

Department of Computer Science
Nirma University

Artificial Intelligence Innovative Assignment

Connect Four Game using Al

Ву

17BCE010 Bharat Adhyaru 17BCE014 Khodidas Chauhan 17BCE015 Chinmaya Kapopara

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About Game

Connect Four is a two-player connection board game, designed by Howard Wexler and NEd Strongin, which is also known as Plot Four, Four Up, Four in a Line, Four in a Row, Drop Four and Gravitrips.

1. Gameplay

Connect Four has two players playing against each other on a 6x7 (6 row, 7 columns) vertically placed board. Each player is represented by tokens of different color pieces.

When a player takes a turn, he/she puts the token in a given column and it falls down to the last empty cell.

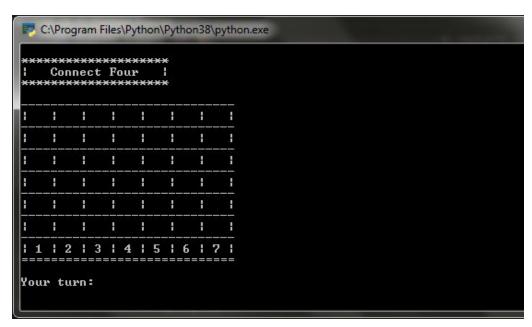
2. Rules & Objective

The Rules are very simple that one cannot place a token in a non-empty column. The Objective of the game is to connect four tokens of the same color either in a row, column or any of the diagonals.

Connect Four is a solved game i.e., its complete state space search tree can be made however, the branching factor is 7 (b=7) and the maximum depth it can go is 42 (m=42). Hence, the total number of nodes in the tree would be **7^42 = 3.1197348e+35**



The Classic Connect Four Board

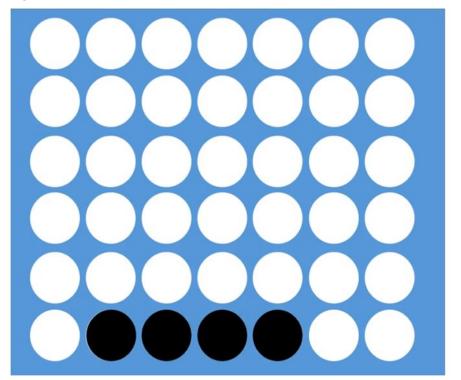


The Connect Four created by us

3. Winning Conditions

As mentioned above, one can win by connecting four tokens in:

a. Row

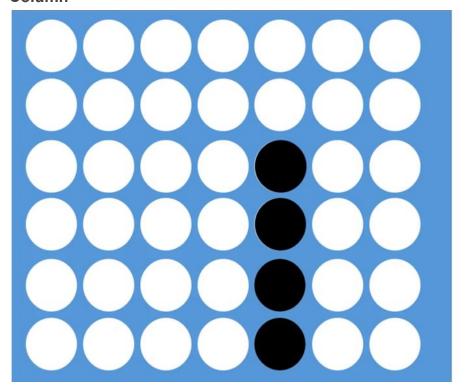


You can get a row horizontally



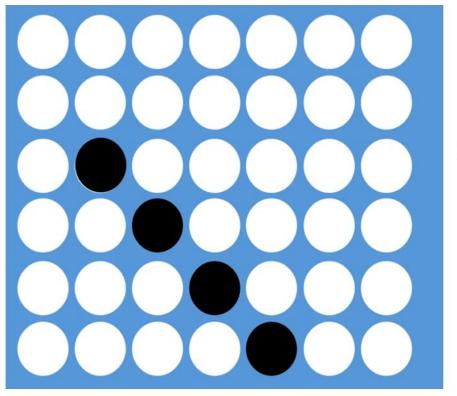
```
C:\Program Files\Python\Python38\python.exe
```

b. Column



You can get a row vertically

c. Diagonals



You can get a row on an angle



```
C:\Program Files\Python\Python38\python.exe
                         1
    1
        1
            1
                     1
                    101
                101
        1
            101
                         1
                             1
        101111
111213141516171
Process returned 0 (0x0) ex
Press any key to continue . . . _
                                execution time : 0.191 s
```

