

**VISVESVARAYA TECHNOLOGICAL
UNIVERSITY**

"JnanaSangama", Belgaum -590014, Karnataka.



**LAB REPORT
on**

Artificial Intelligence (23CS5PCAIN)

Submitted by

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in partial fulfillment for the award of the degree of
BACHELOR OF ENGINEERING
in
COMPUTER SCIENCE AND ENGINEERING



B.M.S. COLLEGE OF ENGINEERING
(Autonomous Institution under VTU)
BENGALURU-560019
Sep-2024 to Jan-2025

**B.M.S. College of Engineering,
Bull Temple Road, Bangalore 560019**
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Department of Computer Science and Engineering



CERTIFICATE

This is to certify that the Lab work entitled “Artificial Intelligence (23CS5PCAIN)” carried out by **Santosh H Jambagi (1BM22CS244)**, who is bonafide student of **B.M.S. College of Engineering**. It is in partial fulfillment for the award of **Bachelor of Engineering in Computer Science and Engineering** of the Visvesvaraya Technological University, Belgaum. The Lab report has been approved as it satisfies the academic requirements in respect of an Artificial Intelligence (23CS5PCAIN) work prescribed for the said degree.

Lab faculty Incharge Name Assistant Professor Department of CSE, BMSCE	Dr. Joythi S Nayak Professor & HOD Department of CSE, BMSCE
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Github Link:

<https://github.com/SantoshJambagi2004/Artificial-Intelligence-Lab>

Program 1

Implement Tic –Tac –Toe Game
Implement vacuum cleaner agent

Algorithm:

7 Tic-Tac-Toe

a) Algorithm: Tic-Tac-Toe()

step 1: Create a 3×3 3-dimensional array filled with underscores ('_') to indicate empty spaces.

step 2: Set current player to 'X'

Validate the input: if $\text{row} < 0$ or $\text{row} > 2$ or $\text{board}[\text{i}][\text{j}] == '_'$

b) Check for winner

check all rows, columns and diagonals to see if any player having any three symbols in line.

c) Check for Draw

If no winner is found, check if the board is full.

d) Switch players

If the game is still ongoing switch current player from 'X' to 'O' or from 'O' to 'X'.

e) Update the board

place the current player's symbol ('X' or 'O') in the selected cell.

f) Display the board

print the board along with its current state whether winner is found or not. Or its a draw.

11/10/24

LAB - 1

→ Implement Vacuum World cleaner

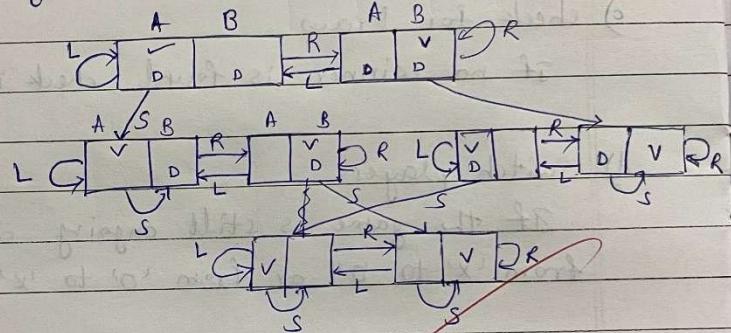
function REFLEX_VACUUM_AGENT ([location, status]) returns
an exact action.

If status = Dirty then return Suck

else If location = A then return Right

else If location = B then return Left

state-space diagram of vacuum world!



Algorithm:

Initial Environment:

- Define grid with cells that are either dirty(1) or clean(0).
- set the initial position of the vacuum cleaner.
- set the goal state : 'A': 0, 'B': 0

The vacuum cleaner can perform the following actions:

1) clean: clean the current cell.

2) Move left: Move to the left cell if it exists

3) Move right: Move to the right cell if it exists.

Initial state : [1, 1] - BAJ

- If the initial state matches the goal state then exit the function.
- Increment the cost for the one suck operation.
- goal state = {'A': 0, 'B': 0}

E	S	I
0	2	F
0	2	E

02
1101m

E	S	I
0	H	O
0	E	F

E	S	I
0	H	F
0	E	F

E	S	I
0	H	F
0	E	O

E	S	I
0	H	I
0	E	F

R

A

U

A

E	S	I
0	H	F
0	E	F

E	S	I
0	H	F
0	E	O

E	S	I
0	H	I
0	E	F

A

V

A

A

E	S	I
0	H	F
0	E	F

E	S	I
0	H	F
0	E	O

E	S	I
0	H	I
0	E	F

A

V

A

A

S

S

S

S

Code: 1: Tic - Tac - Toe

```
import numpy as np

board=np.array([['-','-','-'],['-','-','-'],['-','-','-']])

current_player='X'
flag=0

def check_win():
    for i in range(3):
        if board[i][0] == board[i][1] == board[i][2] != '-':
            return True
    for i in range(3):
        if board[0][i] == board[1][i] == board[2][i] != '-':
            return True
    if board[0][0] == board[1][1] == board[2][2] != '-':
        return True
    if board[0][2] == board[1][1] == board[2][0] != '-':
        return True
    return False

def tic_tac_toe():
    n=0
    print(board)
    while n<9:
        if n%2==0:
            current_player='X'
        else :
            current_player='O'

        row = int(input("Enter row: "))
        col = int(input("Enter column: "))

        if(board[row][col]=='-'):
            board[row][col]=current_player
            print(board)
            flag=check_win();
            if flag==1:
                print(current_player+' wins')
                break
            else:
                n=n+1
        else :
            print("Invalid Position")

    if n==9:
```

```
print("Draw")  
tic_tac_toe()
```

Output:

Win :

```
→ - | - | -  
- | - | -  
- | - | -  
Player X, enter the row (0, 1, 2): 0  
Player X, enter the column (0, 1, 2): 0  
X | - | -  
- | - | -  
- | - | -  
Player O, enter the row (0, 1, 2): 1  
Player O, enter the column (0, 1, 2): 0  
X | - | -  
O | - | -  
- | - | -  
Player X, enter the row (0, 1, 2): 0  
Player X, enter the column (0, 1, 2): 1  
X | X | -  
O | - | -  
- | - | -  
Player O, enter the row (0, 1, 2): 1  
Player O, enter the column (0, 1, 2): 1  
X | X | -  
O | O | -  
- | - | -  
Player X, enter the row (0, 1, 2): 0  
Player X, enter the column (0, 1, 2): 2  
X | X | X  
O | O | -  
- | - | -  
Player X wins!
```

Draw :

```

→ Player 0, enter the row (0, 1, 2): 0
Player 0, enter the column (0, 1, 2): 2
0 | X | 0
X | - | -
- | - | -
Player X, enter the row (0, 1, 2): 2
Player X, enter the column (0, 1, 2): 0
0 | X | 0
X | - | -
X | - | -
Player 0, enter the row (0, 1, 2): 1
Player 0, enter the column (0, 1, 2): 1
0 | X | 0
X | 0 | -
X | - | -
Player X, enter the row (0, 1, 2): 1
Player X, enter the column (0, 1, 2): 2
0 | X | 0
X | 0 | X
X | - | -
Player 0, enter the row (0, 1, 2): 2
Player 0, enter the column (0, 1, 2): 1
0 | X | 0
X | 0 | X
X | 0 | -
Player X, enter the row (0, 1, 2): 2
Player X, enter the column (0, 1, 2): 2
0 | X | 0
X | 0 | X
X | 0 | X
It's a draw!

```

2. Vacuum Cleaner :

```

cost =0
def vacuum_world(state, location):
    global cost
    if(state['A']==0 and state['B']==0):
        print('All rooms are clean')
        return

    if state[location]==1:
        state[location]=0
        cost+=1

```

```

state[location]=(int(input('Is room '+ str(location) +' still dirty :')))

if state[location]==1:
    return vacuum_world(state, location)
else:
    print('Room ' + str(location) + ' cleaned')

next_location='B' if location=='A' else 'A'
if state[next_location]==0:
    state[next_location]=(int(input('Is room '+ str(next_location) +' dirty :')))

print('Moving to room '+str(next_location))
return vacuum_world(state, next_location)

state={}
state['A']=int(input('Enter status of room A :'))
state['B']=int(input('Enter status of room B :'))
location=input('Enter initial location of vacuum (A/B) :')
vacuum_world(state,location)
print("Status = "+str(state))
print('Total cost: ' + str(cost))

```

Output :

```

Enter status of room A : 1
Enter status of room B : 1
Enter initial location of vacuum (A/B) : A
Is room A still dirty : 0
Room A cleaned
Moving to room B
Is room B still dirty : 0
Room B cleaned
Is room A dirty : 0
Moving to room A
All rooms are clean
Status = {'A': 0, 'B': 0}
Total cost: 2

```

Program 2

Implement 8 puzzle problems using Depth First Search (DFS)

Implement Iterative deepening search algorithm

Algorithm:

LAB - 2

8 puzzle problem using BFS & DFS

Initial Setup:

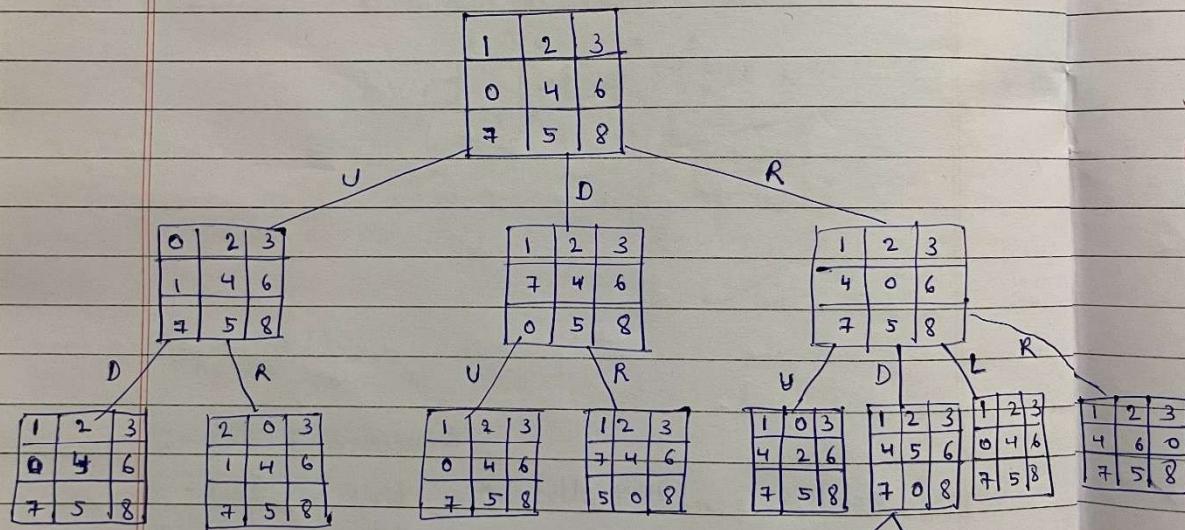
→ Create a 3x3 grid.

→ Represent each state with a tuple containing
+ numbers between (0 to 8)

Goal State:

→ Set a Goal state =

1	2	3
4	5	6
7	8	0



Fin.

Gen.

BFS

Now
10

Date: _____ Page: _____

Finding empty tile: $(a)d + (a)g = (a)$

Find the orgin. i.e. (a) Blank tile (a) \neq zero

Generating possible moves

For each possible direction (up, down, left, right)

- check if the move is valid
- if valid, create a new state by swapping the blank and record the move.

