# Assignment - 4: Tic Tac Toe AI using LMS rule

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SECTION: A

NOTE:

The following program implements Tic Tac Toe for a 4x4 board due to insufficient infrastructure, that is, takes too long to play for a 5x5 board.

```
import numpy as np
import pandas as pd
import copy
import math
from tqdm import tqdm
def initialize board(size = 4):
   Initialize board: (4x4)
    | - || - || - || - |
    _____
   | - || - || - || - |
    ______
    | - || - || - || - |
   | - || - || - || - |
   board = np.ones((size, size), dtype = str)
   board[:,:] = '-'
   return board
def getWinner(board):
       Function to return the winner based on the given board
       +100 : for X
        -100 : for 0
   size = board.shape[0]
   for i in range(size):
       if np.all(board[i,:] == 'X'): # 4 Xs in row i -> X
           return 100
```

```
if np.all(board[i,:] == '0'): # 4 Os in row i \rightarrow 0
            return -100
        if np.all(board[:,i] == 'X'): # 4 Xs in col i -> X
            return 100
        if np.all(board[:,i] == '0'): # 4 Os in col i -> 0
            return -100
   diagonally = [1, 1, 1, 1]
   for i in range(size):
        if board[i,i] != 'X':
            diagonally[0] = 0
        if board[i,i] != '0':
            diagonally[1] = 0
        if board[i,(-i-1)] != 'X':
            diagonally[2] = 0
        if board[i,(-i-1)] !=
                              '0':
            diagonally[3] = 0
   for i in range(4):
        if diagonally[i]:
            if i == 0 or i == 1:
                return 100
            else:
                return -100
   ### check if it was a draw
   if np.sum(board == '-') == 0:
        return 0
   # Otherwise not over
    return -1
def ExctractFeatures(board):
        Function to extract the (size*8 + 1) features from the board
        Explanation of input features given in assignment report
   # get size and create feature vector using it
   size = board.shape[0]
   feature_vector = np.zeros(size*8 + 1, dtype=np.float64)
   # bias
   feature vector[0] = 1.0
   # calculate the number of Xs, Os, and -s in rows and columns
   num_x_row = np.count_nonzero(board == 'X', axis = 1)
   num_x_col = np.count_nonzero(board == 'X', axis = 0)
   num_o_row
                = np.count_nonzero(board == '0', axis = 1)
```

```
num o col = np.count nonzero(board == '0', axis = 0)
num emp row = np.count nonzero(board == '-', axis = 1)
num emp col = np.count nonzero(board == '-', axis = 0)
# diagonal counters
major X = 0
              # Xs in leading
major 0 = 0  # Os in leading
major_emp = 0 # -s in leading
minor_X = 0  # Xs in trailing
minor_O = 0  # Os in trailing
minor emp = 0 # -s in trailing
for i in range(size):
    if board[i,i] == 'X':
        major_X += 1
    if board[i,i] == '0':
        major 0 += 1
    if board[i,i] == '-':
        major emp += 1
    if board[i,-i-1] == 'X':
        minor X += 1
    if board[i,-i-1] == '0':
        minor 0 += 1
    if board[i,-i-1] == '-':
        minor_emp += 1
# add values to feature vector
for i in range(1,size+1): # for each n of {1 to size}
    # for each row/col -> in first 4 of 8 positions
    for j in range(size):
        if num x row[j] == i and num emp row[j] == size-i:
            feature_vector[(i-1)*8 + 1] += 1
        if num_x_col[j] == i and num_emp_col[j] == size-i:
            feature vector[(i-1)*8 + 2] += 1
        if num_o_row[j] == i and num_emp_row[j] == size-i:
            feature vector[(i-1)*8 + 3] += 1
        if num_o_col[j] == i and num_emp_col[j] == size-i:
            feature_vector[(i-1)*8 + 4] += 1
    # diagonals -> in next 4 positions
    if major X == i and major emp == size-i :
        feature_vector[(i-1)*8 + 5]+=1
    if minor_X == i and minor_emp == size-i :
        feature vector[(i-1)*8 + 6]+=1
    if major 0 == i and major emp == size-i :
        feature_vector[(i-1)*8 + 7]+=1
    if minor_0 == i and minor_emp == size-i :
        feature\_vector[(i-1)*8 + 8]+=1
```

