My_Tennis

February 25, 2023

1 My Project

For this project, I will work with the Tennis environment.

1.1 Start with the library setup

```
[]: from unityagents import UnityEnvironment
   import numpy as np
   import torch
   from collections import deque
   from agent import Agent
   #visualizer
   import matplotlib.pyplot as plt
[]: env = UnityEnvironment(file_name="../Tennis_Linux/Tennis.x86_64")
  INFO:unityagents:
   'Academy' started successfully!
  Unity Academy name: Academy
          Number of Brains: 1
          Number of External Brains : 1
          Lesson number: 0
           Reset Parameters :
  Unity brain name: TennisBrain
           Number of Visual Observations (per agent): 0
           Vector Observation space type: continuous
           Vector Observation space size (per agent): 8
           Number of stacked Vector Observation: 3
          Vector Action space type: continuous
          Vector Action space size (per agent): 2
          Vector Action descriptions: ,
[]: # get the default brain
   brain_name = env.brain_names[0]
   brain = env.brains[brain_name]
```

```
[]: # reset the environment
   env_info = env.reset(train_mode=True)[brain_name]
   # number of agents
   num_agents = len(env_info.agents)
   print('Number of agents:', num_agents)
   # size of each action
   action_size = brain.vector_action_space_size
   print('Size of each action:', action_size)
   # examine the state space
   states = env_info.vector_observations
   state_size = states.shape[1]
   print('There are {} agents. Each observes a state with length: {}'.
    →format(states.shape[0], state_size))
   print('The state for the first agent looks like:', states[0])
   # number of episodes
   n_{episodes} = 10000
   # random seed
   SEED = 0
  Number of agents: 2
  Size of each action: 2
  There are 2 agents. Each observes a state with length: 24
  The state for the first agent looks like: [ 0.
                                                                        0.
  0.
               0.
                           0.
    0.
                 0.
                             0.
                                         0.
                                                      0.
                                                                  0.
                                         0.
                                                     -6.65278625 -1.5
     0.
                 0.
                             0.
    -0.
                 0.
                             6.83172083 6.
                                                     -0.
                                                                            1
                                                                  0.
| ]: agent = Agent(state_size=state_size,action_size=action_size, random_seed =__
    →SEED, n_agent=num_agents)
[]: def ddpg(n_episode=10000):
       print('total episode : {}'.format(n_episode) , end="\n")
       scores_window = deque(maxlen=50)
       scores_window_avg = []
       score = []
       scores = []
       reference_score = 0.5
       # stable reaching to target score
       end_condi = 0
       for i_episode in range(1, n_episode+1):
           env_info = env.reset(train_mode=True)[brain_name]
```

```
states = env_info.vector_observations
       score = np.zeros(num_agents)
       agent.reset()
       while(True):
           # Take random action
           actions = agent.act(states)
           # Do env action
           env_info = env.step(actions)[brain_name]
           # Observe states
           next_states = env_info.vector_observations
           # Obtain reward
           reward = env info.rewards
           # Get done flag
           dones = env_info.local_done
           # Do agent stop
           agent.step(states,actions,reward,next_states, dones)
           # update current state
           states = next_states
           # add reward on score
           # score += np.array(reward)
           score += np.max(reward)
           # Done if get done flag
