

Notes

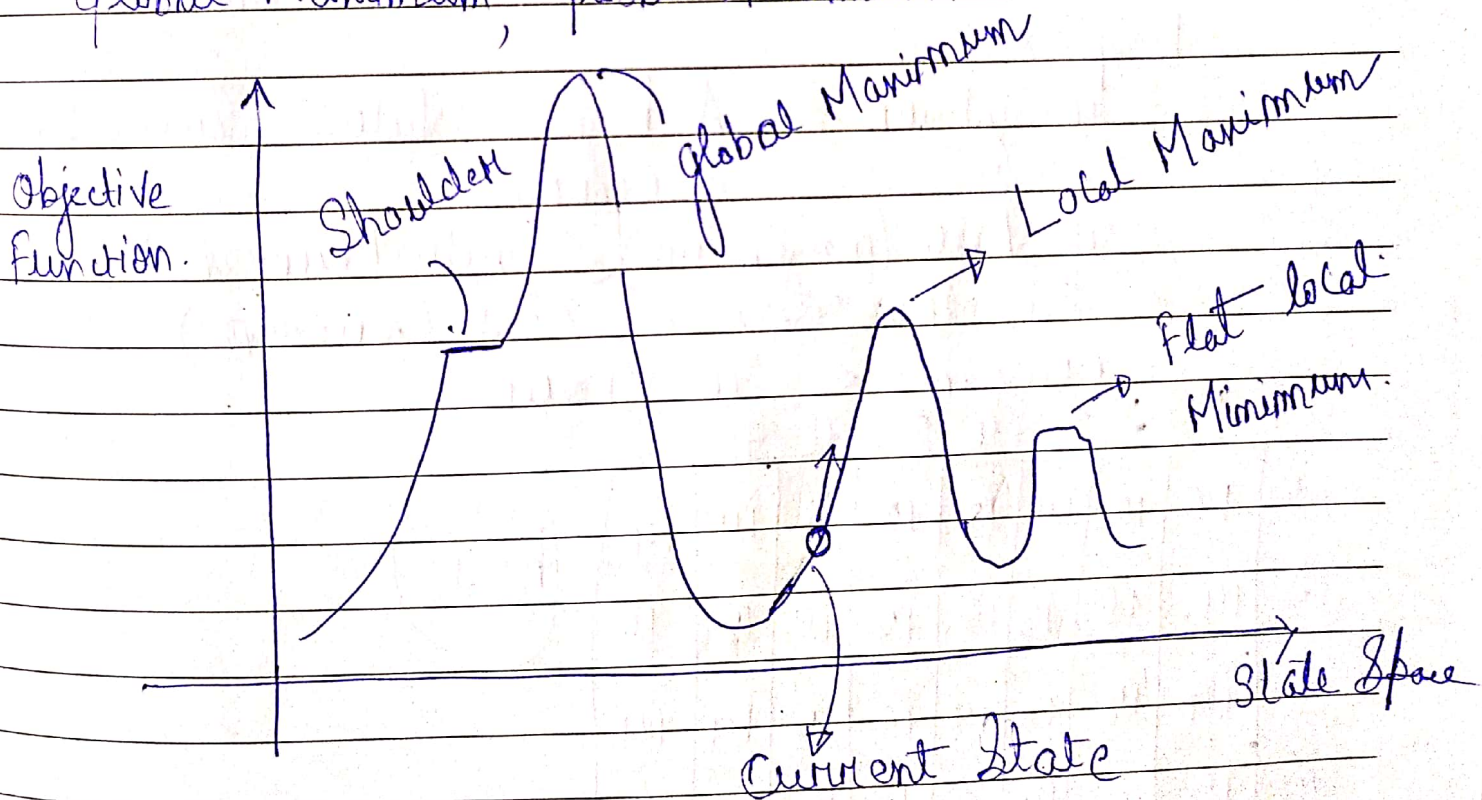
Local Search Algorithms : →

Date | | | |

- * no Path Maintenance
- * operates on a single current state
- * Not Systematic
- * Useful to solve the pure optimization Problem
- * Find the best states according to an objective function heuristic cost function.

State Space landscape

Global Maximum, Global Minimum



* Complete State formulation

Ex: 8-Queens Problem.

