

Artificial Intelligence

- Chapter 1 : Intelligent Agents - Introduction & problem solving
- Chapter 2 : Search Methods
- Chapter 3 : propositional logic & knowledge representation
Semantics, Inference rules
- Chapter 4 : Acting logically :-
Planning - Partial order, scheduling, types of planning
- Chapter 5 : Searching techniques - First order logic - Forward reasoning - Backward reasoning.

Artificial Intelligence :-

AI Comprises of two words :-

- (i) Artificial :- something made by human/non natural thing
- (ii) Intelligence :- ability to understand or think.

- Studies shows how to achieve intelligent behavior through computational means

- This makes AI a branch of Computer Science

John McCarthy, Coined the term in 1956 and defines it as "the science and engineering of making intelligent machines, especially intelligent computer programs".

The definitions of AI are categorized into 4 approaches.

- (i) systems that think like human :- Making m/c's to think as if machines are having sense in full and literal sense.
- (ii) systems that think rationally :- Having reason or understanding.
- (iii) systems that act like human :- Functions similar to human.
- (iv) systems that act rationally :-

In detail :-

- (i) Acting humanly :- Eg, The Turing Test approach.

The Turing Test was proposed by Alan Turing in 1950.

Defn: called as the Imitation Game. "Whether m/c's can think like human".
ask questions of two entities - man & computer
and receive answers from both.

If you can't tell which of the entities is human and which is a computer program, then you consider this computer to be intelligent.

Skills for the computer to pass, it need to possess -
full capabilities:

- (i) Natural language processing
- (ii) Knowledge representation :- To store what it knows or learns
- (iii) Automated reasoning :- To use the stored info. to answer q's

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Defn: called as the Imitation Game "Whether m/c's can think like human". Ask questions of two entities - man & computer and receive answers from both.

If you can't tell which of the entities is human and which is a computer program, then you are fooled and thus computer is considered to be intelligent.

For the computer to pass, it need to possess the foll. capabilities:

- (i) Natural Language Processing
- (ii) Knowledge representation :- To store what it knows or learns
- (iii) Automated reasoning :- To use the stored info. to answer q's, next move

IV) Machine Learning :- To adapt to new circumstances and to detect and extrapolate

To carry out the complete Turing Test, the computer will need:-

- (i) Computer vision → to perceive the objects
- (ii) Robotics → to manipulate objects and move.

Google AI has passed Turing Test

Computer (Player A)	Human Responder (Player B)
Interrogator Player C.	

that has attempted Turing Test:-

- 1) ELIZA → NLP Computer program, which can communicate with human.
- 2) PARRY → simulate a person with most common Chronic mental disorder.
- 3) Eugene Goostman - Convinced 29 % of judges → 13 year old boy

Thinking humanly: The Cognitive Modeling approach

- (i) through introspection :- Trying to capture one's thoughts
- (ii) through psychological experiments :- GPS → checked with memory slips by user

c) Thinking rationally :- This is Prescriptive
Ex) Right thinking

Ex) Socrates is a man
All men are mortal
∴ Socrates is mortal \Rightarrow logic.

d) Acting Rationally :- Doing the Right Thing

Ex) Computer Agents :-
- operating under autonomous control
- Perceiving their environment
- Persisting over a prolonged time period
- Adapting to change.

Foundations of AI:

The various disciplines which contributed ideas & techniques are :

i) Philosophy :-

- (i) How does mental mind arise from physical brain?
- (ii) Can formal rules be used to draw valid conclusions?
- (iii) Where does knowledge come from?
- (iv) How does knowledge lead to action?

Aristotle (384-322 B.C.) was the first to formulate a precise set of laws governing the rational part of mind. He developed philosophy (400 B.C.) made AI possible, by considering the ideas that make one to feel like mind is similar to brain.

Ex) Mathematics, Formal representation & flow of algorithms, computation & probability

Deals with

- (i) Formal rules to draw valid conclusions
- (ii) What can be computed ?
- (iii) How do you deal with uncertainty ?

3. Economics

- How do we make decisions so as to maximize ~~maximize profit~~ profit.
- How do you do this when the payoff may be far² far in the future.

4. Neuroscience

How do brains process info.

5. Psychology : ~~How do~~

How do humans and animals think and act

6. Computer Engineering :-

For AI to succeed intelligence and artificial ^{as} needed. AI programs tend to be large which requires ^{good} ^{mem} _{mem} and CPU's.

7. Control theory and cybernetics :

Self-controlling machines.

8. Linguistics : Language is a model. Modern linguistics & AI are called as computational linguistics or Natural language ~~process~~ processing.

History of AI:

Alan Turing articulated a complete vision of AI in 1950 by article "Computing Machinery and Intelligence". Then he introduced the Turing Test, Machine learning, algorithms, reinforcement algorithm. Aps was the computer program to implement "humanity" approach.

LISP was invented by John McCarthy in 1958 while he was at MIT. In 1963 McCarthy started the AI lab at Stanford.

expert system

Dendral was the project based on AI & it was released in 1960s. AI became an industry in 1980 till now.

1985 - 1995 : Neural N/w gained its popularity.

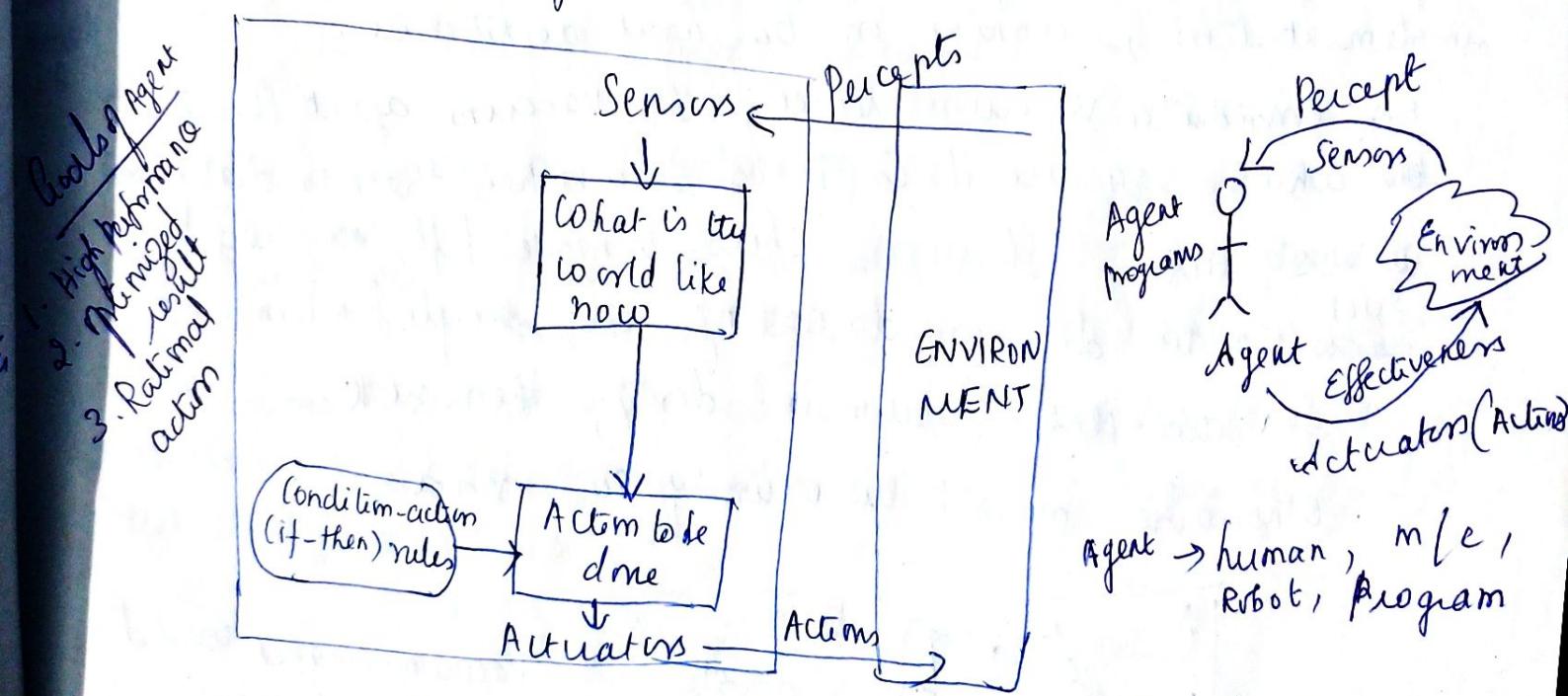
1988 + : Soft Computing.

