

Math 464 - Midterm II Name:

Show your work and explain all answers!

1. (7 points per part) Let X and Y be independent random variables each having the standard normal distribution, $N(0, 1)$.

a) Find the joint density of $U = aX + bY$ and $V = bX - aY$, where $a^2 + b^2 > 0$.

$$\begin{aligned} \begin{pmatrix} u \\ v \end{pmatrix} &= \begin{pmatrix} a & b \\ b & -a \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \\ \begin{pmatrix} x \\ y \end{pmatrix} &= \frac{1}{a^2 + b^2} \begin{pmatrix} a & b \\ b & -a \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} \\ |J| &= \left| \det \left[\frac{1}{a^2 + b^2} \begin{pmatrix} a & b \\ b & -a \end{pmatrix} \right] \right| = \frac{1}{a^2 + b^2} \\ f_{XY}(x, y) dx dy &= \frac{1}{2\pi} e^{-\frac{1}{2}(x^2 + y^2)} dx dy \\ &= \frac{1}{2\pi(a^2 + b^2)} e^{-\frac{1}{2(a^2 + b^2)}(u^2 + v^2)} du dv = f_{UV}(u, v) du dv \end{aligned}$$

b) What are the densities of U and V respectively? Are U and V independent?

$$\begin{aligned} f_U(u) &= \int_{-\infty}^{\infty} f_{UV}(u, v) dv \\ &= \frac{1}{\sqrt{2\pi(a^2 + b^2)}} e^{-\frac{1}{2(a^2 + b^2)}u^2} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi(a^2 + b^2)}} e^{-\frac{1}{2(a^2 + b^2)}v^2} dv \\ &= \frac{1}{\sqrt{2\pi(a^2 + b^2)}} e^{-\frac{1}{2(a^2 + b^2)}u^2} \text{ since the integrand is the density for } N(0, a^2 + b^2) \\ f_V(v) &= \frac{1}{\sqrt{2\pi(a^2 + b^2)}} e^{-\frac{1}{2(a^2 + b^2)}v^2} \text{ by a symmetric argument.} \end{aligned}$$

It is then immediate that $f_{UV}(u, v) = f_U(u)f_V(v)$ and therefore U and V are independent.

2. (9 points) Let X be uniform on $[0, 1]$. Let $Y = -\ln X$; find the density function f_Y .

$$\begin{aligned} P(Y \leq y) &= P(-\ln X \leq y) = P(\ln X \geq -y) \text{ for } 0 \leq y < \infty \text{ (0 otherwise)} \\ &= 1 - P(\ln X < -y) = 1 - P(\ln X \leq -y) \text{ since } P(\ln X = -y) = 0; \\ &= 1 - P(X \leq e^{-y}) = 1 - e^{-y} \text{ for } y \geq 0 \text{ (0 otherwise) since } X \text{ is uniform.} \\ f_Y(y) &= \frac{d}{dy} P(Y \leq y) = e^{-y} \text{ for } y > 0 \text{ (0 otherwise).} \end{aligned}$$

Hence Y has an exponential distribution with parameter 1.

3. (8 points per part) Two random variables X and Y have a joint density of the form

$$f(x, y) = \begin{cases} \frac{c}{(x)^3} & \text{if } 1 < x < \infty \text{ and } 1 < y < x \\ 0 & \text{otherwise} \end{cases}$$

a) Find the value of c that makes $f(x; y)$ into a density. If you are unable to find this constant, just keep calling it c and go on.

$$\begin{aligned} 1 &= c \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = c \int_1^{\infty} dx \int_1^x \frac{1}{x^3} dy = c \int_1^{\infty} \left(\frac{1}{x^2} - \frac{1}{x^3} \right) dx \\ &= c \left(\frac{-1}{x} + \frac{1}{2x^2} \right) \Big|_1^{\infty} = c(1 - 1/2) = c/2 \end{aligned}$$

Hence $c = 2$.

b) Find the density of X .

