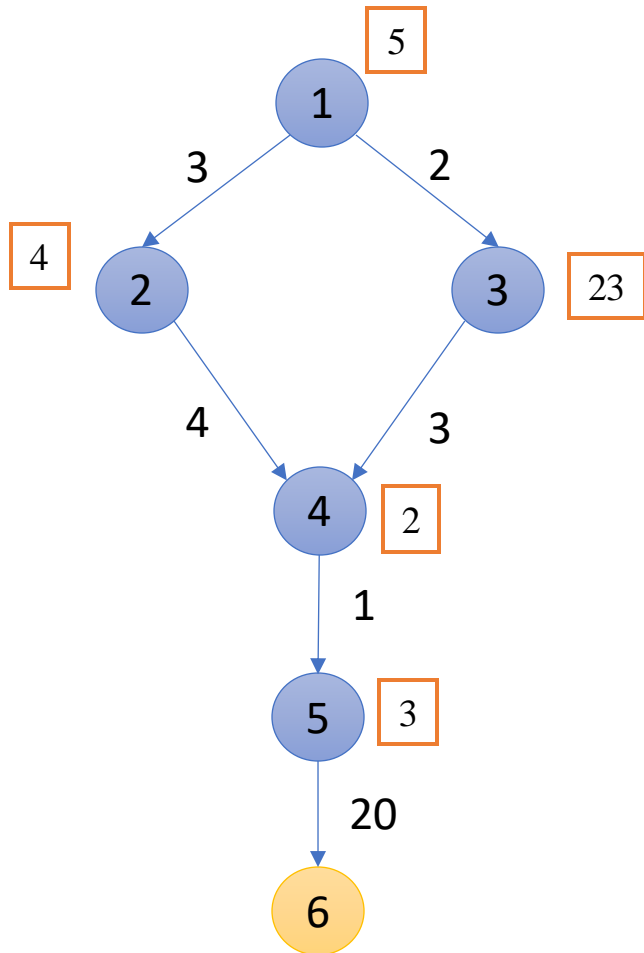


A* Analysis

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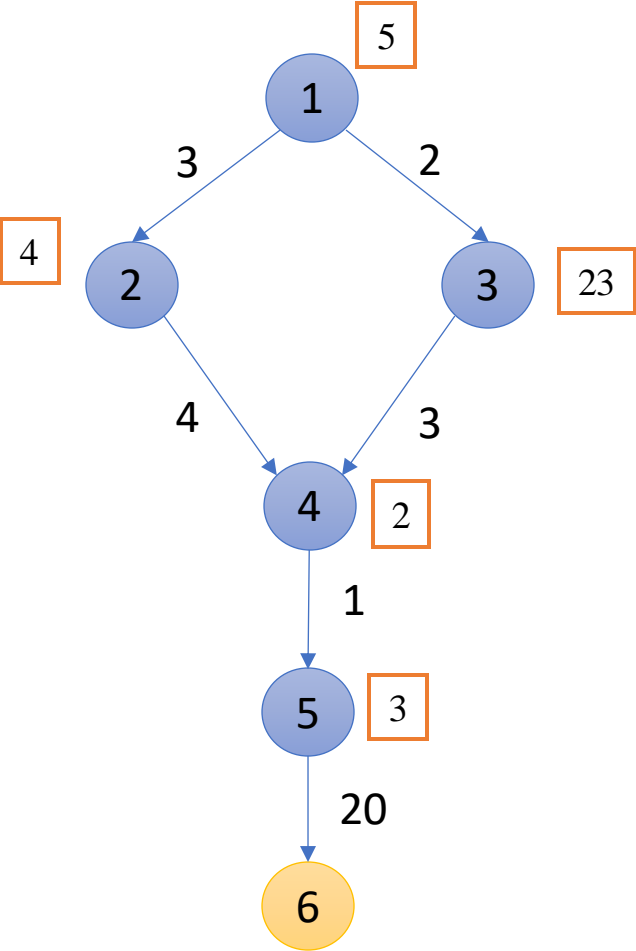
Koustav Rudra

Algorithm A*



| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|---------------|--------|------|---------------|---|
| [1(5)] | 1(5) | N | [2(7),3(25)] | [1(5)] |
| [2(7),3(25)] | 2(7) | N | [3(25),4(9)] | [1(5),2(7)] |
| [3(25),4(9)] | 4(9) | N | [3(25),5(11)] | [1(5),2(7),4(9)] |
| [3(25),5(11)] | 5(11) | N | [3(25),6(28)] | [1(5),2(7),4(9),5(11)] |
| [3(25),6(28)] | 3(25) | N | [6(28),4(7)] | [1(5),2(7), 4(9) ,5(11),3(25)] |

