

# Report

## Learning Algorithm

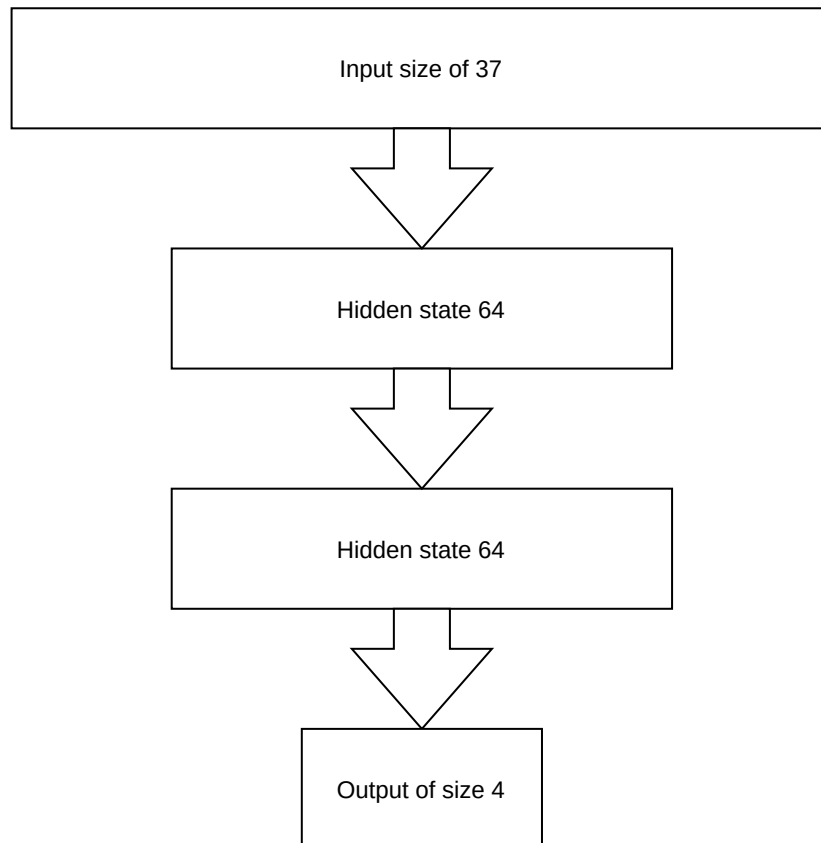
For this project , I have used and modified the code for Deep Q Network (DQN) from the lectures.

In DQN, we minimize the loss given by the minimizing the meas-squared error between the off-policy estimate of target state-value function and its approximate Q function

$$\mathcal{L}(\theta) = \mathbb{E}_{(s,a,r,s') \sim U(D)} \left[ \left( r + \gamma \max_{a'} Q(s', a'; \theta^-) - Q(s, a; \theta) \right)^2 \right]$$

$\theta^-$  and  $\theta$  are the target and approximate Q function parameters and the experience relay stores the samples  $(s, a, r, s')$  in a buffer, which are sampled using the uniform distribution  $U(D)$

The model used for the agent:

