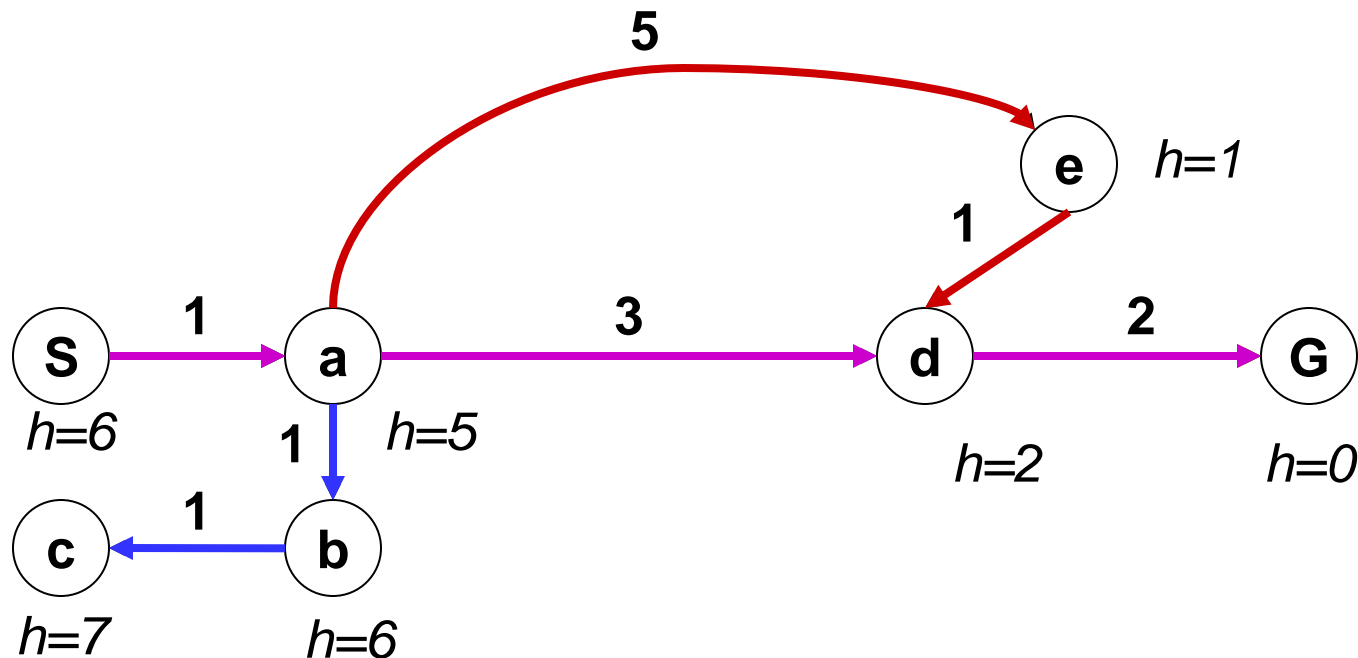


A* Review

- A* uses both backward costs g and forward estimate h : $f(n) = g(n) + h(n)$
- A* tree search is optimal with admissible heuristics (optimistic future cost estimates)
- Heuristic design is key: relaxed problems can help

Combining UCS and Greedy

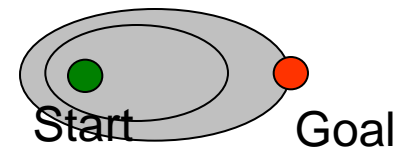
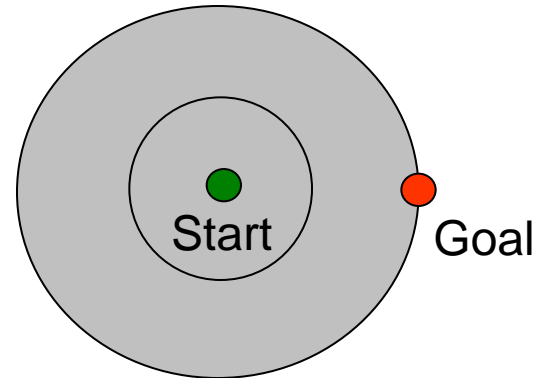
- **Uniform-cost** orders by path cost, or *backward cost* $g(n)$
- **Best-first** orders by goal proximity, or *forward cost* $h(n)$



