

← Principles of Artificial Intelligence (PAI) →

AI is a part of computer science concerns with the designing intelligent computer systems that is the systems which exhibit the characteristics associate with intelligence in human behaviour.

AI contains two terms :-

- 1) Intelligence
- 2) Artificial devices

A system with intelligence is expected to behave as intelligently as a human being. Secondly, A system with intelligence is expected to behave in best possible manners.

19/09/22

~~What is AI? What is the difference between conventional computer and AI computer?~~

Processing

Nature of Input

Explanation

Focus

Maintenance update

Reasoning capability

Application of AI

conventional computer

AI computer

Artificial Intelligence

↓
General
Problem
solving

↓
expert
system

↓
Natural
Lang.
processing

↓
computer
vision

↓
Robotics

↓
computer
Aided
Instruction

↑
Planning

- a) Symbolic Representation
- b) Knowledge Representation
 - Knowledge
 - Incomplete data
 - conflicting data
- c) Ability to learn

Symbolic representation - AI programs mainly deals with non numerical symbols and contrast strongly with commonly accepting view that computer can be used.

AI program can perform numerical calculations also where necessary but their significance or symbolic value will generally enter into decision process performed by program for eg:- in AI

The importance

are:-

- 1) It is the significant of theory — of nature of human intelligence and as such is of great interest to
- 2) It forms the basic belief that it is possible to build the program which can perform intelligent task as performed by human.

Knowledge

Knowledge can present as a modular. There is a correspond external and symbolic system.

This knowledge can be studied and understand in human terms for its representation are numerical.
for eg:-

Office book is modular book

AI program has capability to provide some solution even if all data relevant to the problem is not available at the time.

consequence of data being incom- simply leads to confusion are less certain. In real world, a decision may be wrong in the absence of relevant data. It can happen that the absence of complete data is inherent in the problem as in the game of the hedge.

conflicting data

Data items can even be contradictory to each other. This type of data is called conflicting data or data ~~connected~~ by ~~worried~~ about.

→ ~~for~~ Ability to learn

Intelligent computers are amplitude to learn from their mistakes so that they can improve their performance by taking account of past errors. This is related to capacity for generalizing to drawing analogy and discarding selected information. The function of given computers are learning similar to that of human beings is that of stimulating the machine process. The working of women mind can never be

programmed into machine so that no computer can ever have a mind of its own.

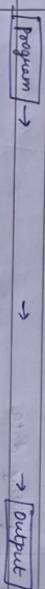
History of AI

First Generation Pre-1950

1) Can a self repairing machine be built which will locate

- Q3) Can a digital computer be programmed to program itself.
Using theoretical concepts from a computer program to learn.

22 Ques Agent :- It can be a program or a self tool



There are 3 parameters for an intelligent system:-

- | | | |
|--------------------------------|-----------------|-----------------------|
| <u>Search Technique</u> | <u>operator</u> | <u>Knowledge base</u> |
| <u>State Space Problems :-</u> | | |

State search Problems :-

A combination of all possible state for a given problem is called state space of problem.

2	1	3	4
5	6	5	9
13	8	11	12
14	15	15	X
11	2	3	4
5	6	7	8
9	10	11	12
13	14	15	X

$$N=4, \quad N(N-1) = 4 \times 3 = 12$$

2) N is even, puzzle is solvable in two conditions.

a) *Wiederholung*

b) metabolic waste

10

1920-1921
Sparta

Sarkar et al. / *Adolescent*

SOMBRERO JABÓN

卷之三

جامعة الملك عبد الله للعلوم والتقنية

THE JOURNAL OF CLIMATE

Revolving Books Books Books

ANSWERED BY [REDACTED] (ANSWERED)

卷之三

卷之三

卷之三

卷之三

卷之三

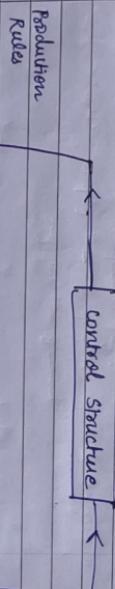
THE JOURNAL OF CLIMATE

Production system

- Rule-based knowledge representation
- Bridge b/w AI Research & expert system.
- Heuristic Model for Human behaviors.
- Strong data-driven nature

New Rules can easily added into system structure of PRODUCTION SYSTEM;

- i) Knowledge Base
- ii) Global Database
- iii) Control Structure



Problem representations

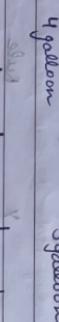
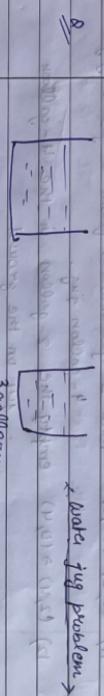
Transforming a problem into component of production system is called Problem representations.

