

# HILL CLIMBING & BEST FIRST SEARCH

Dr. Krishnendu Guha

Assistant Professor (On Contract)

National Institute of Technology (NIT), Jamshedpur

Email: [krishnendu.ca@nitjsr.ac.in](mailto:krishnendu.ca@nitjsr.ac.in)

# HILL CLIMBING

- Hill Climbing is heuristic search used for mathematical optimization problems in the field of AI
- In the above definition, mathematical optimization problems implies that hill climbing solves the problems where we need to maximize or minimize a given real function by choosing values from the given inputs
- Given a large set of inputs and a good heuristic function, it tries to find a sufficiently good solution to the problem
- This solution may not be the global optimal maximum
- Example-[Travelling salesman problem](#) where we need to minimize the distance traveled by salesman
- Types of Hill Climbing: Simple Hill Climbing, Steepest Ascent Hill Climbing, Stochastic Hill Climbing

# FEATURES OF HILL CLIMBING

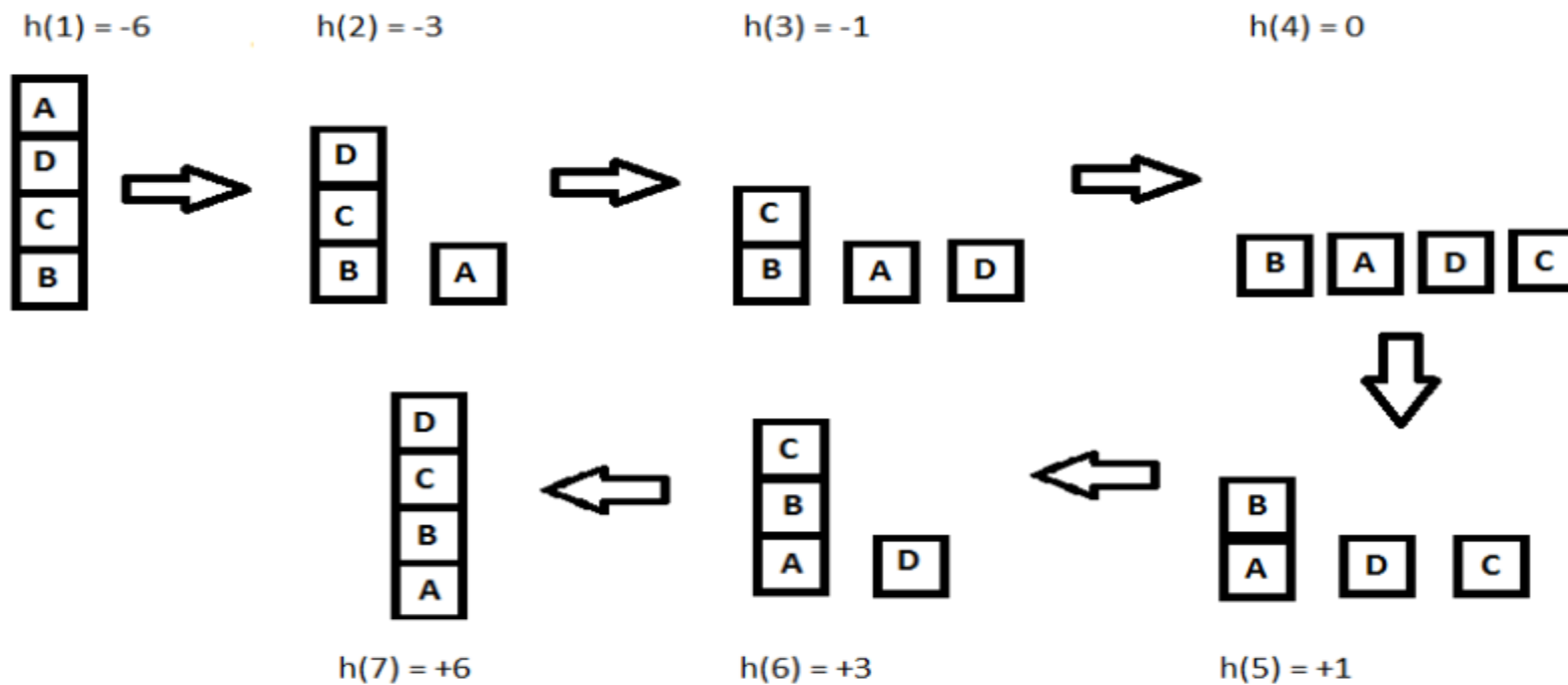
Hill Climbing is a Variant of generate and test algorithm

