

## Assignment 6

Q.1: Given:  $X_1$  and  $X_2$  are independent and  $\sigma_1$  and  $\sigma_2$  subgaussian respectively.

To prove:  $X_1 + X_2$  is  $\sqrt{\sigma_1^2 + \sigma_2^2}$

Proof:

From the properties of subgaussian variables we know:-

$$\forall t \in \mathbb{R}, \mathbb{E}[\exp(tX)] \leq \exp\left(\frac{\sigma^2 t^2}{2}\right)$$

if  $X$  is  $\sigma$ -subgaussian

Thus,

$$\begin{aligned} & \mathbb{E}[\exp(t(X_1 + X_2))] \\ &= \mathbb{E}[\exp(X_1 t), \exp(X_2 t)] \quad \left[ \begin{array}{l} \text{Given: } X_1 \& X_2 \\ \text{are independent} \end{array} \right] \\ &\leq \exp\left(\frac{\sigma_1^2 t^2}{2}\right) \cdot \exp\left(\frac{\sigma_2^2 t^2}{2}\right) \\ &\leq \exp\left(\frac{t^2}{2} (\sigma_1^2 + \sigma_2^2)\right) \end{aligned}$$

$$\leq \exp \left( \frac{t^2}{2} (\sqrt{\sigma_1^2 + \sigma_2^2})^2 \right)$$

Thus :

$X_1 + X_2$  is  $\sqrt{\sigma_1^2 + \sigma_2^2}$  Subgaussian.

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Q.2: To prove:

$$P \left( \sum_{t=1}^n (X_t - \mathbb{E}[X_t]) \geq \epsilon \right) \leq \exp \left( - \frac{2\epsilon^2}{\sum_{t=1}^n (b_t - a_t)^2} \right)$$

Proof:

For all  $1 \leq i \leq n$ , define a new variable  $Z_i$  as the difference between  $X_i$  and its expectation

$$\therefore Z_i = X_i - \mathbb{E}[X_i]$$

This implies:  $\mathbb{E}[Z_i] = 0$

The domain of  $Z_i$  may be bounded inside  $[a_i - \mathbb{E}(X_i), b_i - \mathbb{E}(X_i)]$ . The length of the interval must still have length  $b_i - a_i$ .

Let  $s$  be some positive value.

Then,

$$P\left(\sum_{i=1}^n z_i \geq t\right) = P\left(\exp\left(s \sum_{i=1}^n z_i\right) \geq e^{st}\right)$$

$$\leq \frac{\mathbb{E}\left[\prod_{i=1}^n e^{sz_i}\right]}{e^{st}} \quad \left[\text{Chernoff inequality}\right]$$

————— ①

Also

$$\frac{\mathbb{E}\left[\prod_{i=1}^n e^{sz_i}\right]}{e^{st}} = \frac{\prod_{i=1}^n \mathbb{E}[e^{sz_i}]}{e^{st}} \quad \left[\text{By independence of } z_i\right]$$

$$\leq e^{-st} \prod_{i=1}^n \exp\left(\frac{s^2(b_i - a_i)^2}{8}\right) \quad \left[\text{Hoeffding lemma}\right]$$

$$= \exp\left(-st + \frac{s^2}{8} \sum_{i=1}^n (b_i - a_i)^2\right) \quad \text{————— ②}$$

Let,  $s = \frac{4t}{\sum_{i=1}^n (b_i - a_i)^2}$  and  $t = \epsilon$

Substituting in eq ① we get:

$$P\left(\sum_{i=1}^n z_i \geq \epsilon\right) = P\left(\left(\frac{1}{n} \sum_{i=1}^n x_i - \frac{1}{n} \sum_{i=1}^n \mathbb{E}[x_i]\right) \geq \epsilon\right)$$

$$\leq \exp\left[-\frac{4\epsilon}{\sum_{i=1}^n (b_i - a_i)^2} \epsilon + \frac{1}{8} \left(\frac{4\epsilon}{\sum_{i=1}^n (b_i - a_i)^2}\right)^2 \sum_{i=1}^n (b_i - a_i)^2\right]$$

$$= \exp \left[ \frac{-4\epsilon^2}{\sum_{i=1}^n (b_i - a_i)^2} + \frac{2\epsilon^2}{\left[ \sum_{i=1}^n (b_i - a_i)^2 \right]^2} \sum_{i=1}^n (b_i - a_i)^2 \right]$$