           if np.any(dones):
               break
       scores.append(np.mean(score))
       # average filter for 50 ep
       scores_window.append(np.max(score))
       scores.append(np.max(score))
       scores_window_avg.append(np.mean(scores_window))
       # print scores every 100*N th episode
       if i_episode % 50 == 0:
           print('\rEpisode: \t{} \tScore: \t{:.2f} \tAverage Score: \t{:.2f}'.

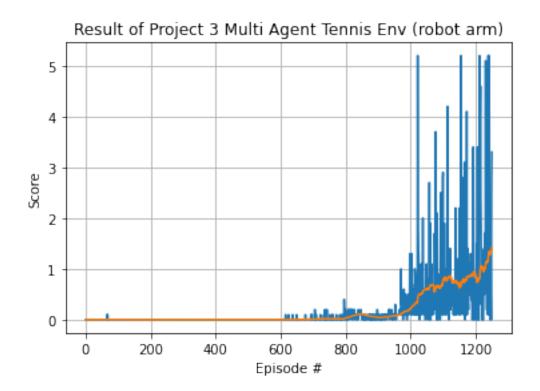
→format(i_episode, np.mean(score), scores_window[-1]), end="")
           if scores_window_avg[-1] > 0.5:
               end condi += 1
       # if the agent is able to receive an average reward(over "100_{\sqcup}
\rightarrow consevcutive episodes") of at least +0.5.
       if scores_window[-1] > 0.5:
           torch.save(agent.actor_local.state_dict(), 'checkpoint_action.pth')
           torch.save(agent.critic_local.state_dict(), 'checkpoint_critic.pth')
           print("score is over 0.5 (sucess standard)")
       if end_condi == 5 and i_episode > 500:
           # end when the agent reaches to target score
```

```
return scores, scores_window_avg
       return scores, scores_window_avg
[]: | scores_window, scores_window_avg = ddpg(n_episode=n_episodes)
  total episode: 10000
  /home/astra/anaconda3/envs/drlnd/lib/python3.6/site-
  packages/torch/nn/functional.py:1340: UserWarning: nn.functional.tanh is
  deprecated. Use torch.tanh instead.
    warnings.warn("nn.functional.tanh is deprecated. Use torch.tanh instead.")
  Episode:
                   950
                           Score: 0.00
                                           Average Score: 0.00score is over 0.5
   (sucess standard)
   score is over 0.5 (sucess standard)
  score is over 0.5 (sucess standard)
  score is over 0.5 (sucess standard)
  score is over 0.5 (sucess standard)
   score is over 0.5 (sucess standard)
  score is over 0.5 (sucess standard)
                   1000
                           Score: 0.10
  Episode:
                                           Average Score: 0.10score is over 0.5
   (sucess standard)
  score is over 0.5 (sucess standard)
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  score is over 0.5 (sucess standard)
                   1050
  Episode:
                           Score: 0.30
                                           Average Score: 0.30score is over 0.5
   (sucess standard)
   score is over 0.5 (sucess standard)
```

```
score is over 0.5 (sucess standard)
Episode:
                1100
                        Score: 2.90
                                        Average Score: 2.90score is over 0.5
(sucess standard)
score is over 0.5 (sucess standard)
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score is over 0.5 (sucess standard)
Episode:
                1150
                        Score: 1.70
                                        Average Score: 1.70score is over 0.5
(sucess standard)
```

```
score is over 0.5 (sucess standard)
                1200
Episode:
                        Score: 0.80
                                        Average Score: 0.80score is over 0.5
(sucess standard)
score is over 0.5 (sucess standard)
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```
score is over 0.5 (sucess standard)
                   1250
                           Score: 3.30
  Episode:
                                           Average Score: 3.30score is over 0.5
   (sucess standard)
env.close()
[]: temp = np.zeros(1250)
   cnt = 0
   print(temp[0])
   print(scores_window[0])
   for i in range(1250):
       temp[i] = scores_window[i*2]
   scores_window = temp
  0.0
  0.0
[]: # plot the scores
   fig = plt.figure()
   ax = fig.add_subplot(111)
   plt.title("Result of Project 3 Multi Agent Tennis Env (robot arm)")
   plt.plot(np.arange(len(scores_window)), scores_window, label = "scores")
   plt.plot(np.arange(len(scores_window_avg)), scores_window_avg, label = "average_
    →of scores")
   plt.grid(True)
   plt.ylabel('Score')
   plt.xlabel('Episode #')
   plt.show()
```



[]: