

MSO 201a: Probability and Statistics
2016-2017-II Semester
Assignment-X

A. Illustrative Discussion Problems

1. Let random vector (X, Y) have the joint p.m.f.

$$f_{X,Y}(x, y) = P(X = x, Y = y) = \frac{(x + y + k - 1)!}{x!y!(k - 1)!} \theta^{x+y} (1 - 2\theta)^k, \quad x, y = 0, 1, 2, \dots,$$

where $k \geq 1$ is an integer and $0 < \theta < \frac{1}{2}$ (we say that (X, Y) has bivariate negative binomial distribution). Find the marginal p.m.f. of X , the conditional distribution of Y given $X = x$ ($x \in \{0, 1, \dots\}$) and the p.m.f. of $Z = X + Y$.

2. Let the bivariate r.v. (X, Y) have joint p.d.f.

$$f_{X,Y}(x, y) = \frac{\Gamma(\theta_1 + \theta_2 + \theta_3)}{\Gamma(\theta_1)\Gamma(\theta_2)\Gamma(\theta_3)} x^{\theta_1-1} y^{\theta_2-1} (1 - x - y)^{\theta_3-1}, \quad x > 0, \quad y > 0, \quad x + y < 1,$$

where $\theta_i > 0$, $i = 1, 2, 3$ (we say that (X, Y) has bivariate beta distribution). Find the marginal p.d.f. of X , the conditional p.d.f. of X given $Y = y$ ($y \in (0, 1)$) and the p.d.f. of $Z = X + Y$.

3. Let X_1, \dots, X_n be i.i.d. A.C. r.v.s with common p.d.f. $f(\cdot)$ and common d.f. $F(\cdot)$. For $r \in \{1, \dots, n\}$, let $X_{r:n} = r$ -th smallest of X_1, \dots, X_n (so that $X_{1:n} \leq X_{2:n} \leq \dots \leq X_{n:n}$ are order statistics of random sample X_1, \dots, X_n).
(a) Find the joint d.f. of r.v. $(X_{1:n}, \dots, X_{n:n})$ and hence show that it is A.C. with a joint p.d.f.

$$g(y_1, \dots, y_n) = \begin{cases} n! \prod_{i=1}^n f(y_i), & \text{if } y_1 < y_2 < \dots < y_n \\ 0, & \text{otherwise} \end{cases}.$$

(b) **(Relationship between beta and binomial distribution)** Let $X \sim \text{Bin}(n, p)$, where $p \in (0, 1)$ and $n \in \mathbb{N}$. Using integration by parts, show that

$$P(\{X \geq r\}) = \sum_{i=r}^n \binom{n}{i} p^i (1-p)^{n-i} = \frac{1}{\beta(r, n-r+1)} \int_0^p t^{r-1} (1-t)^{n-r} dt, \quad r \in \mathbb{N}, \quad r \leq n.$$

Hence show that if $X \sim \text{Bin}(n, p)$, then $P(X \geq r) \leq \binom{n}{r} p^r$.

(c) Using the fact that, for any real number x , the event $\{X_{r:n} \leq x\}$ occurs iff at least r among X_1, \dots, X_n are $\leq x$, find the d.f. of $X_{r:n}, r = 1, \dots, n$. Using the

relation derived in (b) show that $X_{r:n}$ is A.C. with p.d.f.

$$g_r(x) = \frac{n!}{(r-1)!(n-r)!} [F(x)]^{r-1} [1-F(x)]^{n-r} f(x), \quad -\infty < x < \infty.$$

4. (a) **(Poisson approximation to binomial distribution)** Show that $\lim_{n \rightarrow \infty} \binom{n}{r} p_n^r (1-p_n)^{n-r} = e^{-\lambda} \lambda^r / r!$, $r = 0, 1, 2, \dots$, provided $\lim_{n \rightarrow \infty} np_n = \lambda$ (i.e., for large n and small p , so that $np = \lambda > 0$, $\text{Bin}(n, p)$ probabilities can be approximated by $\text{Poisson}(\lambda)$ probabilities);
(b) The probability of hitting a target is 0.001 for each shot. Find the approximate probability of hitting a target at least twice in 5000 shots.
5. Let X be a non-negative and integer valued r.v., satisfying $P(X = 0) = p \in (0, 1)$ and having lack of memory property $P(X \geq n+m | X \geq n) = P(X \geq m)$, $n, m \in \{1, 2, \dots\}$. Show that $X \sim \text{Geometric}(p)$.
6. (a) **(Binomial approximation to hypergeometric distribution)** Show that a hypergeometric distributed can be approximated by a $\text{Bin}(n, p)$ distribution provided N is large ($N \rightarrow \infty$), a is large ($a \rightarrow \infty$) and $a/N \rightarrow p$, as $N \rightarrow \infty$;
(b) Out of 120 applicants for a job, only 80 are qualified for the job. Out of 120 applicants, five are selected at random for the interview. Find the probability that at least two selected applicants will be qualified for the job. Using Problem 6 (a) find an approximate value of this probability.
7. (a) If the m.g.f. of a r.v. X is $(1/3 + 2e^t/3)^5$, $-\infty < t < \infty$, find $P(X \in \{2, 3\})$;
(b) If the m.g.f. of a r.v. X is $e^{4(e^t-1)}$, $-\infty < t < \infty$, find $P(E(X) - 2\sqrt{\text{Var}(X)} < X < E(X) + 2\sqrt{\text{Var}(X)})$.
8. **(Relationship between gamma and Poisson distribution)** Let $X \sim \text{Poisson}(\lambda)$ and let k be a positive integer. Show that $P(X \leq k) = (k!)^{-1} \int_{\lambda}^{\infty} e^{-t} t^k dt$.
9. Each child in a family is equally likely to be a boy or a girl. Find the minimum number of children the family should have so that the probability of it having at least a boy and at least a girl is at least 0.95.
10. **(Relation between binomial and negative binomial probabilities)** Let n be a positive integer, $r \in \{1, 2, \dots, n\}$ and let $p \in (0, 1)$. Using probabilistic arguments and also otherwise show that

$$\sum_{k=r}^n \binom{n}{k} p^k (1-p)^{n-k} = p^r \sum_{k=0}^{n-r} \binom{r+k-1}{k} (1-p)^k,$$

i.e., for $r \in \{1, 2, \dots, n\}$, $P(\text{Bin}(n, p) \geq r) = P(\text{NB}(r, p) \leq n - r)$.

11. A mathematician carries at all times two match boxes, one in his left pocket and one in his right pocket. To begin with each match box contains n matches. Each time the mathematician needs a match he is equally likely to take it from either pocket. Consider the moment when the mathematician for the first time discovers that one of the match boxes is empty. Find the probability that at that moment the other box contains exactly k matches, where $k \in \{0, 1, \dots, n\}$.
12. (**Poisson approximation to negative binomial distribution**) Show that
$$\lim_{r \rightarrow \infty} \binom{r+k-1}{k} p_r^r (1-p_r)^k = \frac{e^{-\lambda} \lambda^k}{k!}, k = 0, 1, 2, \dots,$$
provided $\lim_{r \rightarrow \infty} p_r = 1$ and $\lim_{r \rightarrow \infty} (r(1-p_r)) = \lambda > 0$.
13. (a) Let $F(\cdot)$ be the d.f. of the r.v. X , where $P(X = 1) = p = 1 - P(X = 0)$. Find the distribution of $Y = F(X)$. Is $Y \sim U(0, 1)$?

 (b) Let the r.v. X have the p.d.f. $f(x) = 6x(1-x)$, if $0 \leq x \leq 1$, $= 0$, otherwise. Show that $Y = X^2(3 - 2X) \sim U(0, 1)$.
14. (a) If $X \sim U(0, 1)$, show that of $Y = -\lambda \ln X \sim \text{Exp}(\lambda)$, where $\lambda > 0$;

 (b) Let X_1 and X_2 be independent $N(0, 1)$ r.v.s and let $Y = \frac{X_1}{X_2}$. Show that Y has the Cauchy distribution. What is the distribution of $Y = X^{-1}$?

 (c) Let X and Y be i.i.d. r.v.s. with common p.d.f. $f(x) = \frac{c}{1+x^4}$, $-\infty < x < \infty$, where c is the normalizing constant. Find the p.d.f. of $\frac{X}{Y}$.
15. (a) If $X \sim N(12, 16)$, find $P(X \geq 20)$ (use $\Phi(2) = 0.9772$);

 (b) If $X \sim N(\mu, \sigma^2)$, $P(9.6 \leq X \leq 13.8) = 0.7008$ and $P(X \geq 9.6) = 0.8159$, find μ, σ^2 and $P(X \geq 13.8 | X \geq 9.6)$ (use $\Phi(0.9) = 0.8159$ and $\Phi(1.2) = 0.8849$).
16. For $x > 0$, show that $(x^{-1} - x^{-3})\phi(x) < 1 - \Phi(x) < x^{-1}\phi(x)$ (Hint: Use integration by parts in $(2\pi)^{\frac{1}{2}}(1 - \Phi(x)) = \int_x^\infty t^{-1}(te^{-\frac{t^2}{2}})dt$).
17. Let X be a r.v. with $E(X^m) = m!$, $m = 1, 2, \dots$, and suppose that the m.g.f. of X exists in a neighborhood of zero. Find the p.d.f./p.m.f. of X .
18. (a) Let $X \sim \text{Gamma}(\alpha_1, \theta)$ and $Y \sim \text{Gamma}(\alpha_2, \theta)$ be independent. Show that $U = X+Y$ and $V = \frac{X}{X+Y}$ are independently distributed; also find their distributions.

 (b) Let $W \sim \text{Beta}(\alpha_1, \alpha_2)$ and $T \sim \text{Gamma}(\alpha_1 + \alpha_2, \theta)$ be independent. Using (a), show that $WT \sim \text{Gamma}(\alpha_1, \theta)$.
19. Let $X \sim N(0, 1)$ and $Y \sim N(0, 1)$ be independent. Find p.d.f.s of $\frac{X}{|Y|}$ and $\frac{|X|}{|Y|}$.

