

Multi-Armed Bandits

Given: K arms; for each arm d , reward dist D_d w/ mean μ_d . (unknown)
 T rounds ($T \gg K$) $\in [0, 1]$

Output: Arms ~~d_1, d_2, \dots, d_T~~ played.

Procedure In each round $t=1, \dots, T$,

- Defn (Regret)
1. ALG plays an arm d_t
 2. Reward r_t is sampled indep from D_{d_t}
 3. ALG learns r_t .

Defn (Regret)

$$R(T) = \mu^* \cdot T - \sum_{t=1}^T \mu_{d_t}, \quad \text{where } \mu^* = \max_{d \in K} \mu_d.$$

Generally, consider $\mathbb{E}[R(T)]$ where the expectation is over the randomness of D_d 's & the alg's choices.

Fact (Hoeffding's meq)

Given mutually indep (not necessarily identically dist) X_1, X_2, \dots, X_n , let

$$\bar{X}_n := \frac{X_1 + \dots + X_n}{n} \quad \text{and} \quad \mu_n := \mathbb{E}[\bar{X}_n] = \frac{\mathbb{E}[X_1] + \dots + \mathbb{E}[X_n]}{n}.$$

We then have, for any T , $\Pr \left[|\bar{X}_n - \mu_n| \leq \sqrt{\frac{2 \log T}{n}} \right] \geq 1 - \frac{2}{T^4}.$

Alg 1 (Uniform exploration)

1. Try each arm N times.
2. Select one w/ highest avg reward.
3. Play the chosen one \hat{d} in the remaining rounds.

Obsv At the end of step t , for any arm α ,

$$\Pr[|\bar{\mu}_\alpha - \mu_\alpha| \leq \text{rad}] \geq 1 - \frac{2}{T^4}, \text{ where}$$

$\bar{\mu}_\alpha$ denotes the observed avg reward of α , and $\text{rad} := \sqrt{\frac{2 \ln T}{N}}$.

Let "clean event" be $\bigwedge_{\alpha \in \mathcal{K}} (|\bar{\mu}_\alpha - \mu_\alpha| \leq \text{rad})$.

Obsv

By the union bound, $\Pr[\text{bad event}] \leq \frac{2K}{T^4}$.

Lem Cond on "clean event", if $\hat{\alpha} \neq \alpha^*$, we have $|\mu_{\hat{\alpha}} - \mu_{\alpha^*}| \leq 2\text{rad}$.

Pf) Alg chose $\hat{\alpha}$ instead of α^* since $\bar{\mu}_{\hat{\alpha}} \geq \bar{\mu}_{\alpha^*}$. Due to the condition, $\mu_{\hat{\alpha}} + \text{rad} \geq \bar{\mu}_{\hat{\alpha}} \geq \bar{\mu}_{\alpha^*} \geq \mu_{\alpha^*} - \text{rad}$. \square

Lem $\mathbb{E}[R(t)] \leq Nk + 2\text{rad} \cdot T + o(1)$.

$$\begin{aligned} \text{Pf) } \mathbb{E}[R(t)] &= \mathbb{E}[R(t) | \text{clean}] \Pr[\text{clean}] + \mathbb{E}[R(t) | \text{bad}] \Pr[\text{bad}] \\ &\leq \mathbb{E}[R(t) | \text{clean}] + T \cdot O\left(\frac{K}{T^4}\right). \end{aligned}$$

$$\text{Obsv } \mathbb{E}[R(t) | \text{clean}] \leq N(k-1) + 2\text{rad}(T - Nk). \quad \square$$

Choosing $N = O\left(\left(\frac{T}{K}\right)^{\frac{2}{3}} \cdot (\ln T)^{\frac{1}{3}}\right)$, we have $\mathbb{E}[R(t)] \leq O\left(T^{\frac{2}{3}} \cdot K^{\frac{1}{3}} \cdot (\ln T)^{\frac{1}{3}}\right)$.

Obsv Fix round t and arm α . If α is played $n_{t,\alpha}$ times, where $\bar{\mu}_{t,\alpha}$ is the observed avg reward, we then have

$$\Pr[|\bar{\mu}_{t,\alpha} - \mu_\alpha| \leq \text{rad}_{t,\alpha}] \geq 1 - \frac{2}{T^4}, \text{ where}$$

$$\text{rad}_{t,\alpha} := \sqrt{\frac{2 \ln T}{n_{t,\alpha}}}.$$

See $\text{UCB}_t(\alpha) := \bar{\mu}_{t,\alpha} + \text{rad}_{t,\alpha}$ & $\text{LCB}_t(\alpha) := \bar{\mu}_{t,\alpha} - \text{rad}_{t,\alpha}$.

