Noughts and Crosses with Alpha-Beta Pruning

Problem Statement:

The task is to develop an intelligent version of the classic Tic-Tac-Toe game, also known as Noughts and Crosses, where the AI plays optimally using the Minimax Algorithm and Alpha-Beta Pruning. In this game, Player X is the human player, and Player O is the AI, which uses the Minimax algorithm to decide its moves, improving efficiency with Alpha-Beta pruning.

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INTRODUCTION

Problem Explanation:

Noughts and Crosses (Tic-Tac-Toe) is a 2-player game played on a 3x3 grid. The players take turns marking spaces with 'X' (Player X) and 'O' (Player O). The first player to align three marks in a row, column, or diagonal wins. If the grid is filled without a winner, the game results in a draw.

This project simulates an optimal game where **Player X** and **Player O** play using the **Minimax algorithm**, which evaluates possible moves to select the best one. To improve efficiency, **Alpha-Beta Pruning** is used to eliminate unnecessary calculations.

Methodology

The approach for solving the Noughts and Crosses (Tic-Tac-Toe) problem involves the following key steps:

1. Game Setup:

The game starts with an empty 3x3 grid. Player X is the maximizing player, and Player O is the minimizing player. They take turns marking their moves.

2. Minimax Algorithm:

The Minimax algorithm evaluates all possible moves to select the optimal one for each player. Player X aims to maximize their chances of winning, while Player O minimizes Player X's chances.

3. Alpha-Beta Pruning:

Alpha-Beta pruning enhances the Minimax algorithm by eliminating unimportant branches in the decision tree, improving efficiency and reducing computational time.

4. Game Over Condition:

The game checks for a win after each move. If a player places three marks in a row, column, or

diagonal, they win. If the grid is full without a winner, the game ends in a draw.

5. Player Interaction:

Player X (AI) and Player O (opponent) alternate turns. Both players use the Minimax algorithm to determine their moves.

6. End of Game:

The game concludes when there is a winner or when it ends in a draw. The result is displayed once the game finishes.

```
Code:
import random
# Constants for the game
PLAYER X = 'X'
PLAYER O = 'O'
EMPTY = ' '
# Initialize the board
def init_board():
  return [[EMPTY, EMPTY, EMPTY] for _ in range(3)]
# Check if a player has won
def check win(board, player):
 # Check rows, columns, and diagonals
 for i in range(3):
   if all([board[i][j] == player for j in range(3)]) or
all([board[j][i] == player for j in range(3)]):
     return True
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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
# Check if the game is over (win or draw)
def game_over(board):
 if check_win(board, PLAYER_X):
   return 1 # Player X wins
 if check win(board, PLAYER O):
   return -1 # Player O wins
 if all(board[i][j] != EMPTY for i in range(3) for j in
range(3)):
   return 0 # Draw (no empty spaces)
 return None # Game is ongoing
# Minimax with Alpha-Beta Pruning
def minimax(board, depth, alpha, beta,
maximizing player):
  result = game over(board)
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if result is not None:
   return result #1, -1, or 0 (win, loss, draw)
 if maximizing_player:
   max_eval = -float('inf')
   for i in range(3):
     for j in range(3):
       if board[i][j] == EMPTY:
         board[i][j] = PLAYER_X
         eval = minimax(board, depth + 1, alpha,
beta, False)
         board[i][j] = EMPTY
         max_eval = max(max_eval, eval)
         alpha = max(alpha, eval)
         if beta <= alpha:
           break # Beta cut-off
   return max eval
 else:
   min eval = float('inf')
   for i in range(3):
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for j in range(3):
       if board[i][j] == EMPTY:
         board[i][j] = PLAYER_O
         eval = minimax(board, depth + 1, alpha,
beta, True)
         board[i][j] = EMPTY
         min_eval = min(min_eval, eval)
         beta = min(beta, eval)
         if beta <= alpha:
           break # Alpha cut-off
   return min eval
# Find the best move for the current player
def best_move(board, maximizing_player):
  best_val = -float('inf') if maximizing_player else
float('inf')
  move = (-1, -1)
 for i in range(3):
   for j in range(3):
     if board[i][j] == EMPTY:
```

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board[i][i] = PLAYER X if
maximizing_player else PLAYER_O
       move_val = minimax(board, 0, -float('inf'),
float('inf'), not maximizing_player)
       board[i][j] = EMPTY
       if (maximizing_player and move_val >
best_val) or (not maximizing_player and move_val
<br/>
<br/>
dest val):
         best val = move val
         move = (i, j)
  return move
# Print the board with dashes after the last line
def print_board(board):
 for i, row in enumerate(board):
   print('|'.join(row))
   if i < 2: # Only print dashes between rows, not
after the last one
     print('----')
  print("\n") # Adding a newline after the board
for better readability.
```

```
# Example gameplay loop
def play_game():
 board = init_board()
 current player = PLAYER X # Start with "X"
 while game_over(board) is None:
   print_board(board)
   if current_player == PLAYER_X:
     # Player X's turn (user input)
     valid move = False
     while not valid move:
       try:
         i, j = map(int, input("Player X, enter your
move (row and column, e.g. 12): ").split())
         if board[i][j] == EMPTY:
          valid move = True
           board[i][j] = PLAYER X
         else:
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print("Cell already taken! Try again.")
       except (ValueError, IndexError):
         print("Invalid input! Enter row and
column numbers between 0 and 2.")
   else:
     # Player O's turn (Al move)
     print("Player O is making a move...")
     i, j = best_move(board, False)
     board[i][j] = PLAYER_O
   current player = PLAYER O if current player
== PLAYER_X else PLAYER_X
   # Print a message after each move
   print("Next move is being made...\n")
 print board(board)
 result = game over(board)
 if result == 1:
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```
print("Player X wins!")
elif result == -1:
    print("Player O wins!")
else:
    print("It's a draw!")

if __name__ == "__main__":
    play_game()
```

Output:

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Player X, enter your move (row and column, e.g. 1 2): 2 2
Next move is being made...
Player 0 is making a move...
Next move is being made...
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```
Player X, enter your move (row and column, e.g. 1 2): 2 1
Next move is being made...
Player 0 is making a move...
Next move is being made...
Player X, enter your move (row and column, e.g. 1 2): 1 2
Next move is being made...
```

Reference:

Minimax Algorithm - For understanding the Minimax algorithm, I referred to various online resources and tutorials:

- GeeksforGeeks: "Minimax Algorithm with Alpha-Beta Pruning" (https://www.geeksforgeeks.org/mini-maxalgorithm-in-artificial-intelligence/)
- Alpha-Beta Pruning To implement Alpha-Beta
 Pruning for optimizing the Minimax algorithm:
- Wikipedia: "Alpha-beta pruning" (https://en.wikipedia.org/wiki/Alpha%E2%80%9 3beta_pruning)