

1. 随机向量函数的分布

§3.4 两个随机变量的函数

- 设 $\xi = (X, Y)$ 有联合密度 $p(x, y)$, 求 $Z = f(X, Y)$ 的密度.
- 分布函数法: 第一步, 用 $p(x, y)$ 表达 F_Z ,

$$F_Z(z) = P(f(X, Y) \leq z) = \iint_{\{(x, y): f(x, y) \leq z\}} p(x, y) dx dy.$$

- 第二步, 将** 化为如下积分,

$$** = \int_{-\infty}^z p(u) du.$$

- 结论: $p_Z(z) = p(z)$.

例7.3. 假设 X 与 Y 相互独立, $X \sim \mathcal{P}(\lambda_1)$, $Y \sim \mathcal{P}(\lambda_2)$.

令 $Z = X + Y$. 求: Z 的分布; 在 $Z = n$ 的条件下, X 的条件分布.

- $Z \sim \mathcal{P}(\lambda_1 + \lambda_2): \forall n \geq 0, \quad P(Z = n)$

$$\begin{aligned} &= \sum_{k=0}^n P(X = k, Y = n - k) = \sum_{k=0}^n \frac{\lambda_1^k}{k!} e^{-\lambda_1} \frac{\lambda_2^{n-k}}{(n-k)!} e^{-\lambda_2} \\ &= \frac{1}{n!} \left(\sum_{k=0}^n C_n^k \lambda_1^k \lambda_2^{n-k} \right) e^{-(\lambda_1 + \lambda_2)} = \frac{(\lambda_1 + \lambda_2)^n}{n!} e^{-(\lambda_1 + \lambda_2)}. \end{aligned}$$

- $n = 0$ 时, $P(X = 0|Z = 0) = 1$.

- $n > 0$ 时, 记 $p = \frac{\lambda_1}{\lambda_1 + \lambda_2}$, $q = 1 - p$. 则, $k = 0, 1, \dots, n$,

$$P(X = k|Z = n) = \frac{C_n^k \lambda_1^k \lambda_2^{n-k}}{(\lambda_1 + \lambda_2)^n} = C_n^k p^k q^{n-k},$$

- 条件分布是二项分布.

例7.4. 假设 X 与 Y 相互独立, $X \sim B(n_1, p)$, $Y \sim B(n_2, p)$, $0 < p < 1$. 令 $Z = X + Y$. 求: Z 的分布; 在 $Z = n$ 的条件下, X 的条件分布.

- $n = 0, 1, \dots, n_1 + n_2$. 记 $q = 1 - p$, 则 $P(Z = n)$

$$\begin{aligned} &= \sum_{k=0}^n P(X = k, Y = n - k) = \sum_{k=0}^n C_{n_1}^k C_{n_2}^{n-k} p^{k+n-k} q^{n_1-k+n_2-(n-k)} \\ &= \sum_{k=0}^n C_{n_1}^k C_{n_2}^{n-k} p^n q^{n_1+n_2-n} = C_{n_1+n_2}^n p^n q^{n_1+n_2-n}. \end{aligned}$$

- $n = 0$ 时, $P(X = 0 | Z = 0) = 1$.
- $n > 0$ 时, $k = 0, 1, \dots, n$,

$$P(X = k | Z = n) = \frac{C_{n_1}^k C_{n_2}^{n-k}}{C_{n_1+n_2}^n},$$

- 条件分布是超几何分布.

- 定理4.1. 设 $\xi = (X, Y)$ 有联合密度 $p(x, y)$, $Z = X + Y$. 则

$$p_Z(z) = \int_{-\infty}^{\infty} p(x, z - x) dx.$$

- 证: 第一步,

$$F_Z(z) = P(X + Y \leq z) = \iint_{x+y \leq z} p(x, y) dx dy.$$

- 第二步,

$$\star = \int_{-\infty}^{\infty} \left(\int_{-\infty}^{z-x} p(x, y) dy \right) dx = \int_{-\infty}^z \int_{-\infty}^{\infty} p(x, u - x) dx du$$

- 推论(系4.1): 若 X, Y 相互独立, 分别有密度 p_X, p_Y , 则 $Z = X + Y$ 是连续型, 且

$$p_Z(z) = \int_{-\infty}^{\infty} p_X(x) p_Y(z - x) dx = \int_{-\infty}^{\infty} p_X(z - y) p_Y(y) dy.$$

例4.1 & 4.2. 设 (X, Y) 服从二维正态分布, 联合密度 $p(x, y)$ 为

$$p(x, y) = \hat{C} \exp \left\{ -\frac{u^2 - 2\rho uv + v^2}{2(1 - \rho^2)} \right\}, \quad \left(u = \frac{x - \mu_1}{\sigma_1}, v = \frac{y - \mu_2}{\sigma_2} \right),$$

其中, $\hat{C} = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}}$. 求 $Z = X + Y$ 的密度.