BFS Loop:

while the queue is not empty:

- Dequeue the current state and its path.
- check if the current state matches goal state

2	3
6	0
5	8

22/10/24

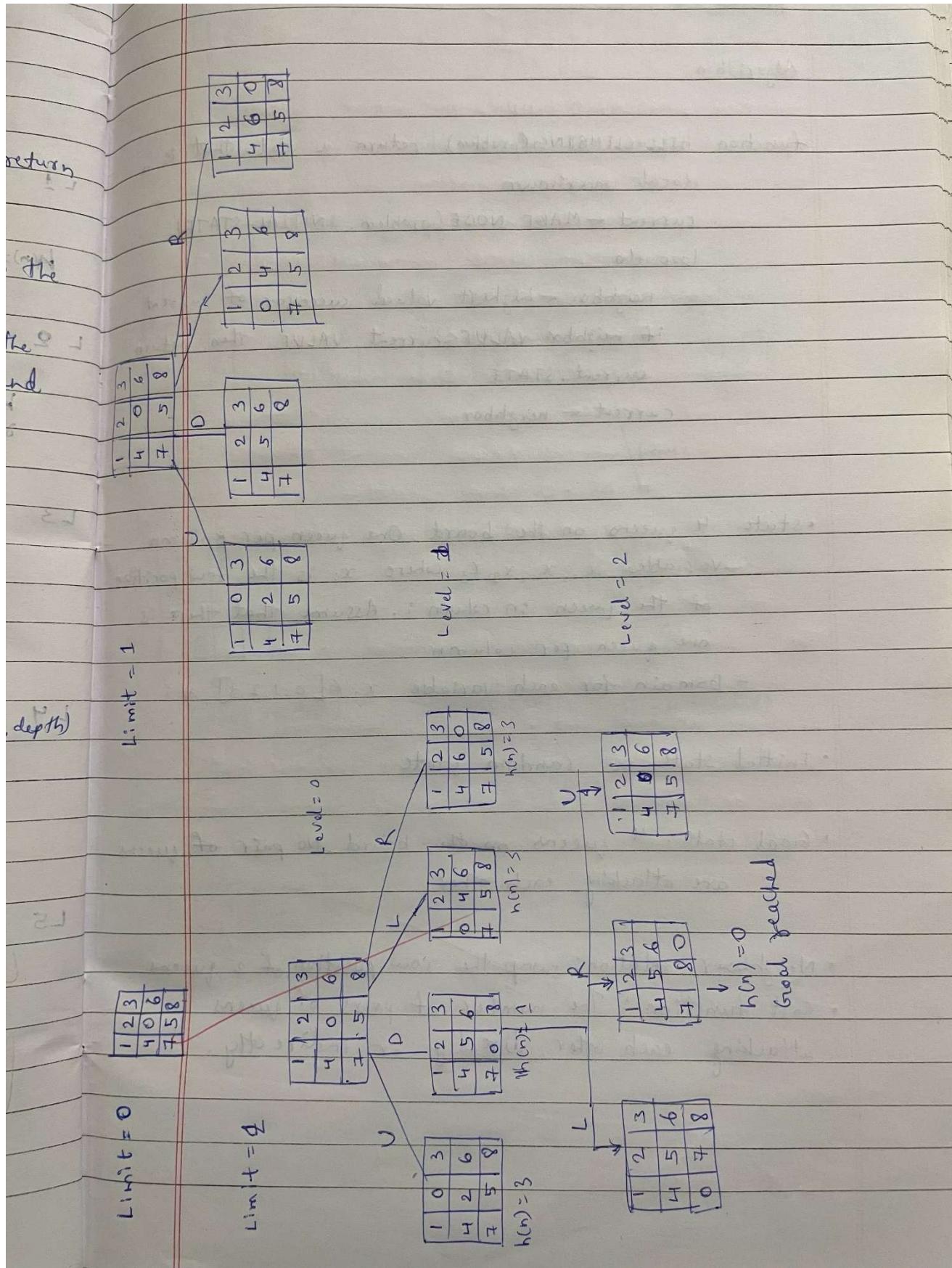
Lab program Implement Iterative Deepening Search Algorithm

- 1) For each child of the current node
 - 2) If it's the current target node return
 - 3) If the current maximum depth is reached, return
 - 4) Set the current node to this node and go back to 1
- 5) After having gone through all children, go to the next child of the parent (the next sibling)
 - 6) After having gone through all children of the start node increase the maximum depth and go back to 1
 - 7) If we have reached all leaf (bottom nodes), the goal node doesn't exist.

```
function ITERATIVE-DEEPENING-SEARCH(problem)
    return a solution or failure
    for depth=0 to << do
        result ← DEPTH-LIMITED-SEARCH(problem, depth)
        if result ≠ cut-off then return result
```

Goal state:

1	2	3
4	5	6
7	8	0



Code:

1: DFS

```
cnt = 0;
def print_state(in_array):
    global cnt
    cnt += 1
    for row in in_array:
        print(' '.join(str(num) for num in row))
    print()

def helper(goal, in_array, row, col, vis):

    vis[row][col] = 1
    drow = [-1, 0, 1, 0]
    dcol = [0, 1, 0, -1]
    dchange = ['U', 'R', 'D', 'L']

    print("Current state:")
    print_state(in_array)

    if in_array == goal:
        print_state(in_array)
        print(f"Number of states : {cnt}")
        return True

    for i in range(4):
        nrow = row + drow[i]
        ncol = col + dcol[i]

        if 0 <= nrow < len(in_array) and 0 <= ncol < len(in_array[0]) and not vis[nrow][ncol]:
            print(f"Took a {dchange[i]} move")
            in_array[row][col], in_array[nrow][ncol] = in_array[nrow][ncol], in_array[row][col]

            if helper(goal, in_array, nrow, ncol, vis):
                return True

            in_array[row][col], in_array[nrow][ncol] = in_array[nrow][ncol], in_array[row][col]

    vis[row][col] = 0
    return False

iniθal_state = [[1, 2, 3], [0, 4, 6], [7, 5, 8]]
```

```
goal_state = [[1, 2, 3], [4, 5, 6], [7, 8, 0]]
visited = [[0] * 3 for _ in range(3)]
empty_row, empty_col = 1, 0
found_soluθon = helper(goal_state, iniθal_state, empty_row, empty_col, visited)
print("Soluθon found:", found_soluθon)
```

Output :

Current state: 1 2 3 0 4 6 7 5 8	Took a L move Current state: 2 3 6 1 4 8 0 7 5	Took a L move Current state: 1 0 2 4 6 3 7 5 8
Took a U move Current state: 0 2 3 1 4 6 7 5 8	Took a L move Current state: 2 3 6 1 0 4 7 5 8	Took a L move Current state: 0 1 2 4 6 3 7 5 8
Took a R move Current state: 2 0 3 1 4 6 7 5 8	Took a D move Current state: 2 3 6 1 5 4 7 0 8	Took a D move Current state: 1 2 3 4 6 8 7 5 0
Took a R move Current state: 2 3 0 1 4 6 7 5 8	Took a R move Current state: 2 3 6 1 5 4 7 8 0	Took a L move Current state: 1 2 3 4 6 8 7 0 5
Took a D move Current state: 2 3 6 1 4 0 7 5 8	Took a L move Current state: 2 3 6 1 5 4 0 7 8	Took a L move Current state: 1 2 3 4 6 8 0 7 5
Took a D move Current state: 2 3 6 1 4 8 7 5 0	Took a D move Current state: 2 4 3 1 0 6 7 5 8	Took a D move Current state: 1 2 3 4 5 6 7 0 8
Took a L move Current state: 2 3 6 1 4 8 7 0 5	Took a R move Current state: 2 4 3 1 6 0 7 5 8	Took a R move Current state: 1 2 3 4 5 6 7 8 0
Took a U move Current state: 2 3 6 1 0 8 7 4 5	Took a U move Current state: 2 4 0 1 6 3 7 5 8	Number of states : 42 Solution found: True

2 : Iterative deepening search

```
class PuzzleState:  
    def __init__(self, board, empty_tile_pos, depth=0, path=[]):  
        self.board = board  
        self.empty_tile_pos = empty_tile_pos # (row, col)  
        self.depth = depth  
        self.path = path # Keep track of the path taken to reach this state  
  
    def is_goal(self, goal):  
        return self.board == goal  
  
    def generate_moves(self):  
        row, col = self.empty_tile_pos  
        moves = []  
        directions = [(-1, 0, 'Up'), (1, 0, 'Down'), (0, -1, 'Left'), (0, 1, 'Right')] # up, down, left, right  
        for dr, dc, move_name in directions:  
            new_row, new_col = row + dr, col + dc  
            if 0 <= new_row < 3 and 0 <= new_col < 3:  
                new_board = self.board[:]  
                new_board[row * 3 + col], new_board[new_row * 3 + new_col] = new_board[new_row *  
3 + new_col], new_board[row * 3 + col]  
                new_path = self.path + [move_name] # Update the path with the new move  
                moves.append(PuzzleState(new_board, (new_row, new_col), self.depth + 1, new_path))  
        return moves  
  
    def display(self):  
        # Display the board in a matrix form  
        for i in range(0, 9, 3):  
            print(self.board[i:i + 3])  
        print(f"Moves: {self.path}") # Display the moves taken to reach this state  
        print() # Newline for better readability  
  
def iddfs(initial_state, goal, max_depth):  
    for depth in range(max_depth + 1):  
        print(f"Searching at depth: {depth}")  
        found = dls(initial_state, goal, depth)  
        if found:  
            print(f"Goal found at depth: {found.depth}")  
            found.display()  
            return found  
    print("Goal not found within max depth.")  
    return None  
  
def dls(state, goal, depth):  
    if state.is_goal(goal):  
        return state
```

```

if depth <= 0:
    return None

for move in state.generate_moves():
    print("Current state:")
    move.display() # Display the current state
    result = dls(move, goal, depth - 1)
    if result is not None:
        return result
return None

def main():
    # User input for initial state, goal state, and maximum depth
    initial_state_input = input("Enter initial state (0 for empty tile, space-separated, e.g. '1 2 3 4 5 6 7 8 0'): ")
    goal_state_input = input("Enter goal state (0 for empty tile, space-separated, e.g. '1 2 3 4 5 6 7 8 0'): ")
    max_depth = int(input("Enter maximum depth: "))

    initial_board = list(map(int, initial_state_input.split()))
    goal_board = list(map(int, goal_state_input.split()))
    empty_tile_pos = initial_board.index(0) // 3, initial_board.index(0) % 3 # Calculate the position of
    the empty tile

    initial_state = PuzzleState(initial_board, empty_tile_pos)

    solution = iddfs(initial_state, goal_board, max_depth)

if __name__ == "__main__":
    main()

```

Output :

```
Enter initial state (0 for empty tile, space-separated, e.g. '1 2 3 4 5 6 7 8 0'): 1 2 3 0 4 6 7 5 8
Enter goal state (0 for empty tile, space-separated, e.g. '1 2 3 4 5 6 7 8 0'): 1 2 3 4 5 6 7 8 0
Enter maximum depth: 2
Searching at depth: 0
Searching at depth: 1

Current state: [0, 2, 3]
[1, 4, 6]
[7, 5, 8]
Moves: ['Up']

Current state: [1, 2, 3]
[7, 4, 6]
[0, 5, 8]
Moves: ['Down']

Current state: [1, 2, 3]
[7, 4, 6]
[0, 5, 8]
Moves: ['Down', 'Up']

Current state: [1, 2, 3]
[7, 4, 6]
[5, 0, 8]
Moves: ['Down', 'Right']

Current state: [1, 2, 3]
[4, 0, 6]
[7, 5, 8]
Moves: ['Right']

Searching at depth: 2
Current state: [0, 2, 3]
[1, 4, 6]
[7, 5, 8]
Moves: ['Up']

Current state: [1, 0, 3]
[4, 2, 6]
[7, 5, 8]
Moves: ['Right', 'Up']

Current state: [1, 2, 3]
[4, 5, 6]
[7, 0, 8]
Moves: ['Right', 'Down']

Current state: [1, 2, 3]
[0, 4, 6]
[7, 5, 8]
Moves: ['Right', 'Left']

Current state: [1, 2, 3]
[4, 6, 0]
[7, 5, 8]
Moves: ['Right', 'Right']

Goal not found within max depth.
```

Program 3

Implement A* search algorithm

Algorithm:

- g) For 8 puzzle problem using A* implementation
 to calculate $f(n)$ using
- $g(n)$ = depth of a node
 - $h(n)$ = heuristic value \rightarrow no. of misplaced tiles
- $$f(n) = g(n) + h(n)$$
- step
- b) $g(n)$ = depth
 $h(n)$ = heuristic value
 \downarrow
 Manhattan distance
- $$f(n) = g(n) + h(n)$$
- step

Algorithm for A* search:

step 1: place the starting node in the OPEN list.

