

# Project 1: Navigation

## Report

To solve this environment I have adapted the DQN agent from demonstration notebooks. It uses a fully connected neural network with 2 hidden layers, each with 64 neurons. The network is learning from replay buffer using soft-updates after every 4 steps.

For training I have also used  $\epsilon$ -greedy policy selection to encourage exploration.  $\epsilon$  starts from 1.0 and goes down to 0.01 with decay factor of 0.995. Other hyper-parameters include: learning rate of the optimizer  $\alpha = 5 \times 10^{-4}$ , soft-update parameter tau  $\tau = 1 \times 10^{-3}$ , and the discount factor  $\gamma = 0.99$ .

Solving the environment means achieving average score over the last 100 episodes of +13. The environment was solved before 1700 episodes. After that, the average score increased slightly but never went over 14. The network weights are saved in "`solved_model.pth`" file.

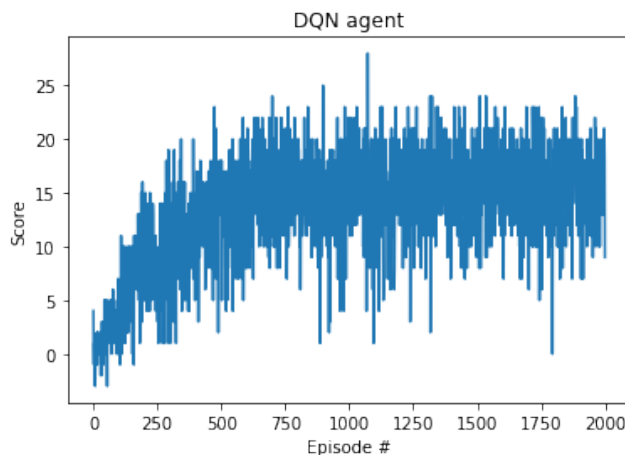


Figure 1: DQN training score over 2000 episodes

There are several potential improvements to this implementation:

1. Learn from pixels
2. Use prioritised experience replay
3. Double DQN
4. Dual DQN