

# CS540 Introduction to Artificial Intelligence

## Lecture 16

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Based on lecture slides by Jerry Zhu and Yingyu Liang

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# UCS Example, Part II

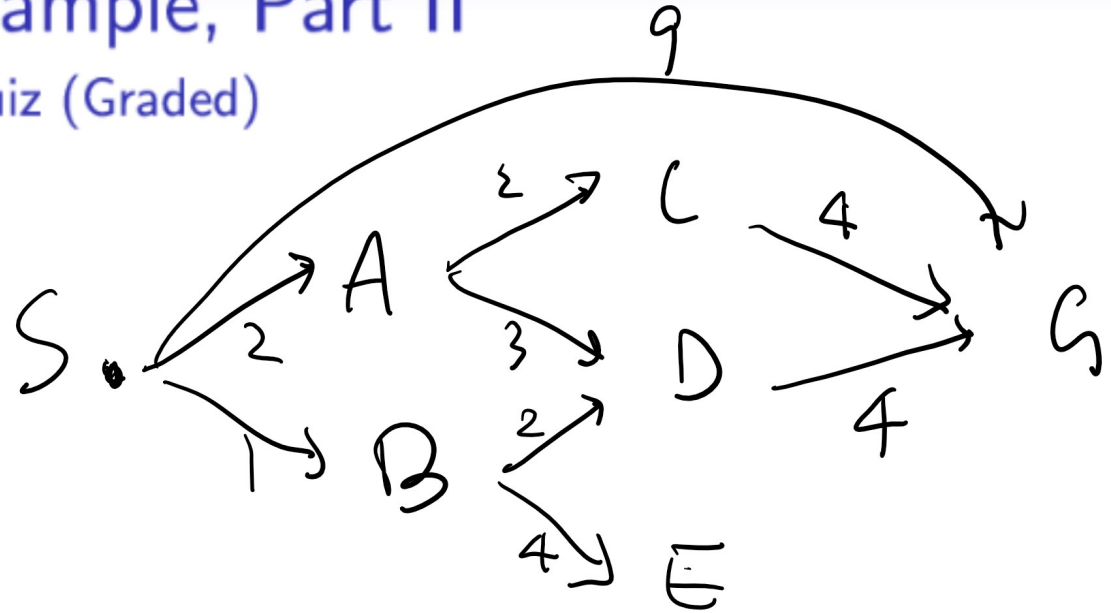
Quiz (Graded)

UCS

expansion path?

Q10

- A: S B A D C E G
- B: S B D G
- C: S A G
- D: S G
- E: S A D B D G





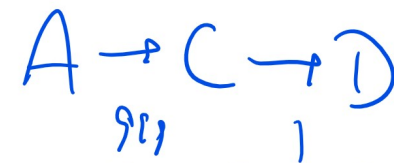
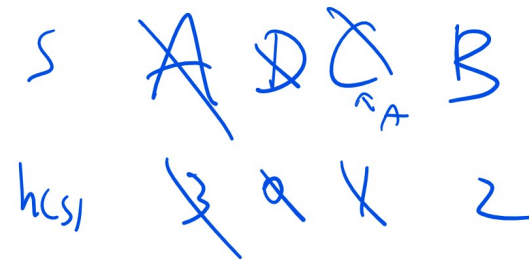
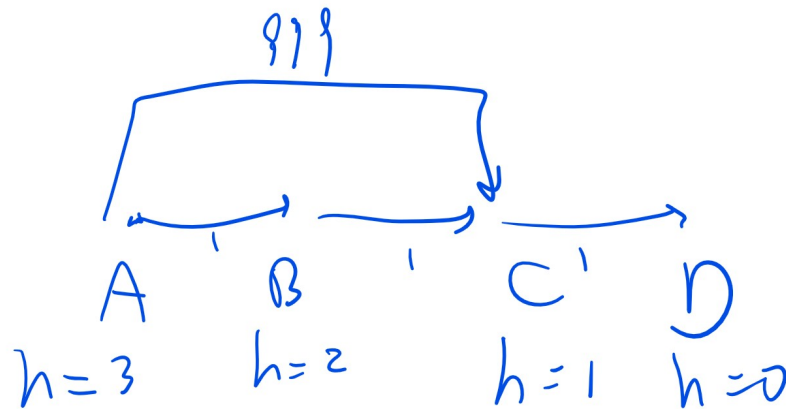


# Best First Greedy Search

## Description



- Expand the vertices with the lowest heuristic cost  $h(s)$  first.
- Use a priority queue based on  $h(s)$ .



= 1000





# Best First Greedy Search

## Algorithm

- Input: a weighted digraph  $(V, E, c)$ , initial states  $I$  and goal states  $G$ , and the heuristic function  $h(s), s \in V$ .
- Output: a path from  $I$  to  $G$ .
- EnQueue initial states into a priority queue  $Q$ . Here,  $Q$  is ordered by  $h(s)$  for  $s \in Q$ .

$$Q = I$$

- While  $Q$  is not empty and goal is not deQueued, deQueue  $Q$  and enQueue its successors.

$$s = Q_{(0)} = \arg \min_{s \in Q} h(s)$$

$$Q = Q + s'(s)$$

# Best First Greedy Search Performance

## Discussion

- Greedy is incomplete.
- Greedy is not optimal.









# A Search

## Algorithm

- Input: a weighted digraph  $(V, E, c)$ , initial states  $I$  and goal states  $G$ , and the heuristic function  $h(s), s \in V$ .
- Output: a path from  $I$  to  $G$ .
- EnQueue initial states into a priority queue  $Q$ . Here,  $Q$  is ordered by  $g(s) + h(s)$  for  $s \in Q$ .

