

Learning Algorithm

The learning algorithm is (Deep Deterministic Policy Gradient) DDPG, which combines DPG (a RL algorithm designed for continuous spaces) with the Deep Learning approach used in Deep Q learning.

The algorithm is actor-critic, where the actor outputs deterministic actions and the critic estimates the Q-value function.

Exploration is added through noise to the deterministic action, implemented through the class OUNoise (hyperparams standard_deviation=0.2)

Experience replay allows breaking correlations of sequential actions. We use a replay buffer of size $1e6$

DDPG uses target networks both for actor and critic, this stabilises learning by not updating the target during a few iterations.

One of the main features of DDPG is the soft updates, which brings stabilisation by gradually updating the target network. This is done through the $\text{TAU} = 1e-3$

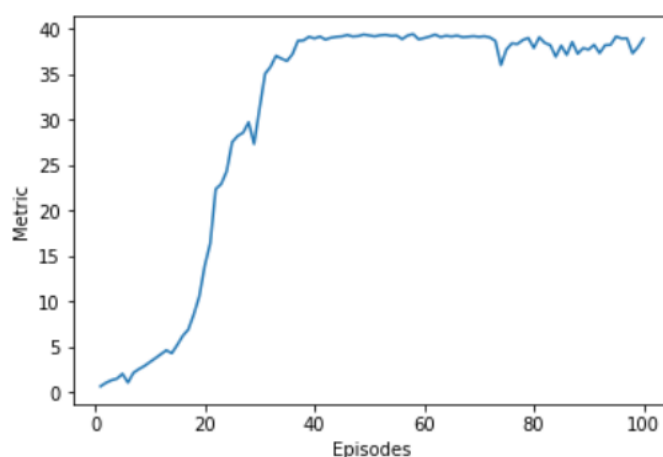
Other hyperparams used are:

```
BATCH_SIZE = 128    # minibatch size
GAMMA = 0.99        # discount factor
LR_ACTOR = 1e-4      # learning rate of the actor
LR_CRITIC = 1e-3     # learning rate of the critic
WEIGHT_DECAY = 0     # L2 weight decay
```

Actor Architecture Layers Size: [128, 256, 4]

Critic Architecture Layers: [128, 256(256+4), 1]

Plot of Rewards



Ideas for Future Work

- 'Hyperparameter Tuning' - Automated strategy to find the best hyperparameters, I have just modified slightly what we learn in the lessons
- Comparison Across other models... PPO, A3C...