$$= \exp \left[ \frac{-4\epsilon^2 + 2\epsilon^2}{\sum_{i=1}^n (b_i - a_i)^2} \right]$$

$$= \exp \left[ \frac{-2\epsilon^2}{\sum_{i=1}^n (b_i - a_i)^2} \right]$$

$$\Rightarrow P \left[ \sum_{i=1}^n (x_i - E[x_i]) \geq \epsilon \right] \leq$$

$$\exp \left[ \frac{-2\epsilon^2}{\sum_{i=1}^n (b_i - a_i)^2} \right]$$


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Q.3: Given: Regret for an optimally tuned  
 ETC algorithm for subgaussian 2-armed bandit  
 with means  $\mu_1, \mu_2 \in \mathbb{R}$  and  $\Delta = |\mu_1 - \mu_2|$  is  
 $\frac{1}{\Delta} \left( \pi \wedge \Delta + 4 \left( 1 + \max \left\{ 0, \log \left( \frac{T \Delta^2}{\pi} \right) \right\} \right) \right)$

$$R_T \leq \min \left\{ T\Delta, \Delta + \frac{4}{\Delta} \left( 1 + \max \left\{ 0, \log \left( \frac{T\Delta^2}{4} \right) \right\} \right) \right\}$$

Show that:  $R_T \leq \Delta + C\sqrt{T}$  where  $C > 0$   
is a universal constant

Proof:

$$R_T \leq \underbrace{\min \left\{ T\Delta, \Delta + \frac{4}{\Delta} \left( 1 + \max \left\{ 0, \log \left( \frac{T\Delta^2}{4} \right) \right\} \right) \right\}}_{\text{2nd term}}$$

↓  
first term

Scenario 1:

If  $\Delta \leq \frac{1}{\sqrt{T}}$ , the 2nd term will become too large.

Hence taking the minimum term, we have:

$$\begin{aligned} R_T &\leq T\Delta \\ &\leq \sqrt{T} \end{aligned}$$

Scenario 2:

If  $\Delta > \frac{1}{\sqrt{T}}$ , first term would be large,

Hence taking the minimum term, we have:

$$R_T \leq \Delta + \frac{4}{\Delta} \left( 1 + \max \left\{ 0, \log \left( \frac{T\Delta^2}{4} \right) \right\} \right)$$

$$\leq \Delta + 4\sqrt{T} + \max_{x>0} \frac{1}{x} \log_+ \left( \frac{T x^2}{4} \right) \quad \text{--- (1)}$$

[with  $\Delta = 1$ ]

$$\text{Let } F = \frac{1}{x} \log\left(\frac{Tx^2}{4}\right) \quad \text{--- (2)} \quad \text{L} \quad \sqrt{T}$$

Differentiating and setting  $F' = 0$ , we have:

$$\frac{dF}{dx} = \frac{1}{x} \frac{d}{dx} \left( \log\left(\frac{Tx^2}{4}\right) \right) + \log\left(\frac{Tx^2}{4}\right) \frac{d}{dx} \left( \frac{1}{x} \right)$$

$$\Rightarrow 0 = \frac{2}{x^2} - \frac{1}{x^2} \log\left(\frac{Tx^2}{4}\right)$$

$$\Rightarrow 0 = \frac{1}{x^2} \left[ 2 - \log\left(\frac{Tx^2}{4}\right) \right]$$

$$\Rightarrow 2 - \log\left(\frac{Tx^2}{4}\right) = 0$$

$$\Rightarrow x^2 = \frac{4}{T} e^2$$

$$\Rightarrow x = \frac{2e}{\sqrt{T}}$$

Substituting value of  $x$  in (2)

$$\begin{aligned} \frac{1}{x} \log\left(\frac{Tx^2}{4}\right) &= \frac{\sqrt{T}}{2e} \log(e^2) \\ &= \sqrt{T} e^{-1} \end{aligned}$$

Substituting this in eq (1), we have

$$R_T \leq \Delta + 4\sqrt{T} + e^{-1}\sqrt{T}$$

$$\leq \Delta + (4 + e^{-1})\sqrt{T}$$

$$\underline{R_T \leq \Delta + C\sqrt{T}} \quad \text{where } C = 4 + e^{-1} \text{ is}$$

always positive

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