In 2003 Human-level AI started emerging.

Due to the availability of Internet, high-speed processor and GPU's, AI is emerging with its charm in recent times.

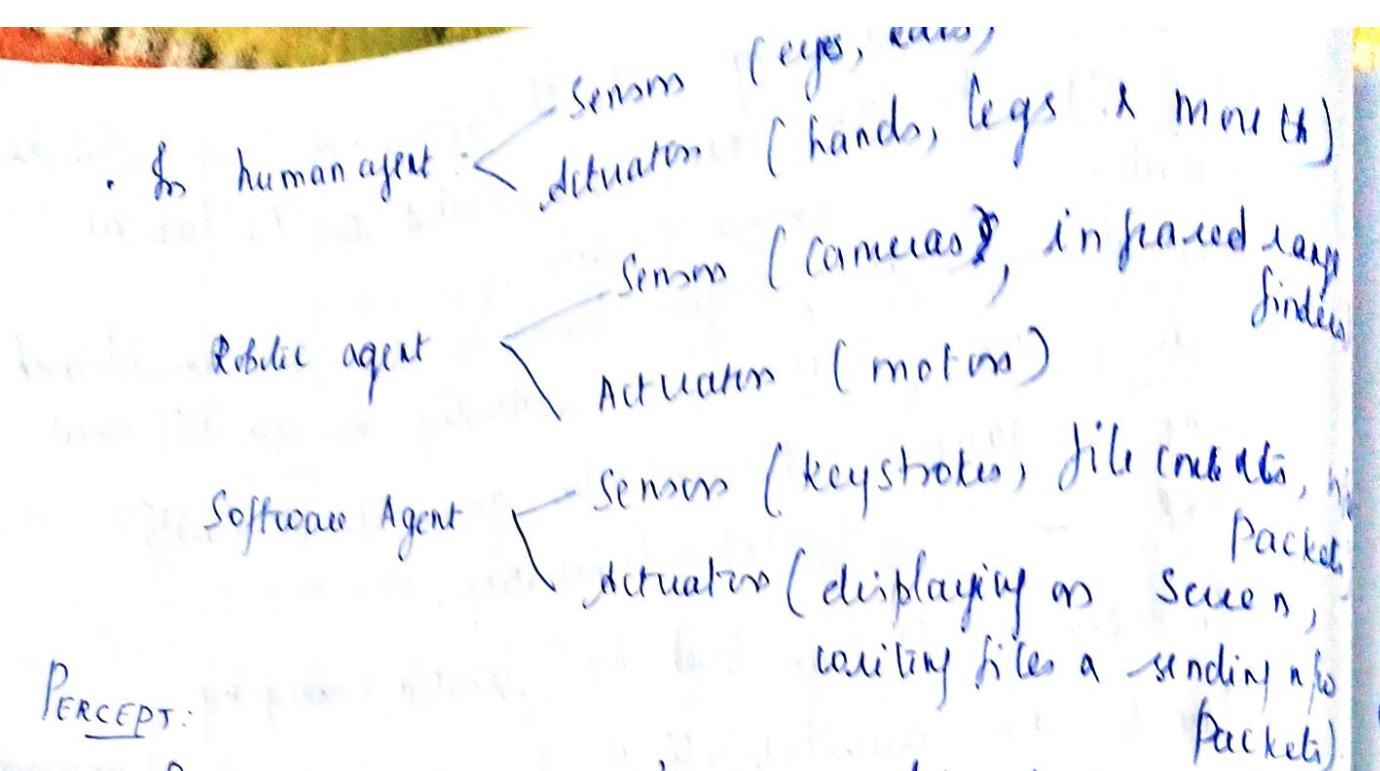
Intelligent Agent :

An agent is anything that can be viewed as perceiving its environment through sensors and sensors acting upon that environment through actuation.



Agent → human, m/e, Robot, Program

Agents interact with environment through Sensors & Actuators
Agent → Percepts → Decisions → Actions



Percept:

Percept refers to the agent's perceptual inputs at any given time.

Percept Sequence:

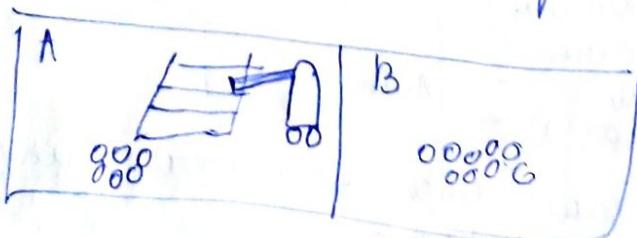
An agent's percept sequence is the complete history every percept the agent has ever perceived.

Agent function: It is a map from the percept sequence to an action.

Agent program: The agent functions for an artificial agent will be implemented by an agent program. The agent programs is a concrete implementation, running in the agent architecture.

Eg) Consider a vacuum cleaner. The vacuum agent perceives which square it is in and whether there is dirt in the square. It can choose to move left, move right, suck up the dirt or do nothing. One simple function is:

- If the agent function is dirty, then suck.
- Otherwise move to the other square



Vacuum cleaner world
with two locations.

Percept Sequence	Action
[A, clean]	Right
[A, Dirty]	suck
[B, clean]	Left
[B, dirty]	suck
[A, clean], [A, clean]	Right
[A, clean], [A, dirty]	suck

Agent function for Vacuum cleaner.

Concept of Rationality:

An agent should act as a Rational agent. A rational agent is one that does the right thing in order to make the environment successful.

Performance measures:

What is rational at any given time depends on 4 things:

- (i) The Performance measure that defines the criterion of success
- (ii) The agent's prior knowledge of the environment
- (iii) The actions that the agent can perform
- (iv) The agent's percept sequence to date.

Omniscience, learning and autonomy :-

Omniscient agent :- knows the actual outcome of its actions & can act accordingly; but impossible in reality.

A rational agent not only gather info., but also learn as much as possible. Successful agents split the task of computing agent function into 3 different periods:

- (i) when the agent is being designed.
- (ii) when it is deliberating in its next actions
- (iii) as it learns from experience.

Nature of the Environment
 specifying the task environment: (PEAS)

Performance, Environment, actuators, sensors
 Taxi Task environment

Agent type	Performance measure	Environment	Actuators	Sensors
Taxi Driver	Safe, fast, legal, comfortable trip, maximize profits	Roads, other traffic, Pedestrians, customers	Steering, accelerator, brake, signal, horn, display	Cameras, GPS, Speedometer, odometer, engine sensors, keyboards, Accelerometer

Properties of Task Environment:

(i) Fully observable / partially observable.

If an agent's sensors give it complete access to the environment at each point in time.

An environment may be partially observable because noisy, inaccurate sensors or because of parts of states are simply missing from the sensor data.

(ii) Deterministic vs non-deterministic (or) stochastic:

Deterministic: If the next state is completely determined by current state

(iii) Episodic Vs non-episodic (or) sequential:

Episodic \rightarrow Agent's experience is divided into atomic episode. Each episode consists of its perceptions and actions and it does not depend

In sequential, the current (could affect the previous episode)
 decision (Eg) chess/ taxi driving

iv) static vs dynamic:

Dynamic: If the environment is changing for agent's actions

Semi-dynamic: If the environment does not change & change after some time.

Eg) Taxi driving is dynamic

Chess is semi dynamic

Crossword puzzles are static

v) discrete vs continuous:

Discrete: The environment has limited no. of distinct, clearly defined Percepts & actions

Continuously: Eg) Chess
The environment changes continuously with range of values.

Eg) Taxi driving.

Structure of agent:

An intelligent agent is a combination of agent Program & Architecture.

Intelligent agent = Agent program + Architecture

Agent program → function that implements the agent mapping from Percepts to actions.

Architecture → Computing device used to run the agent program.

