Homework #4

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Problem 1

Suppose that X is a positive random variable so that P(X > 0) = 1. Prove that for any t < 0, $M_X(t) = E[e^{tX}] < \infty$.

Hint: Find a relationship between e^{tX} and 1, for t < 0. Use the fact that if for all x, $g(x) \le h(x)$ then $\int g(x)dx \le \int h(x)dx$.

$$M_X(t) = E[e^{tX}] = \int_{-\infty}^{\infty} e^{tx} f_X(x) dx$$
$$= \int_{0}^{\infty} e^{tx} f_X(x) dx$$

For t < 0:

$$e^{tx} f_X(x) \le f_X(x) \forall f_X(x)$$

because $e^{kx} \le 1 \forall k < 0$ so it follows:

$$\int_0^\infty e^{tx} f_X(x) dx \le \int_0^\infty f_X(x) dx$$
$$\int_0^\infty e^{tx} f_X(x) dx \le P(X > 0)$$
$$\int_0^\infty e^{tx} f_X(x) dx \le 1$$

Problem 2.

Let X be an exponential random variable with parameter $\lambda > 0$.

a. Use Chebyshev's inequality to find an upper bound on $Pr(|X - E[X]| \ge k)$

$$P(|X - c|) \ge d) \le \frac{E[(X - c)^2]}{d^2}$$

$$P(|X - \mu|) \ge k) \le \frac{E[(X - \mu)^2]}{k^2}$$

$$P(|X - \mu|) \ge k) \le \frac{V[X]}{k^2}$$

$$P(|X - \mu|) \ge k) \le \left(\frac{1}{k\lambda}\right)^2$$

b. Calculate E[|X - E[X]|] and then use Markov's inequality to find a different upper bound on $Pr(|X - E[X]| \ge k)$.

Hint: first remove the absolute value by splitting the integral into two pieces. Each piece can then be integrated using the Exponential CDF (or just integrating directly) and integration by parts.

$$\begin{split} E[|X-E[X]|] &= E[|X-\mu|] = \int_{-\infty}^{\infty} |x-\mu| f_X(x) dx \\ &= \int_{0}^{\infty} |x-\mu| \lambda e^{-\lambda x} dx \\ &= \int_{0}^{1/\lambda} \left(\frac{1}{\lambda} - x\right) \lambda e^{-\lambda x} dx + \int_{1/\lambda}^{\infty} \left(x - \frac{1}{\lambda}\right) \lambda e^{-\lambda x} dx \\ &= \int_{0}^{1/\lambda} e^{-\lambda x} dx - \int_{0}^{1/\lambda} \lambda x e^{-\lambda x} dx + \int_{1/\lambda}^{\infty} \lambda x e^{-\lambda x} dx - \int_{1/\lambda}^{\infty} e^{-\lambda x} dx \\ &= -\frac{e^{-\lambda x}}{\lambda} \Big|_{0}^{1/\lambda} + \lambda \frac{\lambda x + 1}{\lambda^2} e^{-\lambda x} \Big|_{0}^{1/\lambda} - \lambda \frac{\lambda x + 1}{\lambda^2} e^{-\lambda x} \Big|_{1/\lambda}^{\infty} + \frac{e^{-\lambda x}}{\lambda} \Big|_{1/\lambda}^{\infty} \\ &= -\frac{e^{-\lambda x}}{\lambda} \Big|_{0}^{1/\lambda} + \lambda \frac{\lambda x + 1}{\lambda^2} e^{-\lambda x} \Big|_{0}^{1/\lambda} - \lambda \left(\frac{\lambda x}{\lambda^2} e^{-\lambda x} + \frac{e^{-\lambda x}}{\lambda^2}\right) \Big|_{1/\lambda}^{\infty} + \frac{e^{-\lambda x}}{\lambda} \Big|_{1/\lambda}^{\infty} \\ &= \left(-\frac{1}{e\lambda} + \frac{1}{\lambda}\right) + \lambda \left(\frac{2}{e\lambda^2} - \frac{1}{\lambda^2}\right) - \lambda \left(0 + 0 - \frac{2}{e\lambda^2} - \frac{1}{\lambda^2}\right) + \left(0 - \frac{1}{e\lambda}\right) \\ &= -\frac{1}{e\lambda} + \frac{1}{\lambda} + \frac{2}{e\lambda} - \frac{1}{\lambda} + \frac{2}{e\lambda} + \frac{1}{\lambda} - \frac{1}{e\lambda} \\ &= \frac{2}{e\lambda} + \frac{1}{\lambda} \\ &= \frac{4 - 2\lambda}{e\lambda^2} \end{split}$$

Markov's inequality is then:

$$P(|X - \mu| \ge k) = \frac{e - 2}{e \lambda k}$$

(c) For which values of k is the bound in (b) lower than that given in (a)?

$$1 > \frac{P_b}{P_a}$$

$$1 > \frac{(e-2)\lambda^2 k^2}{e\lambda k}$$

$$1 > \frac{e-2}{e}\lambda k$$

$$k < \frac{e}{\lambda(e-2)} \approx \frac{3.784}{\lambda}$$

If the joint probability density of the two random variables X and Y is given by:

$$f_{XY}(x,y) = \begin{cases} 2 & \text{for } 0 < x < y < 1 \\ 0 & \text{otherwise.} \end{cases}$$

a. Find P(Y < 0.5). Hint: Draw a graph.

