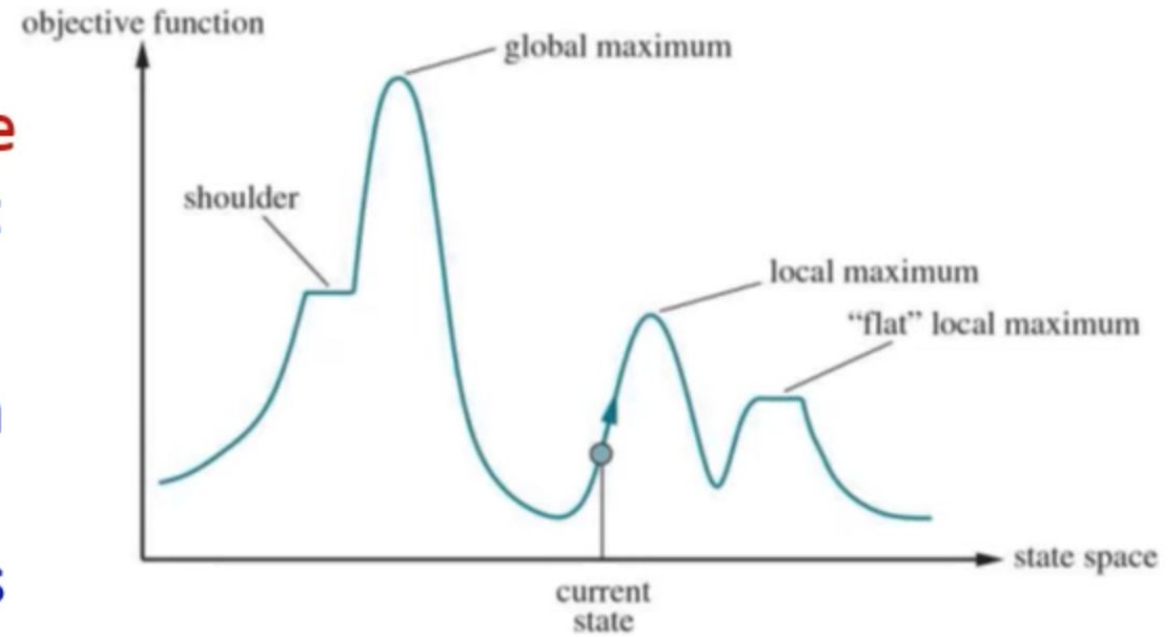


Search Techniques

SIMPLE HILL CLIMBING SEARCH ALGORITHM

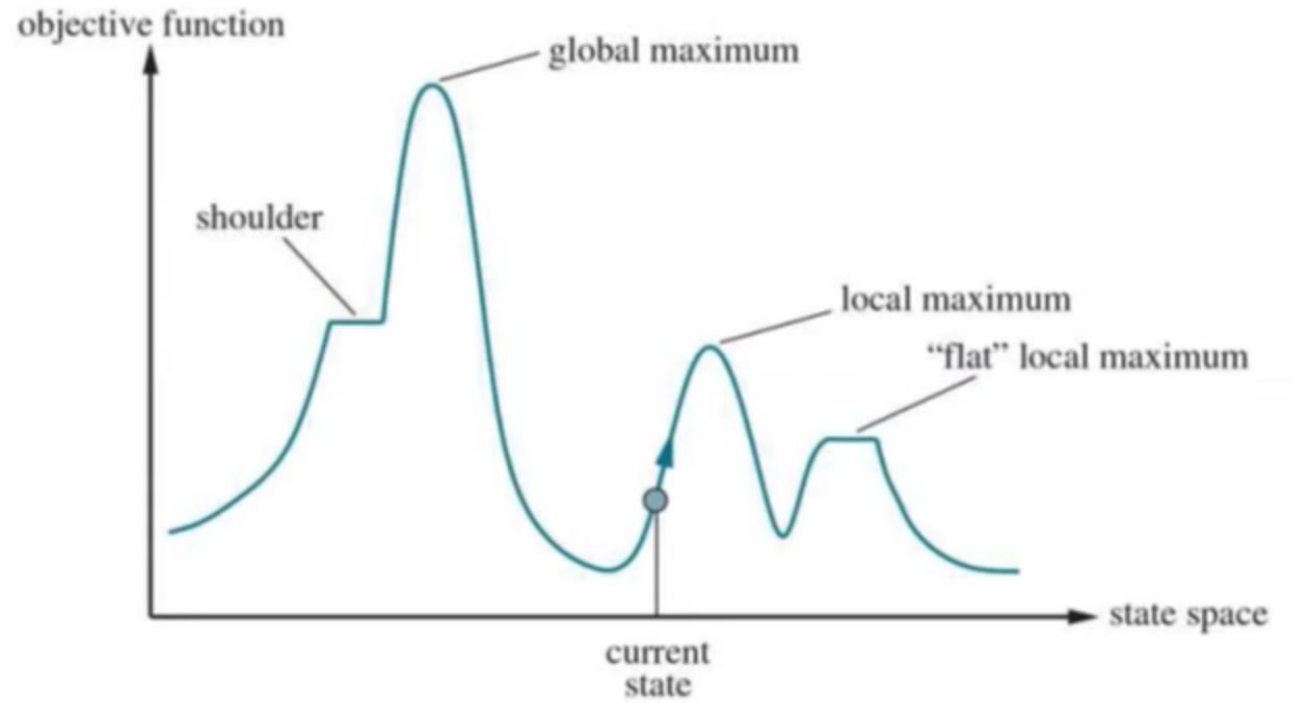
Hill-climbing Search Algorithm

- Hill climbing algorithm is a **Heuristic search** algorithm which continuously moves in the direction of **increasing value** to **find the peak