Types of Intelligent agents:

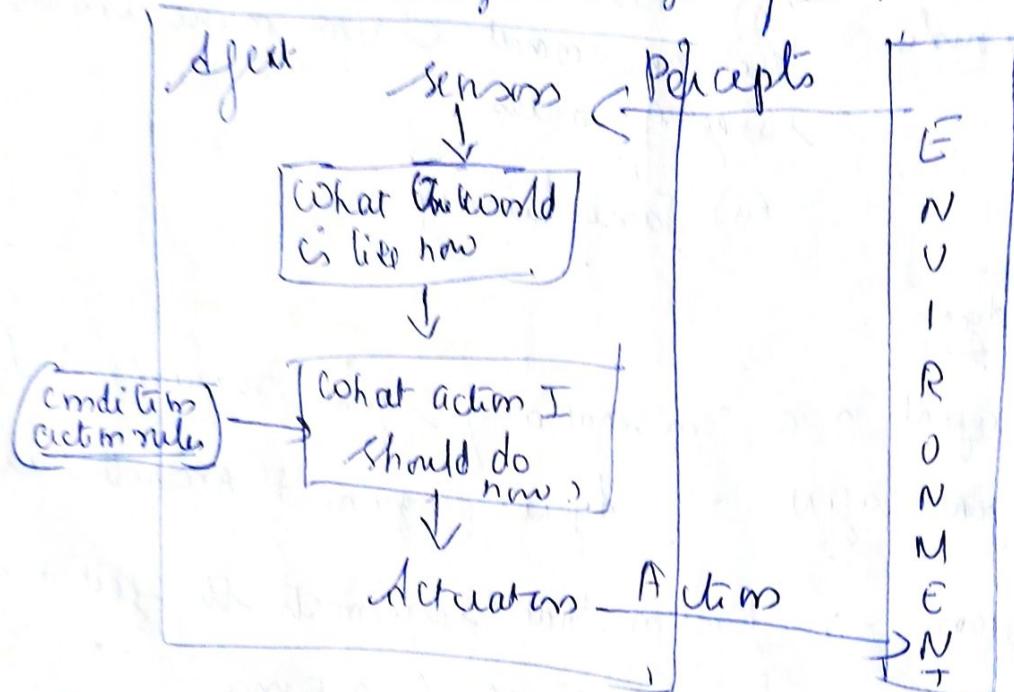
1. simple reflex agent
2. Model based reflex agents
3. Goal based reflex agents
4. Utility based reflex agents
5. Learning agents

i) simple reflex agent:

it responds directly to percepts
ii) more agent select actions on the basis of current percept, i.e.,
the rest of the percept history.

An agent decides about how the Condition action rules affect an agent to make the connection from percept to action.

Condition action rule: if condition then action
rectangle → current internal state of the agents decision process
oval → background info. of an process.



Eg) Agent intelligent program:

function SIMPLE-REFLEX-AGENT (Percept) returns an action

state : rules, a set of condition-action rules

state \leftarrow INTERPRET-INPUT (Percept)

rule \leftarrow RULE-MATCH (state, rules)

action \leftarrow RULE-ACTION [rule]

return action

- INTERPRET-INPUT → function generates an abstracted description of the current state from the percept
- RULE-MATCH → function returns the first rule in the set of rules that matches the given state description
- RULE-ACTION → the selected rule is executed as the action of the given percept.

The correct decision is made, on the basis of current percept i.e., if the environment is fully observable.

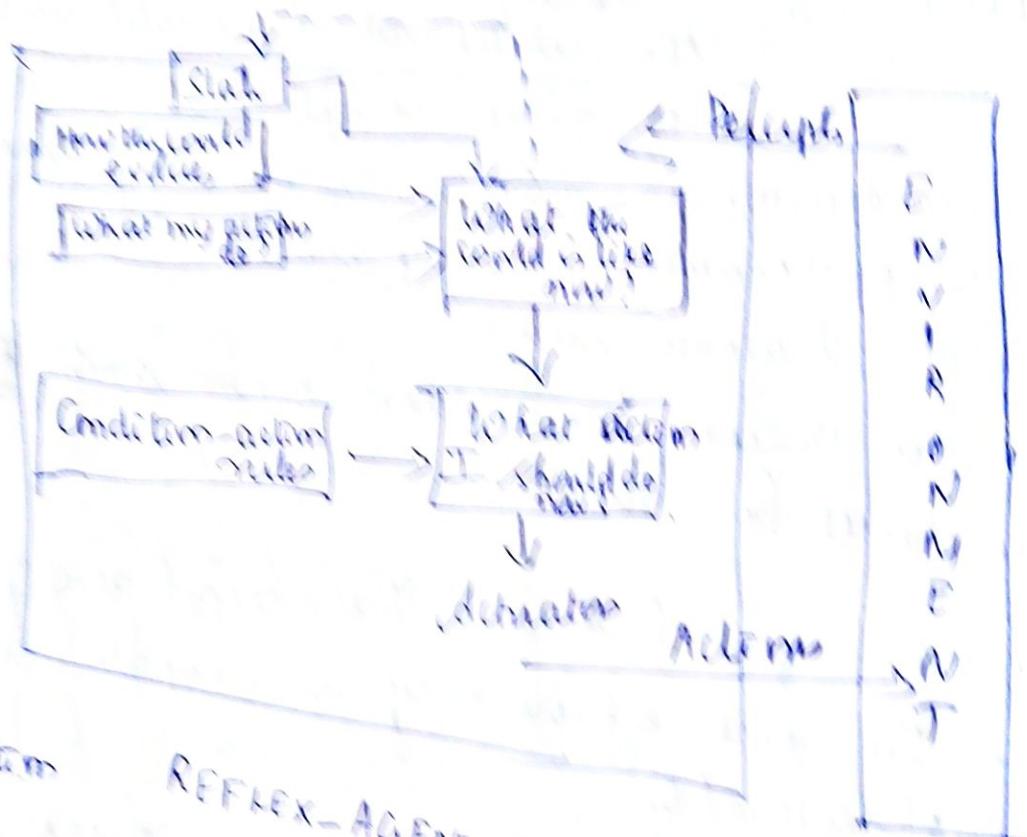
(e.g) Medical diagnosis system:

If the patient has reddish brown spots \Rightarrow start the treatment for measles.

2. Model based agents (Agents that keep track of the world)

- The most effective way is to handle partial observability
 - The agent which combines the current percept with the old internal state to generate updated description of the current state.
 - The current percept is combined with the old internal state and it derives a new current state which is then updated in the state description.
- Update requires 2 kinds of knowledge :
- (i) how the world evolves independently of the agent
 - (ii) Information regarding how agents affect the world

The above knowledge is implemented in simple Boolean circuits or in complex scientific programs called as model of the world. An agent that uses such a model is called as Model-based agent.

