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PAGE:- C PC-33
C-3

PAGE NO.:	

AI

LAB ASSIGNMENT - 2

AIM:- To solve tic tac toe using minmax algorithm.

OBJECTIVE:- To study and implement min-max algorithm for tic tac toe.

THEORY:- Adversarial search:-

This is the search when there is an enemy changing the state of problem every step in a direction you don't want. Eg. Chess etc.

Tic Tac Toe Solving Steps:-

- 1) Check if game has reached terminal state & return a value depending on the outcome.
- 2) Generate all available moves
- 3) Call the minmax fun. on every available step move recursively to reach terminal state.
- 4) Evaluate collection of sorted moves.
- 5) Return optimal move.

MINMAX:- It is a kind of back tracking it used in decision making & game theory to find optimal value of player.

INPUT:- Initial stage

OUTPUT:- Final stage.

FAQ's

Q.1) compare informed search & adversarial search?

→ Adversarial search is a search where we examine the problem which arises when we try to plan ahead of world and other agent are planning against us.

→ Informed search is more careful for large search spaces. It used concept of heuristics

Q.2 What is Alpha-beta pruning?

- = Modified version of min-max algorithm.
- Optimization technique for min-max algorithm.
- Can be applied at any depth of tree sometimes it only prunes entire subtree.

Q.3 Min Max Algorithm.

function MinMax(node, depth, maximizing player):

if depth = 0 or node = terminal node then
return static evaluation of node.

if maximizing player then
return max Eva = ∞ ;

for each child node do

eva = minmax(child, depth-1, false)

max eva = max(max Eva, eva)

return max Eva

else

min Eva = infinity;

for each child of node:

eva = minmax(child, depth-1, true)

min eva = min(min eva, eva)

return min eva;

AI lab 2 Code

```
from math import inf as infinity
from random import choice
import platform
import time
from os import system

HUMAN = -1
COMP = +1
board = [
    [0, 0, 0],
    [0, 0, 0],
    [0, 0, 0],
]

def evaluate(state):
    if wins(state, COMP):
        score = +1
    elif wins(state, HUMAN):
        score = -1
    else:
        score = 0

    return score

def wins(state, player):
    win_state = [
        [state[0][0], state[0][1], state[0][2]],
        [state[1][0], state[1][1], state[1][2]],
        [state[2][0], state[2][1], state[2][2]],
        [state[0][0], state[1][0], state[2][0]],
        [state[0][1], state[1][1], state[2][1]],
        [state[0][2], state[1][2], state[2][2]],
        [state[0][0], state[1][1], state[2][2]],
        [state[2][0], state[1][1], state[0][2]],
    ]
    if [player, player, player] in win_state:
        return True
    else:
        return False

def game_over(state):
    return wins(state, HUMAN) or wins(state, COMP)

def empty_cells(state):
```



```

cells = []

for x, row in enumerate(state):
    for y, cell in enumerate(row):
        if cell == 0:
            cells.append([x, y])

return cells

def valid_move(x, y):
    if [x, y] in empty_cells(board):
        return True
    else:
        return False

def set_move(x, y, player):
    if valid_move(x, y):
        board[x][y] = player
        return True
    else:
        return False

def minimax(state, depth, player):
    if player == COMP:
        best = [-1, -1, -infinity]
    else:
        best = [-1, -1, +infinity]

    if depth == 0 or game_over(state):
        score = evaluate(state)
        return [-1, -1, score]

    for cell in empty_cells(state):
        x, y = cell[0], cell[1]
        state[x][y] = player
        score = minimax(state, depth - 1, -player)
        state[x][y] = 0
        score[0], score[1] = x, y

        if player == COMP:
            if score[2] > best[2]:
                best = score # max value
        else:
            if score[2] < best[2]:
                best = score # min value