State Space Representation

Initial state

Goal state

Rules



- 1) (x, y) if $x < 4 \rightarrow (4, y) \rightarrow$ full 4-gallon jug.
- 2) (x, y) if $y < 3 \rightarrow (x, 3) \rightarrow$ full 3-gallon jug
- 3) (x, y) if $x = d \& d > 0 \rightarrow (x-d, y)$ Pour some water out of 4-gallon jug.
- 4) (x, y) if $y = d \& d > 0 \rightarrow (x, y-d)$ Pour some water out of 3-gallon jug.
- 5) (x, y) if $x > 0 \rightarrow (0, y)$ empty the 4-gallon water on the ground.
- 6) (x, y) if $y > 0 \rightarrow (x, 0)$ empty the 3-gallon water on the ground.

7) (x, y) , $y \leq x+y \leq 4$ and Pour water from 3-gallon

$y > 0 \rightarrow 4, (y-(x+y))$

jug into 4-gallon until jug is full.

8) (x, y) $y \geq x+y \geq 3$ and $x > 0 \rightarrow (x-(3-y), 3)$

Pour water from 4-gallon jug into 3-gallon jug until it is full.

9) (x, y) $y \leq x+y \leq 4$ and $y > 0 \rightarrow (x+y, 0)$

Pour all water of 3-gallon into 4-gallon.

10) (x, y) $y \leq x+y \leq 3$ and $x > 0 \rightarrow (0, x+y)$

Pour all water from 4-gallon to 3-gallon

11) $(0, 2) \rightarrow (2, 0)$ Pour 2-gallons water from 3-gallon jug to 4-gallon jug.

12) $(2, 4) \rightarrow (0, 4)$ empty the 2-gallon in the 4-gallon jug on the ground

8) water jug problem

X (4 gallon) Y (3 gallon) Rule applied

0	0	
4	0	
1	3	1
0	0	
2	1	10
4	1	
3	2	8
2	3	
1	4	
0	1	

Algorithm for breadth first search

- 1) Create a variable called NODE list and set it to initial state.

- 2) until a goal state is found or NODE list is empty to.

- 3.1 Remove first element from NODE list and call it E.

- 3.2 Node list empty quit.

- 3.3 For each way that each rule can match the state described in E do.

- 3.3.1 Apply the new rule to generate a new state.

- 3.3.2 If the new state is a goal state quit and return this state.

- 3.3.3 otherwise add new list to end of NODE list.

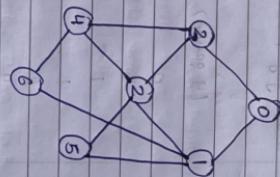
Time/Space complexity = $O(L^d)$

L = branching factor

Path is terminated :-

- Reached dead end
- Produces previous state
- Becomes larger than limit

Backtracking :-



29|9|22

DFS > Depth First Search

$$TC = O(b^d)$$

$$SC = O(d)$$

Begin

- 1) PUSH the starting node at stack pointed to by top.

Stack - top.

- 2) while stack is not empty do Begin
Pop stack to get stack top element.

If stack top element = goal
Return success & stop

Else
push the children of stack top

element in any order into stack

End while

End

If goal=5 ?



$$\text{No. of nodes by branching factor} = b^{d-1}$$

b → branching factor

Time complexity - d → O(n).

Time complexity of DFS → $O(b(b^d + 1))$

Time complexity of BFS → $O(b(b^d + 1))$

Space complexity = b^d

both

BFS for uninformed

techniques

both

BFS for informed

techniques

both

BFS for best-first

techniques

both

BFS for A*

techniques

both

BFS for uniformed

techniques

both

BFS for depth-limited

techniques

both

BFS for iterative-deepening

techniques

both

BFS for bidirectional

techniques

both

BFS for constraint-propagation

techniques

both

BFS for forward-chaining

techniques

both

BFS for backward-chaining

techniques

both

BFS for model-counting

techniques

both

BFS for planning

HILL CLIMBING →

BEGIN

1)

Identify possible starting states and measure the distance of
these distances with the goal node. Push them in a
stack acc. to ascending order of their distance.

2)

Repeat

Pop stack to get the stack top element. If the stack
of elements in the goal, announce it, and exit.

else

PUSH its children into the stack in the ascending order
of distance.
until stack empty

END

Pop climbing is simply a loop which continuously moves
in the direction of increasing value that is uphill.
It terminate when it reaches at the peak where
no node has a higher value.

The algorithm does not maintain a search tree.
so the next descent node data structure need only
record the state and its objective function.

Hill climbing does not look ahead beyond 1
and remains intertwined neighbors of the current state

Hill climbing is sometimes called Greedy local search
because it finds a good neighbor state without thinking
about where to go next.

Hill Climbing often makes very rapid progress towards
a solution because it is quite easy to improve into a
local state.

1)

There are some difficulties in hill climbing
but lower than global maxima, so it is very difficult for
greedy algorithm to navigate all nodes.

