



# Udacity Multi Agent Control Project

Prepared for: Udacity Deep Reinforcement Learning Nano Degree

Prepared by: Divyanshu Kalra

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## SUMMARY

### Algorithm

The algorithm used is Soft Actor Critic Algorithm which is derived from entropy based RL learning.

Explanation:

Similar to TD3 simultaneously learn two Q-functions,  $Q_{\phi_1}$ , and  $Q_{\phi_2}$ , by mean square Bellman error minimisation. The Q- functions are learned by comparing MSE on the same target as shown Equation 1 and the policy is learned by maximising  $Q_{\phi_1}$  similar that of DDPG and TD3 as shown in Equation 2.

$$L(\phi_i, \mathcal{D}) = \mathbb{E}_{(s,a,r,s',d) \sim \mathcal{D}} \left[ \left( Q_{\phi_i}(s, a) - (r + \gamma(1 - d)V_{\psi_{\text{targ}}}(s')) \right)^2 \right]$$

Equation 1.

$$\max_{\theta} \mathbb{E}_{\substack{s \sim \mathcal{D} \\ \xi \sim \mathcal{N}}} [Q_{\phi_1}(s, \tilde{a}_{\theta}(s, \xi)) - \alpha \log \pi_{\theta}(\tilde{a}_{\theta}(s, \xi)|s)]$$

Equation 2.

$$L(\psi, \mathcal{D}) = \mathbb{E}_{\substack{s \sim \mathcal{D} \\ \tilde{a} \sim \pi_{\theta}}} \left[ \left( V_{\psi}(s) - \left( \min_{i=1,2} Q_{\phi_i}(s, \tilde{a}) - \alpha \log \pi_{\theta}(\tilde{a}|s) \right) \right)^2 \right]$$

Equation 3.

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## IMPLEMENTATION DETAILS

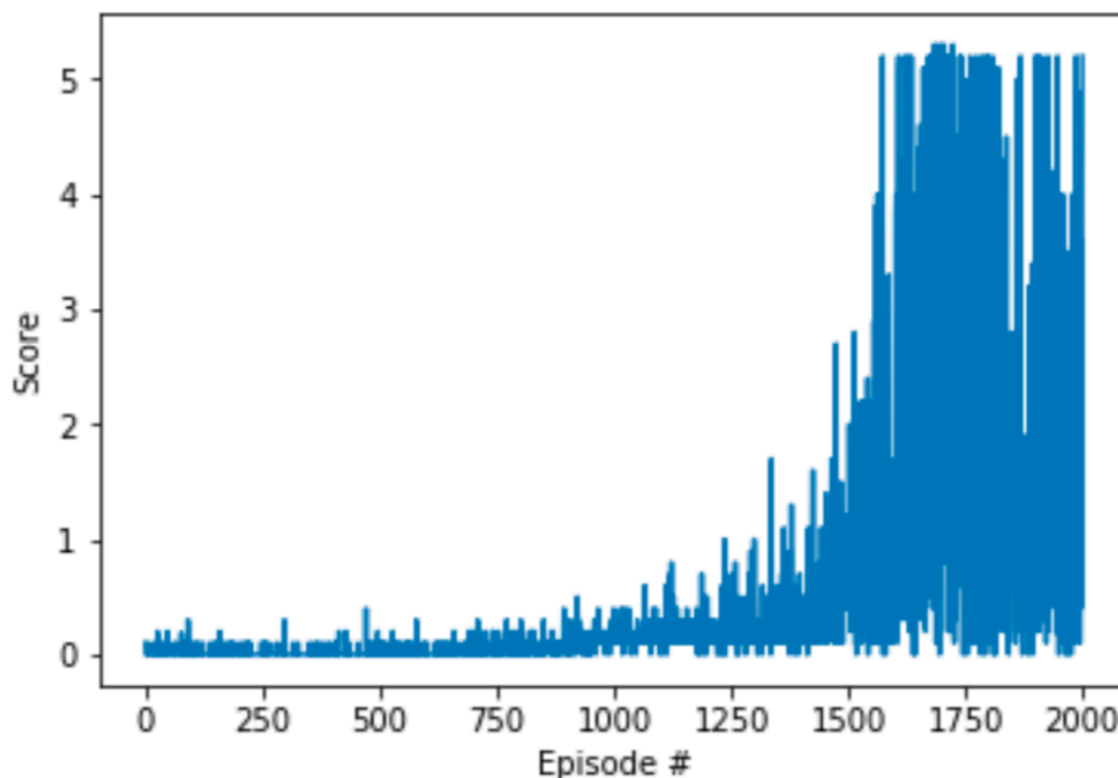
### **Model.py :**

The actor model used for transforming the state information vector into the action vector is written in the class Actor. The actor class also returns the log probability of the action sampled and the mean and standard deviation of the action distribution. The Q - critic model used for transforming the state information vector and action vector into the action-value function is written in the class Critic. The V - critic model used for transforming the state information vector into state value function is written in the class Critic.

### **Tennis.ipynb:**

The Train function runs multiple episodes of the agent's interactions with the environment for a certain number of transitions. The network stores the average reward collected by the agent over the 100 episodes. Once the agent accumulates over the 0.5+ reward in 100 episodes the environment is considered to be solved and the model's weights are saved.

## REWARD PLOT



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## FUTURE IDEAS

- 1) Implementing Actor Attention Critic
- 2) Implementing prioritised replay.