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## "CHAPTER # 01"

"Artificial Intelligence is the science & engineering of making intelligent machines, especially intelligent computer programs."

"Intelligence is the computational part of the ability to achieve goals in the world"

"AI is the study of agents that receive precepts from the environment & perform actions"

→ Rationality is

- ① Maximally achieving pre-defined goals
- ② Goals are expressed in terms of the utility of outcomes

• Turing Test

→ A computer passes the test if a human interrogator, after posing some written questions, cannot tell whether the written responses come from a person or from a computer

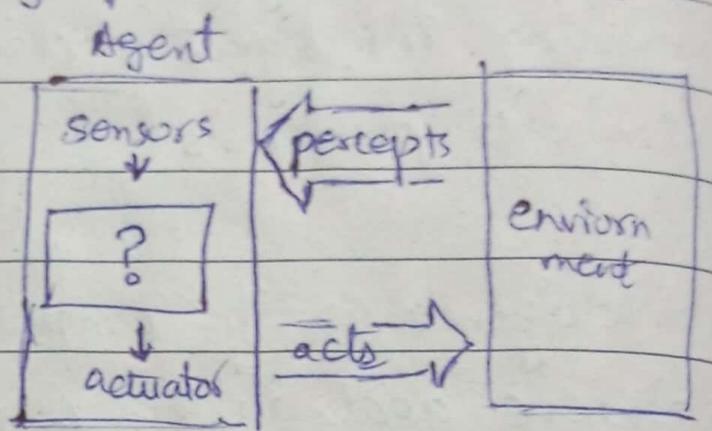
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## o Agents

An agent is anything that can be viewed as perceiving the env through sensors and acting upon that env through actuators.

- percept refers to the content an agent's sensor are perceiving



- percept sequence is the complete history of everything the agent has ever perceived
- An agent's behaviour is described by the "agent function" that maps any given percept sequence to action

- A rational agent is one that does the right thing
- Doing actions in order to modify future percepts - called information gathering
- An agent is autonomous if its behaviour is determined by its own experience (with ability to learn & adapt)

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## • PEAS

### • Performance Measures:

↳ Parameters or metrics on which the performance of the agent would be analyzed

### • Environment:

↳ An env. the agent will be working on.

### • Actuators:

↳ Device responsible for performing actions

### • Sensors:

↳ Device used to perceive.

## Interactive English Tutor Agent:

► P : Student's score on test

► E : Set of students, testing agency

► A : Display of exercises, feedback, speech

► S : Keyboard entry, voice

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## ▷ ENVIRONMENT TYPES

### ① Fully Observable Vs Partially Observable

- ↳ In fully observable the agent has complete & accurate info about the env at any given time. Like in Chess, where entire board state is available
- ↳ In Partially observable the agent has limited or incomplete info about the env due to missing data, noise or constraints. Like in Cards where player cannot see other's cards.
- ↳ If agent has no sensors at all then env is unobservable. Like robot navigating in darkness

### ② Deterministic Vs Stochastic (non deterministic)

- ↳ In deterministic the next state of the env is completely determined by the current state and the agent's action. Eg: Chess
- ↳ Non deterministic the next state has some degree of randomness meaning the same action can lead to diff outcomes. Eg: Rolling a dice

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### ③ Episodic Vs Sequential (non-episodic)

↳ An agent's experience is divided into atomic episodes.

↳ The agent's action in one episode does not affect the future episodes. Each decision is independent of past actions. Eg: Image classification

↳ In sequential the current actions influence future states and rewards. Eg: Chess, Autonomous cars.

### ④ Static Vs Dynamic

↳ In static the env does not change while the agent is deciding on an action. Eg: Chess

↳ In dynamic the env changes over time, even if agent does nothing, requiring real-time decision making.

Eg: Self driving car

### ⑤ Discrete Vs Continuous:

↳ In discrete there are limited no. of distinct, clearly defined percepts & actions. Eg: Chess

↳ In continuous the env has an infinite no. of possible states & actions. Eg: A robot moving in 3D space

Known & The agent knows the rules of the env, like in game of chess.  
Unknowns The agent doesn't fully understand rules of env  
and must learn

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## ① Single Agent Vs Multi Agent

↳ In single agent only one agent interacts with env.