of the mountain or best solution to the problem**.
- It keeps track of **one current state** and on each iteration moves to the **neighboring state with highest value**—that is, it heads in the direction that provides the **steepest ascent**.



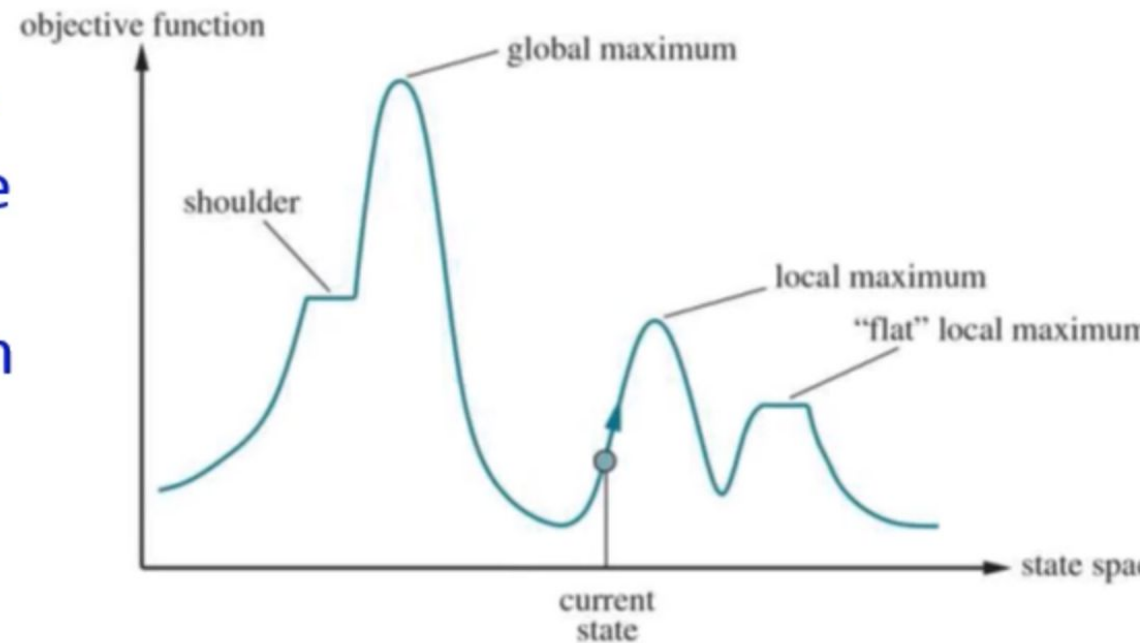
Hill-climbing Search Algorithm...

