# Exercise 3 Probability Theory 2020 Autumn

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## 1 Problem 1

**No**. A counterexample is given as below.

Let  $\Omega$  be (0,1) and the probability measure  $\mathbb{P} = \lambda$ . Define  $X_n$  as

$$X_n = -\frac{1}{nx}, \quad x \in (0,1), n \geqslant 1$$
 (1)

So  $X_{n+1}(w) \geqslant X_n(w)$  for all  $n \geqslant 1$  and  $w \in \Omega$ .

And 
$$X = \lim_{n \to \infty} X_n = 0$$
. So  $E[X] = 0$  but  $E[X_n] = -\infty$ ,  $\lim_{n \to \infty} E[X_n] \neq E[X]$ 

## 2 Problem 2

#### 2.1 (a)

Using  $\int_{\mathbb{R}} f_X(x) dx = 1$ , we have

$$\int_{\mathbb{R}} f_X(x) dx = \int_{-2}^2 c\sqrt{4 - x^2} dx = 2\pi c = 1$$
 (2)

So  $c = \frac{1}{2\pi}$ 

#### 2.2 (b)

Because  $f_X(-x) = f_X(x)$ , for  $k = 1, 3, \dots, 2n + 1, \dots, E[X^k] = 0$ .

For  $k=2n, n=1, 2, \cdots$ 

$$E[X^{k}] = \int_{\mathbb{R}} x^{k} f_{X}(x) dx = \frac{1}{2\pi} \int_{-2}^{2} x^{k} \sqrt{4 - x^{2}} dx$$

$$= \frac{1}{\pi} \int_{0}^{2} x^{k} \sqrt{4 - x^{2}} dx \quad (x = 2\sin\theta)$$

$$= \frac{2^{k+2}}{\pi} \int_{0}^{\frac{\pi}{2}} \sin^{k}\theta \cos^{2}\theta d\theta$$

$$= \frac{2^{k+2}}{\pi} \int_{0}^{\frac{\pi}{2}} \left[ \sin^{k}\theta - \sin^{k+2}\theta \right] d\theta$$
(3)

Denote  $I_n = \int_0^{\frac{\pi}{2}} \sin^n x dx$ ,

$$I_{n} = \int_{0}^{\frac{\pi}{2}} \sin^{n} x dx = \int_{0}^{\frac{\pi}{2}} \sin^{n-1} x d(-\cos x)$$

$$= -\cos x \sin^{n-1} x \Big|_{0}^{\frac{\pi}{2}} + (n-1) \int_{0}^{\frac{\pi}{2}} \sin^{n-2} x \cos^{2} x dx$$

$$= (n-1)(I_{n-2} - I_{n})$$
(4)

So  $I_n = \frac{n-1}{n}I_{n-2}$  and  $I_0 = \frac{\pi}{2}, I_2 = \frac{\pi}{4}$ ,

$$I_{2n} = \frac{(2n-1)(2n-3)\cdots 1}{(2n)(2n-2)\cdots 2} \frac{\pi}{2} = \frac{(2n)!}{(n!)^2} \frac{\pi}{2^{2n+1}}$$

So, For  $k = 2n, n = 1, 2, \cdots$ 

$$E[X^{k}] = \frac{2^{k+2}}{\pi} \int_{0}^{\frac{\pi}{2}} \left[ \sin^{k} \theta - \sin^{k+2} \theta \right] d\theta$$

$$= \frac{2^{k+2}}{\pi} (I_{k} - I_{k+2})$$

$$= \frac{2^{k+2}}{\pi} \left( \frac{(k)!}{((k/2)!)^{2}} \frac{\pi}{2^{k+1}} - \frac{(k+2)!}{((k/2+1)!)^{2}} \frac{\pi}{2^{k+3}} \right)$$

$$= \frac{2(k)!}{((k/2)!)^{2}} - \frac{(k+2)!}{2((k/2+1)!)^{2}}$$

$$= \frac{2}{k+2} \frac{(k)!}{((k/2)!)^{2}}$$
(5)

In summary,

$$E[X^k] = \begin{cases} 0, & k = 2n - 1, n = 1, 2, \dots \\ \frac{2}{k+2} \frac{(k)!}{((k/2)!)^2}, & k = 2n, n = 1, 2, \dots \end{cases}$$
 (6)

#### 3 Problem 3

Yes.

To prove this, we define a series of random variables by replacing  $X_i$  with  $Y_i$  once a time.

Define 
$$Z_0 = X = X_1 + X_2 + \dots + X_n$$
,  $Z_1 = Y_1 + X_2 + \dots + X_n$ ,  $Z_j = \sum_{i=1}^j Y_j + \sum_{i=J+1}^n X_i$ ,  $Z_n = Y = Y_1 + Y_2 + \dots + Y_n$ 

We want to prove  $Z_j$  stochastically dominates  $Z_{j-1}$ , that is,

$$\forall k = 1, 2, \cdots, n, P(Z_{i-1} \geqslant k) \geqslant P(Z_i \geqslant k)$$

Notice that the only difference between  $Z_{j-1}$  and  $Z_j$  is the j-th item,  $X_j$  or  $Y_j$ .

