4. Find the conditional density function and expectation of Y given X when they have joint density function:

(a)
$$f(x, y) = \lambda^2 e^{-\lambda y}$$
 for $0 \le x \le y < \infty$,

(b)
$$f(x, y) = xe^{-x(y+1)}$$
 for $x, y \ge 0$.

$$\hat{\beta}_{x}^{2}: (\alpha) \quad f_{x}(x) = \int_{x}^{\infty} \lambda^{2} e^{-\lambda y} \, dy = \lambda^{2} \int_{x}^{\infty} e^{-\lambda y} \, dy = \lambda^{2} \cdot \left(-\frac{1}{\lambda} e^{-\lambda y}\right)_{x}^{\infty} = \lambda e^{-\lambda x}, \quad 0 \leq x \leq \alpha$$

$$f_{x}(x) = \frac{f(x, y)}{f_{x}(x)} = \frac{\lambda^{2} e^{-\lambda y}}{\lambda e^{-\lambda x}} = \lambda e^{\lambda(x-y)}, \quad 0 \leq x \leq y < \infty$$

$$E[Y|X=x]=\int_{x}^{\infty}y\lambda e^{\lambda(x-y)}dy=x+\frac{1}{\lambda}$$

(b)
$$f_{x}(x) = \int_{0}^{\infty} x e^{-x(y+1)} dy = x \int_{0}^{\infty} e^{-x(y+1)} dy = x \cdot \frac{1}{x} e^{-x(y+1)} \Big|_{0}^{\infty} = e^{-x}$$

$$f_{Y|x}(y|x) = \frac{f(x,y)}{f_{x}(x)} = \frac{x e^{-x(y+1)}}{e^{-x}} = x e^{-xy} . x, y \ge 0$$

$$E[Y|x=x] = \int_{0}^{\infty} y \times e^{-xy} dy = \frac{1}{x}$$

4,6

- **8.** Let X, Y, Z be independent and exponential random variables with respective parameters λ, μ, ν . Find $\mathbb{P}(X < Y < Z)$.
- **9.** Let X and Y have the joint density $f(x, y) = cx(y x)e^{-y}$, $0 \le x \le y < \infty$.
- (a) Find *c*.
- (b) Show that:

$$f_{X|Y}(x \mid y) = 6x(y - x)y^{-3}, \qquad 0 \le x \le y,$$

 $f_{Y|X}(y \mid x) = (y - x)e^{x - y}, \qquad 0 \le x \le y < \infty.$

- (c) Deduce that $\mathbb{E}(X \mid Y) = \frac{1}{2}Y$ and $\mathbb{E}(Y \mid X) = X + 2$.
- 8. $\hat{\mathbf{A}}_{\mathbf{x}}^{\mathbf{x}} : P(\mathbf{x} < \mathbf{y} < \mathbf{z}) = \int_{0}^{\infty} \int_{\mathbf{x}}^{\infty} \int_{\mathbf{y}}^{\infty} \lambda \mu \mathbf{v} e^{-\lambda \mathbf{x} \mu \mathbf{y} \mathbf{v} \cdot \mathbf{z}} \, d\mathbf{x} d\mathbf{y} d\mathbf{z}$ $= \int_{0}^{\infty} \int_{\mathbf{x}}^{\infty} (-\lambda \mu e^{-\lambda \mathbf{x} \mu \mathbf{y} \mathbf{v} \cdot \mathbf{z}} | \mathbf{y}) \, d\mathbf{x} d\mathbf{y}$ $= \int_{0}^{\infty} \int_{\mathbf{x}}^{\infty} \lambda \mu e^{-\lambda \mathbf{x} (\mu + \mathbf{v}) \cdot \mathbf{y}} \, d\mathbf{y} d\mathbf{x}$ $= \int_{0}^{\infty} \frac{\lambda \mu}{\mu + \nu} e^{-(\lambda + \mu + \nu) \cdot \mathbf{x}} \, d\mathbf{x}$ $= \frac{\lambda \mu}{(\mu + \nu)(\lambda + \mu + \nu)}$

