Bios 661: 1-5; Bios 673: 2-6.

- 1. C&B 4.19
- 2. C&B 4.23
- 3. C&B 5.6
- 4. Suppose (X_1, X_2) have the joint pdf

$$f_{X_1, X_2}(x_1, x_2) = \begin{cases} 1 & 0 < x_1 < 1, & 0 < x_2 < 1 \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Let the support of (X_1, X_2) be denoted by the set $S = \{(x_1, x_2) : 0 < x_1 < 1, 0 < x_2 < 1\}$. Draw S on the xy-plane.
- (b) Suppose $Y_1 = X_1 + X_2$ and $Y_2 = X_1 X_2$ with a support \mathcal{T} . Draw \mathcal{T} on the xy-plane.

Solution: The inverse of the transformation is $X_1 = \frac{1}{2}(Y_1 + Y_2)$ and $X_2 = \frac{1}{2}(Y_1 - Y_2)$. Therefore,

$$\mathcal{T} = \{ (y_1, y_2) : 0 < \frac{1}{2} (y_1 + y_2) < 1, \quad 0 < \frac{1}{2} (Y_1 - Y_2) < 1 \}$$

= \{ (y_1, y_2) : -y_1 < y_2, \quad y_2 < 2 - y_1, \quad y_2 < y_1, \quad y_1 - 2 < y_2 \}.

(c) Derive the joint distribution of (Y_1, Y_2) using Jacobin method.

Solution:

$$f_{Y_1,Y_2}(y_1,y_2) = \begin{cases} f_{X_1,X_2}(\frac{1}{2}(y_1+y_2), \frac{1}{2}(y_1-y_2))|J| = \frac{1}{2} & (y_1,y_2) \in \mathcal{T}, \\ 0 & \text{otherwise.} \end{cases}$$

(d) Derive the marginal distribution (pdf) of Y_1 and Y_2 .

Solution:

$$f_{Y_1}(y_1) = \begin{cases} \int_{-y_1}^{y_1} \frac{1}{2} dy_2 = y_1 & 0 < y_1 \le 1\\ \int_{y_1-2}^{2-y_1} \frac{1}{2} dy_2 = 2 - y_1 & 1 < y_1 < 2\\ 0 & \text{otherwise.} \end{cases}$$

Similarly,

$$f_{Y_2}(y_2) = \begin{cases} \int_{-y_2}^{y_2+2} \frac{1}{2} dy_1 = y_2 + 1 & -1 < y_2 \le 0\\ \int_{y_2}^{2-y_2} \frac{1}{2} dy_1 = 1 - y_2 & 0 < y_2 < 1\\ 0 & \text{otherwise.} \end{cases}$$

5. Suppose that random variables X_1 and X_2 are mutually independent and follow N(0,1). Show that the random variable $Y_1 = X_1/X_2$ follows Cauchy distribution with pdf

$$f_{Y_1}(y_1) = \frac{1}{\pi} \frac{1}{1 + y_1^2}, \quad -\infty < y_1 < \infty.$$

Answer the following questions step by step toward the final solution.

(a) Let $Y_2 = X_2$. Show that the Jacobian of the inverse function of y_1 and y_2 is y_2 , where $-\infty < y_2 < \infty$.

Solution: The inverse function of y_1 and y_2 is $X_1 = Y_1Y_2$ and $X_2 = Y_2$ with Jacobian matrix

$$J = \left| \begin{array}{cc} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{array} \right| = \left| \begin{array}{cc} y_2 & y_1 \\ 0 & 1 \end{array} \right|.$$

The determinant is y_2 . Pay attend to the domain of y_2 , which is from $-\infty$ to ∞ .

(b) Derive the joint pdf of Y_1 and Y_2 using the Jacobian method.

Solution: The joint pdf of Y_1 and Y_2 is

$$\begin{split} f_{Y_1,Y_2}(y_1,y_2) &= f_{X_1,X_2}(y_1y_2,y_2)|y_2| \\ &= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{y_1^2y_2^2}{2}\right) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{y_2^2}{2}\right)|y_2|. \end{split}$$

(c) Find the marginal distribution of Y_1 from the joint distribution of Y_1 and Y_2 in (b).

