Batch: IAI-2 Experiment Number: 4

Roll Number: 16010422234 Name: Chandana Ramesh Galgali

Aim of the Experiment: Implementation of Adversarial algorithm : Min-Max for Tic-Tac-Toe Game

Program/Steps:

- 1. Implement two players, 1) AiPlayer and 2) HuPlayer [AI and Human player] for a tic-tac-toe game.
- 2. For AiPlayer implement Minmax algorithm. [For simplicity first consider the start state as given in the figure 2 below. Once the program is working fine with this start state then change the start state to blank game board.]

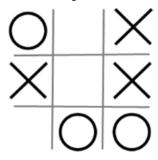


Figure 2: sample start state

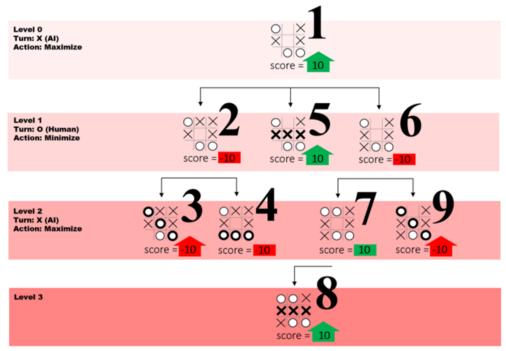


Figure 3 Minimax function call by function call

Code:

```
import math
class TicTacToe:
   def __init__(self):
       self.board = [' ']*9
       self.ai player = None
       self.hu player = None
   def print board(self):
        for i in range(0, 9, 3):
           print(self.board[i], '|', self.board[i+1], '|',
self.board[i+2])
           if i < 6:
                print('----')
   def empty cells(self):
        return [i for i, cell in enumerate(self.board) if cell == ' ']
   def check_winner(self, player):
       win conditions = [(0, 1, 2), (3, 4, 5), (6, 7, 8),
                          (0, 3, 6), (1, 4, 7), (2, 5, 8),
                          (0, 4, 8), (2, 4, 6)]
        for condition in win conditions:
            if all(self.board[i] == player for i in condition):
                return True
        return False
   def check draw(self):
       return ' ' not in self.board
   def game over(self):
        return self.check_winner(self.ai_player) or
self.check winner(self.hu player) or self.check draw()
   def minimax(self, depth, player):
       if player == self.ai player:
           best = [-1, -math.inf]
       else:
           best = [-1, math.inf]
```

```
if depth == 0 or self.game over():
        score = self.evaluate()
        return [-1, score]
    for cell in self.empty_cells():
        self.board[cell] = player
        score = self.minimax(depth - 1, 'O' if player == 'X' else 'X')
        self.board[cell] = ' '
        score[0] = cell
        if player == self.ai player:
            if score[1] > best[1]:
                best = score
        else:
            if score[1] < best[1]:</pre>
                best = score
    return best
def ai turn(self):
    depth = len(self.empty_cells())
    if depth == 0 or self.game over():
        return
    if depth == 9:
        cell = 0
    else:
        cell = self.minimax(depth, self.ai player)[0]
    self.board[cell] = self.ai player
def hu turn(self):
    while True:
        move = int(input('Enter your move (1-9): ')) - 1
        if move in self.empty_cells():
            self.board[move] = self.hu player
            break
        else:
            print('Invalid move! Try again.')
```

```
def evaluate(self):
        if self.check winner(self.ai player):
            return 1
        elif self.check winner(self.hu player):
            return -1
        else:
            return 0
   def play(self):
       print("Welcome to Tic-Tac-Toe!")
       player choice = input("Do you want to be 'X' or 'O'? ").upper()
       if player choice == 'X':
            self.ai player = '0'
            self.hu player = 'X'
       else:
            self.ai player = 'X'
            self.hu player = '0'
       print(f"You are '{self.hu player}', and AI is
{self.ai player}'.")
       print("You play by entering the number of the cell you want to
mark.")
       while not self.game over():
            self.print board()
            if self.hu player == 'X':
                self.hu turn()
            else:
                self.ai turn()
            if self.check winner(self.hu player):
                self.print board()
                print("Congratulations! You win!")
                break
            elif self.check_draw():
                self.print board()
                print("It's a draw!")
                break
```

```
if self.check winner(self.ai player):
                self.print_board()
                print("AI wins! Better luck next time.")
                break
            elif self.check draw():
                self.print_board()
                print("It's a draw!")
                break
            self.ai turn()
            if self.check_winner(self.ai_player):
                self.print board()
                print("AI wins! Better luck next time.")
            elif self.check draw():
                self.print board()
                print("It's a draw!")
                break
            self.print board()
            if self.check_winner(self.hu_player):
                print("Congratulations! You win!")
                break
            elif self.check draw():
                print("It's a draw!")
               break
if name == " main ":
    game = TicTacToe()
    game.play()
```

Output/Result:

```
Welcome to Tic-Tac-Toe!
Do you want to be 'X' or '0'? X
You are 'X', and AI is 'O'.
You play by entering the number of the cell you want to mark.
Enter your move (1-9): 1
x | |
  0
 0
Enter your move (1-9): 9
x | 0 |
 101
| | X
 101
 | | x
Enter your move (1-9): 8
x | o |
 0 |
0 | X | X
X | 0 |
 0
0 | X | X
Enter your move (1-9): 3
x \mid o \mid x
 000
0 | X | X
X | 0 | X
 000
0 | X | X
Enter your move (1-9): 4
x | 0 | x
x | 0 | 0
0 | X | X
It's a draw!
```

Post-Lab Questions:

- 1. Game playing is often called as an
- a) Non-adversarial search
- b) Adversarial search
- c) Sequential search
- d) None of the above
- 2. What are the basic requirements or needs of AI search methods in game playing?
- a) Initial State of the game
- b) Operators defining legal moves
- c) Successor functions
- d) Goal test
- e) Path cost

Outcomes: Analyze and formalize the problem (as a state space, graph, etc.) and select the appropriate search method and write the algorithm

Conclusion (Based on the Results and outcomes achieved):

The implementation of the Min-Max algorithm for the Tic-Tac-Toe game demonstrated its effectiveness in achieving optimal decision-making in adversarial environments. While exhibiting promising results, further research and experimentation are warranted to address scalability challenges and explore avenues for algorithmic improvements in more complex game domains.

References:

- 1. Stuart Russell and Peter Norvig, Artificial Intelligence: A Modern Approach, Second Edition, Pearson Publication
- 2. Luger, George F. Artificial Intelligence: Structures and strategies for complex problem solving, 2009, 6th Edition, Pearson Education