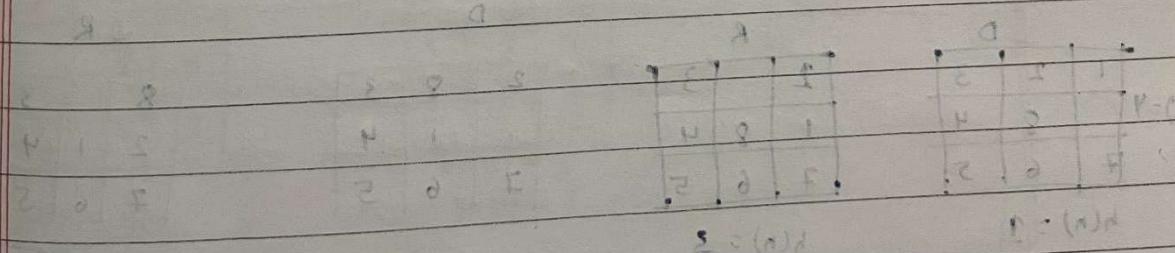
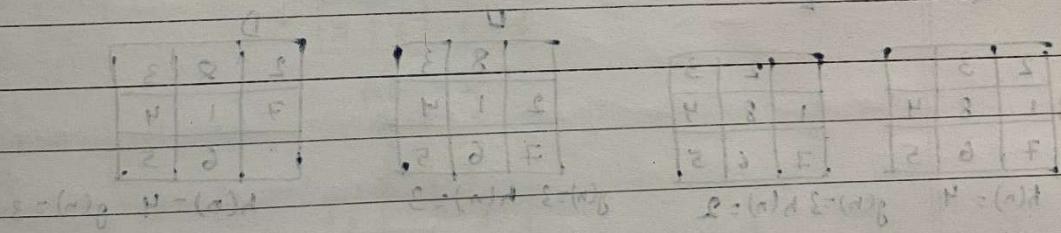
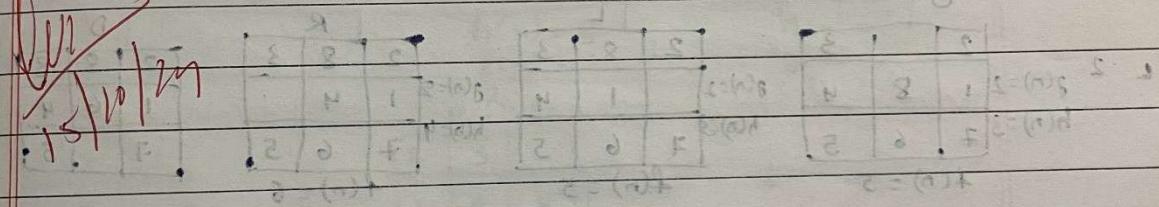
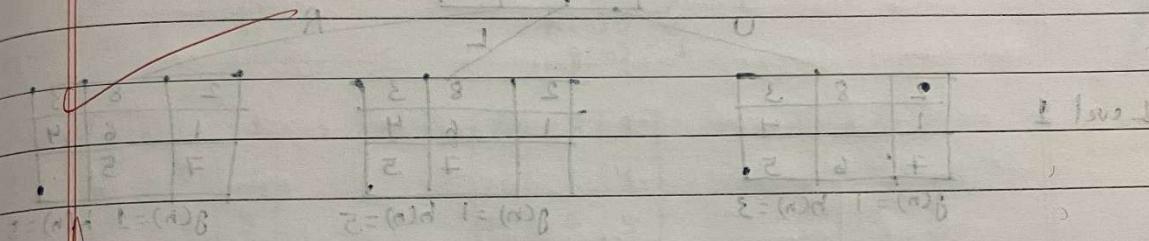
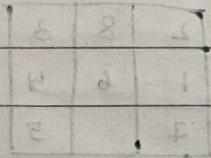
step 2: check if the OPEN list is empty or not, if the list is empty then return failure and stop.

step 3: Select the node from the open list which has the smallest value of evaluation function ($f+g$), if node n is goal node then return success and stop. otherwise

Step 4: Expand node n and generate all its successors and put n into the closed list. For each successor ' n' , check whether ' n ' is already in the OPEN or CLOSED list, if not then compute evaluation function for ' n ' and place into OPEN list.

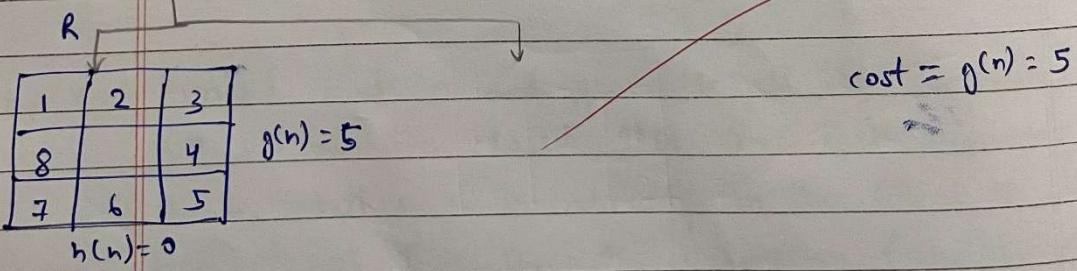
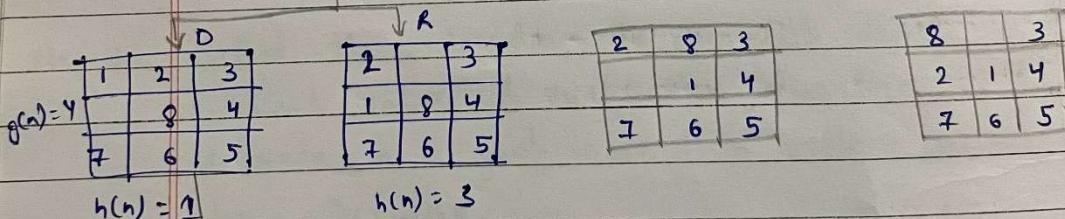
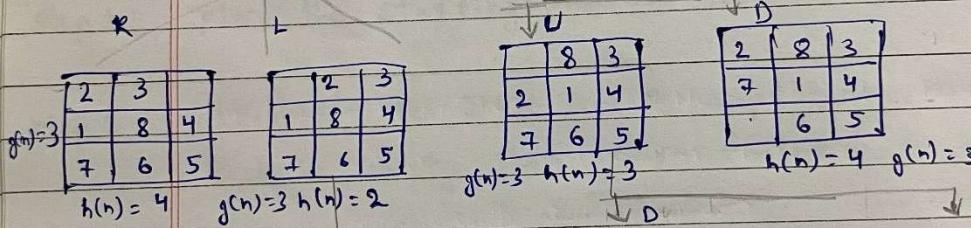
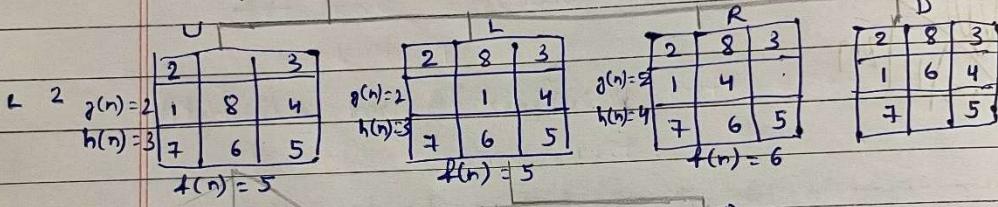
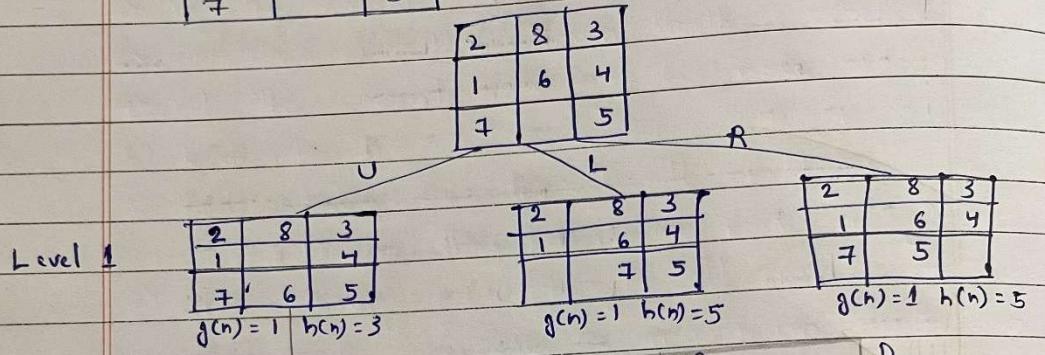
Step 5: Else if node 'n' is already in OPEN and CLOSED, then
it should be attached to the back pointer which
reflects the lowest $j(n)$ value.

Step 6: Return to step 2.