Algorithm A*



| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|---------------|--------|------|--------------|--|
| [3(25),6(28)] | 3(25) | N | [6(28),4(7)] | [1(5),2(7), 4(9) ,5(11),3(25)] |
| [6(28),4(7)] | 4(7) | N | [6(28),5(9)] | [1(5),2(7), 4(9) , 5(11) ,3(25),4(7)] |
| [6(28),5(9)] | 5(9) | N | [6(26)] | [1(5),2(7), 4(9) , 5(11) ,3(25),4(7),5(9)] |
| [6(26)] | 6(26) | Y | | |

A* Analysis

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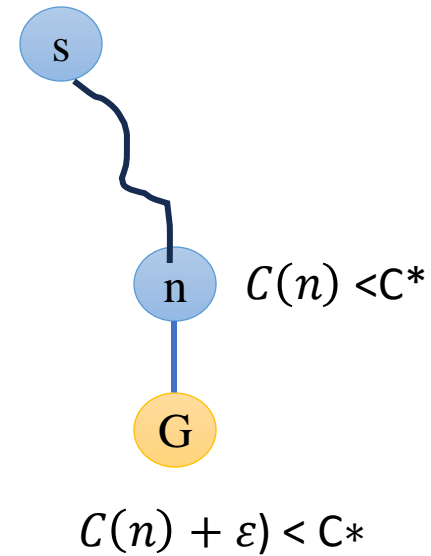
Koustav Rudra

Algorithm A*: Benefit

- Reduces number of expanded nodes
- Performs the lookahead and tells us promising paths
- What about optimality?

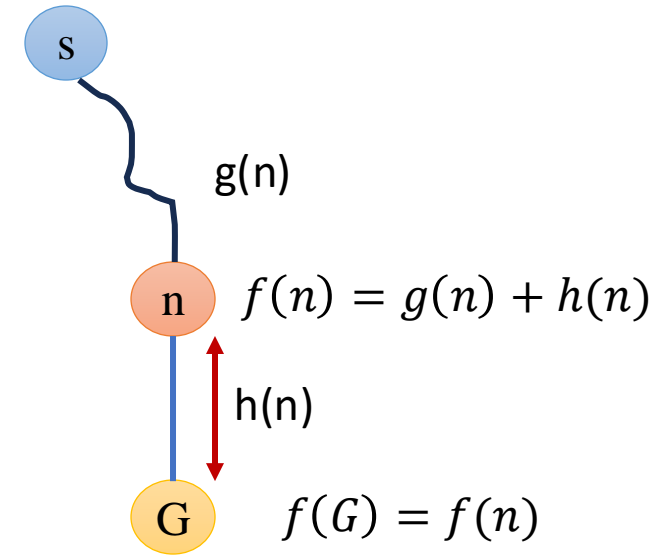
Uniform Cost Search

- **Claim:** If $C(n) < C^*$ (optimal cost) then n must be expanded
- Let algorithm A does not expand n
- For the class of algorithms without any heuristics
 - All states that have cost $< C^*$ will have to be expanded
- Always expands the minimum cost node in your frontier
- When we find the goal
 - All the states that we have in the frontier have cost higher than the goal state



Algorithm A*: Benefit

- **Claim:** $f(n) < C^*$ then n must be expanded
- The heuristic function underestimates
 - $h(n) \leq f^*(n)$
 - Cost of reaching goal from n
 - All costs are +ve
- If we do not expand n, we can't find the goal
- If we have a state whose cost is less than C^*
 - Then every algorithm which guarantees finding optimal solution have to expand it

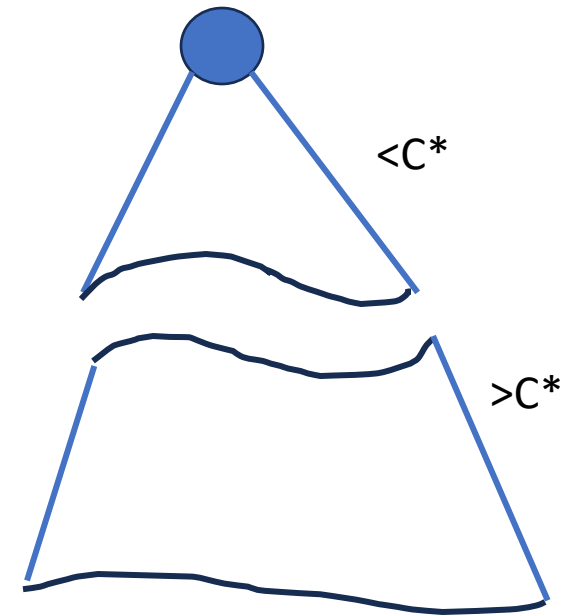


Algorithm A*

- Evaluate the INITIAL state
 - If it is GOAL return it
 - Else $CURRENT \leftarrow INITIAL$
- Loop until the solution is found or no new operators could be applied to CURRENT:
 - Select an operator that has not been applied to the current state [CURRENT] and apply it to produce new state [NEW]
 - Evaluate NEW:
 - If it is GOAL return it
 - Else If $NEW > CURRENT$, $CURRENT \leftarrow NEW$
 - Else go to Loop

