

# **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**

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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 **RAGHAVENDRA R (1BM22CS214)**, 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.

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## LAB 1: Tic-Tac-Toe Game

### Algorithm:

tic-tac-toe: (LAB-1) 24/09/24

man = x : comp = o

0	x	0
0	x	x
x	0	+

Step 1: create a '2D' array & initialize it with '-'  
it is a 3x3 matrix.

Step 2:  $\Rightarrow$  `[['-', '-', '-'], ['-', '-', '-'], ['-', '-', '-']]`

Step 3: Use random to select the player to go first  
(human / computer). First move will be 'x'  
Display the board after the each move

```
def print-board(board):
    for row in board:
        print(' '.join(row))
    print('-' * 5)
```

Step 3: After each move, check for the win of both the  
player and computer and if a win then return  
true for that particular player / for computer

```
def win(board):
    for i in range(3):
        if (board[i][0] == board[i][1] == board[i][2] != ''):
            return True, (board[i][0])
        if (board[0][i] == board[1][i] == board[2][i] != ''):
            return True, (board[0][i])
    // check for diagonals
    if (board[0][0] == board[1][1] == board[2][2] != ''):
        return board[0][0]
    if (board[2][0] == board[1][1] == board[0][2] != ''):
        return board[0][2]
```

Step 4: player must enter the index for the play and the particular place must be allotted be in the board, if not occupied early

```
def player(board):  
    row, col = input :  
    if (board[row][col] == '-'):  
        board[row][col] = 'X'.  
    }
```

Step 5: computer will analyze the moves like it will first prioritize the win moves of its and then avoid the winning possibilities of the player

// check winning move.

```
for i in range(3):
```

```
    for j in range(3):
```

```
        if board[i][j] == '-':
```

```
            board[i][j] = 'O':
```

```
            if check(winner()) == 'O':
```

```
                return // returns if the computer  
                        can win
```

// place at any random pos:

At get a random place and if it is free in the board place it over there..

Step 6: if ~~any~~ any player wins, display it.

Code:

```
import random

def win(board):
    for row in board:
        if row[0] == row[1] == row[2] != "":
            return True
    for col in range(3):
        if board[0][col] == board[1][col] == board[2][col] != "":
            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 printBoard(board):
    print("\n".join([" | ".join(row) for row in board]))

def draw(board):
    return all(cell != "" for row in board for cell in row)

def user_move(board):
    while True:
        try:
            move = int(input("Enter your move (1-9): ")) - 1
            row, col = divmod(move, 3)
            if board[row][col] == "":
                board[row][col] = "X"
                break
            else:
                print("That space is already taken. Try again.")
        except (ValueError, IndexError):
            print("Invalid input. Please enter a number from 1 to 9.")

def computer_move(board):
    while True:
        move = random.randint(0, 8)
        row, col = divmod(move, 3)
        if board[row][col] == "":
            board[row][col] = "O"
            break

def _main():
    board = [["" for _ in range(3)] for _ in range(3)]
```

```

while True:
    printBoard(board)
    user_move(board)
    if win(board):
        printBoard(board)
        print("You win!")
        break
    if draw(board):
        printBoard(board)
        print("It's a draw!")
        break
    computer_move(board)
    if win(board):
        printBoard(board)
        print("Computer wins!")
        break
    if draw(board):
        printBoard(board)
        print("It's a draw!")
        break

if __name__ == "__main__":
    _main()

```

**Output:**

```

| | |
| | |
| | |
Enter your move (1-9): 3
| | X
| O |
| | |
Enter your move (1-9): 2
| X | X
| O |
| | O
Enter your move (1-9): 1
X | X | X
| O |
| | O
You win!
PS C:\Users\Dell\Desktop\BIS>

```

## LAB 2: Vacuum cleaner agent

### Algorithm:

LAB-2 : 1/10/2024

\* write an algorithm and program for a AI controlled vacuum cleaners.

Step 1: create two rooms using two variables A & B. 

A	B
---	---

Step 2: take user input from the user for room A & B as 0 - dirty, 1 - clean.

Step 3: The agent is in "Room-A" and it checks if the room is dirty or not. and prints "A-cleaned" and moves to next room.

⇒ user inputs : A & B. (0 - dirty & 1 - clean)

```
def check_for_clean(var): (A or B).
    while (True):
        if var == 0:
            clean(var);
        else:
            move(var);
    end while
end def
```

Step 4: The room gets cleaned by the agent (vacuum cleaner) and updates the status to 1 from 0 and prints as "cleaned"

```
def clean(var):
    if (var == 0) {
        var == 1.
    end if
    move(var);
end def
```

For 1/10/24



steps: The agents moves to next room and if that room is cleaned come back to previous room and exit the loop if both the rooms are cleaned.

```

def move_agent(var, var):
    while(True):
        if var1 == 0:
            check_clean(var1)

        else
            if var2 == 0:
                check-clean(var2)

            end-if (var1 == var2):
                break;
    end while
end def
  
```

**Code: (2 rooms)**

```

def printArr(arr):
    n=len(arr)
    print(arr[0],arr[1])

def clean(arr,vac):
    if(arr[vac] == 1):
        arr[vac]=0
    if(arr[vac] == 0):
        return

def check(arr):
    if(arr[0]==0 and arr[1]==0):
        return False
    else:
        return True

print("Enter the status of the room(0 for clean; 1 for dirty):")
arr1 = []
for i in range(0,2):
    a=int(input("Status of the room %d:" %i))
  
```

```

    arr1.append(a)

vac=0
while(True):
    printArr(arr1)
    if(check(arr1) == False):
        break
    clean(arr1,vac)
    if(vac==0):
        vac=1
    else:
        vac=0
print("Rooms are cleaned!")

```

Output:

```

i/2room.py
Enter the status of the room(0 for clean; 1 for dirty):
Status of the room 0:1
Status of the room 1:1
1 1
0 1
0 0
Rooms are cleaned!

```

Code: (4 rooms)

```

def printArr(arr):
    for row in arr:
        print(row)
    print()

def clean(arr, x, y):
    if arr[x][y] == 1:
        arr[x][y] = 0

def check(arr):
    for row in arr:
        if 1 in row:
            return True
    return False

directions = [(0, 1), (1, 0), (0, -1), (-1, 0)]

```

```

direction_index = 0 # Start moving right
print("Enter the status of the rooms (0 for clean; 1 for dirty):")
arr1 = []
for i in range(2):
    row = []
    for j in range(2):
        a = int(input(f"Status of room ({i}, {j}): "))
        row.append(a)
    arr1.append(row)

x, y = 0, 0 #Start cleaning from the first room

while True:
    printArr(arr1)
    if not check(arr1):
        break
    clean(arr1, x, y)

    dx, dy = directions[direction_index]
    new_x, new_y = x + dx, y + dy

    if 0 <= new_x < 2 and 0 <= new_y < 2:
        x, y = new_x, new_y
    else:
        direction_index = (direction_index + 1) % 4
        dx, dy = directions[direction_index]
        x, y = x + dx, y + dy #Move in the new direction
print("All rooms are cleaned!")