20. Let X_1, \dots, X_n denote a random sample from the $\text{Exp}(1)$ distribution. Find the marginal distributions of Y_1, \dots, Y_n , where

$$Y_i = \frac{\sum_{j=1}^i X_j}{\sum_{j=1}^{i+1} X_j}, \quad i = 1, \dots, n-1, \quad Y_n = X_1 + \dots + X_n.$$

Are Y_1, \dots, Y_n independent?

21. (a) Let X_1, \dots, X_n ($n \geq 3$) be a random sample from $N(\mu, \sigma^2)$ distribution. Find $E\left(\frac{\bar{X}}{S}\right)$.
(b) Let X_1 and X_2 be i.i.d. $N(0, 1)$ r.v.s. Find the distribution of $Z = \frac{X_1 + X_2}{|X_1 - X_2|}$.
22. Let $X_{(1)}, X_{(2)}, \dots, X_{(n)}$ be the set of order statistics associated with a random sample of size n (≥ 2) from the $\text{Exp}(1)$ distribution.
(a) Let $Z_1 = nX_{(1)}$, $Z_i = (n-i+1)(X_{(i)} - X_{(i-1)})$, $i = 2, \dots, n$. Show that Z_1, \dots, Z_n are i.i.d. $\text{Exp}(1)$ r.v.s.
(b) Using (a) or otherwise, find $E(X_{(r)})$, $\text{Var}(X_{(r)})$ and $\text{Cov}(X_{(r)}, X_{(s)})$, $r < s$.
(c) Show that $X_{(r)}$ and $X_{(s)} - X_{(r)}$ are independent for any $s > r$;
(d) Find the p.d.f. of $X_{(r+1)} - X_{(r)}$;
23. Let X_1, \dots, X_n be i.i.d. nonnegative A.C. r.v.s. with common d.f. $F(\cdot)$. If $E(|X_1|) < \infty$ and $M_n = \max(X_1, \dots, X_n)$, show that

$$E(M_n) = E(M_{n-1}) + \int_0^\infty (F(x))^{n-1} (1 - F(x)) dx.$$

24. Let X and Y be independent and identically distributed $N(0, \sigma^2)$ r.v.s. Let $U = aX + bY$ and $V = bX - aY$ ($a \neq 0, b \neq 0$).
(a) Show that U and V are normally distributed independent random variables;
(b) Show that $\frac{X+Y}{\sqrt{2}}$ and $\frac{X-Y}{\sqrt{2}}$ are independent $N(0, \sigma^2)$ random variables.
25. Let X be a random variable and let α and β be real numbers such that $-\infty < \alpha < \beta < \infty$. Show that $X \sim U(\alpha, \beta)$ if, and only if, $P(\{\alpha < X \leq \beta\}) = 1$ and $P(\{X \in I\}) = P(\{X \in J\})$, for any pair of subintervals I, J of (α, β) having the same length.

B. Practice Problems

1. Let $\underline{X} = (X_1, X_2)$ have the joint p.m.f.

$$f_{\underline{X}}(x_1, x_2) = \begin{cases} \frac{x_1 x_2}{36}, & \text{if } x_1 = 1, 2, 3, x_2 = 1, 2, 3 \\ 0, & \text{otherwise} \end{cases},$$

and let $Y = X_1 + \frac{X_2}{2}$. Find the p.m.f. of Y .

2. (a) Let $X \sim \text{Bin}(n_1, p)$ and $Y \sim \text{Bin}(n_2, p)$ be independent r.v.s. For $t \in \{0, 1, \dots, \min(n_1, n_2)\}$, find the conditional distribution and conditional mean of X given $X + Y = t$.
 (b) Let $X \sim \text{Poisson}(\lambda_1)$ and $Y \sim \text{Poisson}(\lambda_2)$ be independent r.v.s. For $t \in \{0, 1, \dots, \}$, find the conditional distribution and conditional mean of X given $X + Y = t$.
3. Suppose that the number, X , of eggs laid by a bird has the $\text{Poisson}(\lambda)$ distribution, and the probability that an egg would finally develop is $p \in (0, 1)$; here $\lambda > 0$. Under the assumption of independence of development of eggs, show that the number, Y , of eggs surviving has the $\text{Poisson}(\lambda p)$ distribution. Find the conditional distribution of X given $Y = y$.
4. (a) If $X \sim \text{Poisson}(\lambda)$, find $E((2 + X)^{-1})$;
 (b) If $X \sim \text{Geometric}(p)$ and r is a positive integer, find $E(\min(X, r))$.
5. Let $X \sim \text{U}(0, \theta)$, where θ is a positive integer. Let $Y = X - [X]$, where $[x]$ is the largest integer $\leq x$. Show that $Y \sim \text{U}(0, 1)$.
6. If $X \sim \text{U}(0, \theta)$, where $\theta > 0$, find the distribution of $Y = \min(X, \theta/2)$. Calculate $P(\theta/4 < Y < \theta/2)$.
7. Let $X \sim \text{N}(0, 1)$ and let $Y = X$, if $|X| \leq 1$, $= -X$, if $|X| > 1$. Find the distribution of Y .
8. Let X and Y be i.i.d. $\text{U}(-1, 1)$ r.v.s. Find the probability that $x^2 + Xx + Y = 0 > 0$, for every real number x .
9. Let X and Y be respective arrival times of two friends A and B who agree to meet at a spot and wait for the other only for t minutes. Suppose that X and Y are i.i.d. $\text{Exp}(\lambda)$. Show that probability of A and B meeting each other is $1 - e^{-\lambda t}$.
10. Let X and Y be independent r.v.s. with respective p.d.f.s. $f_X(x) = \frac{1}{3}$, if $1 \leq x \leq 4$, $= 0$, otherwise, and $f_Y(y) = e^{-(y-2)}$, if $y \geq 2$ and $= 0$, otherwise. Find the d.f. of $T = \frac{X}{Y}$ and hence find the p.d.f. of T .

11. Let X and Y be i.i.d. $\text{U}(0, 1)$ r.v.s. Find the marginal p.d.f.s of
- $X + Y$, $X - Y$, $\frac{X+Y}{X-Y}$, $|X - Y|$;
 - $\min(X, Y)$, $\max(X, Y)$, $\frac{\min(X, Y)}{\max(X, Y)}$;
 - $X^2 + Y^2$.
12. Let X and Y be i.i.d. $\text{N}(0, 1)$ random variables. Define the random variables R and Θ by $X = R \cos \Theta$, $Y = R \sin \Theta$.
- Show that R and Θ are independent with $\frac{R^2}{2} \sim \text{Exp}(1)$ and $\Theta \sim \text{U}(0, 2\pi)$.
 - Show that $X^2 + Y^2$ and $\frac{X}{Y}$ are independently distributed.
 - Show that $\sin \Theta$ and $\sin 2\Theta$ are identically distributed and hence find the pdf of $T = \frac{XY}{\sqrt{X^2+Y^2}}$.
 - Find the distribution of $U = \frac{3X^2Y - Y^3}{X^2 + Y^2}$.
 - Let U_1 and U_2 be i.i.d. $\text{U}(0, 1)$ r.v.s. Then show that $X_1 = \sqrt{-2 \ln U_1} \cos(2\pi U_2)$ and $X_2 = \sqrt{-2 \ln U_1} \sin(2\pi U_2)$ are i.i.d. $\text{N}(0, 1)$ r.v.s. (This is known as the Box-Muller transformation).
13. Let X_1, X_2, X_3 be i.i.d. $\text{Gamma}(m, 1)$ random variables. Let $Z_1 = X_1 + X_2 + X_3$, $Z_2 = \frac{X_2}{X_1 + X_2 + X_3}$ and $Z_3 = \frac{X_3}{X_1 + X_2 + X_3}$.
- Show that Z_1 and (Z_2, Z_3) are independent and find marginal p.d.f.s of Z_1 , Z_2 and Z_3 .
 - Find $E(Z_1^2 Z_2 Z_3)$.
14. Let $X_{(1)} < X_{(2)} < \dots < X_{(n)}$ be the order statistics associated with a random sample of size n (≥ 2) from the $\text{U}(0, 1)$ distribution. Let $Y_i = \frac{X_{(i)}}{X_{(i+1)}}$, $i = 1, \dots, n-1$, and $Y_n = X_{(n)}$. Show that Y_1, \dots, Y_n are independent and find the p.d.f. of Y_i , $i = 1, \dots, n$.

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 Solutions

Problem No.1 We know that, for a positive integer m and $\lambda \in (-1, 1)$,

$$(-\lambda)^m = 1 + \binom{m}{1} \lambda + \binom{m+1}{2} \lambda^2 + \binom{m+2}{3} \lambda^3 + \dots = \sum_{j=0}^{\infty} \binom{m+j-1}{j} \lambda^j.$$

Clearly the support of X is $S_X = \{0, 1, 2, \dots\}$ and for $\lambda \in S_X$,

$$f_X(x) = P(X=x) = \sum_{y=0}^{\infty} P(X=x, Y=y) = \frac{\lambda^{x+k-1}}{\lambda^x k!} \theta^x (1-\theta)^k \sum_{j=0}^{\infty} \binom{x+y+k-1}{j} \theta^j$$

$$= \binom{x+k-1}{k-1} \theta^x (1-\theta)^k (1-\theta)^{(x+k)}$$

Thus

$$f_X(x) = \begin{cases} \binom{x+k-1}{k-1} \left(\frac{\theta}{1-\theta}\right)^x \left(1 - \frac{\theta}{1-\theta}\right)^k & \text{if } x=0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases} \rightarrow NB(k, \frac{1-\theta}{\theta})$$

For $\lambda \in \{0, 1, 2, \dots\}$,

$$f_{Y|X}(y|x) = \frac{P(X=x, Y=y)}{P(X=x)} = \begin{cases} \binom{x+k-1}{y} \left(\frac{\theta}{1-\theta}\right)^{x+k-y} \theta^y & \text{if } y=0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

$\rightarrow NB(x+k, \theta)$.

