2.2C

March 28, 2021

[6]: # install required system dependencies

```
!apt-get install -y xvfb x11-utils
!apt-get install x11-utils > /dev/null 2>&1
!pip install PyOpenGL==3.1.* \
            PyOpenGL-accelerate==3.1.* \
             gym[box2d] == 0.17.*
 !pip install pyglet
!pip install ffmpeg
 ! pip install pyvirtualdisplay
 !pip install Image
 !pip install gym-maze-trustycoder83
Reading package lists... Done
Building dependency tree
Reading state information... Done
x11-utils is already the newest version (7.7+3build1).
xvfb is already the newest version (2:1.19.6-1ubuntu4.8).
0 upgraded, 0 newly installed, 0 to remove and 30 not upgraded.
Requirement already satisfied: PyOpenGL==3.1.* in /usr/local/lib/python3.7/dist-
packages (3.1.5)
Requirement already satisfied: PyOpenGL-accelerate==3.1.* in
/usr/local/lib/python3.7/dist-packages (3.1.5)
Requirement already satisfied: gym[box2d] == 0.17.* in
/usr/local/lib/python3.7/dist-packages (0.17.2)
Requirement already satisfied: pyglet<=1.5.0,>=1.4.0 in
/usr/local/lib/python3.7/dist-packages (from gym[box2d] == 0.17.*) (1.5.0)
Requirement already satisfied: numpy>=1.10.4 in /usr/local/lib/python3.7/dist-
packages (from gym[box2d] == 0.17.*) (1.18.5)
Requirement already satisfied: cloudpickle<1.4.0,>=1.2.0 in
/usr/local/lib/python3.7/dist-packages (from gym[box2d] == 0.17.*) (1.3.0)
Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages
(from gym[box2d] == 0.17.*) (1.4.1)
Requirement already satisfied: box2d-py~=2.3.5; extra == "box2d" in
/usr/local/lib/python3.7/dist-packages (from gym[box2d] == 0.17.*) (2.3.8)
Requirement already satisfied: future in /usr/local/lib/python3.7/dist-packages
(from pyglet <= 1.5.0, >= 1.4.0 -> gym[box2d] == 0.17.*) (0.16.0)
Requirement already satisfied: pyglet in /usr/local/lib/python3.7/dist-packages
(1.5.0)
```

```
Requirement already satisfied: future in /usr/local/lib/python3.7/dist-packages
     (from pyglet) (0.16.0)
     Requirement already satisfied: ffmpeg in /usr/local/lib/python3.7/dist-packages
     Requirement already satisfied: pyvirtualdisplay in
     /usr/local/lib/python3.7/dist-packages (2.1)
     Requirement already satisfied: EasyProcess in /usr/local/lib/python3.7/dist-
     packages (from pyvirtualdisplay) (0.3)
     Requirement already satisfied: Image in /usr/local/lib/python3.7/dist-packages
     (1.5.33)
     Requirement already satisfied: django in /usr/local/lib/python3.7/dist-packages
     (from Image) (3.1.7)
     Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages
     (from Image) (1.15.0)
     Requirement already satisfied: pillow in /usr/local/lib/python3.7/dist-packages
     (from Image) (7.0.0)
     Requirement already satisfied: sqlparse>=0.2.2 in /usr/local/lib/python3.7/dist-
     packages (from django->Image) (0.4.1)
     Requirement already satisfied: asgiref<4,>=3.2.10 in
     /usr/local/lib/python3.7/dist-packages (from django->Image) (3.3.1)
     Requirement already satisfied: pytz in /usr/local/lib/python3.7/dist-packages
     (from django->Image) (2018.9)
     Requirement already satisfied: gym-maze-trustycoder83 in
     /usr/local/lib/python3.7/dist-packages (0.0.4)
     Requirement already satisfied: pygame==1.9.6 in /usr/local/lib/python3.7/dist-
     packages (from gym-maze-trustycoder83) (1.9.6)
     Requirement already satisfied: numpy==1.18.5 in /usr/local/lib/python3.7/dist-
     packages (from gym-maze-trustycoder83) (1.18.5)
     Requirement already satisfied: gym==0.17.2 in /usr/local/lib/python3.7/dist-
     packages (from gym-maze-trustycoder83) (0.17.2)
     Requirement already satisfied: cloudpickle<1.4.0,>=1.2.0 in
     /usr/local/lib/python3.7/dist-packages (from gym==0.17.2->gym-maze-
     trustycoder83) (1.3.0)
     Requirement already satisfied: pyglet<=1.5.0,>=1.4.0 in
     /usr/local/lib/python3.7/dist-packages (from gym==0.17.2->gym-maze-
     trustycoder83) (1.5.0)
     Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages
     (from gym==0.17.2->gym-maze-trustycoder83) (1.4.1)
     Requirement already satisfied: future in /usr/local/lib/python3.7/dist-packages
     (from pyglet<=1.5.0,>=1.4.0->gym==0.17.2->gym-maze-trustycoder83) (0.16.0)
     We now proceed to initialise the monitor wrapper for Gym so we can visualise the maze and the
     agent on a video
[12]: import sys
      # import pygame
      import numpy as np
      # import math
```

```
# import base64
# import io
# import IPython
import gym
import gym_maze

# from gym.wrappers import Monitor
# from IPython import display
from pyvirtualdisplay import Display
from gym.wrappers.monitoring import video_recorder

d = Display()
d.start()

# Recording filename
video_name = "./vid/Practical_2.mp4"

# Setup the environment for the maze
env = gym.make("maze-sample-10x10-v0")
```