- *Generate and Test*
  - *1. Generate a possible solutions.*
  - *2. Test to see if this is the expected solution.*
  - *3. If the solution has been found quit else go to step 1.*
- We call Hill climbing as a variant of generate and test algorithm as it takes the feedback from test procedure
- Then this feedback is utilized by the generator in deciding the next move in search space

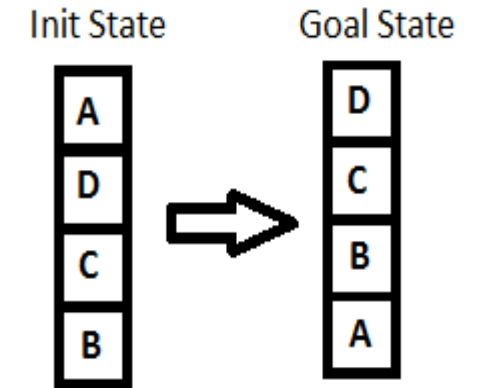
Hill Climbing Uses the Greedy approach

- At any point in state space, the search moves in that direction only which optimizes the cost of function with the hope of finding the optimal solution at the end

- Key point while solving any hill-climbing problem is to choose an appropriate heuristic function.
- Let's define such function  $h$ :
- $h(x) = +1$  for all the blocks in the support structure if the block is correctly positioned otherwise  $-1$  for all the blocks in the support structure



# EXAMPLE



# SIMPLE HILL CLIMBING

- It examines the neighboring nodes one by one and selects the first neighboring node which optimizes the current cost as next node

Algorithm :

- *Step 1 : Evaluate the initial state. If it is a goal state then stop and return success.*
- *Otherwise, make initial state as current state.*
- *Step 2 : Loop until the solution state is found or there are no new operators present which can be applied to current state.*
  - *a) Select a state that has not been yet applied to the current state and apply it to produce a new state.*
  - *b) Perform these to evaluate new state*
    - *i. If the current state is a goal state, then stop and return success.*
    - *ii. If it is better than the current state, then make it current state and proceed further.*
    - *iii. If it is not better than the current state, then continue in the loop until a solution is found.*
- *Step 3 : Exit.*

# STEEPEST ASCENT HILL CLIMBING

- It first examines all the neighboring nodes and then selects the node closest to the solution state as next node

## Algorithm

- *Step 1 : Evaluate the initial state. If it is goal state then exit else make the current state as initial state*
- *Step 2 : Repeat these steps until a solution is found or current state does not change*
  - *i. Let 'target' be a state such that any successor of the current state will be better than it;*
  - *ii. for each operator that applies to the current state*
    - *a. apply the new operator and create a new state*
    - *b. evaluate the new state*
    - *c. if this state is goal state then quit else compare with 'target'*
    - *d. if this state is better than 'target', set this state as 'target'*
    - *e. if target is better than current state set current state to Target*
- *Step 3 : Exit*

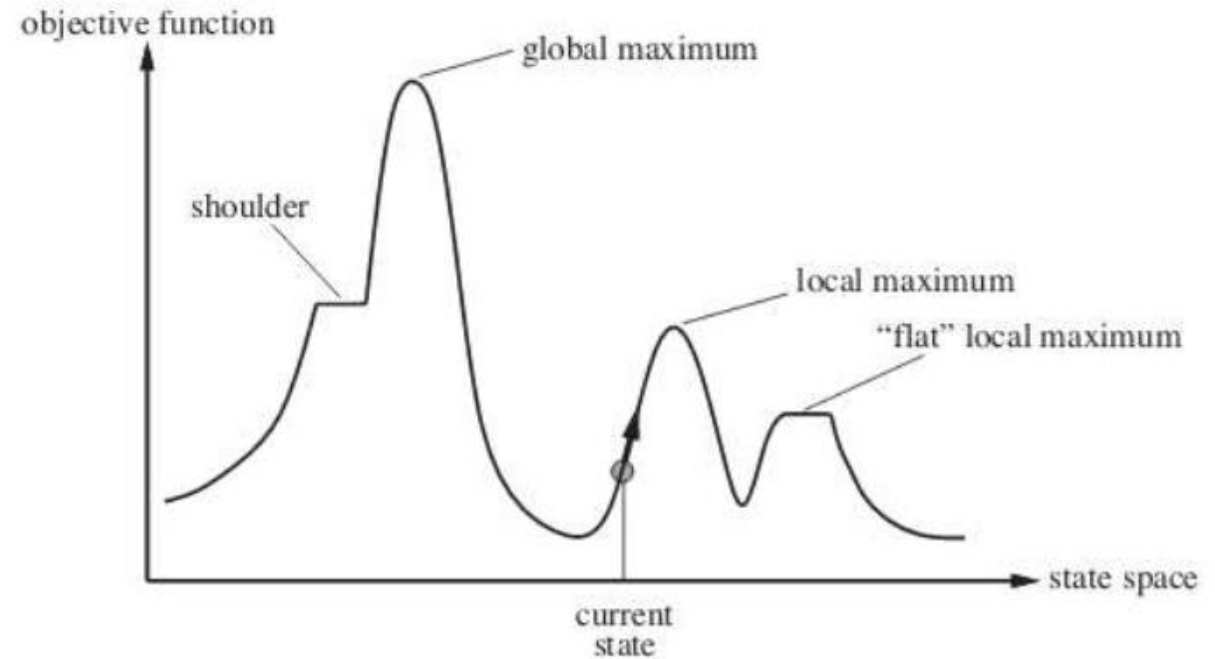
In the case of hill climbing technique we picked any state as a successor which was closer to the goal than the current state whereas, in Steepest-Ascent Hill Climbing algorithm, we choose the best successor among all possible successors and then update the current state