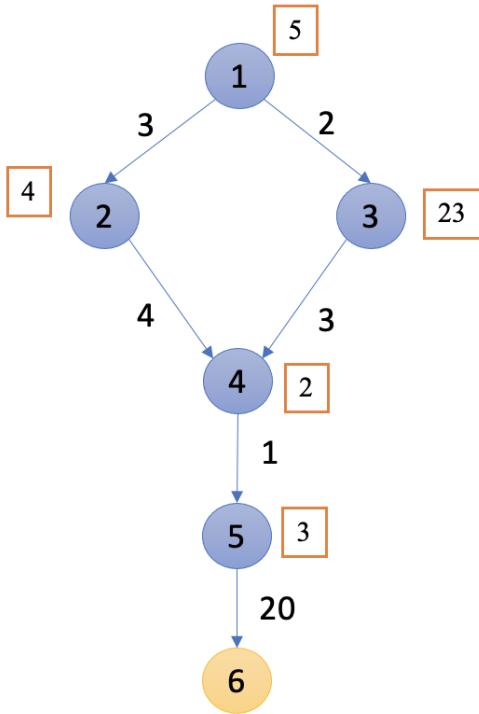
A*: Result

- What is an admissible heuristic?
 - If it always underestimates
 - We always have $h(n) \leq f^*(n)$, where $f^*(n)$ denotes minimum distance to a goal state from n
- For finite state spaces, A* will always terminate



A*: Result

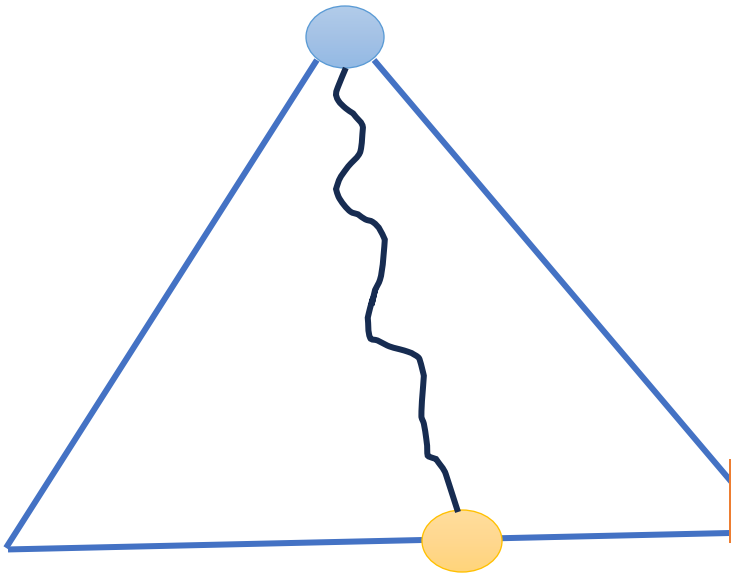
- At any time before A* terminates, there exists in OPEN a state **n**
 - That is on an optimal path from s to a goal state, with
 - $f(n) \leq f^*(s)$



| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|---------------|--------|------|---------------|--|
| [3(25),5(11)] | 5(11) | N | [3(25),6(28)] | [1(5),2(7),4(9),5(11)] |
| [3(25),6(28)] | 3(25) | N | [6(28),4(7)] | [1(5),2(7),4(9),5(11),3(25)] |
| [6(28),4(7)] | 4(7) | N | [6(28),5(9)] | [1(5),2(7),4(9),5(11),3(25),4(7)] |
| [6(28),5(9)] | 5(9) | N | [6(26)] | [1(5),2(7),4(9),5(11),3(25),4(7),5(9)] |
| [6(26)] | 6(26) | Y | | |

A*: Result

- At any time before A* terminates, there exists in OPEN a state **n**
 - That is on an optimal path from s to a goal state, with
 - $f(n) \leq f^*(s)$



| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|---------------|--------|------|---------------|--|
| [3(25),5(11)] | 5(11) | N | [3(25),6(28)] | [1(5),2(7),4(9),5(11)] |
| [3(25),6(28)] | 3(25) | N | [6(28),4(7)] | [1(5),2(7), 4(9) ,5(11),3(25)] |
| [6(28),4(7)] | 4(7) | N | [6(28),5(9)] | [1(5),2(7), 4(9) , 5(11) ,3(25),4(7)] |
| [6(28),5(9)] | 5(9) | N | [6(26)] | [1(5),2(7), 4(9) , 5(11) ,3(25),4(7),5(9)] |
| [6(26)] | 6(26) | Y | | |

