概率统计第七次作业

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2.

$$\begin{split} & :: F(x_i,y_i) = F_x(x_i)F_y(y_j) \\ & :: p_{ij} = F(x_i,y_i) - F(x_i-1,y_i) - F(x_i,y_i-1) + F(x_i-1,y_i-1) \\ & = F_x(x_i)F_y(y_i) - F_x(x_i-1)F_y(y_i) - F_x(x_i)F_y(y_i-1) + F_x(x_i-1)F_y(y_i-1) \\ & = [F_x(x_i) - F_x(x_i-1)]F_y(y_i) - [F_x(x_i) - F_x(x_i-1)]F_y(y_i-1) \\ & = [F_x(x_i) - F_x(x_i-1)][F_y(y_i) - F_y(y_i-1)] \\ & = p_{i\cdot}p_{\cdot j} \end{split}$$

3.

$$\therefore f_x(x) = \frac{\exp(-\frac{1}{2}(\frac{x-\mu_x}{\sigma_x})^2)}{2\pi\sqrt{1-\rho^2}\sigma_x} \int_{-\infty}^{+\infty} \exp(-\frac{(t-\frac{\rho(x-\mu_x)}{\sigma_x})^2}{2(1-\rho^2)}) \mathrm{d}t$$

$$\cdots$$
 由正态分布可知 $\dfrac{1}{\sqrt{2\pi}\sqrt{1-\rho^2}}\int_{-\infty}^{+\infty}\exp(-\dfrac{(t-\dfrac{
ho(x-\mu_x)}{\sigma_x})^2}{2(1-\rho^2)})\mathrm{d}t=1$

$$\therefore f_x(x) = \frac{\exp(-\frac{1}{2}(\frac{x-\mu_x}{\sigma_x})^2)}{\sqrt{2\pi}\sigma_x}$$

$$\therefore X \sim N(\mu_x, \sigma_x^2)$$

4.

3.

(1)

$$1 = \int_{2}^{4} dy \int_{0}^{2} k(6 - x - y) dx = \int_{2}^{4} 2k (5 - y) dy = 8k$$

$$k = \frac{1}{8}$$

$$\therefore P(X < 1, Y < 3) = \int_{2}^{3} dy \int_{0}^{1} \frac{1}{8} (6 - x - y) dx = \int_{2}^{3} (\frac{11}{16} - \frac{y}{8}) dy = \frac{3}{8}$$

(3)

$$\therefore P(X < 1.5) = \int_{2}^{4} dy \int_{0}^{1.5} \frac{1}{8} (6 - x - y) dx = \int_{2}^{4} (\frac{63}{64} - \frac{3y}{16}) dy = \frac{27}{32}$$

(4)

$$\therefore P(X+Y \leqslant 4) = \int_{2}^{4} dy \int_{0}^{4-y} \frac{1}{8} (6-x-y) dx = \int_{2}^{4} \frac{(y-8)(y-4)}{16} dy = \frac{2}{3}$$

4.

(1)

$$P(X < Y) = \int_0^{+\infty} \mathrm{d}y \int_0^y f_x(x) f_y(y) \mathrm{d}x = \int_0^{+\infty} F_x(y) f_y(y) \mathrm{d}y = \int_0^{+\infty} F_x(x) f_y(x) \mathrm{d}x$$

(2)

$$\therefore F_x(x) = \int_0^x \lambda_1 e^{-\lambda_1 x} \mathrm{d}x = 1 - e^{-\lambda_1 x}$$

$$\therefore P(X < Y) = \int_0^{+\infty} (1 - e^{-\lambda_1 x}) \lambda_2 e^{-\lambda_2 x} \mathrm{d}x = \frac{\lambda_1}{\lambda_1 + \lambda_2}$$

5.

$$\therefore F_x(x) = \lim_{y o +\infty} F(x,y) = 1 - e^{-x}, x > 0$$

$$F_y(y)=\lim_{x o +\infty}F(x,y)=1-e^{-y}, y>0$$

$$\therefore F_x(x) = egin{cases} 1-e^{-x}, & x>0 \ 0, & x\leqslant 0 \end{cases}$$

$$F_y(y) = egin{cases} 1-e^{-y}, & x>0 \ 0, & x\leqslant 0 \end{cases}$$

6.