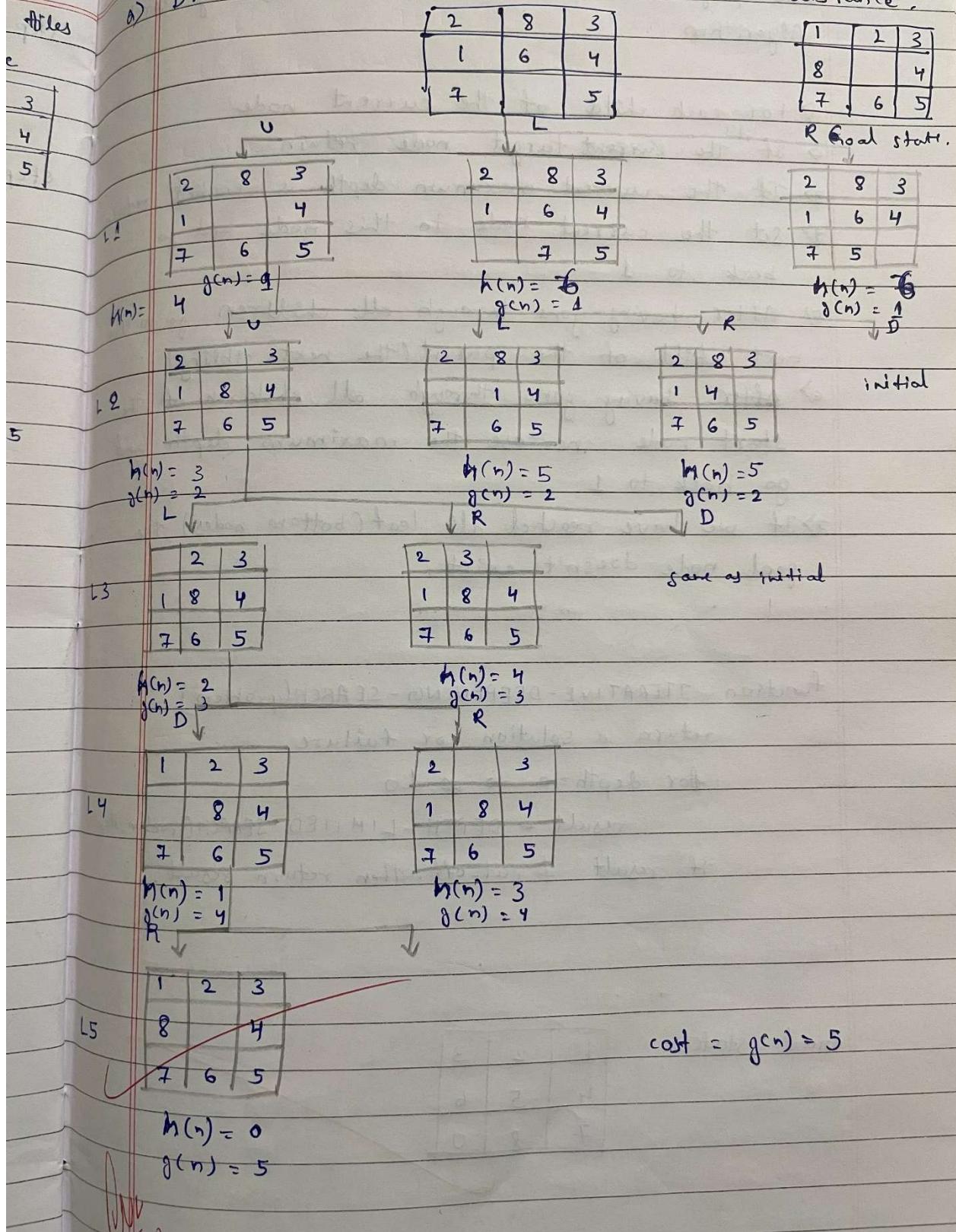


Q) Draw the state space diagram for misplaced tiles

Initial state	Goal state
$\begin{array}{ c c c } \hline 2 & 8 & 3 \\ \hline 1 & 6 & 4 \\ \hline 7 & & 5 \\ \hline \end{array}$	$\begin{array}{ c c c } \hline 1 & 2 & 3 \\ \hline 8 & & 4 \\ \hline 7 & 6 & 5 \\ \hline \end{array}$



a) Draw the state space tree for Manhattan distance.



Code:

Misplaced Tiles :

```
class Node:  
    def __init__(self, state, parent=None, move=None, cost=0):  
        self.state = state  
        self.parent = parent  
        self.move = move  
        self.cost = cost  
  
    def heuristic(self):  
        goal_state = [[1,2,3], [8,0,4], [7,6,5]]  
        count = 0  
        for i in range(len(self.state)):  
            for j in range(len(self.state[i])):  
                if self.state[i][j] != 0 and self.state[i][j] != goal_state[i][j]:  
                    count += 1  
        return count  
  
    def get_blank_position(state):  
        for i in range(len(state)):  
            for j in range(len(state[i])):  
                if state[i][j] == 0:  
                    return i, j  
  
    def get_possible_moves(position):  
        x, y = position  
        moves = []  
        if x > 0: moves.append((x - 1, y, 'Down'))  
        if x < 2: moves.append((x + 1, y, 'Up'))  
        if y > 0: moves.append((x, y - 1, 'Right'))  
        if y < 2: moves.append((x, y + 1, 'Left'))  
        return moves  
  
    def generate_new_state(state, blank_pos, new_blank_pos):  
        new_state = [row[:] for row in state]  
        new_state[blank_pos[0]][blank_pos[1]], new_state[new_blank_pos[0]][new_blank_pos[1]] = \  
            new_state[new_blank_pos[0]][new_blank_pos[1]], new_state[blank_pos[0]][blank_pos[1]]  
        return new_state  
  
def a_star_search(initial_state):  
    open_list = []  
    closed_list = set()  
  
    initial_node = Node(state=initial_state, cost=0)  
    open_list.append(initial_node)
```

```

while open_list:

    open_list.sort(key=lambda node: node.cost + node.heuristic())
    current_node = open_list.pop(0)

    move_description = current_node.move if current_node.move else "Start"
    print("Current state:")
    for row in current_node.state:
        print(row)
    print(f"Move: {move_description}")
    print(f"Heuristic value (misplaced tiles): {current_node.heuristic()}")
    print(f"Cost to reach this node: {current_node.cost}\n")

    if current_node.heuristic() == 0:

        path = []
        while current_node:
            path.append(current_node)
            current_node = current_node.parent
        return path[::-1]
        closed_list.add(tuple(map(tuple, current_node.state)))

        blank_pos = get_blank_position(current_node.state)
        for new_blank_pos in get_possible_moves(blank_pos):
            new_state = generate_new_state(current_node.state, blank_pos, (new_blank_pos[0],
            new_blank_pos[1]))

            if tuple(map(tuple, new_state)) in closed_list:
                continue

            cost = current_node.cost + 1
            move_direction = new_blank_pos[2]
            new_node = Node(state=new_state, parent=current_node, move=move_direction, cost=cost)

            if new_node not in open_list:
                open_list.append(new_node)

return None

initial_state = [[2,8,3], [1,6,4], [7,0,5]]
solution_path = a_star_search(initial_state)

if solution_path:
    print("Solution path:")
    for step in solution_path:

```

```

for row in step.state:
    print(row)
print()
else:
    print("No solution found.")

```

Output :

```

Current state:
[2, 8, 3]
[1, 6, 4]
[7, 0, 5]
Move: Start
Heuristic value (misplaced tiles): 4
Cost to reach this node: 0

Current state:
[2, 8, 3]
[1, 0, 4]
[7, 6, 5]
Move: Down
Heuristic value (misplaced tiles): 3
Cost to reach this node: 1

Current state:
[2, 0, 3]
[1, 8, 4]
[7, 6, 5]
Move: Down
Heuristic value (misplaced tiles): 3
Cost to reach this node: 2

Current state:
[2, 8, 3]
[0, 1, 4]
[7, 6, 5]
Move: Right
Heuristic value (misplaced tiles): 3
Cost to reach this node: 2

Current state:
[0, 2, 3]
[1, 8, 4]
[7, 6, 5]
Move: Right
Heuristic value (misplaced tiles): 2
Cost to reach this node: 3

Current state:
[1, 2, 3]
[0, 8, 4]
[7, 6, 5]
Move: Up
Heuristic value (misplaced tiles): 1
Cost to reach this node: 4

Current state:
[1, 2, 3]
[8, 0, 4]
[7, 6, 5]

Solution path:
[2, 8, 3]
[1, 6, 4]
[7, 0, 5]

[2, 8, 3]
[1, 0, 4]
[7, 6, 5]

[2, 0, 3]
[1, 8, 4]
[7, 6, 5]

[0, 2, 3]
[1, 8, 4]
[7, 6, 5]

[1, 2, 3]
[0, 8, 4]
[7, 6, 5]

[1, 2, 3]
[8, 0, 4]
[7, 6, 5]

```

Code :

Manhattan distance approach

```
class Node:
    def __init__(self, state, parent=None, move=None, cost=0):
        self.state = state
        self.parent = parent
        self.move = move
        self.cost = cost

    def heuristic(self):
        goal_positions = {
            1: (0, 0), 2: (0, 1), 3: (0, 2),
            8: (1, 0), 0: (1, 1), 4: (1, 2),
            7: (2, 0), 6: (2, 1), 5: (2, 2)
        }
        manhattan_distance = 0
        for i in range(len(self.state)):
            for j in range(len(self.state[i])):
                value = self.state[i][j]
                if value != 0:
                    goal_i, goal_j = goal_positions[value]
                    manhattan_distance += abs(i - goal_i) + abs(j - goal_j)
        return manhattan_distance

    def get_blank_position(state):
        for i in range(len(state)):
            for j in range(len(state[i])):
                if state[i][j] == 0:
                    return i, j

    def get_possible_moves(position):
        x, y = position
        moves = []
        if x > 0: moves.append((x - 1, y, 'Down'))
        if x < 2: moves.append((x + 1, y, 'Up'))
        if y > 0: moves.append((x, y - 1, 'Right'))
        if y < 2: moves.append((x, y + 1, 'Left'))
        return moves

    def generate_new_state(state, blank_pos, new_blank_pos):
        new_state = [row[:] for row in state]
        new_state[blank_pos[0]][blank_pos[1]], new_state[new_blank_pos[0]][new_blank_pos[1]] = \
            new_state[new_blank_pos[0]][new_blank_pos[1]], new_state[blank_pos[0]][blank_pos[1]]
        return new_state
```

```

def a_star_search(initial_state):
    open_list = []
    closed_list = set()

    initial_node = Node(state=initial_state, cost=0)
    open_list.append(initial_node)

    while open_list:
        open_list.sort(key=lambda node: node.cost + node.heuristic())
        current_node = open_list.pop(0)

        move_description = current_node.move if current_node.move else "Start"
        print("Current state:")
        for row in current_node.state:
            print(row)
        print(f"Move: {move_description}")
        print(f"Heuristic value (Manhattan distance): {current_node.heuristic()}")
        print(f"Cost to reach this node: {current_node.cost}\n")

        if current_node.heuristic() == 0:
            path = []
            while current_node:
                path.append(current_node)
                current_node = current_node.parent
            return path[::-1]
            closed_list.add(tuple(map(tuple, current_node.state)))

            blank_pos = get_blank_position(current_node.state)
            for new_blank_pos in get_possible_moves(blank_pos):
                new_state = generate_new_state(current_node.state, blank_pos, (new_blank_pos[0], new_blank_pos[1]))

                if tuple(map(tuple, new_state)) in closed_list:
                    continue

                cost = current_node.cost + 1
                move_direction = new_blank_pos[2]
                new_node = Node(state=new_state, parent=current_node, move=move_direction, cost=cost)

                if new_node not in open_list:
                    open_list.append(new_node)

    return None

```

```

initial_state = [[2,8,3], [1,6,4], [7,0,5]]
solution_path = a_star_search(initial_state)

if solution_path:
    print("Solution path:")
    for step in solution_path:
        for row in step.state:
            print(row)
        print()
else:
    print("No solution found.")

```

Output :

```

Current state:
[2, 8, 3]
[1, 6, 4]
[7, 0, 5]
Move: Start
Heuristic value (Manhattan distance): 5
Cost to reach this node: 0

```

```

Current state:
[2, 8, 3]
[1, 0, 4]
[7, 6, 5]
Move: Down
Heuristic value (Manhattan distance): 4
Cost to reach this node: 1

```

```

Current state:
[2, 0, 3]
[1, 8, 4]
[7, 6, 5]
Move: Down
Heuristic value (Manhattan distance): 3
Cost to reach this node: 2

```

```

Current state:
[0, 2, 3]
[1, 8, 4]
[7, 6, 5]
Move: Right
Heuristic value (Manhattan distance): 2
Cost to reach this node: 3

```

```

Current state:
[1, 2, 3]
[8, 0, 4]
[7, 6, 5]
Move: Left
Heuristic value (Manhattan distance): 0
Cost to reach this node: 5

```

```

Solution path:
[2, 8, 3]
[1, 6, 4]
[7, 0, 5]
[2, 8, 3]
[1, 0, 4]
[7, 6, 5]

```

```

[2, 0, 3]
[1, 8, 4]
[7, 6, 5]
[0, 2, 3]
[1, 8, 4]
[7, 6, 5]

```

```

[1, 2, 3]
[0, 8, 4]
[7, 6, 5]
[1, 2, 3]
[8, 0, 4]
[7, 6, 5]

```

Program 4

Implement Hill Climbing search algorithm to solve N-Queens problem

Algorithm:

22/10/24

Implement Hill climbing search to solve N-Queens problem.