- In this algorithm, when it reaches a peak value where no neighbor has a higher value, then it terminates.
- It is also called **greedy local search** as it only searches its good immediate neighbor state and not beyond that.
- Hill Climbing is mostly used when a good heuristic is available.



Hill-climbing Search Algorithm

- Different regions in the state space landscape:
- **Local Maximum** is a state which is better than its neighbor states, but there is also another state which is higher than it.
- **Global Maximum** is the best possible state of state space landscape. It has the highest value of objective function.
- **Current state** is a state in a landscape diagram where an agent is currently present.
- **Flat local maximum** is a flat space in the landscape where all the neighbor states of current states have the same value.
- **Shoulder** is a plateau region which has an uphill edge.



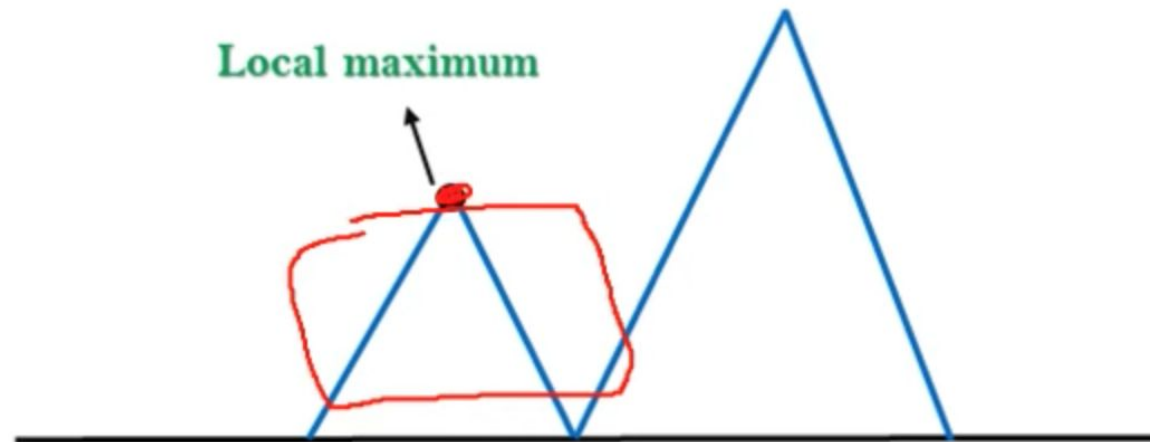
Hill-climbing Search Algorithm...

- The hill-climbing search algorithm, which is the most basic local search technique.
- At each step the **current node is replaced by the best neighbor**.

function HILL-CLIMBING(problem) **returns** a state that is a local maximum
 $current \leftarrow problem.INITIAL$
 while *true* **do**
 $neighbor \leftarrow$ a highest-valued successor state of *current*
 if $VALUE(neighbor) \leq VALUE(current)$ **then return** *current*
 $current \leftarrow neighbor$