所有情况如下:

样本	ннн	ннт	нтн	ТНН	нтт	THT	ттн	TTT
Х	2	2	1	1	1	1	0	0
Υ	3	2	2	2	1	1	1	0

最终可得联合分布律和边缘分布律如下:

Y\X	0	1	2	P(Y=j)
0	1/8	0	0	1/8
1	1/8	2/8	0	3/8

YIX	0	1	2	P(Y=j)
0	0	2/8	1/8	3/8
0	0	0	1/8	1/8
P(X=i)	1/4	2/4	1/4	1

7.

$$f_x = \int_0^x 4.8y(2-x) dy = 2.4(2-x)x^2, 0 \leqslant x \leqslant 1$$

$$f_y = \int_y^1 4.8y(2-x) dx = 2.4(3-4y+y^2), 0 \leqslant y \leqslant 1$$

$$f_x = \begin{cases} 2.4(2-x)x^2, & 0 \leqslant x \leqslant 1\\ 0, & \text{otherwise} \end{cases}$$

$$f_y = \begin{cases} 2.4(3-4y+y^2), & 0 \leqslant y \leqslant 1\\ 0, & \text{otherwise} \end{cases}$$

8.

$$\therefore f_x(x) = egin{cases} \int_x^{+\infty} e^{-y} \mathrm{d}y = e^{-x}, & x > 0 \ 0, & ext{otherwise} \end{cases}$$
 otherwise $f_x(x) = egin{cases} \int_0^y e^{-y} \mathrm{d}x = y e^{-y}, & y > 0 \ 0, & ext{otherwise} \end{cases}$

9.

(1)

$$\therefore 1 = \int_{-1}^{1} dx \int_{x^{2}}^{1} cx^{2}y dy = c \int_{-1}^{1} x^{2} dx \int_{x^{2}}^{1} y dy = c \int_{-1}^{1} x^{2} (\frac{1}{2} - \frac{x^{4}}{2}) dx = \frac{4c}{21}$$
$$\therefore c = \frac{21}{4}$$

(2)

$$f_x(x) = egin{cases} \int_{x^2}^1 rac{21}{4} x^2 y \mathrm{d}y = rac{21}{8} x^2 \left(1 - x^4
ight), & -1 \leqslant x \leqslant 1 \ 0, & ext{otherwise} \end{cases}$$
 otherwise $f_x(x) = egin{cases} \int_{-\sqrt{y}}^{\sqrt{y}} rac{21}{4} x^2 y \mathrm{d}x = rac{7}{2} y^{rac{5}{2}}, & 0 \leqslant y \leqslant 1 \ 0, & ext{otherwise} \end{cases}$

17.

(1)

$$\therefore F_x(x) = F(x,+\infty) = egin{cases} 1 - e^{-ax}, & x \geqslant 0 \ 0, & ext{otherwise} \end{cases}$$

$$F_y(y) = F(+\infty,y) = egin{cases} y, & 0 \leqslant y \leqslant 1 \ 1, & y > 1 \ 0, & ext{otherwise} \end{cases}$$

$$\therefore F_x(x)F_y(y) = F(x,y) = egin{cases} (1-e^{-ax})y, & 0\leqslant y\leqslant 1\ 1-e^{-ax}, & y>1\ 0, & ext{otherwise} \end{cases}$$

 $\therefore X, Y$ 相互独立.

(2)

$$\therefore P(X=x) = \sum_{y=1}^{+\infty} p^2 (1-p)^{x+y-2} = p^2 (1-p)^{x-1} \sum_{y=1}^{+\infty} (1-p)^{y-1} = p(1-p)^{x-1}$$

$$P(Y = y) = p(1 - p)^{y-1}$$

$$P(X = x, Y = y) = P(X = x)P(Y = y) = p^{2}(1 - p)^{x+y-2}$$

 $\therefore X, Y$ 相互独立.

18.

(1)

$$\therefore f(x,y) = egin{cases} rac{1}{2}e^{-rac{y}{2}}, & 0 < x < 1, y > 0 \ 0, & ext{otherwise} \end{cases}$$

(2)

$$\therefore \Delta = 4X^2 - 4Y \geqslant 0$$
, $\mathbb{P}[X^2 \geqslant Y]$

$$\therefore P(X^2 \geqslant Y) = \int_0^1 \mathrm{d}x \int_0^{x^2} \frac{1}{2} e^{-\frac{y}{2}} \mathrm{d}y = 1 - \int_0^1 e^{-\frac{x^2}{2}} \mathrm{d} = 1 - \sqrt{2}\pi (\Phi(1) - \frac{1}{2})$$

6.