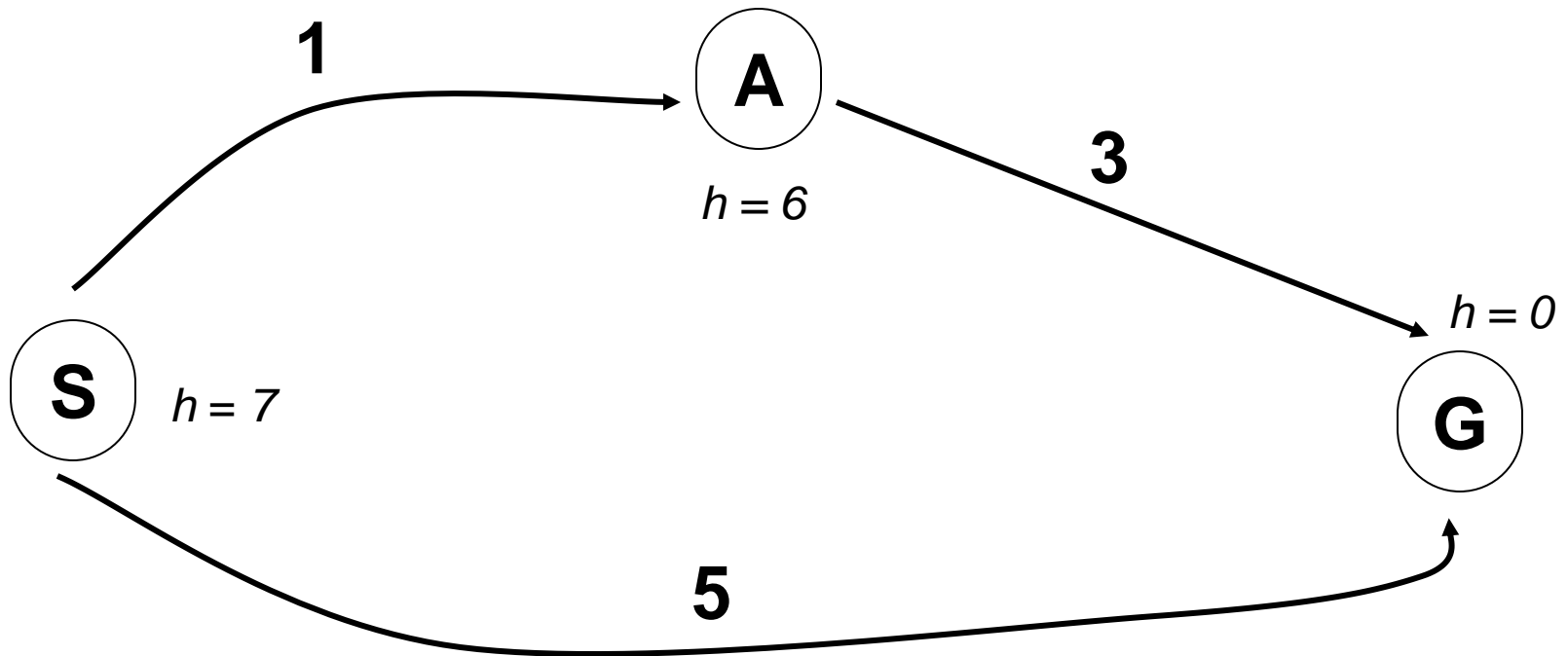
- **A* Search** orders by the sum: $f(n) = g(n) + h(n)$

UCS vs A* Contours

- Uniform-cost expanded in all directions
- A* expands mainly toward the goal, but does hedge its bets to ensure optimality



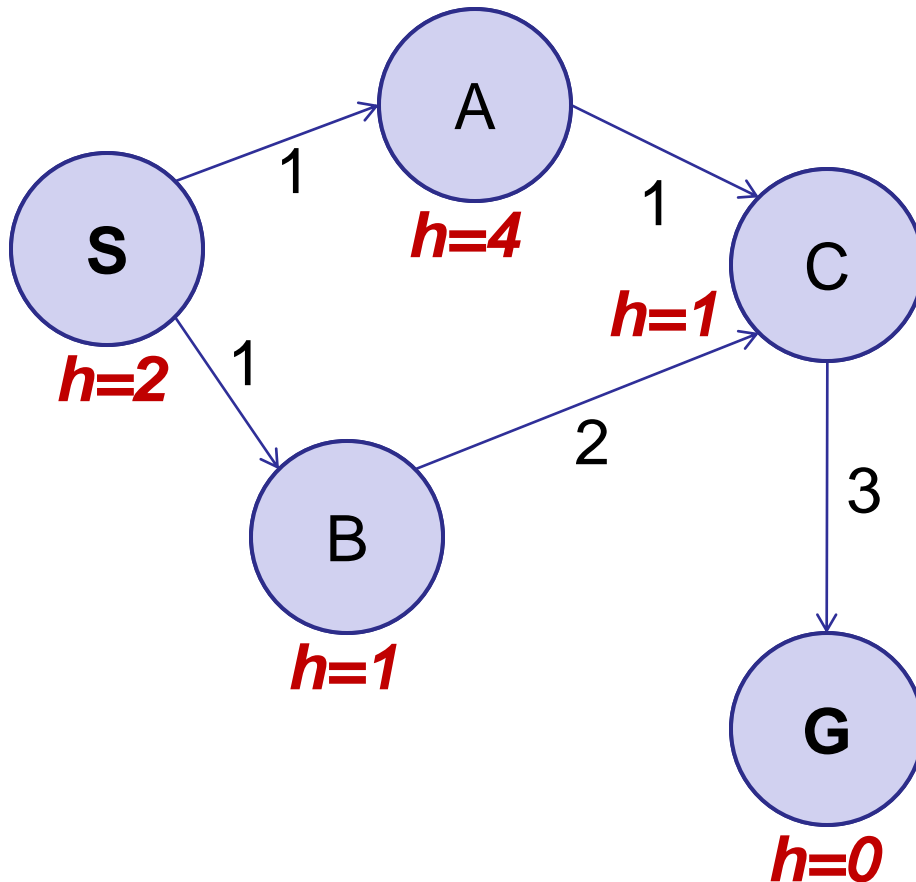
Is A* Optimal?