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_1^x \frac{2}{x^3} dy = \frac{2}{x^2} - \frac{2}{x^3} \text{ for } x > 1 (0 \text{ otherwise})$$

c) Verify that your formula in b) actually satisfies the requirements of a density.

For $x > 1$, $\frac{1}{x^2} > \frac{1}{x^3}$; hence, $f_X(x) \geq 0$. Also, by the calculation on the second line of (a), the total integral of $f_X(x)$ equals 1.

d) Find the conditional expectation $\mathbb{E}(Y|X)$.

$$\begin{aligned} \mathbb{E}(Y|X) &= \int_{-\infty}^{\infty} y f_{Y|X}(y|x) dy = \int_{-\infty}^{\infty} y \frac{f(x, y)}{f_X(x)} dy \\ &= \int_{-\infty}^{\infty} y \frac{2/x^3 \mathbb{I}_{1 < y < x < \infty}}{2(1/x^2 - 1/x^3) \mathbb{I}_{1 < x < \infty}} dy \\ &= \int_1^x \frac{y}{x-1} dy \quad (\text{N.B., } f_{Y|X} \text{ is the uniform density on } (1, x)) \\ &= \frac{1}{x-1} \frac{y^2}{2} \Big|_1^x = \frac{1}{x-1} \frac{x^2 - 1}{2} = \frac{x+1}{2}. \end{aligned}$$

4. (7 points per part) Let X_1, X_2, \dots, X_n be independent random variables each taking values in $\{0, 1, 2, \dots\}$.

- a) Consider the random variable $T = X_1 + \dots + X_n$. Use generating functions to calculate $\mathbb{E}(T)$.

By independence,

$$\begin{aligned} G_T(s) &= G_{X_1}(s) \cdots G_{X_n}(s) \\ \mathbb{E}(T) &= G'_T(1) \\ &= \sum_{i=1}^n G_{X_1}(1) \cdots G_{X_{i-1}}(1) G'_{X_i}(1) G_{X_{i+1}}(1) \cdots G_{X_n}(1) \\ &= \sum_{i=1}^n G'_{X_i}(1) = \sum_{i=1}^n \mathbb{E}(X_i) \end{aligned}$$

- b) Now suppose that the X_i all have the same mass function. Use generating functions to calculate $\mathbb{E}(T^2)$.

By independence and equality of mass functions,

$$\begin{aligned} G_T(s) &= G_X(s)^n \\ G'_T(s) &= nG_X(s)^{n-1}G'_X(s) \\ G''_T(s) &= n(n-1)G_X(s)^{n-2}(G'_X(s))^2 + nG_X(s)^{n-1}G''_X(s) \\ \mathbb{E}(T^2) = G''_T(1) + G'_T(1) &= n(n-1)G'_X(1)^2 + nG''_X(1) + nG'_X(1) \\ &= n(n-1)\mathbb{E}(X)^2 + n\mathbb{E}(X^2) \end{aligned}$$

- c) Still assuming that the X_i all have the same mass function, calculate the variance of T .

$$\begin{aligned} \text{Var}(T) &= \mathbb{E}(T^2) - \mathbb{E}(T)^2 \\ &= n(n-1)\mathbb{E}(X)^2 + n\mathbb{E}(X^2) - (n\mathbb{E}(X))^2 \\ &= -n\mathbb{E}(X)^2 + n\mathbb{E}(X^2) \\ &= n\text{Var}(X) \end{aligned}$$

5. (8 points per part) Let X and Y be independent random variables each having a geometric distribution with parameter p . Set $Z = Y - X$ and $M = \min(X, Y)$.