Eg: Sudoku Solver

↳ In Multi Agent, agent interact with multiple agents.

Eg: Chess

## △ TYPES OF AGENTS

### ① Simple Reflex Agents:

- Acts based on current percept only (ignoring history)
- Use if-then else statements
- Eg: A thermostat turning on/off based on temp.

### ② Model Based Reflex Agent:

- Maintain an internal model of env

↳ it is the knowledge about how things happen in the world

- Consider past states to make better decisions.
- Eg: Self Driving car

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#### ④ Goal Based Agents :

- Optimize actions based Act to achieve specific goal rather than just reacting
- Involves consideration of the previous & future
- Requires planning and decision making
- Eg: A GPS System finding the best route to dest

#### ④ Utility Based Agent:

- Optimize actions based on a utility function
- Utility func to decide which state is better for agent
- They find best way to achieve the goal.
- The utility-based agent is useful when there are multiple possible alternatives, and an agent has to choose in order to perform the best action
- Eg: A stock trading AI maximizing profit while minimizing risk

#### ⑤ Learning Agents :

- Improve their act performance over time by learning from experience
- Use ML techniques to adapt

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- Four conceptual components
  - ① Learning element ② Critic ③ Performance Measure
  - ④ Problem generator
- Eg: Recommendation System

## ► Problem Formulation :

→ Problem Spaces States Space

- ① State : Set of all possible states where you can be
- ② Initial State : The state from where search begins
- ③ Actions : All the possible actions agent can perform.
- ④ Transition Model : Which describes what each action does.

RESULT(s, a) returns the state that results from doing action(a) on state(s)

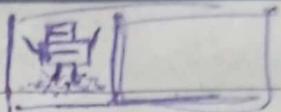
⑤ Action Cost Function : ACTION-Cost(s, a, s') gives numeric cost of applying a on s to reach s'

- A sequence of action forms a path
- A solution is a path from the initial → goal state
- An optimal solution has the lowest path cost among all solutions

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Vaccum World:



- ① States: ~~Two~~ cells
- ② Initial States: Any state can be designated as initial state
- ③ Actions: Suck, moveLeft, moveRight
- ④ Transition Model: Suck removes dirt. Move right & left change the direction by  $90^\circ$
- ⑤ Goal States: The states in which every cell is clean
- ⑥ Action Cost: Each action costs 1

depth of least cost solution

## "SEARCH STRATEGIES" (a) Time Complexity (b)

blind search

Uninformed Search

- Searcher without any prior knowledge (such as closeness or location of goal)

- Time consuming as it explores all possible paths

- DFS, BFS, Uniform Cost Search

- No. of steps, path cost unknown

- agent knows when it reaches a goal

- Operates in a brute force manner

heuristic search

small branching factor of search tree

Informed Search

- Use knowledge such as heuristics, to guide the search

- Less time consuming

- A\*, Heuristic DFS, Heur-BFS, Best-First Search, Hill climbing

- Heuristic is a function that estimate costs or distance from current state to goal state.

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### ► Breadth First Search :

Completeness : Yes , a solution will be found if exists

Time Complexity :  $O(b^{d+1})$

Optimal : Yes.

→ Suppose the branching factor  $b=3$  , and the goal is at depth  $d=20$  then , time to finish =  $3^{20}$

→ Not suitable for large graphs

### ► Depth First Search :

Completeness : Not complete if search space is infinite or contains cycles. Complete in finite space

Optimality : Not guarantee finding the optimal sol .

Time Complexity :  $O(b^d)$

Space Complexity :  $O(b^d)$

### ► Uniform Cost Search : (Cheapest First Search)

↳ Priority working BFS ki hai bas sirf us node ko expand krenge jiski least cost hogi .