• $p_Z(z) = \int_{-\infty}^{\infty} p(x, z - x) dx$. 当 y 取 $z - x$ 时,

$$v = \frac{y - \mu_2}{\sigma_2} = \frac{z - (\mu_1 + \sigma_1 u) - \mu_2}{\sigma_2} = C - \frac{\sigma_1}{\sigma_2} u,$$

其中, $C = (z - \mu_1 - \mu_2)/\sigma_2$.

• 此时, $u^2 - 2\rho uv + v^2$

$$\begin{aligned} &= u^2 - 2\rho u \left(C - \frac{\sigma_1 u}{\sigma_2} \right) + \left(C - \frac{\sigma_1 u}{\sigma_2} \right)^2 \\ &= \left(1 + 2\rho \frac{\sigma_1}{\sigma_2} + \left(\frac{\sigma_1}{\sigma_2} \right)^2 \right) u^2 - 2 \left(\rho + \frac{\sigma_1}{\sigma_2} \right) C u + C^2. \end{aligned}$$

- 目标: 计算 $p_Z(z) = \int_{-\infty}^{\infty} p(x, z-x) dx$. 已有:

$$p(x, z-x) = \hat{C} \left\{ -\frac{Au^2 - 2Bu + C^2}{2(1-\rho^2)} \right\}, \quad \text{其中, } u = \frac{x - \mu_1}{\sigma_1},$$

$$A = 1 + 2\rho \frac{\sigma_1}{\sigma_2} + \left(\frac{\sigma_1}{\sigma_2} \right)^2, \quad B = \left(\rho + \frac{\sigma_1}{\sigma_2} \right) C, \quad C = \frac{z - (\mu_1 + \mu_2)}{\sigma_2}.$$

- 配方:

$$Au^2 - 2Bu + C^2 = A \left(u - \frac{B}{A} \right)^2 - \left(\frac{B^2}{A} - C^2 \right).$$

- 于是, $p_Z(z)$

$$\begin{aligned} &= \hat{C} \exp \left\{ \frac{\frac{B^2}{A} - C^2}{2(1-\rho^2)} \right\} \times \int_{-\infty}^{\infty} \exp \left\{ -\frac{A \left(u - \frac{B}{A} \right)^2}{2(1-\rho^2)} \right\} \sigma_1 du \\ &= \tilde{C} \exp \left\{ \frac{B^2 - AC^2}{2(1-\rho^2)A} \right\}. \quad \tilde{C} = \hat{C} \sigma_1 \sqrt{2\pi \frac{1-\rho^2}{A}} = \frac{1}{\sqrt{2\pi \sigma_2^2 A}}. \end{aligned}$$

- 已有: $p_Z(z) = \tilde{C} \exp \left\{ \frac{B^2 - AC^2}{2(1-\rho^2)A} \right\}$, 其中 \tilde{C} 是常数,

$$A = 1 + 2\rho \frac{\sigma_1}{\sigma_2} + \left(\frac{\sigma_1}{\sigma_2} \right)^2, \quad B = \left(\rho + \frac{\sigma_1}{\sigma_2} \right) C, \quad C = \frac{z - (\mu_1 + \mu_2)}{\sigma_2}.$$

- $B^2 - AC^2$

$$= \left(\left(\rho + \frac{\sigma_1}{\sigma_2} \right)^2 - A \right) C^2 = (\rho^2 - 1) \frac{(z - (\mu_1 + \mu_2))^2}{\sigma_2^2}.$$

- 因此,

$$p_Z(z) = \tilde{C} \exp \left\{ -\frac{(z - \mu)^2}{2\sigma^2} \right\} = \frac{1}{\sqrt{2\pi}\sigma} \exp \left\{ -\frac{(z - \mu)^2}{2\sigma^2} \right\}.$$

其中, $\mu = \mu_1 + \mu_2$, $\sigma^2 = \sigma_2^2 A = \sigma_1^2 + \sigma_2^2 + 2\rho\sigma_1\sigma_2$.

- 特别地, 若 $\rho = 0$ (即 X, Y 相互独立), 则

$$X + Y \sim N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2).$$

- 定理4.2. 设 (X, Y) 有联合密度 $p(x, y)$.

令 $Z = X/Y$ (当 $Y = 0$ 时, 规定 $Z = 0$). 则 Z 为连续型, 且

$$p_Z(z) = \int_{-\infty}^{\infty} |y|p(z y, y)dy.$$

- 证明: 第一步, $\frac{x}{y} \leq z$ 当且仅当 “ $y > 0$ 且 $x \leq yz$ ” 或者 “ $y < 0$ 且 $x \geq yz$.” 于是,

$$F_Z(z) = P(Y > 0, X \leq Yz) + P(Y < 0, X \geq Yz).$$

- $P(Y > 0, X \leq Yz)$

$$\begin{aligned} &= \int_0^{\infty} \int_{-\infty}^{yz} p(x, y) dx dy = \int_0^{\infty} \int_{-\infty}^z p(yu, y) y du dy \\ &= \int_{-\infty}^z \left(\int_0^{\infty} yp(yu, y) dy \right) du. \end{aligned}$$

- 类似处理**, 即可.