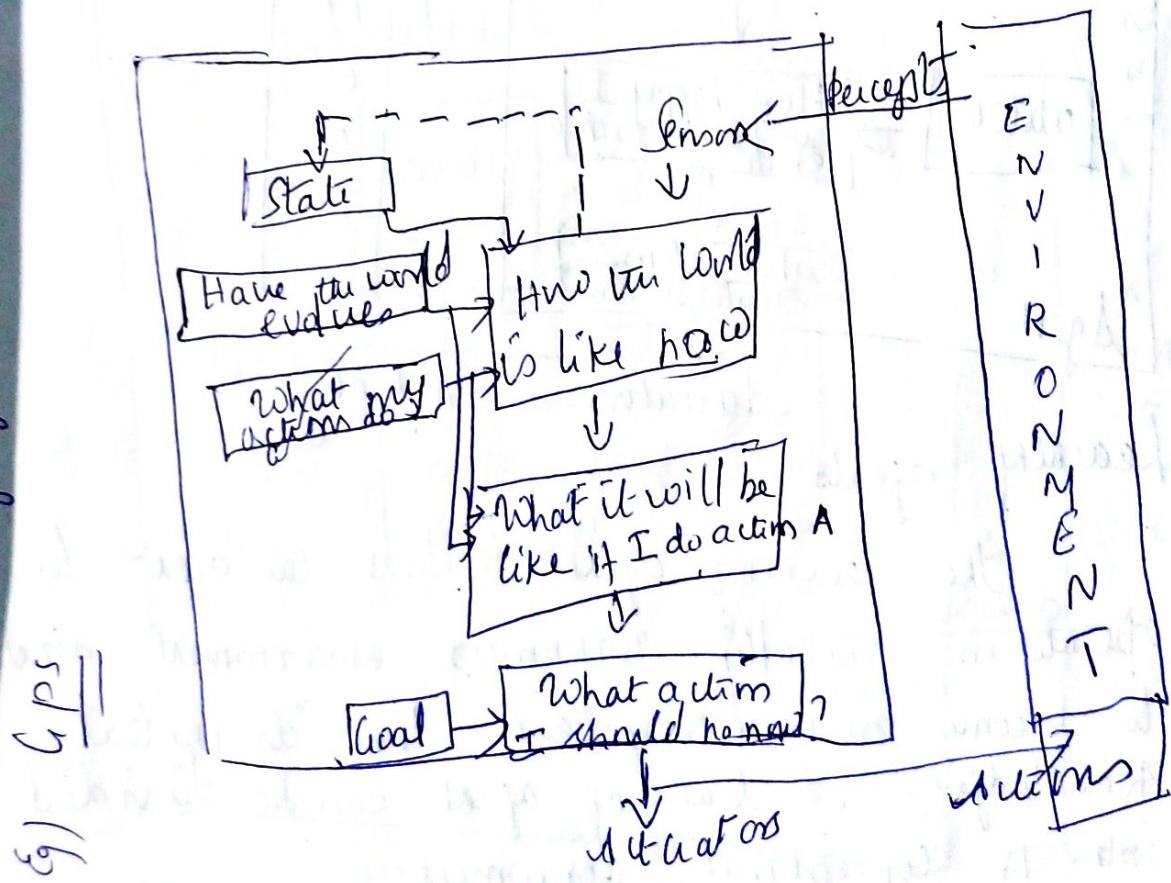


function STATE :
 state ;
 RULE - ACTION [rule]
 State ← UPDATE - STATE (state, action, percept)
 rule → RULE - ACTION [rule]
 action → RULE - ACTION [rule]
 return action

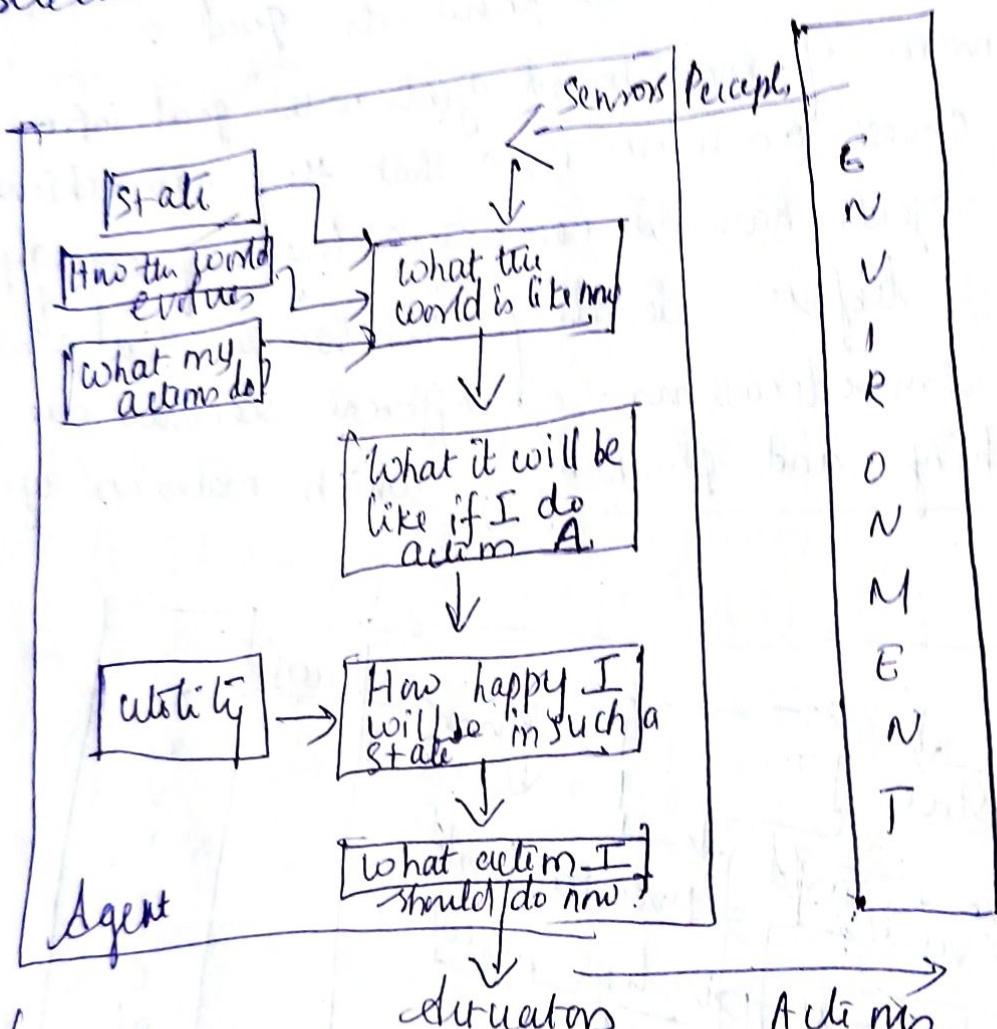
UPDATE - STATE
 state description responsible for creating new information by combining recent and earlier

3. Goal based agents :- (Planning)

- Expansion of model based reflex agents } e.g)
 - Desirable situations (goal)
 - Searching & planning
- The knowledge of the current state environment is not always sufficient to decide for an agent to what to do.
- ① The agent needs to know its goal.
- ② Expansion of model based agents with "goal information".
- ③ They choose an action, so that they can achieve the goal.
- ④ These agents have to consider a long sequence of possible actions before deciding whether the goal is achieved or not.
Such considerations of different scenario are called Searching and planning, which makes an agent proactive.



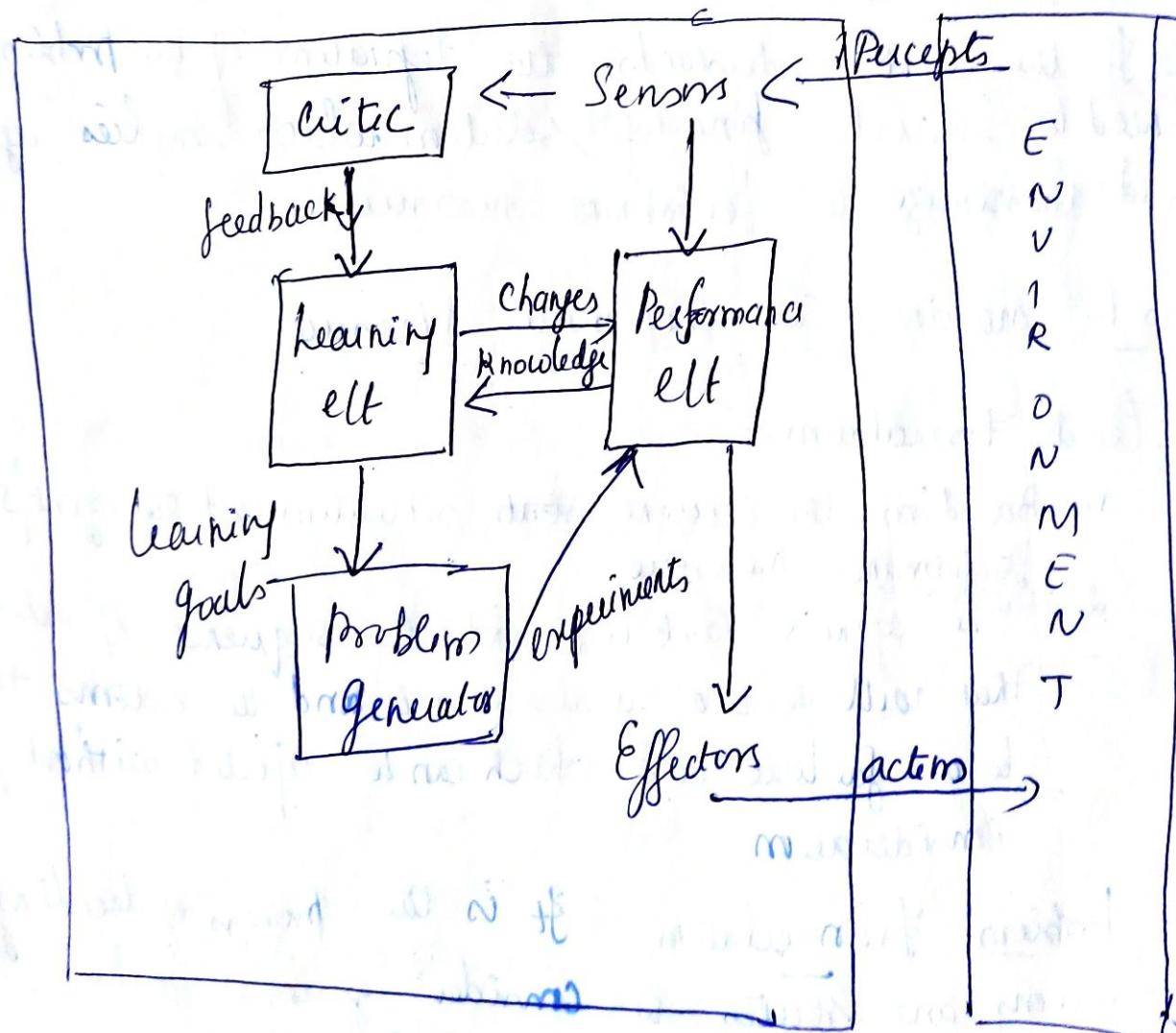
4) utility based agents
 An agent generates a goal state with high quality behaviour. If more than one sequence reaches the goal state then the sequence with more reliable safer and quicker and cheaper than others to be selected?



