

## PROBLEMS - SET 5

**Problem 1.** Let  $(X_n)$  be a i.i.d. sequence of  $\text{Bin}(1, p)$ . Prove the following Chernoff bounds:

$$P(\bar{X}_n \geq (1 + \delta)p) \leq e^{-np \frac{\delta^2}{2+\delta}},$$

and

$$P(\bar{X}_n \leq (1 - \delta)p) \leq e^{-np \frac{\delta^2}{2}}.$$

**Problem 2.** Let  $(X_n)$  be a i.i.d. sequence of  $N(0, 1)$ .

(a) Prove the following Chernoff bound:

$$P(\bar{X}_n \geq \varepsilon) \leq e^{-n \frac{\varepsilon^2}{2}}$$

(b) Obtain a sharper upper tail estimate, proving first the inequality

$$P(X_1 \geq x) \leq \frac{1}{x} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

and deriving then that

$$P(\bar{X}_n \geq \varepsilon) \leq \frac{1}{\varepsilon \sqrt{n}} \frac{1}{\sqrt{2\pi}} e^{-n \frac{\varepsilon^2}{2}}$$

(c) Prove the further inequality

$$P(X_1 \geq x) > \left( \frac{1}{x} - \frac{1}{x^3} \right) \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

**Problem 3.** Let  $(X_n)$  be a i.i.d. sequence of  $\text{Pois}(\lambda)$ . Prove the following Chernoff bounds:

$$P(\bar{X}_n \geq \lambda(1 + \varepsilon)) \leq e^{-n\lambda a(\varepsilon)},$$

and, for  $0 < \varepsilon < 1$

$$P(\bar{X}_n \leq \lambda(1 - \varepsilon)) \leq e^{-n\lambda b(\varepsilon)},$$

where

$$a(\varepsilon) = \frac{\varepsilon^2}{2 + \varepsilon}$$

and

$$b(\varepsilon) = \frac{\varepsilon^2}{2}.$$

*Hint:* prove (and use) the following inequalities

$$(1 + \varepsilon) \log(1 + \varepsilon) - \varepsilon := f(\varepsilon) \geq a(\varepsilon) = \frac{\varepsilon^2}{2 + \varepsilon}, \quad \forall \varepsilon > 0, \quad (0.1)$$

$$\varepsilon + (1 - \varepsilon) \log(1 - \varepsilon) := k(\varepsilon) \geq b(\varepsilon) = \frac{\varepsilon^2}{2} \quad \forall 0 < \varepsilon < 1. \quad (0.2)$$

Note that if  $F, G$  are  $C^2$  functions in  $[0, a]$ ,  $F(0) = G(0)$ ,  $F'(0) = G'(0)$  and  $F''(x) \geq G''(x)$  for all  $x \in [0, a]$ , then  $F(x) \geq G(x)$  for  $x \in [0, a]$ .

**Problem 4.** Let  $(X_n)$  be a i.i.d. sequence with distribution  $\text{Unif}(-1, 1)$

(a) Show that  $m_{X_n}(t) = \frac{\sinh(t)}{t}$ .

(b) Show that there exists  $k > 0$  such that for all  $0 \leq t \leq k$

$$\frac{\sinh(t)}{t} \leq 1 + \frac{t^2}{2}$$

(numerically  $k \simeq 4.75$ .)

(c) Prove the Chernoff bound: for all  $0 < \varepsilon \leq k$

$$\mathbb{P}(\bar{X}_n \geq \varepsilon) \leq e^{-n \frac{\varepsilon^2}{2}}.$$

(d) Without any further computation, explain why the bound for the lower tail

$$\mathbb{P}(\bar{X}_n \leq -\varepsilon) \leq e^{-n \frac{\varepsilon^2}{2}}.$$

also holds for all  $0 < \varepsilon \leq k$ .

**Problem 5.** Let  $(X_n)$  be a i.i.d. sequence such that

$$\varphi_{X_n}(u) = \frac{1}{1 + u^2}.$$

For every  $\varepsilon > 0$  find  $a(\varepsilon) > 0$  (no “nice” form is necessary) such that

$$\mathbb{P}(\bar{X}_n \geq \mathbb{E}(X_1) + \varepsilon) \leq e^{-na(\varepsilon)}$$

$$\mathbb{P}(\bar{X}_n \leq \mathbb{E}(X_1) - \varepsilon) \leq e^{-na(\varepsilon)}.$$

Then extend to the case in which

$$\varphi_{X_n}(u) = \frac{e^{iua}}{1 + u^2}.$$

(Try to make no further calculations).

**Problem 6.** Let  $Y_1, Y_2, \dots, Y_n, Z_1, Z_2, \dots, Z_n$  be independent random variables such that  $Y_i \sim \text{Exp}(1)$  and  $Z_i \sim N(0, 1)$ . Set  $X_i := Y_i + Z_i$ .

(a) Compute mean, variance and moment generating function of  $X_i$ .

(b) For  $0 < \varepsilon < 1$ , determine  $a(\varepsilon) > 0$  such that the following *lower tail Chernoff bound* holds:

$$\mathbb{P}(\bar{X}_n \leq 1 - \varepsilon) \leq e^{-na(\varepsilon)}.$$

*Hint:* use the inequality  $\log(1+t) \geq t - \frac{1}{2}t^2$  for  $t \geq 0$ .

**Problem 7.** (a) Let  $X_1, X_2, \dots, X_n$  be i.i.d random variables with distribution  $N(0, \sigma^2)$ . Show that for every  $\varepsilon > 0$

$$P(\bar{X}_n > \varepsilon) \leq e^{-n\varepsilon^2/2\sigma^2}.$$

(b) A random variable  $X$  with  $E(X) = 0$  is said to be *Subgaussian* for the parameter  $\sigma > 0$  if its moment generating function  $m_X(t)$  is such that

$$m_X(t) \leq e^{t^2\sigma^2/2}$$

for all  $t \in \mathbb{R}$ . Show that the inequality in (a) holds if  $X_1, X_2, \dots, X_n$  are i.i.d random variables, and Subgaussian for the parameter  $\sigma > 0$ .