

Exam in Probability Theory, 6 credits

Exam time:	8-12
Allowed:	Pocket calculator. Table with common formulas and moment generating functions (distributed with the exam). Table of integrals (distributed with the exam). Table with distributions from Appendix B in the course book (distributed with the exam).
Examinator:	Mattias Villani.
Assisting teacher:	Per Sidén, phone 0704-977175
Grades:	Grades: Maximum is 20 points. A=19-20 points B=17-18 points C=12-16 points D=10-11 points E=8-9 points F=0-7 points

- Write clear and concise answers to the questions.
 - Make sure to specify the definition region for all density functions.
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1. The random variable X has the distribution function

$$F_X(x) = \begin{cases} a \left(1 - \frac{1}{x}\right), & 1 < x < \infty \\ 0, & \text{otherwise} \end{cases}$$

and the conditional probability density of Y given X is

$$f_{Y|X=x}(y) = \begin{cases} by, & 0 < y < x \\ 0, & \text{otherwise} \end{cases}.$$

(a) Determine the constant a and the probability density function of X . 1p.

Solution:

$$1 = \lim_{x \rightarrow \infty} F_X(x) = \lim_{x \rightarrow \infty} a \left(1 - \frac{1}{x}\right) = a \Rightarrow a = 1.$$

$$f_X(x) = \frac{d}{dx} F_X(x) = \frac{1}{x^2}, \quad 1 < x < \infty.$$

(b) Determine the constant b as a function of x and compute $E[Y|X=4]$. 1p.

Solution:

$$1 = \int_0^x by dy = \left[\frac{by^2}{2} \right]_0^x = \frac{bx^2}{2} \Rightarrow b = \frac{2}{x^2}.$$

$$E[Y|X=4] = \int y \cdot f_{Y|X=4}(y) dy = \int_0^4 \frac{2y^2}{16} dy = \left[\frac{2}{16} \frac{y^3}{3} \right]_0^4 = \frac{8}{3}.$$

(c) Compute the joint density function of X and Y . Are X and Y independent? 1p.

Solution:

$$f_{X,Y}(x, y) = f_{Y|X=x}(y) f_X(x) = \frac{2y}{x^2} \cdot \frac{1}{x^2} = \frac{2y}{x^4}, \quad \begin{matrix} \max(1, y) < x < \infty \\ 0 < y < x \end{matrix}.$$

X and Y are not independent because of the non-rectangular definition region.

(d) Compute the marginal density of Y and the probability $P(Y < 2)$. 2p.

Solution:

$$f_Y(y) = \int f_{X,Y}(x, y) dx = \int_{\max(1, y)}^{\infty} \frac{2y}{x^4} dx = \left[-\frac{2y}{3x^3} \right]_{\max(1, y)}^{\infty} = \begin{cases} \frac{2}{3}y & , 0 < y < 1 \\ \frac{2}{3y^2} & , 1 < y < \infty \\ 0 & , otherwise \end{cases}$$

$$P(Y < 2) = \int_0^2 f_Y(y) dy = \int_0^1 \frac{2}{3} y dy + \int_1^2 \frac{2}{3y^2} dy = \left[\frac{y^2}{3} \right]_0^1 + \left[-\frac{2}{3y} \right]_1^2 = \frac{1}{3} - \frac{1}{3} + \frac{2}{3} = \frac{2}{3}.$$

2. Suppose that X and Y are random variables with joint density

$$f_{X,Y}(x, y) = \begin{cases} \frac{2}{3}(2x + y) & , \begin{cases} 0 < x < 1 \\ 0 < y < 1 \end{cases} \\ 0 & , otherwise \end{cases}.$$

(a) Compute $E[2X + Y]$. 2p.

Solution:

$$\begin{aligned} E[2X + Y] &= \iint (2x + y) f_{X,Y}(x, y) dx dy = \int_{y=0}^1 \int_{x=0}^1 \frac{2}{3} (2x + y)^2 dx dy \\ &= \int_{y=0}^1 \int_{x=0}^1 \frac{2}{3} (4x^2 + 4xy + y^2) dx dy = \int_{y=0}^1 \left[\frac{2}{3} \left(\frac{4}{3}x^3 + 2x^2y + y^2x \right) \right]_0^1 dy \\ &= \int_{y=0}^1 \frac{2}{3} \left(\frac{4}{3} + 2y + y^2 \right) dy = \left[\frac{2}{3} \left(\frac{4}{3}y + y^2 + \frac{y^3}{3} \right) \right]_0^1 = \frac{2}{3} \left(\frac{4}{3} + 1 + \frac{1}{3} \right) = \frac{16}{9}. \end{aligned}$$

(b) Determine the distribution of $2X + Y$. 3p.