Let $Z = X+T$. Then $S_Z = \{0, 1, 2, \dots\}$ and, for $z \in S_Z$,

$$f_Z(z) = P(X+T=z) = \sum_{x=0}^{\infty} \sum_{y=0}^{\infty} \frac{\lambda^{x+k-1}}{\lambda^x k!} \theta^x (1-\theta)^k$$

$$\stackrel{x+y=z}{=} \frac{\lambda^{z+k-1}}{\lambda^z z!} \theta^z (1-\theta)^k \sum_{x=0}^z \binom{z}{x} = \binom{z+k-1}{z} \theta^z (1-\theta)^k z!$$

Thus

$$f_Z(z) = \begin{cases} \binom{z+k-1}{z} \theta^z (1-\theta)^k & \text{if } z=0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases} \rightarrow NB(k, \theta)$$

Problem No. 2 Let $U = h_1(x, \gamma) = X$ and $Z = h_2(x, \gamma) = X+T$. Then

$S_{X, \gamma} = \{(x, \gamma) \in \mathbb{R}^2 : x \geq 0, \gamma \geq 0, x+\gamma < 1\}$, the transformation

$h = (h_1, h_2) : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ is 1-1 with inverse transformation

$(h_1^{-1}(u, z), h_2^{-1}(u, z))$ where $h_1^{-1}(u, z) = u$, $h_2^{-1}(u, z) = z-u$ the Jacobian

$J = \begin{vmatrix} 1 & 0 \\ -1 & 1 \end{vmatrix} = 1$. Also $(u, z) \in h_1^{-1}(S_{X, \gamma}) \Leftrightarrow u \geq 0, z-u \geq 0, u+z < 1$

$\Rightarrow 0 < u < z < 1$. Thus

$$h_1(S_{X, \gamma}) = \{(u, z) \in \mathbb{R}^2 : 0 < u < z < 1\}.$$

and the joint p.d.f. of (U, Z) is

$$f_{U,Z}(u, z) = f_{X,Y}(u-z-u) I_{\{z \leq u\}} = \begin{cases} \frac{\Gamma(\theta_1 + \theta_2 + \theta_3)}{\Gamma(\theta_1)\Gamma(\theta_2)\Gamma(\theta_3)} u^{\theta_1-1} (z-u)^{\theta_2-1} (1-z)^{\theta_3-1} & \text{if } 0 < u < 1 \\ 0 & \text{otherwise.} \end{cases}$$

Clearly, for $u \notin (0, 1)$, $f_X(u) = f_U(u) = \int_0^1 f_{U,Z}(u, z) dz = 0$.

For $u \in (0, 1)$

$$\begin{aligned} f_X(u) &= f_U(u) = \int_0^u f_{U,Z}(u, z) dz = \frac{\Gamma(\theta_1 + \theta_2 + \theta_3)}{\Gamma(\theta_1)\Gamma(\theta_2)\Gamma(\theta_3)} u^{\theta_1-1} \int_0^1 (z-u)^{\theta_2-1} (1-z)^{\theta_3-1} dz \\ &= \frac{\Gamma(\theta_1 + \theta_2 + \theta_3)}{\Gamma(\theta_1)\Gamma(\theta_2)\Gamma(\theta_3)} u^{\theta_1-1} \int_0^1 ((1-u)(1-z))^{\theta_2-1} ((1-u)z)^{\theta_3-1} (1-u) dz \\ &= \frac{\Gamma(\theta_1 + \theta_2 + \theta_3)}{\Gamma(\theta_1)\Gamma(\theta_2)\Gamma(\theta_3)} u^{\theta_1-1} (1-u)^{\theta_2+\theta_3-1} \int_0^1 z^{\theta_3-1} (1-z)^{\theta_2-1} dz = \frac{1}{\beta(\theta_2, \theta_3)} u^{\theta_1-1} (1-u)^{\theta_2+\theta_3-1}. \end{aligned}$$

Thus $X \sim \text{Beta}(\theta_1, \theta_2 + \theta_3)$.

By symmetry $Y \sim \text{Beta}(\theta_2, \theta_1 + \theta_3)$. Thus, for $y \in (0, 1)$,

$$f_{X,Y}(x, y) = \frac{f_{X,Y}(x, y)}{f_Y(y)} = \begin{cases} \frac{1}{\beta(\theta_2, \theta_3)} \frac{1}{1-y} \left(\frac{x}{1-y}\right)^{\theta_1-1} \left(1-\frac{x}{1-y}\right)^{\theta_3-1} & \text{if } 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

Also

$$f_Z(z) = \int_0^1 f_{U,Z}(u, z) du. \quad \text{Clearly, for } z \notin (0, 1), f_Z(z) = 0.$$

For $z \in (0, 1)$

$$\begin{aligned} f_Z(z) &= \int_0^1 f_{U,Z}(u, z) du = \frac{\Gamma(\theta_1 + \theta_2 + \theta_3)}{\Gamma(\theta_1)\Gamma(\theta_2)\Gamma(\theta_3)} (1-z)^{\theta_3-1} \int_0^z u^{\theta_1-1} (z-u)^{\theta_2-1} du \\ &= \frac{\Gamma(\theta_1 + \theta_2 + \theta_3)}{\Gamma(\theta_1)\Gamma(\theta_2)\Gamma(\theta_3)} (1-z)^{\theta_3-1} \int_0^1 (zu)^{\theta_1-1} (z-zu)^{\theta_2-1} z du = C z^{\theta_1+\theta_2-1} (1-z)^{\theta_3-1} \\ &\Rightarrow Z \sim \text{Beta}(\theta_1 + \theta_2, \theta_3) \end{aligned}$$

Problem No.3 (a) Let S_n denote the set of all permutations $\underline{p} = (p_1 \dots p_n)$ of $(1, \dots, n)$. Since X is A.C.

$$\sum_{\underline{p} \in S_n} P(X_{p_1} < X_{p_2} < \dots < X_{p_n}) = 1 \quad \text{--- (A1)}$$

Also X_1, \dots, X_n are i.i.d $\Rightarrow X_1, \dots, X_n$ are exchangeable

$$\boxed{2/17} \Rightarrow (X_1, \dots, X_n) \stackrel{d}{=} (X_{p_1}, \dots, X_{p_n}) \forall \underline{p} \in S_n \rightarrow (A_2)$$

The joint d.f. of $(X_{1:n}, X_{2:n}, \dots, X_{n:n})$ is

$$\begin{aligned}
 g(y_1, \dots, y_n) &= P(X_{1:n} \leq y_1, X_{2:n} \leq y_2, \dots, X_{n:n} \leq y_n) \\
 &= \sum_{P \in S_n} P(X_{1:n} \leq y_1, X_{2:n} \leq y_2, \dots, X_{n:n} \leq y_n, X_{P_1} < X_{P_2} < \dots < X_{P_n}) \quad (\text{using A}_1) \\
 &= \sum_{P \in S_n} P(X_{P_1} \leq y_1, X_{P_2} \leq y_2, \dots, X_{P_n} \leq y_n, X_{P_1} < X_{P_2} < \dots < X_{P_n}) \\
 &= \sum_{P \in S_n} P(X_1 \leq y_1, X_2 \leq y_2, \dots, X_n \leq y_n, X_1 < X_2 < \dots < X_n) \quad (\text{using A}_2) \\
 &= \ln P(X_1 \leq y_1, X_2 \leq y_2, \dots, X_n \leq y_n, X_1 < X_2 < \dots < X_n) \\
 &= \ln \int_{-\infty}^{y_1} \int_{-\infty}^{y_2} \dots \int_{-\infty}^{y_n} f(x_1, x_2, \dots, x_n) dx_n \dots dx_1 \\
 &\quad X_1 < X_2 < \dots < X_n \\
 &= \ln \int_{-\infty}^{y_1} \int_{-\infty}^{y_2} \dots \int_{-\infty}^{y_n} \left(\prod_{i=1}^n f(x_i) \right) I_A(\underline{x}) dx_n \dots dx_1 \quad \dots (A_3)
 \end{aligned}$$

where $A = \{\underline{x} \in \mathbb{R}^n : -\infty < x_1 < x_2 < \dots < x_n < \infty\}$.

From A_3 , it follows that $(X_{1:n}, X_{2:n}, \dots, X_{n:n})$ is Arc. with joint p.d.f.

$$g(y_1, \dots, y_n) = \left(\prod_{i=1}^n f(y_i) \right) I_A(\underline{y}) = \begin{cases} \ln \prod_{i=1}^n f(y_i), & \text{if } -\infty < y_1 < y_2 < \dots < y_n \\ 0, & \text{otherwise.} \end{cases}$$

(b) For $p \in (0, 1)$, $n \in \mathbb{N}$, $r \in \mathbb{N}$ and $r \leq n$, let

$$I_{r,n}(p) = \frac{1}{p(r, n-r+1)} \int_0^p t^{r-1} (1-t)^{n-r} dt.$$

Integration by parts gives

$$\begin{aligned}
 I_{r,n}(p) &= \binom{n}{r} p^r (1-p)^{n-r} + \frac{1}{\ln \frac{1}{p(r, n-r+1)}} \int_0^p t^r (1-t)^{n-r-1} dt \\
 &= \binom{n}{r} p^r (1-p)^{n-r} + I_{r+1,n}(p) \\
 &= \binom{n}{r} p^r (1-p)^{n-r} + \binom{n}{r+1} p^{r+1} (1-p)^{n-r-1} + I_{r+2,n}(p) \\
 &= \binom{n}{r} p^r (1-p)^{n-r} + \binom{n}{r+1} p^{r+1} (1-p)^{n-r-1} + \dots + \binom{n}{n-1} p^{n-1} (1-p) + I_{n,n}(p) \\
 &= \sum_{i=r}^n \binom{n}{i} p^i (1-p)^{n-i}
 \end{aligned}$$

(c) For $x \in \mathbb{R}$ and $r=1, 2, \dots, n$

$$F_{X_{r:n}}(x) = P(X_{r:n} \leq x) = P(\text{at least } r \text{ of } X_1, \dots, X_n \leq x)$$

$$= \sum_{i=r}^n P(\text{exactly } i \text{ of } X_1, \dots, X_n \leq x)$$

Fix $x \in \mathbb{R}$. Consider a sequence of n trials where i -th trial is designated as success if $\{X_i \leq x\}$ and failure otherwise.