Solution: The marginal distribution of Y_1 is

$$f_{Y_1}(y_1) = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{y_1^2 y_2^2}{2}\right) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{y_2^2}{2}\right) |y_2| dy_2$$

$$= \frac{1}{\pi} \int_0^{\infty} \exp\left(-\frac{(1+y_1^2)y_2^2}{2}\right) y_2 dy_2$$

$$(\text{set } t = y_2^2)$$

$$= \frac{1}{2\pi} \int_0^{\infty} \exp\left(-\frac{(1+y_1^2)t}{2}\right) dt$$

$$= \frac{1}{\pi} \frac{1}{1+y_1^2},$$

where $-\infty < y_1 < \infty$.

6. Let X_1, \ldots, X_n constitute a random sample of size $n(n \geq 3)$ from the parent population

$$f_X(x) = \lambda e^{-\lambda x}, \quad 0 < x < \infty, \quad 0 < \lambda < \infty$$

(a) Find the conditional density function of X_1, \ldots, X_n given that $S = \sum_{i=1}^n X_i = s$.

Solution: Since X_1, \ldots, X_n follows exponential distribution with mean λ^{-1} , $S = \sum_{i=1}^{n} X_i$ follows Gamma distribution with pdf

$$f_S(s) = \frac{1}{\Gamma(n)\lambda^{-n}} s^{n-1} e^{-s\lambda}.$$

The conditional distribution of X_1, \ldots, X_n given that S = s is

$$f_{X_1,...,X_n}(x_1,...,x_n|S=s) = \frac{f_{X_1,...,X_n,S}(x_1,...,x_n,s)}{f_S(s)}$$

$$= \frac{f_{X_1,...,X_n}(x_1,...,x_n)}{f_S(s)}$$

$$= \frac{(n-1)!}{s^{n-1}}.$$

(b) Consider the (n-1) random variables

$$Y_1 = \frac{X_1}{S}, Y_2 = \frac{X_1 + X_2}{S}, \dots, Y_{n-1} = \frac{X_1 + X_2 + \dots + X_{n-1}}{S}.$$

Find the joint distribution of Y_1, Y_2, \dots, Y_{n-1} given that S = s.

Solution: The inverse function of the transformation is

$$X_1 = Y_1 S$$
, $X_2 = (Y_2 - Y_1) S$, ..., $X_{n-1} = (Y_{n-1} - Y_{n-2}) S$.

The Jacobian is S^{n-1} . Hence, the joint pdf of Y_1, Y_2, \dots, Y_{n-1} given that S = s is

$$f_{Y_1,Y_2,\dots,Y_{n-1}}(y_1,y_2,\dots,y_{n-1}|S=s)$$

$$=f_{X_1,X_2,\dots,X_{n-1}}(y_1s,(y_2-y_1)s,\dots,(y_{n-1}-y_{n-2})s|S=s)s^{n-1}$$

$$=f_{X_1,X_2,\dots,X_n}(y_1s,(y_2-y_1)s,\dots,(y_{n-1}-y_{n-2})s,(1-y_{n-1})s|S=s)s^{n-1}$$

$$=\frac{(n-1)!}{s^{n-1}}s^{n-1}=(n-1)!, \quad y_1 < y_2 < \dots < y_{n-1}.$$

The second equation stands because $f_{X_1,X_2,\dots,X_{n-1}}(x_1,x_2,\dots,x_{n-1}|S=s)=f_{X_1,X_2,\dots,X_n}(x_1,x_2,\dots,x_n|S=s)$ since s is given and when one knows $x_1,\dots,x_{n-1},x_n=s-\sum_{i=1}^{n-1}x_i$ is fixed, i.e., no randomness in X_n .

(c) When n=3 and when n=4, find the marginal distribution of Y_1 given that S=s, and then use these results to infer the structure of the marginal distribution of Y_1 given that S=s for any $n\geq 3$.