5- Learning agents:

The learning agent allows the agent to operate in initially unknown environment and to become more competent than its initial knowledge. A learning agent can be divided into 4 conceptual components.

- ① Learning elt: This is responsible for making improvements. It uses the feedback from the critic on how the agent is doing and determines how the performance elt should be modified to do better in future.
- ② Performance element: Responsible for selecting external actions and it is equivalent to agent. It takes in Percepts and decide on actions.
- ③ Critic :- It tells the learning element how well the agent is doing with respect to a fixed performance standard.
- ④ problem generator :- It is responsible for suggesting actions that will lead to new and informative experiences.



PROBLEM SOLVING AGENTS

an important aspect of intelligence is goal-based problem solving. Solution of many problems can be described by finding a sequence of actions that lead to a desirable goal.

Each action changes the state and the aim is to find a sequence of actions and states that lead from the initial (start) state to a final (goal) state.

Problem formulation:

Problem solving is one type of goal based agent, where there are sequences of actions, the agent should select one from these actions which lead to desirable states.

If the agent understands the definition of the problem, we need to search finding solution which implies agent should maximize the performance measures.

Steps to maximize the Performance Measure:

1. Goal Formulation:-

1. Based on the current ~~state~~ situation and the agent's performance measure
2. The agent's task is to find the sequence of actions that will go to a goal state and the actions that lead to a failure case which can be rejected without full consideration.

2. problem formulation: It is the process of deciding what actions and states to consider given a goal.

coach = An agent with several immediate options of

unknown value can examine different possible sequences of actions that leads to the states of known value and then chooses the best sequence.

The process of looking for sequences actions from the current state to reach the goal state is called search.

4. Solution: The search algs takes a problem as input & returns a solution in the form of action sequence.

5. Execution: Once a solution is found, the execution phase consists of carrying out the recommended action.

function SIMPLE-PROBLEM-SOLVING-AGENT (*percept*) returns an action

inputs: *percept*, a percept

state: *seq*, an action sequence, initially empty
state, some description of the current world state
goal, a goal initially null
problem, a problem formulation
state UPDATE-STATE (state, percept)

If *seq* is empty then do

goal FORMULATE-GOAL (state)

problem FORMULATE-PROBLEM (state, goal)

seq SEARCH (problem)

action FIRST(*seq*)

seq REST (*seq*)

return action *E*

State

Observable

Discrete

Deterministic

is called an agent carries out its plan with eye closed.

Well defined problems & solutions.

1) The initial state

2) successor function returns possible actions available

3) Goal test

Determines whether the given state is goal state
eg Goal is to reach a state named as "checkmate".

4) Path cost:

Assigns

a) Step cost: numeric cost to each action.

b) A solution to the problem is the path from the initial state to a goal state.

c) An optimal solution has the lowest path cost.

Toy Problem
Eg. Problem

8 puzzle Problem

The problem solving approach has been applied to a variety of task environments
(1) Toy problems
(2) Real world problems

7	2	4
5		6
8	3	1

start state

	1	2
3	4	5
6	7	8

Goal state

8 puzzle - 3×3 board

with 8 numbered tiles & a blank space.

States: Location of each of the eight tiles & a blank in one of nine squares

Initial State: Any state can be designated as the initial state.

Successor function: All legal states that result from

trying the four actions (left, right, up, down)

Goal State: This checks whether the state matches the goal configuration as shown in Goal state diagram.

Path cost: Each step costs 1, so the path cost is the number of steps in the path

8 puzzle belongs to the family of sliding-block puzzles. These are

called as NP-Complete

8 puzzle has $9! / 2 = 181,440$ reachable states & can be easily solved

15 puzzle (4×4 board) has 1.3 trillion states. It can be solved in few milliseconds by the best search algs

24 puzzle (5×5 board) can be solved as difficult to solve.

2)
Eg)

Romanian Holiday search problem

Eg) Route finding problem

Currently in Arad

Flight should move to Bucharest
Formulate goal: To reach Bucharest

States : various state, sequence of cities : Arad, Sibiu, Fagaras, Buch

Initial state : Arad

Successor function : drive between cities

Goal Test : check whether Bucharest is reached

Path cost : Min. cost

3)

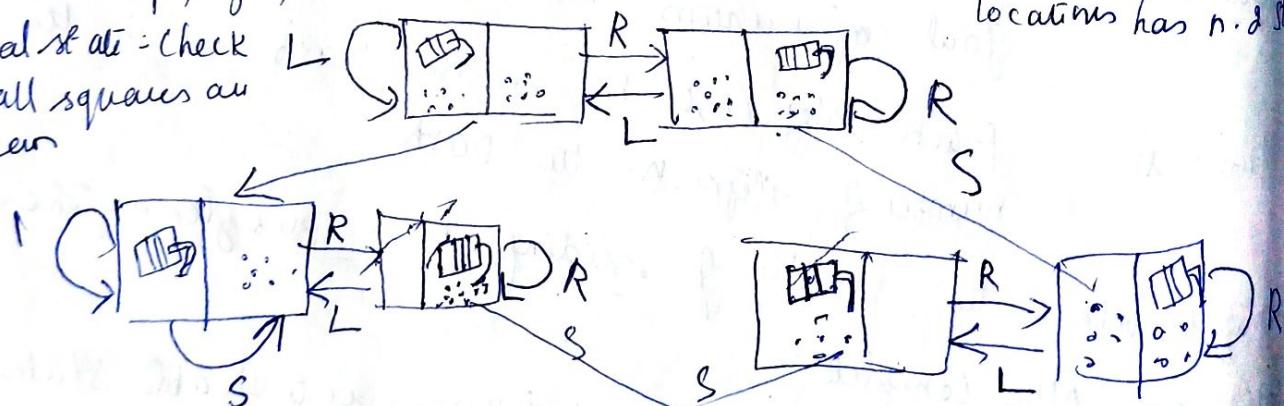
Vacuum World

i) Initial state : Any state

ii) Actions : Left, Right, Suck

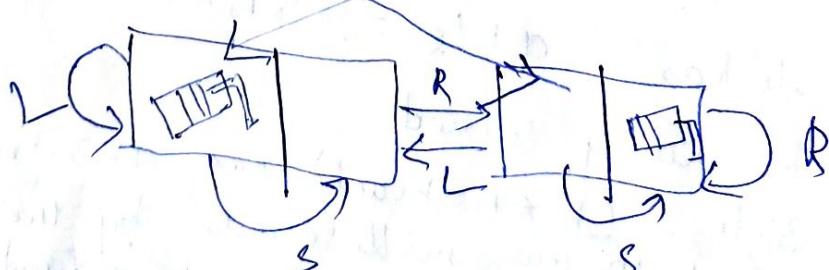
iii) Goal state : Check if all squares are clean

(i) There are $2 \times 2 = 8$ possible states
A larger environment with $n \times n$ locations has $n \cdot d^s$ states



iv) Path cost :

Each step costs 1,
so the path cost is
the no. of steps in the
path



4) 8 queens problem :

Place the queens on the board such that they do not attack with each other.

States : Any arrangement of 0 to 8 queens on a board

Initial state : No queen on board

Successor function : Add a queen to any empty square

Goal test : 8 queens on the board, none attacked

5. Real-world problems :

i) Route-finding problem. → computations, military operations planning & airline travel planning systems

dir-line planning.

- Initial state :

- Successor function :-

- Goal test : Are we at destination by some pre-specified time?

- Path cost : monetary cost, waiting time, flight time, etc.

ii) Travelling Sales Person problem.

alp-hard.

