

\* A search strategy is defined by picking the order of node expansion.

\* Uninformed / Informed Search:

→ Uninformed search algorithms looked through search space for all possible solutions of the problem without having any additional knowledge about search space.

→ Informed search algorithm contains an array of knowledge such as how far we are from the goal, path cost, how to reach to goal node, etc.

\* Heuristics Search:  $h(n)$  heuristic function we use to determine the least cost of each node.

→ Heuristic is a function which is used in Informed search, and it finds the most promising path.

→ Heuristic function estimates how close a state is to the goal state.

→ Goal state  $h(n)$  heuristic value always 0.

→  $g(n)$  calculate start to  $n$  node.  $h(n)$  calculate  $n$  node to goal node.

→  $h(n)$  node  $h(n)$  heuristic value can be Manhattan distance / Euclidean distance we use.

$$\text{Manhattan distance} = |x_2 - x_1| + |y_2 - y_1|$$

$$\text{Euclidean distance} = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

( $x_{\text{start}} - x_{\text{destination}}$ )

\* Admissibility of the heuristic function is given as:  $0 \leq h(n) \leq h^*(n)$ .

Here,  $h(n)$  = heuristic cost, and  $h^*(n)$  = estimated cost.



## \* Greedy Best-first search:

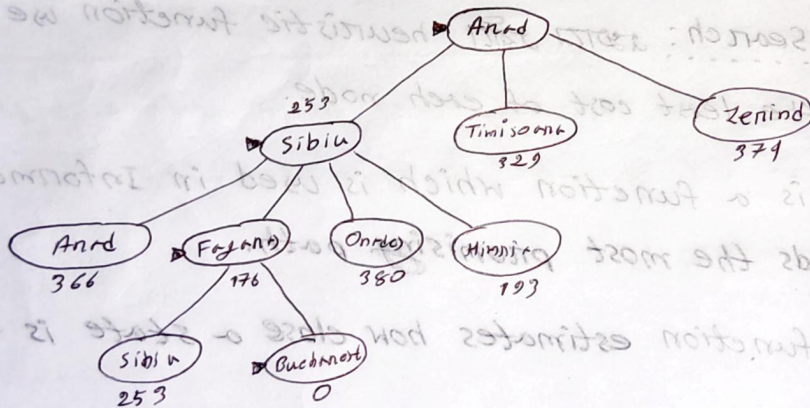
→ इसका मतलब है node का evaluation function काटते हैं।  
 बस हीनुरिस्टिक फंक्शन,  $h(n)$  का based पर calculate  
 करे हैं।

$$f(n) = h(n)$$

→ It uses the heuristic function and search and totally  
 ignores the path cost.

→ इसका मतलब है node का heuristic value को priority  
 देते हैं।

→ The greedy best first algorithm is implemented by the priority  
 queue.



→ 1st optimal solution का मतलब है, जो सबसे कम दूरी का है।

$$h(n) = 140 + 99 + 211 = 450 \text{ Km}$$

$$h^*(n) = 140 + 80 + 97 + 101 = 418 \text{ Km}$$

so, Admissibility:  $0 \leq h(n) \leq h^*(n)$  [not true].

→ 2nd Greedy problem solve का मतलब है, जो algorithm काटते  
 हैं। A\* search. जो optimal solution provide  
 करे।



\*Note \* Google map, puzzle game using A\* search.

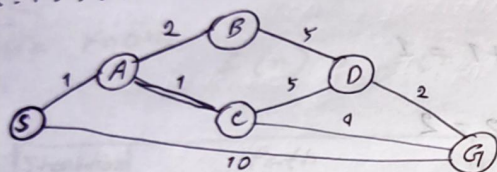
## \* A\* search:

→ It combines the strengths of UCS and greedy best-first search, by which it solve the problem efficiently.

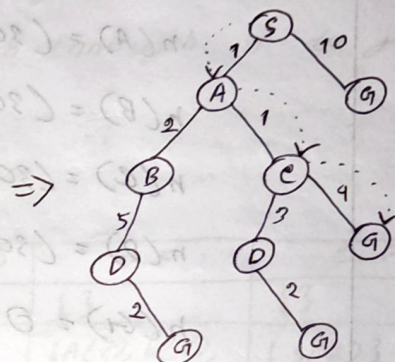
$$f(n) = g(n) + h(n)$$

estimated cost of the cheapest solution ←  $f(n)$   
 path cost ←  $g(n)$   
 heuristic cost ←  $h(n)$

Example:



state	$h(n)$
S	5
A	3
B	4
C	2
D	6
G	0



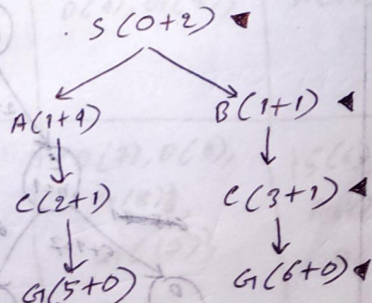
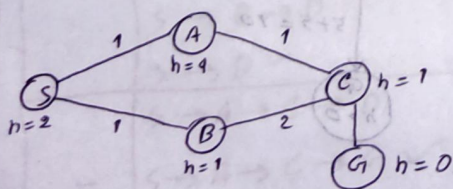
→ Optimal path will properly maintain all the:

① Admissible:  $h(n)$  should be an admissible heuristic for A\* tree search. An admissible heuristic is optimistic in nature.