2)

plateau is an area of the landscape where evaluation
of the function is flat. It can be flat to local maxima from
which no uphill exists.

A hill climbing is not able to find to exit from plateau.

Ridge:-

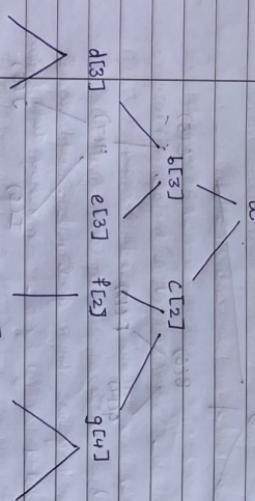
It is an area of the search space where is higher than the
corresponding areas and than itself as a slope but the
orientation of high regions compare to the set of available
nodes and direction in which they move make it impossible
to traverse. To overcome this node apply to a more two or
more nodes before applying the test.

Informed search or Heuristic search →

→ Hill climbing
→ Best First Search

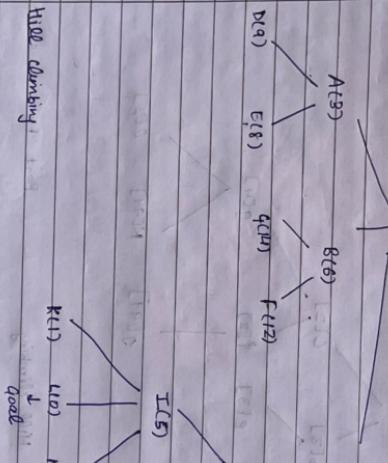
Breadth First Search
BFS - Queue [uninformed search]

when we blindly look for the data
in the provided data structure and stop at
its first occurrence.



Hill climbing
Best First Search

- Start at 'a'
- children of 'a' = $a[1(3), c(2)]$
- Best child $c[2]$ → Best First → 'c'
- children of $c \rightarrow f[j(99)]$
- Not goal → consider $[b(3), f(2), g(4)]$
- Backtrack to 'g' through 'c'
- children of 'g' → 'j[99]
- consider $[b(3), j(99), g(4)]$
- Best child → 'b(3)'
- Best 'b(3)' → goal state
- children of 'b' = $[f(3), e(3)]$
- consider $[d(3), e(3), j(99), g(4)]$
- children of $f \rightarrow j(99)$ → Best First → 'd(3)'
- children of 'd(3)' → 'h(4)'
- consider $[h(4), j(99), g(4)]$
- Best 'h(4)' → goal state



- 1) start at S
 - 2) children of S $[A(3), B(6)]$
 - 3) Best child $A(3)$
 - 4) children of $A(3) \rightarrow [D(9), E(8)]$
 - 5) consider $[D(9), E(8), B(6), C(5)]$
 - 6) Best first $C(5)$.
 - 7) children of $C \rightarrow H(7)$.
 - 8) children $[D(9), E(8), B(6), H(7)]$
 - 9) Best first $B \rightarrow (6)$
 - 10) children of $B \rightarrow [G(14), F(12)]$
 - 11) consider $[K(17), E(8), G(14), F(12), H(7)]$
 - 12) Best first \rightarrow First $\rightarrow H(7)$
 - 13) children of $H \rightarrow I(5), J(6)$
 - 14) consider $[B(6), E(8), G(14), F(12), I(5), J(6)]$
 - 15) Best First $\rightarrow I(5)$.
 - 16) children of $I \rightarrow K(17), L(10), M(2)$.
 - 17) consider $[D(9), E(8), G(14), F(12), K(17), L(10), M(2), J(6)]$
 - 18) Best First $\rightarrow L(10)$
- \hookrightarrow goal state

Hill Climbing

- 1) start at S
 - 2) children of $S [A(3), B(6), C(5)]$
 - 3) Best child $A(3)$
 - 4) children of $A [D(9), E(8)]$
 - 5) Best child E .
 - 6) $E \rightarrow$ [goal state]
 - 7) Backtrack to $C(5)$
 - 8) children of $C \rightarrow H(7)$
 - 9) children of $H \rightarrow [I(5), J(6)]$
 - 10) Best child $I(5)$
 - 11) children of $I(5) \rightarrow [K(17), L(10), M(2)]$
 - 12) Best child $L(10)$.
 - 13) $L \rightarrow$ goal state
- \hookrightarrow goal state

- 1) start at S
 - 2) children of S $[A(3), B(6)]$
 - 3) Best child $A(3)$
 - 4) children of $A(3) \rightarrow [D(9), E(8)]$
 - 5) consider $[D(9), E(8), B(6), C(5)]$
 - 6) Best first $C(5)$.
 - 7) children of $C \rightarrow H(7)$
 - 8) children $[D(9), E(8), B(6), H(7)]$
 - 9) Best first \rightarrow First $\rightarrow H(7)$
 - 10) children of $H \rightarrow I(5), J(6)$
 - 11) consider $[B(6), E(8), G(14), F(12), I(5), J(6)]$
 - 12) Best First $\rightarrow I(5)$.
 - 13) children of $I \rightarrow K(17), L(10), M(2)$.
 - 14) consider $[D(9), E(8), G(14), F(12), K(17), L(10), M(2), J(6)]$
 - 15) Best First $\rightarrow L(10)$
 - 16) children of $L(10) \rightarrow [P(11), Q(12), R(13)]$
 - 17) Best child $P(11)$.
 - 18) $P(11) \rightarrow$ goal state
- \hookrightarrow goal state