Denote the sum of the left n-1 items as Z', and  $Z_{j-1}=Z'+X_j, Z_j=Z'+Y_j$ , also denote  $P(X_i=1)=p_i\geqslant P(Y_i=1)=q_i$ .

$$P(Z_{j-1} \ge k) = P(Z' + X_j \ge k)$$

$$= P(Z' \ge k) + P(Z' = k - 1, X_j = 1)$$

$$= P(Z' \ge k) + P(Z' = k - 1)p_j$$
(7)

In the same way,  $P(Z_i \ge k) = P(Z' \ge k) + P(Z' = k - 1)q_i$ .

Since  $p_j \ge q_j$ , we prove that  $\forall k = 1, 2, \dots, n$ ,  $P(Z_{j-1} \ge k) \ge P(Z_i \ge k)$ 

Therefore,  $\forall k = 1, 2, \cdots, n$ 

$$P(X \geqslant k) = P(Z_0 \geqslant k) \geqslant P(Z_1 \geqslant k) \geqslant \dots \geqslant P(Z_n \geqslant k) = P(Y \geqslant k)$$
(8)

## 4 Problem 4

## 4.1 (a)

Suppose  $V \sim \text{Exp}(\lambda)$ , because U, V are independent,  $\mathbf{E}\left[\frac{V^2}{1+U}\right] = \mathbf{E}[V^2] \, \mathbf{E}\left[\frac{1}{1+U}\right]$ .

$$E[V^2] = \int_0^\infty \lambda x^2 e^{-\lambda x} dx = \int_0^\infty x^2 e^{-x} dx = -\frac{1}{\lambda^2} e^{-x} (2 + 2x + x^2) \Big|_0^\infty = \frac{2}{\lambda^2}$$
(9)

$$E\left[\frac{1}{1+U}\right] = \int_0^1 \frac{1}{1+x} dx = \ln 2 \tag{10}$$

So,

$$E\left[\frac{V^2}{1+U}\right] = E[V^2] E\left[\frac{1}{1+U}\right] = \frac{2\ln 2}{\lambda^2}$$
(11)

## 4.2 (b)

$$P(U \leq V) = \int_0^1 P(V \geq u) f_U(u) du$$

$$= \int_0^1 \int_u^\infty \lambda e^{-\lambda v} dv du$$

$$= \int_0^1 e^{-\lambda u} du$$

$$= \frac{1}{\lambda} (1 - e^{-\lambda})$$
(12)

## 4.3 (c)

Because  $U = \sqrt{Y}, V = \frac{Z}{\sqrt{Y}}$ 

$$f_{Y,Z}(y,z) = f_{U,V}(u(y,z),v(y,z)) \left| \frac{\partial(u,v)}{\partial(y,z)} \right|$$
(13)

And the Jacobian matrix is,

$$|J| = \left| \frac{\partial(u, v)}{\partial(y, z)} \right| = \left| -\frac{\frac{1}{2\sqrt{Y}}}{\frac{Z}{2V\sqrt{Y}}} \quad \frac{0}{\sqrt{Y}} \right| = \frac{1}{2Y}$$
 (14)

So the joint pdf of Y, Z is

$$f_{Y,Z}(y,z) = 1 \times \lambda e^{-\lambda \frac{z}{\sqrt{y}}} \times \frac{1}{2y}$$
 (15)

The support set of Y, Z is  $[0,1] \times [0,\infty)$ .

$$f_{Y,Z}(y,z) = \begin{cases} \frac{\lambda}{2y} e^{-\lambda \frac{z}{\sqrt{y}}}, & (y,z) \in [0,1] \times [0,\infty) \\ 0, & \text{else} \end{cases}$$
 (16)

## 5 Problem 5

First  $X_2 + X_3$  follows the Gamma distribution  $\Gamma(2, \frac{1}{\lambda})$  with  $\frac{1}{\lambda}$  as the scale parameter, to derive this, let  $Z = X_2 + X_3$ ,

$$f_{Z}(z) = \int_{-\infty}^{\infty} f_{X_{2}}(x) f_{X_{3}}(z - x) dx$$

$$= \int_{0}^{z} \lambda e^{-\lambda x} \lambda e^{-\lambda(z - x)} dx$$

$$= \int_{0}^{z} \lambda^{2} e^{-\lambda z} dx$$

$$= \lambda^{2} z e^{-\lambda z}$$

$$(17)$$

So

$$P(X_{1} > X_{2} + X_{3}) = P(X_{1} > Z) = \int_{0}^{\infty} f_{X_{1}}(x)P(Z < x)dx$$

$$= \int_{0}^{\infty} \left( \int_{0}^{x} \lambda^{2}ze^{-\lambda z}dz \right) \lambda e^{-\lambda x}dx$$

$$= \int_{0}^{\infty} \left( 1 - (1 + \lambda x)e^{-\lambda x} \right) \lambda e^{-\lambda x}dx$$

$$= 1 - \int_{0}^{\infty} \lambda e^{-2\lambda x}dx - \int_{0}^{\infty} \lambda^{2}xe^{-2\lambda x}dx$$

$$= 1 - \frac{1}{2} - \frac{1}{4} = \frac{1}{4}$$
(18)