X_1, \dots, X_n are independent \Rightarrow n trials are independent
 X_1, \dots, X_n are identically distributed \Rightarrow probability of success in each trial is the same $= P(X_i \leq x) = F(x)$

Thus

$$F_{X_{r:n}}(x) = \sum_{i=r}^n P(i \text{ successes in } n \text{ trials})$$

$$= \sum_{i=r}^n \binom{n}{i} [F(x)]^i [1-F(x)]^{n-i}$$

$$= \frac{1}{P(r, n-r+1)} \int_0^{x^{n-r}} t^{r-1} (1-t)^{n-r} dt$$

$$= \frac{1}{P(r, n-r+1)} \int_{-\infty}^{F(x)} [F(y)]^{r-1} (1-F(y))^{n-r} f(y) dy$$

[on making the transformation
 $t=F(y) \Rightarrow dt=f(y) dy$]

$\Rightarrow X_{r:n}$ is A.C. with p.d.f.

$$f_{X_{r:n}}(x) = \frac{1}{P(r, n-r+1)} [F(x)]^{r-1} (1-F(x))^{n-r} f(x), -\infty < x < \infty$$

$$= \frac{\Gamma(n)}{\Gamma(r) \Gamma(n-r)} [F(x)]^{r-1} (1-F(x))^{n-r} f(x), -\infty < x < \infty.$$

[Problem No. 4] (a) For $r \in \{0, 1, \dots, n\}$ let

$$A_n = \binom{n}{r} p_n^r (1-p_n)^{n-r} = \frac{1}{r!} \left(1 - \frac{1}{n}\right) \left(1 - \frac{2}{n}\right) \cdots \left(1 - \frac{r-1}{n}\right) \left(n p_n\right)^r \left(1-p_n\right)^{n-r}$$

Let $t_n = n \ln(1-p_n) = n \left[-p_n - \frac{p_n^2}{2} - \frac{1}{1-p_n}\right]$, where $0 < p_n < 1$, $n \geq 1, 2, \dots$.

$n p_n \rightarrow \lambda \Rightarrow p_n \rightarrow 0$ and $t_n \rightarrow 0 \Rightarrow t_n \rightarrow -\lambda \Rightarrow e^{t_n} = (1-p_n)^n \rightarrow e^{-\lambda}$
 Therefore $A_n \xrightarrow{n \rightarrow \infty} \frac{e^{-\lambda} \lambda^r}{r!}, r=0, 1, 2, \dots$

(b) Define, $X = \# \text{ of times the target is hit in 5000 shots}$
 Then $X \sim \text{Bin}(5000, 0.001)$. $\because n=5000$ is large and $p=0.001$ is small. Using (a) (with $\lambda=n p=5$)

$$\text{Required probability} = P(X \geq 2) = 1 - P(X=0) - P(X=1)$$

$$= 1 - e^{-5} - e^{-5} \cdot 5 = 1 - 6e^{-5}.$$

Problem No. 5

We have

$$P(X \geq n+m) = P(X \geq n) P(X \geq m), \quad \forall m, n \in \mathbb{N}, m, n \geq 0$$

$$P(X \geq n+1) = P(X \geq n) P(X \geq 1) = (1-p) P(X \geq n), \quad n \geq 0$$

$$P(X \geq 2) = (1-p) P(X \geq 1) = (1-p)^2. \quad \text{By induction } P(X \geq n) = (1-p)^n, \quad \forall n \in \mathbb{N}$$

$$\Rightarrow P(X=n) = P(X \geq n) - P(X \geq n+1) = p(1-p)^n, \quad n \geq 0$$

$$\Rightarrow P(X=n) = \begin{cases} p(1-p)^n, & \text{if } n \geq 0 \\ 0, & \text{otherwise} \end{cases} \Rightarrow X \sim \text{Geometric}(p).$$

Problem No. 6

(a) Let $X \sim \text{Hyper}(n, N, a)$. The support of X is $S_X = \{m \in \mathbb{N}\}$:

$$\max(0, n-N+a) \leq m \leq \min(n, a)$$

$$n-N+a \geq N\left(\frac{n}{N}-1+\frac{a}{N}\right) \rightarrow 0 \quad \text{and} \quad a = N \frac{a}{N} \rightarrow a \quad \text{as } N \rightarrow \infty$$

$$\text{Thus, as } N \rightarrow \infty, \quad \max(0, n-N+a) \rightarrow 0 \quad \text{and} \quad \min(n, a) \rightarrow n.$$

Also, for $k \in S_X$

$$f_X(k) = \frac{\binom{a}{k} \binom{N-a}{n-k}}{\binom{N}{n}} = \binom{n}{k} \left\{ \prod_{j=0}^{k-1} \frac{a-j}{N-j} \right\} \left\{ \prod_{j=0}^{n-k-1} \frac{N-a-j}{N-k-j} \right\}$$

$$\xrightarrow{N \rightarrow \infty} \binom{n}{k} \left\{ \prod_{j=0}^{k-1} p \right\} \left\{ \prod_{j=0}^{n-k-1} (1-p) \right\} = \binom{n}{k} p^k (1-p)^{n-k}.$$

Thus

$$\lim_{N \rightarrow \infty} f_X(k) = \begin{cases} \binom{n}{k} p^k (1-p)^{n-k}, & \text{if } k = 0, 1, 2, \dots, n \\ 0, & \text{otherwise} \end{cases}$$

(b) Let $X = \# \text{ of applicants in the selected sample that are qualified for the job.}$

$$X \sim \text{Hyper}(80, 5, 120)$$

$$\text{Required probability} = P(X \geq 2) = 1 - P(X=0) - P(X=1)$$

$$= 1 - \frac{\binom{80}{0} \binom{40}{5}}{\binom{120}{5}} - \frac{\binom{80}{1} \binom{40}{4}}{\binom{120}{5}}$$

$N=120$ is large, $a=80$ is large, $p = \frac{a}{N} = \frac{2}{3}$. By (a)

$X \approx \text{Bin}(15, \frac{2}{3})$. Approximate Required Probability

$$= 1 - P(X=0) - P(X=1)$$

$$= 1 - \left(\frac{5}{0}\right) \left(\frac{2}{3}\right)^0 \left(\frac{1}{3}\right)^5 - \left(\frac{5}{1}\right) \left(\frac{2}{3}\right)^1 \left(\frac{1}{3}\right)^4.$$

Problem No. 7

(a) Clearly $X \sim \text{Bin}(5, \frac{2}{3})$

$$P(X \in \{2, 3, 4\}) = \binom{5}{2} \left(\frac{2}{3}\right)^2 \left(\frac{1}{3}\right)^3 + \binom{5}{3} \left(\frac{2}{3}\right)^3 \left(\frac{1}{3}\right)^2 = \frac{40}{81}.$$

(b) Clearly $X \sim \text{Poisson}(4)$. Thus $E(X) = \text{Var}(X) = 4$.

$$P(E(X) - 2\sqrt{\text{Var}(X)} < X < E(X) + 2\sqrt{\text{Var}(X)}) = P(0 < X < 8) \\ = \sum_{j=1}^7 \frac{e^{-4} 4^j}{j!}$$

Problem No. 8

For $k=1, 2, \dots$, let

$$I_k(\lambda) = \frac{1}{k!} \int_0^\lambda e^{-t} t^k dt.$$

Integrating by parts we get

$$I_k(\lambda) = \frac{e^{-\lambda} \lambda^k}{k!} + I_{k-1}(\lambda) = \frac{e^{-\lambda} \lambda^k}{k!} + \frac{e^{-\lambda} \lambda^{k-1}}{(k-1)!} + I_{k-2}(\lambda) \\ = \sum_{j=1}^k \frac{e^{-\lambda} \lambda^j}{j!} + I_0(\lambda) = \sum_{j=0}^k \frac{e^{-\lambda} \lambda^j}{j!} = P(X \leq k).$$

Problem No. 9

Suppose that the family has n children and $X = \# \text{ of boys in the family}$. Then

$$X \sim \text{Bin}(n, \frac{1}{2}).$$

$$P(\text{the family has at least one boy and a girl}) = P(X \geq 1, n-X \geq 1) \\ = P(1 \leq X \leq n-1) = 1 - P(X=0) - P(X=n) = 1 - \left(\frac{1}{2}\right)^n - \left(\frac{1}{2}\right)^n = 1 - \frac{1}{2^n}$$

$$\text{We require } 1 - \frac{1}{2^n} \geq 0.95 \Leftrightarrow 2^n \geq 40 \Leftrightarrow n \geq 6$$

Required Answer = 6

Problem No. 10

Consider a sequence of independent Bernoulli trials with probability of success in each trial being p . Then

$$\sum_{k=r}^n \binom{n}{k} p^k (1-p)^{n-k} = P(\text{at least } r \text{ successes in } n \text{ Bernoulli trials}) \\ = P\left(\bigcup_{k=0}^{n-r} \{\text{r-th success in } (r+k)\text{-th trial}\}\right) \\ = \sum_{k=0}^{n-r} P(\text{r-th success in } (r+k)\text{-th trial}) \\ = \sum_{k=0}^{n-r} \underbrace{\binom{r+k-1}{r-1} p^{r-1} (1-p)^k}_{(r-1) \text{ successes in first } (r+k-1) \text{ trials}} \cdot p \quad \begin{matrix} \rightarrow \\ \text{success in } (r+k)\text{-th trial} \end{matrix} \\ = \text{RHS}$$