Algorithm for best first search

- 1) Put the initial node on a list OPEN.
- 2) If (OPEN is empty) OR (OPEN = GOAL) - terminate.
- 3) Remove best Node from OPEN and call this Node a.
- 4) If ($a = GOAL$) then terminate and success.
- 5) else if node has successors generate all of them find out how far they are from goal node. Sort all the children generated so you by the remaining dist from goal.
- 6) Name this list as CLOSED.
- 7) Replace it. open with closed.
- 8) Go to step 2.

6/10/22

UNIT-01

Example of Hill climbing

$$TC = O(b^d)$$

SC = dia max depth

	Initial state	Goal state	Intermediate state
A	H	H	
B	G	G	
C	F	F	
D	E	E	
E	D	D	
F	C	C	
G	B	B	
H	A	A	

(4+1=5)

Traces possible moves

A	B	C	D	E	F	G	H

$$6+2=8$$

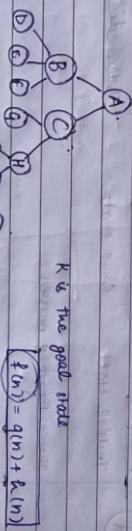
we are using two operators in the given example:-

- R₁ operator - pick up one block and put it on other.
- R₂ operator - pick up one block and put it on other.

- For the calculation of heuristic function add one point for every block which presented at proper position.
- Subtract one point for every block which is sitting on wrong ground position.

- Initially initial state have score of 4 and goal state have score of 8

* Search Algorithms



g(n) is the actual lowest ^{unit} cost of the path from start point to given node.

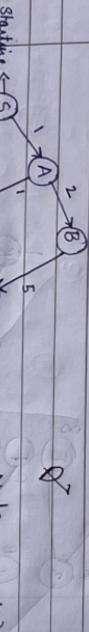
h(n) is the estimated cost of current path from given node to goal state.

Informed search \rightarrow goal state is given.

- 1) place starting Node in OPEN LIST.
 - 2) check if the OPEN LIST is empty or not.
If list is empty then return failure & stop.
 - 3) select If OPEN LIST is not empty then select node from the OPEN LIST which has smallest value of heuristic function ($g(n) + h(n)$) if node n is goal node then return success & stop.
 - 4) otherwise expand node n and generate all of the successors and put n into CLOSED LIST for each successor of n check whether n is already in open or closed list.
 - If not then compute function for n and place into OPEN LIST.
 - 5) else if node n is already in OPEN LIST then it should be attached to back pointer which reflect lower value of g(n) node.
- $f(n) = 5$
 $h(n) = 17$
- from (S \rightarrow to K)

$$f(n) = g(n) + h(n)$$

Advantages & Disadvantages of A* :-



- Adv.) 1) A* algorithm is optimal and complete.
2) This will help to solve tree complex problems.

Disad.) It ~~already~~ does not always produce shortest path sometimes it will not give shortest Path due to heuristic function evaluation.

- ① At S, open list [A, G]
② A is smallest than G
Move to A
③ Closed list

- ④ At A, open list [B, C]
⑤ C is smaller than B
Move to C

- ⑥ Close list, at C, open list

⑦ At C, open list

$$\begin{aligned} S \rightarrow A & \quad S \rightarrow C \\ 1+3=4 & \quad 1+0=1 \\ & \quad \text{A or C} \\ f(A) = 1+3 & = 4 \\ f(C) = 1+0 & = 1 \end{aligned}$$

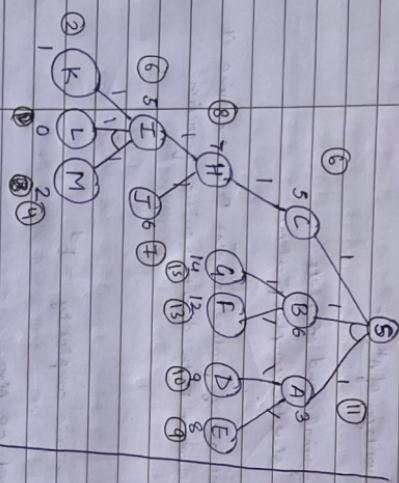
$$\begin{aligned} B \text{ or } C & \\ f(B) = 1+2 & = 3 \\ f(C) = 1+2 & = 3 \end{aligned}$$

