

**COM 5110**  
**Random Processes for Communications - Spring 2018**  
**Homework 2**  
**(Due date: April 26, 2018)**

1. (10 points)

- (a) Find the moment generating function (MGF) of a bilateral exponential distribution defined by

$$f_X(x) = \frac{\mu}{2} e^{-\mu|x|}, \quad -\infty < x < \infty,$$

and under what condition does the MGF exist?

- (b) Find the MGF of a triangular distribution defined by

$$f_X(x) = \frac{1}{a} \left(1 - \frac{|x|}{a}\right), \quad |x| < a.$$

2. (10 points)

- (a) Show that when the degree of freedom in the numerator of  $F$ -distribution is equal to one, the  $F$ -distribution and  $t$ -distribution have a relationship.
- (b) Show that the median of the log-normal distribution  $F_X(x)$  is given by:

$$e^{\mu_Y} = \frac{\mu_X}{\sqrt{1 + \frac{\sigma_X^2}{\mu_X^2}}},$$

where  $Y = \ln X$  has the normal distribution  $N(\mu_Y, \sigma_Y^2)$ .

- (c) Show that the mode of the log-normal distribution  $F_X(x)$  is given by:

$$x_{mode} = e^{\mu_Y - \sigma_Y^2} = \frac{\mu_X}{\left(1 + \frac{\sigma_X^2}{\mu_X^2}\right)^{\frac{3}{2}}}.$$

3. (15 points)

$$\sum_{i=1}^n f_i \log g_i \leq \sum_{i=1}^n f_i \log f_i$$

where  $\mathbf{f} = [f_1, f_2, \dots, f_n]$  and  $\mathbf{g} = [g_1, g_2, \dots, g_n]$  are two probability distributions.

- (a) Use Jensen's inequality.
- (b) Let  $\mathbf{f}$  be fixed. Prove the inequality by solving the following maximization problem

$$\max_{\mathbf{g}} \left\{ \sum_{i=1}^n f_i \log g_i \mid \sum_{i=1}^n g_i = 1 \right\}$$

4. (5 points) Define a random variable  $X_k$  by

$$X_k = c + N_k, \quad k = 1, 2, \dots,$$

where  $c$  is a constant and the  $N_k$  are i.i.d RVs with zero mean. Show that the sequence of sample averages  $\{\bar{X}_n\}$

$$\bar{X}_n = \frac{1}{n} \sum_{k=1}^n X_k, \quad n = 1, 2, \dots,$$

converges in probability to  $c$ .

5. (10 points) Consider the experiment of tossing a fair coin  $n$  times. Let  $T_n$  be the total number of “head (=1)”:

$$T_n = B_1 + B_2 + \dots + B_n,$$

where  $B_i$  are independent binary variables, with  $P[B_i = 1] = P[B_i = 0] = \frac{1}{2}$  for all  $i = 1, 2, \dots, n$ . Assume the threshold value  $b > E[T_n] = \frac{n}{2}$ ; i.e.,  $\frac{1}{2} < \beta < 1$ , where  $\beta = \frac{b}{n}$ .

- (a) Use Markov’s inequality and find an upper bound on the probability that  $T_n$  exceeds  $b = \beta n$ . Determine the upper bound for cases  $n = 100$  and  $n = 1000$  with  $\beta = 0.8$  for both cases.
- (b) Repeat (a) by using Chebyshev’s inequality.
6. (10 points)

- (a) Let  $X$  be a random variable with  $E[X] = 0$  and  $\text{Var}[X] = \sigma^2$ . For any  $a > 0$ , show that

$$P[X \geq a] \leq \frac{\sigma^2}{\sigma^2 + a^2}.$$

- (b) If  $E[X] = \mu$  and  $\text{Var}[X] = \sigma^2$ . Show that

$$P[X \geq \mu + a] \leq \frac{\sigma^2}{\sigma^2 + a^2}.$$

$$P[X \leq \mu - a] \leq \frac{\sigma^2}{\sigma^2 + a^2}.$$

7. (10 points) Let  $S$  be a sum of  $n$  independent standard normal variables  $U_i$ ; i.e.,  $U_i \sim N(0, 1)$ :

$$S = U_1 + U_2 + \dots + U_n.$$

Use Chernoff’s bound to show that  $P[S \geq n\beta] \leq e^{-\frac{n\beta^2}{2}}$ .

8. (10 points) Let  $X_n$  be a sequence of i.i.d RVs with common distribution function  $F_X(x)$  given by

$$F_X(x) = \begin{cases} 0, & \text{if } x \leq 0, \\ x, & \text{if } 0 < x \leq 1, \\ 1, & \text{if } x > 1. \end{cases}$$

Now, we define two sequences  $Y_n$  and  $Z_n$  by

$$Y_n = \max\{X_1, X_2, \dots, X_n\}$$

and

$$Z_n = n(1 - Y_n).$$

Show that  $Z_n$  converges in distribution to a random variable  $Z$  with distribution

$$F_Z(z) = 1 - e^{-z}.$$

9. (10 points) Consider the following example to illustrate that convergence in  $r$ th mean does not imply almost surely convergence. Let  $X_n$  be a sequence of independent RVs defined by

$$X_n = \begin{cases} n^{\frac{1}{2r}}, & \text{with probability } \frac{1}{n}, \\ 0, & \text{with probability } 1 - \frac{1}{n}. \end{cases}$$

and let  $X$  be the degenerate RV

$$X = 0, \text{ with probability one.}$$

- (a) Show that  $X_n$  converges in the  $r$ th mean to  $X = 0$ .  
(b) Show that for  $\epsilon > 0$  and for arbitrary integer  $m > 0$ ,

$$P\left[\bigcap_{n=m}^{\infty} \{\omega : X_n(\omega) < \epsilon\}\right] = 0.$$

- (c) Show that  $X_n$  does not converge almost surely to  $X = 0$ .

10. (10 points) Prove the central limit theory (CLT) for i.i.d RVs (Theorem 11.22 in the textbook) by using the moment generating function (MGF) of the normalized average  $Z_n$ .