

Assignment 2 Report

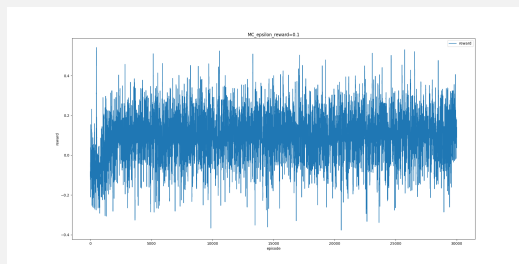
Q1. Discuss and plot learning curves under ϵ values of (0.1, 0.2, 0.3, 0.4) on MC, SARSA, and Q-Learning

1. Below are the learning curve figures of the three methods, we can observe some properties:

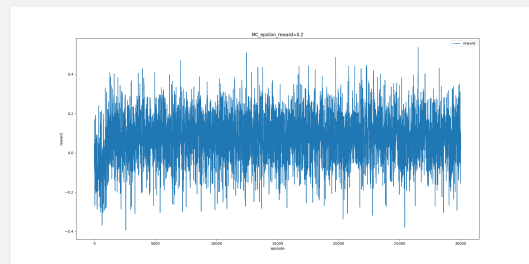
- (a) They all go from low to high.(All three methods.)
- (b) SARSA and Q-learning performs better since it earns more reward rather than gaining oscillating rewards like Monte Carlo methods.
- (c) I didn't discover the difference between each ϵ settings, but I guess when $\epsilon = 0.3$ it performs better(I have this guess from SARSA learning curve)
- (d) In this question, I fixed the max_episodes of MC & SARSA to 30000, Q-learning to 5000
Also in my implementation, I altered the epsilon greedy policy of the first n episodes into random policy
- (e) Implementation of the Average non-discounted Episodic Reward looks like the following:

```
reward_record, loss_record = run_Q_Learning(grid_world, Q_episode  
, epsilon=0.2)  
avg_reward = [sum(row)/len(row) for row in reward_record]  
avg_r = [np.mean(avg_reward[i:i+10]) for i in range(0, len(avg_reward))]
```

