### **Environment code**

not a full version, extra features like the pacman discussed are in the works.

```
In [2]:
        import numpy as np
        import gymnasium as gym
        from gymnasium.spaces import Box, Discrete, Tuple
        from colorama import Fore
        import random
        class Connect4Env(gym.Env):
            def __init__(self, width=7, height=6, connect=4):
                self.num_players = 2
                self.width = width
                self.height = height
                self.connect = connect
                player_observation_space = Box(low=-1, high=1,
                                                shape=(self.width, self.height),
                                                dtype=np.int32)
                self.observation_space = player_observation_space
                self.action_space = Tuple([Discrete(self.width) for _ in range(self.num_players)
                self.state_space_size = (self.height * self.width) ** 3
                self.reset()
            def reset(self):
                Initialises the Connect 4 gameboard.
                self.board = np.full((self.width, self.height), -1)
                self.current_player = 0
                self.winner = None
                return self.get_player_observations()
            # 0 - empty
            # -1 - one player
            # 1 - another player
            def get_player_observations(self):
                transformed_array = np.array([list(map(lambda x: 0 if x == -1 else -1 if x == 0
                transposed_array = np.rot90(transformed_array, k=1)
                return transposed_array
            def step(self, movecol):
                0.00
                Applies a move by a player to the game board in a format which is suitable for a
                if not(movecol >= 0 and movecol <= self.width and self.board[movecol][self.heigh</pre>
                    raise IndexError(f'Invalid move. tried to place a chip on column {movecol} w
                row = self.height - 1
                while row >= 0 and self.board[movecol][row] == -1:
                row += 1
```

```
self.board[movecol][row] = self.current_player
    self.current_player = 1 - self.current_player
   self.winner, reward_vector = self.check_for_episode_termination(movecol, row)
   info = {'legal_actions': self.get_moves(),
            'current_player': self.current_player}
    return self.get_player_observations(), reward_vector, \
           self.winner is not None, info
def check_for_episode_termination(self, movecol, row):
   Check for victories in the current state and generate rewards for the state
   winner, reward_vector = self.winner, [0, 0]
   if self.does_move_win(movecol, row):
       winner = 1 - self.current_player
        if winner == 0: reward_vector = [1, -1]
        elif winner == 1: reward_vector = [-1, 1]
   elif self.get_moves() == []: # A draw has happened
       winner = -1
    return winner, reward_vector
def clone(self):
   st = Connect4Env(width=self.width, height=self.height)
   st.current_player = self.current_player
   st.winner = self.winner
   st.board = np.array([self.board[col][:] for col in range(self.width)])
   return st
def get_moves(self):
    :returns: array with columns where there is a possible move
   if self.winner is not None:
        return []
    return [col for col in range(self.width) if self.board[col][self.height - 1] ==
def does_move_win(self, x, y, me=None):
   Checks whether a newly dropped chip at position param x, param y
   wins the game.
   if me is None:
        me = self.board[x][y]
   for dx, dy in [(0, +1), (+1, +1), (+1, 0), (+1, -1)]:
        p = 1 # positive direction
       while self.is_on_board(x+p*dx, y+p*dy) and self.board[x+p*dx][y+p*dy] == me:
            p += 1
        n = 1 # negative direction
        while self.is_on_board(x-n*dx, y-n*dy) and self.board[x-n*dx][y-n*dy] == me:
            n += 1
        if p + n >= (self.connect + 1): # want (p-1) + (n-1) + 1 >= 4, or more simpl
           return True
    return False
# swaps one random token of player A with another random token of player B
def swap_random_tokens(self):
   while True:
        old_col = random.choice(range(self.height))
        old_row = random.choice(range(self.width))
        old_tile_val = self.board[old_row][old_col]
       # print("old_col: ", old_col)
```

```
# print("old_tile_val: ", old_tile_val)
        if old_tile_val == -1:
            # print("old_tile_val is -1")
            continue
        new_col = random.choice(range(self.height))
        new_row = random.choice(range(self.width))
        new_tile_val = self.board[new_row][new_col]
        # print("new_col: ", new_col)
        # print("new_row: ", new_row)
        # print("new_tile_val: ", new_tile_val)
        if new_tile_val == -1 or new_tile_val == old_tile_val:
            # print("new tile val is -1 or same as the other")
            continue
        if self.does_move_win(new_row, new_col, me=old_tile_val):
            # print("move wins")
            continue
        if self.does_move_win(old_row, old_col, me=new_tile_val):
            # print("move wins")
            continue
        # print("yeeeeeee")
        self.board[old_row][old_col], self.board[new_row][new_col] = self.board[new_
        break
def is_on_board(self, x, y):
    return x \ge 0 and x < self.width and <math>y \ge 0 and y < self.height
def get_result(self, player):
   if self.winner == -1: return 0 # A draw occurred
    return +1 if player == self.winner else -1
def render(self, mode='human'):
   if mode != 'human': raise NotImplementedError('Rendering has not been coded yet'
   for x in range(self.height - 1, -1, -1):
        for y in range(self.width):
            s += \{-1: Fore.WHITE + '.', 0: Fore.RED + 'X', 1: Fore.YELLOW + '0'\}[sel]
            s += Fore.RESET
        s += "\n"
   print(s)
```