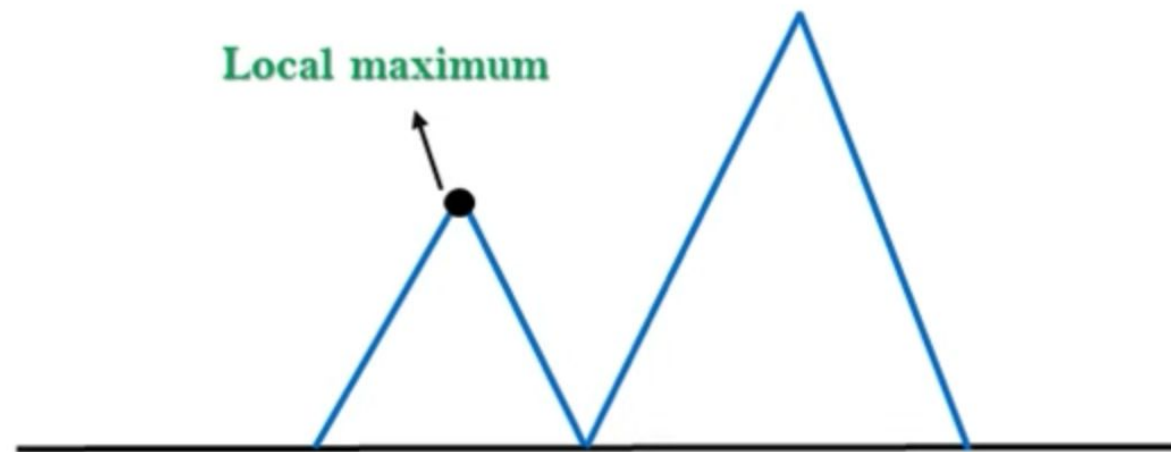
Problems in Hill_Climbing Algorithm

- **1. Local Maximum:** A local maximum is a peak state in the landscape which is better than each of its neighboring states, but there is another state also present which is higher than the local maximum.
- **Solution:** Backtracking technique can be a solution of the local maximum in state space landscape.
- Create a list of the promising path so that the algorithm can backtrack the search space and explore other paths as well.



Problems in Hill_Climbing Algorithm

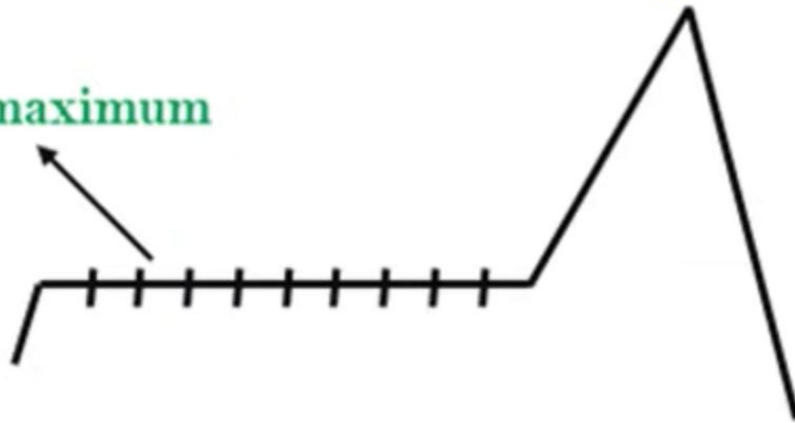
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- **Solution:** Backtracking technique can be a solution of the local maximum in state space landscape.
- Create a list of the promising path so that the algorithm can backtrack the search space and explore other paths as well.



Problems in Hill Climbing Algorithm...

- **2. Plateau:** A plateau is the flat area of the search space in which all the neighbor states of the **current state contains the same value**, because of this algorithm does not find any best direction to move.
- A hill-climbing search might be lost in the plateau area.
- **Solution:** The solution for the plateau is to **take big steps** while searching, to solve the problem.
- **Randomly** select a state which is **far away from the current state** so it is possible that the algorithm could find non-plateau region.

Plateau/Flat maximum



Problems in Hill Climbing Algorithm...

- **3. Ridges:** A ridge is a special form of the local maximum.
- It has an area, which is higher than its surrounding areas, but itself has a slope, and cannot be reached in a single move.
- **Solution:** With the use of **bidirectional search**, or by moving in different directions, we can improve this problem.

Ridge



Physical Annealing

- The Simulated Annealing algorithm is based upon Physical Annealing in real life.
- Physical Annealing is the process of **heating up a material** until it reaches an **annealing temperature** and then
- it will be **cooled down** slowly in order to change the material to a desired structure.
- When the material is **hot**, the molecular structure is **weaker** and is more susceptible to change.
- When the material **cools** down, the molecular structure is **harder** and is less susceptible to change.



Physical Annealing...

- Thermal Dynamics Equation calculates the probability that the Energy Magnitude will increase.