Algorithm

Overview

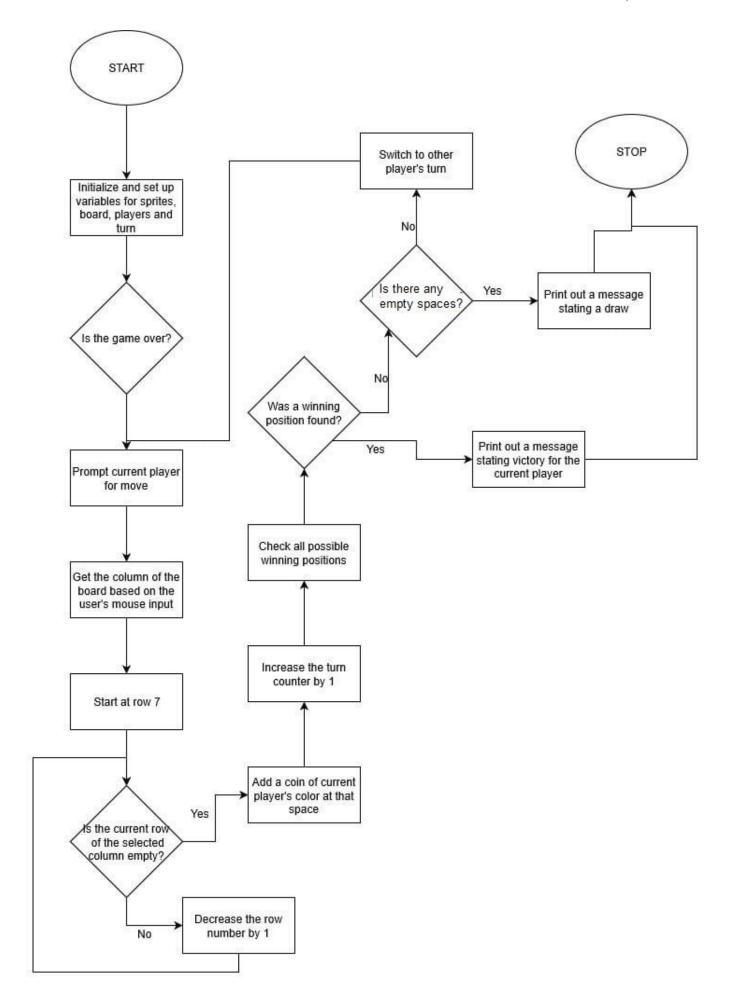
The algorithm uses Minimax for game playing, so as to decide which step to take next. We have also used Alpha-Beta Pruning so as to prune the unnecessary branches in the

state space tree, because as mentioned earlier the total number of nodes reaches as much as 7^42, which is way too large for a normal computer to compute. Hence, by using Alpha-Beta Pruning and limiting the search to a certain depth (here, depth=5) we optimized our algorithm so it can work on a normal computer.

Note: Even after limiting the state space tree search to depth 5, the AI may have to check as much as 117,649 moves (7⁶).

Flowchart

The Flowchart of our algorithm is mentioned below:



Explanation of important functions

1. printBoard() and printInstructions()

As the name suggests printInstructions() prints the instructions at the beginning while printBoard() prints the board after every move of the player as well as AI.

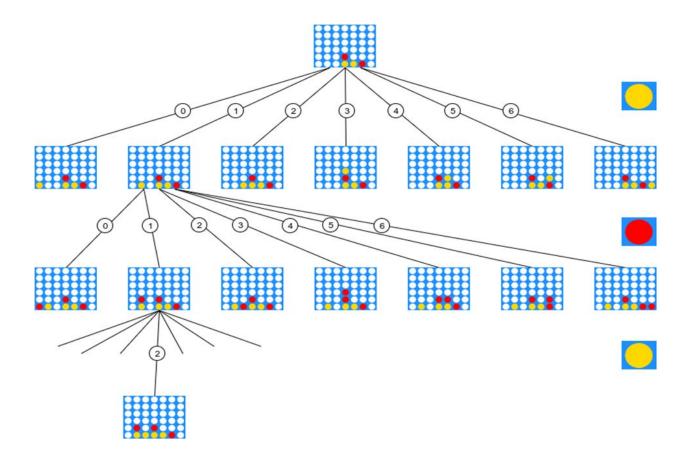
```
0
C:\Program Files\Python\Python38\python.exe
     Rules
Rule 1: Win by connecting four tokens in a row, column or any of the diagonals
Rule 2: Don't let AI win :)
                                                                                              E
    Instructions
You go first
***********
    Connect Four
         .
              1
                  Ŧ
                       1
                            Ŧ
                                1
    ı
         ŧ.
              1
                  Ŧ
                       1
                            1
                                Ŧ
 1 | 2 | 3 | 4 | 5 | 6 | 7 |
Your turn: _
```

The working of printInstructions() and printBoard() functions

2. findBestMove() and Minimax()

In findBestMove() the AI places its token in every possible column cell and then calls the Minimax function for further processing. Hence, findBestMove() is the starting point of initiating Minimax(). Minimax() then creates the state space tree till a game ends condition (win, lose or draw) or upto to a certain depth.

