

2) To find  $C > 0$  such that for all  $n \geq k > 1$  and for all policies  $\pi$  there exists a  $V \in \mathcal{E}_N^K$  such that  $R_n^S(\pi, V) \geq (C\sqrt{k/n})$

$$R_n^S(\pi, V) = E_{V, \pi}[\Delta_{A_{n+1}}(V)]$$

$$= \sum_{i=1}^K P_{V, \pi}(A_{n+1} = i) \Delta_i$$

Fix a policy  $\pi$ . Consider  $K$ -arm gaussian bandit, with variance 1 and mean  $\mu = (\Delta, 0, 0, \dots, 0)$ . Denoting this environment as  $V$ . A policy  $\pi$  and environment  $V$  will give rise to  $P_{V, \pi}$  which will be denoted as  $P_\mu$ .  $\pi$  is the optimal arm.

Let  $i = \underset{j > 1}{\operatorname{argmin}} E_\mu(T_j(n+1))$ .

As  $\sum_{j=1}^K E_\mu(T_j(n+1)) = n+1$ , we can say  $E_\mu(T(i)) \leq \frac{n+1}{K-1}$  from (1)

— (2)

Using this we define the second environment as with mean vector,

$\mu' = (\Delta, 0, 0, \dots, 2\Delta, 0, \dots)$  and gaussian distribution. Where  $2\Delta$  is at the  $i^{\text{th}}$  arm

This environment  $V'$  will give rise to  $P_{V', \pi}$  written as  $P_{\mu'}$

Now,

$$R_n^S(u, v) = \sum_{j=2}^K P_{nv}(j) \Delta \quad [\because \text{For } j=1, \text{ regret} = 0]$$

$$= (1 - P_{nv}(1)) \Delta$$

$$= (1 - E_{\mu}(I_{A_{n+1}=1})) \Delta$$

Consider the event  $A: E(I_{A_{n+1}=1}) \leq 1/2$  [Note that this is a random event as no dependence on  $\mu$ ]

$$\text{Then } R_n^S(u, v) = (1 - E_{\mu}(I_{A_{n+1}=1})) \Delta$$

$$\geq (1 - E_{\mu}(I_{A_{n+1}=1})) P_{\mu}(E(I_{A_{n+1}=1}) \leq 1/2) \Delta$$

$$+ (1 - E(I_{A_{n+1}=1})) P_{\mu}(E(I_{A_{n+1}=1}) > 1/2) \Delta$$

$$\geq (1 - E(I_{A_{n+1}=1})) P_{\mu}(E(I_{A_{n+1}=1}) \leq 1/2) \Delta$$

$$\geq \frac{1}{2} P_{\mu}(E(I_{A_{n+1}=1}) \leq 1/2) \Delta \geq \frac{\Delta}{2} P_{\mu}(A)$$

For  $R_n^S(u, v')$  we have,

$$R_n^S(u, v') = \Delta P_{\mu'}(1) + \sum_{j \neq 1}^K P(j) 2\Delta$$

$$\geq \Delta P_{\mu'}(1)$$

$$= \Delta E_{\mu'}(I_{A_{n+1}=1})$$

Consider the event  $A^c: E(I_{A_{n+1}=1}) > 1/2$  [Again random as no degenerate environment is specified]

$$\therefore R_n^S(u, v') \geq \Delta E(I_{A_{n+1}=1})_{\mu'} \cdot P_{\mu'}((E_{A_{n+1}=1}) > 1/2)$$

$$\geq \frac{\Delta}{2} P_{\mu'}(A^c)$$

$$\begin{aligned} \therefore R_n^S(\pi, \nu) + R_n^S(\pi, \nu') &\geq \frac{\Delta}{2} P_\mu(A) + \frac{\Delta}{2} P_{\mu'}(A') \\ &= \frac{\Delta}{2} (P_\mu(A) + P_{\mu'}(A')) \end{aligned}$$

From Bretagnolle-Huber inequality we know that,

$$P(A) + Q(A') \geq \frac{1}{2} \exp(-D(P, Q))$$

$$\therefore R_n^S(\pi, \nu) + R_n^S(\pi, \nu') \geq \frac{\Delta}{2} \times \frac{1}{2} \exp(-D(P_\mu, P_{\mu'}))$$

Now in  $\mu$  &  $\mu'$  the only difference is the  $i^{\text{th}}$  term  
where  $\mu'_i - \mu_i = 2\Delta$

~~$$\begin{aligned} \therefore D(P_\mu, P_{\mu'}) &= E_\mu \left( \sum_{n+1}^{\infty} I_{A_{n+1}=i} \right) \cdot D[N(0,1), N(2\Delta,1)] \\ &= P_\mu(i) \cdot (2\Delta)^2 \end{aligned}$$~~