~~B or C~~

~~f(C) = 1+2 = 3~~

~~B or C~~

Constraint satisfaction



m_{11}	m_{12}	m_{13}	$\rightarrow m_{11} + m_{12} + m_{13} = 15$
m_{21}	m_{22}	m_{23}	$\downarrow m_{12} + m_{22} + m_{32} = 15$
m_{31}	m_{32}	m_{33}	$\downarrow m_{13} + m_{23} + m_{33} = 15$

$$\rightarrow m_{11} + m_{22} + m_{33} = 15$$

$$\rightarrow m_{11} + m_{22} + m_{33} = 15$$

$$\rightarrow m_{11} + m_{22} + m_{33} = 15$$

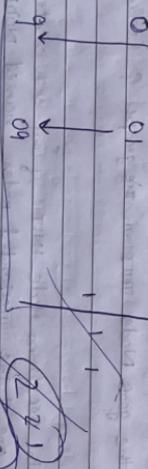
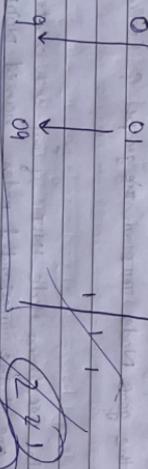
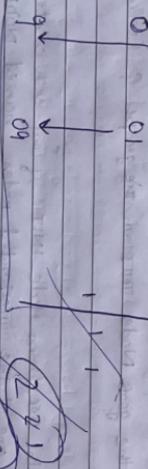
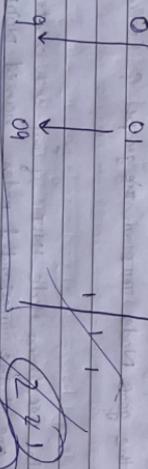
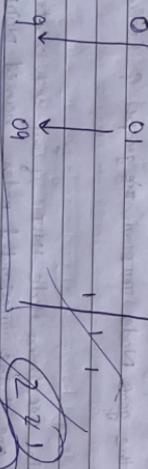
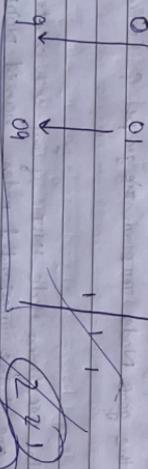
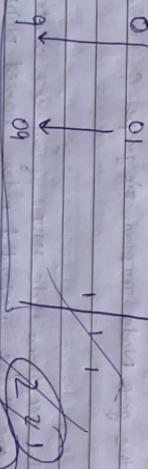
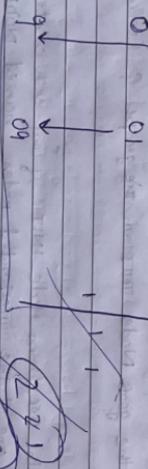
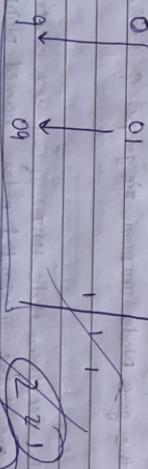
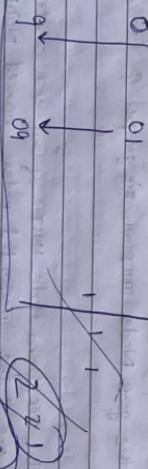
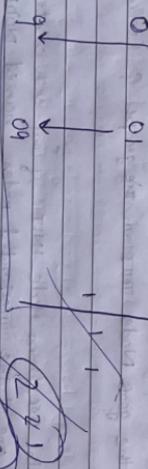
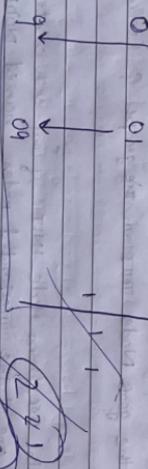
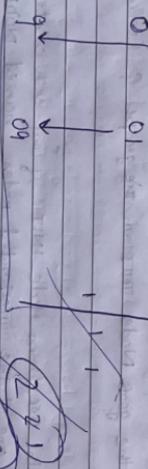
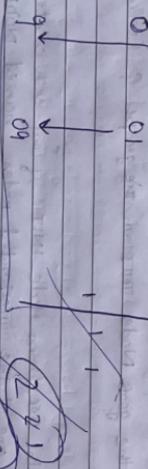