Alg (Successive elimination)

1. Activate all arms.
2. For each phase:
3. play all active arms & update $[LCB, UCB]$.
4. Deactivate all arms whose CB does not overlap of "highest" CB.

Clean event? $\bigwedge_{d \in K, t \in T} (|\hat{\mu}_t(d) - \mu(d)| \leq \text{rad}_t(d)).$

$$\mu(d) \in [LCB_t(d), UCB_t(d)]$$

By the union bound, $\Pr[\text{bad event}] \leq T \cdot K \cdot \frac{2}{T^4} = O(\frac{1}{T^2}).$

Cond on clean event,

- d^* never deactivated.

- For each arm d , the CB of d overlaps CB of d^* at $n_t(d)$ th phase.

$$\Rightarrow \mu(d^*) - \mu(d) \leq 2 \text{rad}_t(d) = 2 \sqrt{\frac{2 \lg T}{n_t(d)}}$$

$$\therefore \text{Contribution of } d \text{ to } \mathbb{E}[R(t) | \text{clean}] = n_t(d) \cdot O(\sqrt{\frac{\lg T}{n_t(d)}}) = O(\sqrt{n_t(d) \lg T})$$

$$\Rightarrow \mathbb{E}[R(t) | \text{clean}] = \sum_{d \in K} O(\sqrt{n_t(d) \lg T}) = O(\sqrt{\lg T}) \sum_{d \in K} \sqrt{n_t(d)} = O(\sqrt{K \lg T})$$

Note: by Jensen's inequality,

$$\frac{1}{K} \sum_{d \in K} \sqrt{n_t(d)} \leq \sqrt{\frac{\sum_{d \in K} n_t(d)}{K}} = \sqrt{\frac{t}{K}} \Rightarrow \sum_{d \in K} \sqrt{n_t(d)} = \sqrt{Kt}$$

Lower bounds

Thm Fix T & K . For any alg, \exists family of instances

$$\text{s.t. } \mathbb{E}[R(t)] \geq \Omega(\sqrt{KT}).$$

Today, wts for $K=2!$

Consider ① I_1 - $D_{d_1} = \text{Ber}(\frac{1+\epsilon}{2})$ ② I_2 - $D_{d_1} = \text{Ber}(\frac{1}{2})$
 $- D_{d_2} = \text{Ber}(\frac{1}{2})$ $- D_{d_2} = \text{Ber}(\frac{1+\epsilon}{2})$

Outline: think of MAB as a seq of "best-arm identification" \rightarrow

in the same setting, the goal is to choose the max-reward arm!
 at round t

Lem Consider a "best-arm identification" w/ $\frac{t}{T} \leq \frac{1}{16\epsilon^2}$. Fix any alg.
 $\exists \alpha \in \{d_1, d_2\}$ st.

$$\Pr[\text{Alg chose } \alpha \mid I_\alpha] < 3/4.$$

(Using this lemma, we have)

$$T \leq \frac{1}{16\epsilon^2} \Rightarrow \epsilon \leq \sqrt{\frac{1}{16T}}$$

Thm Fix T & any ALG. Choose an arm α w/ ϵ , & run ALG on I_α .
 Then, $\mathbb{E}[R(T)] \geq \Omega(\sqrt{KT})$. $w/ \epsilon = O(\frac{1}{\sqrt{T}})$

pf) By choice of ϵ , we can use Lem for each round $t \leq T = \frac{1}{16\epsilon^2}$.

$$\Pr[d_t \neq \alpha] = \underbrace{\Pr[d_t \neq d_1 \mid I_1]}_{\geq \frac{1}{8}} \underbrace{\Pr[I_1]}_{=\frac{1}{2}} + \underbrace{\Pr[d_t \neq d_2 \mid I_2]}_{\geq \frac{1}{4}} \underbrace{\Pr[I_2]}_{=\frac{1}{2}}$$

one would be

$$\therefore \mathbb{E}[R(T)] = \sum_{t=1}^T \Pr[d_t \neq \alpha] \cdot \frac{\epsilon}{2} \geq \frac{\epsilon T}{16} = \Omega(\sqrt{KT})$$