↳ Priority queue use hogi

↳ each node is associated with its cumulative cost

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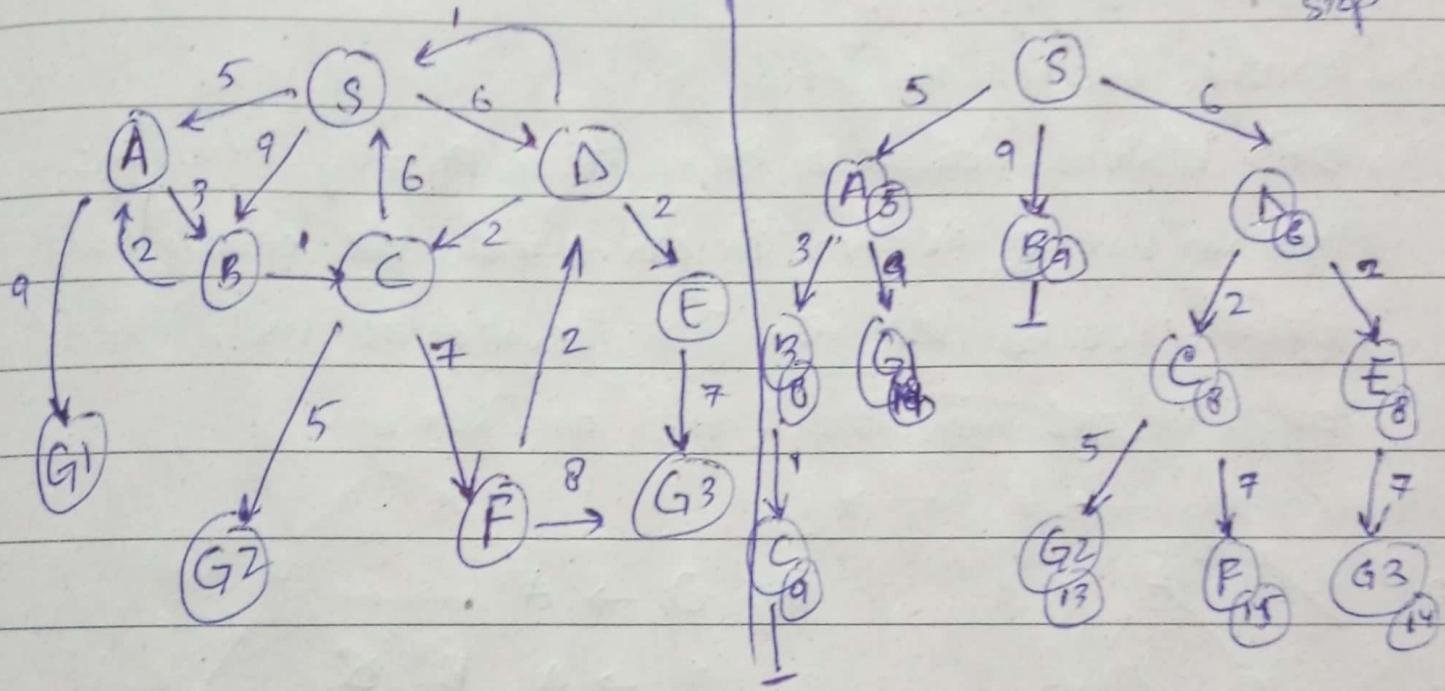
Completeness: Yes

Time Complexity: Larger than  $b^d$  (Space Complexity)

Optimality: Always optimal

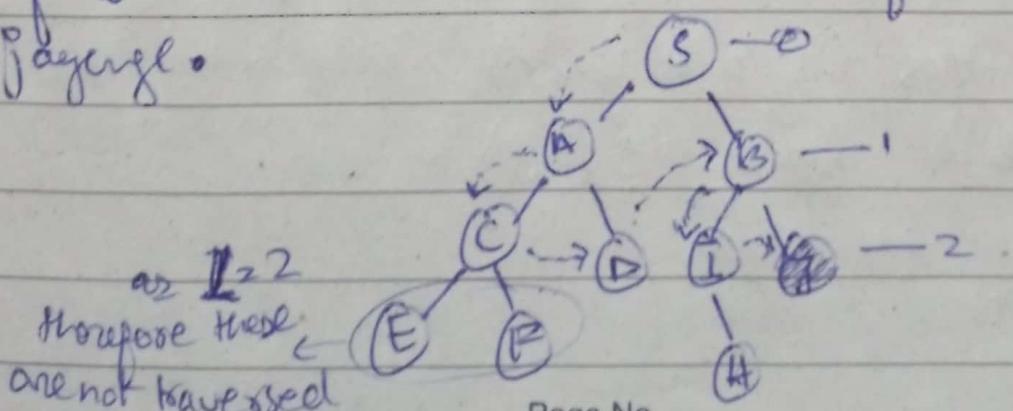
Visited: S, A, D, B, C, E, G<sub>2</sub>