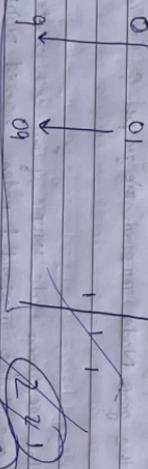
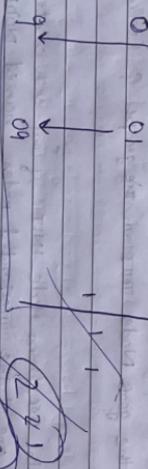
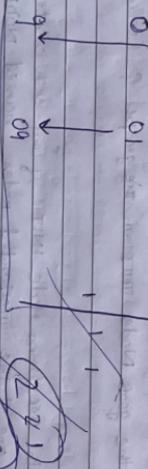
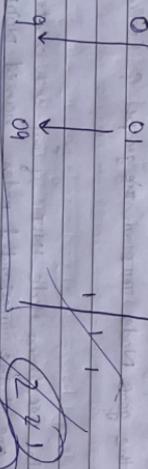
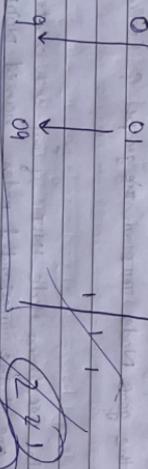
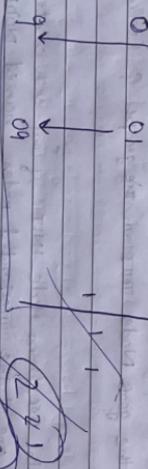
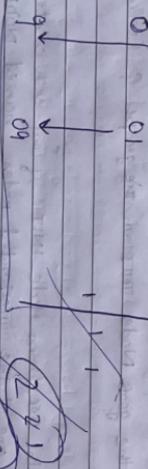
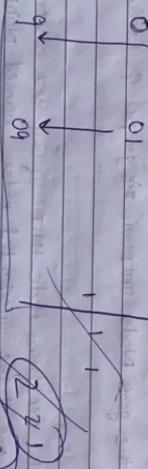
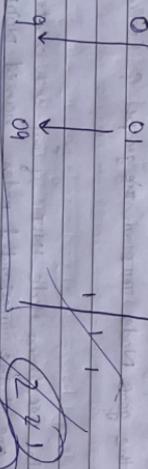
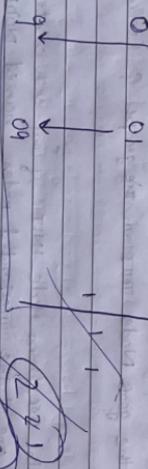
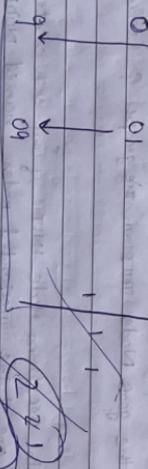
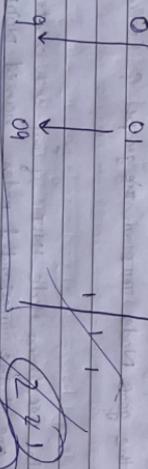
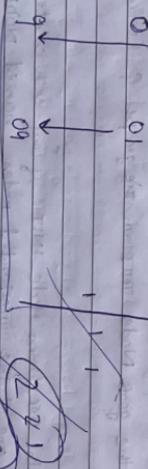
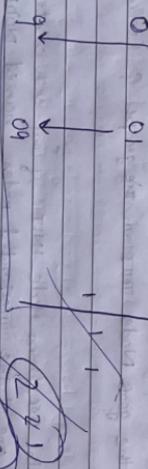
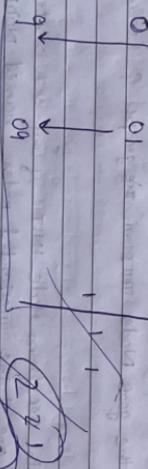
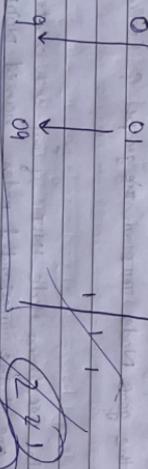
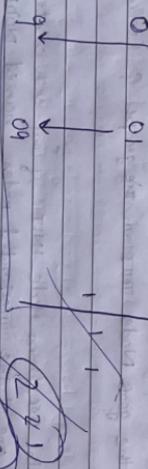
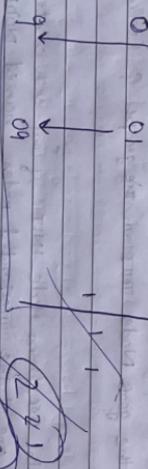
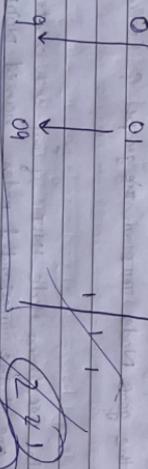
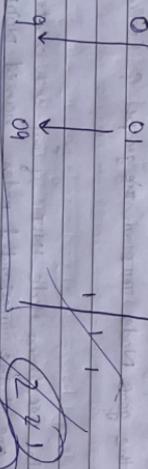
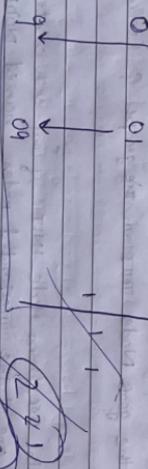
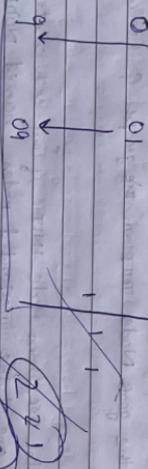
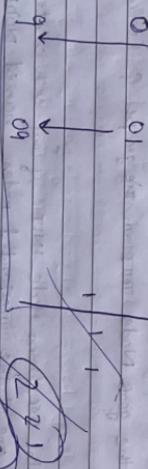
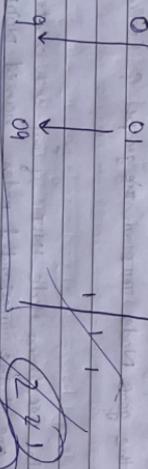