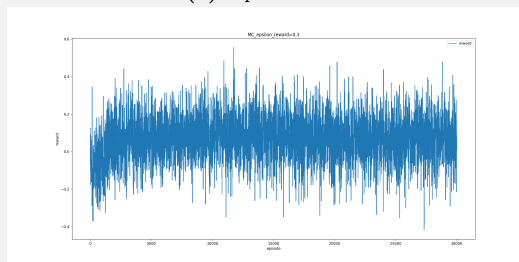
2. Monte Carlo



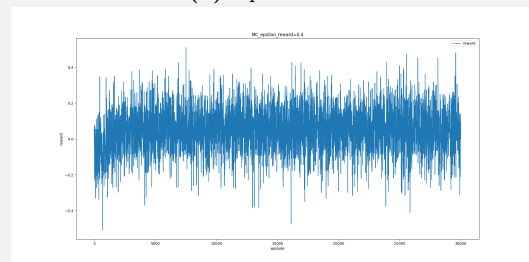
(a). epsilon=0.1



(b). epsilon=0.2



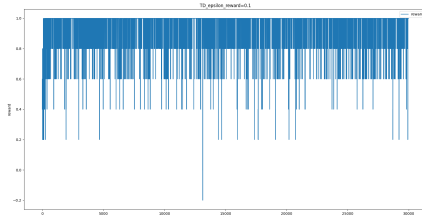
(c). epsilon=0.3



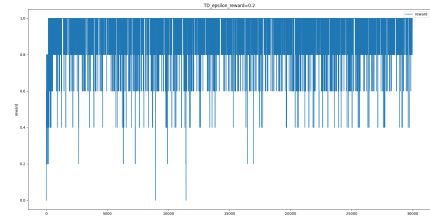
(d). epsilon=0.4

Figure 1: Monte Carlo Learning Curve

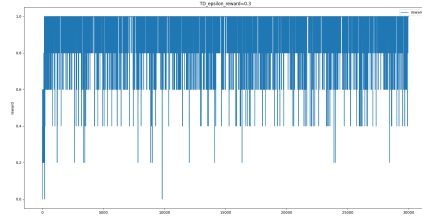
3. SARSA



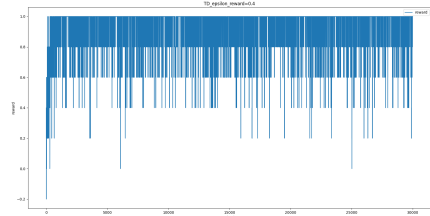
(a). epsilon=0.1



(b). epsilon=0.2



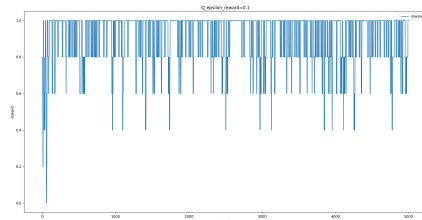
(c). epsilon=0.3



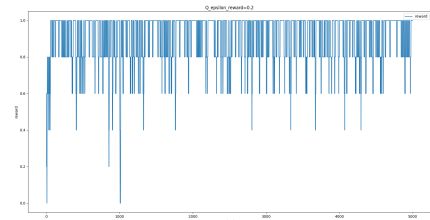
(d). epsilon=0.4

Figure 2: SARSA Learning Curve

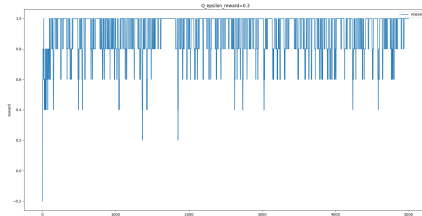
4. Q_learning



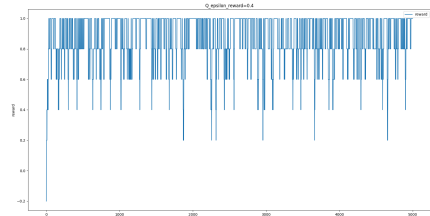
(a). epsilon=0.1



(b). epsilon=0.2



(c). epsilon=0.3



(d). epsilon=0.4

Figure 3: Q Learning Curve

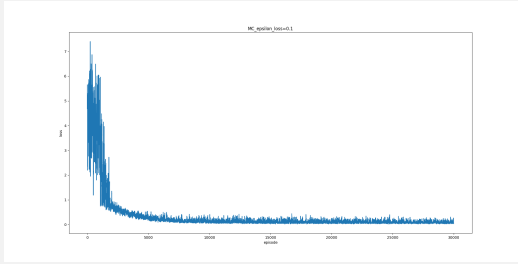
Q2. Discuss and plot loss curves under ϵ values of (0.1,0.2,0.3,0.4) on MC, SARSA, and Q-Learning

1. Below are the loss curve figures of the three methods, we can observe some properties:

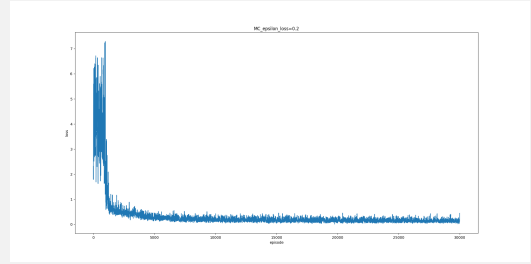
- (a) They all go from high to low.(All three methods.)
- (b) SARSA and Q_learning performs worse since it has bigger bias because of the estimations
- (c) I didn't discover the difference between each ϵ settings, but I guess when $\epsilon = 0.1, 0.4$ it performs better(I have this guess from all three curves)

(d) In this question, I fixed the max_episodes of Monte Carlo & SARSA to 30000, Q_learning to 5000.