Algorithm:

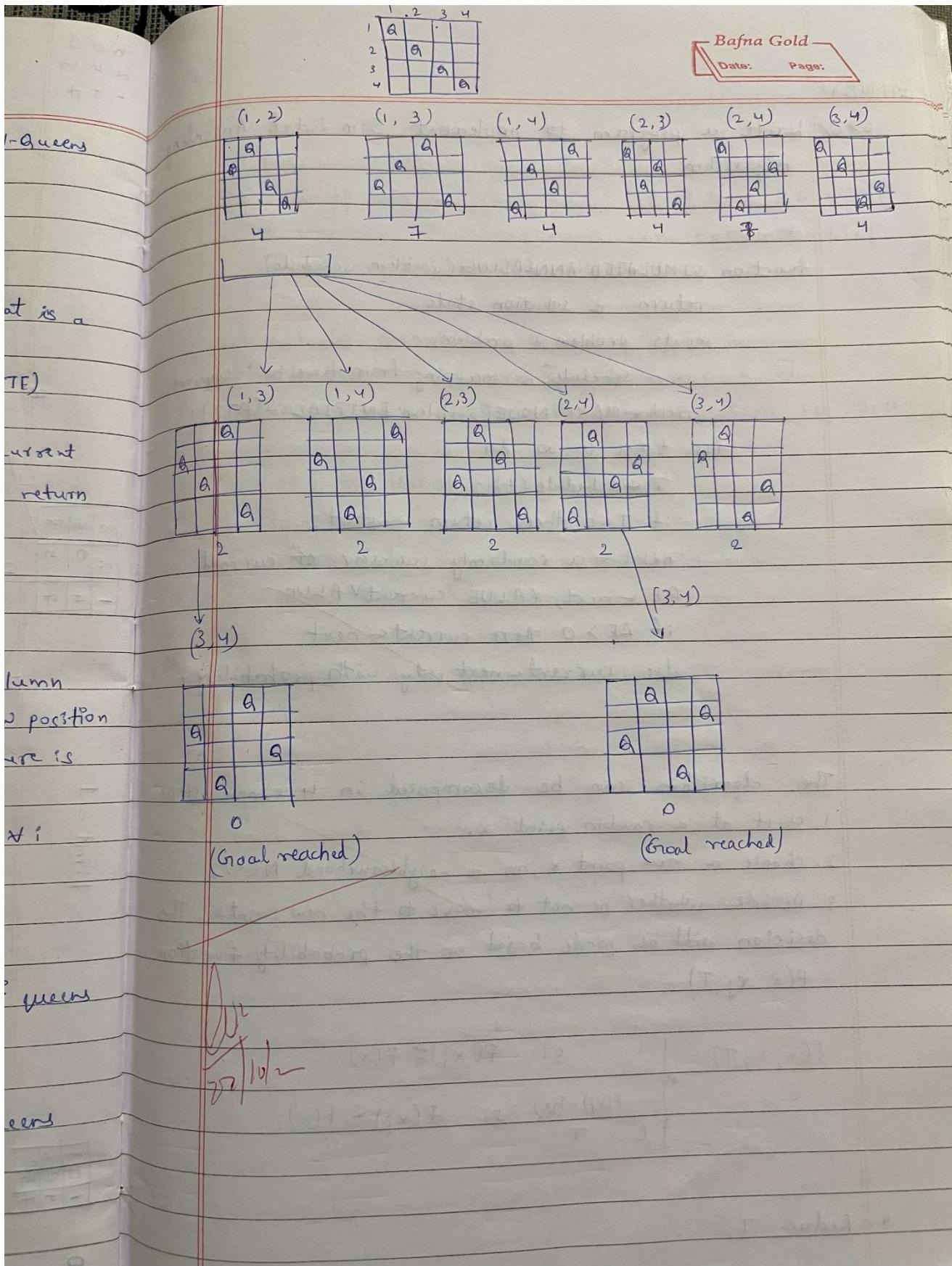
```
function HILL-CLIMBING(problem) return a state that is a local maximum
    current ← MAKE-NODE(problem, INITIAL-STATE)
    loop do
        neighbor ← highest valued successor of current
        if neighbor.VALUE < current.VALUE then return
            current.STATE
        current ← neighbor,
```

- state: 4 queens on the board, one queen per column
 - variables: x_0, x_1, x_2, x_3 where x_i is the row position of the queen in column i . Assume that there is one queen per column
 - Domain for each variable $x_i \in [0, 1, 2, 3], \forall i$

- Initial state: a random state

- Goal state: 4 queens on the board, no pair of queens are attacking each other

- Neighbour relation: swap the row position of 2 queens
- cost function: The number of pairs of queens attacking each other, directly or indirectly.



Code:

```
import random
def calculate_cost(board):

    n = len(board)

    attacks = 0

    for i in range(n):

        for j in range(i + 1, n):

            if board[i] == board[j]: # Same column

                attacks += 1

            if abs(board[i] - board[j]) == abs(i - j): # Same diagonal

                attacks += 1

    return attacks

def get_neighbors(board):

    neighbors = []

    n = len(board)

    for col in range(n):

        for row in range(n):

            if row != board[col]: # Only change the row of the queen

                new_board = board[:]

                new_board[col] = row

                neighbors.append(new_board)

    return neighbors
```

```

def hill_climb(board, max_restarts=100):
    current_cost = calculate_cost(board)
    print("Initial board configuration:")
    print_board(board, current_cost)
    iteration = 0
    restarts = 0

    while restarts < max_restarts: # Add limit to the number of restarts

        while current_cost != 0: # Continue until cost is zero

            neighbors = get_neighbors(board)
            best_neighbor = None
            best_cost = current_cost

            for neighbor in neighbors:

                cost = calculate_cost(neighbor)

                if cost < best_cost: # Looking for a lower cost

                    best_cost = cost
                    best_neighbor = neighbor

            if best_neighbor is None: # No better neighbor found

                break # Break the loop if we are stuck at a local minimum

        board = best_neighbor

```

```

current_cost = best_cost

iteration += 1

print(f"Iteration {iteration}:")
print_board(board, current_cost)

if current_cost == 0:
    break # We found the solution, no need for further restarts
else:
    # Restart with a new random configuration
    board = [random.randint(0, len(board)-1) for _ in range(len(board))]
    current_cost = calculate_cost(board)
    restarts += 1

    print(f"Restart {restarts}:")
    print_board(board, current_cost)

return board, current_cost

def print_board(board, cost):
    n = len(board)

    display_board = ['.' * n for _ in range(n)] # Create an empty board

    for col in range(n):
        display_board[board[col]][col] = 'Q' # Place queens on the board

    for row in range(n):

```

```

print(''.join(display_board[row])) # Print the board

print(f'Cost: {cost}\n')

if __name__ == "__main__":
    n = int(input("Enter the number of queens (N): ")) # User input for N

    initial_state = list(map(int, input(f'Enter the initial state (row numbers for each column, space-separated): ').split())))

    if len(initial_state) != n or any(r < 0 or r >= n for r in initial_state):
        print("Invalid initial state. Please ensure it has N elements with values from 0 to N-1.")

    else:
        solution, cost = hill_climb(initial_state)

        if cost == 0:
            print(f"Solution found with no conflicts:")

        else:
            print(f"No solution found within the restart limit:")

        print_board(solution, cost)

```

Output :

```
Enter the number of queens (N): 4
Enter the initial state (row numbers for each column, space-separated): 0 1 2 3
Initial board configuration:
Q . .
. Q .
. . Q .
. . . Q
Cost: 6

Iteration 1:
. . .
Q Q .
. . Q .
. . . Q
Cost: 4

Iteration 2:
. Q .
Q . .
. . Q .
. . . Q
Cost: 2

Restart 1:
. Q Q Q
. . .
. . .
Q . .
Cost: 4

Iteration 3:
. Q . Q
. . .
. . Q .
Q . .
Cost: 2

Iteration 4:
. Q .
. . Q
. . Q .
Q . .
Cost: 1

Restart 2:
. . . Q
. Q .
. . .
Q . Q .
Cost: 2

Iteration 6:
. . .
. Q .
. . Q Q
Q . .
Cost: 2

Iteration 7:
. . Q .
. Q .
. . . Q
Q . .
Cost: 1

Restart 4:
Q . .
. Q . Q
. . Q .
. . .
Cost: 5

Iteration 8:
Q . .
. Q . Q
. . .
. . .
Cost: 2

Iteration 9:
Q Q .
. . Q
. . .
. . Q .
Cost: 1

Iteration 10:
. Q .
. . Q
Q . .
. . Q .
Cost: 0

Solution found with no conflicts:
. Q .
. . Q
Q . .
. . Q .
Cost: 0
```

Program 5

Simulated Annealing to Solve 8-Queens problem

Algorithm:

29/10/24

6) Q) Write a program to implement Simulated Annealing Algorithm.

```
function SIMULATED ANNEALING(problem, schedule)
    return a solution state.
    inputs: problem, a problem
            schedule, a mapping from time to "temperature"
    current ← MAKE-NODE(problem.INITIAL-STATE)
    for t ← 1 to ∞ do
        T ← schedule(t)
        if T = 0 then return constant
        next ← a randomly successor of current
        ΔE ← next.VALUE - current.VALUE
        if ΔE > 0 then current ← next
        else current ← next only with probability e^ΔE/T
```

The algorithm can be decomposed in 4 simple steps:

1. Start at a random point x .
2. choose a new point x_j on a neighbourhood $N(x)$.
3. Decide whether or not to move to the new point x_j . The decision will be made based on the probability function

$$P(x, x_j, T)$$

$$P(x, x_j, T) = \begin{cases} 1 & \text{if } f(x_j) \leq f(x) \\ e^{\frac{f(x_j) - f(x)}{T}} & \text{if } f(x_j) < f(x) \end{cases}$$

4. Reduce T .

Output:

8 Queens Problem:

The best position found is: [4 1 7 0 3 6 2 5]

The number of queens that are not attacking each other is : 8.0

Tower of Hanoi:

Best state (final configuration): [2 2 2]

Number of correct disks on destination peg: 3.0

Tower of Hanoi configuration:

peg 0: []

peg 1: []

peg 2: [0, 1, 2]

10/11

(Hanoi) T2R1F → 9

(Hanoi) T2F1R → 10

(Hanoi) T1R1F → 11

Code:

```
#!pip install mlrose-hiive joblib
#!pip install --upgrade joblib
#!pip install joblib==1.1.0
import mlrose_hiive as mlrose
import numpy as np

def queens_max(position):
    no_attack_on_j = 0
    queen_not_attacking = 0
    for i in range(len(position) - 1):
        no_attack_on_j = 0
        for j in range(i + 1, len(position)):
            if (position[j] != position[i]) and (position[j] != position[i] + (j - i)) and (position[j] != position[i] - (j - i)):
                no_attack_on_j += 1
        if (no_attack_on_j == len(position) - 1 - i):
            queen_not_attacking += 1
    if (queen_not_attacking == 7):
        queen_not_attacking += 1
    return queen_not_attacking

objective = mlrose.CustomFitness(queens_max)

problem = mlrose.DiscreteOpt(length=8, fitness_fn=objective, maximize=True, max_val=8)
T = mlrose.ExpDecay()

initial_position = np.array([4, 6, 1, 5, 2, 0, 3, 7])

#The simulated_annealing function returns 3 values, we need to capture all 3
best_position, best_objective, fitness_curve = mlrose.simulated_annealing(problem=problem,
                           schedule=T, max_attempts=500,
                           init_state=initial_position)

print('The best position found is:', best_position)
print('The number of queens that are not attacking each other is:', best_objective)
```

Output :

```
The best position found is: [4 0 7 5 2 6 1 3]
The number of queens that are not attacking each other is: 8.0
```

Program 6

Create a knowledge base using propositional logic and show that the given query entails the knowledge base or not.

