

YAI 6<sup>th</sup>  
강화 스터디

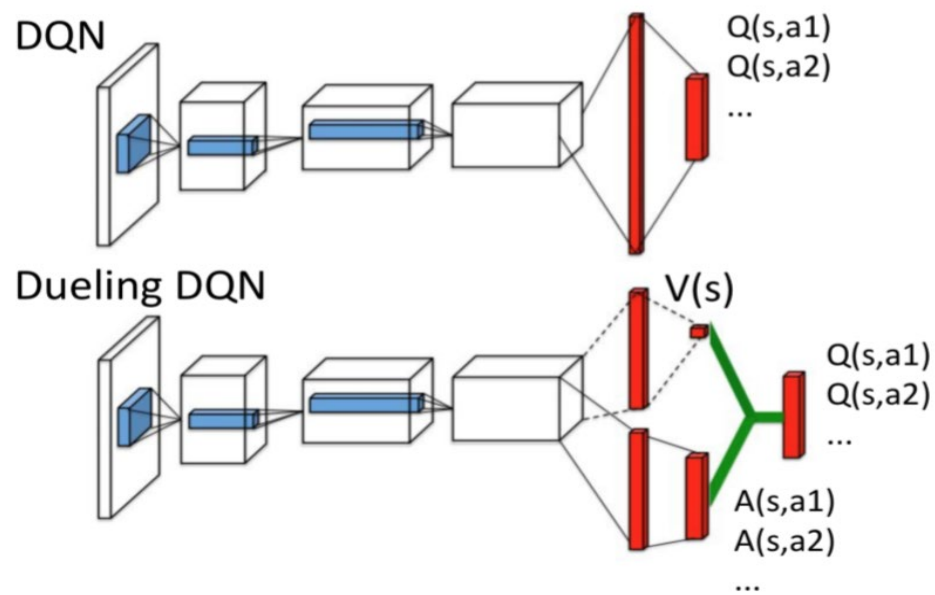
Turtlebot3 RL  
중간 점검

2020-11-17

바로 시작하겠습니다~

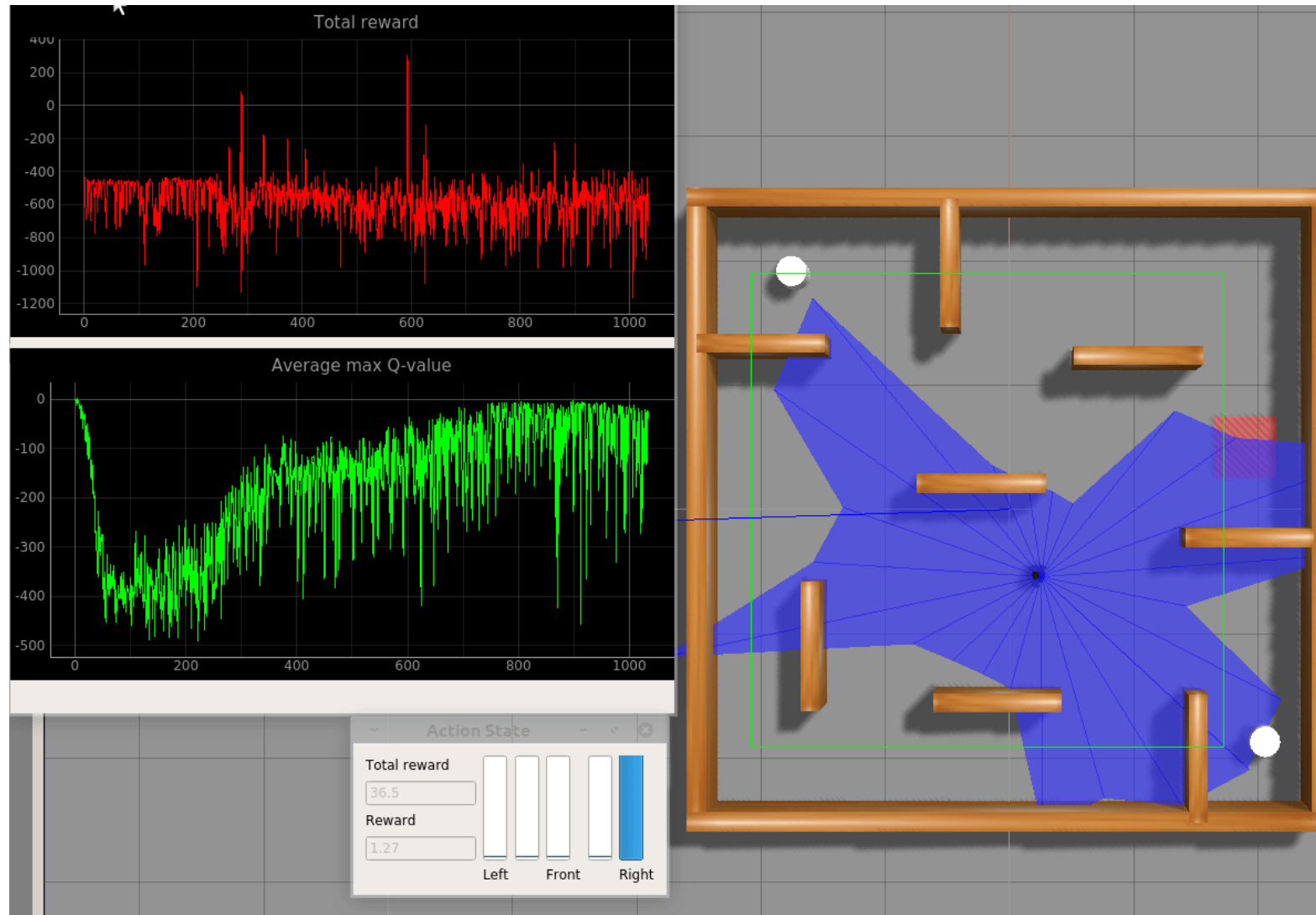
# 01 PER-D3QN

---

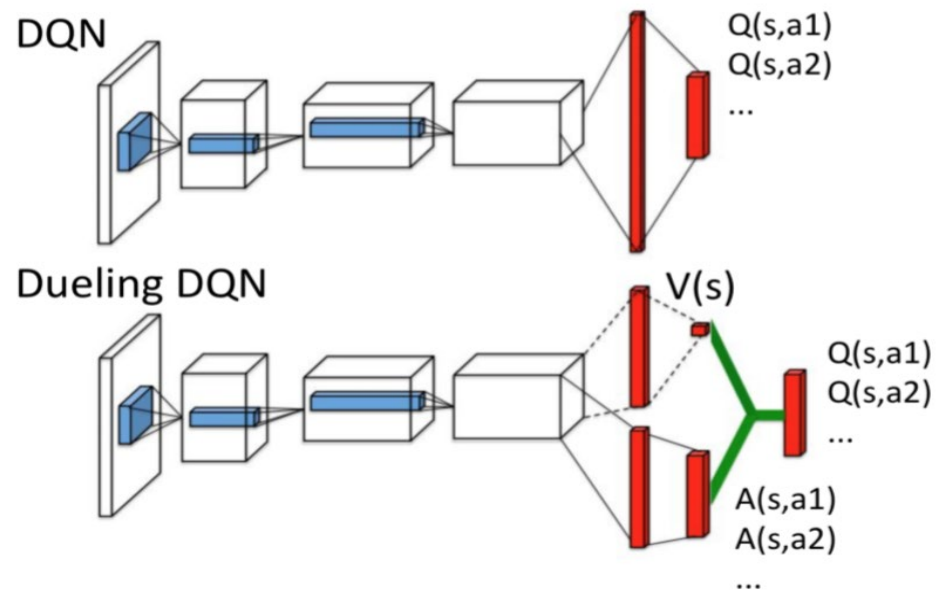


$$Q(s, a; \theta, \alpha, \beta) = V(s; \theta, \beta) + \left( A(s, a; \theta, \alpha) - \frac{1}{|\mathcal{A}|} \sum_{a'} A(s, a'; \theta, \alpha) \right). \quad (9)$$

# 01 PER-D3QN



# 01 PER-D3QN



+

PER

$$Q(s, a; \theta, \alpha, \beta) = V(s; \theta, \beta) + \left( A(s, a; \theta, \alpha) - \frac{1}{|\mathcal{A}|} \sum_{a'} A(s, a'; \theta, \alpha) \right). \quad (9)$$