goal so stop



#### D Depth-Limited Search :

→ DPS ki toha hota hai bs is mein ek depth limit defined hota hai. Us depth limit se neeché nahi jayenge like if  $D=2$  then hum root node se 2nd level tak jayenge.



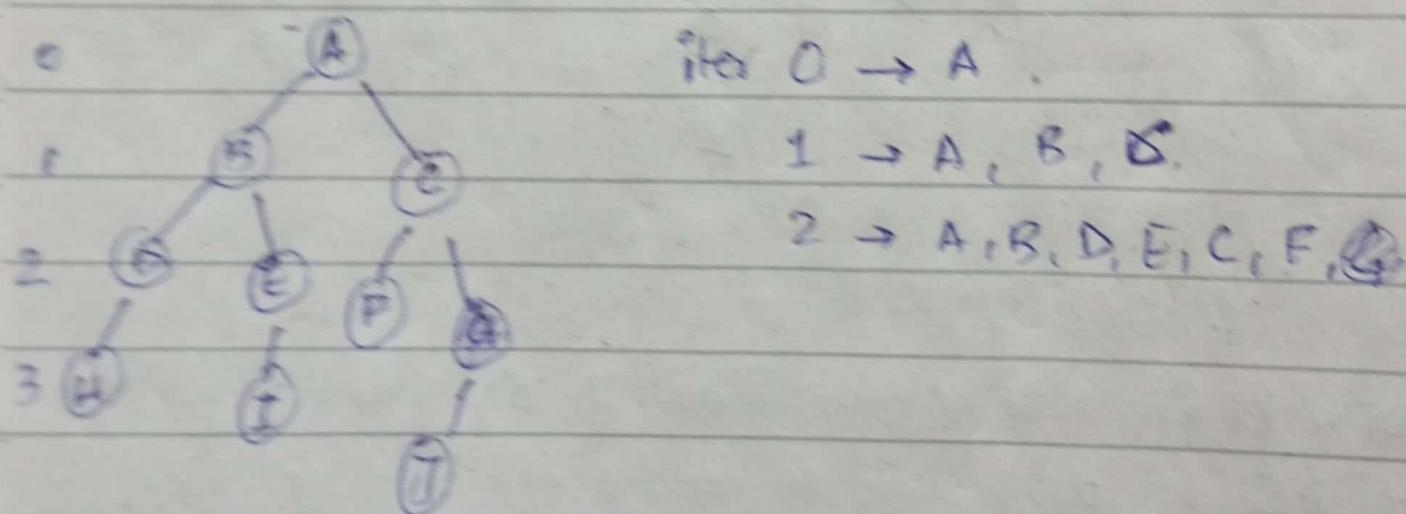
Completeness: No (if goal is beyond  $b^d$ )

Time:  $O(b^d)$  (Space  $\approx O(bd)$ )

Optimal: No (if  $b \leq d$ )

### ▷ Iterative Deepening DFS :

↪ IDFS is the one that has to make multiple iterations to perform search as for iteration  $d$  it explores all nodes at level  $d$ . For iteration  $d+1$  it explores all nodes at level  $d+1$ , then at  $d+2$  and so on.



Completeness: Yes (if  $b$  is finite)

Time:  $O(b^d)$  Space:  $O(bd)$

Optimal: Yes



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## o INFORMED SEARCH ALGO:

→ A heuristic is a function that estimates how close a state is to a goal.

Ex: Manhattan or Euclidean distance for pathing.

