Report

This project DQN model, agent implement project.

State and Action

The state space has 37 dimensions and contains the agent's velocity, along with ray-based perception of objects around agent's forward direction. Given this information, the agent has to learn how to best select actions. Four discrete actions are available, corresponding to:

- **`0`** - move forward.
- **`1`** - move backward.
- **`2`** - turn left.
- **`3`** - turn right.

Learning algorithm

rewards + (gamma * Q_targets_next * (1 - dones))

this is important algorithm, and I am use two Qnetwork (target, main)

Hyper Parameters

n_episodes (int): maximum number of training episodes
max_t (int): maximum number of timesteps per episode
eps_start (float): starting value of epsilon, for epsilon-greedy action selection
eps_end (float): minimum value of epsilon
eps_decay (float): multiplicative factor (per episode) for decreasing epsilon
n_episodes=2000
max_t=1000
eps_start=1.0
eps_end=0.01
eps_decay=0.995

Agent Hyper Parameters

BUFFER_SIZE = int(1e5) # replay buffer size

BATCH_SIZE = 64 # minibatch size

GAMMA = 0.99 # discount factor

TAU = 1e-3 # for soft update of target parameters

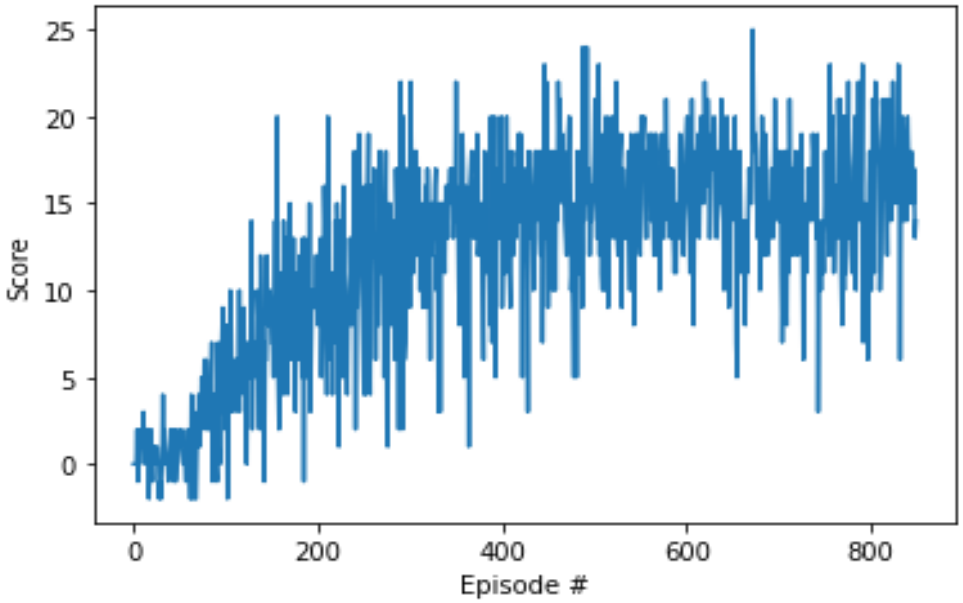
LR = 5e-4 # learning rate

UPDATE_EVERY = 4 # how often to update the network

Model architecture

Model use two 64 Fully Connect layer(hidden layer) and use two Relu Activation function for each hidden layer

Plot of Reward




```
Episode 100    Average Score: 1.35
Episode 200    Average Score: 7.36
Episode 300    Average Score: 11.06
Episode 400    Average Score: 13.69
Episode 500    Average Score: 14.86
Episode 600    Average Score: 15.38
Episode 700    Average Score: 15.66
Episode 800    Average Score: 14.66
Episode 850    Average Score: 16.01
Environment solved in 750 episodes!  Average Score: 16.01
```

Future Improving Idea

We can change and test some hyper parameter and activation function. And little test change value eps(epsilon) value in dqn function