

# Artificial Intelligence

DT-09/09/23

→ AI contains multiple techniques like searching, planning, learning

Deep learning → Artificial neural network.

## History of AI:

1943 - Evaluation of Artificial neuron

1950 - Turing Machine

1956 - Dartmouth conference (organizer - John McCarthy)

1966 - ELIZA - first botchat

1972 - first intelligent robot - WABOT

1974 to 1980 - first winter of AI

1980 - Development of expert system.

1987 to 1993 - Second winter of AI

1997 - IBM Deep Blue. (Chess playing computer)

2002 - people started developing robots in homes

AI in home

→ Roomba - vacuum cleaner

2006 to 2009 - lots of work happened in Artificial network.

Reboot the Artificial neural network

2009 - First self driving car came self

\* Google's first self driving car.

2011 - AI from IBM Watson (quiz show)

2011 to 2014 - Developing of smart phones assistants

Siri, Cortana  
Apple Microsoft Amazon  
Alexa, google now

2014 - Generating Generative Adversarial Network (GAN)

- Image generating
- stable Diffusion
- Dalle

2016 - Google Alpha Go (Game Go)  
defeat the champion of Go game.

2019 - Transformer - NLP

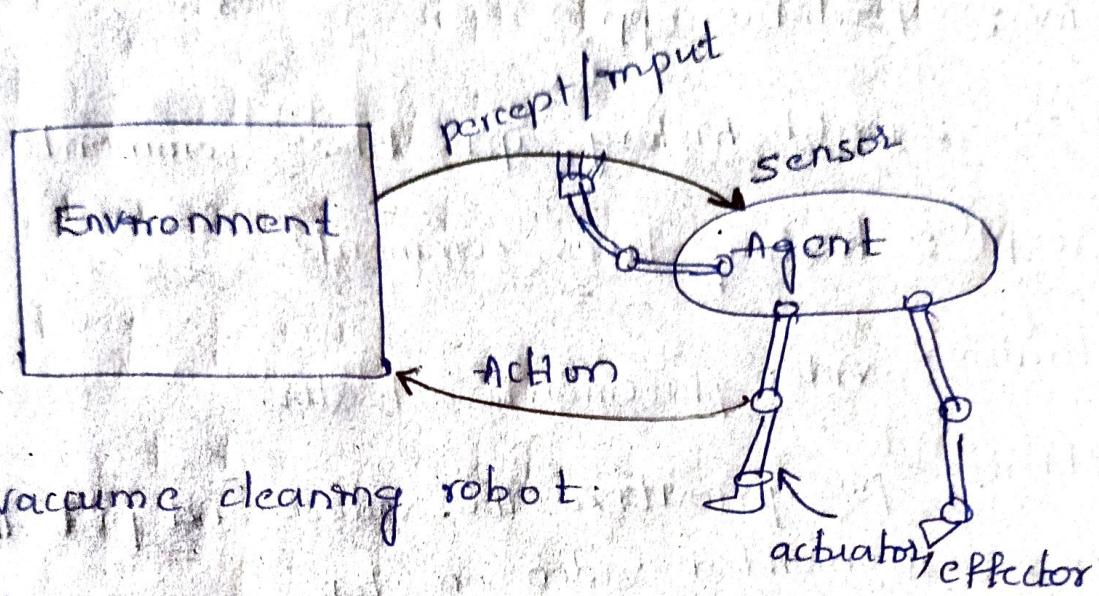
### BOOK:-

A.I - A Modern Approach by Russel & Norvig

History of A.I

# Intelligent Agent:

Agent:



eg: vacuum cleaning robot:

Robotic Agent

- vacuum cleaning robot      Environment: rooms, dirt.
- Sensors - dirt, location sensor.
- percept - clean or not, location A or B.

Actuators: wheel, succession pipe <sup>Actions</sup> move left, right, suck the dirt.

eg: car driving agent

Sensors - camera, gps, auditory sensor

Percept - visual and auditory input, location

Actuators - steering, acceleration, break, start, stop,  
actions - accelerate, apply break etc -

Env - roads, traffic, other cars

\* Software agent      API = application programming interface

- youtube video recommending by agent.

Env : youtube app.

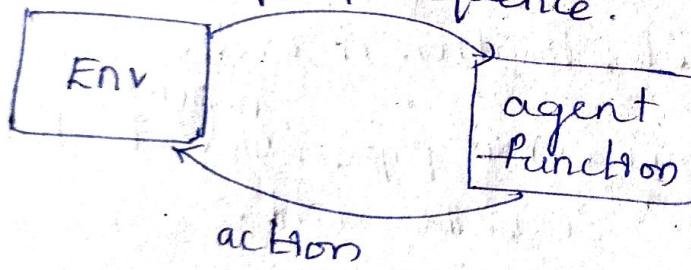
Sensors : watch history api, likes, comments api

Percept : history, likes, comments

Actuator : video placement actuator.

Actions: prioritize the videos and put the recommended video on top.

→ Agent function: mapping b/w percept sequence and the action.



Eg - vacuum cleaning agent.

percept sequences (Agent function)	action
[A, clean]	right
[A, dirty]	clean
[B, clean]	left
[B, dirty]	clean

\* Rational Agent: based on outcome of your action.

- does the right thing

P - Performance measure

E - agent's prior knowledge of Environment

A - Actions

S - Agent's percept sequence.

Ex- case-1: Performance Measure: +1 point for cleaning the room.

case-2: performance measure: +1 point for cleaning room  
-1 point for movement

[A, clean] - do nothing

[A, clean], [A, clean] - do nothing

[A, clean], [A, clean], [A, dirty] - clean

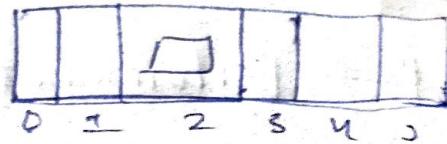
[A, clean], [A, dirty], [A, clean] - right

↳ rational

# \* Properties of Environment:

## ① fully observable vs Partially observable

↓  
you know about the information.



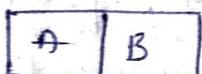
→ Global dart sensor (fully observable)

→ Local dart sensor (Partially observable)

## ② Discrete vs Continuous:

↓  
count the no. of

possible states.



