

1) Best first Search and A* [10]

Consider the search space below, where *S* is the start node and *G1* and *G2* satisfy the goal test. Arcs are labeled with the cost of traversing them and the estimated cost to a goal (the *h* function itself) is reported inside nodes.

For each of the following search strategies, indicate which goal state is reached (if any) and list, in order, all the states popped off of the OPEN list. When all else is equal, nodes should be removed from OPEN in alphabetical order.

a) Best-First-Search (using function *h* only) [3]

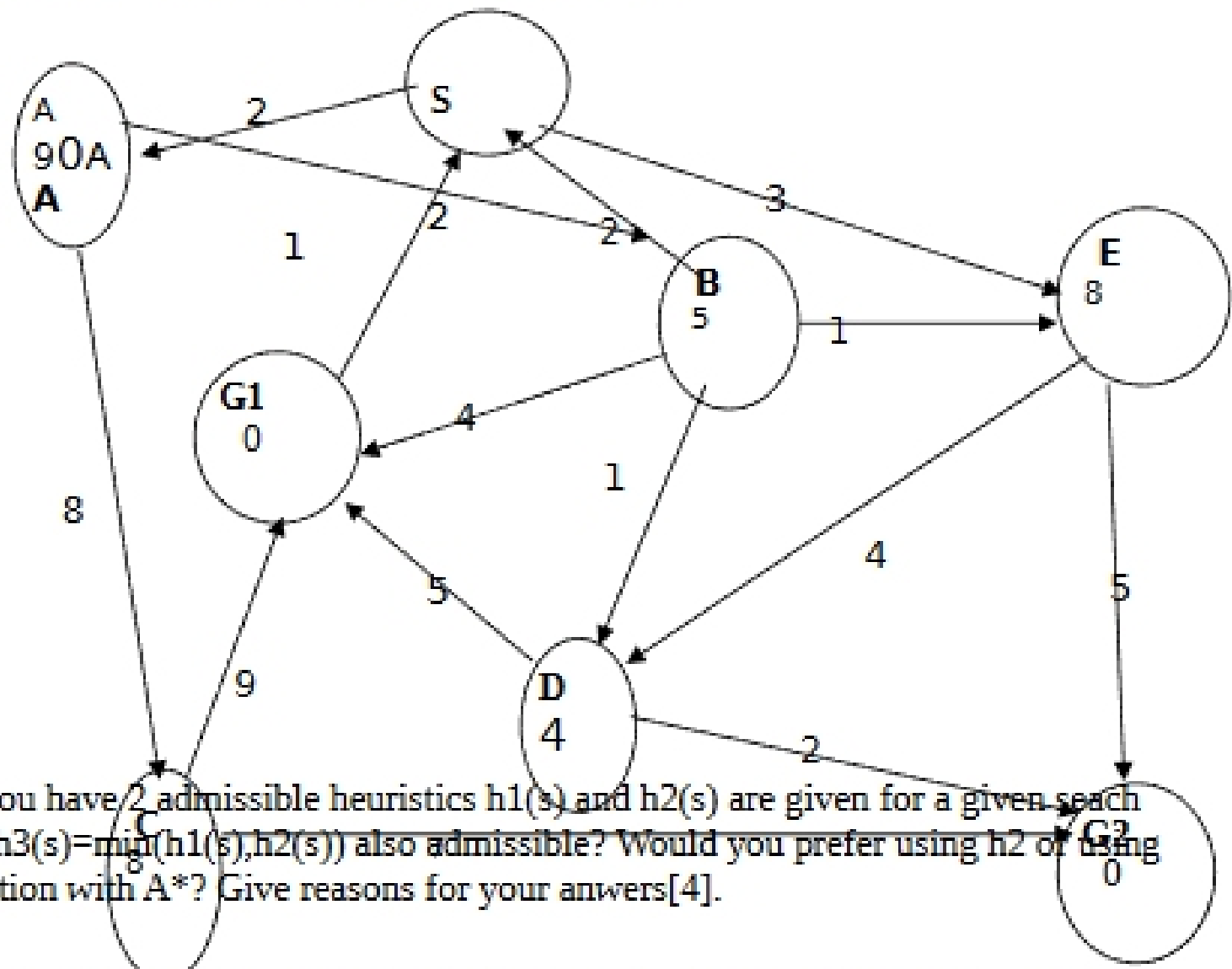
Goal state reached: *G2* [1]

States popped off OPEN: *S*, *E*, *G2* [2]

b) A* (using $f=g+h$)[4]

Goal state reached: *G1* [1]

States popped off OPEN: *S*, *A*, *B*, *G1* [3]



c) Assume you have 2 admissible heuristics $h1(s)$ and $h2(s)$ are given for a given search problem. Is $h3(s) = \min(h1(s), h2(s))$ also admissible? Would you prefer using $h2$ or using $h3$ in conjunction with A*? Give reasons for your answers [4].

Yes, $h3$ is admissible. If $h1$ and $h2$ always underestimate the “true” cost then the lesser of the two will certainly underestimate the true cost as well; therefore, $h3$ is admissible.

I will prefer h_2 , because h_2 is always greater equal than h_3 and therefore it provides a closer approximation of the true cost. As a matter of fact, h_2 dominates h_3 , which translates into equal or better efficiency of the search, as discussed on the bottom of page 106 of our textbook.

2) Local Search

- a) Assume you apply randomized hill climbing to a minimization involving a continuous, differentiable function that has 3 minima. Will it always find the optimal solution? Give reasons for your answer! [3]

No, HC might climb down the wrong minimum depending on the chosen starting point

b) What is the “main” difference between simulated annealing and randomized hill climbing? [2]