$$\begin{array}{ll}
\P. \hat{\mathbb{R}}^{2} : (1) \ f_{x}(x) = C \int_{x}^{\infty} \chi(y - x) e^{-y} dy = Cx \int_{x}^{\infty} y e^{-y} dy - Cx^{2} \int_{x}^{\infty} e^{-y} dy \\
&= Cx(x + 1) e^{-x} - Cx^{2} e^{-x} = Cx e^{-x} \\
1 &= \int_{0}^{\infty} f_{x}(x) = \int_{0}^{\infty} Cx e^{-x} = C \cdot (-x + 1) e^{-x} \Big|_{0}^{\infty} = C \implies C = 1
\end{array}$$

$$f_{X}(x) = Xe^{-x} \qquad f(x,y) = X(y-x)e^{-y}$$

$$f_{Y}(y) = \int_{0}^{y} X(y-x)e^{-y} dx = e^{-y} \int_{0}^{y} X(y-x) dx = e^{-y} \cdot y \int_{0}^{y} X dx - e^{-y} \int_{0}^{y} X^{2} dx = \frac{1}{6}y^{3} \cdot e^{-y}$$

$$\Rightarrow f_{X|Y}(x|y) = \frac{f(x,y)}{f_{Y}(y)} = 6x(y-x) \cdot y^{-3} \qquad f_{Y|X}(y|x) = \frac{f(x,y)}{f_{X}(x)} = (y-x)e^{x-y}$$

10. Let $\{X_r : r \geq 0\}$ be independent and identically distributed random variables with density function f and distribution function F. Let $N = \min\{n \geq 1 : X_n > X_0\}$ and $M = \min\{n \geq 1 : X_0 \geq X_1 \geq \cdots \geq X_{n-1} < X_n\}$. Show that X_N has distribution function $F + (1 - F) \log(1 - F)$, and find $\mathbb{P}(M = m)$.

角年:
$$\{N > n\}$$
 意味着 X_o 是 X_1, \dots, X_N 中最大的 $P(N > n) = \frac{1}{N+1}$

$$P(N = N) = P(N > N-1) - P(N > N) = \frac{1}{N} - \frac{1}{N+1} = \frac{1}{N(N+1)}$$

$$P(X_N \le X) = \sum_{n=1}^{\infty} P(X_N \le X, N = N) = \sum_{n=1}^{\infty} F(X)^{n+1} \cdot \frac{1}{N(N+1)}$$

$$= \sum_{n=1}^{\infty} \frac{F(X)^{n+1}}{N} - \sum_{n=1}^{\infty} \frac{F(X)^{n+1}}{N+1} = F(X) \sum_{n=1}^{\infty} \frac{F(X)^n}{N} - \sum_{n=1}^{\infty} \frac{F(X)^n}{N} + F(X)$$

$$= F(X) + (F(X) - 1) \sum_{n=1}^{\infty} \frac{F(X)^n}{N} = F(X) + (1 - F(X)) \log (1 - F(X))$$

$$P(M = M) = P(X_0 \ge X_1 \ge \dots \ge X_{M-1}) - P(X_0 \ge X_1 \ge \dots \ge X_M)$$

$$= \frac{1}{M} - \frac{1}{(M+1)^{-1}}$$

2. Let X and Y be independent exponential random variables with parameter 1. Find the joint density function of U = X + Y and V = X/(X + Y), and deduce that V is uniformly distributed on [0, 1].

4.7

5. Normal orthant probability. Let X and Y have the bivariate normal density function

$$f(x, y) = \frac{1}{2\pi\sqrt{1-\rho^2}} \exp\left\{-\frac{1}{2(1-\rho^2)}(x^2 - 2\rho xy + y^2)\right\}.$$