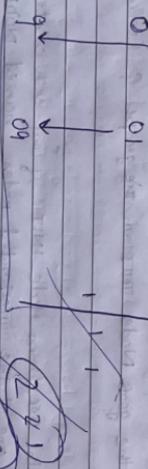
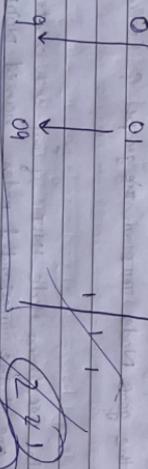
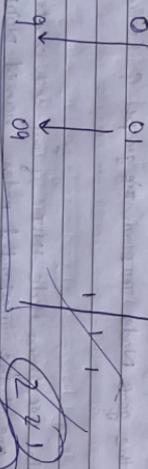
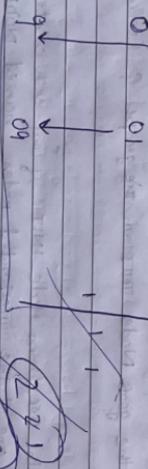
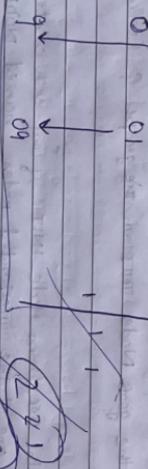
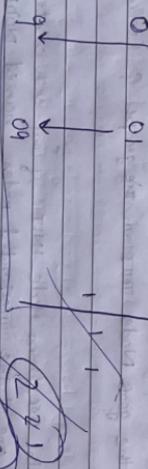
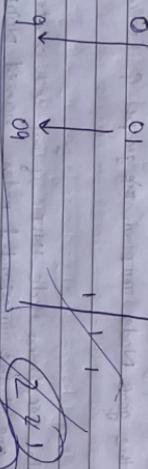
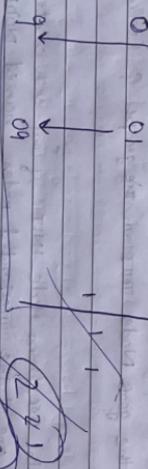
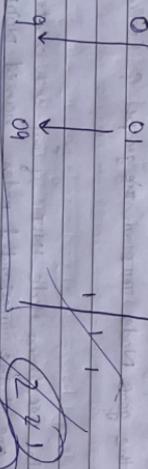
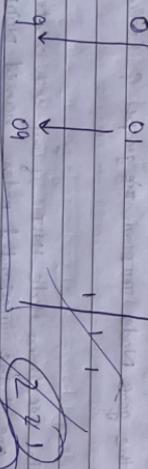
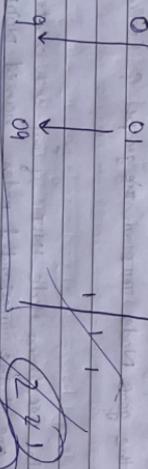
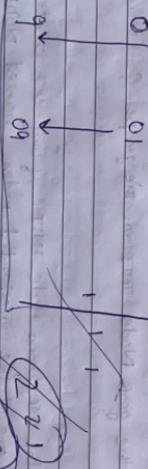
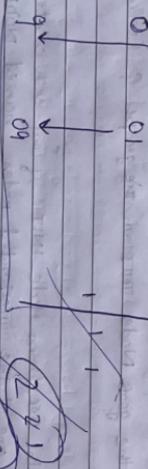
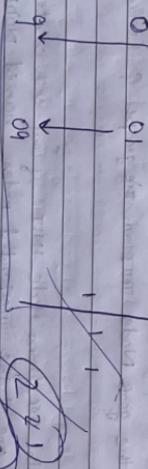
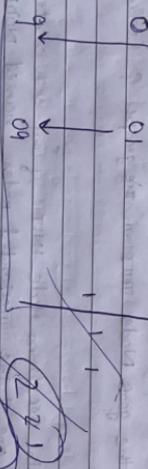
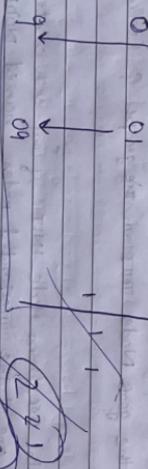
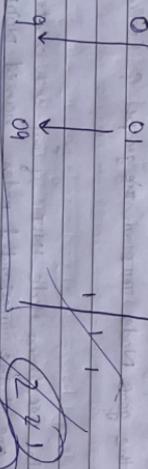
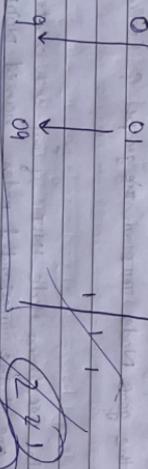
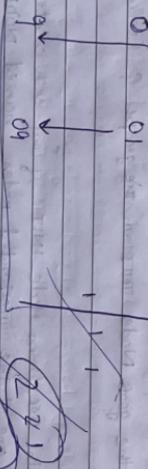
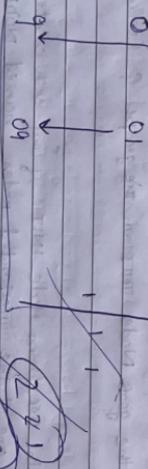
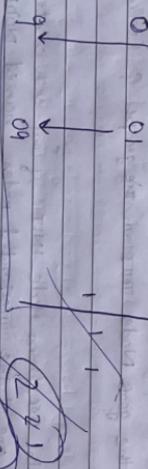
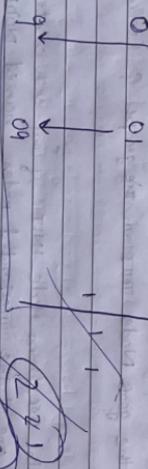
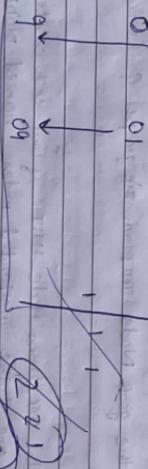
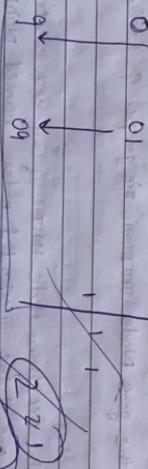
Magic Square

