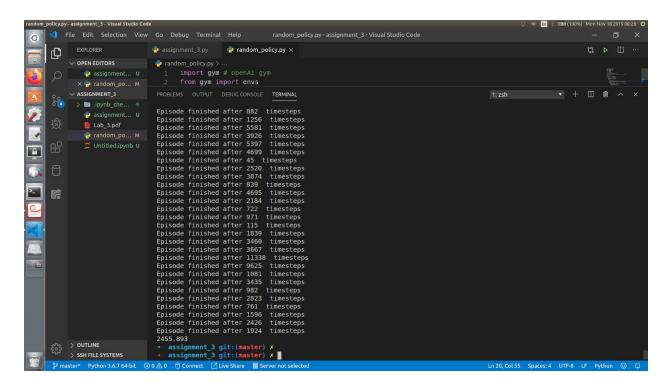
CS529– Applied Artificial Intelligence Lab Assignment - 3

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Entry Number: 2017csb1107

Question 1:

Chose an action randomly and when the reward was **20**, then reseted the environment. The average number of timesteps for 1000 episodes were **2455.893**(Around 2500).



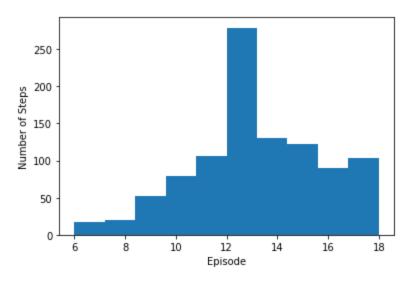
Question 2:

| Discount Factor | Policy Iteration | Value Iteration |
|-----------------|------------------|-----------------|
| 0.99 | 12 * 1100 | 725 |
| 0.95 | 12 * 226 | 146 |
| 0.9 | 12 * 110 | 74 |

| 0.8 | 12 * 53 | 38 |
|-----|---------|----|
| 0.4 | 12 * 14 | 14 |

- For discount factor 0.4, I didn't get the same policy
- Otherwise, Optimal policy is same for Value Iteration and Policy Iteration
- Compared to each other, value-iteration is computationally efficient even though it takes more number of iterations to converge, each iteration is less computationally expensive than policy-iteration.

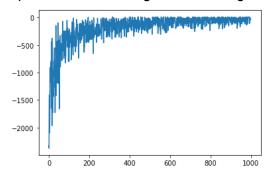
Distribution of Number of steps taken for 1000 episodes is as follows:



Average number of steps is 13.

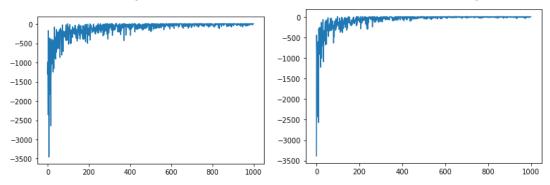
Question 3:

Alpha = 0.05, Convergence around greater than 1000 eps



Alpha = 0.1, Convergence around 600 eps

Alpha = 0.2, Convergence around 500 eps

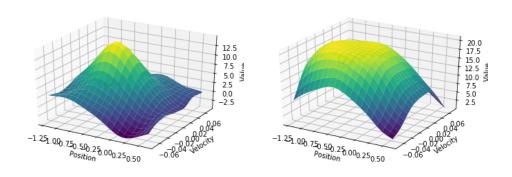


- On Y-axis, I kept total reward of the episode.
- On convergence, the change is total reward becomes constant. It becomes parallel to x-axis.

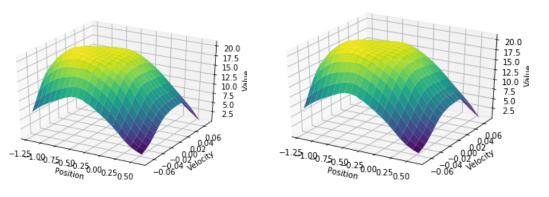
Question 4:

The results were matching the graphs of book. Below are the plots of Mountain Car problem.

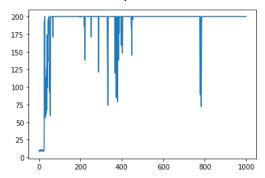
Episode 0 Episode 100



Episode 200 Episode 300



For Cart Pole, the plot of Total reward with num of episodes is :



References:

For Mountain Car:

https://github.com/SamKirkiles/mountain-car-SARSA-AC/blob/master/mountain_car.py

For Cart Pole: https://github.com/ceteke/RL/blob/master/Approximation/Linear%20Sarsa.ipynb