```

### Output:

```

Enter the status of the rooms (0 for clean; 1 for dirty):
Status of room (0, 0): 0
Status of room (0, 1): 0
Status of room (1, 0): 0
Status of room (1, 1): 1
[0, 0]
[0, 1]

[0, 0]
[0, 1]

[0, 0]
[0, 1]

[0, 0]
[0, 0]

All rooms are cleaned!

```

## LAB 3: 8 Puzzle game

### Algorithm:

LAB-3:

08/10/2024

→ write an algorithm & a program to solve 8 puzzle game.

Step 1: Initialize goal state and possible moves.

goal-state =  $[[2, 2, 3], [4, 5, 6], [7, 8, -]]$ .

moves =  $[(0, -1), (1, 0), (0, 1), (0, 1)]$

Step 2: Function to calculate "Manhattan distance".

def manhattan(state):

for i in range(3):

for j in range(3):

if (state[i][j] != '-')

goal-i, goal-j = divmod(state[i][j] - 1, 3).

distance += (abs(i - goal-i) + abs(j - goal-j))

end if

end for

end for

return distance.

Step 3: check if current-state is goal-state:

def check(state):

return goal-state == current-state

Step 4: DFS new neighbours

def neighbours(state):

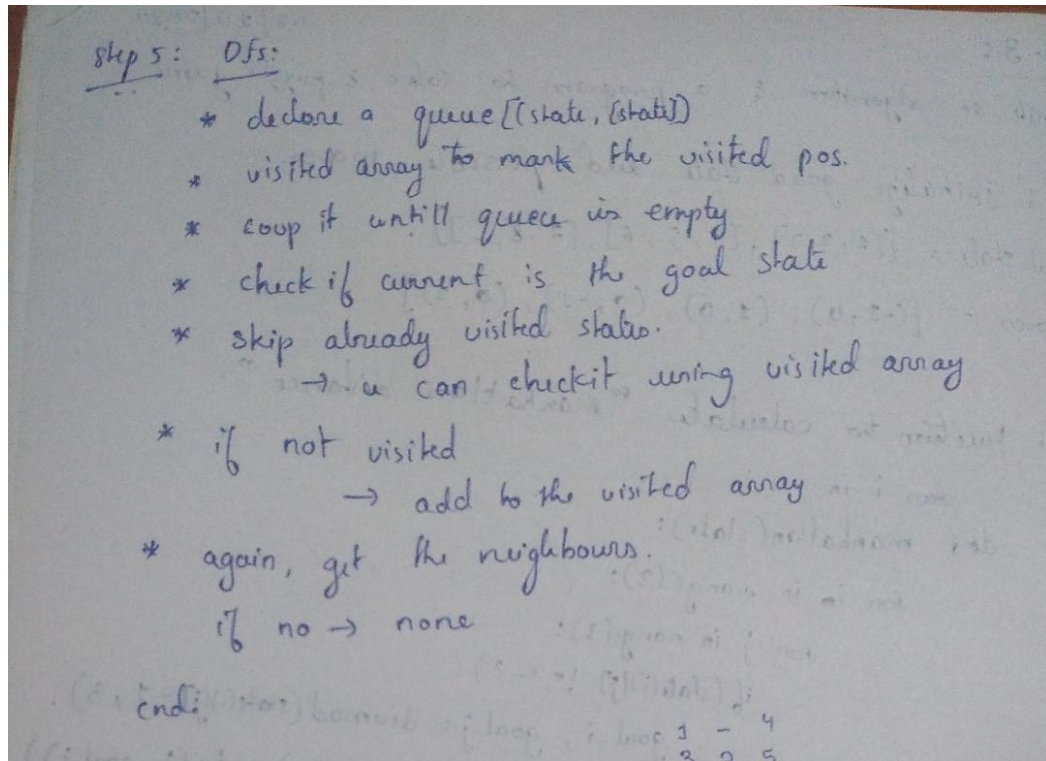
\* Loop the matrix from i to 3 & j to 3

\* for where, if state[i][j] == '-' → then check all the four directions

\* using the moves matrix, you can do that

\* add those to a new array

\* return the neighbours array



Code:

```
class PuzzleState:
    def __init__(self, board, moves=0, previous=None):
        self.board = board
        self.moves = moves
        self.previous = previous
        self.empty_pos = self.find_empty()

    def find_empty(self):
        for i in range(3):
            for j in range(3):
                if self.board[i][j] == 0:
                    return (i, j)

    def manhattan_distance(self):
        dist = 0
        for i in range(3):
            for j in range(3):
                tile = self.board[i][j]
                if tile != 0:
                    target_x = (tile - 1) // 3
                    target_y = (tile - 1) % 3
                    dist += abs(i - target_x) + abs(j - target_y)
        return dist
```

```

def generate_moves(self):
    moves = []
    x, y = self.empty_pos
    directions = [(1, 0), (-1, 0), (0, 1), (0, -1)]

    for dx, dy in directions:
        new_x, new_y = x + dx, y + dy
        if 0 <= new_x < 3 and 0 <= new_y < 3:
            new_board = [row[:] for row in self.board]
            new_board[x][y], new_board[new_x][new_y] = new_board[new_x][new_y],
new_board[x][y]
            moves.append(PuzzleState(new_board, self.moves + 1, self))

    return moves

def dfs(start_board, max_depth):
    stack = [PuzzleState(start_board)]
    visited = set()
    goal_state = [[1, 2, 3], [4, 5, 6], [7, 8, 0]]

    while stack:
        current_state = stack.pop()

        if current_state.board == goal_state:
            return current_state

        visited.add(tuple(map(tuple, current_state.board)))

        if current_state.moves < max_depth:
            for next_state in current_state.generate_moves():
                if tuple(map(tuple, next_state.board)) not in visited:
                    if next_state.manhattan_distance() < 10:
                        stack.append(next_state)

    return None

def print_solution(solution):
    path = []
    while solution:
        path.append(solution.board)
        solution = solution.previous
    for step in reversed(path):
        for row in step:
            print(row)
    print()

```

```

        print(f"Total moves taken to reach the final state: {len(path) - 1}")

initial_board = [[1, 2, 3], [4, 0, 5], [7, 8, 6]]
max_depth = 10
solution = dfs(initial_board, max_depth)
if solution:
    print("Solution found:")
    print_solution(solution)
else:
    print("No solution found.")