# STOCHASTIC HILL CLIMBING

- It does not examine all the neighboring nodes before deciding which node to select
- It just selects a neighboring node at random, and decides (based on the amount of improvement in that neighbor) whether to move to that neighbor or to examine another.

# STATE SPACE DIAGRAM FOR HILL CLIMBING

- State space diagram is a graphical representation of the set of states our search algorithm can reach vs the value of our objective function (the function which we wish to maximize).
- X-axis : denotes the state space ie states or configuration our algorithm may reach.
- Y-axis : denotes the values of objective function corresponding to a particular state.
- The best solution will be that state space where objective function has maximum value (global maximum)



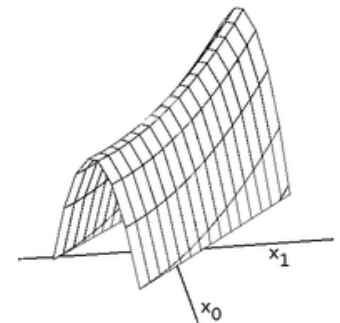
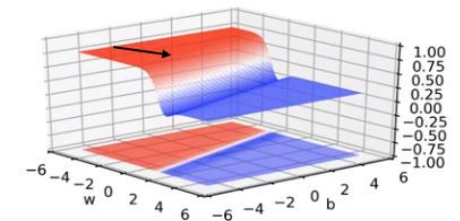
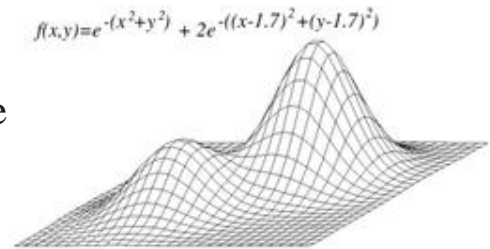


# DIFFERENT REGIONS IN THE STATE SPACE DIAGRAM

- Local maximum : It is a state which is better than its neighboring state however there exists a state which is better than it(global maximum)
  - This state is better because here value of objective function is higher than its neighbors
- Global maximum : It is the best possible state in the state space diagram
  - This because at this state, objective function has highest value
- Plateau/flat local maximum : It is a flat region of state space where neighboring states have the same value
- Ridge : It is region which is higher than its neighbours but itself has a slope.
  - It is a special kind of local maximum
- Current state : The region of state space diagram where we are currently present during the search
- Shoulder : It is a plateau that has an uphill edge

# PROBLEMS IN DIFFERENT REGIONS OF HILL CLIMBING

- Hill climbing cannot reach the optimal/best state(global maximum) if it enters any of the following regions :
- 1. Local maximum : At a local maximum all neighboring states have a value, which is worse than the current state
  - Since hill climbing uses greedy approach, it will not move to the worse state and terminate itself.
  - The process will end even though a better solution may exist.
  - To overcome local maximum problem : Utilize backtracking technique. Maintain a list of visited states. If the search reaches an undesirable state, it can backtrack to the previous configuration and explore a new path.
- 2. Plateau : On plateau all neighbors have same value . Hence, it is not possible to select the best direction.
  - To overcome plateaus : Make a big jump. Randomly select a state far away from current state.
  - Chances are that we will land at a non-plateau region
- 3. Ridge : Any point on a ridge can look like peak because movement in all possible directions is downward
  - Hence the algorithm stops when it reaches this state.
  - To overcome Ridge : In this kind of obstacle, use two or more rules before testing.
  - It implies moving in several directions at once.