Notes

Algorithm:

→ Return a State which is local max^m

Hill Climbing (Problem)

Step 1: Start

inputs: Problem

local Variables: current_node

neighbour_node

current ← Initial-State (Problem)

Loop do

neighbour ← A highest valued Successor of current

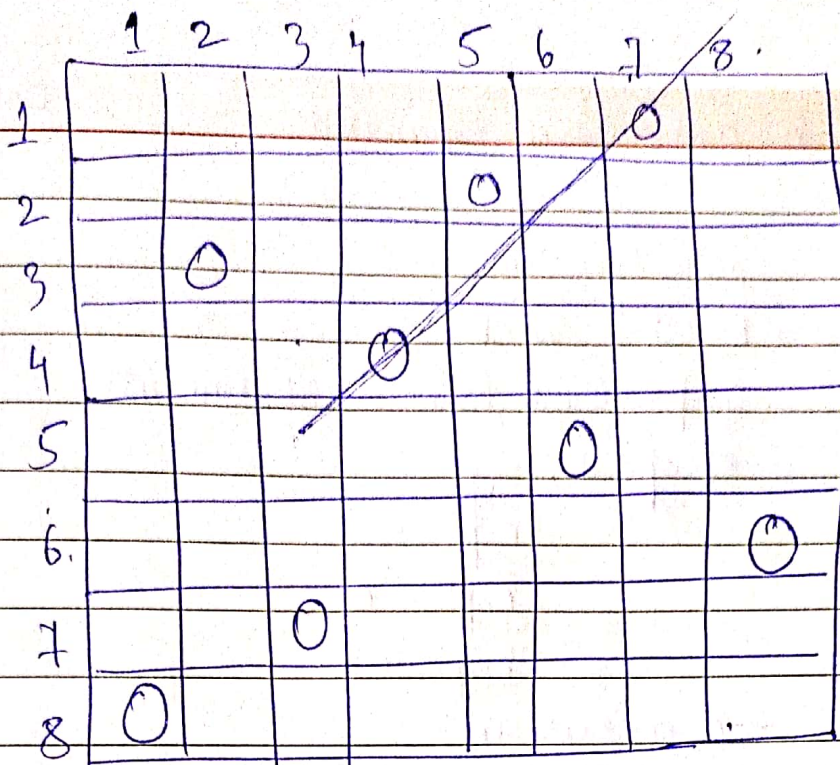
If Value(neighbour) < Value(current)
then return State(current)

current ← neighbour.

1 2 3 4 5 6 7 8

1	18	12	14	13	13	12	14	14
2	14	16	13	15	12	14	12	16
3	14	12	18	13	15	12	14	14
4	15	14	14	13	13	16	13	16
5	13	14	17	15	14	16	16	16
6	17	13	16	18	15	13	15	13
7	18	14	13	15	15	14	13	16
8	14	14	13	17	12	14	12	18

$h = 17$



Hill-climbing: Greedy local search.

Problems: →

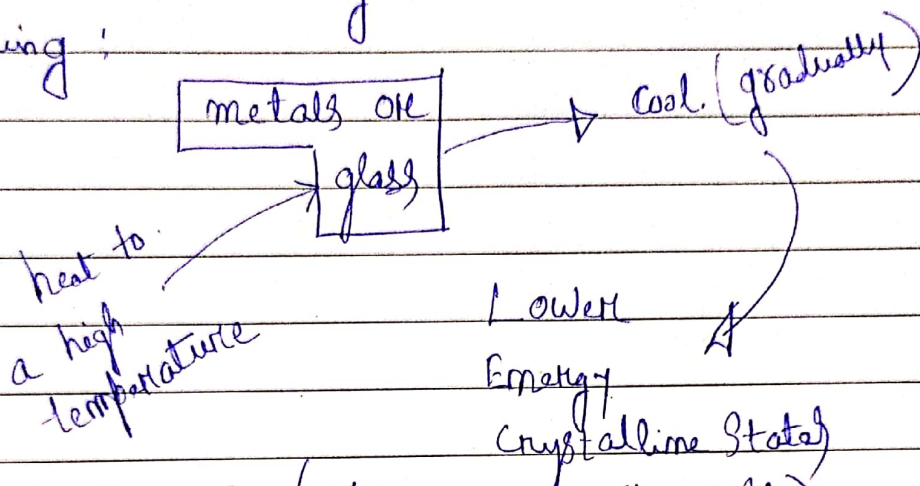
- * Local Maxima
- * Ridges
- * Plateaux

- * Side-Ways moves
- * Stochastic hill climbing
- * First choice " "
- * Random Restart hill climbing

① Simulated Annealing: →

Hill climbing + Random walk

Annealing:



② gradient Descent (Minimizing the cost)

③ Example of the ball on a bumpy surface.

④ ^{SA} Instead of the best move, pick the random moves. If the move improves the situation, then Accepted.

Else

Accept the move with Probability less than 1

The Probability goes down; when the temperature goes down.

Algorithm:

SA(Problem, Schedule) → Return a State

current ← Initial state

for $l \leftarrow 1$ to ∞ , do

$T \leftarrow \text{Schedule}[l]$

$T =$ temperature prob. controlling the downward.

if $T = 0$, then return current

next ← random-successors(current)

$\Delta E \leftarrow \text{Value}[\text{next}] - \text{Value}[\text{current}]$

if $\Delta E > 0$, then current ← next

else

current ← next only with probability $e^{-\Delta E/T}$