m_{11}	m_{12}	m_{13}
m_{21}	m_{22}	m_{23}

$$\rightarrow m_{11} + m_{22} + m_{33} = 15$$

$$\rightarrow m_{11} + m_{22} + m_{33} = 15$$

$$\rightarrow m_{11} + m_{22} + m_{33} = 15$$



← algorithm →

→ 1)

Propagate available constraints. first SET OPEN to set all object which must have value assigned to them in a complete solution.

Then do until an inconsistency is detected or until OPEN is empty.

1.1)

select an object from OPEN & strengthen as much as possible w.r.t. the set of constraints which apply on object.

1.2)

To this set is different from the set which was assigned the last time to object was examined then add to OPEN all object which share any constraint with object.

2)

If the union of the constraints discovered by above solution is defined solution then quit.

3)

If the union of constraints discovered by above solution is contradiction then return failure.

4)

If neither of above action occur then it is necessary to make a guess at domaining in order to proceed.

To do this repeat until a solution is found or all possible solution.

be careful about the boundary condition
using which

Statement of constraints

S E N D
M O R E

1) Values are to be assigned from

M O N E Y

D to 9.

No two letters should assign same value.

2) To some letter occurs more than once it must be assigned some value.

3) The sum of digits must be acc. to given problem.

M=1

S = 8 or 9

O = 0 or 1

N = E or E+1

C₂=1

N+R > 8

E < 9

C₁ = 0
C₁ = 1
E = 2

N = 3

R = 8 or 9

2 + D = 4

2 + D = 10 + Y

D = 8 or 9

D = 8
Y = 0

N = C

← Mean end Analysis →

Till here, the searching strategies only search in one direction either forward or backward.

Mean end analysis uses bi-directional search.

It means in this technique part of the problems are independently solved using forward and backward search simultaneously then the sub-problems solutions are joined to obtain complete solution.

It depicts diff b/w current state and goal state and once such a diff is identified an operator that can reduce the difference must be found.

We apply backward chaining in which operators are selected then sub-goals are setup to establish pre-condition of the operators.

This is called operator sub-volume.

There are certain typical condition on which these subgoals or operators can be applied. These are called pre-condition. The newly generated states after rules are applied are called post condition.

The table that keeps track of the difference between current state and goal state is called difference table.

example :-
GPS
STRIPS

Steps inMEA

- 1) Involves detection of difference b/w current and goal state.
- 2) Once difference identified then find an operator for reducing difference.
- 3) Some time operator cannot applied directly to current state.
- 4) Sub - problem of getting to state where operators applied.
- 5) Operator may not give result in goal state.
- 6) 2nd sub - problem of getting new state from goal state.
- 7) MEA process applied recursively.

a) Three primary goals of mea