↓  
e.g. car driving

④ - clean, clean

clean, dirty

dirty, clean

dirty, dirty

e.g. vacuum cleaning  
in two rooms

## ③ Episodic Vs Sequential:

↓

e.g. car assembly  
line

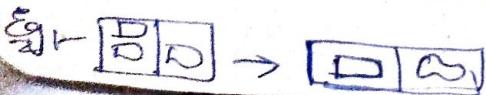
↓

e.g. car driving

## ④ Deterministic Vs stochastic:

↓

next state fully  
depend on current state



↓

next state is partially  
depend on current state

e.g. betting on the  
next game.

## ⑤ Single agent Vs Multi-Agent

↓

↓

Eg:- sudoku

Eg:- chess game

## ⑥ Static Vs Dynamic:

↓

↓

Env not  
changing

environment keep changing.

Eg:- crossword  
puzzle

Eg:- car driving

## ⑦ Known Vs Unknown:

↓

↓

Agent knows the  
environment will change  
for an action.

may not known

Eg:-

At: 11/09/2023

## \* Structure of an agent:

### → Agent Program

- simple reflex agent
- Model based agent
- Goal based agent
- Utility based agent
- Learning agent.

Eg:-

①

Thermostat agent to control room temperature  
- fire

②

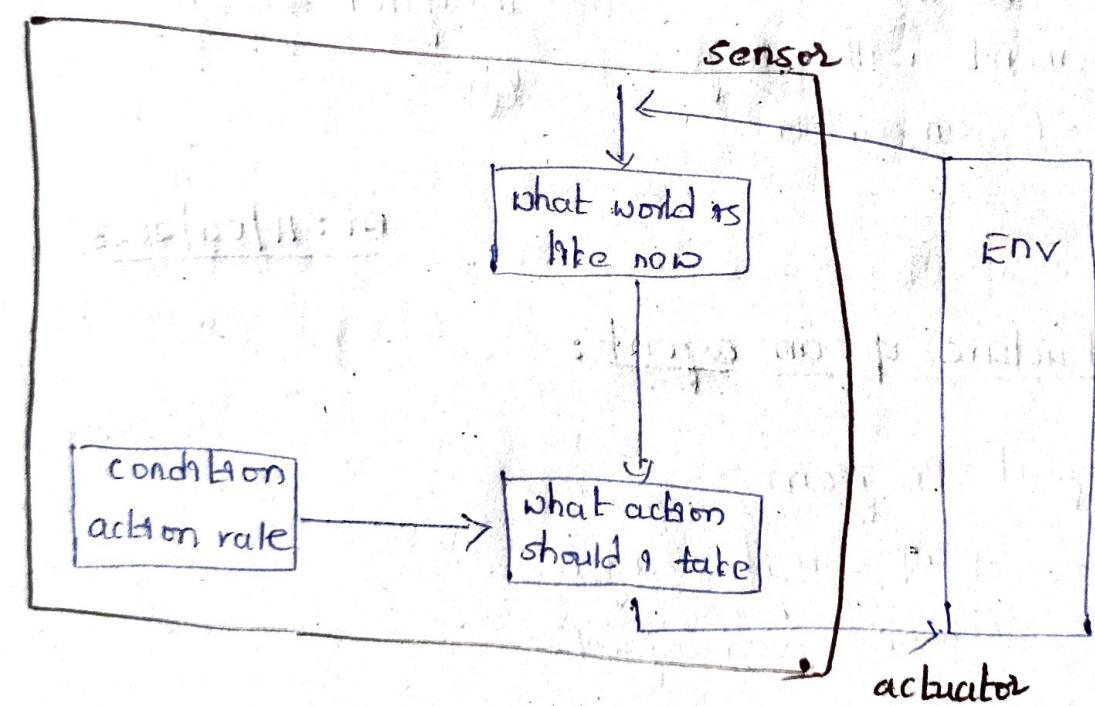
Simple vacuum cleaning agent

- ③ Autonomous delivery drone agent
- ④ Autonomous delivery drone that aims to minimize battery usage.

- ⑤ Teaching agent

- Simple reflex agent:

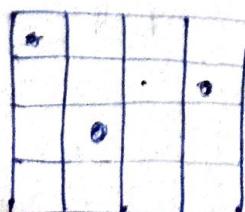
- Ex: Thermostat agent to control room temperature
- select action base on current percept and ignore previous percept history.



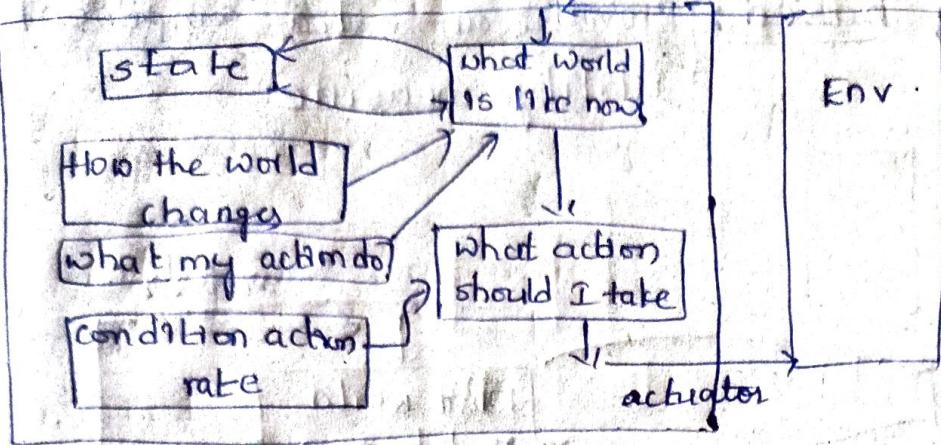
- Model based agent:

It requires representation of the world = Model

e.g. 4-Queens



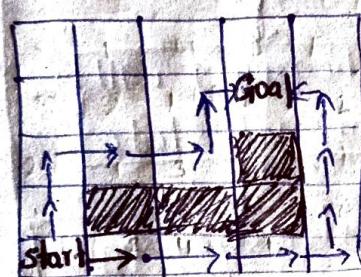
→ Model  
→ 2D array  
→ Graph



## Goal based agent:

e.g. autonomous delivery drone agent.