Algorithm:

12/11/24

LAB 6

- Q) Create a knowledge base using propositional logic and show that the given query entails the knowledge base or not.

function TT-ENTAILS?(KB, α) returns true or false
inputs: KB, the knowledge base, a sentence in propositional logic

α , the query, a sentence in propositional logic
symbols \leftarrow a list of the proposition symbols in KB and α .
return TT-CHECK-ALL(KB, α , symbols, {})

function TT-CHECK-ALL(KB, α , symbols, model) returns true or false

if EMPTY?(symbols) then

: if PL-TRUE?(KB, model) then return PL-TRUE?(α , model)
else return true || when KB is false, always return true

else do

$P \leftarrow$ FIRST(symbols)

rest \leftarrow REST(symbols)

return (TT-CHECK-ALL(KB, α , rest, model \cup { $P = \text{true}$ })

and

TT-CHECK-ALL(KB, α , rest, model \cup { $P = \text{false}$ }))

Propositional Interface?

Ex: $\alpha = A \vee B$ $KB = (AVC) \wedge (BV \neg C)$

Checking that $KB \models \alpha$

A	B	C	AVC	$BV \neg C$	KB	α
F	F	F	F	T	F	F
F	F	T	T	F	F	F
F	T	F	F	T	F	T
<input type="checkbox"/> F	<input type="checkbox"/> T	<input type="checkbox"/> T	T	T	<input type="checkbox"/> T	<input type="checkbox"/> T
<input type="checkbox"/> T	<input type="checkbox"/> F	<input type="checkbox"/> F	T	T	<input type="checkbox"/> T	<input type="checkbox"/> T
<input type="checkbox"/> F	<input type="checkbox"/> F	<input type="checkbox"/> F	T	F	F	T
<input type="checkbox"/> T	<input type="checkbox"/> T	<input type="checkbox"/> F	T	T	<input type="checkbox"/> T	<input type="checkbox"/> T
<input type="checkbox"/> T	<input type="checkbox"/> T	<input type="checkbox"/> T	T	T	<input type="checkbox"/> T	<input type="checkbox"/> T

OP Seen

8/8
12/11/24

Code:

```
import pandas as pd
```

```
# Define the truth table for all combinations of A, B, C
```

```
truth_values = [(False, False, False),  
                (False, False, True),  
                (False, True, False),  
                (False, True, True),  
                (True, False, False),  
                (True, False, True),  
                (True, True, False),  
                (True, True, True)]
```

```
# Columns: A, B, C
```

```
table = pd.DataFrame(truth_values, columns=["A", "B", "C"])
```

```
# Calculate intermediate columns
```

```
table["A or C"] = table["A"] | table["C"]      # A ∨ C
```

```
table["B or not C"] = table["B"] | ~table["C"]  # B ∨ ¬C
```

```
# Knowledge Base (KB): (A ∨ C) ∧ (B ∨ ¬C)
```

```
table["KB"] = table["A or C"] & table["B or not C"]
```

```
# Alpha ( $\alpha$ ): A ∨ B
```

```
table["Alpha ( $\alpha$ )"] = table["A"] | table["B"]
```

```
# Define a highlighting function
```

```
def highlight_rows(row):
```

```
    if row["KB"] and row["Alpha ( $\alpha$ )"]:
```

```
        return ['font-weight: bold; color: black'] * len(row)
```

```
    else:
```

```
        return [""] * len(row)
```

```
# Apply the highlighting function
```

```
styled_table = table.style.apply(highlight_rows, axis=1)
```

```
# Display the styled table
```

```
styled_table
```

Output :

	A	B	C	A or C	B or not C	KB	Alpha (α)
0	False	False	False	False	True	False	False
1	False	False	True	True	False	False	False
2	False	True	False	False	True	False	True
3	False	True	True	True	True	True	True
4	True	False	False	True	True	True	True
5	True	False	True	True	False	False	True
6	True	True	False	True	True	True	True
7	True	True	True	True	True	True	True

Program 7

Implement unification in first order logic

Algorithm:

LAB 7

Q) Implement Unification in First Order Logic.

Algorithm: $\text{Unify}(\psi_1, \psi_2)$

Step 1: If ψ_1 or ψ_2 is a variable or constant, then:

a) If ψ_1 or ψ_2 are identical, then return NIL.

b) Else if ψ_1 is a variable,

a. then if ψ_1 occurs in ψ_2 , then return FAILURE

b. Else return $\{\psi_2 / \psi_1\}$.

c) Else if ψ_2 is a variable,

a. If ψ_2 occurs in ψ_1 , then return FAILURE

b. Else return $\{\psi_1 / \psi_2\}$.

d) Else return FAILURE.

Step 2: If the initial predicate symbol in ψ_1 and ψ_2 are not same, then return FAILURE.

Step 3: If ψ_1 and ψ_2 have a different number of arguments then return FAILURE.

Step 4: Set Substitution set(SUBST) to NIL.

Step 5: For $i=1$ to the number of elements in ψ_1 .

a) Call Unity function with the i th element of ψ_1 and i th element of ψ_2 , and put the result into S.

b) If $S = \text{failure}$ then return failure

c) If $S \neq \text{NIL}$ then do,

a. Apply S to the remainders of both L1 & L2

b. SUBST = APPEND(S, SUBST)

Step 6: Return SUBST

Code:

```
import re

def occurs_check(var, x):
    """Checks if var occurs in x (to prevent circular substitutions)."""
    if var == x:
        return True
    elif isinstance(x, list): # If x is a compound expression (like a function or predicate)
        return any(occurs_check(var, xi) for xi in x)
    return False

def unify_var(var, x, subst):
    """Handles unification of a variable with another term."""
    if var in subst: # If var is already substituted
        return unify(subst[var], x, subst)
    elif isinstance(x, (list, tuple)) and tuple(x) in subst: # Handle compound expressions
        return unify(var, subst[tuple(x)], subst)
    elif occurs_check(var, x): # Check for circular references
        return "FAILURE"
    else:
        # Add the substitution to the set (convert list to tuple for hashability)
        subst[var] = tuple(x) if isinstance(x, list) else x
    return subst

def unify(x, y, subst=None):
    """
    Unifies two expressions x and y and returns the substitution set if they can be unified.
    Returns 'FAILURE' if unification is not possible.
    """
    if subst is None:
        subst = {} # Initialize an empty substitution set

    # Step 1: Handle cases where x or y is a variable or constant
    if x == y: # If x and y are identical
        return subst
    elif isinstance(x, str) and x.islower(): # If x is a variable
        return unify_var(x, y, subst)
    elif isinstance(y, str) and y.islower(): # If y is a variable
        return unify_var(y, x, subst)
    elif isinstance(x, list) and isinstance(y, list): # If x and y are compound expressions (lists)
        if len(x) != len(y): # Step 3: Different number of arguments
            return "FAILURE"

    # Step 2: Check if the predicate symbols (the first element) match
    if x[0] != y[0]: # If the predicates/functions are different
```

```

    return "FAILURE"

# Step 5: Recursively unify each argument
for xi, yi in zip(x[1:], y[1:]): # Skip the predicate (first element)
    subst = unify(xi, yi, subst)
    if subst == "FAILURE":
        return "FAILURE"
    return subst
else: # If x and y are different constants or non-unifiable structures
    return "FAILURE"

def unify_and_check(expr1, expr2):
    """
    Attempts to unify two expressions and returns a tuple:
    (is_unified: bool, substitutions: dict or None)
    """
    result = unify(expr1, expr2)
    if result == "FAILURE":
        return False, None
    return True, result

def display_result(expr1, expr2, is_unified, subst):
    print("Expression 1:", expr1)
    print("Expression 2:", expr2)
    if not is_unified:
        print("Result: Unification Failed")
    else:
        print("Result: Unification Successful")
        print("Substitutions:", {k: list(v) if isinstance(v, tuple) else v for k, v in subst.items()})

def parse_input(input_str):
    """Parses a string input into a structure that can be processed by the unification algorithm."""
    # Remove spaces and handle parentheses
    input_str = input_str.replace(" ", "")

    # Handle compound terms (like p(x, f(y)) -> ['p', 'x', ['f', 'y']])
    def parse_term(term):
        # Handle the compound term
        if '(' in term:
            match = re.match(r'([a-zA-Z0-9_]+)\((.*)\)', term)
            if match:
                predicate = match.group(1)
                arguments_str = match.group(2)
                arguments = [parse_term(arg.strip()) for arg in arguments_str.split(',')]


```

```

        return [predicate] + arguments
    return term

return parse_term(input_str)

# Main function to interact with the user
def main():
    while True:
        # Get the first and second terms from the user
        expr1_input = input("Enter the first expression (e.g., p(x, f(y))): ")
        expr2_input = input("Enter the second expression (e.g., p(a, f(z))): ")

        # Parse the input strings into the appropriate structures
        expr1 = parse_input(expr1_input)
        expr2 = parse_input(expr2_input)

        # Perform unification
        is_unified, result = unify_and_check(expr1, expr2)

        # Display the results
        display_result(expr1, expr2, is_unified, result)

        # Ask the user if they want to run another test
        another_test = input("Do you want to test another pair of expressions? (yes/no): ").strip().lower()
        if another_test != 'yes':
            break

if __name__ == "__main__":
    main()

```

Output :

```
Enter the first expression (e.g., p(x, f(y))): p(b,x,f(g(z)))
Enter the second expression (e.g., p(a, f(z))): p(z,f(y),f(y))
Expression 1: ['p', 'b', 'x', ['f', ['g', 'z']]]
Expression 2: ['p', 'z', ['f', 'y'], ['f', 'y']]
Result: Unification Successful
Substitutions: {'b': 'z', 'x': ['f', 'y'], 'y': ['g', 'z']}
Do you want to test another pair of expressions? (yes/no): yes
Enter the first expression (e.g., p(x, f(y))): p(x,h(y))
Enter the second expression (e.g., p(a, f(z))): p(a,f(z))
Expression 1: ['p', 'x', ['h', 'y']]
Expression 2: ['p', 'a', ['f', 'z']]
Result: Unification Failed
Do you want to test another pair of expressions? (yes/no): yes
Enter the first expression (e.g., p(x, f(y))): p(f(a),g(y))
Enter the second expression (e.g., p(a, f(z))): p(x,x)
Expression 1: ['p', ['f', 'a'], ['g', 'y']]
Expression 2: ['p', 'x', 'x']
Result: Unification Failed
Do you want to test another pair of expressions? (yes/no): no
```

Program 8

Create a knowledge base consisting of first order logic statements and prove the given query using forward reasoning.