A*: Result

- If there is a path from s to a goal state, A* terminates
 - Even if the state space is infinite
 - Need good estimate about heuristic
- What is the worst case scenario?
 - Heuristic is difficult to estimate
- Approximation algorithm

A*: Result

- Algorithm A* is admissible
 - If there is a path from s to a goal state,
 - A* terminates by finding an optimal path

A* Optimality

- Suppose some **suboptimal goal path G_2** has been generated and is in OPEN
- Let **n be an unexpanded node** in OPEN such that **n is on a shortest path** to an **optimal goal G**

$$h(G_2) = h(G) = 0$$

$$f(G_2) = g(G_2) \quad f(G) = g(G)$$

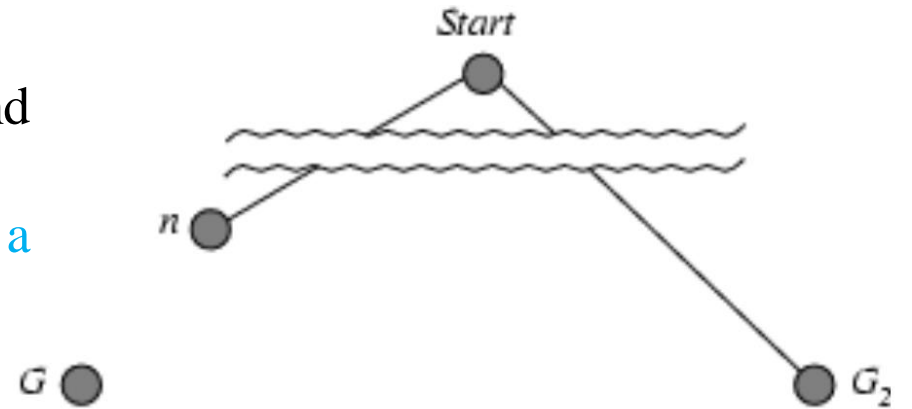
$$g(G_2) > g(G) \quad G_2 \text{ is suboptimal}$$

$$f(G_2) > f(G)$$

$$h(n) \leq h^*(n) \quad h \text{ is admissible, } h^* \text{ is minimal distance}$$

$$g(n) + h(n) \leq g(n) + h^*(n)$$

$$f(n) < f^*(G) \quad f(G_2) > f(n) \quad \text{A* will never select } G_2 \text{ for expansion}$$



A*: Result

- Algorithm A* is admissible
 - If there is a path from s to a goal state,
 - A* terminates by finding an optimal path
- If A1 and A2 are two versions of A* such that A2 is more informed than A1
 - A1 expands at least as many states as does A2
 - $\forall_n h_2(n) > h_1(n)$
 - h_2 is tighter than h_1
 - Claim: $f(n) < C^*$ then n must be expanded

A*: Result

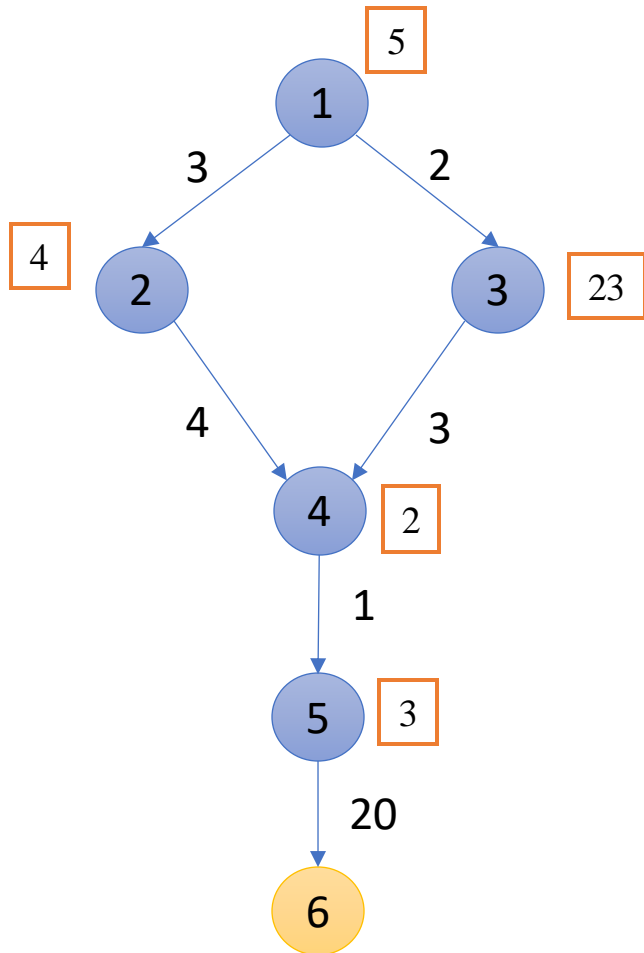
- We have two good heuristic functions h_1 and h_2 but do not know which one is more informed
 - Which one to use?
 - $\forall_n \max\{h_1(n), h_2(n)\}$
- How much effort should we put in computing heuristics?
 - If state space is complex we may invest time

