# Project Report

## Learning algorithm

The learning algorithm used is vanilla Deep Deterministic Policy Gradient Q Learning as described in this paper https://arxiv.org/abs/1509.02971.

#### The actor network has following layers:

- Fully connected layer input: 33 (state size) output: 256
- Fully connected layer input: 256 output 128
- Fully connected layer input: 128 output: 4 (action size)

#### The critic network has following layers:

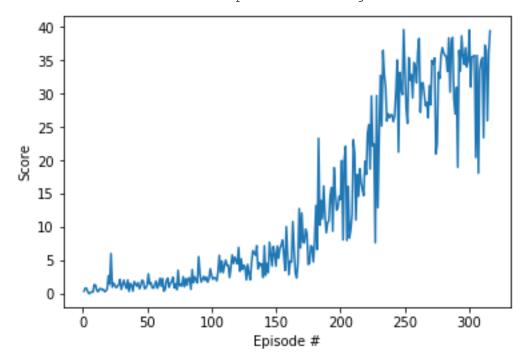
- Fully connected layer input: 33 (state size) output: 256
- Fully connected layer input: 260 output 128
- Fully connected layer input: 128 output: 1

#### Parameters used for the DDPG agent:

- BUFFER SIZE = int(1e5) # replay buffer size
- BATCH SIZE = 128 # minibatch size
- GAMMA = 0.99 # discount factor
- TAU = 1e-3 # for soft update of target parameters
- LR ACTOR = 1e-4 # learning rate of the actor
- LR CRITIC = 1e-4 # learning rate of the critic
- WEIGHT\_DECAY = 0 # L2 weight decay

### Results:

```
Episode 0 Average Score: 0.39
Episode 50 Average Score: 1.17
Episode 100 Average Score: 1.54
Episode 150 Average Score: 3.20
Episode 200 Average Score: 7.06
Episode 250 Average Score: 16.19
Episode 300 Average Score: 27.66
```



## Future work:

- Better hyperparameter tuning
- Solving version 2 of the problem zith 20 simultaneous agents
- Try other algorithms like REINFORCE, TNPG, RWR, REPS, TRPO, CEM, CMA-ES and cmp are them tp DDPG