例4.4. X, Y 相互独立, 都服从 $N(0, 1)$. 求 $Z = X/Y$ 的密度.

• 联合密度:

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

• 因此,

$$\begin{aligned} p_Z(z) &= \int_{-\infty}^{\infty} |y| p(z y, y) dy = \int_{-\infty}^{\infty} |y| \frac{1}{2\pi} \exp \left\{ -\frac{(zy)^2 + y^2}{2} \right\} dy \\ &= \frac{2}{2\pi} \int_0^{\infty} y \exp \left\{ -\frac{(z^2 + 1)y^2}{2} \right\} dy \\ &= \frac{1}{\pi} \int_0^{\infty} e^{-(z^2 + 1)u} du = \frac{1}{\pi(z^2 + 1)}. \end{aligned}$$

- 定理4.3. 假设 $\xi = (X, Y)$ 为连续型, 有密度 $p(x, y)$.
假设

$$\eta = (U, V), \quad \text{其中 } U = f(X, Y), \quad V = g(X, Y).$$

如果(1) $P(\xi \in A) = 1$ 且 $(f, g) : A \rightarrow G$ 是一对一的;

(2) $f, g \in C^1(A)$, 且 $\frac{\partial(u, v)}{\partial(x, y)} \neq 0, \forall (x, y) \in A$,
那么, η 是连续型, 且

$$p_{U, V}(u, v) = p\left(x(u, v), y(u, v)\right) \left| \frac{\partial(x, y)}{\partial(u, v)} \right|, \quad (u, v) \in G.$$

- 证: $\forall D \subseteq G$, 找 $D^* \subseteq A$ 使得 $\eta \in D$ iff $\xi \in D^*$. 于是,

$$P(\star) = P(\star) = \iint_{D^*} p(x, y) dx dy = \iint_D \star \star dudv.$$

例4.5, 4.7, & 习题三、21. 假设 X, Y 相互独立, 都服从 $N(0, 1)$.

- 用极坐标表达:

$$X = R \cos \Theta, \quad Y = R \sin \Theta.$$

- $A = \{(x, y) : x \neq 0, y \neq 0\}$,

$$G = \{(r, \theta) : r > 0, 0 < \theta < 2\pi; \theta \neq \frac{\pi}{2}, \pi, \frac{3\pi}{2}\}.$$

- $\frac{\partial(x, y)}{\partial(r, \theta)} = \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} = r.$

- $p_{R, \Theta}(r, \theta) = \frac{1}{2\pi} e^{-\frac{1}{2}(x^2+y^2)} \left| \frac{\partial(x, y)}{\partial(r, \theta)} \right|$
 $= \frac{1}{2\pi} r e^{-\frac{1}{2}r^2}, r > 0, 0 < \theta < 2\pi; \theta \neq \frac{\pi}{2}, \pi, \frac{3\pi}{2}.$

- R, Θ 独立: $p_{R, \Theta}(r, \theta) = p_R(r) \cdot p_\Theta(\theta).$

- $W := R^2 = X^2 + Y^2 \sim \text{Exp}(\frac{1}{2})$. 因为, $\forall w > 0$,

$$p_W(w) = p_R(r) \frac{dr}{dw} = r \exp\left\{-\frac{r^2}{2}\right\} \cdot \frac{1}{2r} = \frac{1}{2} e^{-\frac{w}{2}}.$$

- $U := e^{-\frac{1}{2}W} \sim U(0, 1)$. 因为, $\forall p \in (0, 1)$,

$$P(U \leq p) = P(W \geq -2 \ln p) = e^{-\frac{1}{2}(-2 \ln p)} = e^{\ln p} = p.$$

- $V := \frac{1}{2\pi} \Theta \sim U(0, 1)$, 且 U 与 V 相互独立.

- $R = \sqrt{-2 \ln U}$, $\Theta = 2\pi V$, 即

$$X = \sqrt{-2 \ln U} \cos(2\pi V), \quad Y = \sqrt{-2 \ln U} \sin(2\pi V).$$

2. 两个随机变量的函数的数学期望

- 随机向量函数的期望(定理4.6):

$$\text{离散型: } Ef(X, Y) = \sum_{i,j} f(x_i, y_j) P(X = x_i, Y = y_j).$$

$$\text{连续型: } Ef(X, Y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) p(x, y) dx dy.$$

- 定理4.4. 若 X 与 Y 相互独立, 则 $EXY = (EX) \cdot (EY)$.
- 例, 连续型的证明:

$$EXY = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xyp_X(x)p_Y(y) dx dy = (EX)(EY).$$

- 定理4.5. 若 X 与 Y 相互独立,
则 $\text{var}(X + Y) = \text{var}(X) + \text{var}(Y)$.
- 证: 左 = $E(X + Y - (EX + EY))^2$
= 右 + $2E(X - EX)(Y - EY)$.