

Problem 1. Suppose X and Y are independent with $X \sim \text{Uniform}(-1, 1)$ and $Y \sim \text{Normal}(0, 1)$. Define a new random variable Z by

$$Z = \begin{cases} X & \text{if } XY > 0 \\ -X & \text{if } XY < 0. \end{cases}$$

For $z > 0$

(a) (10%) What is the distribution of Z ? Prove your answer.

Solutions:
$$P(Z > z) = P(X > z)P(XY > 0) + P(-X > z)P(XY < 0)$$

$$= P(X > z)P(Y > 0) + P(X < -z)P(Y < 0)$$

Since $X \sim \text{Uniform}(-1, 1)$ and $Y \sim \text{Normal}(0, 1)$

$$P(Z > z) = P(X > z)P(Y > 0) + P(X < -z)P(Y < 0)$$

$$= P(X > z)$$

We can also get $P(Z > z) = P(X > z)$ for $z \leq 0$

So $Z \sim X$ $Z \sim \text{Uniform}(-1, 1)$

(b) (4%) Are Z and Y independent? Prove your answer.

Solution: Z and Y are not independent

Since
$$Z = \begin{cases} X & \text{if } XY > 0 \\ -X & \text{if } XY < 0. \end{cases}$$

when $Y > 0$, then $Z > 0$

when $Y < 0$, then $Z < 0$

the sign of Z depends on Y .

So they are not independent.

Problem 2. (20%) Suppose X_1 and X_2 are independent Normal($0, \sigma^2$) random variables. Find the joint density of

$$Y_1 = X_1^2 + X_2^2 \quad \text{and} \quad Y_2 = \frac{X_1^2}{X_1^2 + X_2^2}$$

solutions =

$$\begin{cases} Y_1 = X_1^2 + X_2^2 \\ Y_2 = \frac{X_1^2}{X_1^2 + X_2^2} \end{cases} \rightarrow \begin{cases} X_1^2 = Y_1 Y_2 \\ X_2^2 = Y_1 - Y_1 Y_2 \end{cases}$$

$$\begin{array}{c|c} & x_2 \\ \hline A_3 & A_1 \\ \hline A_4 & A_2 \end{array} \begin{array}{c} \\ \\ x_1 \end{array}$$

$$A_1 = \{x_1, x_2 : x_1 > 0, x_2 > 0\}, \quad A_2 = \{x_1, x_2 : x_1 > 0, x_2 < 0\}$$

$$A_3 = \{x_1, x_2 : x_1 < 0, x_2 > 0\}, \quad A_4 = \{x_1, x_2 : x_1 < 0, x_2 < 0\}$$

For A_1 , $x_1 = \sqrt{y_1 y_2}$, $x_2 = \sqrt{y_1(1-y_2)}$

$$J_1 = \begin{vmatrix} \frac{\sqrt{y_2}}{2\sqrt{y_1}} & \frac{\sqrt{y_1}}{2\sqrt{y_2}} \\ \frac{1}{2\sqrt{y_1}} \sqrt{\frac{1-y_2}{y_1}} & -\frac{1}{2\sqrt{y_1}} \sqrt{\frac{y_1}{1-y_2}} \end{vmatrix} = -\frac{1}{4} \left[\sqrt{\frac{y_2}{1-y_2}} + \sqrt{\frac{1-y_2}{y_2}} \right] = -\frac{1}{4\sqrt{1-y_2} \cdot \sqrt{y_2}}$$

$$\text{and } |J_1| = |J_2| = |J_3| = |J_4| = \frac{1}{4\sqrt{y_2(1-y_2)}}$$

$$f_{Y_1, Y_2}(y_1, y_2) = \sum_{i=1}^4 f_{X_1, X_2}(h_i(x), h_i(y)) \cdot |J_i|$$

$$= \frac{1}{\sqrt{y_2(1-y_2)}} \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{y_1 y_2}{2\sigma^2}} \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{y_1(1-y_2)}{2\sigma^2}}$$

$$= \frac{1}{2\pi\sigma^2} \frac{1}{\sqrt{y_2(1-y_2)}} e^{-\frac{y_1}{2\sigma^2}}$$

where

$$y_1 > 0 \quad \text{and} \quad 0 \leq y_2 \leq 1$$

Problem 3. The random pair (X, Y) has the distribution (joint mass function)

		X			
		1	2	3	
Y	1	1/6	1/12	1/12	$\frac{1}{3}$
	2	1/12	1/6	1/12	$\frac{1}{3}$
	3	1/12	1/12	1/6	$\frac{1}{3}$
		$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	

(a) (7%) Show that X and Y are dependent.

Solution: $P(X=1) = P(Y=1, X=1) + P(Y=2, X=1) + P(Y=3, X=1) = \frac{1}{3}$
 $P(Y=1) = P(X=1, Y=1) + P(X=2, Y=1) + P(X=3, Y=1) = \frac{1}{3}$
 $P(X=1, Y=1) = \frac{1}{6}$

$$P(X=1, Y=1) \neq P(X=1) \cdot P(Y=1)$$

So X and Y are dependent.

(b) (7%) Give a probability table for random variables U and V that have the same marginals as X and Y but are independent.

