### RL Assignment2 Report

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## 1. Convergence of temporal difference value learning

To prove that  $\{V_n\}$  is a Cauchy sequence, we need to show that for any given  $\epsilon > 0$ , there exists a positive integer N, such that for any m > n > N, we have  $|V_m - V_n| < \epsilon$ .

First of all, consider the value of  $|V_n - V_{n-1}|$ 

$$|V_n - V_{n-1}| = |\alpha_n(x_n - V_{n-1})| \le |\alpha_n| \cdot (C_1 + C_2)$$

Denoting  $C_1 + C_2$  as  $C_3$ , we can infer that:

$$|V_m - V_n| = |\sum_{i=n+1}^m (V_i - V_{i-1})|$$

$$\leq \sum_{i=n+1}^m |V_i - V_{i-1}|$$

$$\leq \sum_{i=n+1}^m \frac{C_3}{i^2}$$

$$\leq \sum_{i=N+2}^\infty \frac{C_3}{i^2}$$

Because that  $\sum_{i=1}^{\infty} \frac{1}{i^2} = \frac{\pi^2}{6}$ , as the value of N increases,  $\sum_{\infty}^{i=N+2}$  will converge to 0. Thus, we can indicate that for any  $\epsilon > 0$ , there exists a positive integer N, such that for any m > n > N, we have  $|V_m - V_n| < \epsilon$ .

# 2. Implementation of the SARSA and Q-learning algorithms

#### 2.1. coding

The code and its corresponding results, including the performance of SARSA algorithm and Q-learning algorithm using or not using the target policy with different value of  $\epsilon$  in  $\epsilon-greedy$ , can be found in the folder "code".

### 2.2. discussion

(a) When the value of  $\epsilon$  is very small(e.g. 0.01, 0.05), the algorithms mainly exploit rather than explore, the performance of the algorithms are steady and converge relatively well.

As the value of  $\epsilon$  becomes larger, the agent begins to do more exploration, and the performance of SARSA and Q-learning using the behavor policy becomes more and more unsteady. The cumulative reward of Q-learning with the behavor policy no longer converges as  $\epsilon$  reaches 0.1 and SARSA cannot converge as  $\epsilon$  reaches 0.5. Nonetheless, Q-learning algorithm with the target policy keeps convergence even though  $\epsilon$  is raised to 0.9.

(b) The performance of the target policy is usually more optical than the behavor policy, especially when the value of  $\epsilon$  gets very large, the reason of which might be that the target policy is greedy and always picks the action with highest Q-value.