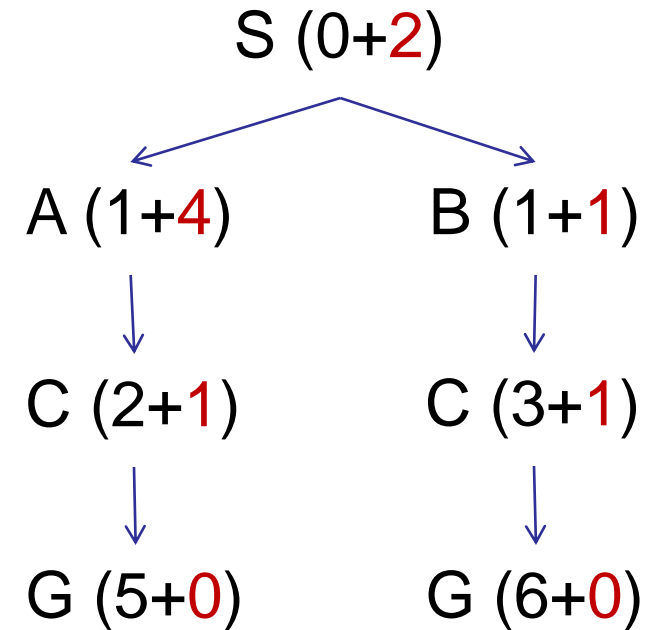
- What went wrong?
- Actual bad goal cost < estimated good goal cost
- We need estimates to be less than actual costs!

A* Graph Search Gone Wrong

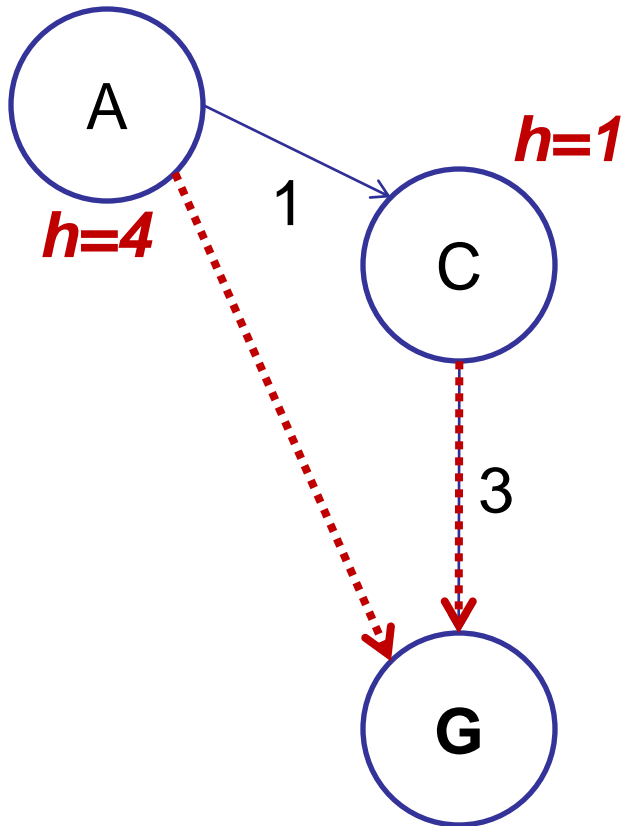
State space graph



Search tree



Consistency



The story on Consistency:

- Definition:
 $\text{cost}(A \text{ to } C) + h(C) \geq h(A)$
- Consequence in search tree:
Two nodes along a path: N_A, N_C
 $g(N_C) = g(N_A) + \text{cost}(A \text{ to } C)$
 $g(N_C) + h(C) \geq g(N_A) + h(A)$
- The f value along a path never decreases
- Non-decreasing f means you're optimal to every state (not just goals)