## Statistics 330 - Final Exam

Total marks: 75.

- 1. Let  $X_1$  and  $X_2$  be independent poisson random variables with  $E(X_1) = \lambda$  and  $E(X_2) = 5\lambda$ .
  - (a) Find  $P(X_1 + X_2 = m)$  for any positive integer m. [4]
  - (b) Assuming that for  $Y = [X_1|X_1 + X_2 = m]$ ,  $P(Y = k) = P(X_1 = k|X_1 + X_2 = m)$ . Identify the distribution of Y. [4]
- 2. Let X be a continuous random variable with the cdf given by

$$F_X(x) = \begin{cases} 1 & 1 \le x \\ x^2 & 0 \le x < 1 \\ 0 & x < 0, \end{cases}$$

- (a) find the pdf of X and show that it is a pdf. [3]
- (b) Identify the pdf of X obtained in (a). [2]
- 3. Let X be a continuous random variable with the pdf given by

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & x > 0 \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Find the cdf of X. [2]
- (b) Find the pdf of  $Y = F_X(u)$ , where  $F_X(\cdot)$  is the cdf of X obtained in (a). [3]
- 4. Let X be a discrete random variable with the pmf P(X = x) > 0 on the set of non-negative integers  $\{0, 1, 2, ....\}$ .
  - (a) Show that the cdf  $F_X(x)$  is a non-decreasing function. [5]
  - (b) Prove the following identity. [4]

$$E(X) = \sum_{n=0}^{\infty} P(X > n).$$

5. Let  $X_1, X_2, X_3$  be independent and identically distributed N(0, 1) random variables. Then define  $Y_1, Y_2$  and  $Y_3$  as

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$$X_1 = Y_1 \cos(Y_2) \sin(Y_3)$$

$$X_2 = Y_1 \sin(Y_2) \sin(Y_3)$$

$$X_3 = Y_1 \cos(Y_3)$$

where  $0 \le Y_1 < \infty, 0 \le Y_2 < 2\pi$  and  $0 \le Y_3 \le \pi$ . Then, show that  $Y_1, Y_2$  and  $Y_3$  are mutually independent. [10]

6. Let  $X_1, ..., X_{50}$  be a random sample of size 50 from N(0, 1). Assume that the moment generating function for a random variable W with distribution  $N(\mu, \sigma^2)$  is given by

$$(1) M_W(t) = e^{\mu t + \sigma^2 \frac{t^2}{2}}.$$

Then, (a) without finding the pdf, find the mgf of  $Z = 5X_1 + 3X_2 - \frac{1}{9}X_{10}$  and (b) compare it with the mgf given in equation (1) to identify the distribution of Z. [5+3]

7. Let  $X_1,...,X_n$  be a random sample from truncated exponential distribution with the pdf

$$f_X(x) = \begin{cases} e^{-(x-\theta)} & x > \theta \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Find the maximum likelihood estimator of  $\theta$  (say  $T_1(\underline{X})$ ). [5]
- (b) Find a method of moment estimator of  $\theta$  (say  $T_2(\underline{X})$ ). [3]
- (c) Evaluate  $B_{T_1}(\theta)$ ,  $B_{T_2}(\theta)$  and  $var(T_1(\underline{X}))$  where,  $B_T(\theta) = E(T) \theta$ . [6]
- 8. Let  $X_1,...,X_n$  be a random sample from  $\Gamma(\alpha,\lambda)$ , where  $\alpha$  is known.
  - (a) Find the maximum likelihood estimator of  $\theta = \frac{1}{\lambda}$  (say  $T(\underline{X})$ ). [3]
  - (b) Find the exact distribution of  $T(\underline{X})$ . [3]
  - (c) Assuming n is large, find the approximate distribution of  $T(\underline{X})$ . [10]

## Formula Sheet

1. Let  $X \sim N(0,1)$ , then

$$f_X(x) = \begin{cases} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} & -\infty < x < \infty \\ 0 & otherwise. \end{cases}$$

