

## Exam in Probability Theory, 6 credits

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

**Solution:**

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

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

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

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:**

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

(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:**

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

Thus

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