# 01 PER-D3QN

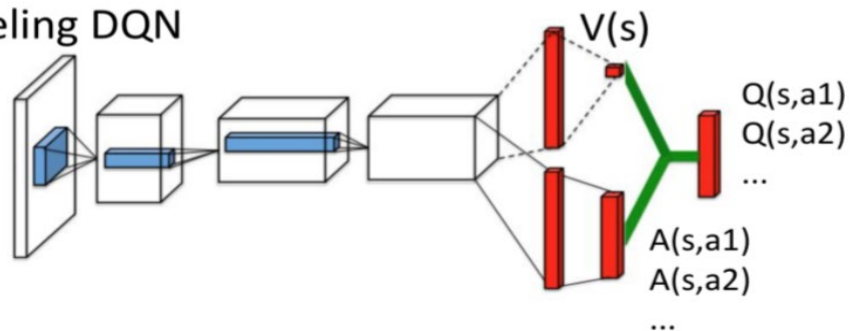
$$p_i = \left| r + \gamma \max_{a'} Q(s_{i+1}, a'; \mathbf{w}^-) - Q(s_i, a_i; \mathbf{w}) \right|$$

$$P(i) = \frac{p_i^\alpha}{\sum_k p_k^\alpha}$$

$$w_i = \left( \frac{1}{N} \cdot \frac{1}{P(i)} \right)$$

memory

Dueling DQN



1. Action

# 01 PER-D3QN

$$p_i = \left| r + \gamma \max_{a'} Q(s_{i+1}, a'; \mathbf{w}^-) - Q(s_i, a_i; \mathbf{w}) \right|$$

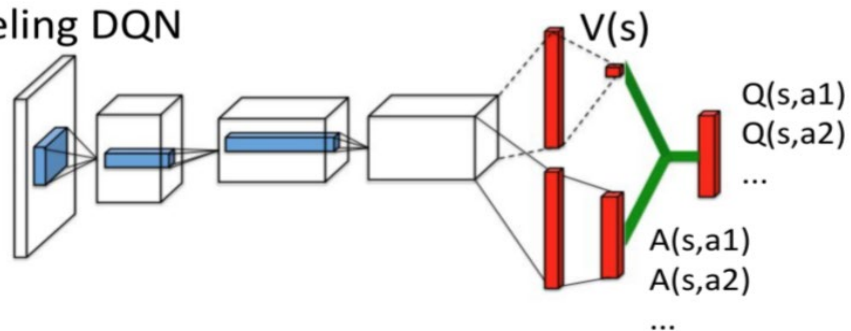
$$P(i) = \frac{p_i^\alpha}{\sum_k p_k^\alpha}$$

$$w_i = \left( \frac{1}{N} \cdot \frac{1}{P(i)} \right)$$

memory

2. Save memory

Dueling DQN



# 01 PER-D3QN

3. Pull out from memory with prioritization

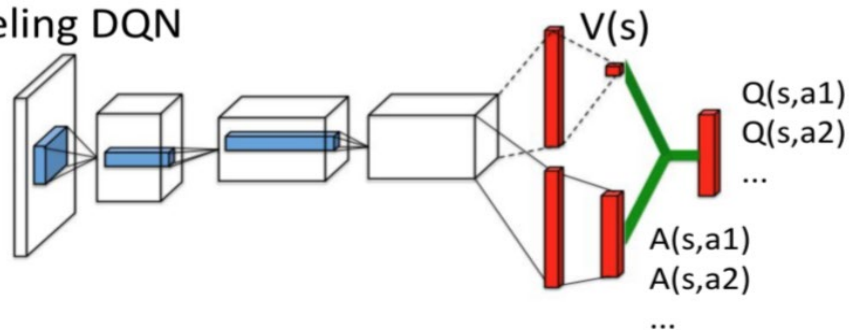
$$p_i = \left| r + \gamma \max_{a'} Q(s_{i+1}, a'; \mathbf{w}^-) - Q(s_i, a_i; \mathbf{w}) \right|$$

$$P(i) = \frac{p_i^\alpha}{\sum_k p_k^\alpha}$$

$$w_i = \left( \frac{1}{N} \cdot \frac{1}{P(i)} \right)$$

memory

Dueling DQN





# 01 PER-D3QN

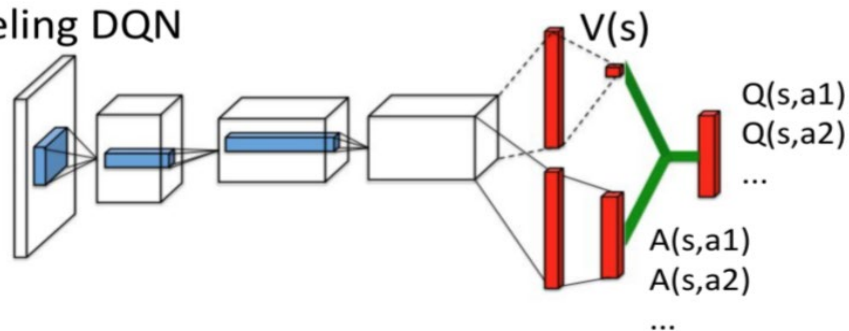
$$p_i = \left| r + \gamma \max_{a'} Q(s_{i+1}, a'; \mathbf{w}^-) - Q(s_i, a_i; \mathbf{w}) \right|$$