```

**Output:**

```

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

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

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

Total moves taken to reach the final state: 2

```



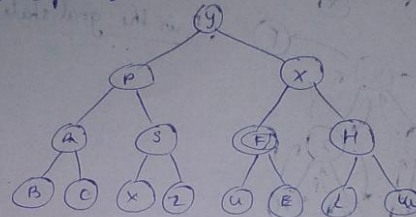
## LAB 4: Iterative deepening search and A\* algorithm

### Algorithm:

LAB - 04:

write an algorithm & code for Iterative deepening depth first search and solve 8 puzzle using A\* Algorithm.

→ Iterative deepening DFS:



\* Iterative deepening DFS is a combination of DFS & BFS. it goes to each level and then does DFS till the level.

\* Step 1: call the limit search function from range(3, max-size)

\* take the goal as i/p.

\* take the graph as i/p.

\* Step 2:

```
def IDDFS(graph, limit, head):
```

```
    for max-depth = 0 to limit:
```

```
        result = DFS(start, domain-head, max-depth).
```

```
        if result:
```

```
            return result
```

```
        else:
```

```
            return 0
```

```
        end if
```

```
    end for
```

\* Step 3:

```
def DepthFirstSearch(head, max-depth, limit):
```

```
    if root head == goal:
```

```
        return head.
```

```
    if depth == limit + 1:
```

```
        return
```

```
    for child in root:
```

```
        dfs(child, max-depth, limit).
```



step 1: define the function Astar with present state & goal state as parameters

```

def Astar(startstate, goalstate):
    for i in range(1, max-depth):
        cost = 0: // g(n).
        if startstate != goalstatevisited[]:
            state[] = generate(startstate).
        for i in states:
            f = cost + manhattan(state, goalstate).
            min = min(min, f). // finds min
        cost++
        visited.append(startstate).
    Astar(state, goalstate)

```

Code:

```

def iterative_deepening_search(graph, start, goal):
    def depth_limited_search(node, goal, depth):
        if depth == 0:
            if node == goal:
                return [node]
            else:
                return None
        elif depth > 0:
            for child in graph.get(node, []):
                result = depth_limited_search(child, goal, depth - 1)
                if result is not None:
                    return [node] + result
            return None
        depth = 0
        while True:
            result = depth_limited_search(start, goal, depth)
            if result is not None:
                return result
            depth += 1
    def get_user_input_graph():
        graph = {}
        num_edges = int(input("Enter the number of edges: "))

```

```

print("Enter each edge in the format 'node1 node2':")
for _ in range(num_edges):
    node1, node2 = input().split()
    if node1 in graph:
        graph[node1].append(node2)
    else:
        graph[node1] = [node2]
    if node2 in graph:
        graph[node2].append(node1)
    else:
        graph[node2] = [node1]
return graph
def main():
    graph = get_user_input_graph()
    start_node = input("Enter the starting node: ")
    goal_node = input("Enter the goal node: ")
    path = iterative_deepening_search(graph, start_node, goal_node)
    if path:
        print(f"Path found: {' -> '.join(path)}")
    else:
        print("No path found")
if __name__ == "__main__":
    main()

```

```

def H_n(state, target):
    return sum(x != y for x, y in zip(state, target))
def F_n(state_with_lvl, target):
    state, lvl = state_with_lvl
    return H_n(state, target) + lvl
def possible_moves(state_with_lvl, visited_states):
    state, lvl = state_with_lvl
    b = state.index(0)
    directions = []
    pos_moves = []
    if b <= 5: directions.append('d')
    if b >= 3: directions.append('u')
    if b % 3 > 0: directions.append('l')
    if b % 3 < 2: directions.append('r')
    for move in directions:
        temp = gen(state, move, b)
        if temp not in visited_states:
            pos_moves.append([temp, lvl + 1])
    return pos_moves
def gen(state, move, b):
    temp = state.copy()

```

```

    if move == 'l': temp[b], temp[b - 1] = temp[b - 1], temp[b]
    if move == 'r': temp[b], temp[b + 1] = temp[b + 1], temp[b]
    if move == 'u': temp[b], temp[b - 3] = temp[b - 3], temp[b]
    if move == 'd': temp[b], temp[b + 3] = temp[b + 3], temp[b]
    return temp
def display_state(state):
    print("Current State:")
    for i in range(0, 9, 3):
        print(state[i:i+3])
    print()
def astar(src, target):
    arr = [[src, 0]]
    visited_states = []
    iterations = 0
    while arr:
        iterations += 1
        current = min(arr, key=lambda x: F_n(x, target))
        arr.remove(current)
        display_state(current[0])
        if current[0] == target:
            return f'Found with {iterations} iterations'
        visited_states.append(current[0])
        arr.extend(possible_moves(current, visited_states))
    return 'Not found'
src = [1, 2, 3, 8, 0, 4, 7, 6, 5]
target = [2, 8, 1, 0, 4, 3, 7, 6, 5]
print(astar(src, target))

```

## Output:

```

i/iter.py
Enter the number of edges: 4
Enter each edge in the format 'node1 node2':
a b
a c
b d
c e
Enter the starting node: a
Enter the goal node: e
Path found: a -> c -> e

```

```

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

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

Found with 40 iterations

```

## LAB 5: Simulated annealing.

### Algorithm:

LAB-05:  
Simulation of Annealing:

objective function:  $x^2 + 5 \sin x$

step 1: define a function called "simulation-Annealing"

```
def simulation-annealing(initialState, initialTemp, coolingRate, it):  
    current = initialState  
    best = current  
    best = objective(current)  
    temp = initialTemp  
    while Temp > 1:  
        for i <- 1 to it:  
            new = neighbour(current)  
            curr = objective(current)  
            new-cost = object(new)  
            if Absolutefail  
            if Function (curr, new-cost, temp) > Rand(0, 1):  
                current = new.  
            if new < best:  
                best = new.  
            temp *= coolingRate  
    return (best, best-cost).
```

step 2: Now define a objective function to change the state.

```
def objective(state):  
    cost = 0  
    for ele in state:  
        cost += ele**2 + sin(ele)  
    return cost.
```

step 3: next function is to check / search for neighbours..

```
def neighbour(state):  
    new = state.copy()  
    ind = Rand(0, Len(state)-1)  
    new[ind] += Rand(-1, 1).  
    return new.
```