a) Show that for integer z and integer $m \geq 0$

$$P(M = m, Z = z) = \begin{cases} P(X = m - z)P(Y = m), & z < 0 \\ P(X = m)P(Y = m + z), & z \geq 0 \end{cases}$$

$$Z < 0 \implies X > Y \implies \min(X, Y) = Y \text{ and}$$

$$Z = z \iff X = Y - z$$

$$\begin{aligned} \implies P(M = m, Z = z) &= P(Y = m, X = Y - z) = P(Y = m, X = m - z) \\ &= P(X = m - z)P(Y = m) \text{ by independence of } X \text{ and } Y; \end{aligned}$$

$$Z \geq 0 \implies X \leq Y \implies \min(X, Y) = X \text{ and}$$

$$Z = z \iff Y = X + z$$

$$\begin{aligned} \implies P(M = m, Z = z) &= P(X = m, Y = X + z) = P(X = m, Y = m + z) \\ &= P(X = m)P(Y = m + z) \text{ by independence of } X \text{ and } Y. \end{aligned}$$

b) Show that part a) implies that for integer z and integer $m \geq 0$,

$$P(M = m, Z = z) = p^2(1-p)^{2m}(1-p)^{|z|}.$$

By (a) the proof of this identity reduces to two cases

$$\begin{aligned} z < 0 : P(M = m, Z = z) &= P(X = m - z)P(Y = m) \\ &= p(1-p)^{m-z}p(1-p)^m = p^2(1-p)^{2m+|z|} \\ z \geq 0 : P(M = m, Z = z) &= P(X = m)P(Y = m + z) \\ &= p(1-p)^mp(1-p)^{m+z} = p^2(1-p)^{2m+|z|} \end{aligned}$$

Note that the probability weights used here for the geometric distribution, $P(X = k) = pq^k$, differ from those we have been using based on the text; however, if I had taken $m \geq 1$ with our usual convention $P(X = k) = pq^{k-1}$ the result would have come out to be the same.

c) Are M and Z independent? Provide a proof or counter-example for your answer.

Since one can write $P(M = m, Z = z) = f(m)g(z)$ where $f(m) = p(1-p)^{2m}$ and $g(z) = p(1-p)^{|z|}$, by a theorem from the text, M and Z are independent. One can in fact directly calculate that $P(M = m) = (1+q)pq^{2m}$ and $P(Z = z) = p(1-p)^{|z|}/(1+q)$.

Extra Credit Problem (10 points) If X and Y are continuous random variables having joint density function $f(x, y)$, find the joint density function of the random variables $W = X^2$ and $Z = Y^2$.

On the domain $x, y > 0$, one has $x = \sqrt{w}, y = \sqrt{z}$ for $z, w > 0$. Hence,

$$|J| = \left| \det \begin{pmatrix} 1/2\sqrt{w} & 0 \\ 0 & 1/2\sqrt{z} \end{pmatrix} \right| = \frac{1}{4\sqrt{wz}}$$

By the change of variables formula,

$$f_{WZ}(w, z) = \frac{1}{4\sqrt{wz}} f_{XY}(\sqrt{w}, \sqrt{z})$$

Alternatively one may observe that for $w, z > 0$

$$\begin{aligned} P(W \leq w, Z \leq z) &= P(-\sqrt{w} \leq X \leq \sqrt{w}, -\sqrt{z} \leq Y \leq \sqrt{z}) \\ &= P(X \leq \sqrt{w}, Y \leq \sqrt{z}) - P(X \leq \sqrt{w}, Y \leq -\sqrt{z}) \\ &\quad - P(X \leq -\sqrt{w}, Y \leq \sqrt{z}) + P(X \leq -\sqrt{w}, Y \leq -\sqrt{z}) \\ f_{WZ}(w, z) &= \frac{\partial^2}{\partial w \partial z} P(W \leq w, Z \leq z) \\ &= \frac{1}{4\sqrt{wz}} (f_{XY}(\sqrt{w}, \sqrt{z}) + f_{XY}(\sqrt{w}, -\sqrt{z}) + f_{XY}(-\sqrt{w}, \sqrt{z}) + f_{XY}(-\sqrt{w}, -\sqrt{z})) \end{aligned}$$