

Lemma 1. *With probability at least $1 - \delta$, for any $(h, k) \in [H] \times [K]$, $\mu^s \in \text{BR}(\pi^s)$, and $\pi \in \Pi$*

$$L_h^{k-1}(\pi^*) - L_h^{k-1}(\pi) \leq -2 \sum_{s=1}^{k-1} \mathbb{E}_{\xi_h \sim \mu^s} [\ell_{\pi^s}(\pi; \xi_h)] + 2 \log(H |\Pi| / \delta).$$

Proof. Given $\pi \in \mathcal{H}$, we denote the random variable $X_{h,\pi}^k$ as

$$X_{h,\pi}^k = \log \left(\frac{\mathbb{P}_{h,\pi^*}(s_{h+1}^k | s_h^k, a_h^k)}{\mathbb{P}_{h,\pi}(s_{h+1}^k | s_h^k, a_h^k)} \right).$$

Now we define a filtration $\{\mathcal{F}_{h,k}\}_{k=1}^K$ as (B.25) in [1]. Thus we have $X_{h,\pi}^k \in \mathcal{F}_{h,k}$. Therefore, by applying Lemma D.1 in [1], we have that with probability at least $1 - \delta$, for any $(h, k) \in [H] \times [K]$, and $\pi \in \Pi$, we have

$$-\frac{1}{2} \sum_{s=1}^{k-1} X_{h,\pi}^s \leq \sum_{s=1}^{k-1} \log \mathbb{E} \left[\exp \left\{ -\frac{1}{2} X_{h,\pi}^s \right\} \middle| \mathcal{F}_{h,s-1} \right] + \log(H |\Pi| / \delta). \quad (1)$$

Meanwhile, by (B.27) in [1], for any $\mu^s \in \text{BR}(\pi^s)$, the conditional expectation equals to

$$\mathbb{E} \left[\exp \left\{ -\frac{1}{2} X_{h,\pi}^s \right\} \middle| \mathcal{F}_{h,s-1} \right] = 1 - \mathbb{E}_{(s_h^s, a_h^s) \sim \mu^s} [D_H(\mathbb{P}_{h,\pi^*}(\cdot | s_h^s, a_h^s) || \mathbb{P}_{h,\pi}(\cdot | s_h^s, a_h^s))]. \quad (2)$$

Denote $\mathbb{E}_{(s_h^s, a_h^s) \sim \mu^s} [D_H(\mathbb{P}_{h,\pi^*}(\cdot | s_h^s, a_h^s) || \mathbb{P}_{h,\pi}(\cdot | s_h^s, a_h^s))]$ as $\mathbb{E}_{\xi_h \sim \mu^s} [\ell_{\pi^s}(\pi; \xi_h)]$. Using the fact $\log(x) \leq x - 1$ and substituting (2) into (1) finishes the proof. \square

Initializing a policy set $\mathcal{H}_\psi \leftarrow \mathcal{H}_{\text{fin}}$, for all $k, l \in [K]$ with $k > l$, if $\pi^k \sim^\psi \pi^l$, we eliminate π^l from \mathcal{H}_ψ . The resulting \mathcal{H}_ψ has the following property by its construction:

- $\mathcal{H}_\psi \subset \mathcal{H}_{\text{fin}}$.
- $n^\psi(\mathcal{H}_\psi) \leq n^\psi(\mathcal{H}_{\text{fin}})$.

Lemma 2. *If for all $k \in [K]$ such that $\pi^k \in \mathcal{H}_\psi$, we have*

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k) \leq c_k,$$

where $\{c_k\}_{k \in [K]}$ is a non-increasing sequence. Then, for all $k \in [K]$, we have

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k) \leq c_k.$$

Proof. By definition, for all $k, l \in [K]$ with $k > l$ and $\pi^k \sim^\psi \pi^l$, we have

$$V(\psi(\pi^k), \pi^k) = V(\psi(\pi^l), \pi^l).$$

Thus,

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k) = V(\psi(\pi^*), \pi^*) - V(\psi(\pi^l), \pi^l).$$

Note that for all $k \in [K]$ such that $\pi^k \in \mathcal{H}_\psi$, we have

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k) \leq c_k,$$

which implies $V(\psi(\pi^*), \pi^*) - V(\psi(\pi^l), \pi^l) \leq c_k$. By the construction rule of \mathcal{H}_ψ , for all $l \in [K]$ with $\pi^l \notin \mathcal{H}_\psi$, we can always find a constant k' such that $k' > l$ and $\pi^{k'} \in \mathcal{H}_\psi$. Thus

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^l), \pi^l) \leq c_{k'} \leq c_l. \quad \square$$

Theorem 3. *Given an MDP with generalized eluder coefficient $d_{\text{GEC}}(\cdot)$ and a finite hypothesis class \mathcal{H}_{fin} with $\pi^* \in \mathcal{H}_{\text{fin}}$, by setting*

$$\eta = \sqrt{\frac{d_{\text{GEC}}(1/\sqrt{HK})}{\log(Hn^\psi(\mathcal{H}_{\text{fin}})/\delta) \cdot HK}},$$

the regret of the MEX algorithm applying on \mathcal{H}_{fin} with oracle ψ after K episodes is upper bounded by, with probability at least $1 - \delta$,

$$\text{Regret}(K) \lesssim \sqrt{d_{\text{GEC}}(1/\sqrt{HK}) \cdot \log(Hn^\psi(\mathcal{H}_{\text{fin}})/\delta) \cdot HK}.$$