*Ans 22/11/2020*

Step 4: a function for acceptance probability.

```

def Function(curr_cost, new_cost, temp):
    if (new_cost < curr_cost):
        return 1
    else:
        return e $\frac{(\text{new\_cost} - \text{current\_cost})}{T}$ 

```

Code:

```

import random
import math

def energy(x):
    return x ** 2 + 5 * math.sin(x) + math.exp(-x)

def adaptive_simulated_annealing(start, temp, cooling_rate, lower_limit,
upper_limit):
    current = start
    current_energy = energy(current)

    while temp > 1:
        # Adaptive step size based on temperature (larger steps when hot)
        step_size = random.uniform(-1, 1) * temp
        new = current + step_size

        # Ensure new solution is within bounds
        if new < lower_limit or new > upper_limit:
            continue

        new_energy = energy(new)

        # If the new spot is better, move there
        if new_energy < current_energy:
            current = new
            current_energy = new_energy
        else:
            # Acceptance probability (explore worse spots)
            probability = math.exp((current_energy - new_energy) / temp)
            if random.uniform(0, 1) < probability:

```

```

        current = new
        current_energy = new_energy

    # Adaptive cooling based on progress
    if abs(new_energy - current_energy) < 0.01:
        temp *= 0.98 # Slow cooling near solution
    else:
        temp *= cooling_rate

    return current

# Run the simulation multiple times from different starting points
best_solution = None
for _ in range(10): # 10 runs
    result = adaptive_simulated_annealing(start=random.uniform(-10, 10), temp=100,
    cooling_rate=0.99, lower_limit=-10, upper_limit=10)
    if best_solution is None or energy(result) < energy(best_solution):
        best_solution = result

print(f"Best solution found: {best_solution}")

```

Output:

```

i/anneal.py
Best solution found: -0.7390095302681031

```



## LAB 6: 8 queens using Hill Climb and A\* Algorithm.

### Algorithm:

\* LAB -6.

1. 8 queens using Hill climbing:

Step 1: define the init function/method to initialize table & size

```
def __init__(self, size, state):
```

```
    self.size = size
```

```
    self.state = state
```

Step

```
    state = random() // generates / allocates random  
                        position for queens
```

```
    eg: [2, 0, 4, 7, 1, 3, 6, 5]
```

Step 2: generate the conflicts

```
    for i → 0 to size:
```

```
        for j → 0 to size:
```

```
            if conflicts = 0:
```

```
                if (state[i] == state[j] && abs(state[i] - state[j]) == abs(i - j))
```

```
                    conflict++
```

```
    return conflicts
```

Step 3: generate the best successors for the current state:

```
def get-best-successor(self):
```

```
    best-conflict = self.get-conflict()
```

```
    best-state = state[i] // current state
```

```
    if (state[now] == col): // queen is not in target pos
```

```
        for i in row:
```

```
            for j in col:
```

```
                if (state[now] == best):
```

```
                    // create a new state & initialize it to
```

```
                    that state
```

```
                    new-state = state
```

```
                    new-state[now] = col
```

```
                    // calculate new conflict
```

```
                    new-conflict = self.get-conflict()
```

```
                    if new-conflict < best-conflict:
```

```
                        new-best-conflict = new-conflict
```

```
                        best-state = new-state
```

```
    return best-state, best-conflict
```

Step 4: def hill climbing (self):  
 current-conflict = get-conflict()  
 new-state, new-conflict = self.get-<sup>get-best-successor</sup>successor()  
 if (new-conflict < current-conflict):  
 return self.state : if current-conflict == 0 then done  
 self.state = new-state

O/p:

## 8. 8 queens using A\* algorithm:

Step 1: initialize:  
 def \_\_init\_\_(self, state=None):  
 self.size = size // 8  
 self.state = random.choice(range(0, self.size)) // allocate random positions for queen

Step 2: is-goal (self):  
 return self.get-conflict == 0

Step 3: get-conflict (self):  
 for i → 0 to size:  
 for j → 0 to size:  
 conflicts = 0  
 if (check for same col & diagonal)  
 conflict++  
 return conflict

Step 4: def get-cost (self): // calls the conflict function  
 return get-conflict()



steps:  
 def get\_successor(state, size):  
     // run two loops to find the successors  
     // store them in successor array  
     successor = board(new-state)  
     - successors.append(Successor.getcost, successor)  
  
steps: // initialize board  
     initial board = board()  
     open set = []  
     heapq.heappush(initial board, open set, (initial board.getcost(), initial  
     while ~~open~~ open set:  
         current cost, board = heapq.heappop(open set)  
         if board.isgoal():  
             return board.state  
         for cost, successor in board.get\_successor():  
             heapq.heappush(open set, (cost + 1), successor).  
  
 op: *Ans 29/10/20*

**Code:**

```

import random

# Helper function to calculate the heuristic (number of conflicts)
def heuristic(board):
    conflicts = 0
    for i in range(len(board)):
        for j in range(i + 1, len(board)):
            if board[i] == board[j] or abs(board[i] - board[j]) == j - i:
                conflicts += 1
    return conflicts

# Hill climbing for 8-queens
def hill_climbing_8_queens():
    n = 8
    # Generate a random initial state
    board = [random.randint(0, n - 1) for _ in range(n)]

```

```

while True:
    current_h = heuristic(board)
    if current_h == 0:
        return board # Solution found

    # Find the best neighbor by moving each queen to every other column in its
row
    best_board = board[:]
    best_h = current_h
    for row in range(n):
        for col in range(n):
            if col == board[row]:
                continue
            new_board = board[:]
            new_board[row] = col
            new_h = heuristic(new_board)

            # If the new board has fewer conflicts, update the best board
            if new_h < best_h:
                best_h = new_h
                best_board = new_board

    # If no improvement, we're stuck in a local minimum; restart
    if best_h >= current_h:
        board = [random.randint(0, n - 1) for _ in range(n)]
    else:
        board = best_board

# Run hill climbing search
solution = hill_climbing_8_queens()
print("Solution board (column positions for each row):", solution)

```

```

import heapq

# Helper function to calculate the heuristic (number of conflicts)
def heuristic(board):
    conflicts = 0
    for i in range(len(board)):
        for j in range(i + 1, len(board)):
            if board[i] == board[j] or abs(board[i] - board[j]) == j - i:
                conflicts += 1
    return conflicts

# A* Search for 8-queens
def a_star_8_queens():

