

## Excercise 3 - Tic-Tac-Toe Game using Minimax and AlphaBeta Pruning Search

**Aim:** Implement a Tic-Tac-Toe Game using Minimax and Alpha-Beta Pruning Search Methods

### Source Code

```
import math

class TicTacToeGame:

    def __init__(self):
        self.board = [' ' for _ in range(9)]
        self.human = 'X'
        self.ai = 'O'
        self.current_player = self.human
        self.states_explored = 0

    def display_board(self):
        print("\n-----")
        for i in range(0, 9, 3):
            print(f"| {self.board[i]} | {self.board[i+1]} | {self.board[i+2]} |")
            print("-----")
        print()

    def is_moves_left(self, board) -> bool:
        if (' ' in board):
            return True
        else:
            return False

    def check_winner(self, board, player):
        winning_combos = [
            (0, 1, 2), (3, 4, 5), (6, 7, 8), #row
            (0, 3, 6), (1, 4, 7), (2, 5, 8), #column
            (0, 4, 8), (2, 4, 6) #diag
        ]

        for a, b, c in winning_combos:
            if board[a] == board[b] == board[c] == player:
                return True
        return False

    def is_terminal(self, board):
        return (self.check_winner(board, self.ai) or
                self.check_winner(board, self.human) or
                not self.is_moves_left(board))

    def utility(self, board) -> int:
        if self.check_winner(board, self.ai):
            return 1
```

```
        elif self.check_winner(board, self.human):
            return -1
        else:
            return 0

def generate_successor_states(self, board, player):
    successors = []
    for i in range(9):
        if board[i] == ' ':
            new_state = list(board)
            new_state[i] = player
            successors.append(new_state)
    return successors

def minimax(self, board, depth, is_maximizing):
    self.states_explored += 1

    score = self.utility(board)

    if self.is_terminal(board):
        return score

    if is_maximizing:
        best_val = -math.inf
        successors = self.generate_successor_states(board, self.ai)
        for state in successors:
            value = self.minimax(state, depth + 1, False)
            best_val = max(best_val, value)
        return best_val
    else:
        best_val = math.inf
        successors = self.generate_successor_states(board, self.human)
        for state in successors:
            value = self.minimax(state, depth + 1, True)
            best_val = min(best_val, value)
        return best_val

def alpha_beta(self, board, depth, alpha, beta, is_maximizing):
    self.states_explored += 1

    if self.is_terminal(board):
        return self.utility(board)

    if is_maximizing:
        best_val = -math.inf
        successors = self.generate_successor_states(board, self.ai)

        for state in successors:
            value = self.alpha_beta(state, depth + 1, alpha, beta, False)
            best_val = max(best_val, value)
            alpha = max(alpha, best_val)

            if beta <= alpha:
                break
        return best_val
    else:
```

```
best_val = math.inf
successors = self.generate_successor_states(board, self.human)

for state in successors:
    value = self.alpha_beta(state, depth + 1, alpha, beta, True)
    best_val = min(best_val, value)
    beta = min(beta, best_val)

    if beta <= alpha:
        break
return best_val

def find_best_move(self, algo):
    self.states_explored = 0

    best_val = -math.inf
    best_move = -1

    for i in range(9):
        if self.board[i] == ' ':
            self.board[i] = self.ai

            if (algo == 1):
                move_val = self.minimax(self.board, 0, False)
            else:
                move_val = self.alpha_beta(self.board, 0, -math.inf, math.inf, False)

            self.board[i] = ' '

            if move_val > best_val:
                best_move = i
                best_val = move_val

    print(f"States explored: {self.states_explored}")
    return best_move

def play_game(self):
    print("<- Tic-Tac-Toe ->")

    algo = int(input(f"Minimax (1) or AlphaBeta Pruning (2): "))
    print("\nYou are 'X' and the AI is 'O'.")
    print("Enter positions 0-8 to play.")
    self.display_board()

    while not self.is_terminal(self.board):
        if self.current_player == self.human:
            while True:
                try:
                    user_input = input(f"Enter move for {self.human} (0-8): ")
                    move = int(user_input)
                    if 0 <= move <= 8 and self.board[move] == ' ':
                        self.board[move] = self.human
                        self.current_player = self.ai
                        break
                except:
                    print("Invalid move.")
```

```
        except ValueError:
            print("Please enter a valid integer between 0 and 8.")
    else:
        move = self.find_best_move(algo)
        self.board[move] = self.ai
        self.current_player = self.human
        print(f"AI placed at pos {move}")

    self.display_board()

    if self.check_winner(self.board, self.ai):
        print("Game Over. AI Wins!")
    elif self.check_winner(self.board, self.human):
        print("Game Over. You Win!")
    else:
        print("Game Over. It's a Draw!")

if __name__ == "__main__":
    game = TicTacToeGame()
    game.play_game()
```