Solution: Define

$$\begin{cases} U &= 2X + Y \\ V &= X \end{cases} \Leftrightarrow \begin{cases} X &= V \\ Y &= U - 2V \end{cases} \Rightarrow |J| = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 0 & 1 \\ 1 & -2 \end{vmatrix} = |-1| = 1.$$

The transformation theorem gives

$$f_{U,V}(u, v) = f_{X,Y}(v, u - 2v) \cdot |J| = \frac{2}{3} (2v + u - 2v) \cdot 1 = \frac{2}{3} u, \quad \begin{cases} 2v < u < 2v + 1 \\ \max(0, \frac{u-1}{2}) < v < \min(1, \frac{u}{2}) \end{cases}.$$

Thus

$$\begin{aligned}
 f_U(u) &= \int f_{U,V}(u,v) dv = \int_{\max(0, \frac{u-1}{2})}^{\min(1, \frac{u}{2})} \frac{2}{3} u dv = \left[\frac{2}{3} uv \right]_{\max(0, \frac{u-1}{2})}^{\min(1, \frac{u}{2})} \\
 &= \begin{cases} \frac{2}{3} u \left(\frac{u}{2} - 0 \right) & , 0 < u < 1 \\ \frac{2}{3} u \left(\frac{u}{2} - \frac{u-1}{2} \right) & , 1 < u < 2 \\ \frac{2}{3} u \left(1 - \frac{u-1}{2} \right) & , 2 < u < 3 \end{cases} = \begin{cases} \frac{u^2}{3} & , 0 < u < 1 \\ \frac{1}{3} u & , 1 < u < 2 \\ u - \frac{u^2}{3} & , 2 < u < 3 \end{cases}
 \end{aligned}$$

3. Let X_1 and X_2 follow a multivariate normal distribution with mean vector $\mu = (1, 0)'$ and covariance matrix

$$\Sigma = \begin{pmatrix} 4 & -1 \\ -1 & 1 \end{pmatrix}.$$

Define Y_1, Y_2 and Y_3 through

$$\begin{cases} Y_1 &= X_1 + X_2 \\ Y_2 &= -X_1 + 2X_2 \\ Y_3 &= X_2 - 1 \end{cases}.$$

- (a) What is the joint distribution of Y_1, Y_2 and Y_3 ? 1.5p.

Solution: Let $Y = (Y_1, Y_2, Y_3)' = BX + b$ with

$$B = \begin{pmatrix} 1 & 1 \\ -1 & 2 \\ 0 & 1 \end{pmatrix}, b = \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}.$$

Then

$$Y \sim N(B\mu + b, B\Sigma B')$$

that is

$$Y \sim N \left(\begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix}, \begin{pmatrix} 3 & -3 & 0 \\ -3 & 12 & 3 \\ 0 & 3 & 1 \end{pmatrix} \right).$$

- (b) Are any of Y_1, Y_2 and Y_3 independent? 1p.

Solution: Yes, Y_1 and Y_3 are independent since element $(3, 1)$ in the covariance matrix is zero. Y_2 is dependent with both others.

- (c) Suppose $X_n \sim \text{Bin}(n, \lambda/n)$. Show that $X_n \xrightarrow{d} \text{Po}(\lambda)$ as $n \rightarrow \infty$. 2.5p.

Solution:

$$\begin{aligned}
 g_{\text{Bin}(n,p)}(t) &= (q + pt)^n \\
 \Rightarrow g_{X_n}(t) &= \left(1 - \frac{\lambda}{n} + \frac{\lambda t}{n} \right)^n \\
 &= \left(1 + \frac{\lambda(t-1)}{n} \right)^n \rightarrow e^{\lambda(t-1)}, n \rightarrow \infty.
 \end{aligned}$$

Since $g_{\text{Po}(\lambda)}(t) = e^{\lambda(t-1)}$ we have

$$X_n \xrightarrow{d} \text{Po}(\lambda)$$

through Theorem 6.4.1.

4. Let $X_k, k = 1, 2, \dots$ be independent random variables, with common density $f_X(x)$ and distribution $F_X(x)$. Also, let N be a positive integer-valued random variable with probability generating function $g_N(t)$. Assume that N and X_1, X_2, \dots are independent. Define

$$Z_N = \max(X_1, X_2, \dots, X_N).$$

(a) Derive the density of $Z_N|N = n$.

2p.

Solution: Denote $Z_N|N = n$ as Z_n . The distribution function of Z_n is

$$F_{Z_n}(z) = P(Z_n \leq z) = P(X_1 \leq z, X_2 \leq z, \dots, X_n \leq z)$$

since Z_n is the maximum and $Z_n \leq z$ is therefore the same as all X_1, \dots, X_n being smaller than z . Now, because of independence and that all X_k have same distribution, we have

$$F_{Z_n}(z) = (F_X(z))^n.$$

Taking the derivative of both sides yields

$$f_{Z_n}(z) = n (F_X(z))^{n-1} f_X(z).$$

(b) Show that $F_{Z_N}(z) = g_N(F_X(z))$.

1.5p.

Solution:

$$\begin{aligned} F_{Z_N}(z) &= \sum_n F_{Z_N|N=n}(z) p_N(n) = \sum_n (F_X(z))^n p_N(n) = E[(F_X(z))^N] \\ &= g_N(F_X(z)). \end{aligned}$$

(c) Now, assume X_1, X_2, \dots are all $U(0, 1)$ -distributed and $N \sim Ge(\frac{1}{2})$. Compute $F_{Z_N}(z)$.

1.5p.

Solution:

$$\begin{aligned} F_X(x) &= x, \quad 0 < x < 1 \\ g_N(t) &= \frac{\frac{1}{2}}{1 - \frac{1}{2}t} = \frac{1}{2-t}. \end{aligned}$$

Thus

$$F_{Z_N}(z) = \frac{1}{2-z}, \quad 0 < z < 1.$$