e.g.-

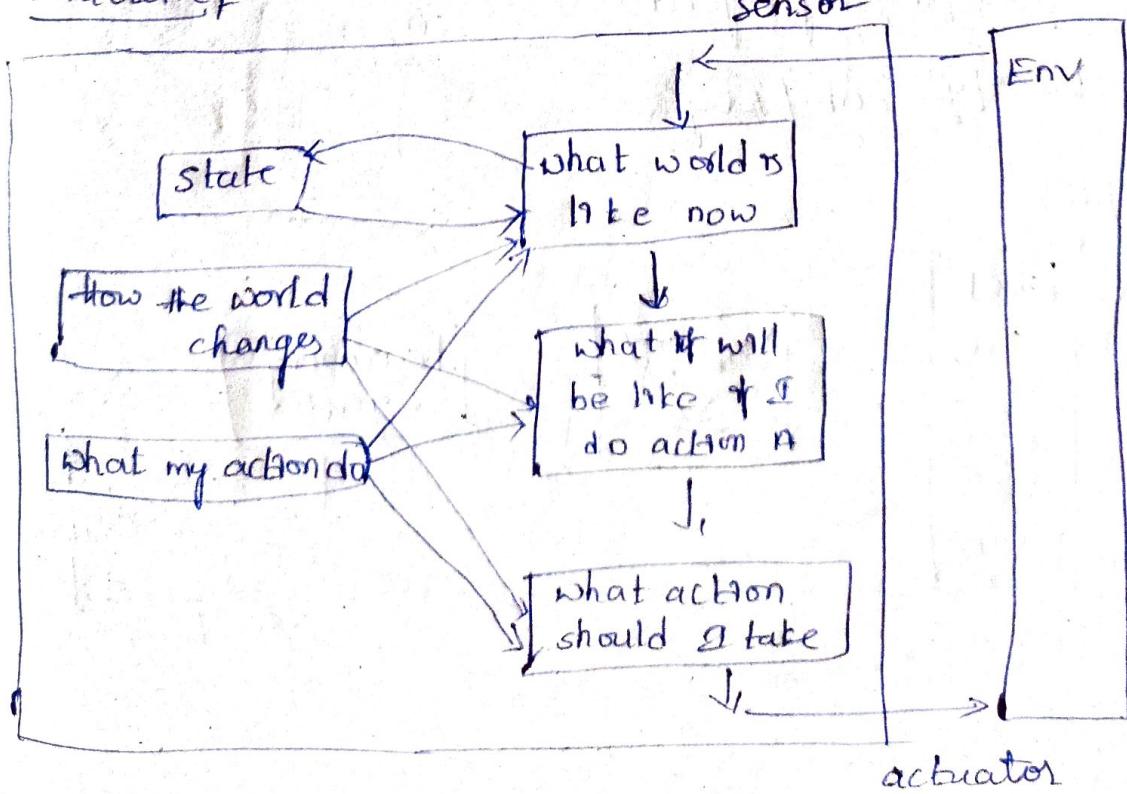


Action

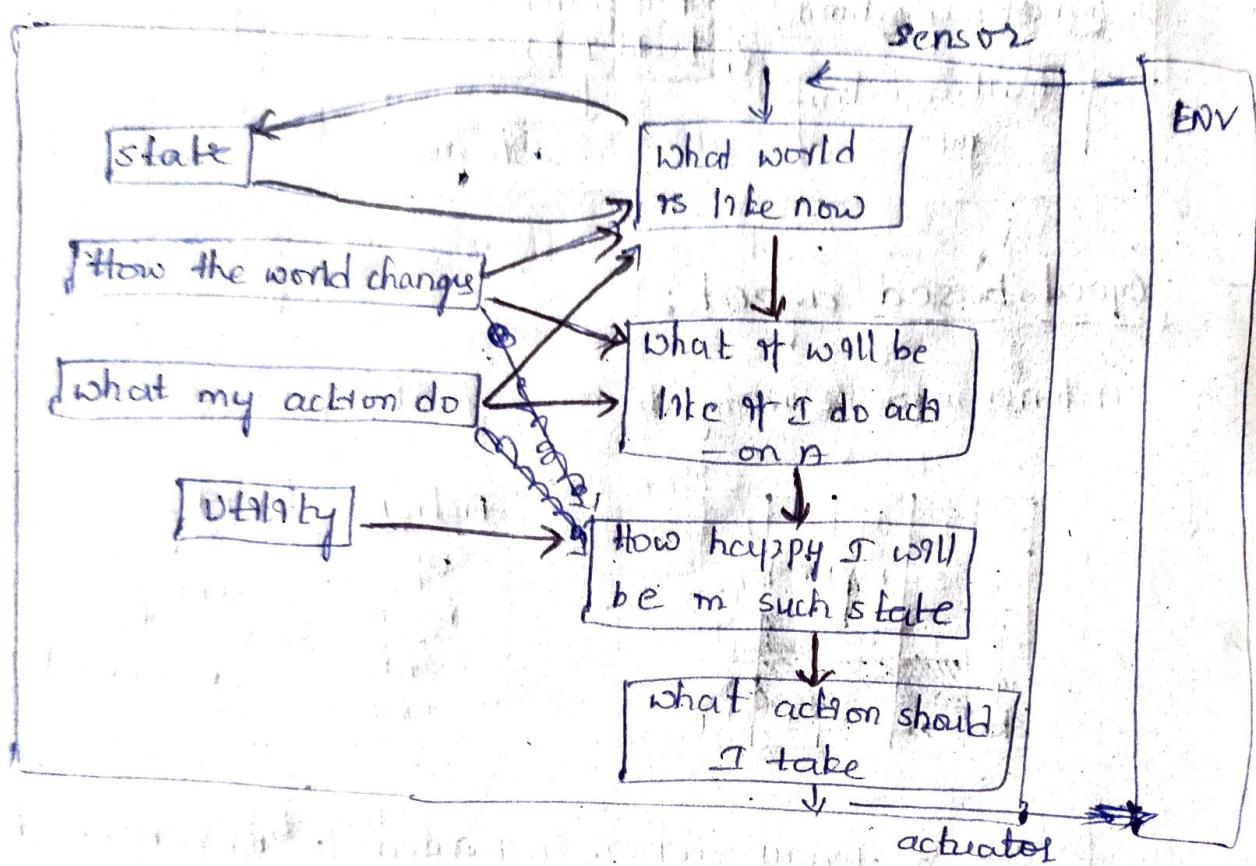
- up
- down
- left
- right

Structures \* Every action intended to minimize the distance from the goal.