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Direct Method: Let $v = 1-p$. We need to show that

$$\sum_{k=v}^n \binom{n}{k} p^k (1-p)^{n-k} = p^v \sum_{k=0}^{n-v} \binom{n+v-1}{k} (1-p)^k$$

$$\Leftrightarrow \sum_{k=v}^n \binom{n}{k} (1-v)^{k-v} v^{n-k} = \sum_{k=0}^{n-v} \binom{n+v-1}{k} v^k.$$

LHS = polynomial of degree $n-v = \sum_{k=0}^{n-v} c_k v^k$

It suffices to show that $c_k = \binom{n+v-1}{k}$, $k=0, 1, 2, \dots, n-v$
for $k \in \{0, 1, 2, \dots, n-v\}$

$$c_k = \text{Coefficient of } v^k \text{ in } \sum_{j=v}^n \binom{n}{j} (1-v)^{j-v} v^{n-j} = \sum_{j=v}^{n-v} \binom{n}{n-j} (1-v)^{n-j-v} v^j$$

$$= \text{Coefficient of } v^k \text{ in } \sum_{j=0}^k \binom{n}{j} (-1)^{k-j} \binom{n-j-v}{k-j} = \binom{n}{k}$$

$$= \sum_{j=0}^k \binom{n}{k-j} (-1)^j \binom{n-k-v+j}{j}$$

But

$$(-1)^j \binom{n-k-v+j}{j} = (-1)^j \frac{(n-k-v+j)(n-k-v+j-1) \cdots (n-k-v+1)}{j!}$$

$$= \frac{(-n+k+v-j)(-n+k+v-j+1) \cdots (-n+k+v-1)}{j!}$$

$$= \binom{-n+k+v-1}{j}$$

$$\Rightarrow c_k = \sum_{j=0}^k \binom{n}{k-j} \binom{-n+k+v-1}{j} = \binom{k+v-1}{k}.$$

Problem No. 11 Let us call the event of choosing box 1 as success and that of choosing box 2 as failure. Then we have a sequence of Bernoulli trials with probability of success in each trial being $p = \frac{1}{2}$.

Required probability = $P(\text{When box 1 is found empty, box 2 has } k \text{ matches})$

+ $P(\text{When box 2 is found empty, box 1 has } k \text{ matches})$

= $P(n-k \text{ failures precede the } (n-k)\text{-th success})$

+ $P(n-k \text{ successes precede the } (n-k)\text{-th failure})$

$$= \binom{2n-k}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{n-k} \times \frac{1}{2} + \binom{2n-k}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{n-k} \times \frac{1}{2} = \binom{2n-k}{n} \left(\frac{1}{2}\right)^{2n-k}$$

Problem No. 12

We will use the Stirling approximation

$$\lim_{n \rightarrow \infty} \frac{\ln}{\sqrt{2\pi} e^n n^{n+\frac{1}{2}}} = 1, \text{ i.e. } \ln \approx \sqrt{2\pi} e^n n^{n+\frac{1}{2}} \text{ for large } n$$

$$\text{Thus LHS} = \frac{\frac{r+k-1}{k}}{\frac{r}{k}} \frac{p_r}{(1-p_r)^k}$$

$$\approx \frac{\sqrt{2\pi} e^{-(r+k-1)} (r+k-1)^{r+k-1}}{\frac{r}{k} \sqrt{2\pi} e^{-(r-1)} (r-1)^{r-1}} \left(1 - \frac{\lambda}{r}\right)^r \left(\frac{\lambda}{r}\right)^k \text{ for large } r$$

$$= \frac{e^{-\lambda} \lambda^k}{\frac{r}{k}} \left(1 - \frac{\lambda}{r}\right)^r \frac{\left(1 + \frac{k-1}{r}\right)^{r+k-1}}{\left(1 - \frac{\lambda}{r}\right)^{r-1}}$$

$$\approx \frac{e^{-\lambda} \lambda^k}{\frac{r}{k}} e^{-\lambda} \frac{e^{k-1}}{e^r} \quad (\text{for large } r)$$

$$= \frac{e^{-\lambda} \lambda^k}{\frac{r}{k}}$$

Problem No. 13

$$(a) F(x) = \begin{cases} 0, & \text{if } x < 0 \\ 1-p, & \text{if } 0 \leq x < 1 \\ 1, & \text{if } x \geq 1 \end{cases} \Rightarrow Y = F(X) = \begin{cases} 1-p, & \text{if } x=0 \\ 1, & \text{if } x \geq 1 \end{cases}$$

Clearly Y is discrete with p.m.b.

$$f_Y(y) = P(Y=y) = \begin{cases} 1-p, & \text{if } y=1-p \\ p, & \text{if } y=1 \\ 0, & \text{otherwise} \end{cases} \quad Y \sim U(0,1).$$

(b) The d.b. of X is

$$F(x) = \int_{-\infty}^x f(t+1) dt = \begin{cases} 0, & \text{if } x < 0 \\ x^2(3-2x), & \text{if } 0 \leq x < 1 \\ 1, & \text{if } x \geq 1 \end{cases} \Rightarrow Y = F(X) \text{ with probability one}$$

$\Rightarrow Y \sim U(0,1)$ (Since X is continuous)

Problem No. 14

$$S_X = [0, 1] \quad f_X(x) = \begin{cases} 1, & 0 < x < 1 \\ 0, & \text{o.w.} \end{cases}$$

$$Y = h(x) = -\lambda \ln x, \text{ where } h(x) = -\lambda \ln x, x > 0.$$

$h: (0, \infty) \rightarrow (0, \infty)$ is 1-1 with inverse transformation $h^{-1}(y) = e^{-\frac{y}{\lambda}}$.

$h(10, \alpha) = (0, \alpha)$. Thus the p.d.f. of γ is

$$f_{\gamma}(y) = b \times (h^{-1}(y)) \mid \frac{d}{dy} h^{-1}(y) \mid I_{(0, \alpha)}^{(\gamma)}$$

$$= b \times (e^{-\frac{y}{\lambda}}) \mid -\frac{1}{\lambda} e^{-\frac{y}{\lambda}} \mid I_{(0, \alpha)}^{(\gamma)} = \begin{cases} \frac{1}{\lambda} e^{-\frac{y}{\lambda}}, & \text{if } y > 0 \\ 0, & \text{otherwise.} \end{cases}$$

(b) The joint p.d.f. of (X_1, X_2) is

$$f_{X_1, X_2}(x_1, x_2) = \frac{1}{2\pi} e^{-\frac{(x_1^2 + x_2^2)}{2}}, (x_1, x_2) \in S_{X_1, X_2} = \mathbb{R}^2.$$

Consider the 1-1 transformation $(X_1, X_2) \rightarrow (Y_1, Y_2)$ defined by

$$Y_1 = \frac{X_1}{X_2}, \quad Y_2 = X_2$$

$$\text{i.e. } X_1 = Y_1 Y_2, \quad X_2 = Y_2$$

$$S_{Y_1, Y_2} = \{(Y_1, Y_2) : -\alpha < Y_2 < \alpha, -\alpha < Y_1 Y_2 < \alpha\} = \mathbb{R}^2.$$

$$J = \begin{vmatrix} 3 & Y_1 \\ 0 & 1 \end{vmatrix} = 3$$

Therefore joint p.d.f. of (Y_1, Y_2) is

$$f_{Y_1, Y_2}(y_1, y_2) = f_{X_1, X_2}(y_1 y_2, y_2) \mid J \mid I_{\mathbb{R}^2}(y_1, y_2)$$

$$= \frac{|J|}{2\pi} e^{-\frac{y_2^2}{2}(1+y_1^2)}, \quad -\alpha < y_2 < \alpha.$$

The p.d.f. of γ is

$$f_{\gamma}(y) = \int_{-\infty}^{\infty} f_{Y_1, Y_2}(y_1, y_2) dy_2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} |J| e^{-\frac{y_2^2}{2}(1+y_1^2)} dy_2 = \frac{1}{\pi} \cdot \frac{1}{1+y_1^2}, \quad -\alpha < y_1 < \alpha$$

$\gamma \sim \text{Cauchy distribution} \approx \text{Student t distribution with 1 d.f.}$

$$(c) \quad f_{X_1, X_2}(x_1, x_2) = \frac{c^2}{(1+x_1^4)(1+x_2^4)}, \quad (x_1, x_2) \in S_{X_1, X_2} = \mathbb{R}^2$$

Consider the 1-1 transformation $(X_1, X_2) \rightarrow (U_1, U_2)$ defined by $U_1 = \frac{X_1}{\sqrt{2}}$, $U_2 = \frac{X_2}{\sqrt{2}}$, i.e., $X_1 = U_1 \sqrt{2}$, $X_2 = U_2 \sqrt{2}$.

$$S_{U_1, U_2} = \{(U_1, U_2) : -\alpha < U_1, U_2 < \alpha, -\alpha < U_2 < \alpha\} = \mathbb{R}^2$$

$$J = \begin{vmatrix} U_2 & U_1 \\ 0 & 1 \end{vmatrix} = U_2$$

Therefore

$$f_{U_1, U_2}(u_1, u_2) = \frac{c^2 / U_2}{(1+u_1^4 u_2^4)(1+u_2^4)}, \quad -\alpha < u_1, u_2 < \alpha.$$

