

**Homework 6**

Due Tuesday, March 12

1. Let  $Y$  be a random variable and let  $U = \ln Y$ , with  $E(U) = 0$ . Show that  $E(Y) \geq 1$ .
2. Show that  $E(S) \leq \sigma$ .
3. (a) Show that  $|E(Y)| \leq E(|Y|)$ . (HINT: You may assume that the absolute value function is convex).  
 (b) Use Cauchy-Schwarz inequality

$$E(|XY|) \leq \sqrt{E(X^2)E(Y^2)}$$

and part (a) to show that

$$|\text{Cov}(Y_1, Y_2)| \leq \sqrt{\text{Var}(Y_1)}\sqrt{\text{Var}(Y_2)}$$

- (c) Use part (b) to conclude that the correlation  $\rho$  between two random variables  $Y_1$  and  $Y_2$  is always between -1 and 1, i.e.,  $-1 \leq \rho \leq 1$ .
4. Assume that  $\text{Var}(S^2) \rightarrow 0$  as  $n \rightarrow \infty$ . Show that for any  $\epsilon > 0$ ,  $P(|S^2 - \sigma^2| \geq \epsilon) \rightarrow 0$  (and hence  $S_n^2 \rightarrow \sigma^2$  in probability) as  $n \rightarrow \infty$ .
5. Let  $X_1, \dots, X_n \sim \text{iid Binomial}(1, p)$ , and let

$$Y_n = \frac{1}{n} \sum_{i=1}^n X_i^2$$

- (a) Show that  $E(X_i^k) = p$  for any integer  $k$ .
- (b) Show that  $E(Y_n) = p$ .
- (c) Show that  $Y_n \rightarrow p$  in probability.
6. Suppose that  $Y \sim \text{Poisson}(\lambda)$ .
  - (a) Let  $U = (Y - \lambda)/\sqrt{\lambda}$ . Show that  $M_U(t) = \exp(\lambda e^{t/\sqrt{\lambda}} - t\sqrt{\lambda} - \lambda)$ .
  - (b) Show that  $M_U(t) \rightarrow \exp(t^2/2)$  as  $\lambda \rightarrow \infty$  (HINT: Use  $e^a = 1 + a + a^2/2! + a^3/3! + \dots$  to expand  $e^{t/\sqrt{\lambda}}$ ).
7. Let  $Y_1, Y_2, \dots$  be a sequence of random variables with probability function

$$P(Y_n = 0) = 1 - \frac{1}{n} \quad \text{and} \quad P(Y_n = n^2) = \frac{1}{n}$$

Show that  $\lim_{n \rightarrow \infty} E(Y_n) = \infty$ .

8. **(5880\*)** Let  $X_1, \dots, X_n \sim \text{iid Uniform}(0, \theta)$ . Let  $Y_n = n(\theta - X_{(n)})$ , where  $X_{(n)} = \max\{X_1, \dots, X_n\}$ . Find the cdf of  $Y_n$ , and show that  $Y_n \rightarrow Y$  in distribution, where  $Y \sim \text{Exponential}(\theta)$ .
9. **(5880\*)** Prove the central limit theorem (CLT). Specifically, if  $Y_1, \dots, Y_n$  are iid with  $E(Y_i) = \mu$  and  $\text{Var}(Y_i) = \sigma^2 < \infty$ , then let

$$U_n = \frac{\bar{Y} - \mu}{\sigma/\sqrt{n}} = \frac{\sum_{i=1}^n Y_i - n\mu}{\sqrt{n}\sigma}$$

and show that

$$U_n \rightarrow Z \sim N(0, 1)$$

in distribution (or equivalently,  $M_{U_n}(t) \rightarrow M_Z(t)$ ) as  $n \rightarrow \infty$ , by following the steps below.

- (a) We may assume that  $\mu = E(Y_i) = 0$ , so that  $U_n = \frac{\sum_{i=1}^n Y_i}{\sqrt{n}\sigma}$ . Show that

$$M_{U_n}(t) = \left( E \left[ \exp \left( \frac{t}{\sqrt{n}\sigma} Y_1 \right) \right] \right)^n = \left( M_{Y_1} \left( \frac{t}{\sqrt{n}\sigma} \right) \right)^n$$

- (b) Using the fact that  $e^a = 1 + a + a^2/2! + \dots$ , show that

$$M_{Y_1} \left( \frac{t}{\sqrt{n}\sigma} \right) = E \left[ \exp \left( \frac{t}{\sqrt{n}\sigma} Y_1 \right) \right] = 1 + \frac{t^2}{2n} + \dots$$

- (c) Assume that the “...” terms are small, so that  $M_{Y_1} \left( \frac{t}{\sqrt{n}\sigma} \right) = 1 + \frac{t^2}{2n}$ . Show that

$$M_{U_n}(t) \rightarrow e^{t^2/2} = M_Z(t) \quad (\text{where } M_Z(t) \text{ is the mgf of } Z \sim N(0, 1))$$

as  $n \rightarrow \infty$ , and thus proving that  $U_n \rightarrow Z$  in distribution.

10. **(5880\*)** Suppose that  $Y_1, \dots, Y_n$  are iid random variables with  $E(Y_i) = \mu \neq 0$  and  $\text{Var}(Y_i) = \sigma^2 < \infty$ . Find the limiting distribution of  $\sqrt{n}(1/\bar{Y} - 1/\mu)$ , i.e., find “?” if  $\sqrt{n}(1/\bar{Y} - 1/\mu) \xrightarrow{d} ?$
11. **(5880\*)** If  $Y_1, \dots, Y_n \sim \text{iid Exponential}(\theta)$  with  $\theta > 0$ , find the limiting distribution of  $\sqrt{n}(\bar{Y}^2 - \theta^2)$ .
12. **(5880\*)** Suppose that  $X_1, \dots, X_n$  are iid Bernoulli( $p$ ), and let  $Y_n = \bar{X}$ .
- (a) Show that  $\sqrt{n}(Y_n - p) \xrightarrow{d} N[0, p(1-p)]$ .
- (b) If  $p \neq 1/2$ , find the limiting distribution of  $\sqrt{n}[Y_n(1 - Y_n) - p(1-p)]$ .
- (c) If  $p = 1/2$ , find the limiting distribution of  $n[Y_n(1 - Y_n) - 1/4]$ .