## 6 Problem 6

The eigenvalues  $Y_1, Y_2$  are the roots of the following equation,

$$(\lambda - X_1)(\lambda - X_2) = X_3^2 \tag{19}$$

## Part 1

First, let  $U = \frac{X_1 + X_2}{2}$ ,  $V = \frac{X_1 - X_2}{2}$ , we prove that  $U, V \stackrel{i.i.d.}{\sim} N(0, 1)$ .

$$f_{U,V}(u,v) = f_{X_1,X_2}(x_1,x_2) \left| \frac{\partial(x_1,x_2)}{\partial(u,v)} \right|$$

$$= \frac{1}{4\pi} e^{-\frac{x_1^2 + x_2^2}{4}} \left\| \begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix} \right\|$$

$$= \frac{1}{2\pi} e^{-\frac{u^2 + v^2}{2}}$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} \times \frac{1}{\sqrt{2\pi}} e^{-\frac{v^2}{2}}$$
(20)

So U, V are independent and follow the standard Gaussian distribution.

#### Part 2

By solving the equation 19 we have

$$\begin{cases}
Y_1 + Y_2 = X_1 + X_2 = 2U \\
|Y_1 - Y_2| = 2\sqrt{\left(\frac{X_1 - X_2}{2}\right)^2 + X_3^2} = 2\sqrt{V^2 + X_3^2}
\end{cases}$$
(21)

Denote 
$$Z_1 = \frac{Y_1 + Y_2}{2}$$
,  $Z_2 = \frac{|Y_1 - Y_2|}{2}$ , so  $Z_1 = U \sim N(0, 1)$ ,  $Z_2 = \sqrt{V^2 + X_3^2}$ .

Because  $X_1, X_2, X_3$  are independent,  $U, V, X_3$  are also independent,  $Z_1$  and  $Z_2$  are independent. Because V and  $X_3$  are standard Gaussian random variables  $Z_2 = \sqrt{V^2 + X_3^2} \sim \chi_2$  (also known as the Rayleigh distribution)

To derive  $f_{Z_2}(z)$ ,

$$F_{Z_2}(z) = P(Z_2 \leqslant z) = P(X_3^2 + V^2 \leqslant z^2) = \iint_{x^2 + y^2 \leqslant z^2} \frac{1}{2\pi} e^{-\frac{x^2 + y^2}{2}} dx dy$$

$$= \frac{1}{2\pi} \int_0^{2\pi} \int_0^z r e^{-\frac{r^2}{2}} dr d\theta$$

$$= \int_0^z r e^{-\frac{r^2}{2}} dr$$

$$= 1 - e^{-\frac{z^2}{2}}$$
(22)

And 
$$f_{Z_2}(z) = F'_{Z_2}(z) = ze^{-\frac{z^2}{2}}, z \geqslant 0$$

#### Part 3

Assume  $Y_1 \geqslant Y_2$ ,

$$f_{Y_1,Y_2}(y_1, y_2) = f_{Z_1,Z_2}(z_1, z_2) \left| \frac{\partial(z_1, z_2)}{\partial(y_1, y_2)} \right|$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{z_1^2}{2}} z_2 e^{-\frac{z_2^2}{2}} \begin{vmatrix} 0.5 & 0.5 \\ 0.5 & -0.5 \end{vmatrix}$$

$$= \frac{1}{4\sqrt{2\pi}} (y_1 - y_2) e^{-\frac{y_1^2 + y_2^2}{4}}, y_1 \geqslant y_2$$

$$(23)$$

Symmetrically, when  $Y_1 \leqslant Y_2$ ,

$$f_{Y_1,Y_2}(y_1,y_2) = \frac{1}{4\sqrt{2\pi}}(y_2 - y_1)e^{-\frac{y_1^2 + y_2^2}{4}}, y_1 \leqslant y_2$$
 (24)

So, the joint pdf of  $Y_1$  and  $Y_2$  is,

$$f_{Y_1,Y_2}(y_1,y_2) \propto |y_1 - y_2| e^{-\frac{y_1^2 + y_2^2}{4}}$$
 (25)

By  $\int_{\mathbb{R}^2} f_{Y_1,Y_2}(y_1,y_2) dy_1 dy_2 = 1$ 

$$f_{Y_1,Y_2}(y_1,y_2) = \frac{1}{8\sqrt{2\pi}} |y_1 - y_2| e^{-\frac{y_1^2 + y_2^2}{4}}$$
(26)