## STP 421 - Problem Set 5 Solutions

1.) Let X and Y be jointly continuous random variables with joint density function

$$f(x,y) = c(y^2 - x^2)e^{-y}, \qquad -y \le x \le y, 0 < y < \infty.$$

- (a) Find c so that f is a density function.
- (b) Find the marginal densities of X and Y.
- (c) Find the expected value of X.

#### **Solutions:**

(a) The normalizing constant c is determined by the condition

$$1 = \int_{\mathbb{R}^2} f(x, y) dx dy$$

$$= c \cdot \int_0^\infty e^{-y} \left( \int_{-y}^y (y^2 - x^2) dx \right) dy$$

$$= c \cdot \int_0^\infty e^{-y} \left( 2y^3 - \frac{2}{3}y^3 \right) dy$$

$$= c \cdot \frac{4}{3} \int_0^\infty e^{-y} y^3 dy$$

$$= c \cdot \frac{4}{3} \Gamma(4).$$

Since  $\Gamma(4) = 3! = 6$ , it follows that  $c = \frac{1}{8}$ .

(b) The marginal density of X at  $x \in \mathbb{R}$  is

$$f_X(x) = \int_R f(x,y)dy$$

$$= \frac{1}{8} \int_{|x|}^{\infty} (y^2 - x^2) e^{-y} dy$$

$$= \frac{1}{8} \left[ \int_0^{\infty} (z + |x|)^2 e^{-z - |x|} dz - x^2 e^{-|x|} \right]$$

$$= \frac{1}{4} (1 + |x|) e^{-|x|},$$

while the marginal density of Y at  $y \in (0, \infty)$  is

$$f_Y(y) = \int_R f(x,y)dx$$

$$= \frac{1}{8} \int_{-y}^y (y^2 - x^2) e^{-y} dx$$

$$= \frac{1}{8} e^{-y} \left( 2y^3 - \frac{2}{3}y^3 \right)$$

$$= \frac{1}{6} y^3 e^{-y},$$

i.e., the marginal distribution of Y is that of a Gamma distributed random variable with parameters  $\alpha = 3$  and  $\lambda = 1$ .

(c) The expected value of X is

$$\mathbb{E}[X] = \int_{-\infty}^{\infty} x f_X(x) dx = 0,$$

since  $x \cdot f_X(x)$  is an odd function.

2.) Let X and Y be independent standard uniform random variables and let a, b and c be positive real numbers. Find the probability that  $aX + bY \le c$ .

# Solution:

Because (X,Y) is uniformly distributed on the unit square  $[0,1] \times [0,1]$ , the probability of an event  $\{(X,Y) \subset A\}$ , where A is a subset of the unit square, is equal to the area of A. In this problem, the set A is the intersection of the region beneath the line  $y = \frac{c}{b} - \frac{a}{b}x$  and the unit square. There are five cases:

$$\mathbb{P}(aX+bY\leq c) = \begin{cases} 1 & \text{if } c\geq a+b \\ 1-\frac{1}{2}\left(1-\frac{c-b}{a}\right)\left(1-\frac{c-a}{b}\right) & \text{if } a+b\geq c\geq \max\{a,b\} \\ \frac{c-b}{a}+\frac{1}{2}\frac{b}{a} & \text{if } a\geq c\geq b \\ \frac{c-a}{b}+\frac{1}{2}\frac{a}{b} & \text{if } b\geq c\geq a \\ \frac{c^2}{2ab} & \text{if } \min\{a,b\}\geq c \end{cases}$$

3.) Show that if X and Y are jointly continuous, then X + Y is a continuous random variable while X, Y and X + Y are not jointly continuous.

#### **Solution:**

Suppose that X and Y are jointly continuous with joint density  $f_{X,Y}(x,y)$ . Then the cumulative distribution function of X + Y is equal to

$$\mathbb{P}{X + Y \le t} = \int_{-\infty}^{\infty} dx \int_{-\infty}^{t-x} dy f_{X,Y}(x,y).$$

Since this function is differentiable with respect to t, it follows that X + Y is a continuous random variable with density

$$p_{X+Y}(t) = \int_{-\infty}^{\infty} dx f_{X,Y}(x, t - x).$$

To show that X, Y, and X + Y are not jointly continuous, let  $A = \{(x, y, z) : z - x - y = 0\}$  and note that  $\mathbb{P}((X, Y, X + Y) \in A) = 1$ . If these variables were jointly continuous, then there would exists a density function  $p : \mathbb{R}^3 \to [0, \infty]$  such that

$$\mathbb{P}(E) = \int_{E} p(x, y, z) dx dy dz$$

for every measurable subset  $E \subset \mathbb{R}^3$ . However, since A is a two-dimensional subspace of  $\mathbb{R}^3$ , we have

$$\int_{A} f(x, y, z) dx dy dz = 0$$

for every measurable function  $f: \mathbb{R} \to [0, \infty]$ . This shows that X, Y, and X + Y do not have a joint density and therefore are not jointly continuous.

4.) Suppose that  $X_1, \dots, X_n$  are independent exponential random variables with parameters  $\lambda_1, \dots, \lambda_n$ . Find the distribution of  $Y = \min\{X_1, \dots, X_n\}$ . Hint: Calculate the probability  $\mathbb{P}(Y > t)$ .

# Solution:

Since Y > t if and only if  $X_i > t$  for all  $i = 1, \dots, n$ , it follows that

$$\mathbb{P}(Y > t) = \mathbb{P}(X_1 > t, \dots, X_n > t)$$

$$= \prod_{i=1}^{n} \mathbb{P}(X_i > t)$$
 (by independence)
$$= \prod_{i=1}^{n} e^{-\lambda_i t}$$

$$= \exp\left\{-t \sum_{i=1}^{n} \lambda_i\right\},$$

which shows that Y is exponentially distributed with rate parameter  $\sum_{i=1}^{n} \lambda_i$ .