminate until a goal node is selected for expansion. However, a path to a goal node might be reached long before that node is selected for expansion. Why not terminate as soon as a goal node has been found? Illustrate your answer with an example.

**Answer:**

One cannot stop after the a goal node is found while expanding its predecessor since that can lead to a suboptimal solution. For instance, consider the simple “diamond” search graph with the given costs and heuristics in Figure 4.

Consider the trace of A\* for this problem shown in the table below. If we stopped searching after encountering the goal state (D) for the first time after expanding node C in step 3, then we would have been lead to believe that the best path to the goal had a cost of 7. However, by waiting until the the goal state has the lowest cost of nodes in the frontier, we guarantee that the lowest cost path through node B is found.

	Action	Open Nodes	$f(n)$	$g(n)$	$h(n)$
1	Initial State	(A, 6)	6	0	6
2	Expand A				
	Add B	(B, 6)	6	3	3
	Add C	(B, 6), (C, 5)	5	2	3
3	Expand C				
	Add D	(B, 6), (D, 7)	7	7	0
4	Expand B				
	Replace D	(D, 6)	6	6	0
5	Expand D				
	At Goal State		-	-	-

5. For the sliding block puzzle, specify a heuristic function  $H$ , and show the search tree produced by algorithm A\* using this function. Show the first 10 nodes expanded.

**Answer:**

A simple heuristic is taking half the sum of the number of white tiles to the right of each black tile,  $w_i$  since it will take at least one move to move past each black tile. Dividing by half ensures that we never overestimate since it's possible for a white tile to hop two black tiles in a single move.

$$h(n) = \frac{1}{2} \sum_i w_i$$

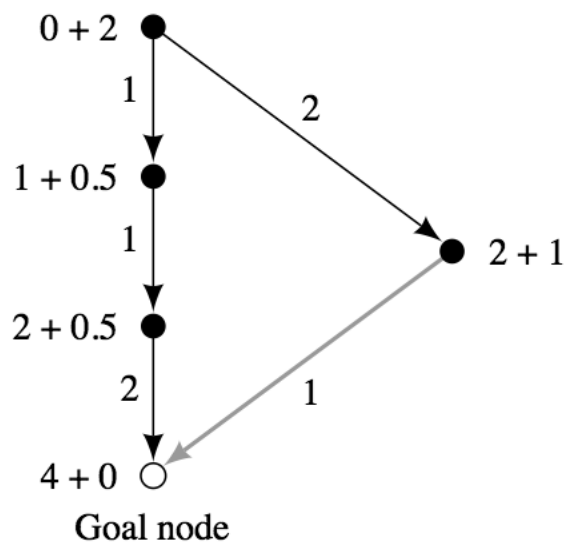


Figure 1: Example where stopping after finding a goal state for the first time fails for A\* search.

	Action	Open Nodes	$f(n)$	$g(n)$	$h(n)$
1	Initial State	(BBBWWWE, 4.5)	4.5	0	4.5
2	Expand BBBWWWE				
	Add BBBWWEW	(BBBWWWEW, 5.5), (BBBWEWW, 5.5)	5.5	1	4.5
	Add BBBWEWW	(BBBEWWW, 5.5)	5.5	1	4.5
	Add BBBEWWW		6.5	2	4.5
3	Expand BBBWWEW				
	Add BBEWWBW	(BBBWEWW, 5.5), (BBBEWWW, 6.5)	6.5	3	3.5
		(BBEWWBW, 6.5)			
4	Expand BBBWEWW				
	Add BBEWBWW	(BBBEWWW, 6.5), (BBEWWBW, 6.5)	6	2	4
	Add BEBWBWW	(BBEWBWW, 6.0), (BEBWBWW, 7.0)	7	3	4
5	Expand BBEWBWW				
	Add BBWEBWW	(BBBEWWW, 6.5), (BBEWWBW, 6.5)	7	3	4
	Add BBWWBEW	(BEBWBWW, 7.0), (BBWEBWW, 7.0)	7.5	4	3.5
	Add EBBWBWW	(BBWWBEW, 7.5), (EBBWBWW, 7.0)	7	3	4
6	Expand BBBEWWW				
	Add BBEBWWW	(BBEWWBW, 6.5), (BEBWBWW, 7.0)	7.5	3	4.5
	Add BEBBWWW	(BBWEBWW, 7.0), (EBBWBWW, 7.0)	7.5	3	4.5
	Add EBBBWWW	(BBWWBEW, 7.5), (BBEBWWW, 7.5)	8.5	4	4.5
		(BEBBWWW, 7.5), (EBBBWWW, 8.5)			
7	Expand BBEWWBW				
	Add BEBWWBW	(BEBWBWW, 7.0), (BBWEBWW, 7.0)	7.5	4	3.5
	Add EBBWWBW	(EBBWBWW, 7.0), (BBWWBEW, 7.5)	7.5	4	3.5
	Add BBWEWBW	(BBEBWWW, 7.5), (BEBBWWW, 7.5)	7.5	4	3.5
	Add BBWWEBW	(EBBBWWW, 8.5), (BEBWWBW, 7.5)	7.5	4	3.5
		(EBBWWBW, 7.5), (BBWEWBW, 7.5)			
		(BBWWEBW, 7.5)			

6. Prove the following properties on algorithm A\*.

- (a) A heuristic function is monotone if for every node  $n$  and its child node  $n'$

$$h(n) \leq h(n') + c(n, n')$$

Prove that is  $h_1$  and  $h_2$  are both monotone, so also is  $h = \max(h_1, h_2)$

**Answer:**

$$\begin{aligned}
h(n) &= \max(h_1(n), h_2(n)) \\
&\leq \max(h_1(n') + c(n, n'), h_2(n') + c(n, n')) \\
&\leq \max(h_1(n'), h_2(n')) + c(n, n') \\
&\leq h(n') + c(n, n')
\end{aligned}$$

- (b) Prove that if  $h$  is monotone then it is also admissible.