## Structure



- Utility based: Provide an extra component of utility measure.

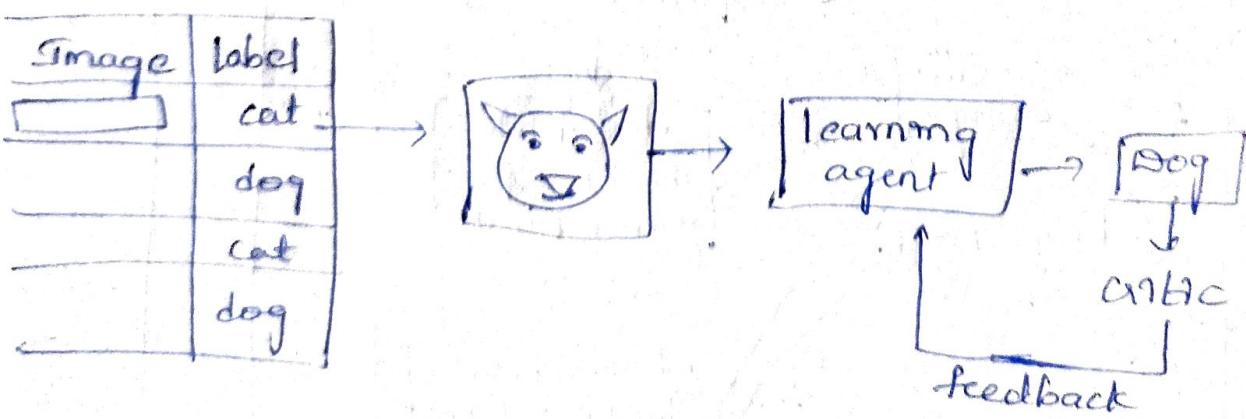


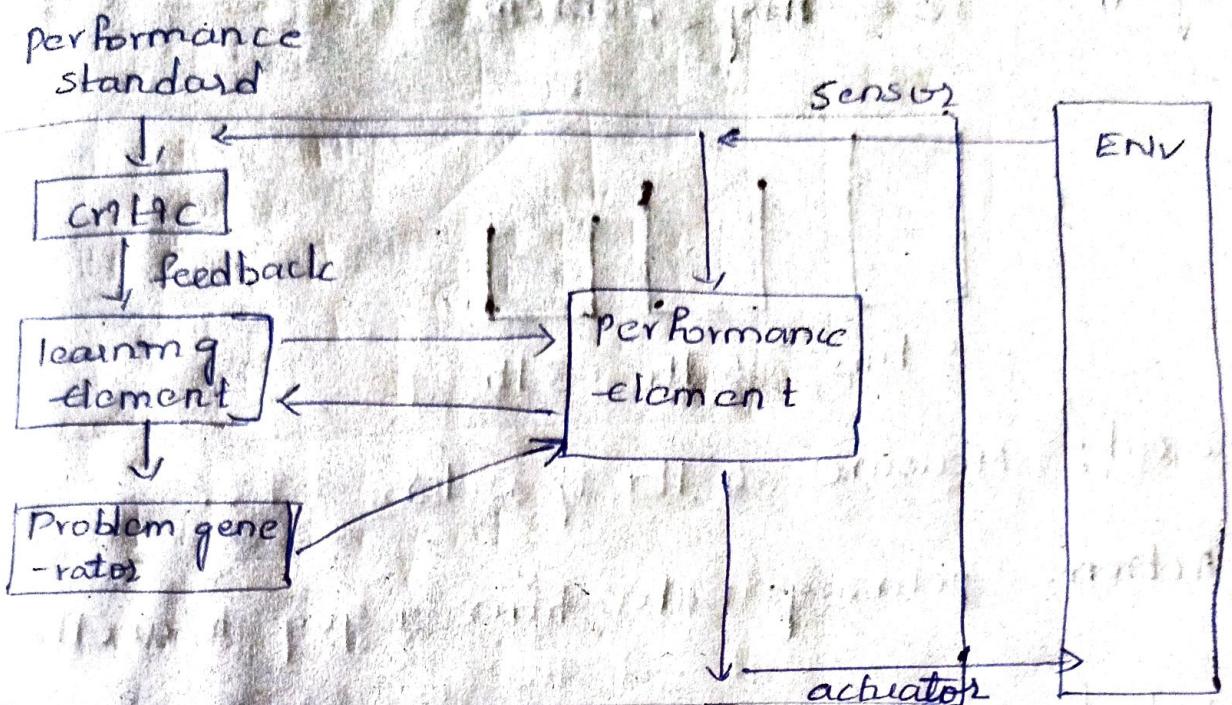
- Learning agent:

- cricket playing agent

e.g.: classify cat / dog

Examples





dt: 13/09/2023

### \* Problem Solving by Searching:

- Goal based agent can be implemented using searching.
- Task is to find sequence of actions take from initial state to goal state.
- easy problem/ environment / world
- Discrete
- static
- single agent
- known
- deterministic
- Action will never fail.

# Eg-①: 3 - Jug Problem

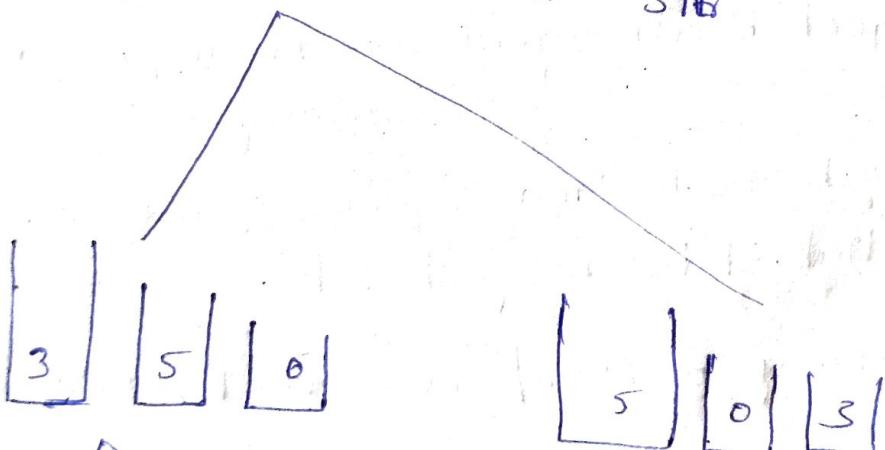
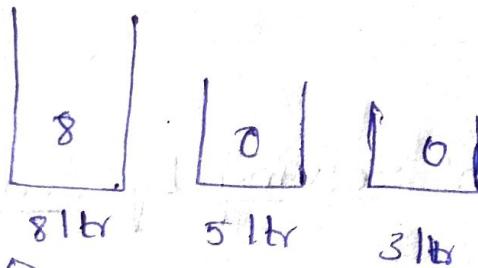


