



Problem statement: Tic-Tac-Toe solver

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Introduction

Tic-Tac-Toe is a widely known two-player game played on a 3x3 grid, where players alternate turns placing their marks (X or O) in an attempt to form a winning row, column, or diagonal.

This project implements an AI-powered Tic-Tac-Toe Solver using the Minimax Algorithm with Alpha-Beta Pruning . The AI plays optimally, ensuring it either wins the game or forces a draw.

Methodology

1. Game Representation:

- The Tic-Tac-Toe board is modeled as a 3x3 matrix .
- Players make moves by placing 'X' or 'O' on the board.
- The game terminates when a player wins or when all cells are filled (tie).

2. Minimax Algorithm:

- The AI evaluates all possible moves recursively to determine the best possible decision.
- It assigns scores based on outcomes:
 - +1 for an AI (X) win
 - -1 for an opponent (O) win
 - 0 for a draw

3. Alpha-Beta Pruning:

- The Alpha-Beta Pruning technique is incorporated to enhance the efficiency of the Minimax Algorithm .
- This optimization helps in reducing unnecessary computations by eliminating branches that don't need to be explored.

4. Implementation Details:

- Developed in Python .
- Executes in a loop where the AI and human player take turns.
- Utilizes recursion to compute the best move using Minimax.

Code

python

```
def print_board(board):
```

```
    """Prints the current state of the Tic-Tac-Toe board."""
```

```
    for row in board:
```

```
        print(" | ".join(row))
```

```
    print("-" 9)
```

```
def check_winner(board):
```

```
    """Checks if there's a winner or a tie."""
```

```
    Check rows
```

```
    for row in board:
```

```
        if row[0] == row[1] == row[2] != ' ':
```

```
            return row[0]
```

```
    Check columns
```

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

```
        if board[0][col] == board[1][col] == board[2][col] != ' ':
```

```
            return board[0][col]
```

```
    Check diagonals
```

```
    if board[0][0] == board[1][1] == board[2][2] != ' ':
```

```
        return board[0][0]
```

```
    if board[0][2] == board[1][1] == board[2][0] != ' ':
```

```
return board[0][2]
```

```
Check for tie
```

```
for row in board:
```

```
    if ' ' in row:
```

```
        return None    No winner yet
```

```
return 'Tie'
```

```
def minimax(board, depth, is_maximizing):
```

```
    """Minimax algorithm with alpha-beta pruning."""
```

```
    winner = check_winner(board)
```

```
    if winner:
```

```
        if winner == 'X':
```

```
            return 1
```

```
        elif winner == 'O':
```

```
            return -1
```

```
    else:
```

```
        return 0
```

```
if is_maximizing:
```

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

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

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

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

```
                board[i][j] = 'X'
```

```
                score = minimax(board, depth + 1, False)
```

```

        board[i][j] = ' '   Backtrack
        best_score = max(score, best_score)
    return best_score
else:
    best_score = float('inf')
    for i in range(3):
        for j in range(3):
            if board[i][j] == ' ':
                board[i][j] = 'O'
                score = minimax(board, depth + 1, True)
                board[i][j] = ' '   Backtrack
                best_score = min(score, best_score)
    return best_score

```

```

def find_best_move(board):
    best_score = -float('inf')
    best_move = (-1, -1)

    for i in range(3):
        for j in range(3):
            if board[i][j] == ' ':
                board[i][j] = 'X'
                score = minimax(board, 0, False)
                board[i][j] = ' '   Backtrack
                if score > best_score:

```

```
        best_score = score
        best_move = (i, j)
    return best_move
```

Example usage:

```
board = [[' ', ' ', ' '], [' ', ' ', ' '], [' ', ' ', ' ']]
print_board(board)
```

while True:

```
    move = find_best_move(board)
    board[move[0]][move[1]] = 'X'
    print_board(board)
    winner = check_winner(board)
```

if winner:

```
    print(f"{winner} wins!")
    break
```

Placeholder for user input (replace with actual user input)

```
row, col = map(int, input("Enter row and column (0-2, space-separated):
").split())
```

```
board[row][col] = 'O'
```

```
print_board(board)
```

```
winner = check_winner(board)
```

if winner:


```
print(f"{winner} wins!")
```

```
break
```

```
---
```

Output/Result

1. AI makes an optimal move:

- The AI determines the best possible move using Minimax Algorithm .
- AI places 'X' in an advantageous position.

2. Game Progression:

- The AI and the human player take turns making moves.
- The board updates after each move.

3. Final Outcome:

- The game concludes when a player wins or the board is filled.
- The AI ensures a win or forces a tie.

References/Credits

- Minimax Algorithm : Fundamental AI decision-making technique.
- Alpha-Beta Pruning : Optimization method to enhance efficiency.
- Python Implementation: Developed and executed in [Google Colab](#) .