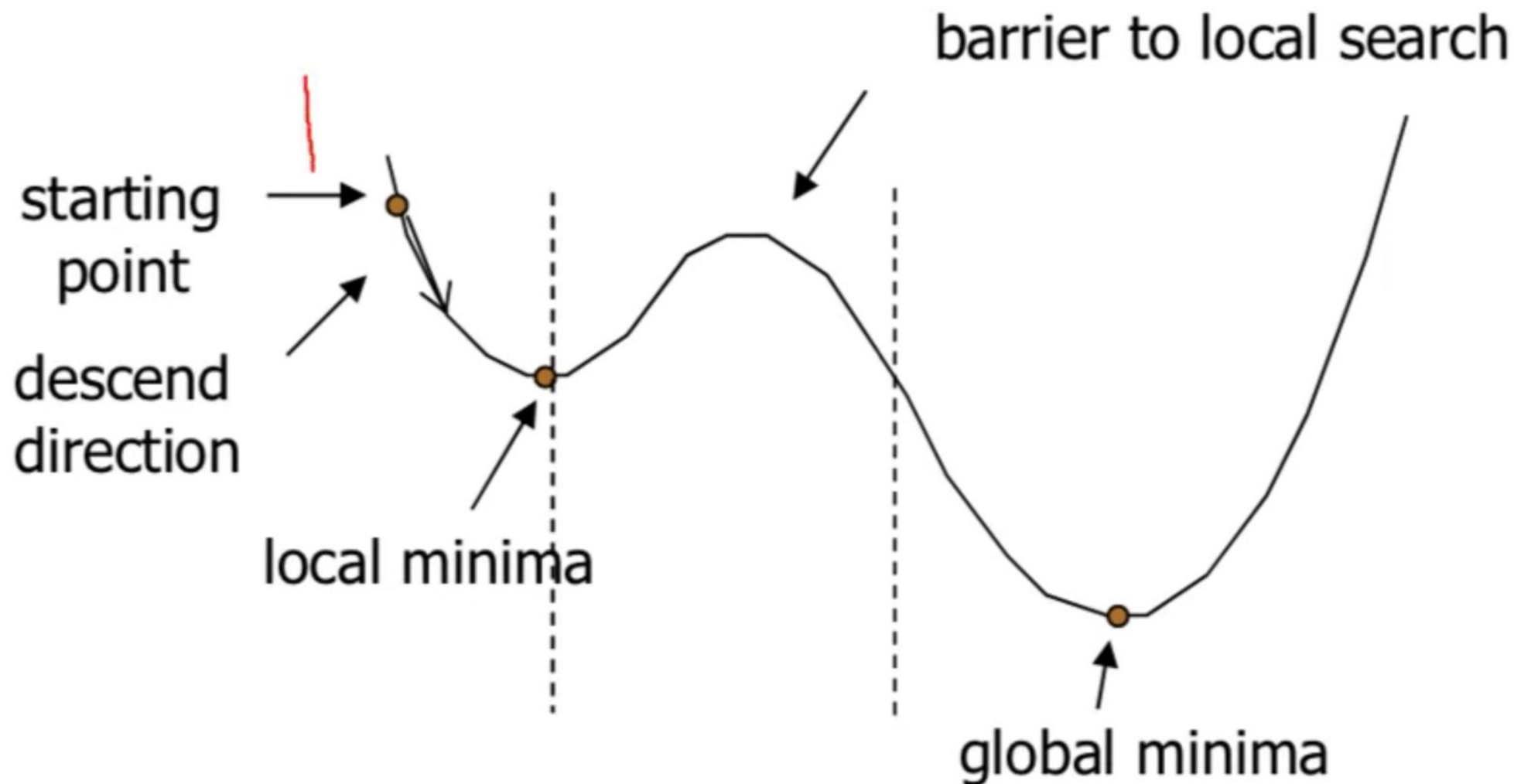
$$P(\Delta E) = e^{-\frac{\Delta E}{k*t}}$$

- Where ΔE - Energy Magnitude
- t - temperature
- k - Boltzmann constant.