Algorithm:

LAB 8

- A) Create a knowledge base consisting of first order logic statements and prove the given query using forward reasoning.

function FOL-FC-Ask(KB, α) returns a substitution or false

inputs: KB, the knowledge base, a set of first-order definite clauses

α , the query, an atomic sentence

local variables: new, the new sentences inferred on each iteration

repeat until new is empty

new $\leftarrow \emptyset$

for each rule in KB do:

$(p_1 \wedge \dots \wedge p_n \Rightarrow q) \leftarrow \text{STANDARDIZE-VARIABLES}(\text{rule})$

for each θ such that $\text{SUBST}(\theta, p_1 \wedge \dots \wedge p_n) = \text{SUBST}(\theta, p'_1 \wedge \dots \wedge p'_n)$

for some $(p'_1 \dots p'_n \text{ in } \text{KB})$

$\eta' \leftarrow \text{SUBST}(\theta, \eta)$

if η' does not unify with some sentence already in KB or new then

add η' to new

$\phi \leftarrow \text{UNIFY}(\eta', \alpha)$

if ϕ is not fail then return ϕ

add new to KB

return false

Representation in FOL:

→ It is a crime for an American to sell weapons to hostile nations.

Let's say p, q, and r are variables

$\text{American}(p) \wedge \text{Weapon}(q) \wedge \text{Sells}(p, q, r) \wedge \text{Hostile}(r) \Rightarrow \text{Criminal}(p)$

→ Country A has some missiles

$\exists x \text{ owns}(A, x) \wedge \text{Missile}(x)$

Existential Instantiation, introducing a new constant T1:

$\text{Owns}(A, T1)$

$\text{Missile}(T1)$

→ All the missiles were sold to country A by Robert

$\forall x \text{ Missile}(x) \wedge \text{Owns}(A, x) \Rightarrow \text{Sells}(\text{Robert}, x, A)$

→ Missiles are weapons

$\text{Missiles}(x) \Rightarrow \text{weapons}(x)$

Robert is an American

$\text{American}(\text{Robert})$

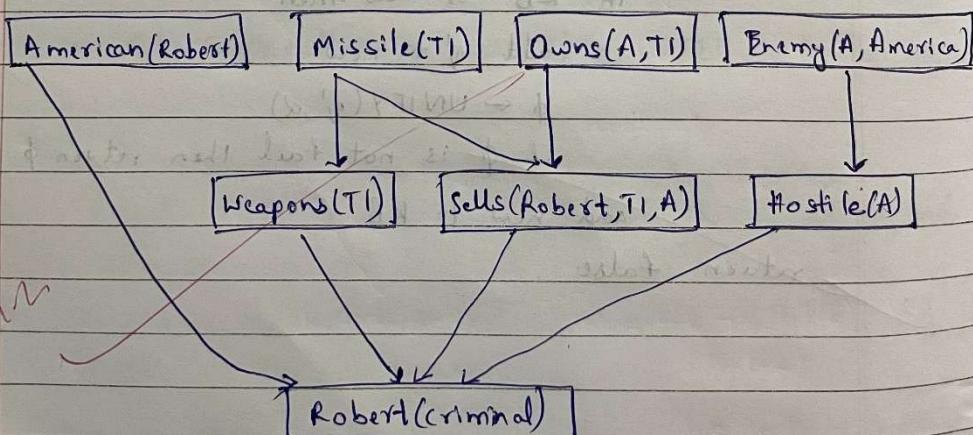
→ Enemy of America is known as hostile

$\forall x \text{ Enemy}(x, \text{America}) \Rightarrow \text{Hostile}(x)$

→ The country A, an enemy of America

$\text{Enemy}(A, \text{America})$

To Prove: Robert is Criminal $\text{Criminal}(\text{Robert})$



$\text{American}(p) \wedge \text{Weapon}(q) \wedge \text{Sells}(p, q, r) \wedge \text{Hostile}(r) \Rightarrow \text{Criminal}(p)$

a)

1.

2.

3.

4.

a.

b.

c.

d.

e.

f.

g.

Code:

```
class KnowledgeBase:
    def __init__(self):
        self.facts = set() # Set of known facts
        self.rules = [] # List of rules

    def add_fact(self, fact):
        self.facts.add(fact)

    def add_rule(self, rule):
        self.rules.append(rule)

    def infer(self):
        inferred = True
        while inferred:
            inferred = False
            for rule in self.rules:
                if rule.apply(self.facts):
                    inferred = True

# Define the Rule class
class Rule:
    def __init__(self, premises, conclusion):
        self.premises = premises # List of conditions
        self.conclusion = conclusion # Conclusion to add if premises are met

    def apply(self, facts):
        if all(premise in facts for premise in self.premises):
            if self.conclusion not in facts:
                facts.add(self.conclusion)
                print(f"Inferred: {self.conclusion}")
                return True
        return False

# Initialize the knowledge base
kb = KnowledgeBase()

# Facts in the problem
kb.add_fact("American(Robert)")
kb.add_fact("Missile(T1)")
kb.add_fact("Owns(A, T1)")
kb.add_fact("Enemy(A, America)")

# Rules based on the problem
# 1. Missile(x) implies Weapon(x)
kb.add_rule(Rule(["Missile(T1)"], "Weapon(T1)"))
```

```

# 2. Enemy(x, America) implies Hostile(x)
kb.add_rule(Rule(["Enemy(A, America)"], "Hostile(A)"))

# 3. Missile(x) and Owns(A, x) imply Sells(Robert, x, A)
kb.add_rule(Rule(["Missile(T1)", "Owns(A, T1)"], "Sells(Robert, T1, A)"))

# 4. American(p) and Weapon(q) and Sells(p, q, r) and Hostile(r) imply Criminal(p)
kb.add_rule(Rule(["American(Robert)", "Weapon(T1)", "Sells(Robert, T1, A)", "Hostile(A)"],
    "Criminal(Robert)"))

# Infer new facts based on the rules
kb.infer()

# Check if Robert is a criminal
if "Criminal(Robert)" in kb.facts:
    print("Conclusion: Robert is a criminal.")
else:
    print("Conclusion: Unable to prove Robert is a criminal.")

```

Output :

```

Inferred: Weapon(T1)
Inferred: Hostile(A)
Inferred: Sells(Robert, T1, A)
Inferred: Criminal(Robert)
Conclusion: Robert is a criminal.

```

Program 9

Create a knowledge base consisting of first order logic statements and prove the given query using Resolution

Algorithm:

LAB 9

- a) Convert a given first order logic statement into Resolution

Basic steps for proving a conclusion S given premises

Premise₁, ..., Premise_n

1. Convert all sentences to CNF
2. Negate conclusion S & convert result to CNF
3. Add negated conclusion S to the premise clauses
4. Repeat until contradiction or no progress is made
 - a. Select 2 clauses (call them parent clauses)
 - b. Resolve them together, performing all required unifications
 - c. If resolvent is the empty clause, a contradiction has been found (i.e. S follows from the premises)
 - d. If not, add resolvent to the premises

If we succeed in step 4, we have proved the conclusion

Representation in FOL:

Given KB or premises:

- a. John likes all kind of food a. $\forall x : \text{food}(x) \rightarrow \text{likes}(\text{John}, x)$
- b. Apple and vegetables are food b. $\text{food}(\text{Apple}) \wedge \text{food}(\text{vegetables})$
- c. Anything anyone eats and not killed is food c. $\forall x \forall y : \text{eats}(x, y) \wedge \neg \text{killed}(x) \rightarrow \text{food}(y)$
- d. Anil eats peanuts & still alive d. $\text{eats}(\text{Anil}, \text{Peanuts}) \wedge \text{alive}(\text{Anil})$
- e. Harry eats everything that Anil eats e. $\forall x : \text{eats}(\text{Anil}, x) \rightarrow \text{eats}(\text{Harry}, x)$
- f. Anyone who is alive implies not killed f. $\forall x : \text{alive}(x) \rightarrow \neg \text{killed}(x)$
- g. Anyone who is not killed implies alive g. $\forall x : \neg \text{killed}(x) \rightarrow \text{alive}(x)$
- h. Prove by resolution that:
John likes peanuts h. $\text{likes}(\text{John}, \text{peanuts})$

① Eliminate Implication:

- $\forall x \rightarrow \text{food}(x) \vee \text{likes}(\text{John}, x)$
- $\text{food}(\text{Apple}) \wedge \text{food}(\text{vegetables})$
- $\forall x \forall y \rightarrow [\text{eats}(x, y) \wedge \neg \text{killed}(x)] \vee \text{food}(y)$
- $\text{eats}(\text{Anil}, \text{Peanuts}) \wedge \text{alive}(\text{Anil})$
- $\forall x \neg \text{eats}(\text{Anil}, x) \vee \text{eats}(\text{Harry}, x)$
- $\forall x \rightarrow [\neg \text{killed}(x)] \vee \text{alive}(x)$
- $\forall x \neg \text{alive}(x) \vee \neg \text{killed}(x)$
- $\text{likes}(\text{John}, \text{Peanuts})$

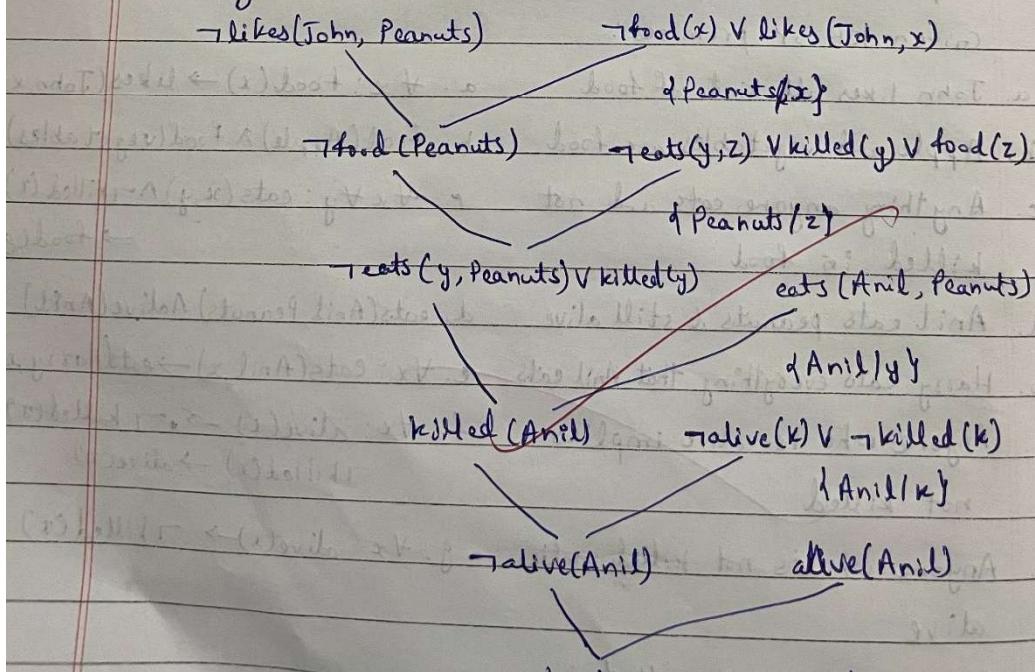
② Move Negation

(\neg)owards b-side

③ Rename variables or standardize variables. ④ Drop Universal Quantification

- $\forall x \neg \text{food}(x) \vee \text{likes}(\text{John}, x)$
 - $\text{food}(\text{Apple}) \wedge \text{food}(\text{vegetables})$
 - $\forall y \forall z \neg \text{eats}(y, z) \vee \text{killed}(y) \vee \text{food}(z)$
 - $\text{eats}(\text{Anil}, \text{Peanuts}) \wedge \text{alive}(\text{Anil})$
 - $\forall w \neg \text{eats}(\text{Anil}, w) \vee \text{eats}(\text{Harry}, w)$
 - $\forall g [\text{killed}(g)] \vee \text{alive}(g)$
 - $\forall k \neg \text{alive}(k) \vee \neg \text{killed}(k)$
 - $\text{likes}(\text{John}, \text{Peanuts})$
- $\neg \text{food}(x) \vee \text{likes}(\text{John}, x)$
 - $\text{food}(\text{Apple})$
 - $\neg \text{eats}(y, z) \vee \text{killed}(y) \vee \text{food}(z)$
 - $\neg \text{eats}(\text{Anil}, \text{Peanuts})$
 - $\neg \text{alive}(\text{Anil})$
 - $\neg \text{eats}(\text{Anil}, w) \vee \text{eats}(\text{Harry}, w)$
 - $\text{killed}(g) \vee \text{alive}(g)$
 - $\neg \text{alive}(k) \vee \neg \text{killed}(k)$
 - $\text{likes}(\text{John}, \text{Peanuts})$

Proof by Resolution!