A* Analysis

20/01/2025

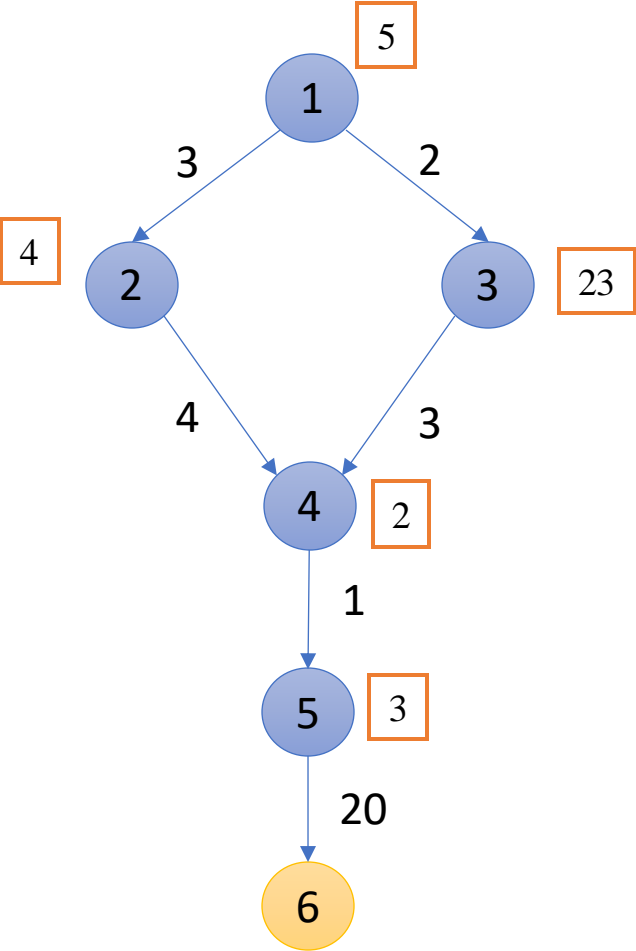
Koustav Rudra

Algorithm A*



| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|---------------|--------|------|---------------|---|
| [1(5)] | 1(5) | N | [2(7),3(25)] | [1(5)] |
| [2(7),3(25)] | 2(7) | N | [3(25),4(9)] | [1(5),2(7)] |
| [3(25),4(9)] | 4(9) | N | [3(25),5(11)] | [1(5),2(7),4(9)] |
| [3(25),5(11)] | 5(11) | N | [3(25),6(28)] | [1(5),2(7),4(9),5(11)] |
| [3(25),6(28)] | 3(25) | N | [6(28),4(7)] | [1(5),2(7), 4(9) ,5(11),3(25)] |