8 ltrs    5 ltrs    3 ltrs

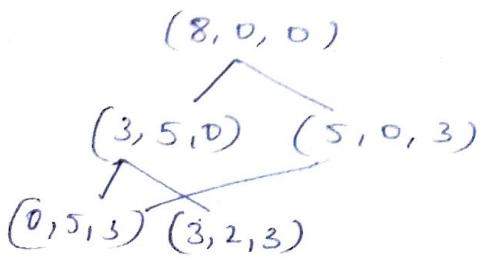
Goal: Measure 4 ltrs of water.

Action: pouring water from 1 jug to another

Initial state:



[0][5][3]    [3][2][3]



state space  
graph.

$(8, 0, 0)$  = initial

$(3, 5, 0)$

$(3, 2, 3)$

$(6, 1, 0)$

$(6, 0, 2)$

$(1, 5, 2)$

$(1, 4, 3)$

= goal.

another way

## Eq-@; 8 Puzzle Problem

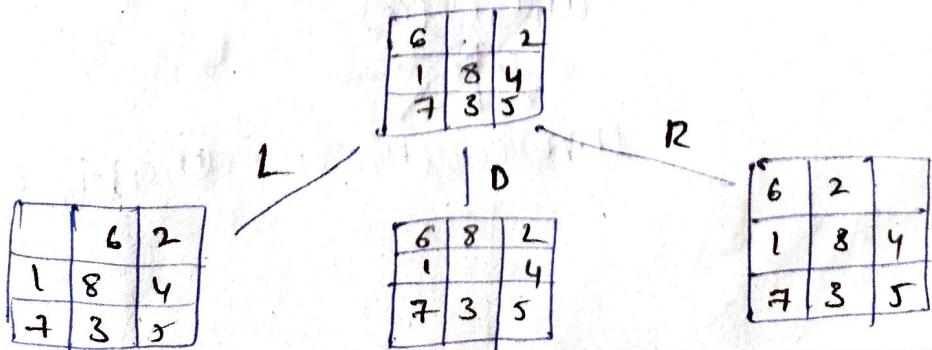
6		2
1	8	4
7	3	5

goal = state

1	2	3
8		4
7	6	5

Initial state

Actions : Move Left, Right, Up, Down.



Eg-③: Man, Lion, Goat, Cabbage.



- only man can drive the boat
- At time only one item he can take.
- If man is not present

L will eat the G

G " " " " C

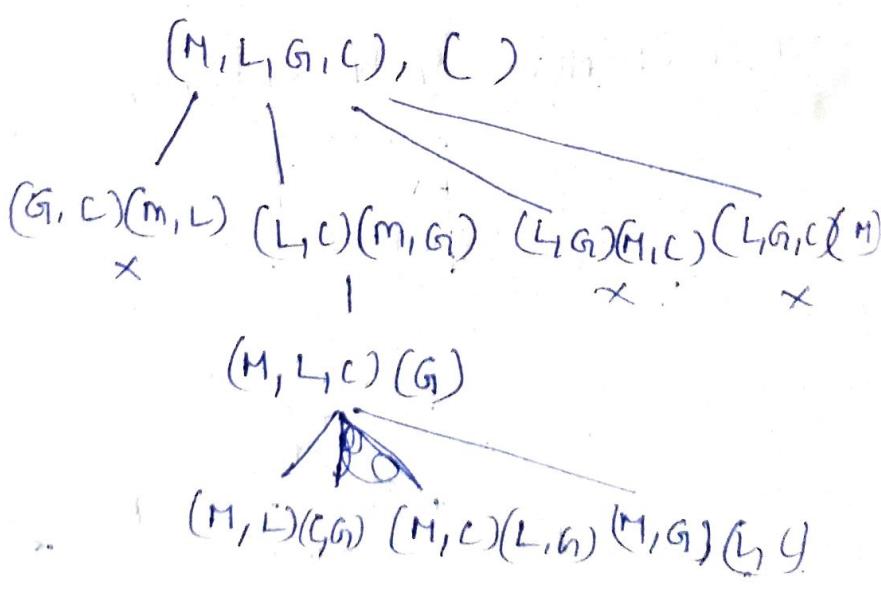
M L G C → M G

M L G → A C G

M L G → M L C

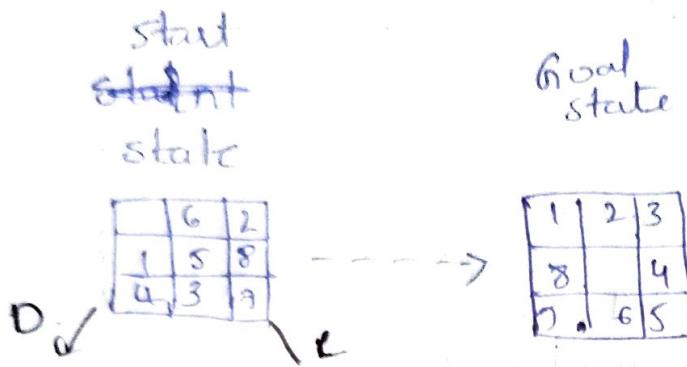
A G → M G L C

Initial state:



- Travelling Sales person
  - n-queens
  - Towers of Hanoi
  - Maze.
  - state space :
  - state space is implicit.
- 
- generate
  - select next state
  - from available candidate states
  - Goal test.
- Domain Independent algorithm :
  - Two function
- MoreGen(n)
- take a state input and return set of states that are reachable in step from input state.
- GoalTest(n)
- return true if input state is goal state otherwise return false.
- One set to maintain state of the node
    - open set.