## MCTS (Monte Carlo Tree Search)

# print("old\_row: ", old\_row)

```
In [3]: import numpy as np
import random
import math
```

```
import copy
import pickle
from multiprocessing import Pool, cpu_count, Manager
class Node:
   def __init__(self, state, parent=None, move=None):
        self.state = state
        self.parent = parent
        self.move = move
        self.children = []
        self.visits = 0
        self.value = 0
    def is_fully_expanded(self):
        return len(self.children) == len(self.state.get_moves())
    def best_child(self, c_param=1.4):
        choices_weights = [
            (child.value / child.visits) + c_param * math.sqrt((2 * math.log(self.visits)
            for child in self.children
        return self.children[np.argmax(choices_weights)]
    def most_visited_child(self):
        visits = [child.visits for child in self.children]
        return self.children[np.argmax(visits)]
    def expand(self):
        moves = self.state.get_moves()
        for move in moves:
            if not any(child.move == move for child in self.children):
                next_state = copy.deepcopy(self.state)
                next_state.step(move)
                child_node = Node(next_state, self, move)
                self.children.append(child_node)
                return child node
    def rollout(self):
        current_state = copy.deepcopy(self.state)
        while current_state.winner is None:
            possible_moves = current_state.get_moves()
            move = random.choice(possible_moves)
            current_state.step(move)
        return current_state.get_result(1 - current_state.current_player)
    def backpropagate(self, result):
        self.visits += 1
        self.value += result
        if self.parent:
            self.parent.backpropagate(-result)
class MCTS:
   def __init__(self, env, q_table, num_simulations=1000):
        self.env = env
        self.q_table = q_table
        self.num_simulations = num_simulations
    def search(self, state):
        root = Node(state)
        for _ in range(self.num_simulations):
            node = root
            while node.is_fully_expanded() and node.children:
                node = node.best_child()
            if not node.is_fully_expanded():
```

```
node = node.expand()
            result = node.rollout()
            node.backpropagate(result)
            self.update_q_table(node)
        return root.most_visited_child().move
    def update_q_table(self, node):
        while node:
            state_str = ''.join(map(str, node.state.board.flatten()))
            key = (state_str, node.move)
            if key not in self.q_table:
                self.q_table[key] = 0
            self.q_table[key] += node.value
            node = node.parent
episode = 0
def get_state_action_key(state, action):
    state_str = ''.join(map(str, state.flatten()))
    return (state_str, action)
def run_episode(args):
    global episode
    episode += 1
    q_table, num_simulations = args
    env = Connect4Env()
    mcts = MCTS(env, q_table, num_simulations=num_simulations)
    state = env.reset()
    done = False
    while not done:
        if env.current_player == 0:
            action = mcts.search(env.clone())
        else:
            possible_moves = env.get_moves()
            action = random.choice(possible_moves)
        next_state, reward, done, info = env.step(action)
        key = get_state_action_key(state, action)
        if key not in q_table:
            q_table[key] = 0
        q_table[key] += reward[env.current_player]
        state = next_state
        if done:
            break
    print(f'episode {episode} done')
    return q_table
def train_mcts(num_episodes=100, num_simulations=1000):
    with Manager() as manager:
        q_table = manager.dict()
        with Pool(processes=cpu_count()) as pool:
            q_tables = pool.map(run_episode, [(q_table, num_simulations) for _ in range(
        q_table = dict(q_table) # Convert manager.dict() to a regular dictionary
        with open('q_table.pkl', 'wb') as f:
            pickle.dump(q_table, f)
```

```
print("Q-table generated and saved to 'q_table.pkl'")
```

