## 第八周 条件分布与条件期望

## 8.4 全期望公式(下)

例 8.4.1 随机变量 X 的密度函数为阶梯形函数,  $f(x) = \begin{cases} 2/3, & 0 \le x < 1 \\ 1/3, & 1 \le x < 2, \\ 0, & x \ge 2 \end{cases}$ 

设定事件 $A = \{X$ 落入区间 $[0,1)\}$ , $B = \{X$ 落入区间 $[1,2)\}$ ,试利用全期望公式计算期望E(X)和方差Var(X)。

解: 先计算事件 A、B 发生的概率, 得到

$$P(A) = \int_0^1 f(x) dx = \int_0^1 \frac{2}{3} dx = \frac{2}{3}, \qquad P(B) = \int_1^2 f(x) dx = \int_1^2 \frac{1}{3} dx = \frac{1}{3}$$

$$f(x|A) = \begin{cases} 1, & 0 \le x < 1 \\ 0, & \text{其他} \end{cases}$$
,  $f(x|B) = \begin{cases} 1, & 1 \le x < 2 \\ 0, & \text{其他} \end{cases}$ 

$$E(X|A) = \int_0^1 x \cdot f(x|A) dx = \int_0^1 x \cdot 1 dx = \frac{1}{2},$$

$$E(X|B) = \int_1^2 x \cdot f(x|B) dx = \int_1^2 x \cdot 1 dx = \frac{3}{2}$$

$$E(X^2|A) = \int_0^1 x^2 \cdot f(x|A) dx = \int_0^1 x^2 \cdot 1 dx = \frac{1}{3}$$
,

$$E(X^{2}|B) = \int_{1}^{2} x^{2} \cdot f(x|B) dx = \int_{1}^{2} x^{2} \cdot 1 dx = \frac{7}{3}$$

$$E(X) = P(A)E(X|A) + P(B)E(X|B) = \frac{1}{3}\frac{1}{2} + \frac{2}{3}\frac{3}{2} = \frac{7}{6}$$

$$E(X^{2}) = P(A)E(X^{2}|A) + P(B)E(X^{2}|B) = \frac{1}{3}\frac{1}{3} + \frac{2}{3}\frac{7}{3} = \frac{5}{3}$$

$$Var(X) = E(X^{2}) - E(X) = \frac{5}{3} - \left(\frac{7}{6}\right)^{2} = \frac{11}{36}$$

例 8.4.2 
$$(X,Y) \sim N(0,0,\sigma_1^2,\sigma_2^2,\rho)$$
, 证明:  $E(XY) = \rho \sigma_1 \sigma_2$ .

证明: 
$$E(XY) = E[E(XY|X)] = E[X \cdot E(Y|X)]$$

$$f(y|x) = \frac{f(x,y)}{f_X(x)} = \frac{\frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left\{\frac{-1}{2(1-\rho^2)} \left[\frac{x^2}{\sigma_1^2} - 2\rho\frac{xy}{\sigma_1\sigma_2} + \frac{y^2}{\sigma_2^2}\right]\right\}}{\frac{1}{\sqrt{2\pi}\sigma_1} \exp\left\{-\frac{x^2}{2\sigma_1^2}\right\}}$$

$$= \frac{1}{\sqrt{2\pi}\sigma_2\sqrt{1-\rho^2}} \exp\left\{-\frac{1}{2\sigma_2^2(1-\rho^2)} \left(y-\rho\frac{\sigma_2}{\sigma_1}x\right)^2\right\}$$

$$E(Y|X=x) = \rho \frac{\sigma_2}{\sigma_1} x \implies E(Y|X) = \rho \frac{\sigma_2}{\sigma_1} X \implies$$

$$E(XY) = E[E(XY \mid X)] = E[X \cdot E(Y \mid X)] = E(\rho \frac{\sigma_2}{\sigma_1} X^2) = \rho \frac{\sigma_2}{\sigma_1} \sigma_1^2 = \rho \sigma_1 \sigma_2.$$

例 8.4.3 (随机多个独立随机变量的和的期望) 假设某医生每天门诊挂号的病人数为N,是服从参数为a的泊松分布随机变量。又假设每位病人门诊看病的时间也为随机的,均服从参数为b的指数分布随机变量,且相互独立。这名医生总的门诊看病时间记为T,求E(T)和Var(T)

解: 设N 位病人的看病时间分别为 $X_1, X_2, \cdots, X_N$ ,  $X_k \sim Exp(b)(k=1,2,\cdots,N)$ 则 $T = X_1 + X_2 + \cdots + X_N$ , 利用全期望公式,

$$E(T) = E(E(T|N)) = E\left(E\left(\sum_{k=1}^{N} X_{k} \middle| N\right)\right) = \sum_{n=1}^{\infty} P(N=n) \cdot E\left(\sum_{k=1}^{N} X_{k} \middle| N=n\right)$$
$$= \sum_{n=1}^{\infty} P(N=n) \cdot E\left(\sum_{k=1}^{n} X_{k}\right) = \sum_{n=1}^{\infty} P(N=n) \cdot n \cdot E(X_{1})$$

$$= E(X_{1}) \sum_{n=1}^{\infty} n \cdot P(N = n) = E(X_{1}) E(N)$$

$$E(T^{2}) = E(E(T^{2}|N)) = \sum_{n=1}^{\infty} P(N = n) \cdot E(\sum_{k=1}^{N} X_{k})^{2} | N = n$$

$$= \sum_{n=1}^{\infty} P(N = n) \cdot E(\sum_{k=1}^{n} X_{k})^{2} = \sum_{n=1}^{\infty} P(N = n) \cdot E(\sum_{k=1}^{n} X_{k}^{2} + \sum_{1 \le i, j \le n, i \ne j} X_{i} X_{j})$$

$$= \sum_{n=1}^{\infty} P(N = n) \cdot (n \cdot E(X_{1}^{2}) + n \cdot (n-1) \cdot E(X_{1})^{2})$$

$$= \sum_{n=1}^{\infty} P(N = n) \cdot (n \cdot E(X_{1}^{2}) - n \cdot E(X_{1})^{2} + n^{2} \cdot E(X_{1})^{2})$$

$$= (E(X_{1}^{2}) - E(X_{1})^{2}) \cdot \sum_{n=1}^{\infty} n \cdot P(N = n) + E(X_{1})^{2} \cdot \sum_{n=1}^{\infty} n^{2} \cdot P(N = n)$$

$$= Var(X_{1}) \cdot E(N) + E(X_{1})^{2} \cdot E(N^{2})$$

因为N服从参数为a的泊松分布,X,服从参数为b的指数分布,所以

$$E(N) = Var(N) = a$$
,  $E(N^2) = a^2 + a$ ,  $E(X_1) = \frac{1}{b}$ ,  $Var(X_1) = \frac{1}{b^2}$ 

$$E(T) = E(X_1)E(N) = \frac{a}{b},$$

$$E(T^2) = Var(X_1) \cdot E(N) + E(X_1)^2 \cdot E(N^2) = \frac{a^2 + 2a}{b^2}, \quad Var(T) = E(T^2) - E(T)^2 = \frac{2a}{b^2}$$

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