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Language: Python3

```
import sys
print(sys.version)

3.10.12 (main, Nov 20 2023, 15:14:05) [GCC 11.4.0]
```

List of Libraries used: Numpy ,Tensorflow , keras.

```
print(tensorflow.__version__)
print(keras.__version__)
print(np.__version__)
print(matplotlib.__version__)
2.15.0
2.15.0
1.23.5
3.7.1
```

RL FRAMEWORK:

DEEP Q NETWORK

Learning rate = 0. 00025

Dense(128, activation='relu', input shape=(9,)

Dense(128, activation='relu')

Dense(9, activation='linear')

loss=mean_squared_error

optimizer = RMS propogation

A similar style for target and train networks .

No of episodes – 10,000

Epsilon - 0.4

Discount Factor - 0.7

Minimum size of experience replay buffer – 4096

Maximum size of experience replay buffer – 1048576

Evaluation on random player mode. 1000 times

Wins: 546 Draws: 181 Loss: 273

DETAILLS ABOUT ENVIRONMENT CLASS:

I have created the tic tac toe environment.

1.Initialization:

- The **Tictactoe_v0** class is initialized with an empty game board, winning positions, and the current player's turn.
- The winning positions are defined for rows, columns, and diagonals.

2. Reset Method:

- The **reset** method resets the game board, player turn, and player marks based on whether the human or the agent should make the first move.
- If the agent starts, it calls the **env_act** method to simulate the agent's move.

3. Winning Check:

- The **check_win** method examines the board for winning positions or a tie.
- It returns:
 - 1 if the player wins.
 - -1 if the opponent wins.
 - 0 in case of a tie.
 - True if the game is done, and False otherwise.

4. Environment Action:

- The **env_act** method simulates the environment taking an action.
- It selects a random available move unless there is an opportunity for the player or opponent to win, in which case it makes a strategic move.
- It updates the board, checks for a win or tie, and switches the current player's turn.

5. Step Method:

- The **step** method is designed for the interaction between the agent and the environment during the training loop.
- It takes an action as input, updates the board, checks for a win or tie, and switches the current player's turn.
- It returns the current state of the board, the reward based on the game outcome, whether the game is done.