Proof. We decompose the regret into two terms,

$$\begin{aligned} \text{Regret}(K) &\triangleq \sum_{k=1}^K V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^*) \\ &= \underbrace{\sum_{k=1}^K V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k)}_{\text{Term (i)}} + \underbrace{\sum_{k=1}^K V(\psi(\pi^k), \pi^k) - V(\psi(\pi^k), \pi^*)}_{\text{Term (ii)}}. \end{aligned}$$

Term (i). By the choice of π^k , we have

$$V(\psi(\pi^*), \pi^*) - \eta \sum_{h=1}^H L_h^{k-1}(\pi^*) \leq V(\psi(\pi^k), \pi^k) - \eta \sum_{h=1}^H L_h^{k-1}(\pi^k)$$

for all $k \in [K]$. Thus,

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k) \leq \eta \sum_{h=1}^H (L_h^{k-1}(\pi^*) - L_h^{k-1}(\pi^k)). \quad (3)$$

Applying Lemma 1, we have that with probability at least $1 - \delta$, for any $(h, k) \in [H] \times [K]$ and all $\pi \in \mathcal{H}_\psi$,

$$L_h^{k-1}(\pi^*) - L_h^{k-1}(\pi) \leq -2 \sum_{s=1}^{k-1} \mathbb{E}_{\xi_h \sim \psi(\pi^s)} [\ell_{\pi^s}(\pi; \xi_h)] + 2 \log(H|\mathcal{H}_\psi|/\delta).$$

Substituting the above equation into (3) gives us, with probability at least $1 - \delta$, for all $k \in [K]$ with $\pi^k \in \mathcal{H}_\psi$, we have

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k) \leq -2\eta \sum_{h=1}^H \sum_{s=1}^{k-1} \mathbb{E}_{\xi_h \sim \psi(\pi^s)} [\ell_{\pi^s}(\pi^k; \xi_h)] + 2H\eta \log(H|\mathcal{H}_\psi|/\delta)$$

We define c_k as

$$c_k \triangleq -2\eta \sum_{h=1}^H \sum_{s=1}^{k-1} \mathbb{E}_{\xi_h \sim \psi(\pi^s)} [\ell_{\pi^s}(\pi^k; \xi_h)] + 2H\eta \log(H|\mathcal{H}_\psi|/\delta).$$

The sequence $\{c_k\}_{k \in [K]}$ is a non-increasing sequence. Applying Lemma 2 gives us, with probability at least $1 - \delta$, for all $k \in [K]$, we have

$$V(\psi(\pi^*), \pi^*) - V(\psi(\pi^k), \pi^k) \leq c_k.$$

Summing over $[K]$ gives us, with probability $1 - \delta$,

$$\begin{aligned} \text{Term (i)} &\leq \sum_{k=1}^K c_k \\ &= -2\eta \sum_{k=1}^K \sum_{h=1}^H \sum_{s=1}^{k-1} \mathbb{E}_{\xi_h \sim \psi(\pi^s)} [\ell_{\pi^s}(\pi^k; \xi_h)] + 2H\eta \log(H |\mathcal{H}_\psi| / \delta) \\ &\leq -2\eta \sum_{k=1}^K \sum_{h=1}^H \sum_{s=1}^{k-1} \mathbb{E}_{\xi_h \sim \psi(\pi^s)} [\ell_{\pi^s}(\pi^k; \xi_h)] + 2H\eta \log(H n_\psi(\mathcal{H}_{\text{fin}}) / \delta). \end{aligned}$$

Term (ii). Follow the proof of Theorem 4.4 in [1],

$$\text{Term (ii)} \leq 2\eta \sum_{k=1}^K \sum_{h=1}^H \sum_{s=1}^{k-1} \mathbb{E}_{\xi_h \sim \psi(\pi^s)} [\ell_{\pi^s}(\pi^k; \xi_h)] + \frac{d_{\text{GEC}}(\varepsilon_{\text{conf}})}{8\eta} + \sqrt{d_{\text{GEC}}(\varepsilon_{\text{conf}})HK} + \varepsilon_{\text{conf}}HK.$$

Combining Term (i) and Term (ii).

$$\begin{aligned} \text{Regret}(K) &= \text{Term (i)} + \text{Term (ii)} \\ &\leq 2\eta KH \log(H n^\psi(\mathcal{H}_{\text{fin}}) / \delta) + \frac{d_{\text{GEC}}(\varepsilon_{\text{conf}})}{8\eta} + \sqrt{d_{\text{GEC}}(\varepsilon_{\text{conf}})HK} + \varepsilon_{\text{conf}}HK. \end{aligned}$$

Set $\varepsilon_{\text{conf}} = 1 / \sqrt{HK}$ and

$$\eta = \sqrt{\frac{d_{\text{GEC}}(1 / \sqrt{HK})}{\log(H n^\psi(\mathcal{H}_{\text{fin}}) / \delta) \cdot HK}}$$

leads to the proof. □

Bibliography

- [1] Zhihan Liu, Miao Lu, Wei Xiong, Han Zhong, Hao Hu, Shenao Zhang, Sirui Zheng, Zhuoran Yang, and Zhaoran Wang. One Objective to Rule Them All: A Maximization Objective Fusing Estimation and Planning for Exploration. may 2023.