$$P(i) = \frac{p_i^\alpha}{\sum_k p_k^\alpha}$$

$$w_i = \left( \frac{1}{N} \cdot \frac{1}{P(i)} \right) \quad 4. \text{ Importance sampling}$$

memory

Dueling DQN



# 01 PER-D3QN

$$p_i = \left| r + \gamma \max_{a'} Q(s_{i+1}, a'; \mathbf{w}^-) - Q(s_i, a_i; \mathbf{w}) \right|$$

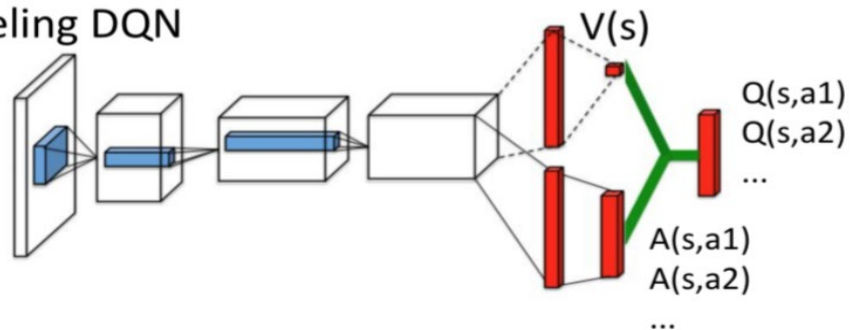
$$P(i) = \frac{p_i^\alpha}{\sum_k p_k^\alpha}$$

$$w_i = \left( \frac{1}{N} \cdot \frac{1}{P(i)} \right)$$

memory

5. Learn

Dueling DQN



# 01 PER-D3QN

$$p_i = \left| r + \gamma \max_{a'} Q(s_{i+1}, a'; \mathbf{w}^-) - Q(s_i, a_i; \mathbf{w}) \right|$$

$$P(i) = \frac{p_i^\alpha}{\sum_k p_k^\alpha}$$

$$w_i = \left( \frac{1}{N} \cdot \frac{1}{P(i)} \right)$$

---

```
def appendMemory(self, state, action, reward, next_state, done):
    #edit5
    td_error = reward + self.discount_factor * np.max(self.target_model.predict(next_state.reshape(1, len(next_state)))) - self.model.predict(state.reshape(1, len(state)))

    p = np.abs(td_error + 0.0000001) ** self.alpha
    #edit5

    self.memory.append((state, action, reward, next_state, done))

    #edit5
    self.priority.append(p)
    #edit5


def get_XY_batch(self):
    X_batch = np.zeros(batch_size)
    prob = self.get_prob() / sum

    sample_indices = np.argsort(-prob)[len(prob)-batch_size:]
    batch_size=batch_size

    importance = [0]*batch_size
    importance = np.array(importance[sample_indices])
    sample_idx = np.arange(len(sample_indices))

    return sample_idx, importance


i = importance[i] ** self.beta


self.model.fit(X_batch, Y_batch, batch_size=self.batch_size, epochs=1, verbose=0, sample_weight=importance batch)
```

# 01 PER-D3QN

---

```
def appendMemory(self, state, action, reward, next_state, done):  
    #(edit5)  
    td_error = reward + self.discount_factor * np.max(self.target_model.predict(next_state), axis=-1) - self.current_model.predict(state, action)  
  
    p = np.abs(td_error + 0.0000001) ** self.alpha  
    #(edit5)  
  
    self.memory.append((state, action, reward, next_state, done))  
  
    #(edit5)  
    self.priority.append(p)  
    #(edit5)
```

# 01 PER-D3QN

---

```
def get_PER_batch(self):  
    p_sum = np.sum(self.priority)  
    prob = self.priority / p_sum  
  
    sample_indices = choices(range(len(prob)), k=self.batch_size, weights=prob)  
  
    importance = (1/prob) * (1/len(self.priority))  
    importance = np.array(importance)[sample_indices]  
    samples = np.array(self.memory)[sample_indices]  
  
    return samples, importance
```