Robot navigation

Automatic assembly navigation

Internet Searching

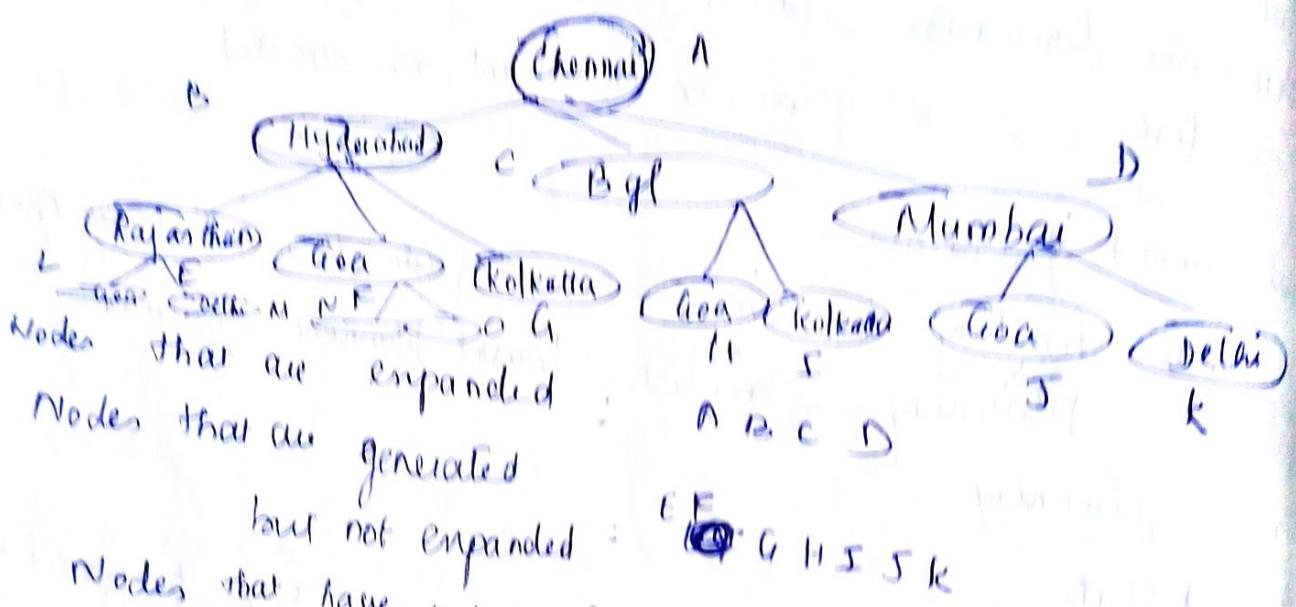
Search for solutions

Search Tree: A search tree is generated by the

- (i) Initial state
- (ii) Successor function

→ State space.

Ex) The foll. is a search tree for finding a route from Chennai to Delhi



Initial state: Root of the search tree

The first step is to test whether this is a goal state. Function to the current state is expanded by applying the successor function to the current state.

The choice of which state to expand is determined by the search strategy.

There are infinite no. of paths in the state space so the search tree has an infinite no. of nodes.

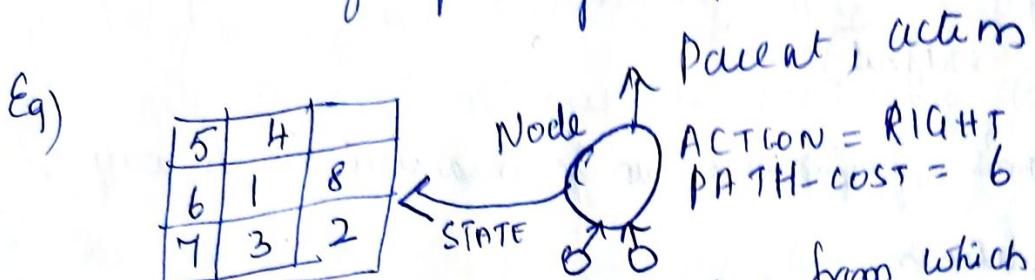
(i) STATE: a state in the state space with 5 components.

(ii) PARENT NODE: the node in the state space to which the node is attached in the search tree that generated it.

ACTION :- the action that was applied to the parent to generate the node.

PATH-COST: the cost from parent node denoted by $g(n)$

DEPTH : the no. of steps along the path from the initial state



Nodes are the data structure from which the search tree is generated

Fringe: Fringe is a collection of nodes that have been generated but not yet been expanded. Each element of the fringe is a leaf node; i.e., a node with no successors in the tree. The collection of these nodes are implemented as a queue.

MAKE-QUEUE(element...) \Rightarrow creates a queue with the given elements

EMPTY(queue) \Rightarrow returns true only if there are no more elements in the queue

FIRST(queue) \Rightarrow return FIRST(queue) and removes it from the queue

INSERT(element, queue) \Rightarrow inserts an element into the queue and returns the resulting queue

INSERT-ALL (elements, queue) \Rightarrow inserts a set of elements into the queue and returns the resulting queue.

Measuring Problem-solving Performance

The job of problem-solving algorithm is either failure or success.
(Some algorithms might stuck in an infinite loop and not return an output).

The algorithm's performance can be measured in 4 ways:

- (i) completeness
- (ii) optimality
- (iii) time complexity
- (iv) space complexity

Uninformed Search Strategies

Strategies that know whether one non-goal state is more promising than another are called "Informed Search" strategies.

- ① Breadth-first Search
- (i) Depth-first Search
- (ii) Uniform-cost Search
- (iv) Depth-limited Search
- v) Iterative deepening search

Breadth-first Search (BFS)

Searches breadth-wise.

Traverses the tree where each node is a

solution potential candidate for

It expands nodes from the root of the tree and then generates one level of the tree at a time until a solution is found.

Initially the queue contains just the root.

In each iteration, nodes at the head of the queue are removed and then expanded.

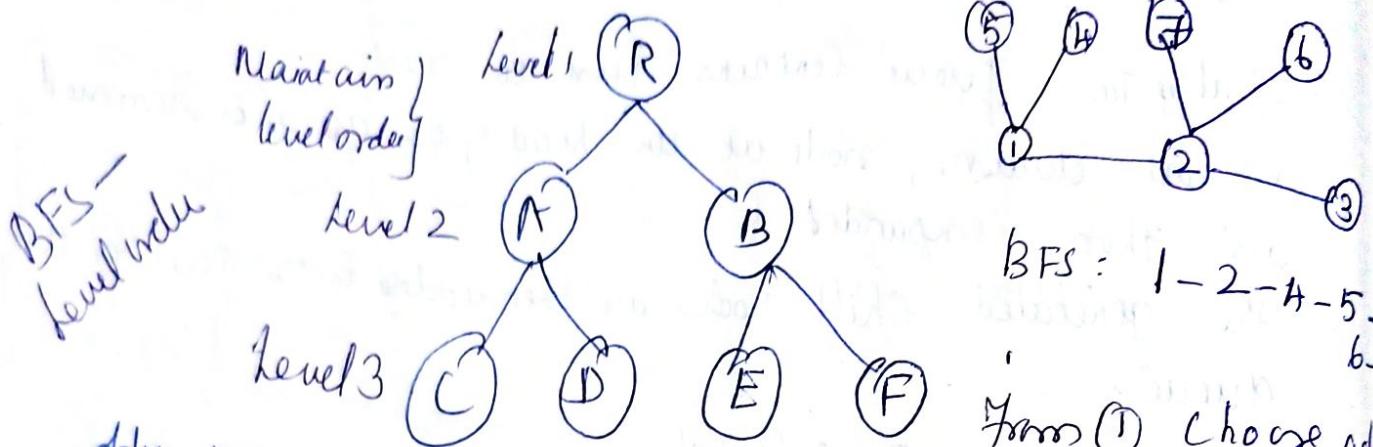
The generated child nodes are then added to the tail of the queue.

