# DOCUMENTATION FOR ARTIFIAL INTELLIGENCE ASSIGNMENT-2

Question-1 Hill climbing informed searching algorithm.

```
INPUT-
      Start state- 2 8 3
                1 6 4
                7 0 5
      Goal state- 1 2 3
               8 0 4
               7 6 5
OUTPUT-
      2 8 3
      1 6 4
      7 0 5
      Node: 0
      Depth: 0
      Moves: []
      1.sum of Manhattan distance of each tile from the goal position
      2.number of tiles displaced from their destined position
      [18]
      2 8 3
      1 0 4
      7 6 5
      Node: 1
      Depth: 1
      Moves: ['up']
      2 8 3
      1 6 4
      0 7 5
      Node: 2
      Depth: 1
      Moves: ['left']
      2 8 3
      1 6 4
      7 5 0
      Node: 3
      Depth: 1
      Moves: ['right']
                       -----
      [16, 16, 20]
2 0 3
      1 8 4
      7 6 5
      Node: 4
      Depth: 2
      Moves: ['up', 'up']
```

```
2 8 3
0 1 4
7 6 5
Node: 5
Depth: 2
Moves: ['up', 'left']
2 8 3
1 4 0
 7 6 5
Node: 6
Depth: 2
Moves: ['up', 'right']
 [14, 14, 16]
0 2 3
 1 8 4
7 6 5
Node: 7
Depth: 3
Moves: ['up', 'up', 'left']
2 3 0
 1 8 4
7 6 5
Node: 8
Depth: 3
Moves: ['up', 'up', 'right']
 [12, 14]
 1 2 3
0 8 4
7 6 5
Node: 9
Depth: 4
Moves: ['up', 'up', 'left', 'down']
 [12]
 1 2 3
7 8 4
0 6 5
Node: 10
Depth: 5
Moves: ['up', 'up', 'left', 'down', 'down']
1 2 3
8 0 4
7 6 5
Node: 11
Moves: ['up', 'up', 'left', 'down', 'right']
Success
Time: 0.014479875564575195
 HILL CLIMBING -
```

Hill climbing is a inform search algorithm and it based on Greedy Local Search.

<sup>-&</sup>gt; Local search: use single current state and move to neighboring states.

<sup>-&</sup>gt; Are also useful for pure optimization problems. Find best state according to

some objective function.

- -> State space landscape.
  - Location (defined by state).
- -> Elevation.
  - Defined by the value of the heuristic function or objective function.
- -> Elevation corresponds to cost.
  - Global minimum (aim is to find the lowest valley).
- -> Elevation corresponds to an objective function.
  - Global maximum (aim is to find the highest peak).
- -> Complete algorithm always finds a goal if one exists.
- -> Optimal algorithm always finds global minimum/maximum.

### HILL CLIMBING SEARCH

- -> "is a loop that continuously moves in the direction of increasing value".
- -> It terminates when a peak is reached.
- -> No neighbor has higher value.
- -> does not maintain any search tree.
- -> Current node data structure records state and objective function value.
- -> does not look ahead beyond the immediate neighbors of the current state.
- -> chooses randomly among the set of best successors, if there is more than one.
- -> Hill-climbing a.k.a. greedy local search.
- -> Grabs a good neighbor state without thinking ahead about where to go next.
- -> Makes very rapid progress towards a solution.
- -> Quite easy to improve a bad state.
- -> Some problem spaces are great for hill climbing and others are terrible.

#### ALGORITHM

```
function HILL-CLIMBING(problem) return a state that is a
  (global) maximum
  input: problem, a problem
  local variables: current, a node.
      neighbor, a node.
  Current <--MAKE-NODE(INITIAL-STATE[problem])
  loop do
      neighbor <-- a highest valued successor of current
      if VALUE [neighbor] ≤ VALUE[current] then return
  STATE[current]
      current <-- neighbor</pre>
```

#### ADVANTAGE

- 1. Uses very little memory.
- 2. Finds often reasonable solutions in large or infinite state spaces.

### **DRAWBACKS**

- 1. Local Maxima
  - (A) peaks that aren't the highest point in the space (below global maxima).

- (B) higher than each of its neighboring states.
- 2. Plateau: region where evaluation function is flat.
  - (A) Flat local maximum: no uphill exit exists.
  - (B) Shoulder: possible to make progress.
  - (C) Could give the search algorithm no direction (random walk) for local maximum.
- 3. Ridges:
  - (A) flat like a plateau, but with drop-offs to the sides; steps to the North, East, South and West may go down, but a combination of two steps (e.g. N, W) may go up.
  - (B) results in a sequence of local maximum not connected to each other.

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Question-2 Simulated Annealing informed search.