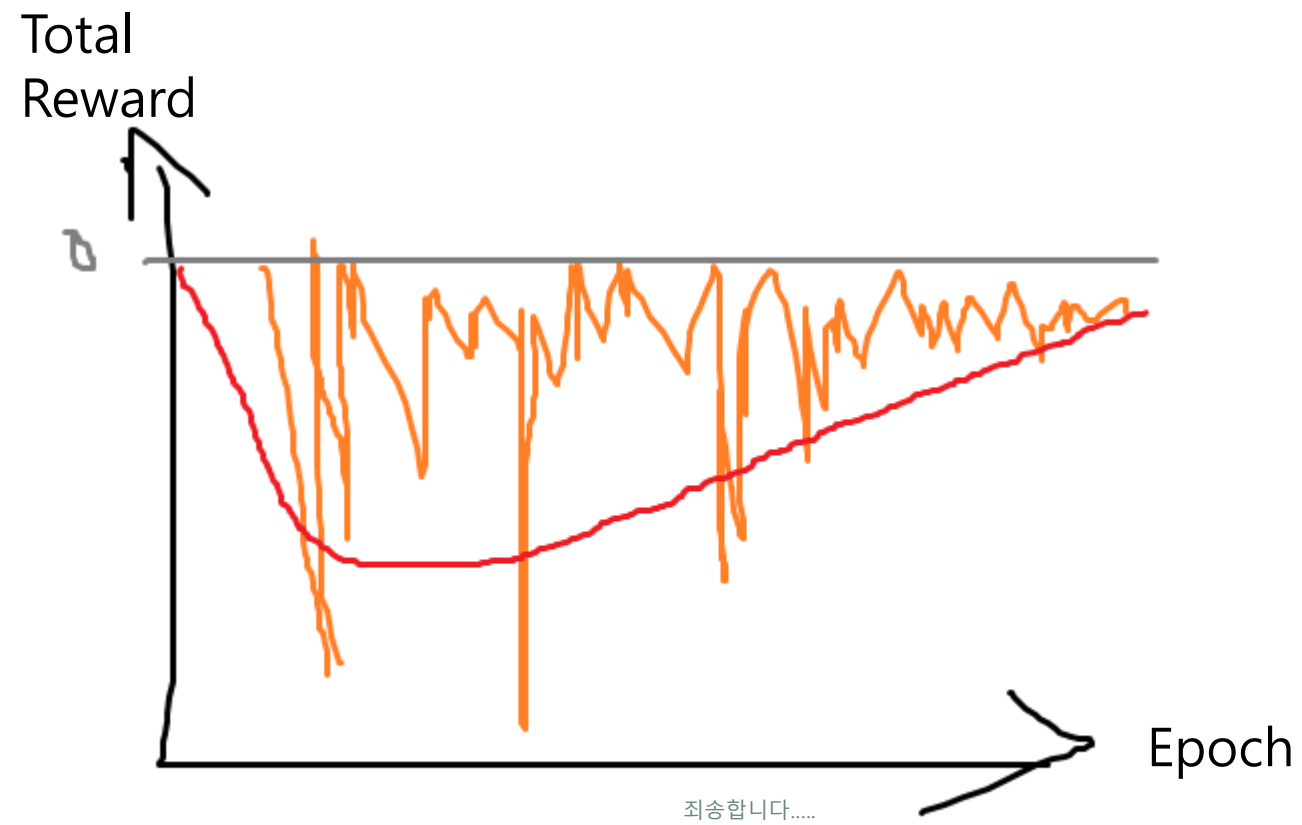
# 01 PER-D3QN

---

```
i = importance[i] ** self.beta
```

```
self.model.fit(X_batch, Y_batch, batch_size=self.batch_size, epochs=1, verbose=0, sample_weight=importance_batch)
```

# 01 PER-D3QN\_result



# 02 Agent & Environment

---

1. Rewards = {'right\_direction' = +1,  
'wrong\_direction' = -4  
'far\_away\_from\_goal' = -2\*\*distance\_ratio  
'collision' = -150  
'goal' = +200  
'near\_obstacle' = -5  
}

2. Actions = rotation right? rotate left?





# 02 Agent & Environment

---

1. Rewards = {'right\_direction' = +1,  
'wrong\_direction' = -4  
'far\_away\_from\_goal' = -2\*\*distance\_ratio  
'collision' = -150  
'goal' = +200  
'near\_obstacle' = -5  
'time\_since\_goal' -= 1  
}

