

#1

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MAS 480 HW3

Concentration Inequalities

$$\psi_Y(\theta) = \log \mathbb{E}[e^{\theta Y}] \text{ by definition.}$$

$$\Rightarrow e^{\psi_Y(\theta)} = \mathbb{E}[e^{\theta Y}] = \int f_Y(y) e^{\theta y} dy$$

$$\frac{\partial}{\partial \theta} \Rightarrow \psi'_Y(\theta) e^{\psi_Y(\theta)} = \int y e^{\theta y} f_Y(y) dy = \mathbb{E}[Y e^{\theta Y}]$$

$$\frac{\partial}{\partial \theta} \Rightarrow (\psi''_Y(\theta) + (\psi'_Y(\theta))^2) e^{\psi_Y(\theta)} = \int y^2 e^{\theta y} f_Y(y) dy = \mathbb{E}[Y^2 e^{\theta Y}]$$

$$\begin{aligned} \Rightarrow \psi''_Y(\theta) &= e^{-\psi_Y(\theta)} \mathbb{E}[Y^2 e^{\theta Y}] - (\psi'_Y(\theta))^2 \\ &= \frac{\mathbb{E}[Y^2 e^{\theta Y}]}{\mathbb{E}[e^{\theta Y}]} - \frac{(\mathbb{E}[Y e^{\theta Y}])^2}{(\mathbb{E}[e^{\theta Y}])^2} \quad (\because e^{\psi_Y(\theta)} = \mathbb{E}[e^{\theta Y}]) \end{aligned} \quad \dots (1)$$

$$\text{Var}(Y) \leq \mathbb{E}[(Y - \frac{b+a}{2})^2] \leq \frac{1}{4}(b-a)^2. \quad (\because |Y - \frac{b+a}{2}| \leq \frac{b-a}{2} \text{ always}) \dots (2)$$

$$\hookrightarrow \because \frac{\partial}{\partial \mu} \mathbb{E}[(Y - \mu)^2] = 2\mu - 2\mathbb{E}[Y] = 0 \text{ gives } \mu = \mathbb{E}[Y].$$

Thus, $\mathbb{E}[(Y - \mu)^2]$ minimized at $\mu = \mathbb{E}[Y]$.

$$\frac{\mathbb{E}[Y^2 e^{\theta Y}]}{\mathbb{E}[e^{\theta Y}]} = \int y^2 e^{\theta y} \cdot \frac{1}{\mathbb{E}[e^{\theta Y}]} f_Y(y) dy$$

$$\text{Let } L(y) = \frac{e^{\theta y}}{\mathbb{E}[e^{\theta Y}]} f_Y(y) dy, \quad z = \int L(y) dy.$$

$$\text{Then, } \int dz = \int L(y) dy = \int \frac{e^{\theta y}}{\mathbb{E}[e^{\theta Y}]} f_Y(y) dy = \frac{\mathbb{E}[e^{\theta Y}]}{\mathbb{E}[e^{\theta Y}]} = 1.$$

Thus, z is a probability measure.

$$\frac{\mathbb{E}[Y^2 e^{\theta Y}]}{\mathbb{E}[e^{\theta Y}]} = \int y^2 L(y) dy = \int y^2 dz = \mathbb{E}_z[Y^2] \dots (3)$$

$$\frac{\mathbb{E}[Y e^{\theta Y}]}{\mathbb{E}[e^{\theta Y}]} = \int y L(y) dy = \int y dz = \mathbb{E}_z[Y] \dots (4)$$

$$\begin{aligned} \text{Thus, } \psi_Y''(\theta) &\stackrel{(1)(3)(4)}{=} \mathbb{E}_z[Y^2] - \mathbb{E}_z[Y]^2 \\ &\stackrel{(2)}{=} \text{Var}_z[Y] \stackrel{\text{Now,}}{\leq} \frac{1}{4}(b-a)^2. \end{aligned}$$

$$\psi_Y'(\theta) = \frac{\mathbb{E}[Y]}{\mathbb{E}[1]} = 0. \quad \psi_Y(\theta) = \log \mathbb{E}[1] = \log 1 = 0.$$

$$\psi_Y(\theta) = \int_0^\theta \psi_Y'(t) dt = \int_0^\theta \int_0^t \psi_Y''(u) du dt$$

$$(\theta \geq 0) \Rightarrow \psi_Y(\theta) \leq \int_0^\theta \int_0^t \frac{1}{4}(b-a)^2 du dt = \int_0^\theta \frac{1}{4}(b-a)^2 t dt = \frac{1}{8}(b-a)^2 \theta^2$$

$$(\theta < 0) \Rightarrow \psi_Y(\theta) = \int_0^\theta \int_0^t \psi_Y''(u) du dt = \int_0^\theta - \int_t^0 \psi_Y''(u) du dt$$

$$= \int_0^0 \int_t^0 \varphi_Y''(u) du dt \leq \int_0^0 \int_t^0 \frac{1}{4}(b-a)^2 du dt$$

$$= \int_0^0 \frac{1}{4}(b-a)^2 (-t) dt = -\frac{1}{8}(b-a)^2 0^2 - (-\frac{1}{8}(b-a)^2 \theta^2) \\ = \frac{1}{8}(b-a)^2 \theta^2.$$

Thus, $\varphi_Y(\theta) \leq \frac{1}{8}(b-a)^2 \theta^2 \quad \forall \theta \in \mathbb{R}$, proving Y is sub-Gaussian with parameter $\frac{b-a}{2}$.

#2 By Theorem 2,

$$S_n = X_1 + \dots + X_n, \quad X_i \sim \text{Bernoulli}(p) \Rightarrow 0 \leq X_i \leq 1.$$

$$P(|S_n - np| \geq 0.1np) \leq e^{-\left(\frac{0.2np}{\sum (b_i - a_i)^2}\right)} = e^{-0.2p} \dots \textcircled{1}$$

By Theorem 3,

$$\begin{aligned} P(S_n - np \geq 0.1np) &= P(S_n \geq (1+0.1)np) \\ &\leq \left(\frac{e^{0.1}}{1.1^{1.1}}\right)^{np} \dots \textcircled{2} \leq e^{-\frac{1}{3}(0.1)^2 np} \dots \textcircled{4} \end{aligned}$$

$$\begin{aligned} P(S_n - np \leq -0.1np) &= P(S_n \leq (1-0.1)np) \\ &\leq \left(\frac{e^{-0.1}}{0.9^{0.9}}\right)^{np} \dots \textcircled{3} \leq e^{-\frac{1}{2}(0.1)^2 np} \dots \textcircled{5} \end{aligned}$$

(n, p)	theorem 2		theorem 3		Program
	①	② + ③	④ + ⑤		
(10, 0.2)	0.96079	1.98007	1.98341	0.69801	
(100, 0.2)	0.96079	1.80938	1.84034	0.62315	
(1000, 0.2)	0.96079	0.73494	0.88130	0.11405	
(10000, 0.2)	0.96079	0.00009	0.00132	0.00001	

For $n=10, 100$, theorem 2 gives better bound than theorem 3:

$$\text{Program} < \textcircled{1} < \textcircled{2} + \textcircled{3} < \textcircled{4} + \textcircled{5}$$

For $n=1000, 10000$, theorem 3 gives better bound than theorem 2:

$$\text{Program} < \textcircled{2} + \textcircled{3} < \textcircled{4} + \textcircled{5} < \textcircled{1}$$

this is because ① is constant wrt n while ②+③, ④+⑤ are n -dependent. Also, tail bound by software is of course smallest among all.