

Probability Theory: HW5

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Exercise 6.36

We have,

$$f(x, y, z) = \begin{cases} 8xyz & \text{if } 0 < x, y, z < 1, \\ 0 & \text{otherwise} \end{cases}$$

To find if X, Y, Z are independent, we can find their marginal. So we have,

$$\begin{aligned} f_X(x) &= \int_0^1 \int_0^1 8xyz \, dy \, dz \\ &= \int_0^1 [8xz \frac{y^2}{2}]_0^1 \, dz \\ &= \int_0^1 4xz \, dz \\ &= \int_0^1 4x \frac{z^2}{2}]_0^1 \, dz \\ &= \int_0^1 4x \frac{1}{2} \, dz \\ &= 2x \end{aligned}$$

Similarly we get $f_Y(y) = 2y$ and $f_Z(z) = 2z$. Now note that we have $f(x, y, z) = f_X(x)f_Y(y)f_Z(z)$ which means that the r.v are independent.

Now to find $\mathbb{P}(X > Y)$. Here Z can take on any value so we have,

$$\begin{aligned} P(X > Y) &= \int_0^1 \int_0^1 \int_y^1 8xyz \, dx \, dy \, dz \\ &= \int_0^1 \int_0^1 4yz[x^2]_y^1 \, dy \, dz \\ &= \int_0^1 \int_0^1 4yz[1 - y^2] \, dy \, dz \\ &= \int_0^1 \int_0^1 4yz - 4y^3z \, dy \, dz \\ &= \int_0^1 [2y^2z - y^4z]_0^1 \, dz \\ &= \int_0^1 [2z - z] \, dz \\ &= \int_0^1 [z] \, dz \\ &= \frac{z^2}{2}]_0^1 \\ &= \frac{1}{2} \end{aligned}$$

Now note that $\mathbb{P}(X > Y)$ and $\mathbb{P}(Y > Z)$ will be the same value as X, Y, Z all have the same distribution and pdf so it is also equal to $\frac{1}{2}$.

Exercise 6.45

We have, X, Y and the joint density,

$$f(x, y) = \begin{cases} \frac{1}{2}(x + y)e^{-x-y} & \text{if } x, y > 0 \\ 0 & \text{otherwise} \end{cases}$$

Now to find the density function of $X + Y$. We can use the convolution formula to achieve this. So the pdf of $X + Y$ evaluated at z would be when $y = z - x$ and we integrate x from 0 to z as both $x, y > 0$ so $z - x > 0, x < z$. This gives us,

$$\begin{aligned} f_{X+Y}(z) &= \int_0^z \frac{1}{2}(x+z-x)e^{-x-z+x} dx \\ &= \int_0^z \frac{1}{2}(z)e^{-z} dx \\ &= \frac{1}{2}(z)e^{-z}[x]_0^z \\ &= \frac{1}{2}z^2e^{-z} \quad \text{for } z > 0 \end{aligned}$$

Exercise 6.55

We have,

$$f(x, y) = \begin{cases} \frac{1}{4}e^{-\frac{1}{2}(x+y)} & \text{if } x, y > 0 \\ 0 & \text{otherwise} \end{cases}$$

First we need to joint density of $U = \frac{1}{2}(X - Y)$ and $V = Y$. We rewrite this as functions of U and V to get, $Y = V$ and $X = 2U + V$. Now the jacobian of this is $2 \cdot 1 = 2$. So the joint of U, V is,

$$\begin{aligned} f_{U,V}(u, v) &= 2f(2u + v, v) \\ &= 2 \frac{1}{4}e^{-\frac{1}{2}(2u+v+v)} \\ &= \frac{1}{2}e^{-u-v} \quad \text{for } u, v \in A \end{aligned}$$

Now to find the density of U . We have,

$$f_U(u) = \int_{-\infty}^{\infty} \frac{1}{2}e^{-u-v} dv$$

Now if $u \geq 0$ the lower limit is 0 and if $u < 0$, note that $X \geq 0$ which means $2u + v \geq 0$ or that $v \geq -2u$ so lower limit is $-2u$. For both we have,

$$\begin{aligned} f_U(u) &= \int_0^{\infty} \frac{1}{2}e^{-u-v} dv \\ &= \int_0^{\infty} \frac{1}{2}e^{-u}e^{-v} dv \\ &= \frac{1}{2}e^{-u}[-e^{-v}]_0^{\infty} \\ &= \frac{1}{2}e^{-u} \end{aligned}$$

Now for $u < 0$ we have,

$$\begin{aligned} f_U(u) &= \int_{-2u}^{\infty} \frac{1}{2} e^{-u-v} dv \\ &= \int_{-2u}^{\infty} \frac{1}{2} e^{-u} e^{-v} dv \\ &= \frac{1}{2} e^{-u} [-e^{-v}]_{-2u}^{\infty} \\ &= \frac{1}{2} e^{-u} e^{2u} \\ &= \frac{1}{2} e^u \\ &= \frac{1}{2} e^{-|u|} \end{aligned}$$