### minimax

```
In [4]: | ## minimax
        import math
        import multiprocessing as mp
        def minimax(node, depth, alpha, beta, maximizingPlayer):
            if depth == 0 or node.winner is not None:
                 return node.get_result(maximizingPlayer)
            valid_moves = node.get_moves()
            if maximizingPlayer:
                value = -math.inf
                for move in valid_moves:
                     child = node.clone()
                     child.step(move)
                    value = max(value, minimax(child, depth-1, alpha, beta, False))
                     alpha = max(alpha, value)
                     if alpha >= beta:
                         break
                 return value
            else:
                 value = math.inf
                for move in valid_moves:
                     child = node.clone()
                     child.step(move)
                    value = min(value, minimax(child, depth-1, alpha, beta, True))
                     beta = min(beta, value)
                     if alpha >= beta:
                         break
                 return value
        def minimax_worker(child, depth, alpha, beta, maximizingPlayer, return_dict, idx):
            return_dict[idx] = minimax(child, depth, alpha, beta, maximizingPlayer)
        def minimax_decision(env, depth=4):
            manager = mp.Manager()
            return_dict = manager.dict()
            jobs = []
            best_value = -math.inf
            best_move = None
            valid_moves = env.get_moves()
            for idx, move in enumerate(valid_moves):
                child = env.clone()
                child.step(move)
                p = mp.Process(target=minimax_worker, args=(child, depth-1, -math.inf, math.inf,
                 jobs.append(p)
                p.start()
            for job in jobs:
                job.join()
            for idx, move in enumerate(valid_moves):
                value = return_dict[idx]
                if value > best_value:
                     best_value = value
                     best_move = move
            return best_move
```

