

Week 9 - Lecture 41

An environment :

$$\mathcal{V} = (\mu_1, 0, 0, 0 \dots 0)$$

$$\text{where } \mu_1 = [0, 1/2]$$

$$i = \arg \min_{j > 1} \mathbb{E}[N_j(T)] \quad \text{where } i \text{ has been played less than}$$

$T/(K-1)$ rounds

$$\therefore \mathbb{E}[N_i(T)] \leq T/(K-1)$$

2nd env:

$$\mathcal{V}' = (\mu_1, 0, 0, \dots, \underset{\uparrow_i}{2\mu_1}, \dots, 0)$$

We know :

$$R_T(\pi, \mathcal{V}) = \sum_{i=1}^K \Delta_i \mathbb{E}_{\mathcal{V}}[N_i(T)]$$

$$1) \quad R_T(\pi, \mathcal{V}) \geq P_{\mathcal{V}}(N_1(T) \leq T/2) \frac{T \mu_1}{2}$$

optimal arm has been played less than $T/2$ times

[1 is optimal arm]

Thus sub-optimal arm has been played $T/2$ times.

And on these rounds would have incurred regret

of $T\mu_1/2$.

$$2) R_T(\pi, \nu') \geq P_{\nu'}(N_1(T) > T/2) \frac{T\mu_1}{2}$$

\therefore

$$\begin{aligned} R_T(\pi, \nu) + R_T(\pi, \nu') \\ \geq \frac{T\mu_1}{2} \left(\underbrace{P_{\nu}(N_1(T) \leq T/2)}_{\text{Event } A} + \underbrace{P_{\nu'}(N_1(T) > T/2)}_{\text{Event } A^c} \right) \end{aligned}$$

[Event A & A^c w.r.t. two diff. probability measures]

\therefore Using the divergence result :

$$\geq \frac{T\mu_1}{2} \frac{1}{2} \exp(-D(P_{\nu} \parallel P_{\nu'}))$$

$$\left[\begin{array}{l} \text{Underlying sample space for both environments} \\ \Omega = \mathbb{R}^T \times [K]^T \end{array} \right]$$

Invoking the lemma proven in the earlier lecture:

$$D(P_{\nu} \parallel P_{\nu'}) = \sum_{k=1}^K \mathbb{E}_{\nu}[N_k(T)] D(P_k \parallel P'_k)$$

Since ν & ν' are same except at position i ,
 \therefore apart from at position i , everything would
 be zero, \therefore we have:

$$D(P_{\psi} \parallel P_{\psi'}) = \mathbb{E}_{\psi} [N_i(T)] D(P_i \parallel P'_i)$$

The environments are gaussian \therefore we can get expression for divergence : \hookrightarrow [Assume $\sigma^2 = 1$]

$$D(P_{\psi} \parallel P_{\psi'}) = \mathbb{E}_{\psi} [N_i(T)] \frac{(0 - 2\mu_1)^2}{2\sigma^2}$$

$$= \mathbb{E}_{\psi} [N_i(T)] \frac{2\mu_1^2}{\sigma^2}$$

$$\leq \left(\frac{T}{K-1} \right) 2\mu_1^2 \quad \left[\text{Assume } \sigma^2 = 1 \right]$$

Plugging this into the regret summation:

$$R_T(\pi, \psi) + R_T(\pi, \psi') \geq \frac{T\mu_1}{4} \exp\left(-\frac{2T\mu_1^2}{K-1}\right)$$

$$\text{Let } \mu_1 = \sqrt{\frac{K-1}{4T}} \quad \left[\begin{array}{l} \text{Already assumed that} \\ T > K-1 \end{array} \right]$$

$$\leq \sqrt{K_4} = \frac{1}{2} \quad \left[\begin{array}{l} \text{Thus satisfied } \mu_1 = [0, 1/2] \\ \& 2\mu_1 = [0, 1] \end{array} \right]$$

\therefore

$$R_T(\pi, \psi) + R_T(\pi, \psi')$$

$$\geq \frac{1}{2 \times 4} \sqrt{T(K-1)} \frac{1}{\sqrt{e}}$$

For minmax lower bound the regret to be incurred would be max of $1/2$ of the summation of regrets

\therefore

$$2 \max \{R_T(\pi, v), R_T(\pi, v')\}$$

$$\geq R_T(\pi, v) + R_T(\pi, v')$$

$$\Rightarrow \max \{R_T(\pi, v), R_T(\pi, v')\}$$

$$\geq (R_T(\pi, v) + R_T(\pi, v')) / 2$$

$$\therefore R_T^*(\varepsilon^K) \geq \frac{1}{16\sqrt{e}} \sqrt{T(K-1)}$$

Hence Proved