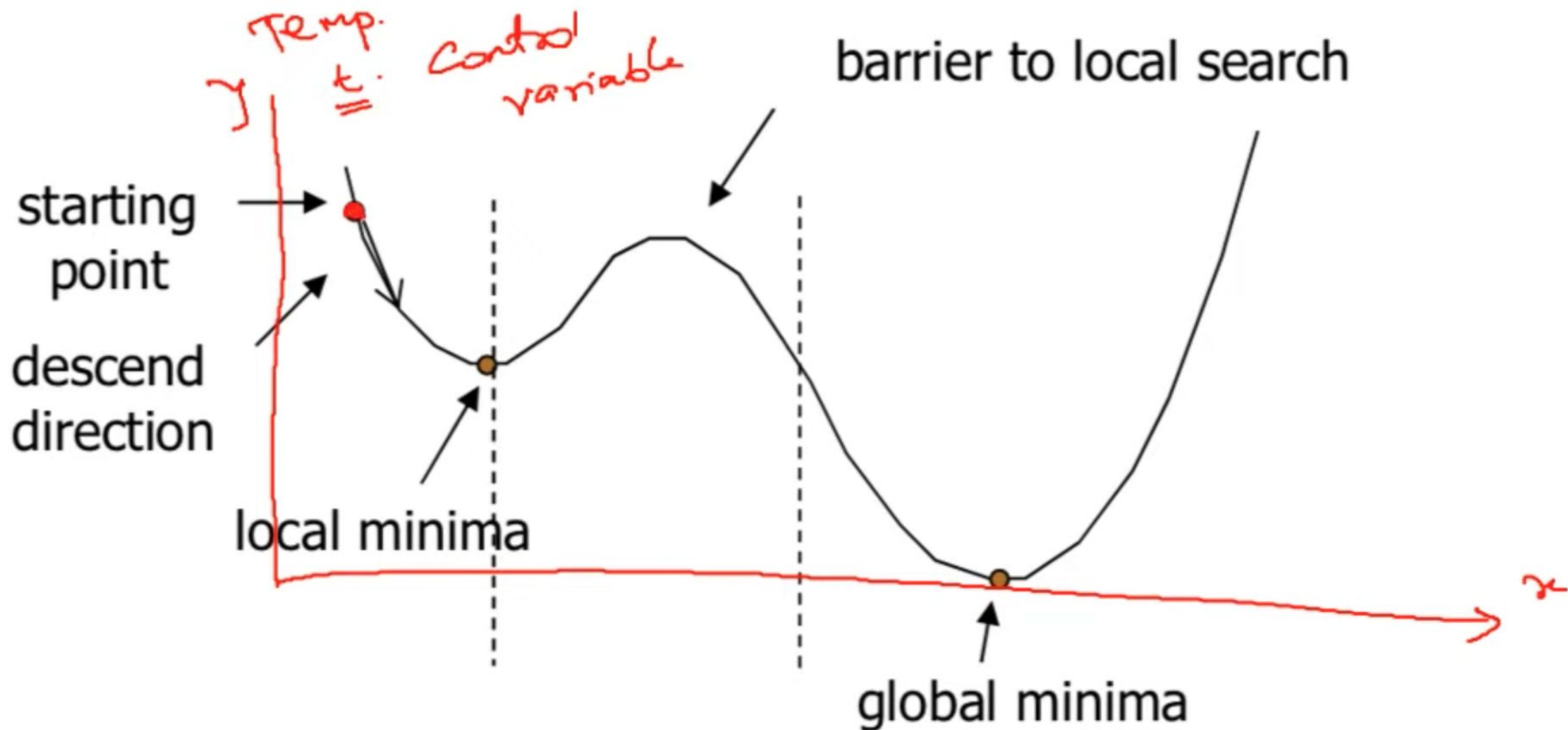
Simulated Annealing

- **Simulated Annealing** is a **stochastic global search optimization algorithm** and it is modified version of stochastic hill climbing.
- This algorithm appropriate for **nonlinear objective functions** where other local search algorithms do not operate well.
- **The simulated-annealing solution is to start by shaking hard (i.e., at a high temperature) and**
- **then gradually reduce the intensity of the shaking (i.e., lower the temperature).**
- Simulated Annealing (SA) is very useful for situations where there are **a lot of local minima.**