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$$\begin{aligned}
 f_{U_1}(u_1) &= \int_{-\infty}^{\infty} f_{U_1 U_2}(u_1, u_2) du_2 = 2c^2 \int_0^{\infty} \frac{u_2}{(1+u_1^4 u_2^4)(1+u_2^4)} du_2 \\
 &= 2c^2 \int_0^{\infty} \frac{u_2}{1-u_1^4} \left(\frac{1}{1+u_2^4} - \frac{u_1^4}{1+u_1^4 u_2^4} \right) du_2 \\
 &= \frac{2c^2}{1-u_1^4} \int_0^{\infty} \frac{u_2}{1+u_2^4} du_2 - \frac{2c^2 u_1^4}{1-u_1^4} \int_0^{\infty} \frac{u_2}{1+u_1^4 u_2^4} du_2 \\
 &= \frac{c^2}{1-u_1^4} \int_0^{\infty} \frac{du_3}{1+u_3^2} - \frac{c^2 u_1^4}{1-u_1^4} \times \frac{1}{u_1^4} \int_0^{\infty} \frac{du_3}{1+u_3^2} = \frac{c^2 \pi}{2(1-u_1^4)} (1-u_1^{-2}) \\
 &= \frac{c\pi}{2} \cdot \frac{1}{1+u_1^{-2}} = \frac{1}{\pi} \cdot \frac{1}{1+u_1^{-2}}, \quad (c = \frac{2}{\pi L}, \text{ since } \int_{-\infty}^{\infty} f_{U_1}(u_1) du_1 = 1) \\
 &\quad -\infty < u_1 < \infty.
 \end{aligned}$$

Problem No. 15

$$\begin{aligned}
 (a) P(X \geq 20) &= 1 - P(X \leq 20) = 1 - \Phi\left(\frac{20-\mu}{\sigma}\right) = 1 - \Phi(12) \\
 &= 1 - 0.9772 = 0.228
 \end{aligned}$$

$$\begin{aligned}
 (b) P(9.6 \leq X \leq 13.8) &= 0.7008 \Rightarrow P(X \leq 13.8) - (1 - P(X \geq 9.6)) = 0.7008 \\
 \Rightarrow P(X \leq 13.8) &= 0.7008 + .1841 = .8849 \\
 \Rightarrow \Phi\left(\frac{13.8-\mu}{\sigma}\right) &= 0.8849 \Rightarrow \frac{13.8-\mu}{\sigma} = 1.2 \Rightarrow \mu + 1.2\sigma = 13.8 \\
 &\dots \quad (A)
 \end{aligned}$$

$$\begin{aligned}
 P(X \geq 9.6) &= .8159 \Rightarrow 1 - \Phi\left(\frac{9.6-\mu}{\sigma}\right) = .8159 \Rightarrow \Phi\left(-\frac{9.6-\mu}{\sigma}\right) = .8159 \\
 \Rightarrow -\frac{9.6-\mu}{\sigma} &= 0.9 \Rightarrow \mu - 0.9\sigma = 9.6 \dots \quad (B) \\
 (A) + (B) &\Rightarrow \mu = 9.6, \quad \sigma = 3.5 \\
 P(X \geq 13.8 | X \geq 9.6) &= 1 - P(X < 13.8 | X \geq 9.6) = 1 - \frac{P(9.6 \leq X < 13.8)}{P(X \geq 9.6)} \\
 &= 1 - \frac{.7008}{.8159} = .1411.
 \end{aligned}$$

Problem No. 16

$$\sqrt{2\pi} (1 - \Phi(x)) = \int_x^{\infty} e^{-t^2/2} dt = \int_x^{\infty} \frac{1}{t} \frac{(t e^{-t^2/2})}{DF} dt$$

Using $\frac{d}{dt} (-e^{-t^2/2}) = t e^{-t^2/2}$ we get

$$\sqrt{2\pi} (1 - \Phi(x)) = \frac{e^{-x^2/2}}{x} - \int_x^{\infty} \frac{1}{t^2} e^{-t^2/2} dt \leq \frac{e^{-x^2/2}}{x}$$

Also

$$\begin{aligned}
 \sqrt{2\pi} (1 - \Phi(x)) &= \frac{e^{-x^2/2}}{x} - \int_x^{\infty} \frac{1}{t^3} (t e^{-t^2/2} dt) = \frac{e^{-x^2/2}}{x} - \frac{e^{-x^2/2}}{x^2} + \int_x^{\infty} \frac{3}{t^4} e^{-t^2/2} dt \\
 &\geq \frac{e^{-x^2/2}}{x} - \frac{e^{-x^2/2}}{x^3}.
 \end{aligned}$$

Problem No. 17

$$M_X(t) = \sum_{j=0}^{\infty} \frac{t^j}{j!} E(X^j) = \sum_{j=0}^{\infty} t^j = (1-t)^{-1}, \quad |t| < \theta, \quad t > 0$$

$\Rightarrow X \sim \text{Exp}(1) \Rightarrow X$ is A.c. with p.d.f. $f_{X|A}(x) = \begin{cases} e^{-x} & \text{if } x > 0 \\ 0 & \text{O.w.} \end{cases}$

Problem No. 18

$$f_{X,Y}(x,y) = \frac{1}{\Gamma(\alpha_1)\Gamma(\alpha_2)} x^{\alpha_1-1} y^{\alpha_2-1} e^{-\frac{x+y}{\theta}}, \quad (x,y) \in S_{X,Y} = [0, \theta]^2$$

(a) Consider the H transformation $(x,y) \rightarrow (U,V)$ defined by

$$U = x+y, \quad V = \frac{x}{x+y} \Rightarrow x = UV, \quad y = U(1-U).$$

$$S_{U,V} = \{(u,v) : uv \geq 0, u(1-u) \geq 0\} = \{(u,v) : u \geq 0, 0 \leq u \leq 1\} = [0, \theta] \times [0, 1].$$

$$J = \begin{vmatrix} u & u \\ 1-u & -u \end{vmatrix} = -u$$

Therefore $f_{U,V}(u,v) = \frac{1}{\Gamma(\alpha_1)\Gamma(\alpha_2)\theta^{\alpha_1+\alpha_2}} (uv)^{\alpha_1-1} (u(1-u))^{\alpha_2-1} e^{-\frac{u}{\theta}(1-u)}, \quad I_{S_{U,V}}^{(u,v)}$.

$$f_{U,V}(u,v) = \begin{cases} \left(\frac{1}{\Gamma(\alpha_1)\alpha_2} u^{\alpha_1-1} e^{-\frac{u}{\theta}} \right) \left(\frac{1}{\beta(\alpha_1, \alpha_2)} u^{\alpha_2-1} (1-u)^{\alpha_2-1} \right), & u \geq 0, v \geq u, \\ 0, & \text{O.w.} \end{cases}$$

$\Rightarrow U \sim \text{Gamma}(\alpha_1 + \alpha_2, \theta)$ and $V \sim \text{Beta}(\alpha_1, \alpha_2)$ are independent.

(b) Using (a), we can write $W = \frac{X}{X+Y}$ and $T = X+Y$,

where $X \sim \text{Gamma}(\alpha_1, \theta)$ and $Y \sim \text{Gamma}(\alpha_2, \theta)$ are

independent. Then $WT = X \sim \text{Gamma}(\alpha_1, \theta)$.

Problem 19

$X \sim N(0,1)$ and $Y^2 \sim \chi_1^2$ are independent

$$\Rightarrow \frac{X}{\sqrt{Y^2}} \sim t_1 \quad (\text{Student t distribution with 1 d.f.})$$

$$\Rightarrow Z = \frac{X}{\sqrt{Y^2}} \sim t_1 \quad (\text{or Cauchy distribution})$$

The p.d.f. of Z is $f_Z(z) = \frac{1}{\pi} \cdot \frac{1}{1+z^2}, \quad -\infty < z < \infty$

Thus $U = \frac{|X|}{\sqrt{Y^2}} = |Z|$ has p.d.f. $f_{U|u} = \begin{cases} \frac{2}{\pi} \cdot \frac{1}{1+u^2}, & \text{if } u > 0 \\ 0, & \text{O.w.} \end{cases}$

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(use Jacobian method)

Problem No. 20

$$f_{\underline{x}}(x_1, \dots, x_n) = \prod_{i=1}^n f(x_i) = e^{-\sum_{i=1}^n x_i}, \quad \underline{x} \in S_{\underline{x}} = [0, a]^n$$

Consider the 1-1 transformation $(x_1, \dots, x_n) \rightarrow (y_1, \dots, y_n)$
defined by

$$y_1 = \frac{x_1}{x_1 + x_2}, \quad y_2 = \frac{x_1 + x_2}{x_1 + x_2 + x_3}, \quad y_3 = \frac{x_1 + x_2 + x_3}{x_1 + x_2 + x_3 + x_4}, \dots, \quad y_n = \frac{\sum_{i=1}^n x_i}{\sum_{i=1}^n x_i},$$

$$y_n = \sum_{i=1}^n x_i, \quad \text{i.e.}$$

$$x_1 = \prod_{i=1}^n y_i, \quad x_2 = \left(\prod_{i=2}^n y_i\right)(1-y_1), \quad x_3 = \left(\prod_{i=3}^n y_i\right)(1-y_1)(1-y_2), \dots, \quad x_n = y_{n-1} y_n (1-y_{n-1}),$$

$$x_n = y_n (1-y_{n-1}).$$

$$S_{\underline{y}} = \{ \underline{y} : \prod_{i=1}^n y_i > 0, (\prod_{i=2}^n y_i)(1-y_1) > 0, (\prod_{i=3}^n y_i)(1-y_1)(1-y_2) > 0, \dots, y_{n-1} y_n (1-y_{n-1}) > 0, y_n (1-y_{n-1}) > 0 \}$$

$$= \{ \underline{y} : 0 \leq y_i \leq 1, i=1, \dots, n-1, \quad y_n > 0 \} = [0, 1]^{n-1} \times [0, \infty)$$

$$J = \begin{vmatrix} y_2 \cdots y_n & y_1 y_3 \cdots y_n & y_1 y_2 y_4 \cdots y_n \cdots y_1 \cdots y_{n-3} y_{n-1} y_n & y_1 \cdots y_{n-2} y_{n-1} y_n \\ -y_2 \cdots y_n & y_3 \cdots y_n (1-y_1) & y_2 y_4 \cdots y_n (1-y_1) \cdots y_2 \cdots y_{n-3} y_{n-1} y_n (1-y_1) & y_2 \cdots y_{n-2} y_{n-1} (1-y_1) y_2 \cdots y_n \\ 0 & -y_3 \cdots y_n & y_4 \cdots y_n (1-y_2) \cdots y_3 \cdots y_{n-3} y_{n-1} y_n (1-y_2) & y_3 \cdots y_{n-2} y_{n-1} (1-y_2) y_3 \cdots y_n \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots y_{n-1} y_n (1-y_{n-1}) y_{n-2} y_n (1-y_{n-1}) \cdots \\ 0 & 0 & 0 & \cdots -y_{n-1} y_n & y_n (1-y_{n-1}) y_{n-1} (1-y_{n-1}) \\ 0 & 0 & 0 & \cdots 0 & -y_n & 1-y_n \end{vmatrix}$$

$$= (-1)^{n-1} (-y_2 y_3 \cdots y_n) (-y_3 \cdots y_n) \cdots (-y_{n-1} y_n) (-y_n)$$