#### scenarios

```
def minimax_vs_mcts(q_table, minimax_depth=4):
In [5]:
             env = Connect4Env()
            mcts = MCTS(env, q_table, num_simulations=1000)
             state = env.reset()
             done = False
             print("Starting a game of Connect 4!")
            env.render()
            while not done:
                 if env.current_player == 0:
                     print("MCTS Agent's turn:")
                     action = mcts.search(env.clone())
                 else:
                     print("Minimax Agent's turn:")
                     action = minimax_decision(env.clone(), minimax_depth)
                 state, reward, done, info = env.step(action)
                 env.render()
                 if done:
                     if env.winner == -1:
                         print("It's a draw!")
                     elif env.winner == 0:
                         print("MCTS Agent wins!")
                     else:
                         print("Minimax Agent wins!")
                     break
In [7]: def mcts_vs_mcts(q_table):
            env = Connect4Env()
            mcts = MCTS(env.clone(), q_table, num_simulations=1000)
            mcts2 = MCTS(env.clone(), q_table, num_simulations=1000)
            state = env.reset()
             done = False
             print("Starting a game of Connect 4!")
            env.render()
             while not done:
                 if env.current_player == 0:
                     print("MCTS1 Agent's turn:")
                     action = mcts.search(env.clone())
                 else:
                     print("MCTS2 Agent's turn:")
                     action = mcts2.search(env.clone())
                 state, reward, done, info = env.step(action)
                 env.render()
                 if done:
                     if env.winner == -1:
                         print("It's a draw!")
                     elif env.winner == 0:
                         print("MCTS1 Agent wins!")
                     else:
                         print("MCTS2 Agent wins!")
                     break
```

```
In [10]: def human_vs_mcts(q_table):
             env = Connect4Env()
             mcts = MCTS(env, q_table, num_simulations=1000)
             state = env.reset()
             done = False
             print("Starting a game of Connect 4!")
             env.render()
             while not done:
                  if env.current_player == 0:
                      print("MCTS Agent's turn:")
                      action = mcts.search(env.clone())
                 else:
                      valid_move = False
                      while not valid_move:
                          try:
                              action = int(input("Your turn! Enter the column number (0-6): "))
                              if action in env.get_moves():
                                  valid_move = True
                              else:
                                  print("Invalid move. Column is full or out of range. Try again."
                          except ValueError:
                              print("Invalid input. Please enter a number between 0 and 6.")
                 state, reward, done, info = env.step(action)
                 env.render()
                 if done:
                      if env.winner == -1:
                          print("It's a draw!")
                      elif env.winner == 0:
                          print("MCTS Agent wins!")
                      else:
                          print("You win!")
                      break
```

# Agent vs Agent play

Both agents utilize the same Q-table. Agents are able to draw, agents are able to win, which means this is not a fully optimal solution since we know the game is solved, so we are not fully ready with that.

Below we are able to run the code for 2 agents to play against each other with a pre-trained Q-table. Settings: 100 episodes, 1000 simulation depth.

```
MCTS1 Agent's turn:
X . . . . .
MCTS2 Agent's turn:
0.....
X . . . . . .
MCTS1 Agent's turn:
0.....
\mathsf{X}\ldots\mathsf{X}\ldots
MCTS2 Agent's turn:
0.....
X0.X...
MCTS1 Agent's turn:
X . . . . . .
0.....
X0.X...
MCTS2 Agent's turn:
X . . . . .
00 . . . .
X0 . X . . .
MCTS1 Agent's turn:
XX . . . .
00....
X0.X...
MCTS2 Agent's turn:
XX . . . . .
00....
X00X . . .
MCTS1 Agent's turn:
```

```
. X . . . .
XX . . . . .
00....
X00X...
MCTS2 Agent's turn:
. X . . . . .
XX . . . .
00....
X00X0..
MCTS1 Agent's turn:
. X . . . . .
\mathsf{XX}\dots\dots
00X....
X00X0..
MCTS2 Agent's turn:
0X . . . .
XX . . . .
00X....
X00X0..
MCTS1 Agent's turn:
. X . . . . .
\mathsf{0X}\dots\dots
XX . . . . .
00X....
X00X0..
MCTS2 Agent's turn:
. 0 . . . . .
. X . . . . .
0X . . . .
\mathsf{XX}\dots\dots
00X....
X00X0..
MCTS1 Agent's turn:
.0....
XX . . . .
0X . . . .
ХХ . . . . .
00X....
X00X0..
MCTS2 Agent's turn:
. 0 . . . .
XX . . . . .
0X . . . .
XX . . . . .
00X....
X00X00.
MCTS1 Agent's turn:
. 0 . . . . .
XX . . . . .
\mathsf{0X}\dots\dots
```

```
XXX . . . .
           00X....
           X00X00.
           MCTS2 Agent's turn:
           . 0 . . . . .
           XX . . . . .
           0X0....
           XXX . . . .
           00X....
           X00X00.
           MCTS1 Agent's turn:
           .0....
           XX . . . . .
0X0 . . . .
           \mathsf{XXX}\dots
           00XX...
           X00X00.
           MCTS1 Agent wins!
In [ ]:
```