② Consistency: Second required condition is consistency for only A\* graph-search.

→ Time complexity:  $O(b^d)$ .

→ Space complexity:  $O(b^d)$ .



→ Admissibility is not enough to maintain completeness and optimality under A\* graph search.



\* Consider the last 2 digits of your ID is 30 and the calculation formula of  $h(n)$  for all nodes are as follows:

∴ The heuristic values of S, A, B, C, D, and G will be as follows -

$$h(S) = (30 \% 4) + 4 = 2 + 4 = 6$$

$$h(A) = (30 \% 7) + 3 = 2 + 3 = 5$$

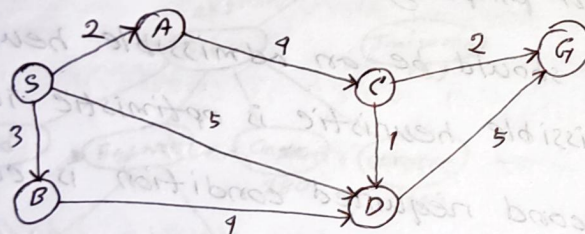
$$h(B) = (30 \% 5) + 2 = 0 + 2 = 2$$

$$h(C) = (30 \% 3) + 1 = 0 + 1 = 1$$

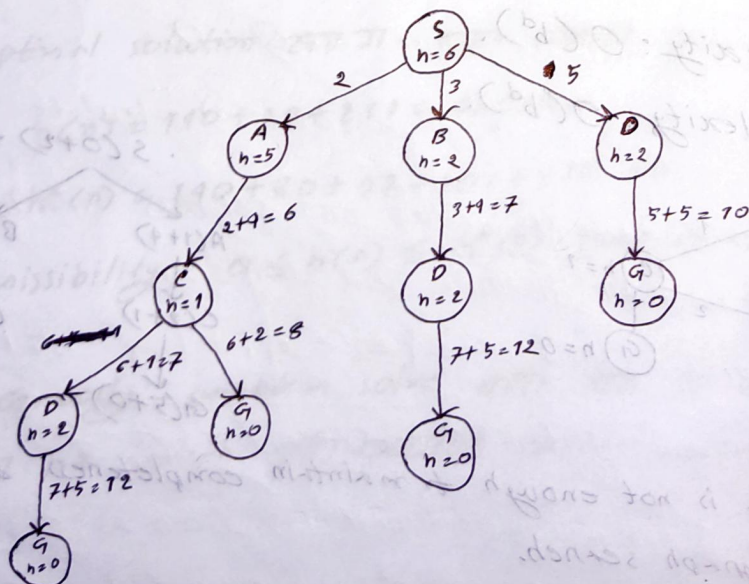
$$h(D) = (30 \% 6) + 2 = 0 + 2 = 2$$

$$h(G) = 0$$

The given state space graph:



The corresponding search tree:





Now, I will apply A\* search algorithm on the search tree:  
Here,

O-F = open fringe

C-F = close fringe

$g(n)$  = Actual cost from start node to  $n$  node.

$h(n)$  = Estimated cost from  $n$  to goal node.

$f(n)$  = Estimated total cost of path through  $n$  to goal.