(on performing $R_1 \rightarrow R_1 + \cdots + R_n$)
and expanding

$$= y_2 y_3 \cdots y_{n-1} y_n^{n-1}$$

Thus

$$f_{\underline{y}}(\underline{y}) = y_2 y_3 \cdots y_{n-1} y_n^{n-1} e^{-\sum y_i} I_{S_{\underline{y}}}(\underline{y}) = \prod_{i=1}^n g_i(y_i) \pm S_{\underline{y}}$$

$$\text{where } g_i(y_i) = \begin{cases} i y_i^{i-1}, & \text{if } 0 < y_i < 1 \\ 0, & \text{otherwise} \end{cases}, \quad i=1, \dots, n-1, \quad g_n(y_n) = \begin{cases} \frac{e^{-y_n} y_n^{n-1}}{1-y_n}, & \text{if } y_n > 0 \\ 0, & \text{otherwise} \end{cases}$$

$\Rightarrow \gamma_1, \dots, \gamma_n$ are independent with $f_{\gamma_i}(y) = \begin{cases} e^{-y}, & y > 0 \\ 0, & \text{otherwise} \end{cases}$

$i=1, \dots, n$ and $f_{\gamma_n}(y) = \begin{cases} \frac{1}{\sqrt{\pi}} e^{-y} y^{n-1}, & \text{if } y > 0 \\ 0, & \text{o.w.} \end{cases}$

Problem No. 21 (a) We know that \bar{x} and s^2 are independent, $\bar{x} \sim N(\mu, \sigma^2)$ and $\frac{(n-1)s^2}{\sigma^2} \sim \chi^2_{n-1}$. Therefore

$$\begin{aligned} E\left(\frac{\bar{x}}{s}\right) &= E\left(\bar{x} \cdot \frac{1}{s}\right) = E(\bar{x}) E\left(\frac{1}{s}\right) = \mu E\left(\frac{1}{s}\right) \\ \frac{(n-1)s^2}{\sigma^2} \sim \chi^2_{n-1} &\Rightarrow E\left(\frac{1}{\sqrt{n-1}s}\right) = E\left(\frac{1}{\sqrt{\chi^2_{n-1}}}\right) = \frac{1}{2^{\frac{n-1}{2}} \Gamma\left(\frac{n-1}{2}\right)} \int_0^\infty \frac{1}{\sqrt{\pi}} e^{-\frac{x}{2}} x^{\frac{n-1}{2}-1} dx \\ &= \frac{2^{\frac{n-2}{2}} \Gamma\left(\frac{n-2}{2}\right)}{2^{\frac{n-1}{2}} \Gamma\left(\frac{n-1}{2}\right)} \Rightarrow E\left(\frac{1}{s}\right) = \frac{1}{2} \sqrt{\frac{n-1}{2}} \frac{\Gamma\left(\frac{n-2}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)} \\ \Rightarrow E\left(\frac{\bar{x}}{s}\right) &= \sqrt{\frac{n-1}{2}} \frac{\Gamma\left(\frac{n-2}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)} \frac{\mu}{\sigma} \end{aligned}$$

(b) Let $\gamma_1 = x_1 + x_2$ and $\gamma_2 = x_1 - x_2$. Then

$$\begin{aligned} M_{\gamma_1, \gamma_2}(t_1, t_2) &= E[e^{t_1 \gamma_1 + t_2 \gamma_2}] = E[e^{(t_1+t_2)x_1} e^{(t_1-t_2)x_2}] \\ &= E[e^{(t_1+t_2)x_1}] E[e^{(t_1-t_2)x_2}] \\ &= M_{x_1}(t_1+t_2) M_{x_2}(t_1-t_2) \quad (\text{x_1 and x_2 independent}) \end{aligned}$$

$$\Rightarrow M_{\gamma_1, \gamma_2}(t_1, t_2) = e^{\frac{(t_1+t_2)^2}{2}} e^{\frac{(t_1-t_2)^2}{2}} = e^{\frac{2+t^2}{2}} e^{\frac{2+L^2}{2}}, \quad (t_1, t_2) \in \mathbb{R}^2$$

$\Rightarrow \gamma_1$ and γ_2 are independent, $\gamma_1 \sim N(10, 2)$, $\gamma_2 \sim N(0, 2)$

$\Rightarrow \frac{\gamma_1}{\sqrt{2}} \sim N(0, 1)$, $\frac{\gamma_2}{\sqrt{2}} \sim N(0, 1)$ are independent

$\Rightarrow \frac{\gamma_1}{\sqrt{2}} \sim N(0, 1)$ and $\frac{\gamma_2^2}{2} \sim \chi^2_1$ are independent

$$\Rightarrow Z = \frac{\gamma_1}{\sqrt{2}} = \frac{\frac{\gamma_1}{\sqrt{2}}}{\sqrt{\frac{\gamma_2^2}{2}}} = \frac{N(0, 1)}{\sqrt{\chi^2_1 / 1}} \text{ independent}$$

$\Rightarrow Z \sim t_1$ (t distribution with 1 d.f. or Cauchy dist.)

$$\Rightarrow f_Z(z) = \frac{1}{\pi} \cdot \frac{1}{1+z^2}, \quad -\infty < z < \infty$$

Problem No. 22 $f_{\underline{x}}(\underline{x}) = \prod_{i=1}^n f(x_i) = e^{-\sum_{i=1}^n x_i}, \underline{x} \in S_{\underline{x}} = [0, a]^n.$

By Problem 3, joint p.d.f. of $(x_{(1)}, \dots, x_{(n)})$ is

$$g(y_1, \dots, y_n) = \begin{cases} \ln e^{-\sum_{i=1}^n y_i}, & \text{if } 0 < y_1 < \dots < y_n < a \\ 0, & \text{otherwise.} \end{cases}$$

Consider the 1-1 transformation $(x_{(1)}, \dots, x_{(n)}) \rightarrow (z_1, \dots, z_n)$ defined by $z_1 = n x_{(1)}, z_2 = (n-1)(x_{(2)} - x_{(1)}), z_i = (n-i+1)$

$(x_{(1)} - x_{(i-1)}), i = 3, \dots, n$, i.e.,

$$x_{(1)} = \frac{z_1}{n}, x_{(2)} = \frac{z_1}{n} + \frac{z_2}{n-1}, x_{(3)} = \frac{z_1}{n} + \frac{z_2}{n-1} + \frac{z_3}{n-2}, \dots,$$

$$x_{(n)} = \frac{z_1}{n} + \frac{z_2}{n-1} + \dots + \frac{z_{n-1}}{2} + z_n$$

$$S_{\underline{z}} = \{\underline{z} : 0 \leq \frac{z_1}{n} < \frac{z_1}{n} + \frac{z_2}{n-1} \leq \frac{z_1}{n} + \frac{z_2}{n-1} + \frac{z_3}{n-2} \leq \dots \leq \frac{z_1}{n} + \frac{z_2}{n-1} + \dots + \frac{z_{n-1}}{2} + z_n\}$$

$$= \{\underline{z} : z_1 \geq 0, z_2 \geq 0, \dots, z_n \geq 0\}$$

$$J = \begin{vmatrix} \frac{1}{n} & 0 & 0 & \dots & 0 & 0 & 0 \\ \frac{1}{n} & \frac{1}{n-1} & 0 & \dots & 0 & 0 & 0 \\ \frac{1}{n} & \frac{1}{n-1} & \frac{1}{n-2} & \dots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \frac{1}{2} & \frac{1}{2} & 1 \end{vmatrix} = \frac{1}{n!}$$

Thus $f_{\underline{z}}(\underline{z}) = \frac{1}{n!} \ln e^{-\sum_{i=1}^n z_i} \cdot \mathbb{1}_{S_{\underline{z}}}(\underline{z})$

$$= \begin{cases} \prod_{i=1}^n e^{-z_i}, & \text{if } z_i \geq 0, i = 1, \dots, n \\ 0, & \text{otherwise.} \end{cases}$$

$\Rightarrow z_1, z_2, \dots, z_n$ are c.c.d. EX 1(1)

(b) $x_{(r)} = \frac{z_1}{n} + \frac{z_2}{n-1} + \dots + \frac{z_r}{n-r+1}$

$$x_{(n)} = \frac{z_1}{n} + \frac{z_2}{n-1} + \dots + \frac{z_r}{n-r+1} + \frac{z_{r+1}}{n-r} + \dots + \frac{z_n}{n-n+1} \quad (r < n)$$

$$\Rightarrow E(X_{(r)}) = \frac{1}{h} + \frac{1}{h-1} + \cdots + \frac{1}{h-r+1}$$

$$\text{Var}(X_{(r)}) = \frac{1}{h^2} + \frac{1}{(h-1)^2} + \cdots + \frac{1}{(h-r+1)^2}$$

$$\text{Cov}(X_{(r)}, X_{(n)}) = \text{Cov}\left(\frac{Z_1}{h} + \cdots + \frac{Z_r}{h-r+1}, \frac{Z_1}{h} + \cdots + \frac{Z_n}{h-n+1}\right)$$

$$= \text{Var}(X_{(r)}) = \frac{1}{h^2} + \frac{1}{(h-1)^2} + \cdots + \frac{1}{(h-r+1)^2}$$

$$(c) \quad X_{(r)} = \sum_{i=1}^r \frac{Z_i}{h-i+1}, \quad X_{(n)} - X_{(r)} = \sum_{i=r+1}^n \frac{Z_i}{h-i+1}, \quad 1 \leq r < n \leq n$$