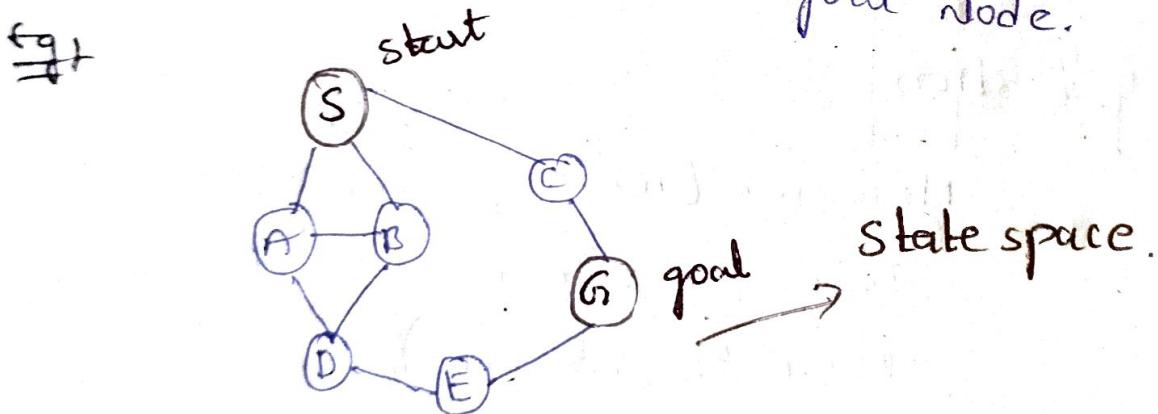
## → state Space Search:



$(M, L, G, C) \rightarrow ( ) (M, L, G, C)$

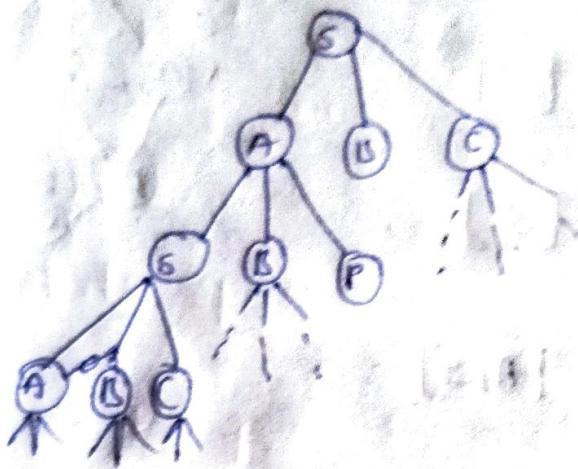
→ MoveGen(N): Returns the next states that can be reached in one step.

→ Goal Test(N): Return true if N is a goal node.



Nodes	More Gen(N)	Goal Test(N)
S →	(A, B, C)	F
A →	(S, B, D)	F
B →	(S, A, D)	F
C →	(S, G)	F
D →	(A, B, E)	F
E →	(D, G)	F
G →	(C, E)	T

# State Space Search Tree:



→ Algorithms for searching: ② types

1) Blind search / Uninformed search.

- DFS
- BFS
- DFID

2) Heuristic search / Informed search

- Best first search
- A\* search.

① Blind search:

② Simple search:

- we use two lists to maintain the status of the nodes.

③ OPEN - List of nodes that generated but not inspected

④ CLOSED - nodes already inspected / tested

open

[s]

[A, B, C]

↑

[D, B, C]

↑

[E, B, C]

↑

[E, G, B, C]

↑

Goal test(a) = T.

else

return

Algorithm:

OPEN  $\leftarrow$  [s]

CLOSED  $\leftarrow$  []

while (OPEN is not empty)

( $\bigcirc$ )

do pick a node n from OPEN

OPEN  $\leftarrow$  OPEN - n

CLOSED  $\leftarrow$  CLOSED ~~+~~ U n

If (Goal test(n) = True)

then return n

else

children = MoveGen(n)

closed

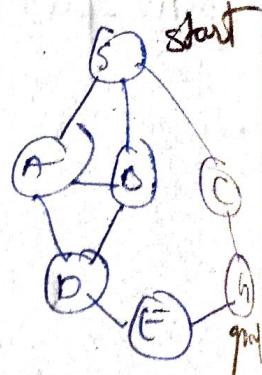
[]

[s]

[A, S]

[D, A, S]

[E, D, A, S]



OPEN  $\leftarrow$  OPEN U (children - CLOSED)

return FAILURE

(NODE, parent) - To find the path from start  $\rightarrow$  goal

simple search with path information:

OPEN

CLOSED

$[(S, \text{null})]$

$[ ]$

$[(A, S), (B, S), (C, S)]$

$[(S, \text{null})]$

$[(D, A), (B, S), (C, S)]$

$[(A, S), (S, \text{null})]$

$[(E, D), (B, S), (C, S)]$

$[(D, A), (A, S), (S, \text{null})]$

$[(G, E), (B, S), (C, S)]$

$[(E, D), (D, A), (A, S), (S, \text{null})]$

G-E-D-A-S-null

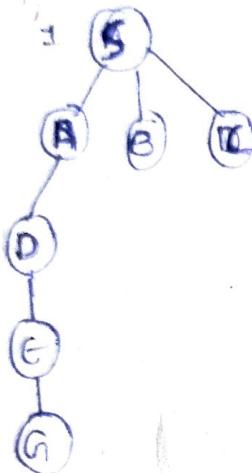
$[(G, E), (E, D), (D, A), (A, S), (S, \text{null})]$

construct  $\blacktriangleleft$

# Example search with Path Information:

DFS & BFS

DFS



OPEN

[S]

[ABC]

[DBC]

[EBC]

[GBC]

EMPTY

CLOSED

[]

[S]

[AS]

[DAS]

[EDAS]

[GEDAS]  
↑  
goal

OPEN

[S]

[ABC]

[BCD]

[CD]

[DG]

[G]

[]

EMPTY

CLOSED

[]

[S]

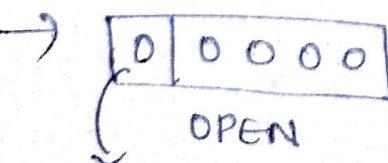
[AS]

[BAS]

[CBAS]

[DCBAS]