Algorithm A*

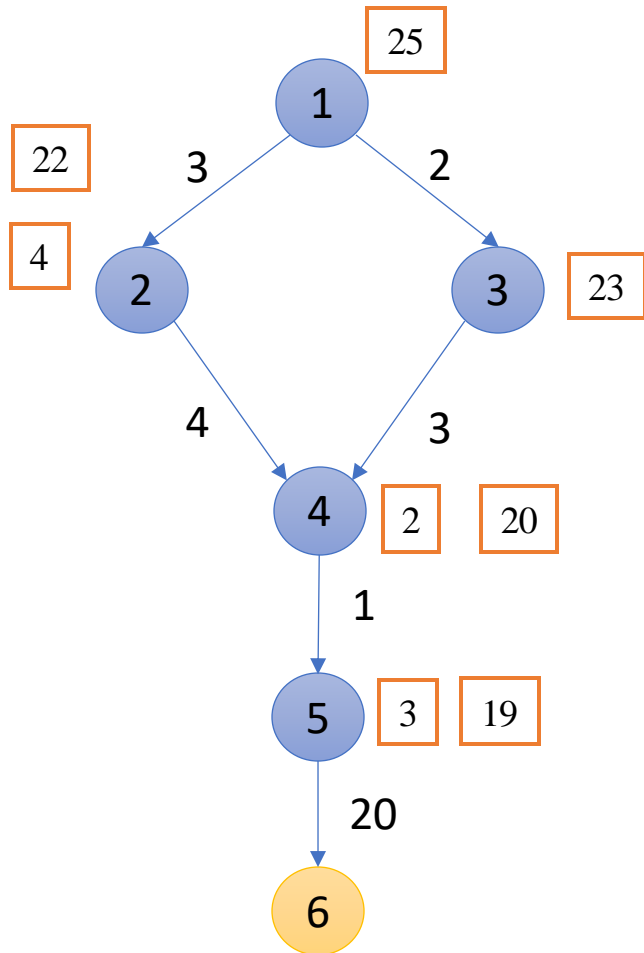


| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|---------------|--------|------|--------------|--|
| [3(25),6(28)] | 3(25) | N | [6(28),4(7)] | [1(5),2(7), 4(9) ,5(11),3(25)] |
| [6(28),4(7)] | 4(7) | N | [6(28),5(9)] | [1(5),2(7), 4(9) , 5(11) ,3(25),4(7)] |
| [6(28),5(9)] | 5(9) | N | [6(26)] | [1(5),2(7), 4(9) , 5(11) ,3(25),4(7),5(9)] |
| [6(26)] | 6(26) | Y | | |

Monotone Heuristics

- An admissible heuristic function, $h()$, is monotonic if for every successor m of n :
 - $h(n) - h(m) \leq C(n, m)$

Algorithm A*



| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|---------------|--------|------|---------------|---------------------------------|
| [1(25)] | 1(25) | N | [3(25),2(7)] | [1(25)] |
| | | | [3(25),2(25)] | |
| [3(25),2(25)] | 3(25) | N | [2(25),4(7)] | [1(25),3(25)] |
| | | | [2(25),4(25)] | |
| [2(25),4(25)] | 2(25) | N | [4(25)] | [1(25),3(25),2(25)] |
| [4(25)] | 4(25) | N | [5(9)] | [1(25),3(25),2(25),4(25)] |
| | | | [5(25)] | |
| [5(25)] | 5(25) | N | [6(26)] | [1(25),3(25),2(25),4(25),5(25)] |
| [6(25)] | 6(25) | Y | | |

Monotone Heuristics

- An admissible heuristic function, $h()$, is monotonic if for every successor m of n :
 - $h(n) - h(m) \leq C(n, m)$
- If the monotone heuristic is satisfied, then A^* has already found an optimal path to the state it selects for expansion
 - $f(m) = g(m) + h(m)$
 - $f(m) = g(n) + C(n, m) + h(m)$
 - $f(m) \geq g(n) + h(n)$
 - $f(m) \geq f(n)$

$f(n)$ is non-decreasing along any path

If $h(n)$ is consistent/monotone, A^* using GRAPH-SEARCH is optimal

Monotone Heuristics

- An admissible heuristic function, $h()$, is monotonic if for every successor m of n :
 - $h(n) - h(m) \leq C(n, m)$
- If the monotone heuristic is satisfied, then A^* has already found an optimal path to the state it selects for expansion
- How to convert a non-monotonic heuristic to monotonic one?

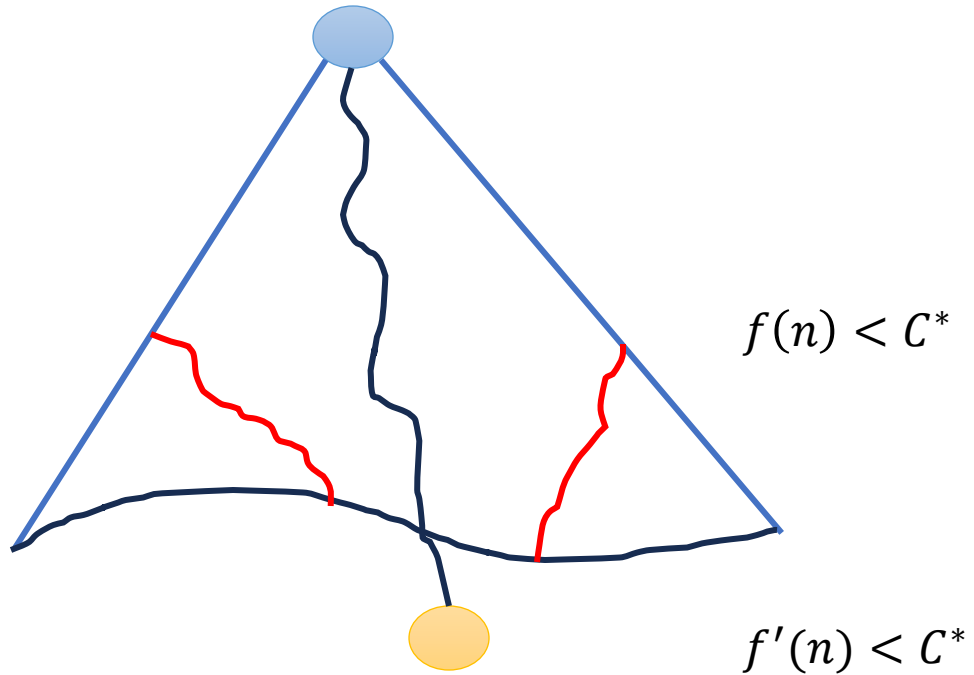
Pathmax

- Converts a non-monotonic heuristic to a monotonic one
 - During generation of the successor, m , of n we set:
 - $h'(m) = \max\{h(m), h(n) - C(n, m)\}$
 - Use $h'(m)$ as heuristic at m