The working of findBestMove() and Minimax() is shown below:



3. Heuristics

The heuristics are precisely designed in such a way that AI focuses a little more on winning rather than focusing on defending and winning equally. We achieved the heuristics value (mentioned in code) by rigorous testing and some online help.

a. winning_move()

This function checks that in the current state of board is there any possibility for any player (Al or player) to win. If yes, it returns true and the game ends right then, else it returns false and the game continues.

b. anyMovesLeft()

This function simply checks if there is/are move(s) left i.e., it checks if there is any empty cell. If yes, it returns **True** and the game continues, else it returns **False** and that indicates the game is drawn.

c. evaluateBoard()

This function simply evaluates the board, after every move made by the AI, in Minimax() and assigns a heuristic value based on the state of the board.

It assigns value in following ways:

- 1. +10,000 if AI is winning
- 2. -10,000 if AI is loosing
- 3. 100 if found |X|X|X|X|
- 4. 5 if found as |X|X|X|_| or |X|X|_|X| and so on
- 5. 2 if found as |X|X|_|_| or |X|_|X|_| and so on

The heuristics for point 3 to 5 are multiplied by -1 if it's the player's turn else it is passed as it is in case of Al turns (because Al is the maximizer player here).

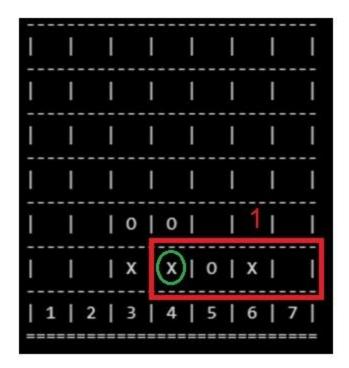
d. checkRow(), checkCol(), checkDiag1(), checkDiag2()

The evaluateBoard() takes help of above mentioned functions to assign heuristic value to the board state.

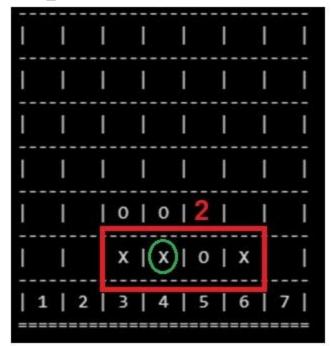
The rowCheck function evaluates all permutations of a token by placing it in four different positions of four different rows, it can be a part of. And then sums up the score of the heuristic values. This is done in order to take the AI towards a state (of board) where it has the highest chances of winning or lowest chances of losing.

The working of rowCheck() is mentioned below: (The token whose permutations are checked is mentioned in green circle)

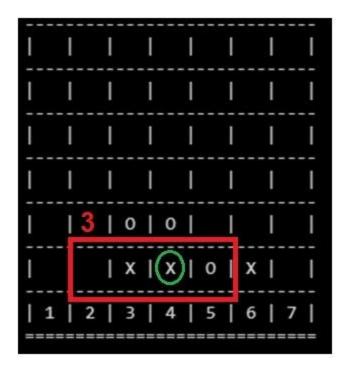
Let say the score of below mentioned state is **score_1 = 0** (as per heuristics)



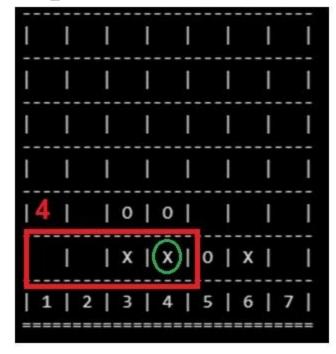
score_2 = 0



 $score_3 = 0$



score_4 = 2



Hence, the final score turns out to be final_score = score_1 + score_2 + score_3 + score_4 final_score = 2

The final_score's value is returned to evaluateBoard() by rowCheck().

The same concept of checking all permutations of a token in a column and both diagonals, and assigning a score, is done by checkCol() and checkDiag1() & checkDiag2(), respectively.