```

```

n = 8
open_set = []
# Initial state: empty board
heapq.heappush(open_set, (0, [])) # (f, board)

while open_set:
    f, board = heapq.heappop(open_set)

    # Goal check
    if len(board) == n and heuristic(board) == 0:
        return board

    # Generate successors
    row = len(board)
    for col in range(n):
        new_board = board + [col]
        if heuristic(new_board) == 0: # No conflicts so far
            g = row + 1
            h = heuristic(new_board)
            heapq.heappush(open_set, (g + h, new_board))

    return None # No solution found

# Run A* search
solution = a_star_8_queens()
print("Solution board (column positions for each row):", solution)

```

## Output:

```

i/Hill.py
Solution board (column positions for each row): [3, 1, 6, 4, 0, 7, 5, 2]
PS C:\Users\Dell\Desktop\BTS>

```

```

i/AA.py
Solution board (column positions for each row): [0, 4, 7, 5, 2, 6, 1, 3]
PS C:\Users\Dell\Desktop\BTS>

```

## LAB 7: knowledge base using propositional logic

### Algorithm:

LAB-07:

→ Propositional Logic:

1)  $P \rightarrow Q$  (if  $P$  is true, then  $Q$  must be true).

2)  $P$  (we know  $P$  is true).

→ we can infer  $Q$ .

→ Formal Representations of Entailment:

$P, (P \rightarrow Q) \models Q$ .

"given  $P \in P \rightarrow Q$ , it logically follows that  $Q$  must be true". ( $\models$  represents logical entailment)

Step 1:

Knowledge Base:

- 1) Alice is the mother of Bob
- 2) Bob is the father of Charlie
- 3) A father is a parent
- 4) A mother is a parent
- 5) All parents have children
- 6) if someone is a parent, their children are siblings
- 7) Alice is married to David

Hypothesis:

- "charlie is a sibling of Bob".

1)  $M(Alice, Bob)$

2)  $F(Bob, Charlie)$

3)  $Father(x) \rightarrow Parent(x)$

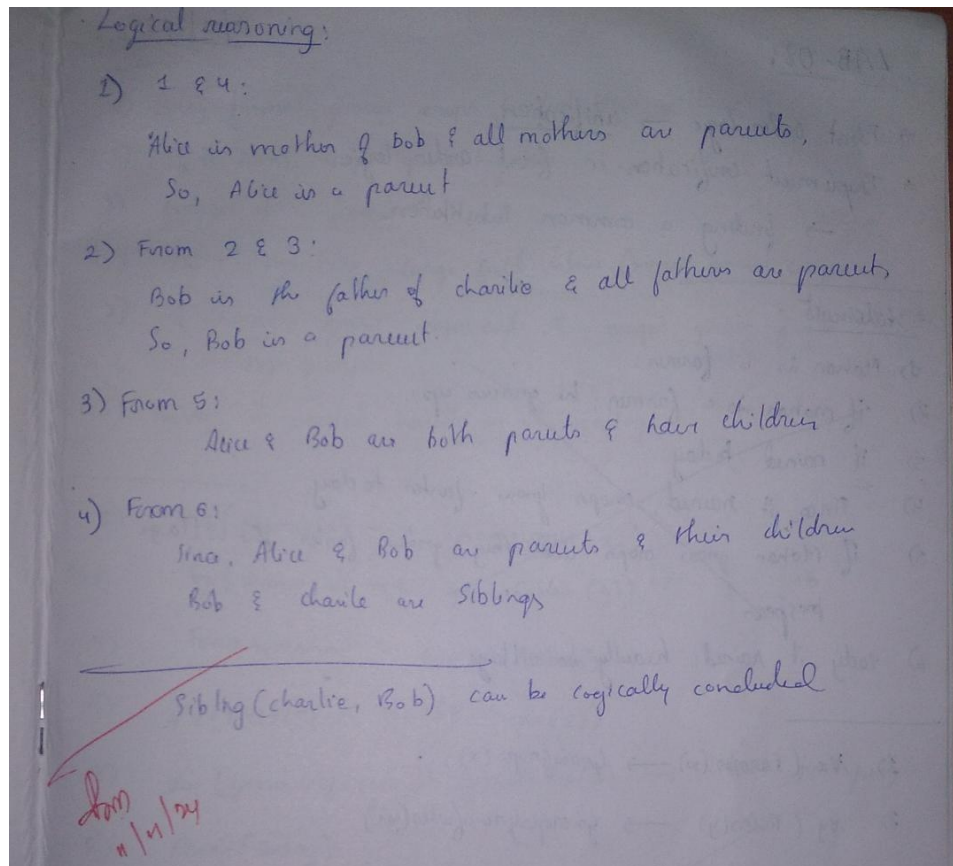
4)  $Mother(x) \rightarrow Parent(x)$

5)  $Parent(x) \rightarrow HasChildren(x)$

6)  $Parent(x) \wedge HasChildren(x)$

→ ~~sibling(children(x))~~

~~7) Married(Alice, David)~~



### Code:

```
# Define the knowledge base
knowledge_base = {
    "parent": {
        "Alice": ["Bob"], # Alice is the parent of Bob
        "Bob": ["Charlie"] # Bob is the parent of Charlie
    },
    "married": {
        "Alice": "David" # Alice is married to David
    }
}

# Define helper functions
def is_parent(parent, child):
    """Check if a person is a parent of a given child."""
    return child in knowledge_base["parent"].get(parent, [])

def get_children(parent):
    """Retrieve all children of a given parent."""
    return knowledge_base["parent"].get(parent, [])
```

```

def are_siblings(person1, person2):
    """Check if two people are siblings."""
    for parent, children in knowledge_base["parent"].items():
        if person1 in children and person2 in children:
            return True
    return False

# Hypothesis: "Charlie is a sibling of Bob"
hypothesis = are_siblings("Charlie", "Bob")

# Solve and print the result
if hypothesis:
    print("The hypothesis 'Charlie is a sibling of Bob' is TRUE.")
else:
    print("The hypothesis 'Charlie is a sibling of Bob' is FALSE.")