A*: Result

- Heuristic Function is weak
 - Lots of states having cost $< C^*$
 - Need to explore lots of states
 - Less pruning
- Heuristic function is zero
 - Uniform cost search
- Heuristic function is overestimating
 - More pruning
 - Less number of states to explore
 - Does that always help?

A*: Result



When to go for overestimating?

Are we satisfied with sub-optimal solution?

In some cases we may get 10-50 times faster solution

Inadmissible Heuristic

- Advantages

- In many cases, inadmissible heuristics can cause better pruning, and
- Significantly reduce the search time

- Drawbacks

- A* may terminate with a suboptimal solution

Algorithm A*: Result

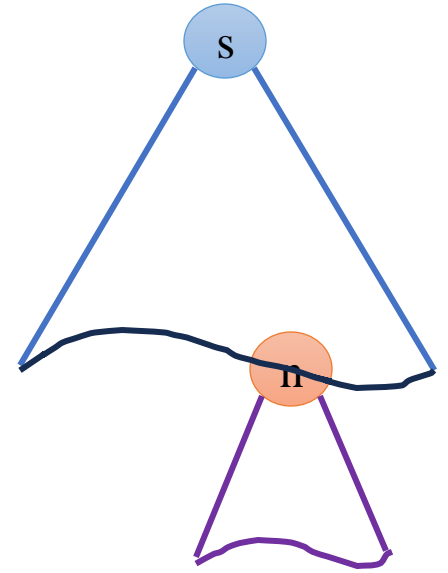
- Is A* good?
 - Uses too much memory
- Why should we concern about the shift of nodes from CLOSED to OPEN?
 - We want to expand a node only once
 - Derive a set of states should be expanded by any admissible algorithm
 - Any optimal algorithm has to expand the nodes $f(n) < C^*$

Algorithm A*: Result

- Complexity would be linear in terms of expanded nodes
 - If we can ensure every node is expanded only once
- If S is the set of nodes which must be visited by any admissible algorithm
 - Then the complexity is linear in S
 - $S = \{n | f(n) < C^*\}$
- Any algorithm which is admissible or any algorithm which guarantees to give us the optimal solution
 - Will have to expand this set of states
- Algorithm that is linear in the size of S is asymptotically optimal
 - Any algorithm that is admissible and guarantees optimal solution will have to explore $|S|$ nodes

Iterative Deepening A* (IDA*)

- Set $C = f(s)$
- Perform DFBB with cut-off C
 - Expand a state, n , only if its f -value is less than or equal to C
 - If a goal is selected for expansion then return C and terminate
- Update C to the minimum f -value which exceeded C among states which were examined
- Go to Step 2



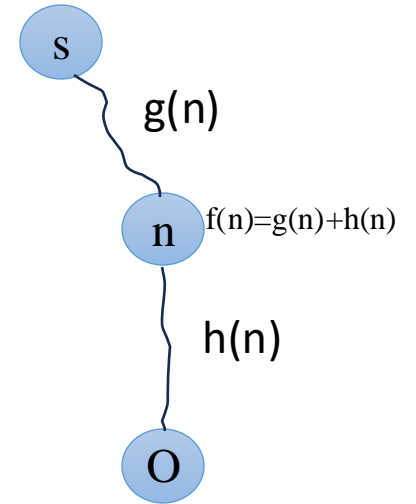
Iterative Deepening A* (IDA*)

- In the worst case, only one new state is expanded in each iteration
 - If A* expands N states, then IDA* can expand:
 - $1 + 2 + 3 + \dots + N = O(N^2)$
- Space Complexity:
 - Linear
- IDA* is asymptotically optimal

Variations

BEST-FIRST Tree Search

- **Initialize:** Set $OPEN = \{s\}$, $CLOSED = \{\}$, $f(s) = h(s)$
- **Fail:**
 - If $OPEN = \{\}$, Terminate with failure
- **Select:** Select the minimum cost state, n , from $OPEN$ and save in $CLOSED$
- **Terminate:**
 - If $n \in G$, terminate with success



BEST-FIRST Tree Search

- **Expand:**

- For each successor, m , of n :
 - If $m \notin [\text{OPEN} \cup \text{CLOSED}]$
 - Set $f(m) = h(m)$
 - Insert m in OPEN