Code

```
# ***************
# Crafted from scratch by
# 17BCE010 Bharat Adhyaru
# 17BCE014 Khodidas Chauhan #
# 17BCE015 Chinmaya Kapopara #
# **********
#global variables
player = 'X'
opponent = '0'
moves count = 0
best row = -1
best col = -1
winner = None
def printInstructions():
   print(" -----")
   print("| Rules |\n -----")
   print("Rule 1: Win by connecting four tokens in a row, column or
any of the diagonals")
   print("Rule 2: Don't let AI win :)")
   print("\n -----")
   print(" | Instructions |\n -----")
   print("You = \'X\'\nAI = '0'")
   print("You go first\n")
   print("\n**************")
   print("| Connect Four |\n************\n")
```

```
def printBoard(board):
    for i in range(6):
       print("-----")
       print("|", end="")
       for j in range(7):
           print("",board[i][j], end = "")
           print(" |", end="")
       print()
   print("----")
   print(" | 1 | 2 | 3 | 4 | 5 | 6 | 7 |")
   print("=======\n")
# Improving heuristics because the program cannot run
# on its max recursion limit(10**9)
# Because, legal moves present at every depth is 7 (b = 7)
\# and max depth that can be achieved is 42 (m = 42)
# Hence, the time complexity is O(b^m) = O(7^42) = 3.1197348e+35 !!!
# Hence, the search needs to be limited to a certain depth
# but still we need to find the best move
# So the heuristics have been improved in the following way
# for row:
# return 0 if found as |X| |0|X| or |X|0| | | and so on...
# return 1 if found as |X|_{-}|_{-}| (Note: the remaining three entries
must be empty)
# return 2 if found as |X|X| | |
# return 5 if found as |X|X|X|_|
# same applies to columns and diagonals
def evaluateTokenCount(array, main token):
   token count = 0
   empty count = 0
   #count both
   token count = array.count(main token)
    empty count = array.count(' ')
    if token count == 4:
                                                 # return 100 if
found |X|X|X|X|
       return 100
    elif token count == 3 and empty count == 1: # return 5 if
```

```
found as |X|X|X|_{-}| or |X|X|_{-}|X| and so on
        return 5
    elif token count == 2 and empty count == 2: # return 2 if
found as |X|X| \mid | or |X| \mid |X| \mid and so on
        return 2
    #elif token count == 1 and empty count == 3: # return 1 if
found as |X|X| \mid | or |X| \mid |X| \mid and so on
        #return 1
    else:
        return 0
                                                     # cases like
|X|_{0} = |X|_{0} = |X|_{0} = |X|_{0} and so on
def checkRow(board, r, c):
    # make subarrays to check for four consecutive places for all the
four subarrays
    # say row = |X| | |X| | 0| and our c = 4 (board[r][c] = X),
referring this token as MX
    # so four different possibilities that need to be checked are:
    # 1. |X| | |MX|
    # 2. |_|_|MX|_|
    # 3. | |MX| |0|
    # 4. |MX| |0| |
    # because every possibility adds (or reduces) the score and all
of them contributes to score
    # so the final score, for this function would be sum of all
possibilities
    score = 0
    for j in range(4):
        start = c-j
        end = start+3
        if start >= 0 and end < cols:</pre>
            row_array = [board[r][i] for i in range(start, end+1)]
            score += evaluateTokenCount(row array, board[r][c])
#passing main token
    return score
#-----
def checkCol(board, r, c):
```

```
# make subarrays to check for four consecutive places for all the
four subarrays
   score = 0
   for j in range(4):
        start = r+j
       end = start-3
        if start < rows and end >= 0:
            col_array = [board[i][c] for i in range(start, end-1,
-1)]
           score += evaluateTokenCount(col array, board[r][c])
    return score
#-----
def checkDiag1(board, r, c):
   # make subarrays to check for four consecutive places for all the
four subarrays for one diagonal - top left to bottom right
   score = 0
   for j in range(4):
        start row = r-j
        start col = c-i
       end row = start row + 3
       end col = start col + 3
        if start row >= 0 and end_row < rows and start_col >= 0 and
end_col < cols:</pre>
           diag1 array = [board[start row+i][start col+i] for i in
range(4)]
           score += evaluateTokenCount(diag1_array, board[r][c])
    return score
#-----
def checkDiag2(board, r, c):
   # make subarrays to check for four consecutive places for all the
four subarrays for other diagonal - bottom left to top right
   score = 0
    for j in range(4):
        start row = r-j
        start col = c+j
       end row = start row + 3
       end col = start col - 3
```

```
if start row >= 0 and end row < rows and start col < cols and</pre>
end col >= 0:
           diag2 array = [board[start row+i][start col-i] for i in
range(4)]
           score += evaluateTokenCount(diag2 array, board[r][c])
    return score
#-----
def evaluateBoard(board):
   turn = -1
                    #best score for Maximizer
   max score = -1
   min score = 1
                      #best score for Minimizer
   best score = 0 #best (absolute value) of above two
   #checking for each cell of board