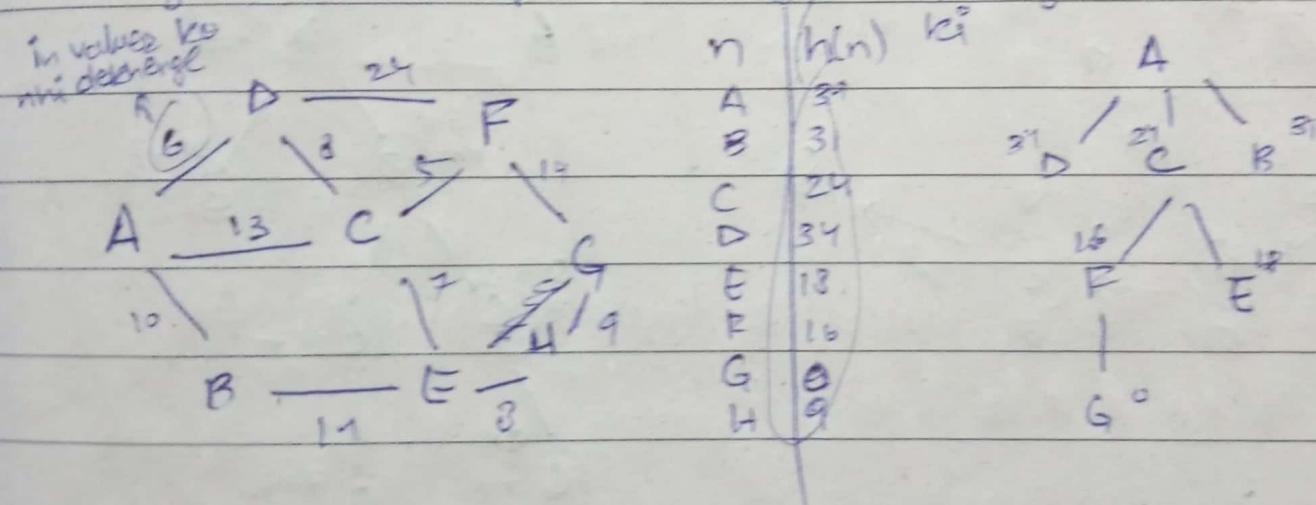
→ Optimality depends upon Heuristic.

## ▷ Best First Search:

↳ Ismein har node ki ek heuristic value di hoga.

Ye ek estimated value hoga. Jis node ki subse kam

hoga use ko explore korange. → ye is estimated values ka n se good node tak path chne



→ Does not give optimal sol.

→ Not complete (May stuck in infinite loop)

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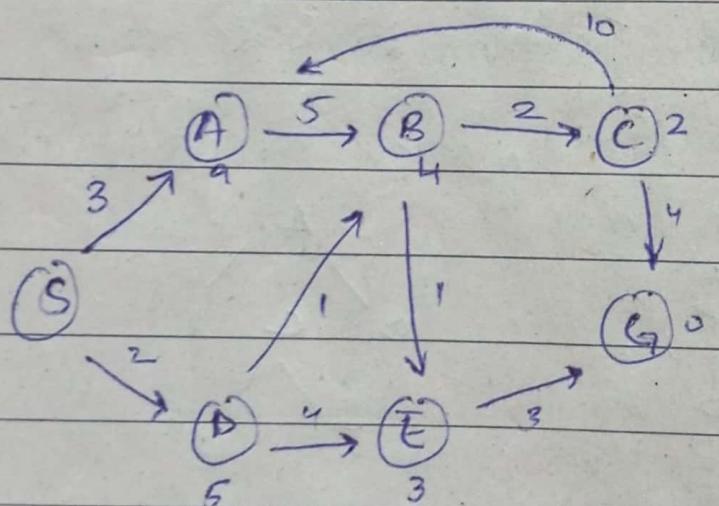
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## A\* Search

It is first  $h(n)$  ko nahi dekhte blke actual cost ko bhi consider karte hain.

