# Q Player's Interleaved Tabular Q-Learning Algorithm

Can be found in the file ./tictactoe/players.py. This contains a class QPlayer which contains the functions train\_single\_game and train that implement the following pseudo-code.

### Algorithm 1 Q Player's interleaved tabular Q-Learning algorithm

```
1: Initialize Q_1, Q_2 arbitrarily
 2: for a total of N games do
          Initialize s as the empty board
 3:
 4:
          Take \epsilon-max-greedy action a and observe next state \tilde{s}
          Take \epsilon-min-greedy action \tilde{a} and observe next state s'
 5:
          while s' non-terminal do
 6:
 7:
               if in state s Player 1 is to move then
                    Q_1(s, a) \leftarrow Q_1(s, a) + \alpha [0 + \gamma \max_{a'} Q_1(s', a') - Q_1(s, a)]
 8:
                    Take \epsilon-max-greedy action a' and observe next state \tilde{s}'
 9:
               else if in state s Player 2 is to move then
10:
                    Q_2(s, a) \leftarrow Q_2(s, a) + \alpha [0 + \gamma \min_{a'} Q_2(s', a') - Q_2(s, a)]
11:
                    Take \epsilon-min-greedy action a' and observe next state \tilde{s}'
12:
               (s, a, \tilde{s}, \tilde{a}, s') \leftarrow (\tilde{s}, \tilde{a}, s', a', \tilde{s}')
13:
          If Player 1 won, do r \leftarrow 1, else if Player 2 won, do r \leftarrow -1, else, do r \leftarrow 0
14:
          if in state s Player 1 is to move then
15:
               Q_1(s,a) \leftarrow Q_1(s,a) + \alpha |r - Q_1(s,a)|
16:
               Q_2(\tilde{s}, \tilde{a}) \leftarrow Q_2(\tilde{s}, \tilde{a}) + \alpha [r - Q_2(\tilde{s}, \tilde{a})]
17:
          else if in state s Player 2 is to move then
18:
               Q_2(s,a) \leftarrow Q_2(s,a) + \alpha | r - Q_2(s,a) |
19:
               Q_1(\tilde{s}, \tilde{a}) \leftarrow Q_1(\tilde{s}, \tilde{a}) + \alpha [r - Q_1(\tilde{s}, \tilde{a})]
20:
21: return Q_1, Q_2
```

**Remark.** Given a game state s, the  $\epsilon$ -max-greedy action is a random action with probability  $\epsilon$  or equal to  $\arg \max_a Q_1(s,a)$  with probability  $1-\epsilon$ . The  $\epsilon$ -min-greedy action is a random action with probability  $\epsilon$  or equal to  $\arg \min_a Q_1(s,a)$  with probability  $1-\epsilon$ .

**Parameters.** Have to be chosen at the beginning.

- 1. Step-size  $\alpha \in (0,1]$
- 2. Greedy-factor  $\epsilon \in [0, 1]$
- 3. Discount factor  $\gamma$
- 4. Number of games  $N \in \mathbb{N}$ .

# TD Player's Adjusted Tabular TD( $\lambda$ ) Algorithm

Can be found in the file ./tictactoe/players.py. This contains a class TDPlayer which contains the functions train\_single\_game and train that implement the following pseudo-code.

### **Algorithm 2** TD Player's adjusted tabular TD( $\lambda$ ) algorithm

```
1: Initialize V arbitrarily
 2: Choose \alpha \in (0,1] and \epsilon \in [0,1]
 3: for a total of N games do
          Initialize s as the empty board
 4:
          Take \epsilon-max-min-greedy action a and observe next state s'
 5:
 6:
          Initialize empty dictionary z
          while s' non-terminal do
 7:
               If \mathbf{z}(s) is empty, do \mathbf{z}(s) \leftarrow 1, else do \mathbf{z}(s) \leftarrow \mathbf{z}(s) + 1
 8:
               for all x in z do
 9:
                    V(x) \leftarrow V(x) + \alpha[0 + \gamma V(s') - V(s)]\mathbf{z}(x)
10:
11:
                    \mathbf{z}(x) \leftarrow \lambda \gamma \mathbf{z}(x)
               Take \epsilon-max-min-greedy action a and observe next state s''
12:
13:
               (s,s') \leftarrow (s',s'')
          If \mathbf{z}(s) is empty, do \mathbf{z}(s) \leftarrow 1, else do \mathbf{z}(s) \leftarrow \mathbf{z}(s) + 1
14:
          If Player 1 won, do r \leftarrow 1, else if Player 2 won, do r \leftarrow -1, else, do r \leftarrow 0
15:
          for all x in z do
16:
               V(x) \leftarrow V(x) + \alpha[r - V(s)]\mathbf{z}(x)
17:
          V(s') \leftarrow r(s')
18:
19: \mathbf{return}\ V
```

**Remark.** Given a game state s, the  $\epsilon$ -max-min-greedy action is a random action with probability  $\epsilon$  or with probability  $1 - \epsilon$  it is equal to  $\arg \max_a V(s_a)$  if it is Player 1's turn and  $\arg \min_a V(s_a)$  if it is Player 2's turn. State  $s_a$  is the state that follows after applying action a to s.

Parameters. Have to be chosen at the beginning.

- 1. Step-size  $\alpha \in (0,1]$
- 2. Greedy-factor  $\epsilon \in [0, 1]$
- 3. Discount factor  $\gamma$
- 4. Number of games  $N \in \mathbb{N}$ .

# Deep Player's Adjusted Approximate Monte Carlo Algorithm

Can be found in the file ./tictactoe/players.py and ./connectfour/players.py. They contain a class DeepPlayer which contain the functions train\_single\_game and train that implement the following pseudo-code.

#### Algorithm 3 Deep Player's adjusted approximate Monte Carlo algorithm

```
1: Initialize \theta \in \mathbb{R}^d arbitrarily
 2: for i = 1, ..., N do
          Initialize s_0 as the empty board
 3:
          Generate a sequence of states s_0, ..., s_T according to \epsilon-max-min-greedy policy
 4:
          If Eva won, do r \leftarrow 1, else if Odin won, do r \leftarrow -1, else, do r \leftarrow 0
 5:
          for t = T, ..., 0 do
 6:
               \theta \leftarrow \theta - \alpha \left[ v(\theta, s_t) - (\gamma^{T-t}r + 1)/2 \right] \nabla v(\theta, s_t)
 7:
          if i \mod n \equiv 0 then
 8:
                \alpha \leftarrow \alpha \cdot b and \epsilon \leftarrow \epsilon \cdot c
 9:
10: \mathbf{return}\ V
```

**Remark.** Given a game state s, the  $\epsilon$ -max-min-greedy action is a random action with probability  $\epsilon$  or with probability  $1 - \epsilon$  it is equal to  $\arg \max_a V(s_a)$  if it is Player 1's turn and  $\arg \min_a V(s_a)$  if it is Player 2's turn. State  $s_a$  is the state that follows after applying action a to s.

Parameters. Have to be chosen at the beginning.

- 1. Step-size  $\alpha \in (0,1]$
- 2. Greedy-factor  $\epsilon \in [0, 1]$
- 3. Discount factor  $\gamma$
- 4. Number of games  $N \in \mathbb{N}$ .
- 5. Decrease parameters  $n \in \mathbb{N}$ ,  $b, c \in [0, 1]$
- 6. Function approximator v which is based on a Neural network