Algorithm: Breadth - First Search

1. Create a variable called NODE-LIST and set it to the initial state.
2. Loop until a goal state is found or NODE-LIST is empty
 - a) Remove an first element E, from the NODE-LIST
if NODE-LIST was empty then quit.
 - b) For each way that each rule can match the state described in E do:
 - i) apply the rule to generate a new state
 - ii) if the new state is the goal state, quit and return this state
 - iii) otherwise add this state to the end of NODE-LIST

Note: BFS always finds a shortest path to a goal.
since each node is generated in constant time,
the amount of time used by Breadth first search is
proportional to the no. of nodes generated, which is a constant.

since the nodes at level d is b^d , the total no. of nodes in the worst case is $b + b^2 + b^3 + \dots + b^d$ i.e., $O(b^{d+2})$

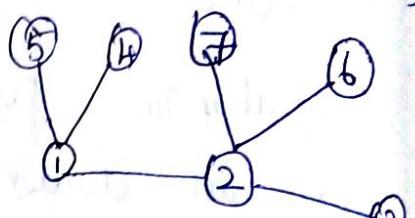


Advantage:

1. BFS will never trap exploring unwanted path.
2. If there is a solution BFS will definitely find it.
3. If there is more than one solution, BFS can find the minimal one that requires less no. of steps.

Drawbacks:

1. Memory requirement is saved, since each level must store no. of nodes stored.
2. The space complexity of BFS is $O(b^d)$. If the solution is further away from the root, BFS will consume lot of time.

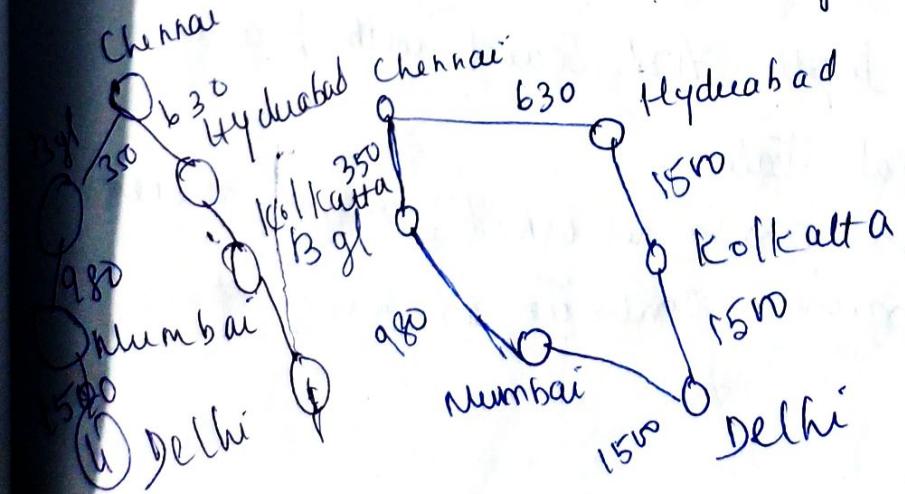


BFS: 1-2-A-5-6

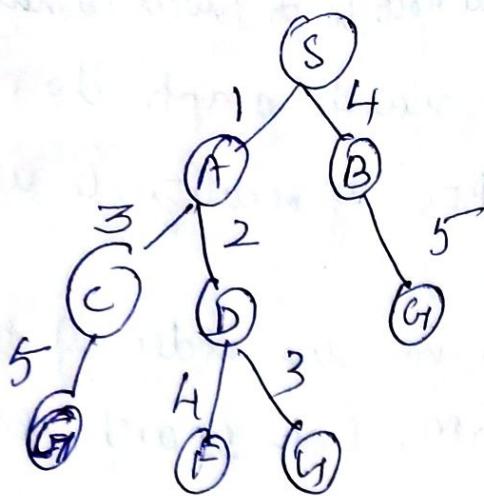
From ① choose adjacent vertices. It can be in any order.

2) Uniform Cost Search (find goal node with lowest cumulative cost)

- (i) If all the edges in the search graph do not have the same cost, then BFS generalizes to uniform-cost search.
- (ii) Instead of expanding nodes in the order of their depth from the root, uniform cost-search expands nodes in the order of their cost from the root.
- (iii) At each step, the next step n to be expanded is one whose cost $g(n)$ is lowest where $g(n)$ is the sum of the edge costs from the root to node n. The nodes are stored in a priority queue. This algorithm is also known as Dijkstra's single-source shortest algorithm.
- (iv) Whenever a node is chosen for expanding by uniform cost search, a lowest-cost path to that node has been found.
- (v) The worst case time complexity of uniform cost-search is $O(b^c/m)$, where c is the cost of an optimal solution and m is the min-edge cost.



Chennai - Bgl - mumbai -
Delhi - 283
Chennai - Hyderabad -
Kolkatta - Delhi -
3630



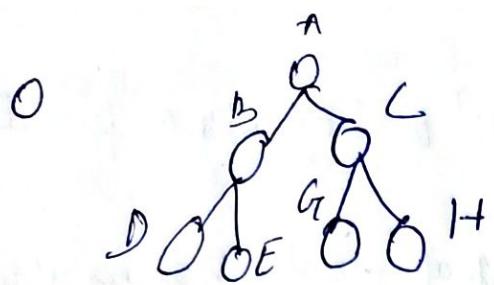
S - A — D — G

Depth first Search (DFS)

DFS searches deeper into problem space. BFS always successor of the deepest unexpanded node. It uses last-in first-out stack for keeping the unexpanded nodes.

Algorithm :- Depth first Search

1. If the initial state is a goal state, quit
2. Otherwise loop until success or failure is signalled.
 - (a) Generate a state say E and let it the successor of the initial state. If there is no successor, signal failure.
 - (b) Call Depth-First Search with E as the initial state.
 - (c) If success is returned, signal success and terminate in this loop.



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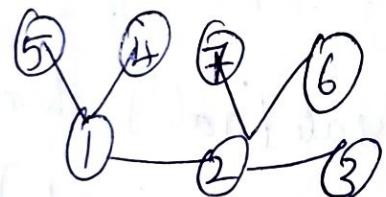
Advantages :-

- 1) Memory requirement is linear with respect to Search graph, where as in BFS we require more space.
- 2) Time complexity $O(b^d)$, since it generates the same set of nodes as BFS, but in different order.
- 3) If DFS search finds solution without exploring much in a path, then time & space it takes will be less.

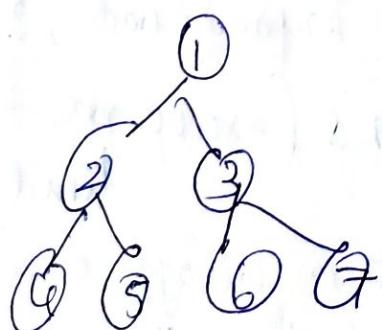
Drawbacks

- ① Not guaranteed to find the solution.
- ② There is no guarantee to find a minimal solution, if more than one solution exists.

Eg)



DFS: ①, 2, 3, 6, 7, 4, 5



1, 2, 4, 5, 3, 6, 7

Pre-order traversal

4. Depth Limited Search (III to 4)
DFS will not find a goal if it searches down a path that has infinite length. So in general, DFS is not guaranteed to find a solution. So it is not complete. This problem is eliminated by limiting the depth of the search to some value l . This is another way of preventing DFS from finding a goal, the goal is deeper than l will not be found.

function Depth-Limited-Search (problem, limit) return solution
return Recursive-DLS (make-node (initial-state [problem]))

function Recursive-DLS (node, problem, limit) return solution / fail / cutoff
cut-off-occurred ? false

IF Goal-Test (problem, state[node]) then
 return Solution (node)
else if Depth [node] = limit then return cut off
else for each successor in Expand (node, problem)
 result Recursive-DLS (successor, problem)

if result = cutoff then return cutoff
else if result not = failure then return result
if cutoff occurred ? then return cutoff else return

Depth-limited search can be terminated with 2 conditions of failure.

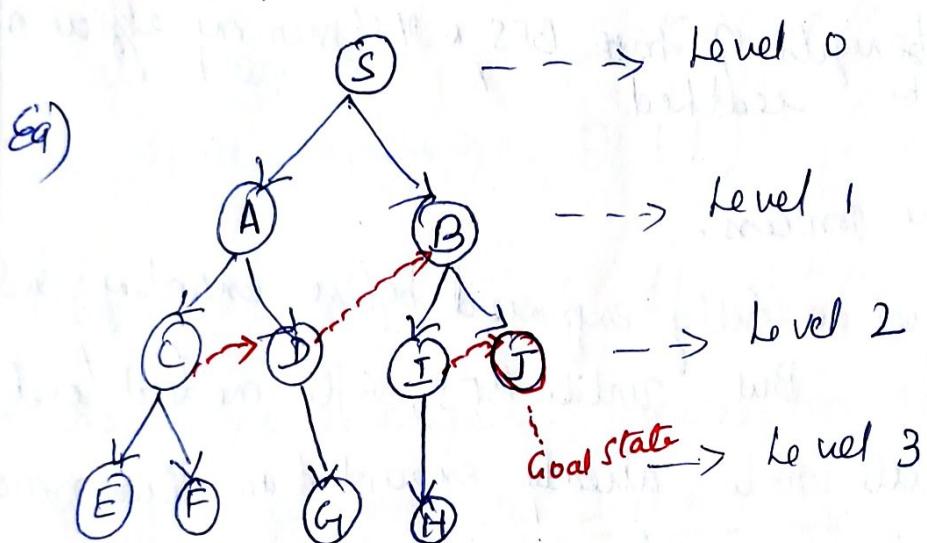
- (i) Standard failure value : It indicates that problem does not have any solution.
- (ii) Cutoff failure value : It defines no solution for the problem within a given depth limit.