So we have,

$$f_U(u) = \frac{1}{2} e^{-|u|} \quad \text{for } u \in \mathbb{R}$$

Exercise 6.61

We have X, Y are independent r.v that are exponential with parameter λ . Now let $U = X$ and $V = X + Y$. Rewriting this we have $X = V - U$ and $Y = U$ the Jacobian of which is 1. So we have,

$$\begin{aligned} f_{U,V} &= f(v-u, u) \\ &= f(v-u)f(u) \\ &= \lambda^2 e^{-\lambda(v-u+u)} \\ &= \lambda^2 e^{-\lambda v} \end{aligned}$$

Now to find the conditional given $X + Y = a$ or $V = a$ we integrate over U first to get the marginal of V . Note that $U = Y$ so $u \geq 0$. Now as $X = a - Y$ the maximum value Y can take is a as X is also non-negative. So our lower bound is 0 and upper is a so we have,

$$\begin{aligned} f_V(v) &= \int_0^v \lambda^2 e^{-\lambda v} du \\ &= \lambda^2 e^{-\lambda v} [u]_0^v \\ &= \lambda^2 e^{-\lambda v} v \end{aligned}$$

So this gives us $f_{U|V}(u | v) = f_{X|X+Y} = \frac{f(u,v)}{f_V(v)} = \frac{1}{v}$. So here we have $v = a$ which gives us $f_{U|V}(u | a) = \frac{1}{a}$ which is uniform in the interval $(0, a)$. Hence knowing $X + Y$ gives us no useful information about the distribution of X in the interval $(0, a)$.

Exercise 6.70

We have X, Y are uniformly distributed on the uni disk so,

$$f_{X,Y}(x,y) = \begin{cases} \pi^{-1} & \text{if } x^2 + y^2 \leq 1, \\ 0 & \text{otherwise} \end{cases}$$

To find $E\sqrt{X^2 + Y^2}$ we have,

$$E\sqrt{X^2 + Y^2} = \int \int \sqrt{x^2 + y^2} \pi^{-1} dx dy$$

Now changing to polar coordinates we have,

$$\begin{aligned}
E\sqrt{X^2 + Y^2} &= \int_0^{2\pi} \int_0^1 \sqrt{r^2} r \pi^{-1} dr d\theta \\
&= \int_0^{2\pi} \int_0^1 r^2 \pi^{-1} dr d\theta \\
&= \int_0^{2\pi} [r^3]_0^1 \frac{1}{3} \pi^{-1} d\theta \\
&= \int_0^{2\pi} \frac{1}{3} \pi^{-1} d\theta \\
&= \frac{1}{3} \pi^{-1} 2\pi \\
&= \frac{2}{3}
\end{aligned}$$

For $E[X^2 + Y^2]$ we have

$$\begin{aligned}
E[X^2 + Y^2] &= \int_0^{2\pi} \int_0^1 r^2 r \pi^{-1} dr d\theta \\
&= \int_0^{2\pi} \int_0^1 r^3 \pi^{-1} dr d\theta \\
&= \int_0^{2\pi} [r^4]_0^1 \frac{1}{4} \pi^{-1} d\theta \\
&= \int_0^{2\pi} \frac{1}{4} \pi^{-1} d\theta \\
&= \frac{1}{4} \pi^{-1} 2\pi \\
&= \frac{2}{4} = \frac{1}{2}
\end{aligned}$$

Exercise 7.10

We have X is uniformly distributed on (a, b) and we need to find $E(X^k)$. If x is uniformly distributed then the density function is,

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{if } x \in [a, b] \\ 0 & \text{otherwise} \end{cases}$$

So we have,

$$\begin{aligned}
E(X^k) &= \int_a^b \frac{x^k}{b-a} dx \\
&= \frac{1}{b-a} \left[\frac{x^{k+1}}{k+1} \right]_a^b \\
&= \frac{1}{b-a} \left(\frac{b^{k+1} - a^{k+1}}{k+1} \right) \\
&= \frac{b^{k+1} - a^{k+1}}{(b-a)(k+1)}
\end{aligned}$$