Show that X and $Z = (Y - \rho X)/\sqrt{1 - \rho^2}$ are independent N(0, 1) variables, and deduce that

$$\mathbb{P}(X > 0, Y > 0) = \frac{1}{4} + \frac{1}{2\pi} \sin^{-1} \rho.$$

$$\mathcal{V}_{\Gamma}: \left| \frac{\Im(X, \xi)}{\Im(X, y)} \right| = \left| \frac{1}{1 - \rho^2} \frac{1}{\sqrt{1 - \rho^2}} \right| = \frac{1}{\sqrt{1 - \rho^2}}$$

$$\chi^2 - 2\rho \times y + y^2 = (\chi^2 + \xi^2) (1 - \rho^2)$$

$$f_{x,z}(x,\xi) = \frac{1}{2\pi(1-\rho^2)} \exp\left(-\frac{1}{2(1-\rho^2)}(x^2+\xi^2)(1-\rho^2)\right) \cdot \sqrt{1-\rho^2} \, dxd\xi$$

$$= \frac{1}{2\pi} e^{-\frac{1}{2}(x^2+\xi^2)} = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\xi^2} = f_x(x) \cdot f_{\xi}(\xi)$$

⇒ X、そ独を且同 N(0,1)分布

$$\{x>0.7>0\} = \{x>0.2> - \frac{(x)}{\sqrt{1-\rho^2}}\}$$

tana =
$$-\frac{\rho}{\sqrt{1-\rho^2}}$$
 $\Rightarrow \alpha = -\arctan\frac{\rho}{\sqrt{1-\rho^2}}$

$$P(x>0, 7>0) = \int_{\alpha}^{\frac{\pi}{2}} \int_{\alpha}^{\infty} \frac{1}{2\pi} e^{-\frac{1}{2}t^{2}} \cdot r dr d\theta = \int_{\alpha}^{\frac{\pi}{2}} \frac{1}{2\pi} d\theta = \frac{1}{2\pi} (\frac{\pi}{2} - \alpha) = \frac{1}{4} + \frac{1}{2\pi} \sin^{2}\theta$$

9. A point (X, Y, Z) is picked uniformly at random inside the unit ball of \mathbb{R}^3 . Find the joint density of Z and R, where $R^2 = X^2 + Y^2 + Z^2$.

角中·单位球体积V= 等

P(Rsr. Z=Z)= 3/4 /2 T(12-W2)dw

求导得 f(v, ≥)=3/2 r as (≥)<r<1.



4.9

3. Let $\mathbf{X} = (X_1, X_2, \dots, X_n)$ have the $N(\boldsymbol{\mu}, \mathbf{V})$ distribution, and show that $Y = a_1 X_1 + a_2 X_2 + \dots + a_n X_n$ has the (univariate) $N(\boldsymbol{\mu}, \sigma^2)$ distribution where

$$\mu = \sum_{i=1}^{n} a_i \mathbb{E}(X_i), \qquad \sigma^2 = \sum_{i=1}^{n} a_i^2 \operatorname{var}(X_i) + 2 \sum_{i < j} a_i a_j \operatorname{cov}(X_i, X_j).$$

$$Var(\Upsilon) = Var(\sum_{i=1}^{n} \alpha_{i}X_{i}) = \sum_{i=1}^{n} Var(\alpha_{i}X_{i}) + 2\sum_{i < j} Cov(\alpha_{i}X_{i}, \alpha_{j}X_{j})$$
$$= \sum_{i=1}^{n} \alpha_{i}^{2} Var(X_{i}) + 2\sum_{i < j} \alpha_{i}\alpha_{j} Cov(X_{i}, X_{j})$$

4.9

7. Let the vector $(X_r: 1 \le r \le n)$ have a multivariate normal distribution with covariance matrix $\mathbf{V} = (v_{ij})$. Show that, conditional on the event $\sum_{i=1}^{n} X_r = x$, X_1 has the N(a, b) distribution where $a = (\rho s/t)x$, $b = s^2(1 - \rho^2)$, and $s^2 = v_{11}$, $t^2 = \sum_{ij} v_{ij}$, $\rho = \sum_{i} v_{i1}/(st)$.

$$i\mathcal{F}: \alpha = \mathcal{F}[X_1 | \sum_{i=1}^{n} X_i = x]$$
 $b = Var(X_1 | \sum_{i=1}^{n} X_i = x)$

Lem 7. Let X and Y have a bivariate normal density with zero means, variances σ^2 , τ^2 , and correlation ρ . Show that:

(a)
$$\mathbb{E}(X \mid Y) = \frac{\rho \sigma}{\tau} Y$$
,

(b)
$$var(X \mid Y) = \sigma^2 (1 - \rho^2),$$