~~But  $P_\mu(i) \leq 1/K$~~

~~$$\therefore D(P_\mu, P_{\mu'}) \leq \frac{1}{K} (2\Delta)^2$$~~

~~$$\therefore R_n^S(\pi, \nu) + R_n^S(\pi, \nu') \geq \frac{\Delta}{2} \exp\left(-\frac{1}{K} (2\Delta)^2\right)$$~~

$$\begin{aligned} \therefore D(P_\mu, P_{\mu'}) &= E_\mu(T_i(n+1)) \cdot D[N(0,1), N(2\Delta,1)] \\ &\leq \frac{n+1}{K-1} \frac{(2\Delta)^2}{2} \text{ from (2)} \end{aligned}$$

$$\therefore R_n^S(\pi, \nu) + R_n^S(\pi, \nu') \geq \frac{\Delta}{4} \exp\left(-\frac{n 2\Delta^2}{K}\right)$$

Choose  $\Delta = \sqrt{\frac{K-1}{4(n+1)}}$ , as it is given that  $n \geq K \geq 1$   
 $\Delta \leq 1/2$ , [We need  $\Delta \leq 1/2$  as we used  $2\Delta$  as mean and we have bounded means in  $[0, 1]$ ]

$$R_n^S(\pi, \nu) + R_n^S(\pi, \nu') \geq \frac{1}{4} \sqrt{\frac{K-1}{n+1}} \exp(-1/2)$$

$$\Rightarrow 2 \max(R_n^S(\pi, \nu), R_n^S(\pi, \nu')) \geq \frac{1}{8} \sqrt{\frac{K-1}{n+1}} \exp(-1/2)$$

Let  $R_n^S(\pi, \nu) \geq R_n^S(\pi, \nu')$ , then

$$R_n^S(\pi, \nu) \geq \frac{1}{16} \frac{1}{\sqrt{e}} \sqrt{\frac{K-1}{n+1}}$$

$$\geq \frac{1}{27} \sqrt{\frac{K-1}{n+1}}$$

Now, notice that, we need a  $\lambda$  such that

$$\frac{K-1}{n+1} \geq \lambda \frac{K}{n} \quad \text{where } n \geq K \geq 1$$

Choose  $\lambda = 1/3$

$$\Rightarrow \frac{K-1}{n+1} \geq \frac{1}{3} \frac{K}{n}$$

$$\Rightarrow 3nK - 3n \geq Kn + K$$

$$\Rightarrow 2nK - 3n \geq K$$

$$\Rightarrow n(2K-3) \geq K$$

$$\Rightarrow (2K-3) \geq \frac{K}{n} > 0$$

Now  $K \in [2, 3, \dots, \infty)$

$$\therefore 2K \in [4, 6, \dots, \infty)$$

$\therefore 2K-3 > 0$ , hence relation is satisfied for  $\lambda = 1/3$

$$\therefore R_n^S(\pi, \nu) \geq \frac{1}{27\sqrt{3}} \sqrt{\frac{K}{n}} //$$



Q2) Choose  $\mathcal{K}$  the environment of  $K$ -arm gaussian bandit, with  
variance of all arms as one and means as

$$\mu \equiv (\Delta, 0, 0, \dots, 0)$$

The policy is U.E

$$R_n^S(U.E, \nu) = \sum_{j=2}^K \Delta_j P_{\pi_\nu}(j)$$

$$= \sum_{j=2}^K \Delta P_{\pi_\nu}(j)$$

$$= (1 - P_{\pi_\nu}(1)) \Delta$$

$$\text{Now } P_{\pi_\nu}(1) = P_{\pi_\nu}(\hat{\mu}_1 > \max_{j \neq 1} \hat{\mu}_j)$$

$$\leq P_{\pi_\nu}(\hat{\mu}_1 > \hat{\mu}_i) \text{ for some } i \neq 1$$

But as  $\mu_1$  and  $\mu_i$  are gaussian and hence subgaussian  
by appropriately changing the means we know that

$$P_{\pi_\nu}(\hat{\mu}_1 > \hat{\mu}_i) \leq \exp\left(-\frac{n}{K} \frac{\Delta^2}{4}\right) \quad (1)$$

$$\begin{aligned} \dots R_n^S(U.E) &= (1 - P_{\pi_\nu}(1)) \Delta \\ &\geq \left(1 - \exp\left(-\frac{n}{K} \frac{\Delta^2}{4}\right)\right) \Delta \quad [\text{from (1)}] \end{aligned}$$

$$\text{let } \delta = \sqrt{\frac{4K \log K}{n}}$$

$$\therefore R_n^S(U.E) \geq \left(1 - \frac{1}{K}\right) \sqrt{\frac{4K \log K}{n}}$$

Now  $K \in [2, 3, 4, \dots, \infty)$

$$\therefore 1 - \frac{1}{K} \geq 1 - \frac{1}{2} = \frac{1}{2}$$

$$\therefore R_n^s(CU-B, U) \geq \left(1 - \frac{1}{K}\right) \sqrt{\frac{4K \log K}{n}}$$

$$\geq \frac{1}{2} \sqrt{\frac{4K \log K}{n}}$$

$$\geq \underline{\underline{\sqrt{\frac{K \log K}{n}}}}$$