   for j in range(cols):
       for i in range(rows):
           score = 0
           if board[i][j] == ' ' and i < rows-1: #finds first</pre>
non-empty cell
               continue
           else:
               if board[i][j] == ' ': #if cell of a column is
empty no need to evaluate
                   continue
               #else continue below
               score += checkRow(board, i, j) + checkCol(board, i,
j) + checkDiag1(board, i, j) + checkDiag2(board, i, j) #sum of all
four
               if board[i][j] == 'X':
                   min score = max(min score, score)
               else:
                   max score = max(max score, score)
    return int(max score-min score/1.2)
                                               # the div is kept so
that AI focus little more on winning rather than
                                               # keeping defending
and winning exactly(or almost) equal
                                               # in case of
```

```
(max score - min score), the winning and defending is almost equal
                                                # experiment with the
division value and you will get what we mean
def anyMovesLeft(board):
   global rows, cols
   for i in range(rows):
        for j in range(cols):
            if board[i][j] == ' ':
                return True
   return False
#-----
def winning move(board, token):
     # Check horizontal locations for win
     for c in range(cols-3):
          for r in range(rows):
                if board[r][c] == token and board[r][c+1] == token
and board[r][c+2] == token and board[r][c+3] == token:
                     return True
     # Check vertical locations for win
     for c in range(cols):
          for r in range(rows-3):
                if board[r][c] == token and board[r+1][c] == token
and board[r+2][c] == token and board[r+3][c] == token:
                     return True
     # Check positively sloped diagonals for win
     for c in range(cols-3):
          for r in range(rows-3):
                if board[r][c] == token and board[r+1][c+1] == token
and board[r+2][c+2] == token and board[r+3][c+3] == token:
                     return True
     # Check negatively sloped diagonals for win
     for c in range(cols-3):
          for r in range(3, rows):
```

```
if board[r][c] == token and board[r-1][c+1] == token
and board[r-2][c+2] == token and board[r-3][c+3] == token:
                      return True
def minimax(board, depth, is maximising player, alpha, beta):
    global player, opponent, moves count
    moves count += 1
    score = 0
    if winning_move(board, opponent):
        return 10000 - depth
                                         #for faster victory
    elif winning move(board, player):
                                       #for slower draw/defeat
        return -10000 + depth
    if anyMovesLeft(board) == False:
        return 0
    #limiting to a certain depth
    if depth == 5:
                               #It will check at most 7<sup>6</sup> moves
(depth 0 is also included) -> 7^6 = 117,649
          return evaluateBoard(board)
    #if AI's turn
    if is_maximising_player == True:
        best val = -1000000
        val = -1
        for j in range(cols):
            for i in range(rows):
                if board[i][j] == ' ' and i < rows-1: #finds first</pre>
non-empty cell
                    continue
                elif i > 0:
                    if board[i][j] != ' ':
                        i -= 1
                    board[i][j] = opponent
                    val = minimax(board, depth + 1, False, alpha,
beta)
                    best val = max(best val, val)
```

```
alpha = max(alpha, best val)
                    #undo
                    board[i][j] = ' '
                    #Pruning step
                    if beta <= alpha:</pre>
                        return best val
                    #go for next column
                    break
                if i == 0: #if first row is non-empty go to next
column
                    break
        return best_val
    #if player's turn
    if is_maximising_player == False:
        best val = 1000000
        val = -1
        for j in range(cols):
            for i in range(rows):
                if board[i][j] == ' ' and i < rows-1: #finds first</pre>
non-empty cell
                    continue
                elif i > 0:
                    if board[i][j] != ' ':
                        i -= 1
                    board[i][j] = player
                    val = minimax(board, depth + 1, True, alpha,
beta)
                    best val = min(best val, val)
                    beta = min(beta, best val)
                    #undo
                    board[i][j] = ' '
                    #Pruning Step
                    if beta <= alpha:</pre>
                        return best val
                    #go for next column
                    break
                if i == 0: #if first row is non-empty go to next
column
```

```
break
       return best_val
#-----
def findBestMove(board):
    global best row, best col, opponent, cols, rows