Code:

```
KB = {  
    "food(Apple)": True,  
    "food(vegetables)": True,  
    "eats(Anil, Peanuts)": True,  
    "alive(Anil)": True,  
    "likes(John, X)": "food(X)", # Rule: John likes all food  
    "food(X)": "eats(Y, X) and not killed(Y)", # Rule: Anything eaten and not killed is food  
    "eats(Harry, X)": "eats(Anil, X)", # Rule: Harry eats what Anil eats  
    "alive(X)": "not killed(X)", # Rule: Alive implies not killed  
    "not killed(X)": "alive(X)", # Rule: Not killed implies alive  
}
```

Function to evaluate if a predicate is true based on the KB

```
def resolve(predicate):  
    # If it's a direct fact in KB  
    if predicate in KB and isinstance(KB[predicate], bool):  
        return KB[predicate]
```

If it's a derived rule

```
if predicate in KB:  
    rule = KB[predicate]  
    if " and " in rule: # Handle conjunction  
        sub_preds = rule.split(" and ")  
        return all(resolve(sub.strip()) for sub in sub_preds)  
    elif " or " in rule: # Handle disjunction  
        sub_preds = rule.split(" or ")  
        return any(resolve(sub.strip()) for sub in sub_preds)  
    elif "not " in rule: # Handle negation  
        sub_pred = rule[4:] # Remove "not "  
        return not resolve(sub_pred.strip())  
    else: # Handle single predicate  
        return resolve(rule.strip())
```

If the predicate is a specific query (e.g., likes(John, Peanuts))

```
if "(" in predicate:  
    func, args = predicate.split("(")  
    args = args.strip(")").split(", ")  
    if func == "food" and args[0] == "Peanuts":  
        return resolve("eats(Anil, Peanuts)") and not resolve("killed(Anil)")  
    if func == "likes" and args[0] == "John" and args[1] == "Peanuts":  
        return resolve("food(Peanuts)")
```

Default to False if no rule or fact applies

```
return False
```

Query to prove: John likes Peanuts

```
query = "likes(John, Peanuts)"  
result = resolve(query)  
  
# Print the result  
print(f"Does John like peanuts? {'Yes' if result else 'No'}")
```

Output:

```
Does John like peanuts? Yes
```

Program 10

Implement Alpha-Beta Pruning.

Algorithm:

Q) Implement Alpha Beta Pruning Algorithm

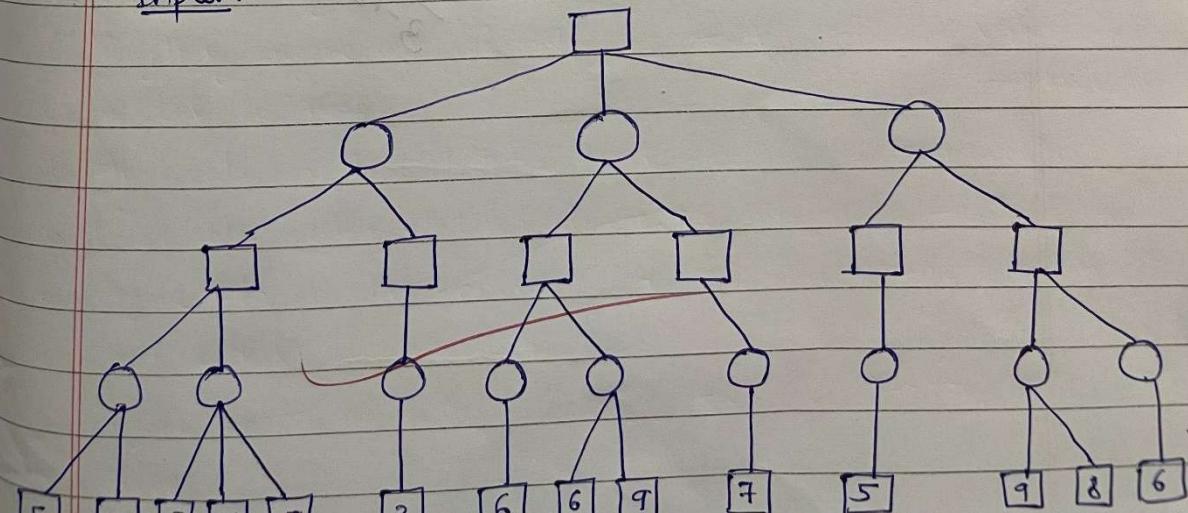
- Alpha(α) - Beta(β) proposes to find the optimal path without looking at every node in the game tree.
- Max contains Alpha(α) and Min contains Beta(β) bound during the calculation.
- In both MIN and MAX node, we return where $\alpha > \beta$ which compares with its parent node only.
- Both minimax and Alpha(α) - Beta(β) cut-off give same path.
- Alpha(α) - Beta(β) gives optimal solution as it takes less time to get the value for the root node.

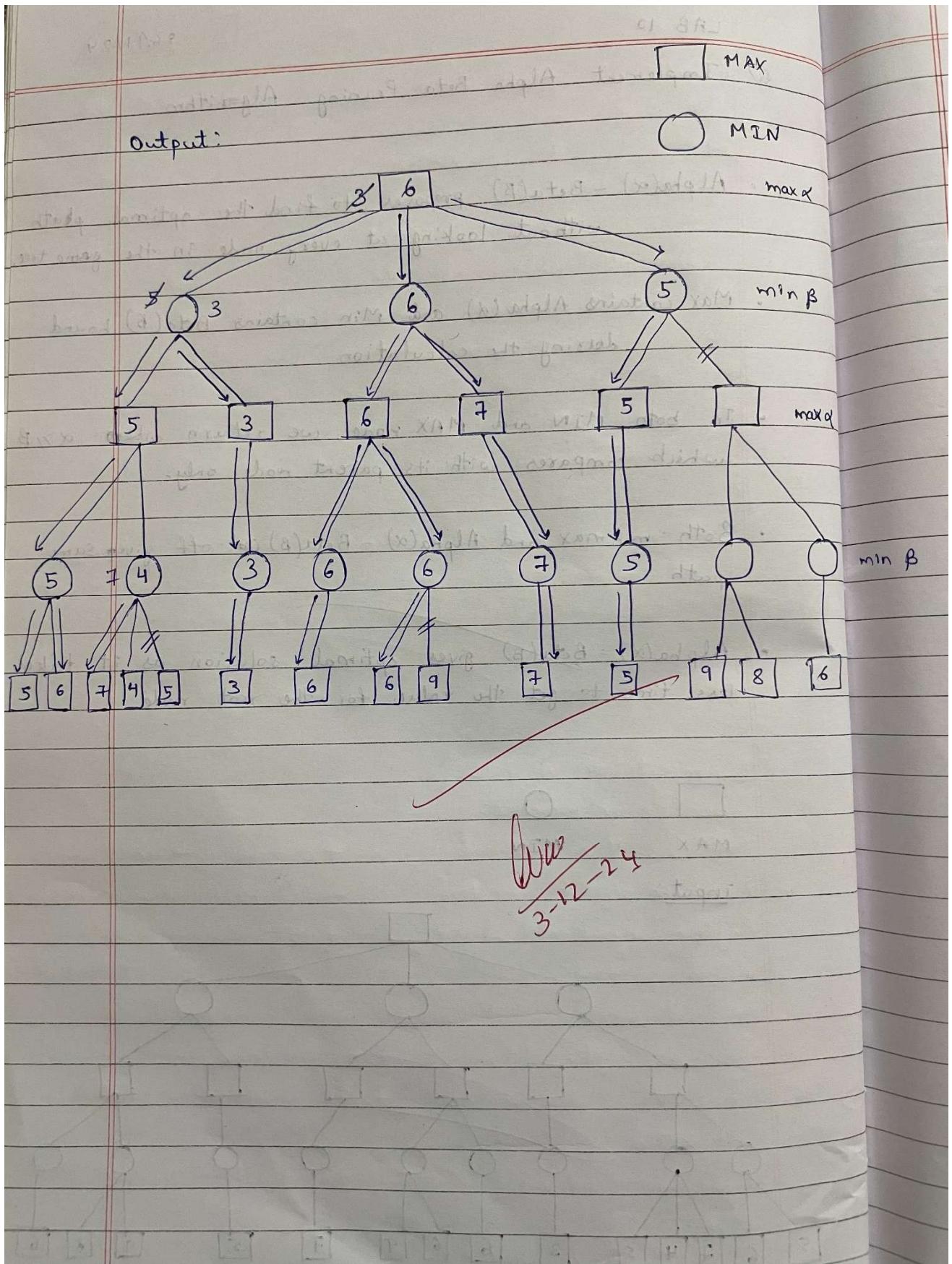


MAX



MIN

Input:



Code:

```
import math
def minimax(node, depth, is_maximizing):
    """
    Implement the Minimax algorithm to solve the decision tree.
```

Parameters:

node (dict): The current node in the decision tree, with the following structure:

```
{  
    'value': int,  
    'left': dict or None,  
    'right': dict or None  
}
```

depth (int): The current depth in the decision tree.

is_maximizing (bool): Flag to indicate whether the current player is the maximizing player.

Returns:

int: The utility value of the current node.

```
"""
```

Base case: Leaf node

```
if node['left'] is None and node['right'] is None:  
    return node['value']
```

Recursive case

```
if is_maximizing:  
    best_value = -math.inf  
    if node['left']:  
        best_value = max(best_value, minimax(node['left'], depth + 1, False))  
    if node['right']:  
        best_value = max(best_value, minimax(node['right'], depth + 1, False))  
    return best_value  
else:  
    best_value = math.inf  
    if node['left']:  
        best_value = min(best_value, minimax(node['left'], depth + 1, True))  
    if node['right']:  
        best_value = min(best_value, minimax(node['right'], depth + 1, True))  
    return best_value
```

Example usage

```
decision_tree = {  
    'value': 5,  
    'left': {  
        'value': 6,  
        'left': {  
            'value': 7,  
            'left': {
```

```
'value': 4,
'left': None,
'right': None
},
'right': {
    'value': 5,
    'left': None,
    'right': None
}
},
'right': {
    'value': 3,
    'left': {
        'value': 6,
        'left': None,
        'right': None
    },
    'right': {
        'value': 9,
        'left': None,
        'right': None
    }
},
'right': {
    'value': 8,
    'left': {
        'value': 7,
        'left': {
            'value': 6,
            'left': None,
            'right': None
        },
        'right': {
            'value': 9,
            'left': None,
            'right': None
        }
    },
    'right': {
        'value': 8,
        'left': {
            'value': 6,
            'left': None,
            'right': None
        },
        'right': None
    }
}
```

```
        }
    }
}

# Find the best move for the maximizing player
best_value = minimax(decision_tree, 0, True)
print(f"The best value for the maximizing player is: {best_value}")
```

Output:

```
The best value for the maximizing player is: 6
```