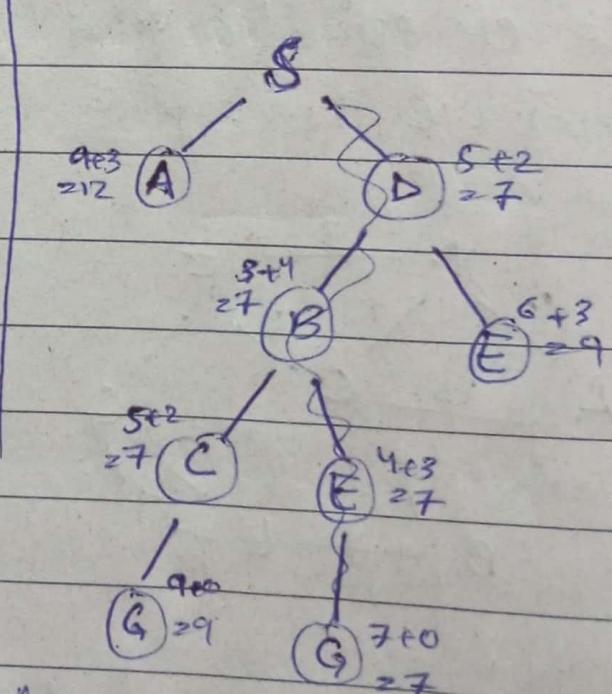
$$f(n) = \underbrace{g(n)}_{\substack{\text{Actual cost from} \\ S \rightarrow n}} + \underbrace{h(n)}_{\text{estimation cost from } n \rightarrow G}$$

Firka  $f(n)$  Ram hoga wko choose heng



FRS

Finds optimal solution



Agr do nodes ki same value aati h to dono ko explore kرنge

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- Admissibility For Heuristics

- $h(n)$  must be admissible

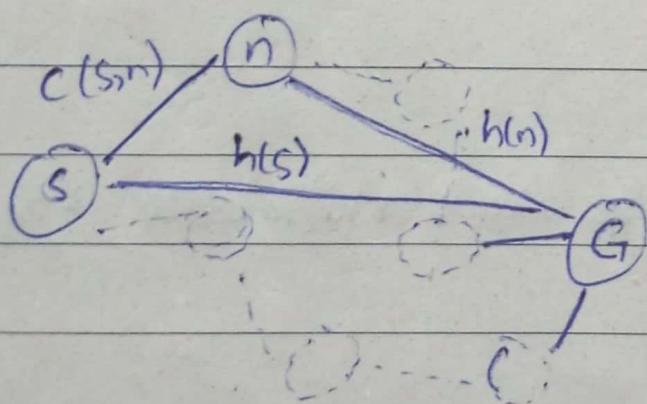
- An admissible heuristic is one that never overestimates the cost to reach the goal.

- Admissible heuristic are by nature optimistic bcz they think the cost of solving problem is less than it actually is.

- Straight line dis is admissible.

- Consistency For heuristic

$$h(s) \leq c(s,n) + h(n)$$



$$h(s) \leq c(s,n) + h(n)$$

It is like triangle inequality

- ▷ Weighted A\* Search

$$f(n) = g(n) + W \times h(n)$$

- ⇒ doesn't gives optimal

- ⇒ but gives quickly

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► Hill Climbing : (Local Search, Greedy Approach, No backtrace)

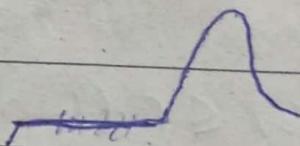
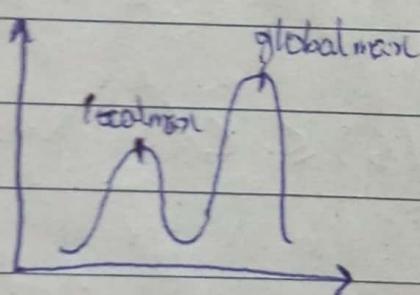
↳ This algo continuously moves in the direction of incr. elevation / value to find the peak or best sol to the problem.

↳ It terminates when it reaches the peak value where no neighbours has a higher value.