## Input/Output

```
• vicfic@NAG3:~/stud/ai/ex3$ python tic-tac-toe.py  
<- Tic-Tac-Toe ->  
Minimax (1) or AlphaBeta Pruning (2): 1
```

You are 'X' and the AI is 'O'.  
Enter positions 0-8 to play.

```
-----  
| | | |  
-----  
| | | |  
-----  
| | | |  
-----
```

Enter move for X (0-8): 4

```
-----  
| | | |  
-----  
| | x | |  
-----  
| | | |  
-----
```

States explored: 55504  
AI placed at pos 0

```
-----  
| o | | |  
-----  
| | x | |  
-----  
| | | |  
-----
```

Enter move for X (0-8): 1

```
-----  
| o | x | |  
-----  
| | x | |  
-----  
| | | |  
-----
```

States explored: 1054  
AI placed at pos 7

```
-----  
| o | x | |  
-----  
| | x | |  
-----  
| | o | |  
-----
```

```
• vicfic@NAG3:~/stud/ai/ex3$ python tic-tac-toe.py  
<- Tic-Tac-Toe ->  
Minimax (1) or AlphaBeta Pruning (2): 2
```

You are 'X' and the AI is 'O'.  
Enter positions 0-8 to play.

```
-----  
| | | |  
-----  
| | | |  
-----  
| | | |  
-----
```

Enter move for X (0-8): 4

```
-----  
| | | |  
-----  
| | x | |  
-----  
| | | |  
-----
```

States explored: 6138  
AI placed at pos 0

```
-----  
| o | | |  
-----  
| | x | |  
-----  
| | | |  
-----
```

Enter move for X (0-8): 1

```
-----  
| o | x | |  
-----  
| | x | |  
-----  
| | | |  
-----
```

States explored: 373  
AI placed at pos 7

```
-----  
| o | x | |  
-----  
| | x | |  
-----  
| | o | |  
-----
```

Enter move for X (0-8): 3

```
-----  
| 0 | X |   |  
-----  
| X | X |   |  
-----  
|   | 0 |   |  
-----
```

States explored: 50  
AI placed at pos 5

```
-----  
| 0 | X |   |  
-----  
| X | X | 0 |  
-----  
|   | 0 |   |  
-----
```

Enter move for X (0-8): 2

```
-----  
| 0 | X | X |  
-----  
| X | X | 0 |  
-----  
|   | 0 |   |  
-----
```

States explored: 4  
AI placed at pos 6

```
-----  
| 0 | X | X |  
-----  
| X | X | 0 |  
-----  
| 0 | 0 |   |  
-----
```

Enter move for X (0-8): 8

```
-----  
| 0 | X | X |  
-----  
| X | X | 0 |  
-----  
| 0 | 0 | X |  
-----
```

Game Over. It's a Draw!

vicfic@NAG3:~/stud/ai/ex3\$

Enter move for X (0-8): 3

```
-----  
| 0 | X |   |  
-----  
| X | X |   |  
-----  
|   | 0 |   |  
-----
```

States explored: 38  
AI placed at pos 5

```
-----  
| 0 | X |   |  
-----  
| X | X | 0 |  
-----  
|   | 0 |   |  
-----
```

Enter move for X (0-8): 2

```
-----  
| 0 | X | X |  
-----  
| X | X | 0 |  
-----  
|   | 0 |   |  
-----
```

States explored: 4  
AI placed at pos 6

```
-----  
| 0 | X | X |  
-----  
| X | X | 0 |  
-----  
| 0 | 0 |   |  
-----
```

Enter move for X (0-8): 8

```
-----  
| 0 | X | X |  
-----  
| X | X | 0 |  
-----  
| 0 | 0 | X |  
-----
```

Game Over. It's a Draw!

vicfic@NAG3:~/stud/ai/ex3\$

## **Learning Outcomes**

- Implemented the Minimax and Alpha-Beta Pruning Search Methods.
- Learned to write a algorithmic opponent in an adversarial environment
- Understood the efficiancy that the Alpha Beta Pruning approach has over general Minimax in the amount of states explored.