CS285 Homework 1

Jeremiah Chen

June 5, 2025

1 Analysis

1.1 Part 1

Using Hint 1, we have:

$$p_e(t) = \Pr\left(\bigcup_{t=1}^{T} \pi_{\theta}(a_t \neq \pi^*(s_t) \mid s_t)\right)$$

$$\leq \sum_{t=1}^{T} \Pr\left(\pi_{\theta}(a_t \neq \pi^*(s_t) \mid s_t)\right)$$

$$= \sum_{t=1}^{T} \mathbb{E}_{P_{\pi^*(s_t)}} \pi_{\theta}(a \neq \pi^*(s_t) \mid s_t)$$

$$\leq \varepsilon T$$

Since

$$p_{\pi_{\theta}}(s_{t}) = (1 - p_{e}(t)) \cdot p_{\pi^{*}}(s_{t}) + p_{e}(t) \cdot p_{\text{mistake}}(s_{t})$$
$$|p_{\pi_{\theta}}(s_{t}) - p_{\pi^{*}}(s_{t})| \le p_{e}(t) \cdot |p_{\pi^{*}}(s_{t}) - p_{\text{mistake}}(s_{t})|$$

As a result:

$$\sum_{s_t} |p_{\pi_{\theta}}(s_t) - p_{\pi^*}(s_t)| \le \sum_{s_t} p_e(t) \cdot |p_{\pi^*}(s_t) - p_{\text{mistake}}(s_t)|$$

$$\le \max_t p_e(t) \sum_{s_t} (p_{\pi^*}(s_t) + p_{\text{mistake}}(s_t))$$

$$\le 2\varepsilon T$$

1.2 Part 2

Notice that

$$J(\pi^*) - J(\pi_{\theta}) = \sum_{t} \mathbb{E}_{P_{\pi^*(s_t)}} r(s_t) - \sum_{t=1}^{T} \mathbb{E}_{P_{\pi_{\theta}}(s_t)} r(s_t)$$

$$\leq \sum_{t} \sum_{s_t} |P_{\pi^*}(s_t) - P_{\pi_{\theta}}(s_t)| \cdot r(s_t)$$

$$\leq 2\varepsilon T \sum_{t} r(s_t) \qquad \cdots$$

Therefore, for the first case $\forall t < T, r(s_t) = 0$

$$2\varepsilon T \sum_{t} r(s_{t}) \le 2\varepsilon T \cdot R_{\max}$$
$$= \mathcal{O}(T\varepsilon)$$

For an arbitrary case:

$$2\varepsilon T \sum_{t} r(s_{t}) \le 2\varepsilon T^{2} \cdot R_{\max}$$
$$= \mathcal{O}(T^{2}\varepsilon)$$

2 Editing Code

See the file hw1.py for the code implementation.

3 Behavioral Cloning

See the file under textttperformance/BC/ for the results of the behavioral cloning task.

4 DAgger

See the file under textttperformance/DAgger/ for the results of the DAgger task.