# BEST FIRST SEARCH

- In BFS and DFS, when we are at a node, we can consider any of the adjacent as next node
- So both BFS and DFS blindly explore paths without considering any cost function
- The idea of Best First Search is to use an evaluation function to decide which adjacent is most promising and then explore
- Best First Search falls under the category of Heuristic Search or Informed Search
- We use a priority queue to store costs of nodes. So the implementation is a variation of BFS, we just need to change Queue to Priority Queue

### Algorithm of Best First Search:

Best-First-Search(Graph g, Node start)

1) Create an empty PriorityQueue

    PriorityQueue pq;

2) Insert "start" in pq.

    pq.insert(start)

3) Until PriorityQueue is empty

    u = PriorityQueue.DeleteMin

    If u is the goal

        Exit

    Else

        For each neighbor v of u

            If v "Unvisited"

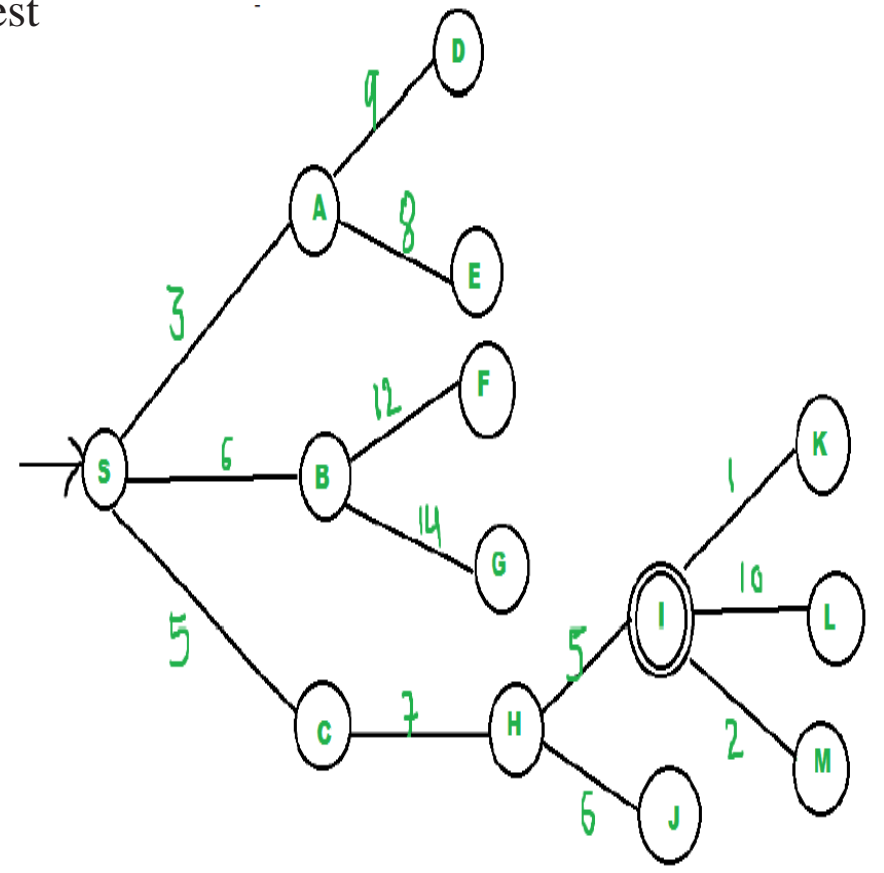
                Mark v "Visited"

                pq.insert(v)

                Mark v "Examined"

End procedure

- We start from source "S" and search for goal "I" using given costs and Best First search.
- pq initially contains S
- We remove s and process unvisited neighbors of S to pq.
- pq now contains {A, C, B} (C is put before B because C has lesser cost)
- We remove A from pq and process unvisited neighbors of A to pq.
- pq now contains {C, B, E, D}
- We remove C from pq and process unvisited neighbors of C to pq.
- pq now contains {B, H, E, D}
- We remove B from pq and process unvisited neighbors of B to pq.
- pq now contains {H, E, D, F, G}
- We remove H from pq. Since our goal "I" is a neighbor of H, we return.



# ANALYSIS

- **Worst case time complexity**
- The worst case time complexity for Best First Search is  $O(n * \text{Log } n)$  where  $n$  is number of nodes
- In worst case, we may have to visit all nodes before we reach goal
- Note that priority queue is implemented using Min(or Max) Heap, and insert and remove operations take  $O(\text{log } n)$  time.
  
- **Performance of the algorithm**
- Performance of the algorithm depends on how well the cost or evaluation function is designed.