□

KL-divergence

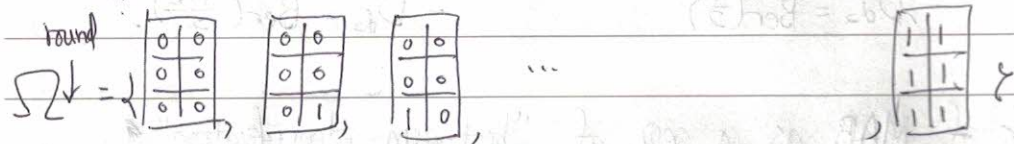
$$KL(p, q) = \sum_{x \in \Omega} p(x) \cdot \ln \frac{p(x)}{q(x)}$$

Fact (Chain rule) For $p = p_1 \times p_2 \times \dots \times p_n$ & $q = q_1 \times q_2 \times \dots \times q_n$ where
 p_i, q_i on the same sample space, $KL(p, q) = \sum_{i=1}^n KL(p_i, q_i)$.

Fact (Pinsker's Ineq) For any event $A \subset \Omega$, $2(p(A) - q(A))^2 \leq KL(p, q)$!

Fact $KL(\text{Ber}(\frac{1+\epsilon}{2}), \text{Ber}(\frac{1}{2})) \leq 2\epsilon^2$ & $KL(\text{Ber}(\frac{1}{2}), \text{Ber}(\frac{1+\epsilon}{2})) < \epsilon^2 \quad \forall \epsilon < \frac{1}{2}$.

pf of Lem) Consider all possible outcomes. For example, $t=3$,



Let $p_1(\omega)$ & $p_2(\omega)$ be the prob of outcome ω in I_1 & I_2 , resp.

e.g.,

$$p_1 \left(\begin{array}{|c|c|c|} \hline 0 & 0 & 1 \\ \hline 0 & 0 & 1 \\ \hline 0 & 0 & 1 \\ \hline \end{array} \right) = \left(\frac{1-\epsilon}{2} \right)^3 \cdot \left(\frac{1}{2} \right)^3 \quad p_2 \left(\begin{array}{|c|c|c|} \hline 1 & 1 & 0 \\ \hline 1 & 1 & 0 \\ \hline 1 & 1 & 0 \\ \hline \end{array} \right) = \left(\frac{1}{2} \right)^3 \left(\frac{1-\epsilon}{2} \right)^2 \left(\frac{1+\epsilon}{2} \right)$$

As ALG is not aware of the real dist, the output of ALG depends on the outcome, i.e., $ALG: \Omega \rightarrow d_1, d_2$.

Let $E := \{ \omega : ALG(\omega) = d_1 \}$. Then $\bar{E} := \{ \omega : ALG(\omega) = d_2 \}$.

* Spse t.c. that $p_1(E) \geq 3/4$ & $p_2(\bar{E}) \geq 3/4$. We then have

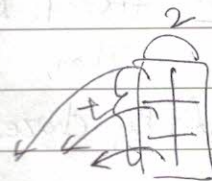
$$p_1(E) - p_2(E) \geq 3/4 - 1/4 = 1/2. \dots (a)$$

However, observe that

$$2(p_1(E) - p_2(E))^2 \leq KL(p_1, p_2) = \sum_{\text{each cell}} KL(p_1^{cell}, p_2^{cell})$$

$$\leq \frac{t}{2} \cdot 2\epsilon^2$$

$$\Rightarrow p_1(E) - p_2(E) \leq \frac{\epsilon \sqrt{t}}{2} \leq \frac{1}{2\sqrt{2}} < \frac{1}{2} \quad \text{since } \frac{t}{2} \leq \frac{1}{6\epsilon^2} \Rightarrow \epsilon \leq \frac{1}{\sqrt{3t}}$$



~~For general K,~~

For general K , $I_d: P_d = \text{Ber}(\frac{1+\epsilon}{2})$, the other $\text{Ber}(\frac{1}{2})$.

Lem Consider a "best-arm identification" w/ $t \leq \frac{cK}{\epsilon^2}$ (for a small enough absolute constant c). Fix any alg, \exists at least $\frac{K}{3}$ arms d st.

$$P_r[ALG \text{ chose } d \mid I_d] < 3/4.$$