We now proceed to perform a simple Q-tracking algorithm. The method here is quite sub-optimal since it employs a random choice to pick between exploration and exploitation. It does illustrate, however, how Q-value tracking can be done using the maze.

```
[13]: def get_Q_table_ind (size, x, y):
       x_max, y_max = size
        return y_max * x + y
      def policy (Q_table, current_state, epsilon):
        current_state_in_Q = get_Q_table_ind(SIZE, current_state[0], current_state[1])
        if np.random.uniform(0,1) < epsilon:</pre>
            return env.action_space.sample()
        else:
            return int(np.argmax(Q_table[current_state_in_Q]))
      def Q update (Q table, current state, next state, action, reward,
       →learning_rate):
       current_state_in_Q = get_Q_table_ind(SIZE, current_state[0], current_state[1])
       next_state_in_Q = get_Q_table_ind(SIZE, next_state[0], next_state[1])
        Q_table[current_state_in_Q, action] = (1-learning_rate)_
       →*Q_table[current_state_in_Q, action] +learning_rate*(reward +
       →max(Q table[next state in Q,:]))
      def Q_update_new (Q_table, current_state, next_state, action, reward, alpha,_
       ⇒gamma):
        current_state_in_Q = get_Q_table_ind(SIZE, current_state[0], current_state[1])
        next_state_in_Q = get_Q_table_ind(SIZE, next_state[0], next_state[1])
```

```
Q_table[current_state_in_Q, action] = Q_table[current_state_in_Q, action] +

→alpha * (reward - gamma * Q_table[current_state_in_Q, action])
```

```
[9]: def run(env, epsilon, gamma):
      n_actions = env.action_space.n
       current_state = env.reset()
       Q_table = np.zeros((XSIZE * YSIZE, n_actions))
       N_count = np.zeros(XSIZE * YSIZE)
       rewards = []
       for e in range(50):
           total_reward = 0
           # We are not done yet
           done = False
           for i in range(MAX_ITERATION):
               env.unwrapped.render()
               vid.capture frame()
               action = policy(Q_table, current_state, epsilon)
               next_state, reward, done, _ = env.step(action)
               \# Q\_update(Q\_table, current\_state, next\_state, action, reward, 
     \rightarrow learning_rate)
               current_state_in_Q = get_Q_table_ind(SIZE, current_state[0],__
     N_count[current_state_in_Q] += 1
               Q_update_new (Q_table, current_state, next_state, action, reward, 1.0u
     →/N_count[current_state_in_Q], gamma)
               total_reward = total_reward + reward
               # If the episode is finished, we leave the for loop
               if done:
                   break
               current_state = next_state
           #Show reward
           print("Total episode reward:", total_reward)
           rewards.append(total_reward)
           current_state = env.reset()
       return rewards
```

```
[14]: MAX_ITERATION = 500
XSIZE = 10
YSIZE = 10
SIZE = [XSIZE, YSIZE]
# Setup the video
```

```
vid = None
vid = video_recorder.VideoRecorder(env, video_name)

epsilon = 0
gamma = 0.5

rewards_0_05 = run(env, epsilon, gamma)

print("Video successfuly saved.")
vid.close()
vid.enabled = False
```

Total episode reward: -0.5000000000000003 Total episode reward: 0.711999999999997 Total episode reward: 0.69699999999998 Total episode reward: 0.65899999999998 Total episode reward: 0.647999999999997 Total episode reward: 0.64499999999998 Total episode reward: 0.615999999999997 Total episode reward: 0.535999999999996 Total episode reward: -0.5000000000000003 Total episode reward: 0.55299999999997 Total episode reward: -0.5000000000000003 Total episode reward: -0.50000000000000003 Total episode reward: 0.68899999999998 Total episode reward: -0.5000000000000003 Total episode reward: 0.66799999999997 Total episode reward: -0.5000000000000003 Total episode reward: -0.5000000000000003 Total episode reward: -0.50000000000000003 Total episode reward: 0.75199999999998 Total episode reward: -0.5000000000000003 Total episode reward: -0.5000000000000003

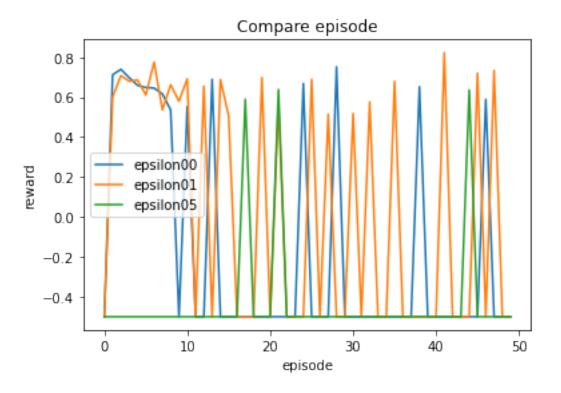
```
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: 0.65199999999997
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: -0.50000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: 0.588999999999997
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Total episode reward: -0.5000000000000003
Video successfuly saved.
```

```
[17]: epsilon = 0.1
gamma = 0.5
rewards_01_05 = run(env, epsilon, gamma)
```