```

**Output:**

```

i/entail.py
The hypothesis 'Charlie is a sibling of Bob' is FALSE.

```



## LAB 8: Unification in First Order Logic.

### Algorithm:

Statements:

- 1) Every farmer grows crops to feed their village
- 2) if it rains, crops grow together
- 3) Mohan is a farmer
- 4) The village has large field where farmers grow crops
- 5) if someone grows crops and the crops grow faster, village prospers
- 6) Today, it rained heavily in the village

- 1)  $\forall x (\text{Farmer}(x) \rightarrow \text{growscrops}(x))$
- 2)  $\forall y (\text{Rains}(y) \rightarrow \text{growsFaster}(y))$
- 3)  $\text{Farmer}(\text{Mohan}) \wedge \text{Livesin}(\text{Mohan})$
- 4)  $\exists z (\text{Field}(z) \wedge \text{UsedFor}(\text{crops}(z)))$
- 5)  $\forall w (\text{growscrops}(w) \wedge \text{growsFaster}(w) \rightarrow \text{village prospers})$
- 6)  $\text{Rains}(\text{Today})$

- 2) ③  $\rightarrow$  ~~note~~  $\text{Farmer}(\text{Mohan})$
- ①  $\rightarrow$  if Mohan is a farmer, he grows crops ( ).
- ⑥  $\rightarrow$  it rained today ( $\text{Rains}(\text{Today})$ )
- ②  $\rightarrow$   $\text{growsgrowsFaster}(\text{Today})$
- ⑤  $\rightarrow$  if Mohan grows crops.
- ⑥ By combining above  
 $\text{villageProspers}(\text{Mohan}, \text{Today}) = \text{TRUE}$

Code:

```
def unify(x, y, subst=None):
    """
    Unification Algorithm: Unifies two terms, X and Y.
    """
    if subst is None:
        subst = {}

    if x == y: # Step 1(a): If X and Y are identical
        return subst
    elif isinstance(x, str) and x.islower(): # Step 1(b): If X is a variable
        return unify_variable(x, y, subst)
    elif isinstance(y, str) and y.islower(): # Step 1(c): If Y is a variable
        return unify_variable(y, x, subst)
    elif isinstance(x, tuple) and isinstance(y, tuple): # Step 2: Check predicates
        and arguments
        if x[0] != y[0] or len(x) != len(y): # Predicate symbol or argument count
            mismatch
            return None
        for x_i, y_i in zip(x[1:], y[1:]): # Step 5: Recurse through arguments
            subst = unify(x_i, y_i, subst)
            if subst is None:
                return None
        return subst
    else:
        return None # Step 1(d): Failure case

def unify_variable(var, x, subst):
    """
    Unify variable with another term.
    """
    if var in subst:
        return unify(subst[var], x, subst)
    elif occurs_check(var, x, subst): # Check if var occurs in x
        return None
    else:
        subst[var] = x
        return subst

def occurs_check(var, x, subst):
    """
    Check if a variable occurs in a term.
    """
```



```

    if var == x:
        return True
    elif isinstance(x, tuple):
        return any(occurs_check(var, xi, subst) for xi in x)
    elif isinstance(x, str) and x in subst:
        return occurs_check(var, subst[x], subst)
    return False

# Test cases for unification
x1 = ("P", "a", "x")
y1 = ("P", "a", "b")

x2 = ("Q", "x", ("R", "x"))
y2 = ("Q", "a", ("R", "a"))

print("Unifying", x1, "and", y1, "=>", unify(x1, y1)) #
print("Unifying", x2, "and", y2, "=>", unify(x2, y2))

```

**Output:**

```

i/unify.py
Unifying ('P', 'a', 'x') and ('P', 'a', 'b') => {'x': 'b'}
Unifying ('Q', 'x', ('R', 'x')) and ('Q', 'a', ('R', 'a')) => {'x': 'a'}

```

## LAB 9: knowledge base consisting of first order logic.

### Algorithm:

LAB-09

First order logic  $\rightarrow$  Prove the given query using forward reasoning

$\rightarrow$  Starts with a base state and uses the inference rules and available knowledge in the forward direction till it reaches the goal state

---

Question:

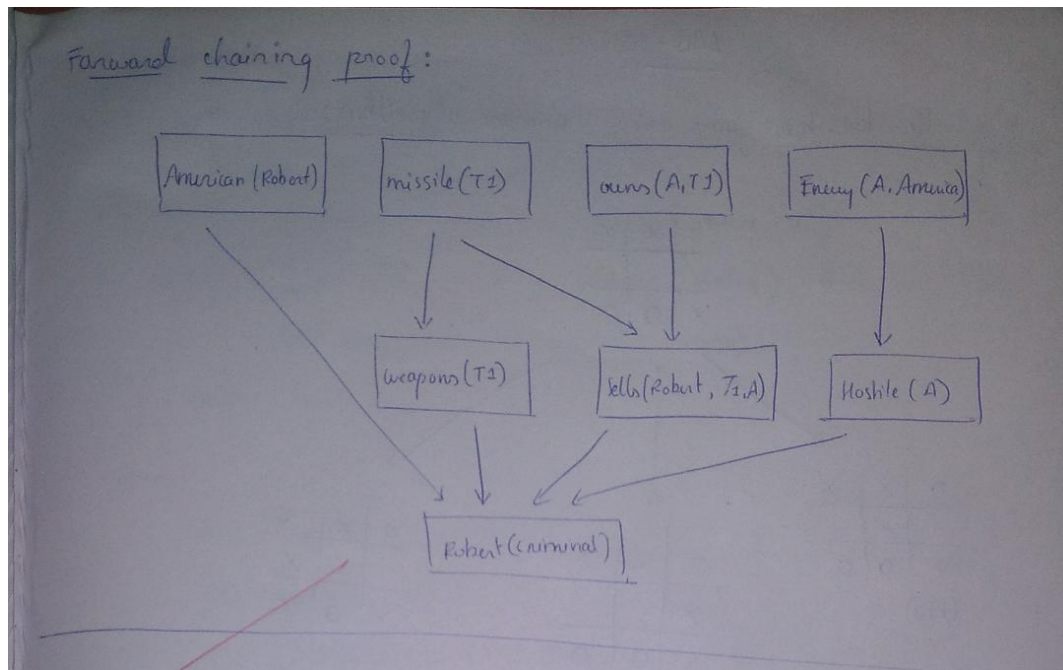
As per the law, it is a crime for an american to sell weapons to hostile nation. Country A, an enemy of America, has some missiles, and all the missiles were sold to it by Robert, who is an american citizen

$\rightarrow$  Prove that - "Robert is criminal".