BEST-FIRST Tree Search with pruning

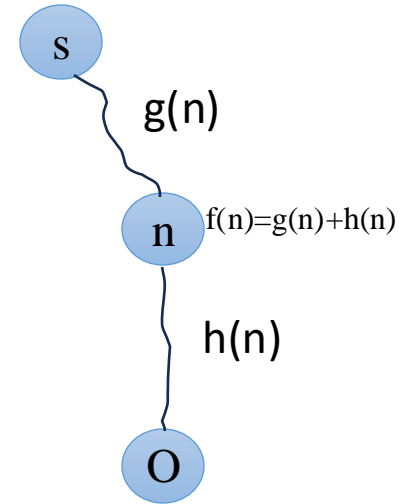


- **Loop:**

- Go to step 2

BEST-FIRST Tree Search [pruning]

- **Initialize:** Set $OPEN=\{s\}$, $CLOSED = \{ \}$, $f(s) = h(s)$, **CB**
- **Fail:**
 - If $OPEN=\{ \}$, Terminate with failure
- **Select:** Select the minimum cost state, **n**, from OPEN and save in CLOSED
- **Terminate:**
 - If $n \in G$ and $f(n) < CB$, $CB = f(n)$, Go to Step 2
 - Else terminate



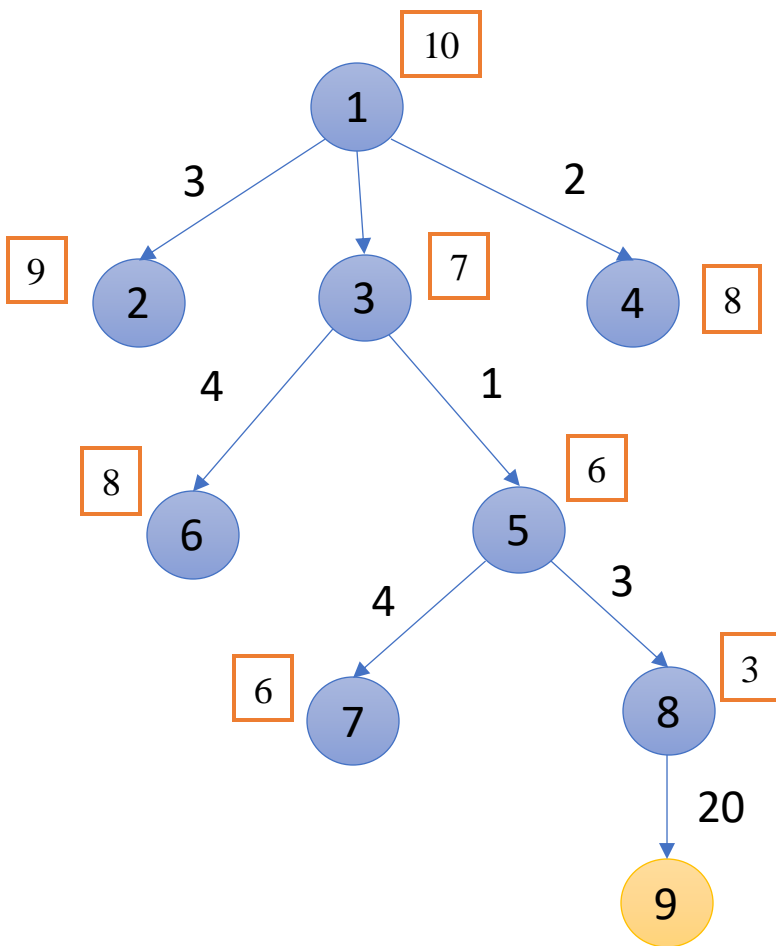
BEST-FIRST Tree Search [pruning]

- **Expand:**
 - If $f(n) \leq CB$
 - For each successor, m , of n :
 - If $m \notin [OPEN \cup CLOSED]$
 - Set $f(m) = h(m)$
 - Insert m in OPEN



- **Loop:**
 - Go to step 2

BEST-FIRST Tree Search



| OPEN SET | SELECT | GOAL | EXPANDED | CLOSED |
|----------------------------|--------|------|----------------------------|------------------------|
| [1(10)] | 1(10) | N | [2(9),3(7),4(8)] | [1(10)] |
| [2(9),3(7),4(8)] | 3(7) | N | [2(9),4(8),6(8),5(6)] | [1(10),3(7)] |
| [2(9),4(8),6(8),5(6)] | 5(6) | N | [2(9),4(8),6(8),7(6),8(3)] | [1(10),3(7),5(6)] |
| [2(9),4(8),6(8),7(6),8(3)] | 8(3) | N | [2(9),4(8),6(8),7(6),9(0)] | [1(10),3(7),5(6),8(3)] |
| [2(9),4(8),6(8),7(6),9(0)] | 9(0) | Y | | |

Thank You