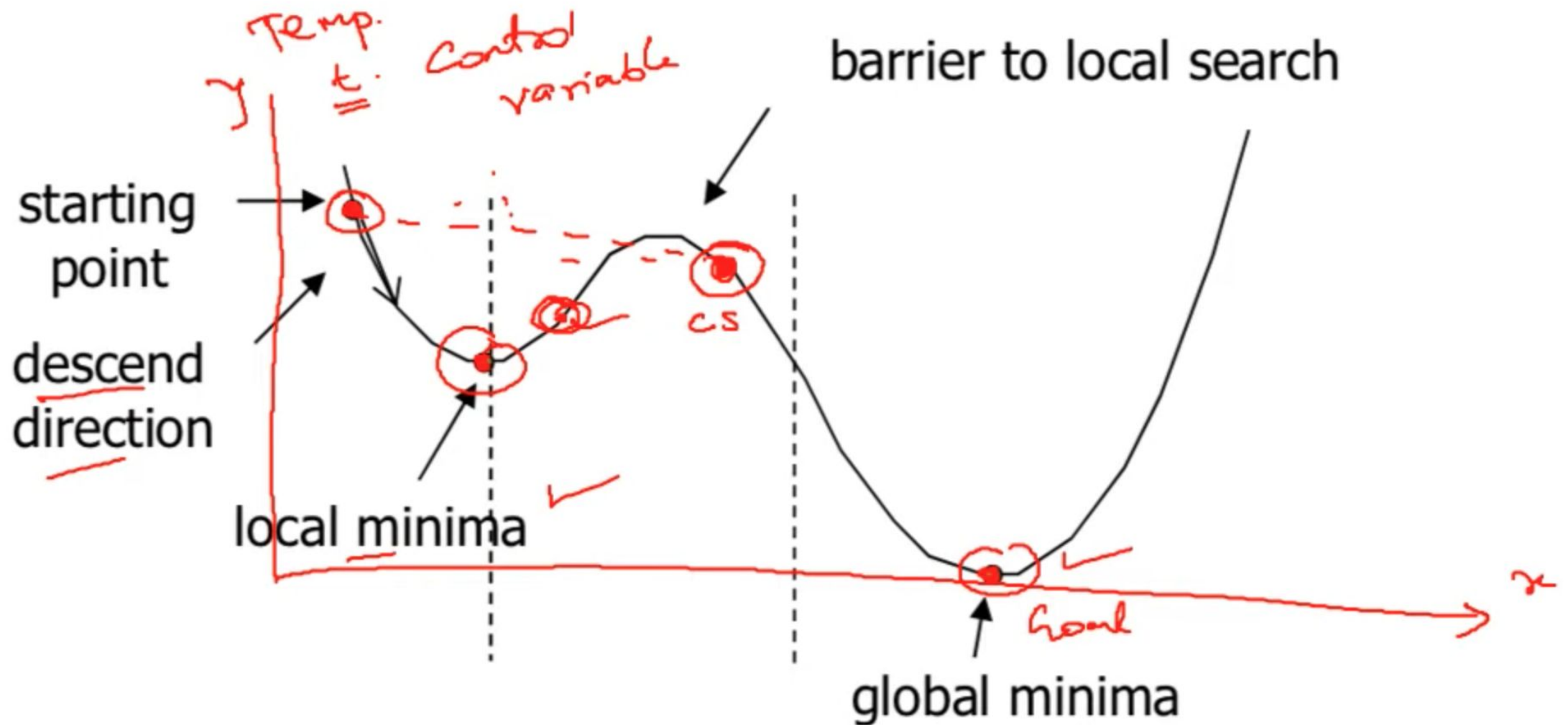
Simulated Annealing - State Space Diagram