$$P(Y < 0.5) = \int_0^1 \int_0^y f(x, y) dx dy$$
$$= \int_0^1 \int_0^y 2 dx dy = 2 \frac{y^2}{2} \Big|_0^1$$
$$= 1$$

b. Find P(X < 0.5).

$$P(X < 0.5) = \int_0^{0.5} \int_x^1 f(x, y) dy dx$$
$$= \int_0^1 (2 - x) dx = 2x \Big|_0^{0.5} - \frac{x^2}{2} \Big|_0^{0.5}$$
$$= 1 - \frac{1}{8} = \frac{7}{8}$$

c. Find P(X + Y < 1)

$$\begin{split} P(X+Y<1) &= \int_0^{0.5} \int_x^{1-x} f(x,y) dy dx \\ &= 2 \int_0^{0.5} (1-2x) dx = 2 \left(x |_0^{0.5} - 2 \frac{x^2}{2} |_0^{0.5} \right) \\ &= 2 \frac{1}{4} = \frac{1}{2} \end{split}$$

d. Find the joint CDF F(x,y). (You will need to consider several cases.)

$$F(x,y) = \begin{cases} 0 & \text{if } x < 0 \text{ or } y < 0 \\ 0 & \text{if } y > 1 \text{ or } x > y \\ 2xy & \text{if } y < 1 \text{ and } x < y \end{cases}$$

Suppose that X and Y are continuous random variables with the following joint cdf:

$$F_{XY}(x,y) = \widetilde{F}(x)\widetilde{F}(y) \tag{1}$$

where $\widetilde{F}(\cdot)$ is a (univariate) cdf.

(Aside) It follows from (1) that X and Y are independent random variables with the same marginal distribution.

(a) For a fixed value t, express $P(X \le t, Y \le t)$ in terms of $\widetilde{F}(\cdot)$.

Hint: Recall the definition of a joint cdf $F_{XY}(x,y)$.

By definition $F(x,y) = P(X \le x, Y \le y)$ so it follows that:

$$P(X \le t, y \le t) = F_{XY}(t, t) = \tilde{F}(t)\tilde{F}(t) = \tilde{F}^2(t)$$

(b) Let $T = \max(X, Y)$. Using your answer to (a) find the cdf $F_T(t)$ and the pdf $f_T(t)$ for T:

Hint: If $X \le t$ and $Y \le t$, what can one say about $\max(X, Y)$?

Because $T = \max(X, Y)$, for $T \le t$ to hold both $X \le t$ and $Y \le t$ have to be satisfied:

By definition:

$$f_T(t) = \frac{\partial}{\partial t} F_T(t) = \frac{\partial}{\partial t} \tilde{F}^2(t)$$
$$= 2\tilde{F}(t)\tilde{f}(t)$$

If the joint CDF for two random variables, X and Y, is given by:

$$F_{XY}(x,y) = \begin{cases} (1 - e^{-x^2})(1 - e^{-y^2}) & \text{for } x > 0, y > 0 \\ 0 & \text{otherwise} \end{cases}$$

Find the joint pdf $f_{XY}(x,y)$.

Let X and Y have the joint density function (here k > 0 is an unknown constant):

$$f_{XY}(x,y) = \begin{cases} k(x-y), & 0 \le y \le x \le 1; \\ 0 & \text{otherwise,} \end{cases}$$

- (a) Sketch the support of the joint density. You will need to use this in answering the next three question parts.
- (b) Find the constant of proportionality k.
- (c) Find the marginal densities $f_X(x)$ and $f_Y(y)$.
- (d) Find the conditional densities of $f_{Y|X}(y \mid x)$ and $f_{X|Y}(x \mid y)$. Remember to carefully state for which values of the conditioning variable these are defined.

If the joint probability density of the two random variables X and Y is given by:

$$f_{XY}(x,y) = \begin{cases} 4xy & \text{for } 0 < y < 1, \ 0 < x < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Let
$$D = (X - Y)$$
 and $S = (X + Y)$.

- (a) Solve for X and Y in terms of D and S.
- (b) Find the support of $f_{DS}(d, s)$ and draw a sketch.
- (c) Find the joint pdf $f_{DS}(d, s)$.
- (d) Find the marginal densities $f_D(d)$ and $f_S(s)$.
- (e) Find the conditional densities of $f_{D|S}(d \mid s)$ and $f_{S|D}(s \mid d)$. Remember to carefully state for which values of the conditioning variable these are defined.

The random variables (X, Y) have the following joint pdf:

$$f_{XY}(x,y) = \begin{cases} 2y & \text{for } 0 < x < 1 \text{ and } 0 < y < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Consider the transformation: U = XY and V = X.

- (a) Draw the support of the joint density $f_{XY}(x,y)$ for (X,Y).
- (b) Solve for X and Y in terms of U and V.
- (c) Carefully find the support for the joint density for (U, V). Then find the joint density $f_{UV}(u, v)$, remember to include the support.
- (d) Find the marginal densities $f_U(u)$ for U and $f_V(v)$ for V.