

CIS 419/519: Homework 6

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Although the solutions are entirely my own, I consulted with the following people and sources while working on this homework: <https://www.youtube.com/watch?v=kNPGXgzxoHw>

PART I: PROBELM SET

1 Reinforcement Learning I

The reward does not communicate the goal to the robot well. The robot can only get a positive reward at the end of the maze while all the other points have 0 reward. That is to say, the robot does not know what to do until it first reaches the exit and it keeps wandering around before that. A better reward would be giving all the failed runs a negative reward so that the robot would know that it should try to navigate to the exit as quick as possible.

2 Reinforcement Learning II

(a) The signs do not matter in continuing tasks. For episodic tasks, however, the sign of these rewards do matter as in problem I.

(b) Based on the fact that:

$$R_t = \sum_{k=0}^{\infty} \gamma^k r_{t+k+1}$$

Adding a constant C to all the rewards will simply yield a new reward \tilde{R}_t such that:

$$\tilde{R}_t = \sum_{k=0}^{\infty} \gamma^k (r_{t+k+1} + C)$$

which is also equivalent to:

$$\sum_{k=0}^{\infty} \gamma^k (r_{t+k+1}) + \sum_{k=0}^{\infty} C \gamma^k$$

That is to say:

$$\tilde{R}_t = R_t + \sum_{k=0}^{\infty} C\gamma^k$$

Putting this in to the Value function $V^\pi(s) = \mathbb{E}_\pi[R_t \mid s_t = s]$

We can easily obtain the fact that

$$\tilde{V}_\pi(s) = \mathbb{E}_\pi[G_t + \sum_{k=0}^{\infty} \gamma^k C \mid S_t = s] = V_\pi(s) + \sum_{k=0}^{\infty} \gamma^k C$$

Since the discounted factor γ is always smaller than 1, the second term in the equation above can be simplified as $\frac{C}{1-\gamma}$ based on the geometric series sum.

(c) Based on the facts above, that is to say, the constant K added to the values of all states is simply

$$\frac{C}{1-\gamma}$$