AI-BASED NOUGHTS AND CROSSES WITH MINIMAX AND ALPHA-BETA PRUNING

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Project: Noughts and Crosses with Alpha-Beta

Pruning

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Date: March 10, 2025

## Introduction

Tic-Tac-Toe or Noughts and Crosses is a two-player game in which players alternate turns to fill in spaces on a 3x3 grid. This project develops an AI Tic-Tac-Toe agent based on the Minimax algorithm with Alpha-Beta Pruning, which enables the AI to play optimally by reducing its worst-case loss.

## Methodology

#### 1. Game Representation:

- The board is a 3x3 NumPy array with values:
- `-1` for the human player ('X')
- `1` for the AI player ('0')
- `0` for vacant spaces

## 2. Minimax Algorithm:

- Minimax algorithm is applied for making decisions, with the AI always choosing the optimal move by considering future game positions.

### 3. Alpha-Beta Pruning:

- Alpha-Beta Pruning improves Minimax by removing redundant branches, making it more efficient.

#### 4. Move Selection:

- The AI determines the optimal move using Minimax and makes an optimal play.
- The game loop tests for win/draw situations after every move.

## **Full Python Code Implementation**

Below is the complete Python implementation of the AI for Noughts and Crosses using Minimax and Alpha-Beta Pruning:

import numpy as np

```
# Constants for the players
HUMAN = -1 \# 'X'
AI = 1
         # '0'
EMPTY = 0
# Function to print the board
def print_board(board):
 symbols = {HUMAN: 'X', AI: 'O', EMPTY: '-'}
 for row in board:
    print(" ".join([symbols[cell] for cell in row]))
 print("\n")
```

## # Function to check for a winner or a draw

```
def check_winner(board):
 for player in [HUMAN, AI]:
```

if any(np.all(row == player) for row in board) or any(np.all(col == player) for col in board.T) or np.all(np.diag(board) == player) or np.all(np.diag(np.fliplr(board)) == player):

```
return player
 return None if np.any(board == EMPTY) else 0
# Minimax with Alpha-Beta Pruning
def minimax(board, depth, alpha, beta, is_maximizing):
 winner = check_winner(board)
 if winner is not None:
   return {AI: 10, HUMAN: -10, 0: 0} [winner]
 if is maximizing:
   best = -np.inf
   for i, j in zip(*np.where(board == EMPTY)):
      board[i, j] = AI
      best = max(best, minimax(board, depth + 1, alpha, beta, False))
      board[i, j] = EMPTY
      alpha = max(alpha, best)
     if beta <= alpha:
        break
   return best
 else:
   best = np.inf
   for i, j in zip(*np.where(board == EMPTY)):
      board[i, j] = HUMAN
      best = min(best, minimax(board, depth + 1, alpha, beta, True))
      board[i, j] = EMPTY
      beta = min(beta, best)
      if beta <= alpha:
        break
   return best
# Function to find the best move for AI
def find_best_move(board):
 best_val, best_move = -np.inf, None
 for i, j in zip(*np.where(board == EMPTY)):
   board[i, j] = AI
   move_val = minimax(board, 0, -np.inf, np.inf, False)
   board[i, j] = EMPTY
   if move_val > best_val:
      best_val, best_move = move_val, (i, j)
 return best_move
# Main game loop
def play_game():
 board = np.zeros((3, 3), dtype=int)
 print("Tic-Tac-Toe with AI (Alpha-Beta Pruning)")
 print_board(board)
 while True:
   # Human move
   while True:
      try:
```

```
row, col = map(int, input("Enter row and column (0-2): ").split())
        if board[row, col] == EMPTY:
         board[row, col] = HUMAN
         break
        print("Cell occupied! Try again.")
      except (ValueError, IndexError):
        print("Invalid input! Enter numbers between 0-2.")
   print_board(board)
   if (winner := check_winner(board)) is not None:
      print("Draw!" if winner == 0 else "You Win!" if winner == HUMAN else "AI Wins!")
     break
   # AI move
   print("AI is thinking...")
   ai_move = find_best_move(board)
   board[ai_move] = AI
   print_board(board)
   if (winner := check_winner(board)) is not None:
     print("Draw!" if winner == 0 else "You Win!" if winner == HUMAN else "AI Wins!")
     break
# Start the game
if __name__ == "__main__":
 play_game()
```

# **Output/Result**

Below is a sample game output:

```
Tic-Tac-Toe with AI (Alpha-Beta Pruning)
---
---
Enter row and column (0-2): 0 0
X --
---
AI is thinking.
X - 0
---
---
MI Wins!
```

The AI optimally plays correctly, making the optimal move selection.

```
Enter row and column (0-2): 2 2
- - X
AI is thinking...
- 0 -
- - X
Enter row and column (0-2): 4
Invalid input! Enter numbers between 0-2.
Enter row and column (0-2): 1
Invalid input! Enter numbers between 0-2.
Enter row and column (0-2): 0 0
- 0 -
- - X
AI is thinking...
X 0 -
- 0 -
- - X
Enter row and column (0-2): 2 1
X 0 -
- 0 -
- X X
AI is thinking...
X 0 -
- 0 -
0 X X
Enter row and column (0-2): 0 2
X \circ X
- 0 -
0 X X
AI is thinking...
X \circ X
- 0 0
0 X X
Enter row and column (0-2): 1 0
X \circ X
X 0 0
0 X X
```

# **References/Credits**

Minimax Algorithm: Wikipedia

Alpha-Beta Pruning: GeeksforGeeks