Simulated Annealing - State Space Diagram

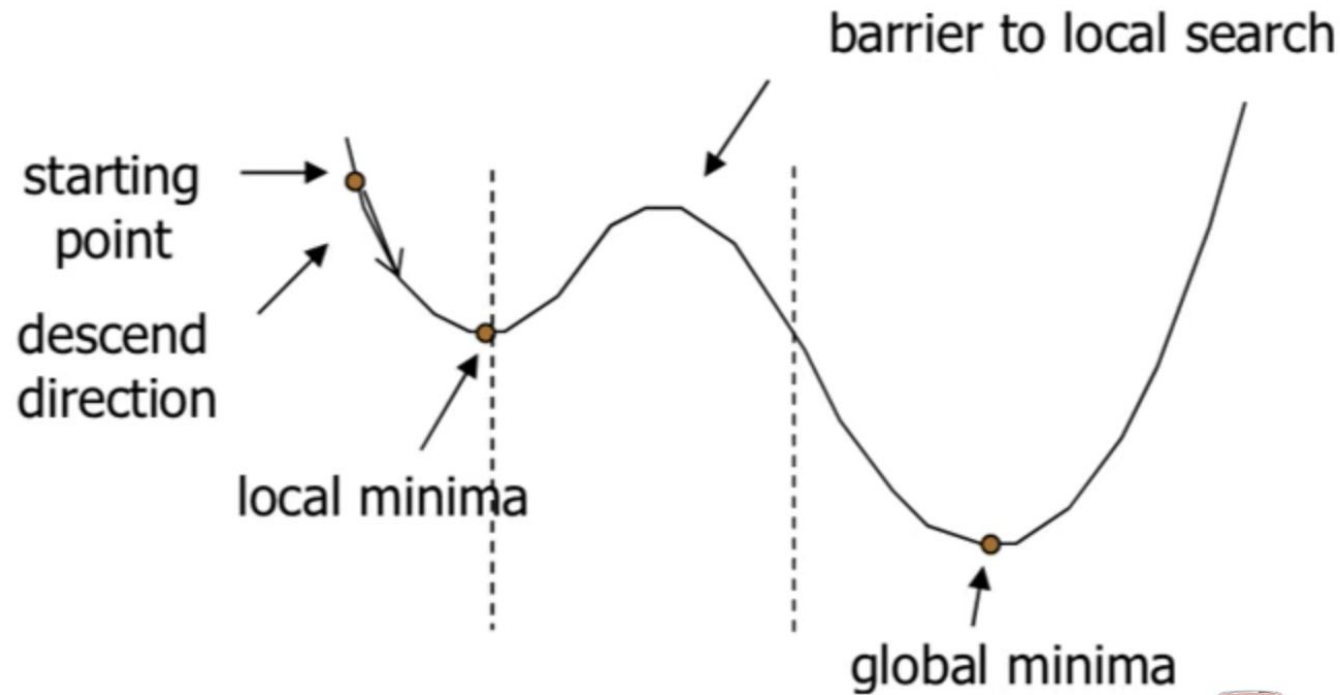


Simulated Annealing - State Space Diagram



Simulated Annealing- Example- ping-pong ball

- Imagine the task of getting a ping-pong ball into the deepest crevice in a very bumpy surface.
- If we just let the ball roll, it will come to rest at a local minimum.
- The trick is to shake just hard enough, to bounce the ball out of local minima then the ball will reach the global minimum.



Simulated Annealing Algorithm

function SIMULATED-ANNEALING(*problem*, *schedule*) **returns** a solution state

current \leftarrow *problem*.INITIAL

for $t = 1$ **to** ∞ **do**

$T \leftarrow$ *schedule*(t)

if $T = 0$ **then return** *current*

next \leftarrow a randomly selected successor of *current*

$\Delta E \leftarrow$ VALUE(*current*) $-$ VALUE(*next*)

if $\Delta E > 0$ **then** *current* \leftarrow *next*

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

Simulated annealing...

- Simulated annealing was used to solve VLSI layout problems
- It has been applied widely to factory scheduling and
- other large-scale optimization tasks.