---

Given facts:

- $\rightarrow$  Robert is a american  $\Rightarrow$  American(Robert)
- $\rightarrow$  country A is enemy of america.  $\Rightarrow$  Enemy(A, america).
- $\rightarrow$  It is a crime to sell weapons to hostile nations  $\Rightarrow$   
 $American(p) \wedge weapons(q) \wedge sells(p, q, n) \wedge Hostile(n) \Rightarrow criminal(p)$
- $\rightarrow$  country A has some missiles  
 $\exists x owns(A, x) \wedge missile(x)$
- $\rightarrow$  owns(A, T1)
- $\rightarrow$  missile(T1).
- $\rightarrow$  All of missiles were sold to country A by Robert  
 $\forall x Missiles(x) \wedge owns(A, x) \Rightarrow sells(Robert, x, A)$
- $\rightarrow$  Missiles are weapons.  
 $Missiles(x) \Rightarrow weapons(x)$
- $\rightarrow$  Enemy of america is known as hostile.  
 $\forall x Enemy(x, America) \Rightarrow Hostile(x)$



Code:

```

# Define the knowledge base (KB) as a set of facts
KB = set()

# Premises based on the provided FOL problem
KB.add('American(Robert)')
KB.add('Enemy(America, A)')
KB.add('Missile(T1)')
KB.add('Owns(A, T1)')

# Define inference rules
def modus_ponens(fact1, fact2, conclusion):
    """ Apply modus ponens inference rule: if fact1 and fact2 are true, then
    conclude conclusion """
    if fact1 in KB and fact2 in KB:
        KB.add(conclusion)
        print(f"Inferred: {conclusion}")

def forward_chaining():
    """ Perform forward chaining to infer new facts until no more inferences can be
    made """
    # 1. Apply: Missile(x) → Weapon(x)
    if 'Missile(T1)' in KB:
        KB.add('Weapon(T1)')
        print(f"Inferred: Weapon(T1)")

    # 2. Apply: Sells(Robert, T1, A) from Owns(A, T1) and Weapon(T1)

```

```

if 'Owns(A, T1)' in KB and 'Weapon(T1)' in KB:
    KB.add('Sells(Robert, T1, A)')
    print(f"Inferred: Sells(Robert, T1, A)")

# 3. Apply: Hostile(A) from Enemy(A, America)
if 'Enemy(America, A)' in KB:
    KB.add('Hostile(A)')
    print(f"Inferred: Hostile(A)")

# 4. Now, check if the goal is reached (i.e., if 'Criminal(Robert)' can be
inferred)
if ('American(Robert)' in KB and
    'Weapon(T1)' in KB and
    'Sells(Robert, T1, A)' in KB and
    'Hostile(A)' in KB):
    KB.add('Criminal(Robert)')
    print("Inferred: Criminal(Robert)")

# Check if we've reached our goal
if 'Criminal(Robert)' in KB:
    print("Robert is a criminal!")
else:
    print("No more inferences can be made.")

# Run forward chaining to attempt to derive the conclusion
forward_chaining()

```

Output:

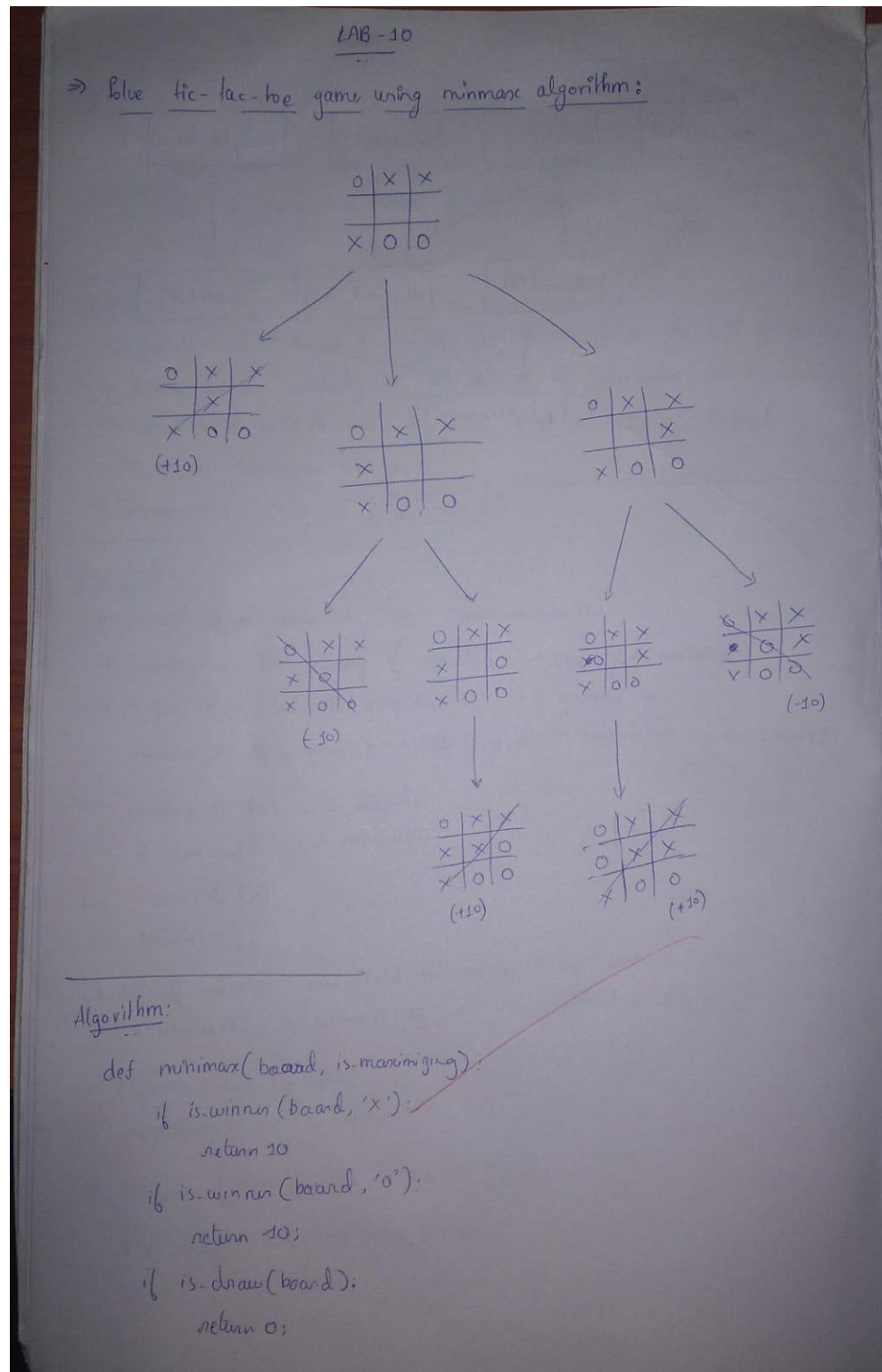
```

i/forward.py
Inferred: Weapon(T1)
Inferred: Sells(Robert, T1, A)
Inferred: Hostile(A)
Inferred: Criminal(Robert)
Robert is a criminal!