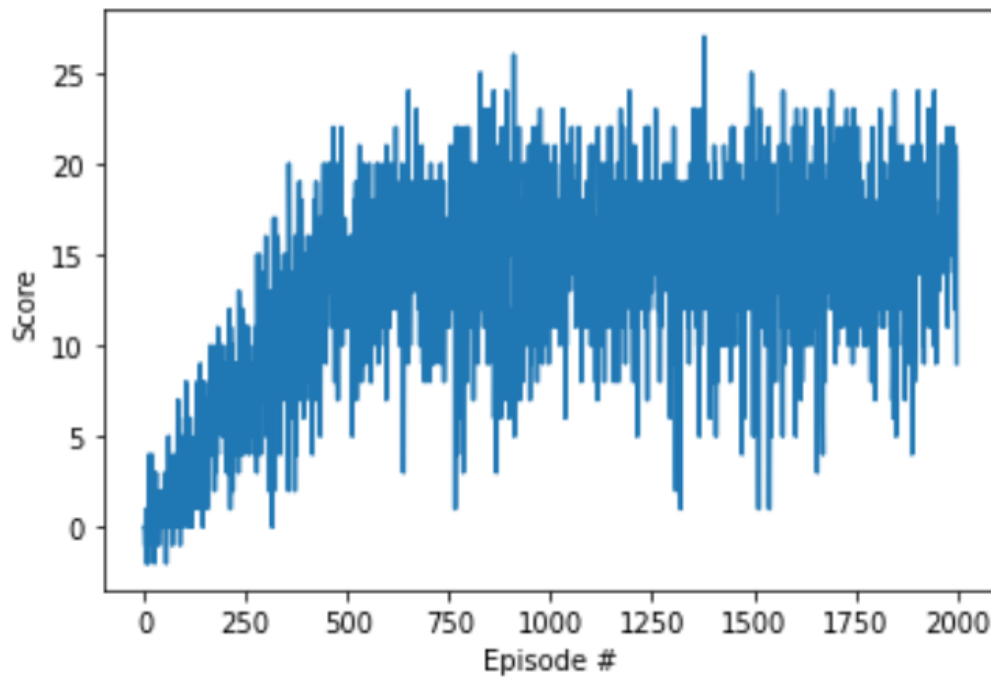


The hyperparameters used:

- Batch size of 64

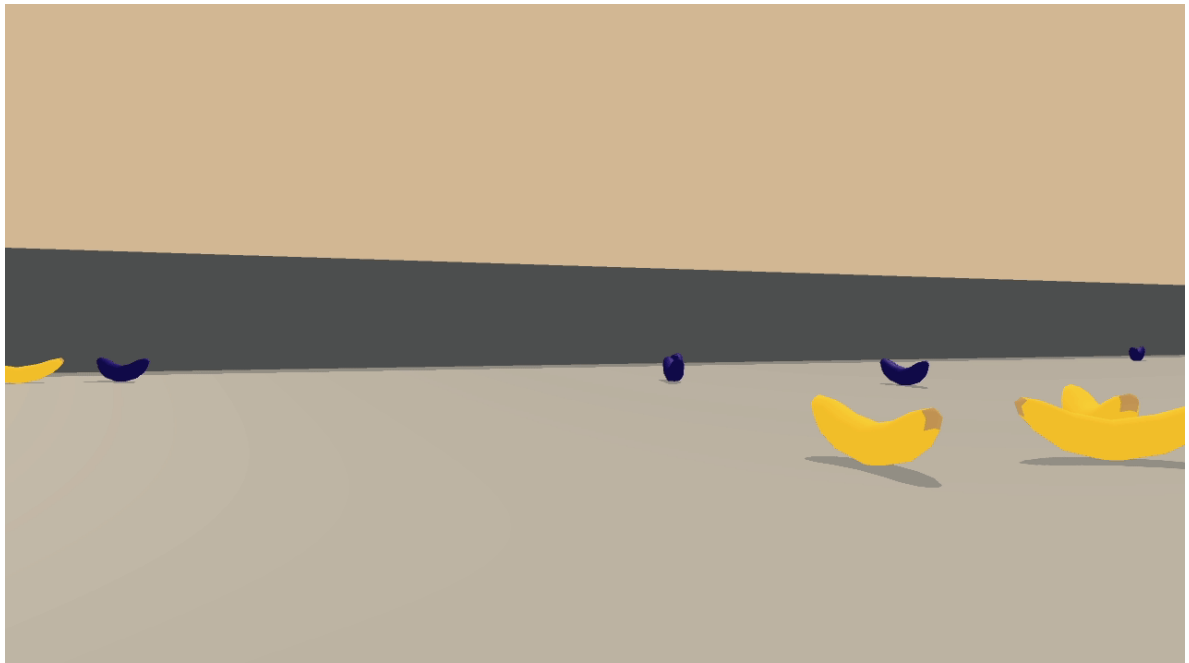
- Discount factor of 0.99
- Tau ( $\tau$ ) which updated the target Q function is 1e-3
- Learning rate 5e-4
- Update every 4 steps

## Plot of Rewards



The environment was solved by episode 500 as can be seen in `Navigation.ipynb`

## Trained agent behavior



## Ideas for Future Work

- Will extend the the DQN to dueling DQN and also solve the learning from pixels challenge.
- Invest in more time to explore different hyper-parameters and ablations studies