

# 1. Analysis

Consider a discrete MDP, horizon  $T$ , an expert policy  $\pi^*$

Assume:

$$\mathbb{E}_{p_{\pi^*}(s)} \pi_\theta(a \neq \pi^*(s) | s) = \frac{1}{T} \sum_{t=1}^T \mathbb{E}_{p_{\pi^*}(s_t)} \pi_\theta(a_t \neq \pi^*(s_t) | s_t) \leq \varepsilon$$

1. Show that  $\sum_{s_t} |p_{\pi}(s_t) - p_{\pi^*}(s_t)| \leq 2T\varepsilon$

$$1) \frac{1}{T} \sum_{t=1}^T \mathbb{E}_{p_{\pi^*}(s_t)} \pi_\theta(a_t \neq \pi^*(s_t) | s_t) = \frac{1}{T} \cdot \sum_{t=1}^T \sum_{s_t} p_{\pi^*}(s_t) \pi_\theta(a_t \neq \pi^*(s_t) | s_t) \leq \varepsilon$$

$$\sum_{t=1}^T \sum_{s_t} p_{\pi^*}(s_t) \pi_\theta(a_t \neq \pi^*(s_t) | s_t) \leq T\varepsilon \quad , \quad \sum_{s_t} p_{\pi^*}(s_t) \pi_\theta(a_t \neq \pi^*(s_t) | s_t) \leq \varepsilon$$

2) By the union bound inequality

$$\sum_{s_t} \pi_\theta(a_t \neq \pi^*(s_t) | s_t) \leq \sum_{t=1}^T \sum_{s_t} p_{\pi^*}(s_t) \pi_\theta(a_t \neq \pi^*(s_t) | s_t) \leq T\varepsilon$$

$$3) p_{\pi_\theta}(s_t) = (1 - \varepsilon)^t p_{\pi^*}(s_t) + (1 - (1 - \varepsilon)^t) p_{\pi_{\text{mistake}}}(s_t)$$

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$$\sum_{s_t} |p_{\pi_\theta}(s_t) - p_{\pi^*}(s_t)| = \sum_{s_t} (1 - (1 - \varepsilon)^t) |p_{\pi_{\text{mistake}}}(s_t) - p_{\pi^*}(s_t)| \leq 2T\varepsilon$$

$$\sum_{s_t} |p_{\pi_\theta}(s_t) - p_{\pi^*}(s_t)| = \sum_{s_t} (1 - (1 - \varepsilon)^t) |p_{\pi_{\text{mistake}}}(s_t) - p_{\pi^*}(s_t)|$$

2. Assume  $|r(s_t)| \leq R_{\max}$ ,  $J(\pi) = \sum_{t=1}^T \mathbb{E}_{p_{\pi}(s_t)} r(s_t)$

$$\begin{aligned} &\leq \sum_{s_t} 2(1 - (1 - \varepsilon)^t) = \sum_{s_t} 2\varepsilon t \\ &= 2T\varepsilon \quad (\because \sum_{s_t} t = T) \end{aligned}$$

a) Show that  $J(\pi^*) - J(\pi_\theta) = O(T\varepsilon)$  s.t.  $r(s_t) = 0 \quad \forall t < T$

$$\begin{aligned} J(\pi^*) - J(\pi_\theta) &= \sum_{t=1}^T \mathbb{E}_{p_{\pi^*}(s_t)} r(s_t) - \sum_{t=1}^T \mathbb{E}_{p_{\pi_\theta}(s_t)} r(s_t) \\ &= \sum_{t=1}^T \left( \sum_{s_t} (p_{\pi^*}(s_t) - p_{\pi_\theta}(s_t)) \right) \cdot r(s_t) \\ &= (p_{\pi^*}(s_T) - p_{\pi_\theta}(s_T)) \cdot r(s_T) \leq 2T\varepsilon \cdot R_{\max} \quad O(T\varepsilon) \end{aligned}$$

$$b) J(\pi^*) - J(\pi_\theta) = \sum_{t=1}^T \sum_{s_t} (p_{\pi^*}(s_t) - p_{\pi_\theta}(s_t)) \cdot r(s_t)$$

$$\leq \sum_{t=1}^T \sum_{s_t} |p_{\pi_\theta}(s_t) - p_{\pi^*}(s_t)| \cdot R_{\max}$$

$$\leq \sum_{t=1}^T 2T\varepsilon \cdot R_{\max} = 2R_{\max} \cdot T \cdot \varepsilon \quad O(T^2\varepsilon)$$