```

## LAB 10: Min-max algorithm and alpha-beta pruning.

Algorithm:





if is\_maximizing:

best\_score = -float('inf')

for move in possible\_moves(board):

new\_board = make\_move(board, move, 'x')

score = minimax(new\_board, False) # 'o' turn

best\_score = max(best\_score, score)

return best\_score

else:

best\_score = float('inf')

for move in possible\_moves(board):

new\_board = make\_move(board, move, 'o')

~~new\_board =~~

score = minimax(new\_board, True) # 'x' turn

best\_score = min(best\_score, score)

return best\_score

---

def make\_move(board, move, player):

new\_board = board[:]

new\_board[move] = player

return new\_board

---

INB-11

→ solve 8 queens problems using alpha-beta pruning.

Algorithm:

```
def alpha_beta(self, board, col, alpha, beta, maximizing_player):
```

```
    if col >= self.size:
```

```
        return 0, [row[:] for row in board]
```

```
    if maximizing_player:
```

```
        max_eval = float('-inf')
```

```
        best_board = None
```

```
        for row in range(self.size):
```

```
            if self.is_safe(board, row, col):
```

```
                board[row][col] = 1
```

```
                eval_score, potential_board = self.alpha_beta_search
```

```
                    (board, col+1, alpha, beta, False)
```

```
                board[row][col] = 0
```

```
                if eval_score > max_eval:
```

```
                    max_eval = eval_score
```

```
                    best_board = potential_board
```

```
                alpha = max(alpha, eval_score)
```

```
                if beta <= alpha
```

```
                    break
```

```
        return max_eval, best_board
```

```
    else:
```

```
        min_eval = float('inf')
```

```
        beta = min(beta, eval_score)
```

```
        if beta <= alpha:
```

```
            break
```

```
    return min_eval, best_board
```

## Code:

```
import math

# Constants for the players
AI = 'X'
HUMAN = 'O'
EMPTY = '_'

# Function to print the board
def print_board(board):
    for row in board:
        print(" ".join(row))
    print()

# Function to check if a player has won
def check_winner(board, player):
    # Check rows, columns, and diagonals
    for row in board:
        if all(cell == player for cell in row):
            return True
    for col in range(3):
        if all(row[col] == player for row in board):
            return True
    if all(board[i][i] == player for i in range(3)) or all(board[i][2 - i] == player
for i in range(3)):
        return True
    return False

# Function to check if the game is a draw
def is_draw(board):
    return all(cell != EMPTY for row in board for cell in row)

# Minimax algorithm
def minimax(board, depth, is_maximizing):
    if check_winner(board, AI):
        return 10 - depth
    if check_winner(board, HUMAN):
        return depth - 10
    if is_draw(board):
        return 0

    if is_maximizing:
        best_score = -math.inf
        for i in range(3):
```



```

        for j in range(3):
            if board[i][j] == EMPTY:
                board[i][j] = AI
                score = minimax(board, depth + 1, False)
                board[i][j] = EMPTY
                best_score = max(best_score, score)
    return best_score
else:
    best_score = math.inf
    for i in range(3):
        for j in range(3):
            if board[i][j] == EMPTY:
                board[i][j] = HUMAN
                score = minimax(board, depth + 1, True)
                board[i][j] = EMPTY
                best_score = min(best_score, score)
    return best_score

# Function to find the best move for AI
def find_best_move(board):
    best_score = -math.inf
    move = (-1, -1)
    for i in range(3):
        for j in range(3):
            if board[i][j] == EMPTY:
                board[i][j] = AI
                score = minimax(board, 0, False)
                board[i][j] = EMPTY
                if score > best_score:
                    best_score = score
                    move = (i, j)
    return move

# Example usage
if __name__ == "__main__":
    # Initialize a sample board
    board = [
        ['X', 'O', 'X'],
        ['O', 'X', 'O'],
        ['_', '_', '_']
    ]
    print("Current Board:")
    print_board(board)

    best_move = find_best_move(board)
    print(f"The best move for AI is: {best_move}")

```

```

class EightQueens:
    def __init__(self, size=8):
        self.size = size

    def is_safe(self, board, row, col):
        """Check if placing a queen at board[row][col] is safe."""
        for i in range(col):
            if board[row][i] == 1: # Check this row on the left
                return False

        for i, j in zip(range(row, -1, -1), range(col, -1, -1)): # Check upper
diagonal
            if board[i][j] == 1:
                return False

        for i, j in zip(range(row, self.size), range(col, -1, -1)): # Check lower
diagonal
            if board[i][j] == 1:
                return False

        return True

    def alpha_beta_search(self, board, col, alpha, beta, maximizing_player):
        """Alpha-Beta Pruning Search."""
        if col >= self.size: # If all queens are placed
            return 0, [row[:] for row in board] # Return 0 as heuristic since it's
a valid solution

        if maximizing_player:
            max_eval = float('-inf')
            best_board = None
            for row in range(self.size):
                if self.is_safe(board, row, col):
                    board[row][col] = 1
                    eval_score, potential_board = self.alpha_beta_search(board, col
+ 1, alpha, beta, False)
                    board[row][col] = 0
                    if eval_score > max_eval:
                        max_eval = eval_score
                        best_board = potential_board
                    alpha = max(alpha, eval_score)
                    if beta <= alpha: # Beta cutoff
                        break
            return max_eval, best_board
        else:
            min_eval = float('inf')

```

```

        best_board = None
        for row in range(self.size):
            if self.is_safe(board, row, col):
                board[row][col] = 1
                eval_score, potential_board = self.alpha_beta_search(board, col
+ 1, alpha, beta, True)
                board[row][col] = 0
                if eval_score < min_eval:
                    min_eval = eval_score
                    best_board = potential_board
                beta = min(beta, eval_score)
                if beta <= alpha: # Alpha cutoff
                    break
        return min_eval, best_board

    def solve(self):
        """Solve the 8-Queens problem."""
        board = [[0] * self.size for _ in range(self.size)]
        _, solution = self.alpha_beta_search(board, 0, float('-inf'), float('inf'),
True)
        return solution

    def print_board(self, board):
        """Print the chessboard."""
        for row in board:
            print(" ".join("Q" if col else "." for col in row))
        print()

if __name__ == "__main__":
    game = EightQueens()
    solution = game.solve()
    if solution:
        print("Solution found:")
        game.print_board(solution)
    else:
        print("No solution exists.")

```

**Output:**

```
i/minmax.py
Current Board:
X O X
O X O
- - -

The best move for AI is: (2, 0)
```

```
i/alpha.py
Solution found:
. Q . . . . .
. . . . Q . . .
. . . . . . Q .
Q . . . . . . .
. . Q . . . . .
. . . . . . . Q
. . . . . Q . .
. . . Q . . . .
```