2. Actions = rotation right? rotate left? **move faster!**

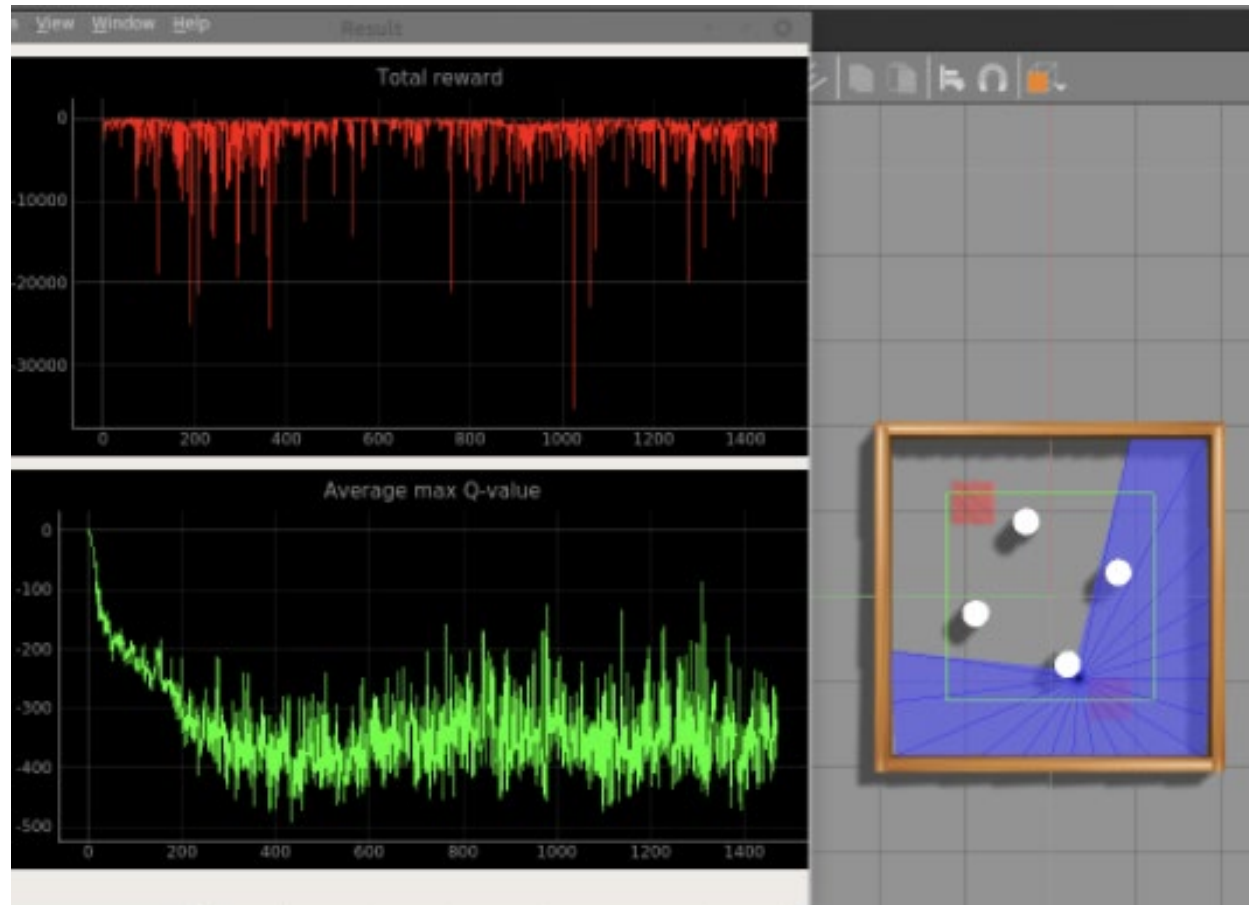
**3. Easier environment**

**4. Back to DDQN**



# 02 Agent & Environment

---



# 03 Building

---



jwill1994/ros2\_ml ☆

By [jwill1994](#) • Updated 6 days ago

Container



laphisboy/xx\_ros\_tb3 ☆

By [laphisboy](#) • Updated 13 days ago

ubuntu 16:04 with ROS kinetic and Turtlebot3 with referencing ROBOTIS e-Manual

Container

# 06 Q's?



Q's

# Fin

감사합니다~