[GDCBAS]  
↑  
goal

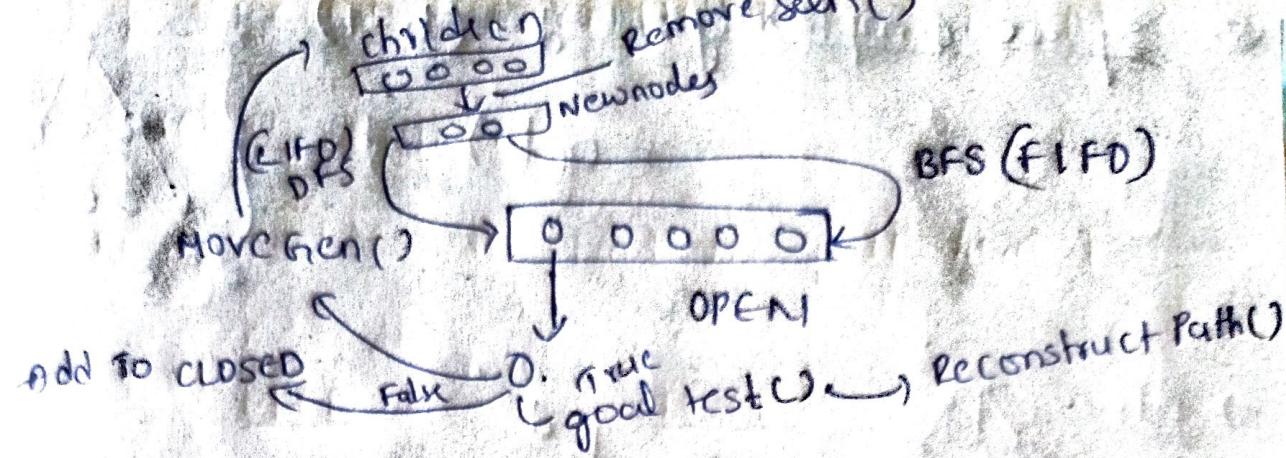


OPEN

we pick from begin  
and we add to  
closed.

CLOSED

→ head [2, 3, 4, 5]  
= 2 - (first element)  
→ tail [2, 3, 4, 5]  
= 3, 4, 5 (except first)



$$\text{APPEND} \Rightarrow [2, 3] : [4, 5] = [2, 3, 4, 5]$$

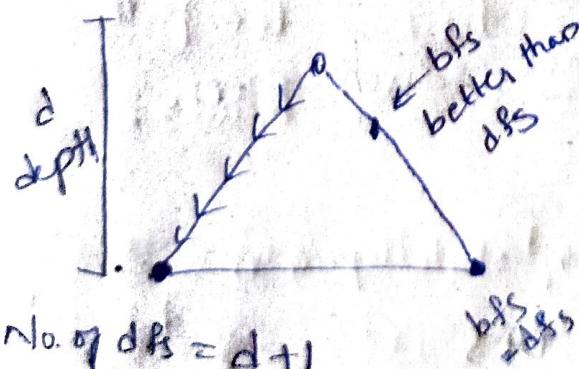
$\leftarrow$  assignment

= comparison.

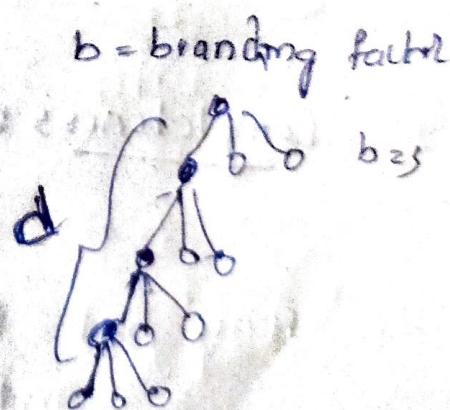
Dt: 25/09/23

→ DFS & BFS Comparison:

- | DFS                                    | BFS         |
|--|-------------|
| Time = $\Theta(b^d)$                   | $O(b^d)$    |
| Space - Linear                         | Exponential |
| Quality of solution (shortest path)    |             |
| Completeness (premises of goal exists) |             |
- Time Complexity: no. of nodes ~~inspected~~ inspected (closed)

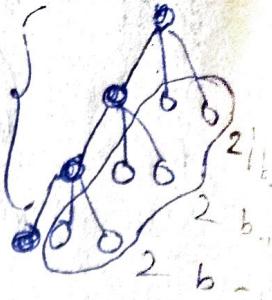


$$\text{No. of BFS} = b^{d-1}$$

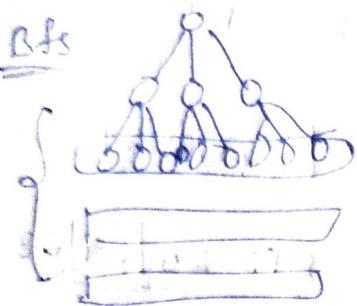


② Space Complexity :  $d * (b - 1)$

$$= O(d) \text{ (BFS)}$$



⇒ BFS



$$\Rightarrow b^d$$

③ shortest path :-

DFS

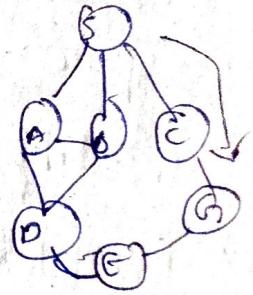
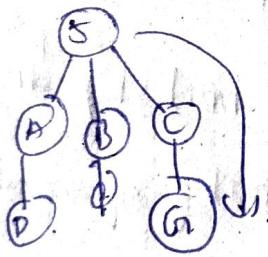
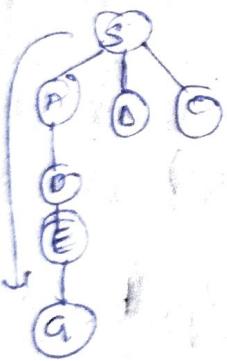
Not guaranteed

BFS

Always return the  
shortest path.

DFS

BFS



④

Completeness :-

Infinite search space.

→ May not terminate if state space is  
infinite → DFS

→ Always terminate if goal exists - BFS

# Depth First Iterative Deepening (DFID)

1) Time Complexity :

$$b^d + b^{d-1} + \dots + b^2 + b + 1 = O(b^d)$$

2) Space Complexity : Linear

3) Shortest Path : Yes

4) Completeness : Yes

Algorithm: depth information for each node need to be maintained.