    best val = -10000000
    # alpha & beta
    alpha = -10000000
    beta = 10000000
    #try to find the best move by placing token in every column
    for j in range(cols):
       val = 0
       #find last empty cell in that column
       for i in range(rows):
            if board[i][j] == ' ' and i < rows-1: #finds first</pre>
non-empty cell
               continue
           elif i > 0:
                                                   #could enter if
board[i][j] != ' ' or i == 5: and do nothing if first cell is
occupied
               if board[i][j] != ' ':
                                                  #first non-empty
cell, could be i == 5 as well
                   i -= 1
                                                  #go to previous
empty cell
               # if board[i][j] == ' ':
                                                 #last cell of col
empty i == 5
                   #do nothing
               board[i][j] = opponent
               #uncomment below three statements to visualize
working of Minimax
               #----- Visualizer ------
               #print("=====Before=====")
               #printBoard(board)
               #print("i,j =",i,j," | val =",val)
               #call minimax
               val = minimax(board, 0, False, alpha, beta)
```

```
# uncomment below line to visualize just heuristics
                print("col,row =",j+1,i+1," | val =",val)
                #undo move
                board[i][j] = ' '
                if val > best val:
                    best row = i
                    best col = j
                    best val = val
                break
            if i == 0: #if first row is non-empty go to next
column
                break
    board[best_row][best_col] = opponent
#-----
#main()
rows, cols = (6, 7)
#initializations
board = [[' ' for j in range(cols)] for i in range(rows)]
printInstructions()
printBoard(board)
player col = 0
winner = None
# Game loop
while anyMovesLeft(board) == True:
    player col = input("Your turn: ")
    #invalid input if not a number
    if not player col.isnumeric():
        print("Invalid input!! Enter choice from 1 to 7. Try
again.\n")
        continue
    player col = int(player col)
    #invalid index
    if player col > 7 or player col < 1:</pre>
        print("Invalid column index!! Enter choice from 1 to 7. Try
```

```
again.\n")
      continue
   player_col -= 1  #indexing starts from 1 for the consumer
   #invalid move
   if board[0][player col] != ' ': #topmost row of that
column not empty
      print("Invalid move!! Column not empty. Try again.\n")
      continue
   #insert into that cols
   for i in range(rows):
       if board[i][player_col] == ' ' and i < rows-1: #finds first</pre>
non-empty cell
          continue
      elif i > 0:
                                               #could enter
if board[i][j] != ' ' or i == 5: and do nothing if first cell is
occupied
          if board[i][player col] != ' ':
                                               #first
non-empty cell, could be i == 5 as well
             i -= 1
                                               #go to
previous empty cell
          board[i][player_col] = player
   print("\n -----")
   printBoard(board)
   #check for winner
   if winning move(board,player):
      winner = player
      break
   #AI's turn
   print(" -----")
   findBestMove(board)
   printBoard(board)
   print("Moves checked by AI
=",moves_count,"\n----\n")
```

```
moves count = 0
   #check for winner again
    if winning move(board,opponent):
       winner = opponent
       break
if winner == player or winner == None:
    printBoard(board)
print("\n************")
print("| Game Over |\n**********\n")
if winner == opponent:
    print("You Lost!!")
   print("Better luck next time :)")
elif winner == player:
   print("You Won!!")
   print("AI: Hats off to You. You defeated me") #This is never
gonna happen :)
else:
   print("Game Tied!!\n")
   print("AI: Told you, I'm unbeatable ;)")
# The most important thing we noted is that, heuristics is everything
for this kind of board games
# If we change the heuristic values in evaluateTokenCount(), the
moves of AI will change
# Same applies to return statement of evaluateBoard()
# because alpha-beta pruning and minimax at last depends on the score
they recieve
# ***************
# Crafted from scratch by
# 17BCE010 Bharat Adhyaru
# 17BCE014 Khodidas Chauhan #
# 17BCE015 Chinmaya Kapopara #
# **********
```

Conclusion

In this innovative assignment, we applied our knowledge of the Minimax algorithm and Alpha-Beta Pruning algorithm to develop the Connect Four game from scratch. While developing the game, we did some rigorous testing of how AI is behaving in different situations which in turn taught us what is the importance of heuristics and how differently AI reacted by a slight change of heuristics value. Hence, we learnt that the main hero for AI based board games is heuristics.

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Made by
17BCE010 Bharat Adhyaru
17BCE014 Khodidas Chauhan
17BCE015 Chinmaya Kapopara