Solution: When n=3,

$$f_{Y_1}(y_1|S=s) = \int_{y_2}^1 2! dy_2 = 2(1-y_1), \quad 0 < y_1 < 1.$$

When n=4,

$$f_{Y_1}(y_1|S=s) = \int_{y_1}^1 \int_{y_2}^1 3! dy_3 dy_2 = 3(1-y_1)^2, \quad 0 < y_1 < 1.$$

For a general n,

$$f_{Y_1}(y_1|S=s) = (n-1)(1-y_1)^{n-2}, \quad 0 < y_1 < 1.$$

7. (Bios 673 class material) A certain simple biological system involves exactly two independently functioning components. If one of these two components fails, then entire systems fails. For i = 1, 2, let Y_i be the random variable representing the time to failure of the ith component, with the pdf of Y_i being

$$f_{Y_i}(y_i) = \theta_i e^{-\theta_i y_i}, \quad 0 < y_i < \infty, \quad \theta_i > 0.$$

Clearly, if this biological system fails, then only two random variables are observable, namely U and W, where $U = \min(Y_1, Y_2)$ and

$$W = \begin{cases} 1, & \text{if } Y_1 < Y_2, \\ 0, & \text{if } Y_2 < Y_1. \end{cases}$$

(a) Show that the joint distribution $f_{U,W}(u,w)$ of random variables U and W is

$$f_{U,W}(u,w) = \theta_1^{(1-w)} \theta_2^w e^{-(\theta_1 + \theta_2)u}, \quad 0 < u < \infty, \quad w = 0, 1.$$

Solution: Working on the derivation of $P(U \le u, W = 0)$ and $P(U \le u, W = 1)$, one have

$$\begin{split} P(U \leq u, W = 0) &= P(Y_2 \leq u, Y_2 < Y_1) \\ &= \int_0^u \int_{y_2}^\infty \left(\theta_1 e^{-\theta_1 y_1}\right) \left(\theta_2 e^{-\theta_2 y_2}\right) dy_1 dy_2 \\ &= \left(\frac{\theta_1}{\theta_1 + \theta_2}\right) \left\{1 - e^{-(\theta_1 + \theta_2)u}\right\} \end{split}$$

$$f_{U,W}(u, w = 0) = \theta_1 e^{-(\theta_1 + \theta_2)u}, \quad 0 < u < \infty.$$
 (1)

$$\begin{split} P(U \leq u, W = 1) &= P(Y_1 \leq u, Y_1 < Y_2) \\ &= \int_0^u \int_{y_1}^\infty \left(\theta_1 e^{-\theta_1 y_1}\right) \left(\theta_2 e^{-\theta_2 y_2}\right) dy_2 dy_1 \\ &= \left(\frac{\theta_2}{\theta_1 + \theta_2}\right) \left\{1 - e^{-(\theta_1 + \theta_2)u}\right\} \end{split}$$

$$f_{U,W}(u, w = 1) = \theta_2 e^{-(\theta_1 + \theta_2)u}, \quad 0 < u < \infty.$$
 (2)

Combining (1) and (2), one has

$$f_{U,W}(u,w) = \theta_1^{(1-w)} \theta_2^w e^{-(\theta_1 + \theta_2)u}, \quad 0 < u < \infty, \quad w = 0, 1.$$

(b) Find the marginal distribution $f_W(w)$ of the random variable W.

Solution: By definition of marginal distribution, one can have

$$f_W(w) = \int_0^\infty f_{U,W}(u, w) du = \int_0^\infty \theta_1^{(1-w)} \theta_2^w e^{-(\theta_1 + \theta_2)u} du$$
$$= \theta_1^{(1-w)} \theta_2^w (\theta_1 + \theta_2)^{-1} = \left(\frac{\theta_1}{\theta_1 + \theta_2}\right)^{(1-w)} \left(\frac{\theta_2}{\theta_1 + \theta_2}\right)^w$$

(c) Find the marginal distribution $f_U(u)$ of the random variable U.

Solution: Similar to (b), one can have

$$f_U(u) = \sum_{w=0}^{1} f_{U,W}(u, w) = (\theta_1 + \theta_2)e^{-(\theta_1 + \theta_2)u}.$$

(d) Show that U and W are independent.

Solution: Since $f_{U,W}(u,w) = f_U(u)f_W(w)$, one can claim U are W are independent.