We know,

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

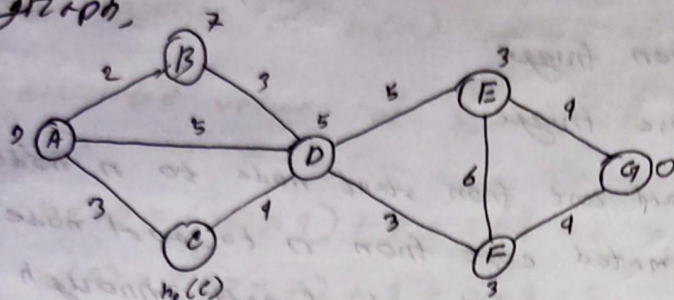
Iteration	Path	$g(n)$	$h(n)$	$f(n)$	O-F	C-F
init	S	0	6	6	{}	{}
1	S → A	2	5	7	{A(7), B(5), D(7)}	{S(6)}
	S → B ✓	3	2	5		
	S → D	5	2	7		
2	S → A ✓	2	5	7	{A(7), D(7), D(9)}	{S(6), B(5)}
	S → B → D	7	2	9		
	S → D	5	2	7		
3	S → A → C ✓	6	1	7	{A(7), D(7), D(9), C(7)}	{S(6), B(5), A(7)}
	S → B → D	7	2	9		
	S → D	5	2	7		
4	S → A → C → D	7	2	9	{D(7), D(9), D(11), G(8)}	{S(6), B(5), A(7), C(7)}
	S → A → C → G	8	0	8		
	S → B → D	7	2	9		
	S → D ✓	5	2	7		
5	S → A → C → D	7	2	9	{D(7), D(9), G(8), G(10)}	{S(6), B(5), A(7), C(7), D(9)}
	S → A → C → G ✓	8	0	8		
	S → B → D	7	2	9		
	S → D → G	10	0	10		
6	So this iteration we find the goal node G in the close fringe. So, it will return the path S → A → C → G and will optimal cost 8.				{D(7), D(9), G(10)}	{S(6), B(5), A(7), C(7), D(9), G(8)}



Given graph,

4 No Qns Ans (b)

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(i) In order for  $h_2(C)$  to be admissible,

$$0 < h_2(C) \leq \text{Optimal path cost from } C \text{ to } G$$

$$\Rightarrow 0 < h_2(C) \leq \text{cost}(C \text{ to } D) + \text{cost}(D \text{ to } F) + \text{cost}(F \text{ to } G)$$

$$\Rightarrow 0 < h_2(C) \leq 1 + 3 + 4$$

$$\Rightarrow 0 < h_2(C) \leq 11$$

so,  $h_2(C)$  is admissible for  $0 < h_2(C) \leq 11$ .

(ii) In order for  $h_2(C)$  to be consistent,

$$0 \leq h_2(C) - h_2(A) \leq \text{cost}(A \text{ to } C)$$

$$\Rightarrow 0 \leq h_2(C) - 0 \leq 3$$

$$\Rightarrow 0 \leq h_2(C) \leq 3$$

$$0 \leq h_2(C) - h_2(D) \leq \text{cost}(C \text{ to } D)$$

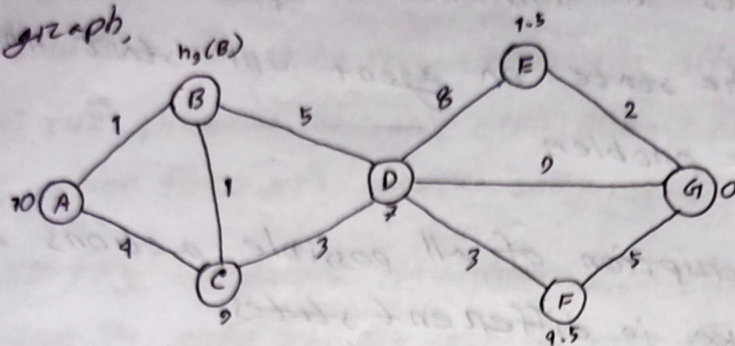
$$\Rightarrow 0 \leq h_2(C) - 5 \leq 1$$

$$\Rightarrow 0 \leq h_2(C) \leq 6$$

so,  $h_2(C)$  is consistent for  $0 \leq h_2(C) \leq 3$ .



(i) Given graph,



If  $h_3(B)$  denotes the heuristic cost of node B then in order for the heuristic function  $h_3$  to be admissible, it must fulfill the following condition

$$0 \leq h_3(B) \leq \text{Optimal total path cost from B to goal node.}$$

If G is the goal node; Optimal total path cost of B is  
 $\text{cost}(B \text{ to } C) + \text{cost}(C \text{ to } D) + \text{cost}(D \text{ to } F) + \text{cost}(F \text{ to } G)$   
 $= 1 + 3 + 3 + 5 = 12$

So, for  $h_3$  to be admissible,  $h_3(B) \leq 12$

So, any value of  $h_3(B)$  less than or equal to 12 and greater than or equal to 0 will make  $h_3$  admissible. [However  $h_3(B)$  cannot be 0 since B is not the goal node].

(ii) It is possible to formulate a given problem in artificial intelligence if an AI program is capable of solving or can be trained to solve it and if the problem fulfills the



necessary <sup>certainty</sup> criteria of problem formulation, such as having clearly defined states, initial state, goal state, etc.

The main components to formulate a problem are:

1. Initial state: The state an agent will start when solving the problem.
2. Actions: A description of all possible actions an agent can take in different states.
3. Transition Model: A description of what each action does.
4. Goal Test: A test to check if agent's current state is a goal, which indicates that the problem has been solved.
5. Path cost function: Assigns path cost to each of an agent's path. Agent will choose cost function that reflects its performance measure.