From (a) it follows that (Z_1, \dots, Z_r) and (Z_{r+1}, \dots, Z_n) are independent $\Rightarrow X_{(r)}$ and $X_{(n)} - X_{(r)}$ are independent

$$(d) \quad Z_{r+1} = (h-r)(X_{(r+1)} - X_{(r)}) \sim \text{Exp}(1) \quad (\text{by (a)})$$

$$\Rightarrow f_{Z_{r+1}}(z) = \begin{cases} e^{-z} & \text{if } z > 0 \\ 0 & \text{o.w.} \end{cases}$$

Therefore the p.d.f. of $S = X_{(r+1)} - X_{(r)}$ is

$$f_S(s) = \begin{cases} (h-r)e^{-(h-r)s} & \text{if } s > 0 \\ 0 & \text{o.w.} \end{cases}$$

Problem No. 23 By Problem 3 η_n is of A.C. with p.d.f.

$$f_{\eta_n}(x) = n F^{(n)}(x) f(x), \quad -\infty < x < \infty,$$

where f is the p.d.f. of X_1 .

Clearly the d.f. of η_n is

$$\begin{aligned} F_{\eta_n}(x) &= P(\eta_n \leq x) = P(X_1 \leq x, X_2 \leq x, \dots, X_n \leq x) \\ &= \prod_{i=1}^n P(X_i \leq x) = [F(x)]^n, \quad -\infty < x < \infty. \end{aligned}$$

Also $P(\eta_n > 0) = 1$. Therefore

$$\begin{aligned} E(\eta_n) &= \int_0^\infty [1 - F_{\eta_n}(x)] dx = \int_0^\infty [1 - (F(x))^n] dx \\ &= \int_0^\infty [1 - (F(x))^{n-1}] dx + \int_0^\infty (F(x))^{n-1} (1 - F(x)) dx \\ &= E(\eta_{n-1}) + \int_0^\infty (F(x))^{n-1} (1 - F(x)) dx. \end{aligned}$$

Problem No. 24 (a)

$$\begin{aligned}
 M_{U,V}(t_1, t_2) &= E(e^{t_1 U + t_2 V}) = E(e^{(t_1 a + t_2 b)x} e^{(t_1 b - t_2 a)y}) \\
 &= E(e^{(t_1 a + t_2 b)x}) E(e^{(t_1 b - t_2 a)y}) \quad (x \text{ and } y \text{ are indept}) \\
 &= M_x(t_1 a + t_2 b) M_y(t_1 b - t_2 a) \\
 &= e^{\frac{(t_1 a + t_2 b)^2 \sigma^2}{2}} e^{\frac{(t_1 b - t_2 a)^2 \sigma^2}{2}} \\
 &= e^{\frac{(a+b)^2 \sigma^2 + t_1^2}{2}} e^{\frac{(a-b)^2 \sigma^2 + t_2^2}{2}} \\
 &= M_{U,V}(t_1, 0) M_{U,V}(0, t_2), \quad (t_1, t_2) \in \mathbb{R}^2
 \end{aligned}$$

$\Rightarrow U \sim N(0, (a^2 + b^2)\sigma^2)$ and $V \sim N(0, (a^2 + b^2)\sigma^2)$ are i.i.d.

(b) Follows from (a) by taking $a=b=\frac{1}{\sqrt{2}}$.

Problem No. 25 First suppose that $X \sim U(\alpha, \beta)$. Then

$$f_X(x) = \begin{cases} \frac{1}{\beta-\alpha}, & \text{if } \alpha < x < \beta \\ 0, & \text{otherwise.} \end{cases}$$

$$\text{clearly } P(\alpha < X \leq \beta) = \int_{\alpha}^{\beta} f_X(x) dx = 1 \quad \subseteq (\alpha, \beta)$$

Also if $I = (c, d)$ ($c < d$) or $[c, d]$, $(c, d]$, $[c, d)$ then

$$P(X \in I) = \int_c^d f_X(x) dx = \frac{d-c}{\beta-\alpha} \rightarrow \text{depends only on length of } I$$

Conversely suppose that $P(\alpha < X \leq \beta) = 1$ and $P(X \in I) = P(X \in J)$ for any pair of subintervals $I, J \subseteq (\alpha, \beta)$ having the same length. Then clearly

$$F_X(x) = P(X \leq x) = 1 - P(X > x) \leq 1 - P(\alpha < X \leq \beta) = 0$$

$$\Rightarrow F_X(x) = 0. \text{ And } F_X(\beta) = P(X \leq \beta) \geq P(\alpha < X \leq \beta) = 1 \Rightarrow F_X(\beta) = 1.$$

Thus $F_X(x) = 0$ if $x \leq \alpha$ and $F_X(x) = 1$ if $x \geq \beta$.

For $\alpha < x < \beta$, $F_X(x) = P(X \leq x) = P(\alpha < X \leq x + (\beta - x)) = h(x - \alpha)$,

where $h(\lambda) = P(\alpha < X \leq \alpha + \lambda)$, $0 < \lambda \leq \beta - \alpha$.

$$\begin{aligned}
 h(\lambda_1 + \lambda_2) &= P(\alpha < x \leq \alpha + \lambda_1 + \lambda_2) = P(\alpha < x \leq \alpha + \lambda_1) + P(\alpha + \lambda_1 < x \leq \alpha + \lambda_1 + \lambda_2) \\
 &= h(\lambda_1) + h(\lambda_2) \quad (\text{Since } P(x \in I) \text{ depends only} \\
 &\quad \text{on length of } I) \\
 &= h(\lambda_1) + h(\lambda_2), \quad \lambda_1 \geq 0, \quad \lambda_2 \geq 0, \quad \lambda_1 + \lambda_2 \leq \beta - \alpha
 \end{aligned}$$

Define $g(x) = h((\beta - \alpha)x)$, $0 < x \leq 1$.

$$\text{Then } g(\lambda_1 + \lambda_2) = g(\lambda_1) + g(\lambda_2), \quad \lambda_1 \geq 0, \quad \lambda_2 \geq 0, \quad \lambda_1 + \lambda_2 \leq 1$$

By induction

$$g(\lambda_1 + \lambda_2 + \dots + \lambda_n) = g(\lambda_1) + g(\lambda_2) + \dots + g(\lambda_n), \quad \lambda_i \geq 0, \quad i=1(1)n$$

$$\lambda_1 + \dots + \lambda_n \leq 1, \quad n \geq 1 \dots$$

$$\Rightarrow g(x) \geq n g\left(\frac{x}{n}\right), \quad 0 < x \leq 1 \quad (\text{by taking } \lambda_i = \frac{x}{n}, \quad i=1 \dots n)$$

$$g(mx) = m g(1), \quad 0 < y \leq \frac{1}{m} \quad (\text{by taking } m=n, \text{ and } \lambda_i = y, \quad i=1(1)m)$$

Now let $m, n \in \{\frac{1}{2}, \dots, 1\}$ and $\frac{m}{n} \in (0, 1)$ (i.e. $m < n$)

$$g\left(\frac{m}{n}\right) = m g\left(\frac{1}{n}\right) \quad (\text{from (B)}, \quad \text{since } \frac{1}{n} < \frac{1}{m})$$

$$= \frac{m}{n} g(1) \quad (\text{from (A)})$$

$$\Rightarrow g(r) = r g(1), \quad \forall \text{ rational } r \in (0, 1)$$

Fix $x \in (0, 1)$. Choose sequences $\{u_n\}_{n \geq 1}$ and $\{v_n\}_{n \geq 1}$

of rationals in $(0, 1)$ such that $u_n \uparrow x$ and $v_n \downarrow x$.

Note that $u_n \uparrow$ and therefore

$$g(u_n) \leq g(x) \leq g(v_n), \quad \forall n \geq 1$$

$$\Rightarrow u_n g(1) \leq g(x) \leq v_n g(1), \quad \forall n \geq 1$$

$$\Rightarrow x g(1) \leq g(x) \leq x g(1) \quad (\text{by taking limits})$$

$$\Rightarrow g(x) = x g(1), \quad \forall x \in (0, 1)$$

$$\Rightarrow h((\beta - \alpha)x) = x h(\beta - \alpha) = x, \quad \forall x \in (0, 1)$$

$$\text{Hence } h(\beta - \alpha) = P(\alpha < x \leq \beta) = 1$$

$$\Rightarrow h(\lambda) = \frac{\lambda}{\beta - \alpha}, \quad \forall 0 < \lambda \leq \beta - \alpha$$

$$\Rightarrow F_x(\alpha + \lambda) = F_x(\alpha + \lambda) - F_x(\alpha) = P(\alpha < x \leq \alpha + \lambda) = h(\lambda) = \frac{\lambda}{\beta - \alpha},$$

$$\quad \forall 0 < \lambda \leq \beta - \alpha$$

$$\Rightarrow F_x(x) = \frac{x - \alpha}{\beta - \alpha}, \quad \forall \alpha < x \leq \beta$$

$$\Rightarrow F_x(x) = \begin{cases} 0 & \text{if } x < \alpha \\ \frac{x - \alpha}{\beta - \alpha} & \text{if } \alpha \leq x < \beta \\ 1 & \text{if } x \geq \beta \end{cases} \Rightarrow X \sim U(\alpha, \beta)$$