设该正态分布的随机变量被划分成 $egin{pmatrix} X \\ Y \end{pmatrix} = egin{pmatrix} x \\ y \end{pmatrix}$,其中 y 是一个向量. 则有概率密度函数

$$f(x,y) = (2\pi)^{-rac{n}{2}} |\Sigma|^{-rac{1}{2}} \exp\left(-rac{1}{2}egin{pmatrix} x - \mu_x \ y - \mu_y \end{pmatrix}^T egin{pmatrix} \Sigma_{xx} & \Sigma_{xy} \ \Sigma_{yx} & \Sigma_{yy} \end{pmatrix}^{-1} egin{pmatrix} x - \mu_x \ y - \mu_y \end{pmatrix}
ight)$$

由 Schur 补可知

$$\begin{pmatrix} \Sigma_{xx} & \Sigma_{xy} \\ \Sigma_{xy} & \Sigma_{yy} \end{pmatrix} = \begin{pmatrix} I & 0 \\ \Sigma_{yx}\Sigma_{xx}^{-1} & I \end{pmatrix} \begin{pmatrix} \Sigma_{xx} & 0 \\ 0 & \Sigma_{yy} - \Sigma_{yx}\Sigma_{xx}^{-1}\Sigma_{xy} \end{pmatrix} \begin{pmatrix} I & \Sigma_{xx}^{-1}\Sigma_{xy} \\ 0 & I \end{pmatrix}$$

并且我们有

$$\begin{pmatrix} I & 0 \\ \Sigma_{yx}\Sigma_{xx}^{-1} & I \end{pmatrix}^{-1} = \begin{pmatrix} I & 0 \\ -\Sigma_{yx}\Sigma_{xx}^{-1} & I \end{pmatrix}, \begin{pmatrix} I & \Sigma_{xx}^{-1}\Sigma_{xy} \\ 0 & I \end{pmatrix}^{-1} = \begin{pmatrix} I & -\Sigma_{xx}^{-1}\Sigma_{xy} \\ 0 & I \end{pmatrix}$$

$$\Theta_{yy} = \Sigma_{yy} - \Sigma_{yx} \Sigma_{xx}^{-1} \Sigma_{xy}$$

将其带入 f(x,y) 可得

$$\begin{split} f(x,y) &= (2\pi)^{-\frac{n}{2}} |\Sigma|^{-\frac{1}{2}} \exp\left(-\frac{1}{2} \begin{pmatrix} x - \mu_x \\ y - \mu_y \end{pmatrix}^T \begin{pmatrix} \Sigma_{xx} & \Sigma_{xy} \\ \Sigma_{yx} & \Sigma_{yy} \end{pmatrix}^{-1} \begin{pmatrix} x - \mu_x \\ y - \mu_y \end{pmatrix}\right) \\ &= (2\pi)^{-\frac{n}{2}} |\Sigma|^{-\frac{1}{2}} \exp(-\frac{1}{2} (x - \mu_x)^T \Sigma_{xx}^{-1} (x - \mu_x) \\ &- \frac{1}{2} (y - (\mu_y + \Sigma_{yx} \Sigma_{xx}^{-1} (x - \mu_x)))^T \Theta_{yy} (y - (\mu_y + \Sigma_{yx} \Sigma_{xx}^{-1} (x - \mu_x)))) \\ &= (2\pi)^{-\frac{1}{2}} |\Sigma_{xx}|^{-\frac{1}{2}} \exp\left(-\frac{1}{2} (x - \mu_x)^T \Sigma_{xx}^{-1} (x - \mu_x)\right) \\ &\cdot (2\pi)^{-\frac{n-1}{2}} |\Theta_{yy}|^{-\frac{1}{2}} \exp\left(-\frac{1}{2} (y - (\mu_y + \Sigma_{yx} \Sigma_{xx}^{-1} (x - \mu_x)))^T \Theta_{yy} (y - (\mu_y + \Sigma_{yx} \Sigma_{xx}^{-1} (x - \mu_x)))\right) \end{split}$$

因为 $f(x,y) = p(x,y) = p(x)p(y|x) = f_x(x)p(y|x)$, 有

$$f_x(x) = (2\pi)^{-rac{1}{2}} |\Sigma_{xx}|^{-rac{1}{2}} \exp\left(-rac{1}{2}(x-\mu_x)^T \Sigma_{xx}^{-1}(x-\mu_x)
ight)$$

$$p(y|x) =$$

$$(2\pi)^{-\frac{n-1}{2}}|\Theta_{yy}|^{-\frac{1}{2}}\exp\left(-\frac{1}{2}(y-(\mu_y+\Sigma_{yx}\Sigma_{xx}^{-1}(x-\mu_x)))^T\Theta_{yy}(y-(\mu_y+\Sigma_{yx}\Sigma_{xx}^{-1}(x-\mu_x)))\right)$$

可以看出, x 的概率密度函数 $f_x(x)$ 依然满足正态分布

所以
$$X \sim N(\mu_x, \Sigma_{xx})$$