### ▼ Determine next possible moves

```
def getPossibleStates(board, player):
    ""
    Function to determine the next possible moves from the current board state.
    For every empty slot '-':
        Replace with either 'X' or '0' and append it to a list
        return list
    ""
    board_states = []
    size = board.shape[0]

for i in range(size):
    for j in range(size):
        if board[i,j] == '-':
            temp_board = copy.deepcopy(board)
            temp_board[i,j] = player
            board_states.append(temp_board)

return board_states
```

## Get the value of the given board

```
def XFWT_single(board):
     X : Extract
     F : Features
     W : Weighted
     T : Total
     For a given board, extract the features and then
      return weighted sum of features.
   # extract features and calculate value
   feature_vector = ExctractFeatures(board)
   board_value = np.dot(weights,feature_vector.T)
   return board_value
def XFWT(boards):
     X : Extract
      F : Features
     W : Weighted
     T : Total
```

```
board_values = []

for board in boards:
    # extract features and calculate value
    feature_vector = ExctractFeatures(board)
    board_value = np.dot(weights,feature_vector.T)
    board_values.append(board_value)

return board_values
```

#### Printing Board

```
def printBoard(board):
    size = board.shape[0]

print("----"*size)
for i in range(size):
    for j in range(size):
        print("| {} | ".format(board[i][j]), end='')
    print()
    print("----"*size)
```

# ▼ Training function

```
def train(size, alpha, num_iters, weights):
   for epoch in tqdm(range(num_iters)):
        curr_board_history = []
        board = initialize board(size)
        current = 'X'
        # till not over
        while (getWinner(board) == -1):
            # get next possible states, and calculate their values
            next_states = np.array(getPossibleStates(board, current))
            np.random.shuffle(next_states)
            next_values = XFWT(next_states)
            # append board to history and set the next board state as the one with maximum va
            curr board history.append(board)
            board = next_states[np.argmax(next_values)]
            # randomly change player to get optimal training
            rand val = np.random.randn(1)
            if (rand val > 0):
                current = 'X' if (current == '0') else '0'
```

```
curr board history.append(board)
        result = getWinner(board)
        # update weights
        for idx, board_state in enumerate(curr_board_history):
            X = ExctractFeatures(board state)
            if ((idx+2) < len(curr_board_history)):</pre>
                weights += alpha * (XFWT_single(curr_board_history[idx+2])
                              - XFWT single(board state))* X
            else:
                weights += alpha * (result - XFWT_single(board_state)) * X
# TRAINING
size = 4
alpha = 0.05
num iters = 1000
num features = size*8 + 1
weights = np.random.randn(num_features)*0.01 # initialize weights
train(size, alpha, num_iters, weights) # train model
           | 1000/1000 [00:22<00:00, 45.27it/s]
```

#### ▼ Test Function

```
def play(size):
   board = initialize_board(size)
   current = 'X'
   # till game is not over
   while (getWinner(board) == -1):
        if current == 'X':
          print('Computer\'s turn : ')
          next_states = np.array(getPossibleStates(board, current))
          next values = XFWT(next states)
          board = next_states[np.argmax(next_values)]
          printBoard(board)
        else:
          print('Human\'s move : ')
          while (True):
            a = int(input('Enter x coordinate : '))
            b = int(input('Enter y coordinate : '))
            if board[a,b] == '-':
              break
            else:
              print("Please select an open position. Try again.")
```

```
board[a,b] = '0'
         printBoard(board)
       current = 'X' if (current == '0') else '0'
   result = getWinner(board)
   if result == 100:
     print('Computer Wins!')
   elif result == -100:
     print('Human Wins!')
   else:
     print('Draw!')
print("Let the game begin : ")
play(size)
    | X || - || U || U |
    | - || - || - || X |
    Computer's turn :
    -----
    | X || X || - || O |
    | 0 || - || x || 0 |
    | X || - || 0 || 0 |
    | X || - || - || X |
    Human's move :
    Enter x coordinate: 0
    Enter y coordinate : 2
    _____
    | x || x || o || o |
    | 0 || - || x || 0 |
    | X || - || 0 || 0 |
    | X || - || - || X |
    -----
    Computer's turn :
    | x | | x | | o | | o |
    | 0 || - || X || 0 |
    | X || - || 0 || 0 |
    | X || - || X || X |
    _____
    Human's move :
    Enter x coordinate : 1
```

Enter y coordinate : 1

x	>	· 〈	   0		0
0	(	)	   X		0
x	-	-	0		0
X	-	·	   X		x
Compu	ıter		+11	rn	
X					0
	>		   0	     	0   0
X	)	· 〈   ·	   0    x		<u>:</u>
X	0	· · · · · · · · · · · · · · · · · · ·	   0    X    0		 0   