### INPUT-

Start state- 2 8 3 1 6 4 7 0 5

Goal state- 1 2 3 8 0 4 7 6 5

### OUTPUT-

Started running simulated annealing:

goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[2, 8, 3, 1, 6, 4, 7, 0, 5]

Choosing random movement:E state after move:[2, 8, 3, 1, 6, 4, 7, 5, 0] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[2, 8, 3, 1, 6, 4, 7, 5, 0]

Choosing random movement:W state after move: [2, 8, 3, 1, 6, 4, 7, 0, 5] goal: [1, 2, 3, 8, 0, 4, 7, 6, 5] state: [2, 8, 3, 1, 6, 4, 7, 0, 5]

Choosing random movement:N state after move: [2, 8, 3, 1, 0, 4, 7, 6, 5] goal: [1, 2, 3, 8, 0, 4, 7, 6, 5] state: [2, 8, 3, 1, 0, 4, 7, 6, 5]

Choosing random movement:S state after move:[2, 8, 3, 1, 6, 4, 7, 0, 5]

goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[2, 8, 3, 1, 6, 4, 7, 0, 5]

Choosing random movement:N state after move: [2, 8, 3, 1, 0, 4, 7, 6, 5] goal: [1, 2, 3, 8, 0, 4, 7, 6, 5] state: [2, 8, 3, 1, 0, 4, 7, 6, 5]

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Choosing random movement:N state after move: [2, 8, 3, 0, 6, 4, 1, 7, 5] goal: [1, 2, 3, 8, 0, 4, 7, 6, 5] state: [2, 8, 3, 0, 6, 4, 1, 7, 5]

Choosing random movement:N state after move:[0, 8, 3, 2, 6, 4, 1, 7, 5] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[0, 8, 3, 2, 6, 4, 1, 7, 5]

Choosing random movement: E state after move: [8, 0, 3, 2, 6, 4, 1, 7, 5] goal: [1, 2, 3, 8, 0, 4, 7, 6, 5] state: [8, 0, 3, 2, 6, 4, 1, 7, 5]

Choosing random movement:W

state after move: [0, 8, 3, 2, 6, 4, 1, 7, 5] goal: [1, 2, 3, 8, 0, 4, 7, 6, 5] state: [0, 8, 3, 2, 6, 4, 1, 7, 5]

Choosing random movement: E state after move: [8, 0, 3, 2, 6, 4, 1, 7, 5] goal: [1, 2, 3, 8, 0, 4, 7, 6, 5] state: [8, 0, 3, 2, 6, 4, 1, 7, 5]

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Choosing random movement:S state after move:[8, 3, 4, 2, 6, 0, 1, 7, 5] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[8, 3, 4, 2, 6, 0, 1, 7, 5]

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Choosing random movement:S state after move:[8, 6, 3, 2, 0, 4, 1, 7, 5] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[8, 6, 3, 2, 0, 4, 1, 7, 5]

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Choosing random movement:N state after move:[8, 6, 3, 2, 7, 0, 1, 5, 4] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[8, 6, 3, 2, 7, 0, 1, 5, 4]

Choosing random movement:N state after move:[8, 6, 0, 2, 7, 3, 1, 5, 4] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[8, 6, 0, 2, 7, 3, 1, 5, 4]

Choosing random movement:W state after move:[8, 0, 6, 2, 7, 3, 1, 5, 4] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[8, 0, 6, 2, 7, 3, 1, 5, 4]

Choosing random movement:S state after move:[8, 7, 6, 2, 0, 3, 1, 5, 4] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[8, 7, 6, 2, 0, 3, 1, 5, 4]

Choosing random movement:N state after move:[8, 0, 6, 2, 7, 3, 1, 5, 4] goal:[1, 2, 3, 8, 0, 4, 7, 6, 5] state:[8, 0, 6, 2, 7, 3, 1, 5, 4]

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```
Choosing random movement:W
state after move:[8, 6, 3, 2, 7, 4, 1, 0, 5]
goal:[1, 2, 3, 8, 0, 4, 7, 6, 5]
state:[8, 6, 3, 2, 7, 4, 1, 0, 5]
```

## Simulated Annealing-

- -> Hill climbing that does not make downhill move is incomplete.
- -> Pure random walk: choosing successor from a list is complete but inefficient.
- -> Solution: hill climbing + random walk =simulated annealing to ensure both efficiency and completeness.

Use a more complex Evaluation Function:

- -> Do sometimes accept candidates with higher cost to escape from local optimum.
  - Idea: but gradually decrease their size and frequency.
- -> Adapt the parameters of this evaluation function during execution.
- -> Based upon the analogy with the simulation of the annealing of solids.

#### Others Names-

- 1. Monte Carlo Annealing
- 2. Statistical Cooling
- 3. Probabilistic Hill Climbing
- 4. Stochastic Relaxation
- 5. Probabilistic Exchange Algorithm

# Analogy-

- ->Slowly cool down a heated solid, so that all particles arrange in the ground energy state.
- -> At each temperature wait until the solid reaches its thermal equilibrium.
- -> Probability of being in a state with energy E: Pr { E = E } = 1/Z(T) . exp (-E/k B .T)
- E Energy
- T Temperature
- kB Boltzmann constant
- *Z*(*T*) *Normalization factor (temperature dependent)*

# \_ At a fixed temperature T:

- Perturb (randomly) the current state to a new state
- ΔE is the difference in energy between new and current state
- If  $\Delta E < 0$  (new state is lower), accept new state as current state.
- If  $\Delta E \ge 0$ , accept new state with probability  $Pr(accepted) = exp(-\Delta E/kB.T)$ .
- Eventually the systems evolves into thermal equilibrium

## at temperature T.

• When equilibrium is reached, temperature *T* can be lowered and the process can be repeated.

# Difference between Hill Climbing and Simulated Annealing

- -> *Unlike hill climbing, it does not always pick the best move.*
- -> *SA picks* the move randomly.
- -> *If situation improves then accept the move.*
- -> Otherwise, accept the move with some probability.
- -> Probability decreases as the temperature goes down.
- -> Probability decreases exponentially with the badness of the move.
- -> Bad moves are more likely to be allowed at the start when temperature is high, and they are unlike when temperature decreases.