```

```
return best
```

```
def render(state, c_choice, h_choice):
    chars = {
        -1: h_choice,
        +1: c_choice,
        0: ' '
    }
    str_line = '-----'

    print('\n' + str_line)
    for row in state:
        for cell in row:
            symbol = chars[cell]
            print(f'| {symbol} |', end='')
        print('\n' + str_line)
```

```
def ai_turn(c_choice, h_choice):
    depth = len(empty_cells(board))
    if depth == 0 or game_over(board):
        return

    print(f'Computer turn [{c_choice}]')
    render(board, c_choice, h_choice)

    if depth == 9:
        x = choice([0, 1, 2])
        y = choice([0, 1, 2])
    else:
        move = minimax(board, depth, COMP)
        x, y = move[0], move[1]

    set_move(x, y, COMP)
    time.sleep(1)
```

```
def human_turn(c_choice, h_choice):
    depth = len(empty_cells(board))
    if depth == 0 or game_over(board):
        return

    move = -1
    moves = {
        1: [0, 0], 2: [0, 1], 3: [0, 2],
        4: [1, 0], 5: [1, 1], 6: [1, 2],
        7: [2, 0], 8: [2, 1], 9: [2, 2],
    }
```

```

print(f'Human turn [{h_choice}]\n')
render(board, c_choice, h_choice)

while move < 1 or move > 9:
    try:
        move = int(input('Use numpad (1..9): '))
        coord = moves[move]
        can_move = set_move(coord[0], coord[1], HUMAN)

        if not can_move:
            print('Bad move')
            move = -1
    except (EOFError, KeyboardInterrupt):
        print('Bye')
        exit()
    except (KeyError, ValueError):
        print('Bad choice')

def main():
    #clean()
    h_choice = '' # X or O
    c_choice = '' # X or O
    first = '' # if human is the first

    while h_choice != 'O' and h_choice != 'X':
        try:
            print('')
            h_choice = input('Choose X or O\nChosen: ').upper()
        except (EOFError, KeyboardInterrupt):
            print('Bye')
            exit()
        except (KeyError, ValueError):
            print('Bad choice')

    if h_choice == 'X':
        c_choice = 'O'
    else:
        c_choice = 'X'

    while first != 'Y' and first != 'N':
        try:
            first = input('First to start?[y/n]: ').upper()
        except (EOFError, KeyboardInterrupt):
            print('Bye')
            exit()
        except (KeyError, ValueError):
            print('Bad choice')

```

```

while len(empty_cells(board)) > 0 and not game_over(board):
    if first == 'N':
        ai_turn(c_choice, h_choice)
        first = ''

    human_turn(c_choice, h_choice)
    ai_turn(c_choice, h_choice)

    if wins(board, HUMAN):
        print(f'Human turn [{h_choice}]')
        render(board, c_choice, h_choice)
        print('YOU WIN!')
    elif wins(board, COMP):
        print(f'Computer turn [{c_choice}]')
        render(board, c_choice, h_choice)
        print('YOU LOSE!')
    else:
        render(board, c_choice, h_choice)
        print('DRAW!')

    exit()

if __name__ == '__main__':
    main()

```

'''

OUTPUT :-

Choose X or O

Chosen: X

First to start?[y/n]: y

Human turn [X]

```

-----
|  |  |  |  |
-----
|  |  |  |  |
-----
|  |  |  |  |
-----

```

Use numpad (1..9): 1

Computer turn [O]

```

-----
| X |  |  |  |
-----
|  |  |  |  |
-----

```

```
|  |  |  |
-----
```

Human turn [X]

```
-----
| X |  |  |
-----
```

```
|  | O |  |
-----
```

```
|  |  |  |
-----
```

Use numpad (1..9): 3

Computer turn [O]

```
-----
| X |  |  | X |
-----
```

```
|  | O |  |
-----
```

```
|  |  |  |
-----
```

Human turn [X]

```
-----
| X | O |  | X |
-----
```

```
|  | O |  |
-----
```

```
|  |  |  |
-----
```

Use numpad (1..9): 8

Computer turn [O]

```
-----
| X | O |  | X |
-----
```

```
|  | O |  |
-----
```

```
|  | X |  |
-----
```

Human turn [X]

```
-----
| X | O |  | X |
-----
```

```
| O | O |  |
-----
```

```
|  | X |  |
-----
```

Use numpad (1..9): 6

Computer turn [0]

```
-----
| x || o || x |
-----
| o || o || x |
-----
|   || x ||   |
-----
```

Human turn [X]

```
-----
| x || o || x |
-----
| o || o || x |
-----
|   || x || o |
-----
```

Use numpad (1..9): 7

```
-----
| x || o || x |
-----
| o || o || x |
-----
| x || x || o |
-----
```

DRAW!

...