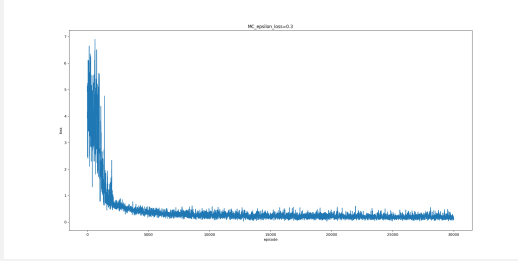
2. Monte Carlo



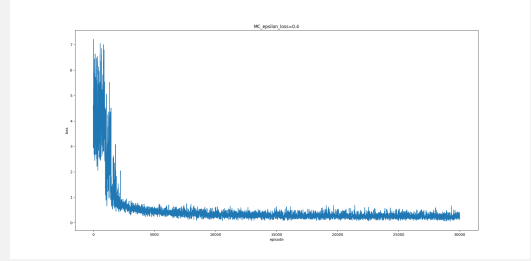
(a). epsilon=0.1



(b). epsilon=0.2



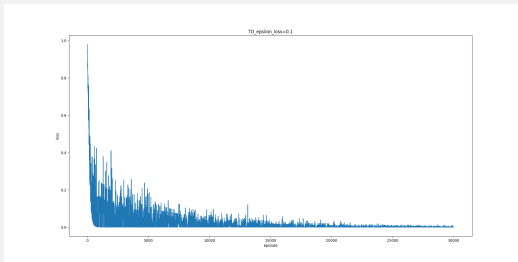
(c). epsilon=0.3



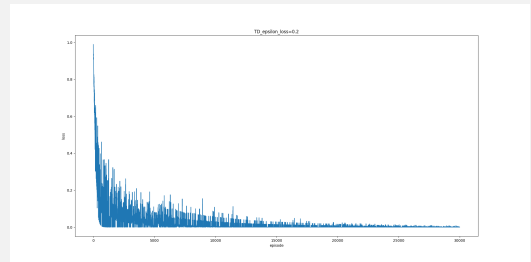
(d). epsilon=0.4

Figure 4: Monte Carlo Loss Curve

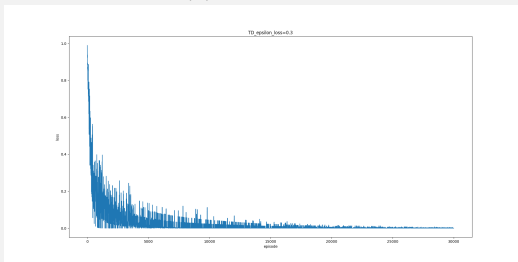
3. SARSA



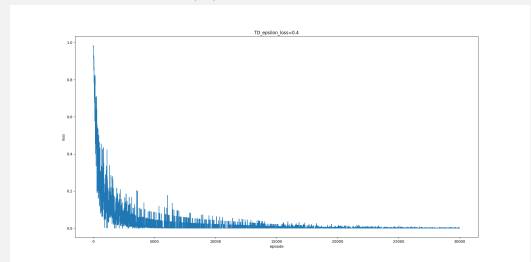
(a). epsilon=0.1



(b). epsilon=0.2



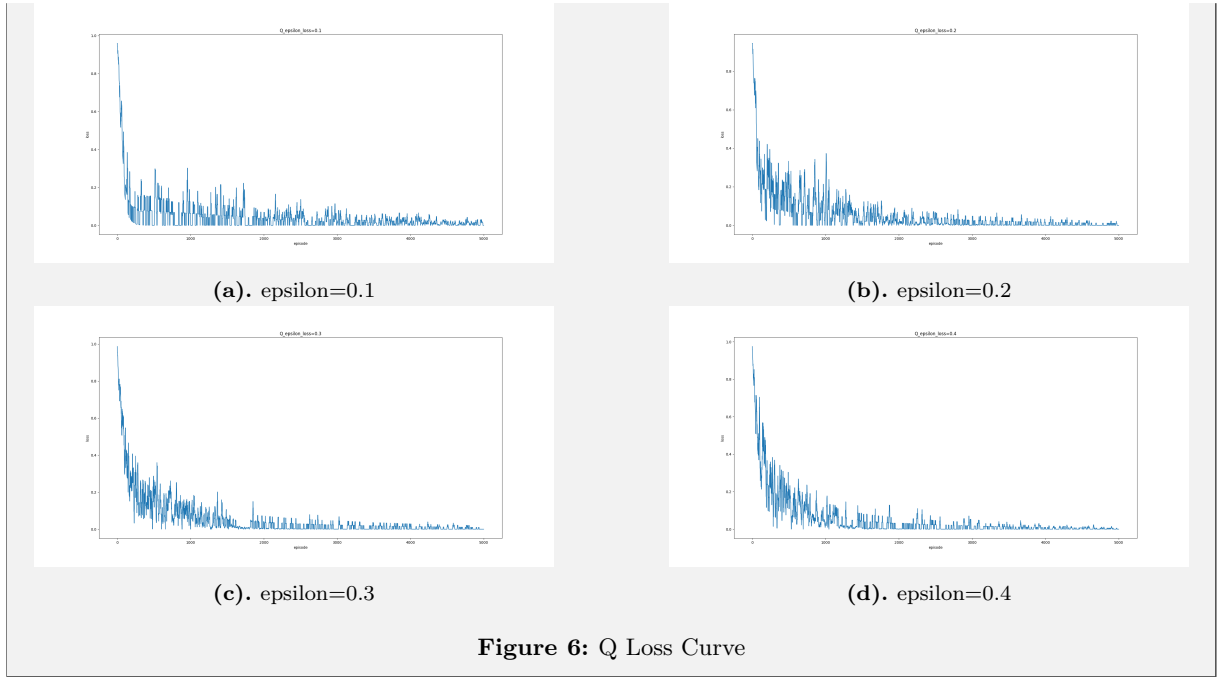
(c). epsilon=0.3



(d). epsilon=0.4

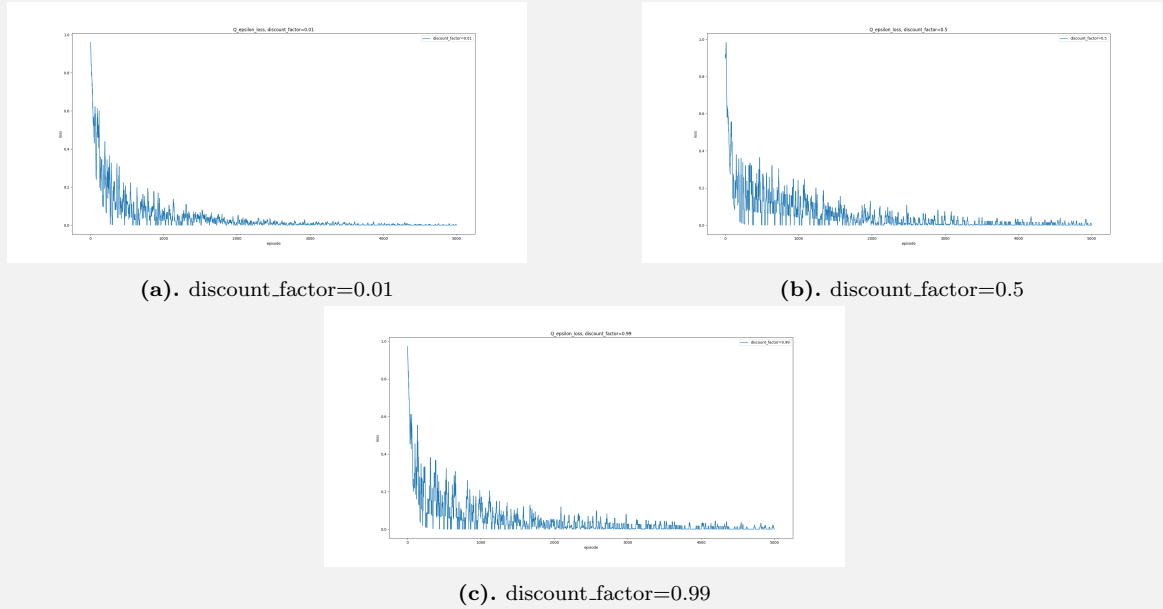
Figure 5: SARSA Loss Curve

4. Q_learning

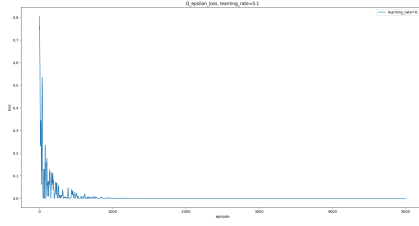


Q3. Discuss and plot ...

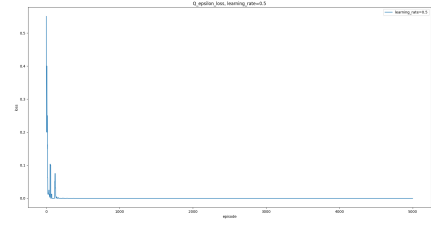
- Below are the figures of different settings of the three methods, we can observe some properties: (In this question, I fixed the max_episodes settings identical with the previous two questions, other hyperparameters are set to default value if not discussed in each section below.)