Solution: We get $P(U=1) = P(U=2) = P(U=3) = \frac{1}{3}$
 $P(V=1) = P(V=2) = P(V=3) = \frac{1}{3}$

		U		
		1	2	3
V	1	$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$
	2	$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$
	3	$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$

U and V are indep.

where

$$P(U=i, V=j) = P(U=i)P(V=j)$$

$$i = 1, 2, 3$$

$$j = 1, 2, 3$$

Problem 4.

(a) (6%) State the definition of a two-parameter exponential family (2pef), that is, give a general expression for the density function of a 2pef.

$$f(x) = h(x) c(\theta) \exp \left[\sum_{i=1}^2 w_i(\theta) t_i(x) \right]$$

(b) (6%) Does the Gamma(α, β) family with both α and β unknown form a 2pef? Justify your answer.

$$\begin{aligned} f(x) &= \frac{1}{\Gamma(\alpha) \beta^\alpha} x^{\alpha-1} e^{-x/\beta}, \quad 0 \leq x < \infty \\ &= \frac{1}{\Gamma(\alpha) \beta^\alpha} e^{-x/\beta} \cdot e^{(\alpha-1) \log x}, \quad \theta = (\alpha, \beta) \\ &= \underbrace{\Gamma(\alpha)}_{h(x)} \cdot \underbrace{\frac{1}{\Gamma(\alpha) \beta^\alpha}}_{c(\theta)} \exp \left[\underbrace{-\frac{1}{\beta}}_{w_1(\theta)} \cdot \underbrace{x}_{t_1(x)} + \underbrace{(\alpha-1)}_{w_2(\theta)} \cdot \underbrace{\log x}_{t_2(x)} \right] \end{aligned}$$

So Gamma(α, β) family with α and β unknown forms a 2pef.

(c) (5%) Does the Uniform(a, b) family with both a and b unknown form a 2pef? Justify your answer.

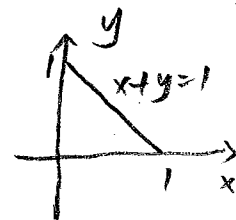
$$f(x) = \frac{1}{b-a}, \quad a \leq x \leq b, \quad \theta = (a, b)$$

Since the support $\{x: a \leq x \leq b\}$ depends on the parameter $\theta = (a, b)$, so the Uniform(a, b) doesn't form a 2pef.

Problem 5. Suppose X, Y have joint density $f_{X,Y}(x, y) = 6(1-x-y)$ in the region where $x > 0$, $y > 0$, and $x + y < 1$ (and zero outside this region).

(a) (5%) Find $f_X(x)$.

$$\begin{aligned} f_X(x) &= \int_0^{1-x} 6(1-x-y) dy \\ &= -3(1-x-y)^2 \Big|_0^{1-x} \\ &= 3(1-x)^2 \quad \checkmark \quad 0 < x < 1 \end{aligned}$$



(b) (5%) Find $f_{Y|X}(y|x)$ for $0 < x < 1$.

$$f(y|x) = \frac{f(x, y)}{f(x)} = \frac{6(1-x-y)}{3(1-x)^2} = \frac{2(1-x-y)}{(1-x)^2} \quad \checkmark$$

where $0 < x < 1$ and $0 < y < 1-x$ $\checkmark$

$\checkmark$ (c) (5%) Find $E(Y|X=x)$ for $0 < x < 1$.

$$\begin{aligned} E(Y|X=x) &= \int_0^{1-x} y \cdot f(y|x) dy \\ &= \int_0^{1-x} \frac{2y(1-x-y)}{(1-x)^2} dy \\ &= \frac{1}{(1-x)^2} \left[(1-x)y^2 - \frac{2}{3}y^3 \right] \Big|_0^{1-x} \\ &= \frac{1-x}{3} \quad \checkmark \quad \text{for } 0 < x < 1 \end{aligned}$$

No work is required in the remaining problems. You will receive full credit for stating the correct answers.

Problem 6.

- (a) (4%) State a general expression for $\text{Var}(Y|X)$ as a difference of two terms.

$$\text{Var}(Y|X) = E(Y^2|X) - [E(Y|X)]^2$$

- (b) (4%) Use the answer to part (a) and obtain an expression for $E[\text{Var}(Y|X)]$. Simplify where possible.

$$E[\text{Var}(Y|X)] = EY^2 - E\{[E(Y|X)]^2\}$$

- (c) (4%) Obtain an expression for $\text{Var}[E(Y|X)]$ as a difference of two terms. Simplify where possible.

$$\text{Var}[E(Y|X)] = E\{[E(Y|X)]^2\} - (EY)^2$$

Problem 7. Suppose $X_1, X_2, \dots, X_n$ are iid from a distribution with density f and cdf F .

(a) (4%) State a formula for the cdf of $X_{(j)}$, the j -th order statistic.

$$F_{X_{(j)}}(x) = \sum_{k=j}^n \binom{n}{k} [F(x)]^k [1-F(x)]^{n-k}$$

(b) (4%) State a formula for the density of $X_{(j)}$.

$$f_{X_{(j)}}(x) = n \binom{n-1}{j-1} f(x) \cdot [F(x)]^{j-1} \cdot [1-F(x)]^{n-j}$$