Exercise 7.36

We have X_1, X_2, \dots is a sequence of uncorrelated random variables with variance σ^2 and $S_n = X_1 + \dots + X_n$. We have assuming $m < n$ that,

$$\begin{aligned}
Var(S_n + S_m) &= Var(S_m) + Var(S_n) + 2Cov(s_m, s_n) \\
2Cov(S_m, S_n) &= Var(S_n + S_m) - (Var(S_m) + Var(S_n)) \\
&= Var(X_1 + \dots + X_n + X_1 + \dots + X_m) \\
&\quad - (Var(X_1 + \dots + X_m) + Var(X_1 + \dots + X_n)) \\
&= Var(2X_1 + \dots + 2X_m + X_{m+1} + \dots + X_n) \\
&\quad - (2Var(X_1) + \dots + 2Var(X_m) + Var(X_{m+1}) + \dots + Var(X_n)) \\
&= 4(Var(X_1) + \dots + Var(X_m)) + Var(X_{m+1}) + \dots + Var(X_m) \\
&\quad - 2(Var(X_1) + \dots + Var(X_m)) - (Var(X_{m+1}) + \dots + Var(X_n)) \\
&= 2(Var(X_1) + \dots + Var(X_m)) \\
&= 2(Var(X_1 + \dots + X_m)) \\
&= 2Var(S_m)
\end{aligned}$$

So we get,

$$\begin{aligned}
2Cov(S_m, S_n) &= 2Var(S_m) \\
Cov(S_m, S_n) &= Var(S_m)
\end{aligned}$$

And we get,

$$\begin{aligned}
Var(S_m) &= Var(X_1) + \dots + Var(X_m) \\
&= m\sigma^2 \quad \text{if } m < n
\end{aligned}$$

Problem 6

We have X_1, X_2, \dots, X_n are independent r.v each with distribution function F and density f . $U = \min\{X_1, \dots, X_n\}$ and $V = \max\{X_1, \dots, X_n\}$. The joint cdf of U, V can be seen as,

$$F_{U,V}(u, v) = \mathbb{P}(U \leq u, V \leq v)$$

Now note that for $V \leq v$ we need each $X_n \leq v$ as V is the max so there cannot be a X_n with value greater than v . Now if we have each $X_n \leq v$. For U note that if we need $U > u$ then we need all $X_n > u$ so for $U \leq u$ we have for all X_n that $u \leq X_n \leq v$. So now our solution would be the probability that all $V \leq v$ minus that when $U \geq u$ and $V \leq v$ this is,

$$F_{U,V}(u, v) = (F(v))^n - (F(v) - F(u))^n$$

So,

$$\begin{aligned}
\frac{\partial}{\partial u} \frac{\partial}{\partial v} F_{U,V}(u, v) &= \frac{\partial}{\partial u} \frac{\partial}{\partial v} ((F(v))^n - (F(v) - F(u))^n) \\
&= \frac{\partial}{\partial u} [n(F(v))^{n-1} f(v) - n(F(v) - F(u))^{n-1} f(v)] \\
&= n f(v) \frac{\partial}{\partial u} [(F(v))^{n-1} - (F(v) - F(u))^{n-1}] \\
&= n f(v) [-(n-1)(F(v) - F(u))^{n-2} (-f(u))] \\
&= n(n-1) f(u) f(v) (F(v) - F(u))^{n-2}
\end{aligned}$$

Problem 20

We have X and Y are r.v with vector (X, Y) uniformly distributed in $R = \{(x, y) : 0 < y < x < 1\}$. We need to find joint probability of (X, Y) first.

Note that the area of the regions in $\int_0^1 \int_y^1 dx dy = \int_0^1 (1-y) dy = \frac{1}{2}$. So the joint probability of X, Y is $\frac{1}{2} = 2$ for $0 < y < x < 1$. Now to find $\mathbb{P}(X + Y < 1)$. Note that as $x + y < 1$ we have $x < 1 - y$ but we also have $x > y$ so we get $y < x < 1 - y$. Now for bounds on y note that y goes from 0 to 1. But y can't be greater than $\frac{1}{2}$ as that would be we have $x > 1/2$ and $x + y > 1/2$ So we have,

$$\begin{aligned}\mathbb{P}(X + Y < 1) &= \int_0^{\frac{1}{2}} \int_y^{1-y} 2 dx dy \\ &= \int_0^{\frac{1}{2}} 2[1 - y - y] dy \\ &= \int_0^{\frac{1}{2}} 2[1 - 2y] dy \\ &= \int_0^{\frac{1}{2}} 2 - 4y dy \\ &= [2y - 2y^2]_0^{\frac{1}{2}} \\ &= 1 - \frac{1}{2} = \frac{1}{2}\end{aligned}$$

Now to find $f_X(x)$ and EX . We have the joint as 2 and to get f_X we need to integrate over y . As y goes from 0 to x we have,

$$\begin{aligned}f_X(x) &= \int_0^x 2 dy \\ &= 2[y]_0^x \\ &= 2x\end{aligned}$$

So the expectation of X is $\int_0^1 x \cdot 2x dx = \frac{2x^3}{3}]_0^1 = \frac{2}{3}$. Now we have $f_{Y|X}(y | x) = \frac{f(x,y)}{f_X(x)} = \frac{2}{2x} = \frac{1}{x}$. So $E(Y | X = x) = \int_0^x y \frac{1}{x} dy = \frac{y^2}{2x}]_0^x = \frac{x^2}{2x} = \frac{x}{2}$ for $0 < x < 1$