2. Let  $X \sim Beta(\alpha, \beta)$ , then

$$f_X(x) = \begin{cases} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} & 0 < x < 1\\ 0 & otherwise. \end{cases}$$

3. Let  $X \sim \Gamma(\alpha, \lambda)$ , then

$$f_X(x) = \begin{cases} \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha - 1} e^{-\lambda x} & 0 < x < \infty \\ 0 & otherwise. \end{cases}$$

4. Let  $X \sim Poi(\lambda)$ , then

$$P(X=x) = \frac{\lambda^x}{x!}e^{-\lambda}$$
 for  $x=0,1,2,...$  and zero elsewhere.

5. Let  $X \sim Bin(n, p)$ , then

$$P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$$
 for  $k = 0, 1, 2, ..., n$  and zero elsewhere.

6. Some properties of  $\Gamma(\cdot)$ :

- (a)  $\Gamma(\alpha+1) = \alpha\Gamma(\alpha)$ , for any positive real number  $\alpha$ .
- (b)  $\Gamma(\frac{1}{2}) = \sqrt{\pi}$ .
- (c)  $\Gamma(0) = \Gamma(1) = 1$ .
- (d)  $\Gamma(k+1) = k!$ , for any positive integer k.

7. Determinant of a  $3 \times 3$  matrix:

Let 
$$J = \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix}$$
 then,  $det(J) = |J| = a \cdot \left| \begin{pmatrix} e & f \\ h & i \end{pmatrix} \right| - b \cdot \left| \begin{pmatrix} d & f \\ g & i \end{pmatrix} \right| + c \cdot \left| \begin{pmatrix} d & e \\ g & h \end{pmatrix} \right|$ 

8. A function  $g(\cdot)$  is said to be **non-decreasing** if  $g(b) \geq g(a)$  whenever b > a.

9. Let  $X_1,...,X_n$  be a random sample such that the pdf of  $X_i \sim f(x)$  for i=1,...,n, then the k-th order statistics is denoted by  $X_{(k)}$ . The pdf of  $Y=X_{(k)}$  is given by

$$f_{X_{(k)}}(x) = \frac{n!}{(k-1)! (n-k)!} (F(x))^{k-1} f(x) (1 - F(x))^{n-k}$$

for  $1 \leq k \leq n$ . The joint pdf of  $Y = (X_{(i)}, X_{(j)})$  for i < j, is given by

$$f_Y(u,v) = \frac{n!}{(i-1)! (j-1-i)! (n-j)!} (F(u))^{i-1} f(u) (F(v) - F(u))^{j-1-i} f(v) (1 - F(v))^{n-j}.$$

- 10. Central Limit Theorem: Let  $Z_1, ..., Z_n$  be identically distributed random variables with  $E(Z_1) = \mu$  and  $var(Z_1) = \sigma^2$ , then for large values of n, the distribution of  $\frac{\sqrt{n}(\bar{Z}-\mu)}{\sigma}$  can be approximated by N(0,1).
- 10. **Delta Method**: If Y is a random variable such that  $\sqrt{n}(Y \mu) \sim N(0, \sigma^2)$ , then for any one-to-one function  $g(\cdot)$ ,  $\sqrt{n}(g(Y) g(\mu)) \sim N(0, (g'(\mu))^2 \sigma^2)$ .
- 11. Let  $X_1, ..., X_m$  be a random sample from
- (a)  $Poi(\lambda)$ , then  $\sum_{i=1}^{m} X_i \sim Poi(m\lambda)$ .
- (b) Bin(k, p), then  $\sum_{i=1}^{m} X_i \sim Bin(mk, p)$ .
- (c)  $\Gamma(\alpha, \lambda)$ , then  $\sum_{i=1}^{m} X_i \sim \Gamma(m\alpha, \lambda)$ .
- 12. Trigonometric relations:
- (a)  $\frac{d}{dx}sin(x) = cos(x)$
- (b)  $\frac{d}{dx}cos(x) = -sin(x)$
- (c)  $sin^2(x) + cos^2(x) = 1$ .