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$$s = Q_{(0)} = \arg \min_{s \in Q} g(s) + h(s)$$

$$Q = Q + s'(s)$$



# A Star Search

## Description

- $A^*$  search is A search with an admissible heuristic.

# Admissible Heuristic

## Definition

- A heuristic is admissible if it never over estimates the true cost.

$$0 \leq h(s) \leq h^*(s)$$

# Admissible Heuristic 8 Puzzle Example

## Quiz (Graded)

Q14 A, B, D, E

Which ones (select multiple) of the following are admissible heuristic function for the 8 Puzzle?

1	2	3
4	5	6
7	8	

A:  $h(s)$  = number of tiles in the wrong position.

B:  $h(s) = 0$ .

C:  $h(s) = 1$ .

D:  $h(s)$  = sum of Manhattan distance between each tile and its goal location.

E:  $h(s)$  = sum of Euclidean distance between each tile and its goal location.

UCS  
 $h(G) = 1 \neq h^*(G) = 0$

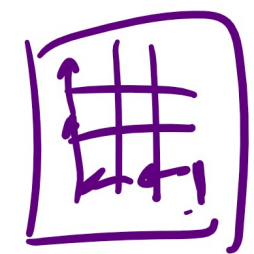
jump anywhere

move without other tiles

$0 \leq h_E \leq h_D \leq h^*$

$h(s) \leq h^*(s) \forall s$

move diagonally





# Admissible Heuristic General Example

## Quiz (Graded)

Q1b A C E

• Which ones (select ~~multiple~~) of the following are admissible heuristic function?

• A:  $h(s) = h^*(s)$ .

~~• B:  $h(s) = \max\{2, h^*(s)\}$ .~~

• C:  $h(s) = \min\{2, h^*(s)\}$ .

~~• D:  $h(s) = h^*(s) - 2$ .~~

• E:  $h(s) = \sqrt{h^*(s)}$ .

$$0 \leq h(s) \leq h^*(s)$$

$$h^*(G) = 0$$

$h^*$  must be integer

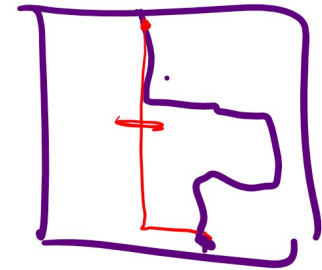
# Dominated Heuristic

## Definition

- One heuristic,  $h_1$ , is dominated by another,  $h_2$ , if:

$$h_1(s) \leq h_2(s) \leq h^*(s), \forall s \in S$$

*h<sub>2</sub> is always closer to h\**



$h(s) =$  Manhattan distance from cell to goal  
for HW.

- If  $h_2$  dominates  $h_1$ , then  $h_2$  is better than  $h_1$  since  $A^*$  using  $h_1$  expands at least as many states (or more) than  $A^*$  using  $h_2$ .
- If  $h_2$  dominated  $h_1$ ,  $A^*$  with  $h_2$  is better informed than  $A^*$  with  $h_1$ .



# A Star Search Maze Example

## Quiz (Graded)

# A Star Search with Revisit Example, Part I

## Quiz (Graded)

- Given the following adjacency matrix. Find A\* Search expansion path.

—	S	A	B	C	D	E	G
S	$h = 6$	2	1	—	—	—	9
A	—	$h = 0$	—	2	3	—	—
B	—	—	$h = 6$	—	2	4	—
C	—	—	—	$h = 4$	—	—	4
D	—	—	—	—	$h = 1$	—	4
E	—	—	—	—	—	$h = 10$	—
G	—	—	—	—	—	—	$h = 0$



# A Star Search with Revisit, Part I

## Algorithm

- Input: a weighted digraph  $(V, E, c)$ , initial states  $I$  and goal states  $G$ , and the heuristic function  $h(s), s \in V$ .
- Output: a path with minimum cost from  $I$  to  $G$ .
- EnQueue initial states into a priority queue  $Q$ . Here,  $Q$  is ordered by  $g(s) + h(s)$  for  $s \in Q$ .

$$Q = I$$

$$g(I) = 0$$

$$g(s) = \infty, \text{ for } s \notin I$$

- Initialize the list of visited vertices,  $P$ .

$$P = \emptyset$$

# A Star Search with Revisit, Part II

## Algorithm

- While  $Q$  is not empty and goal is not deQueued, deQueue  $Q$ , put it on  $P$  and enQueue its successors to  $Q$ , and update the cost functions.

$$s = Q_{(0)} = \arg \min_{s \in Q} g(s) + h(s)$$

$$P = P + s$$

$$Q = Q + s'(s), \text{ update } g(s') = \min \{g(s'), g(s) + c(s, s')\}$$



# A Search Performance

## Discussion

- $A^*$  is complete.
- $A^*$  is optimal.

# Iterative Deepening A Star Search

## Discussion

- $A^*$  can use a lot of memory.
- Do path checking without expanding any vertex with  $g(s) + h(s) > 1$ .
- Do path checking without expanding any vertex with  $g(s) + h(s) > 2$ .
- ...
- Do path checking without expanding any vertex with  $g(s) + h(s) > d$ .

# Iterative Deepening A Star Search Performance

## Discussion

- IDA\* is complete.
- IDA\* is optimal.
- IDA\* is more costly than A\*.

# Beam Search

## Discussion

- Version 1: Keep a priority queue with fixed size  $k$ . Only keep the top  $k$  vertices and discard the rest.
- Version 2: Only keep the vertices that are at most  $\epsilon$  worse than the best vertex in the queue.  $\epsilon$  is called the beam width.

# Beam Search Performance

## Discussion

- Beam is incomplete.
- Beam is not optimal.