### 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

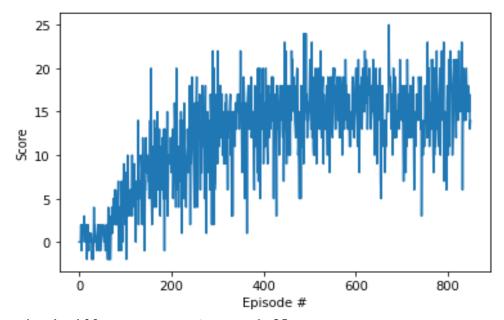
GAMMA = 0.99 # discount factor

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

LR = 5e-4 # learning rate

UPDATE\_EVERY = 4 # how often to update the network

### **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 Average Score: 15.38 Episode 600 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