(node, parent, depth)

→ Depth Bound DFS (DBDFS)

→ DFS(depth Bound)

→ DFID → for depth = 1 to n

DBDFS(depth)

DFDFS: depth Bound = 2

OPEN

CLOSED

[ (S, ml, 0) ] [ ]

[ (A, S, 1), (B, S, 1), (C, S, 1) ] [ (S, ml, 0) ]

[ (D, A, 2), (B, S, 1), (C, S, 1) ] [ (A, S, 1), (S, ml, 0) ]

OPEN

CLOSE

$\left[ (B, S_1, 1), (C, S_1, 1) \right]$

$\left[ (D, A_1, 2), (A_1, S_1, 1), (S, n_1, 0) \right]$

$\left[ (C, S_1, 1) \right]$

$\left[ (B, S_1, 1), (D, A_1, 2), (A_1, S_1, 1), (S, n_1, 0) \right]$

$\left[ (G, C_1, 2) \right]$

P  
goal ← stop

$\left[ (C, S_1, 1), (B, S_1, 1), (D, A_1, 2), (A_1, S_1, 1), (S, n_1, 0) \right]$

path:  $n_1 \rightarrow S \rightarrow C \rightarrow G$

At - 27/09/23

## \* DFID Example:

### MoveGen

S → DCBA

A → SBJE

B → SFA

C → SDHG

D → SIC

E → AJK

F → BKJ

G → CL

H → CML

I → DH

J → AFE

K → EF

L → GHM

M → LH

a. show the order in which DFID applies goal test to nodes.

b. what is the path it finds?

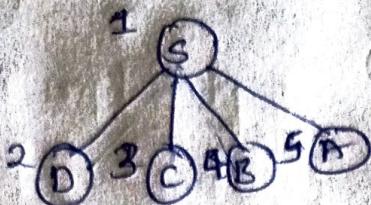
c. show the order in which BFS applies goal test to nodes.

d. How many nodes BFS will open.

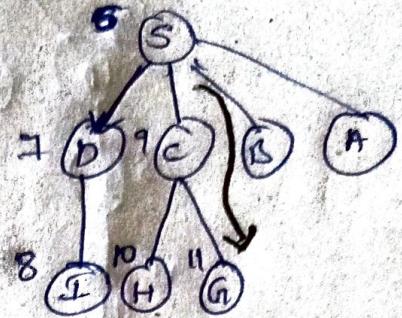
e. D.FS

f. No. of nodes opened on DFS

depth Bound = 1



depth Bound = 2



a. Order

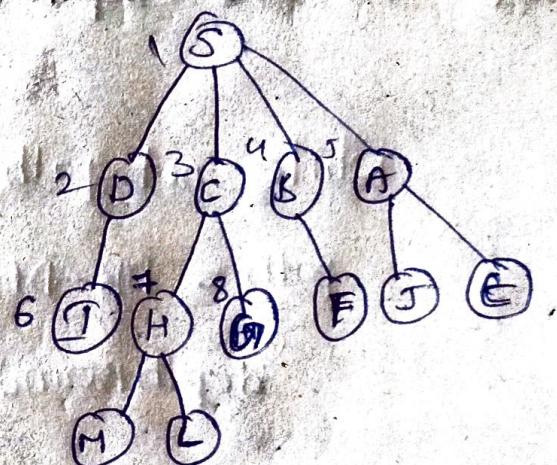
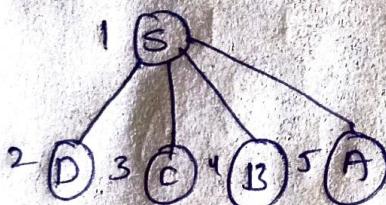
S D C B A

S D I C H G

b. Path

S → C → G

c.



order

S D C B A I H G

path →

S → C → G.

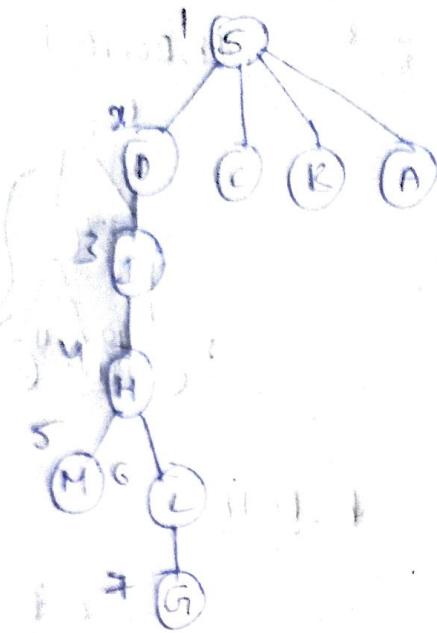
d. No. of nodes opened = 13

D) DFS

order → S - D - I - H - M - L - G

path → S - D - I - H - L - G

No. of nodes opened = 10



Alg: DB-DFS (depth Bound)

while not path:

path = DB-DFS (dept Bound)

depthBound + 1

Alg: DFIDS()

count  $\leftarrow$  1 // path length

path  $\leftarrow$  empty list

depthBound  $\leftarrow$

repeat

previousCount  $\leftarrow$  count

(count, path)  $\leftarrow$  DB-DFS

depth Bound  $\leftarrow$  depth Bound + 1

until

(Path is not empty) or (previousCount = count)

return path.

DB·DFS( $s$ , depth Bound)

count  $\leftarrow 0$

OPEN  $\leftarrow (s, \text{null}, 0) : []$

CLOSED  $\leftarrow []$

while OPEN is not empty

nodePair  $\leftarrow$  head OPEN

( $N$ ,  $_$ , depth)  $\leftarrow$  nodePair

↳ placeholder - used when variable is not needed to be stored

if Goal test( $N$ ) = True

return Reconstruct Path(nodePair, CLOSED)

else

CLOSED  $\leftarrow$  nodePair : CLOSED

if depth < depth Bound

children  $\leftarrow$  Move Gen( $N$ )

newNodes  $\leftarrow$  Remove seen (children, OPEN, CLOSED)

newPairs  $\leftarrow$  Make Node pair (newNodes,  $N$ , depth)

OPEN  $\leftarrow$  newPairs : tail OPEN

count  $\leftarrow$  count + length of newPairs

else

OPEN  $\leftarrow$  tail OPEN

return (count, EmptyList)