Total episode reward: -0.5000000000000003 Total episode reward: -0.50000000000000003 Total episode reward: 0.62099999999998 Total episode reward: -0.5000000000000003 Total episode reward: 0.78499999999998 Total episode reward: -0.5000000000000003 Total episode reward: 0.681999999999997 Total episode reward: 0.61099999999998 Total episode reward: 0.555999999999996 Total episode reward: 0.57999999999996 Total episode reward: 0.569999999999996 Total episode reward: 0.707999999999997 Total episode reward: -0.5000000000000003 Total episode reward: 0.585999999999996 Total episode reward: 0.531999999999996 Total episode reward: 0.573999999999996 Total episode reward: -0.50000000000000003 Total episode reward: 0.557999999999996 Total episode reward: -0.5000000000000003 Total episode reward: 0.70099999999998 Total episode reward: -0.5000000000000003 Total episode reward: -0.5000000000000003 Total episode reward: 0.611999999999997 Total episode reward: -0.5000000000000003 Total episode reward: -0.50000000000000003 Total episode reward: -0.5000000000000003 Total episode reward: 0.665999999999997

```
Total episode reward: -0.50000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: 0.75299999999998
    Total episode reward: -0.50000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.50000000000000003
    Total episode reward: 0.645999999999997
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: 0.625999999999997
    Total episode reward: -0.50000000000000003
[]: epsilon = 0.5
     gamma = 0.5
    rewards_05_05 = run(env, epsilon, gamma)
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.50000000000000003
    Total episode reward: -0.50000000000000003
    Total episode reward: -0.50000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
```

```
Total episode reward: -0.5000000000000003
    Total episode reward: 0.637999999999997
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.50000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.50000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.50000000000000003
    Total episode reward: 0.635999999999997
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
    Total episode reward: -0.5000000000000003
[]: import matplotlib.pyplot as plt
    plt.plot(rewards_0_05, label = "epsilon00")
    plt.plot(rewards_01_05, label = "epsilon01")
    plt.plot(rewards_05_05, label = "epsilon05")
    plt.xlabel('episode')
    plt.ylabel('reward')
    plt.title('Compare episode')
    plt.legend()
    plt.show()
```



```
[15]: epsilon = 0.1
gamma = 0.0
rewards_01_00 = run(env, epsilon, gamma)
```

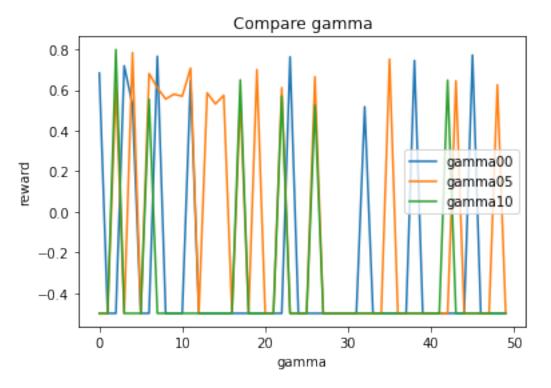
Total episode reward: 0.68399999999997 Total episode reward: -0.5000000000000003 Total episode reward: -0.50000000000000003 Total episode reward: 0.71999999999998 Total episode reward: 0.52799999999996 Total episode reward: -0.50000000000000003 Total episode reward: -0.50000000000000003 Total episode reward: 0.76699999999998 Total episode reward: -0.5000000000000003 Total episode reward: -0.5000000000000003 Total episode reward: -0.5000000000000003 Total episode reward: 0.64799999999997 Total episode reward: -0.5000000000000003 Total episode reward: -0.50000000000000003 Total episode reward: -0.5000000000000003

```
Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: 0.76399999999998
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: 0.517999999999996
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: 0.74599999999998
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: 0.77299999999998
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.50000000000000003
[16]: epsilon = 0.1
      gamma = 1.0
     rewards_01_10 = run(env, epsilon, gamma)
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
```

```
Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: 0.64999999999997
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: 0.570999999999997
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.50000000000000003
     Total episode reward: 0.525999999999996
     Total episode reward: -0.5000000000000003
     Total episode reward: 0.64899999999998
     Total episode reward: -0.50000000000000003
     Total episode reward: -0.5000000000000003
     Total episode reward: -0.5000000000000003
[18]: import matplotlib.pyplot as plt
     plt.plot(rewards_01_00, label = "gamma00")
     plt.plot(rewards_01_05, label = "gamma05")
     plt.plot(rewards_01_10, label = "gamma10")
     plt.xlabel('gamma')
     plt.ylabel('reward')
     plt.title('Compare gamma')
```

Total episode reward: -0.5000000000000003

```
plt.legend()
plt.show()
```



<IPython.core.display.HTML object>