**Answer:**

At the goal node,  $h(n^*) = 0$ , so one step away from the goal,  $h(n) \leq h(n') + c(n, n') \leq c(n, n^*)$ . By induction,  $h(n) \leq c(n, n^*)$  from any node, thus it is admissible since it always underestimates the path cost to the goal state.

- (c) Prove that if the heuristic function is monotone then A\* will never reopen any nodes.

**Answer:**

$$\begin{aligned}
 f(n) &= g(n) + h(n) \\
 &\leq g(n) + h(n') + c(n, n') \\
 &\leq g(n) + c(n, n') + h(n') \\
 &\leq g(n') + h(n') \\
 &\leq f(n')
 \end{aligned}$$

Thus, we have proven that  $f(n)$  is always greater at a successor node, so once a node has been visited, it cannot be visited again at a smaller cost and thus will never be reopened.

- (d) Prove or give a counter example: if for every node  $n$ ,  $h_1(n) \geq h_2(n)$ , then A\* with  $h_1$  always expands less nodes than A\* with  $h_2$

**Answer:**

Well, since  $h_1(n) \geq h_2(n)$  we could trivially say that if the two heuristics are equal then  $h_2$  will not result in expanding less nodes than  $h_1$ , but I assume that is not the intent of the question, so let's consider  $h_1(n) < h_2(n)$ .

For any given node  $n$ , it will be placed in the open list with a value of  $f(n) = g(n) + h(n)$ . Since  $f_1(n) \geq f_2(n)$ , every node evaluated by  $f_1$  functions will have an estimated value at least as large as  $f_2$ . Since the node with the smallest value is chosen for expansion, one of two possibilities exist

- $f_1(n) \geq f_2(n)$  and  $n$  has the smallest value. In this case the same node is expanded by both heuristics.
- $f_1(n) \geq f_2(n)$  and there exists a node  $n'$  in the open list such that  $f_1(n) \geq f_1(n')$  and  $f_2(n) \leq f_2(n')$ . In this case, A\* will select node  $n'$  first, and, if that node leads to a solution state, node  $n$  will not be expanded. Note that the inequality for  $f_2$  is valid since it just shows that  $f_2$  underestimated the path cost to the goal from node  $n'$ .

Thus we have shown that A\* with heuristic  $h_1$  can only expand as many or less nodes than  $h_2$ .

- (e) Let  $h$  be an admissible function and let  $f(n) = w \cdot g(n) + (1 - w)h(n)$ ,  $0 \leq w \leq 1$ . Will A\* with  $f$  find an optimal solution when  $w = \frac{1}{4}$ ?,  $w = \frac{1}{2}$ ?,  $w = \frac{3}{4}$ ? Can you provide a general rule? (note, that  $f$  here denotes an arbitrary evaluation function, not necessarily an exact one).

**Answer:**

It is not guaranteed to find an optimal solution for  $w = \frac{1}{4}$  and is guaranteed for  $w = \frac{1}{2}$  and  $\frac{3}{4}$ . The general rule is, for constants  $a$  and  $b$  where  $f(n) = a \cdot g(n) + b \cdot h(n)$ , if  $\frac{a}{b} \geq 1$  then A\* is guaranteed to find an optimal solution.

For this problem,  $a = w, b = 1 - w$ , so, solving the ratio,

$$\begin{aligned}
 \frac{a}{b} &\geq 1 \\
 \frac{w}{1-w} &\geq 1 \\
 w &\geq 1-w \\
 w &\geq \frac{1}{2}
 \end{aligned}$$

7. Extra-credit problem, no solution.

8. On page 108, we defined the relaxation of the 8-puzzle in which a tile can move from square A to square B if B is blank. The exact solution of this problem defined **Gaschnig's heuristic**. Explain

why Gashnig's heuristic is at least as accurate as  $h_1$  (misplaced tiles), and show cases where it is more accurate than both  $h_1$  and  $h_2$ . Can you suggest a way to calculate Gaschnig's heuristic efficiently?

**Answer:**

Since the misplaced tile heuristic can place any tile in any other position in one move and Gaschnig's heuristic can only move a tile to the blank spot, Gaschnig's will always take *at least* one move to get a tile to its proper position, and may require two moves if the blank location is in its final position and tiles are still misplaced. Thus, Gaschnig's heuristic *dominates*  $h_1$  since it is always returns more or an equal number of moves.

1	2	3
8		4
7	6	5

Target

1	2	3
6		4
8	7	5

State

True Shortest Path Cost: 4

Gaschnig's: 4

Misplaced Tiles: 3

1	2	3
8		4
7	6	5

Target

1	2	3
7		4
8	6	5

State

True Shortest Path Cost: > 3

Gaschnig's: 3

Manhattan: 2

Some possible pseudo-code for computing Gaschnig's heuristic is:

```

moves = 0
while not in goal state:
    if blank in goal position:
        swap blank with any mismatch
    else
        swap blank with matched tile
    moves++
return moves

```