Problems :

① Local Maximum

② Plateau / Flat Maximum



↳ ye algo saari states ka track nhi rakhta & current state ko dekhta hai

- It is not guaranteed to find the optimal soln.
- It is highly sensitive to the initial state
- It doesn't maintain a search history
- Req less computation

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- We can perform multiple runs at random initial state to get best possible results.
- Stochastic Hill Climbing: chooses at random from among the uphill moves

### ▷ Local Beam Search :

↳ Take care of space complexity (constant)

↳ Beam width is given ( $B$ )  $\Rightarrow$

↳ Soozi states ko save kiske rakhta, jitni beam value hogi utni hi states save kiske rakhega.

### ▷ Simulated Annealing :

function SIMULATED ANNEALING return a sol state

input: problem, schedule (a mapping from time to temperature)

local variables: current, next,  $T$  (a "temperature" controlling probability of downward steps)

current  $\leftarrow$  MAKE-NODE ("INITIAL-STATE [problem]")

for  $t \leftarrow 1$  to  $\infty$  do:  $\rightarrow T=0$  means that I am frozen, i.e. frozen means i cannot make any other moves, so return current state

$T \leftarrow$  schedule [ $t$ ]

if  $T > 0$  then set new current

next  $\leftarrow$  a randomly selected successor of current

$\Delta E \leftarrow$  VALUE [next] - VALUE [current]

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Raju

Temp is too cold  
 $E^{-\frac{\Delta E}{kT}} = e^{-\infty} \rightarrow 0$   
Non prob  $\geq 0$  so it  
will never make a move

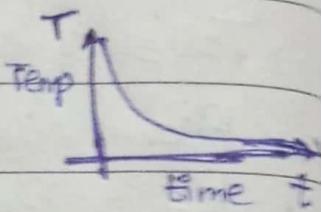
$\Delta E < 0$   $\rightarrow$  both term  
-ve value negl. Tab  
opp.  $\Delta E > 0$   $\rightarrow$  pos.  $\Delta E$   $\rightarrow$  pos.  $\Delta E$   $\rightarrow$  pos.

If  $\Delta E > 0$  then current  $\leftarrow$  next  
else current  $\leftarrow$  next only with prob  $e^{\Delta E / kT}$

→ It's also rein hum bad moves bhi lesakte hain (like we can also go down the hill)

→ Lekin jab T ki value boht high hogi tab hi bad more lena ki probab zyada hogi.

→ T ki value gradually ~~↓~~ low hoti jayegi.



high T: prob of "locally bad" move is higher

low T: u u . u & + lower

## ▷ Genetic Algorithm

- ① Initialize the population (solution states) randomly or with potential good solutions
- ② Compute fitness of each ~~set~~ population  $\rightarrow$  initial fitness value
- ③ Select parents using a selection procedure
- ④ Create offspring by crossovers & mutation operators
- ⑤ Compute the fitness of the new offspring
- ⑥ Select members of population to die using a selection process
- ⑦ Go to step 2 until termination criteria are met.

6+5+3+3+3+2

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→ Hamare pras mukhtaliif diff sol<sup>n</sup> hui or unmain se best sol<sup>n</sup> extract karna hui

• Types of Crossover → process of combining two chromosomes (strings) to produce new offspring

① Single Point Crossover:

at string 1 → 11011; 00100 110110

string 2 → 11010; 11000011110

offspring 1 → 11011; 11000011110

offspring 2 → 11010; 00100 110110

② Two Point crossover

↳ 2 points se break karke combine kardenge

③ Uniform Crossover:

↳ random bits ko include kardenge lek ek offspring bnega

Jindai zayeda fitness percentage  
Logi unke choose karte  $24/78 = 0.307$

$$0.307 \times 100 = 30.7\%$$

Chance of being chosen

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24748552

24  
31%

32752411

24752411

32748552 → 32748152

32752411

23  
29%

24748552

24752411

24752411 → 24752411

24415124

20  
28%

32752411

32752124

32752124 → 32752124

32543213

11  
14%

24415124

24415411

24415411 → 24415417

• initial population  
function

• selection

• crossover

• mutation

SEND

4 3 8 2

MORE

1 0 5 3

MONEY

1 0 4 3 5