- discount_factor: In this section, we can find out that as the discount_factor increases, the loss converges slower.



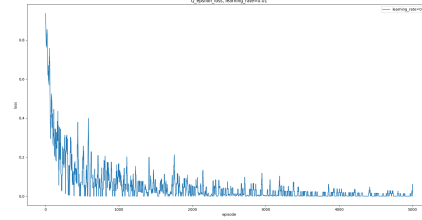
- learning_rate: In this section, we can find out that as the discount_factor decreases, the loss converges slower.



(a). learning_rate=0.1



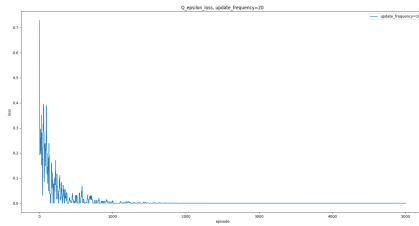
(b). learning_rate=0.5



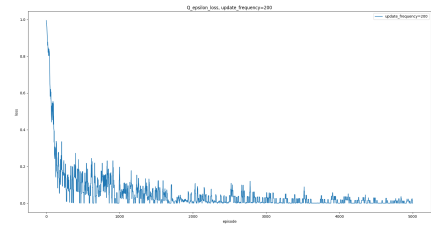
(c). learning_rate=0.01

Figure 8: Q Learning Loss Curve Different Learning Rate

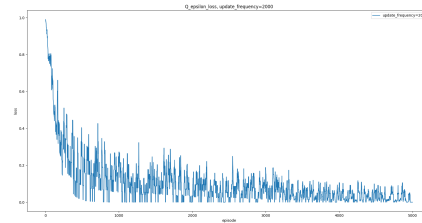
4. Update Frequency: In this section, we can find out that as the discount_factor increases, the loss converges slower.



(a). update_frequency=20



(b). update_frequency=200



(c). update_frequency=2000

Figure 9: Q Learning Loss Curve Different Update Frequency