Advantages

Memory efficient

Drawbacks

- (i) incompleteness
- (ii) If the problem has more than one solution, it may not be optimal.



Limit - Level 2 \Rightarrow When level 3 is reached,
Search terminates.

Time Complexity = $O(b^l)$.

Space complexity = $O(b \times l)$.

5. Iterative Deepening Search

If a depth-limited depth first search limited depth l does not find the goal, try it again with $l+1$. Continue this until goal is found.

Make depth-limited depth first search complete by applying it greater values for the depth limit l .

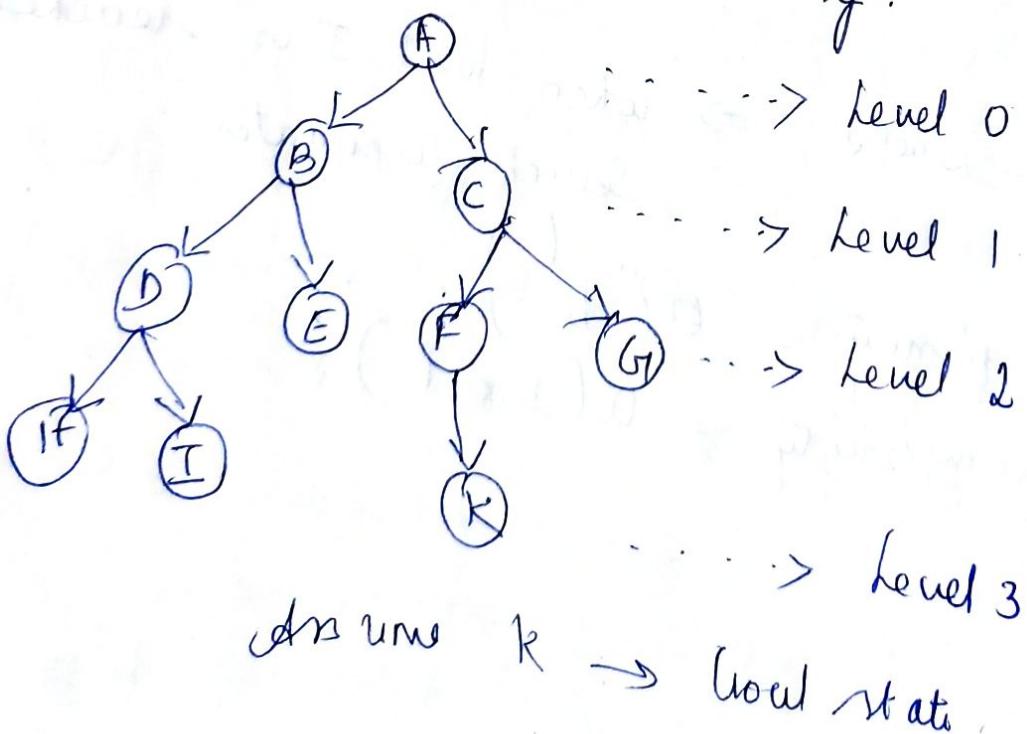
It is a combination of BFS & DFS.
Advantage:

(fast search)

Drawback: Incorporate benefits of both BFS & DFS (memory efficiency). Goal can be reached.

Repeats the process.

Like BFS, a level is fully explored before extending to the next level. But unlike BFS, after one level is fully explored, all nodes already expanded are thrown away. The search starts with a clear memory.



1st iteration $\rightarrow A$: not goal

2nd iteration $\rightarrow A, B, C$: not goal

3rd iteration $\rightarrow A, B, D, E, C, F, G$: not goal

4th iteration $\rightarrow A, B, D, H, I, E, C, F, K, G$ \rightarrow Goal State is reached

The alpm is complete if the branching factor is finite.

Time Complexity:

branching factor : b

depth : d

worst case time complexity : $O(b^d)$

space complexity : $O(bd)$.

b. Bidirectional Search Algorithm

- (i) As the name implies, searches in two directions at the same time one forward from the initial state and the other backward from the goal.
- (ii) This is done by expanding tree with branching factor b and the distance from start to goal d.
- (iii) The search stops when ~~over~~ searches from both directions meet in the middle.
- (iv) Bidirectional search is a brute-force search algm, that requires an explicit goal state instead of simply a test for a goal condition.

v) once the search is over, the path from the initial state is concatenated with the inverse of the path from the goal state to form the complete solution.

vi) Bidirectional search still guarantees optimal solution.

Time complexity: $O(b^{d/\alpha})$ since each search needs only to proceed to half the solution path.

Space complexity: $O(b^{d/\alpha})$

Advantages

- (i) Speed - sum of time taken by 2 searches (forward and backward) is much less than $O(b^d)$.
- (ii) It requires less memory.

Drawbacks

- (i) Implementation of bidirectional search is difficult because additional logic must be included to decide which search tree to extend at each step.
- (ii) Goal state need to be known in advance.
- (iii) The algs must be too efficient to find the intersection of two search trees.
- (iv) It is not always possible to search backwards through possible states.

Summary of algorithms

Criterium	BFS	DFS	Depth-limited	Iterative deepening	Bidirectional search
Completeness?	YES	NO	YES, if $I \geq d$	YES	YES
Time	b^{d+1}	b^m	b^l	b^d	$b^{d/2}$
Space	b^{d+l}	b^m	b^l	b^d	$b^{d/2}$
Optimal?	YES	NO	NO	YES	YES

Searching with Partial information

If the knowledge of states/actions is incomplete, it leads to 3 distinct problems:

- Sensorless problem: If no sensors, it could be in one of several possible initial states, and each action might lead to one of several possible successor states.
- Contingency problem: If the environment is partially observable or if actions are uncertain (adversarial) then the agent's percepts provide new information after each action. Each possible percept defines a contingency that must be planned for.

A problem is called adversarial if the uncertainty is caused by actions of another agent.

Exploration problems:

When the states and actions of the environment are unknown, the agent must act to discover them as an explore-and-exploit problem.

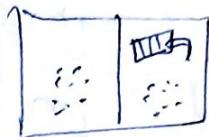
Sensorless problems

A Vacuum agent knows all effects of its action, by sensors. The 8 possible states of Vacuum are:

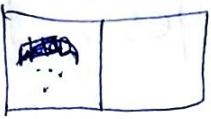
1.



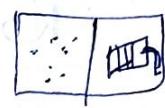
(1)



3



(2)



5.



(3)



6.



(4)



(5)



(6)



(7)



(8)



- No sensors
- Initial state (1, 2, 3, 4, 5, 6, 7, 8)
- After action [RIGHT] the state (2, 4, 6, 8)
- After action [SUCK] the states (4, 8)
- After action [LEFT] the states (3, 7)
- After action [SUCK] the state (8)

∴ [right, suck, left, &suck] can be caused without any sensor

Partial knowledge of states & actions

lead to Contingency problem

If uncertainty is caused by actions of another agent:
Adversarial problem

When states and actions are unknown: Exploration Problem

Eg) According to Murphy's Law:

suck can dirty a clean carpet

$$\text{Percept} = [L, \text{Dirty}] = \{1, 3\}$$

$$[\text{suck}] = \{5, 7\}$$

$$- [\text{RIGHT}] = \{6, 8\}$$

$$[\text{suck}] \in \{6\} = \{8\} \Rightarrow \text{success}$$

$$\text{But, } [\text{suck}] \in \{8\} \Rightarrow \text{failure}$$

Belief-State: no fixed action sequences guarantee solution.

Relax requirement:

$$[\text{suck}, \text{RIGHT}], \text{ if } \cancel{[\text{dirty}]} \text{ then } [\text{suck}]$$

contingency planning