## 概率统计(A)课程作业: 离散变量的数字特征

周书予

2000013060@stu.pku.edu.cn

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(1)

$$\begin{cases} 1 = \sum_{n \geqslant 0} \mathbb{P}(X = n) = \sum_{n \geqslant 0} A \cdot \frac{B^n}{n!} = A e^B \\ a = \mathbb{E}[X] = \sum_{n \geqslant 1} n \mathbb{P}(X = n) = \sum_{n \geqslant 1} A \cdot \frac{B^n}{(n-1)!} = AB \sum_{n \geqslant 0} \frac{B^n}{n!} = AB e^B \end{cases} \Rightarrow \begin{cases} A = e^{-a} \\ B = a \end{cases}$$

(2)

 $\Leftrightarrow \frac{a}{1+a} = p.$ 

$$\mathbb{E}[X] = \sum_{n \geqslant 0} n \mathbb{P}(X = n) = \frac{1}{1+a} \sum_{n \geqslant 0} n p^n = \frac{1}{1+a} \cdot \frac{p}{(1-p)^2} = a$$

$$\mathbb{E}[X^2] = \sum_{n \geqslant 0} n^2 \mathbb{P}(X = n) = \frac{1}{1+a} \sum_{n \geqslant 0} n^2 p^n = \frac{1}{1+a} \cdot \frac{p(1+p)}{(1-p)^3} = a(1+2a)$$

$$\text{Var}[X] = \mathbb{E}[X^2] - \mathbb{E}[X]^2 = a + a^2$$

 $\mathbf{2}$ 

$$\begin{split} \mathbb{E}\left[\min\{X,Y\}\right] &= \sum_{n\geqslant 0} n \mathbb{P}\left(\min\{X,Y\} = n\right) \\ &\leqslant \sum_{n\geqslant 0} n (\mathbb{P}\left(X = n\right) \mathbb{P}\left(Y \geqslant n\right) + \mathbb{P}\left(Y = n\right) \mathbb{P}\left(X \geqslant n\right)) \\ &\leqslant \sum_{n\geqslant 0} n (\mathbb{P}\left(X = n\right) + \mathbb{P}\left(Y = n\right)) \\ &= \mathbb{E}\left[X\right] + \mathbb{E}\left[Y\right] \\ &= \mathbb{E}\left[X + Y\right] \end{split}$$

由于 X,Y > 0, 故后者收敛能推出前者也收敛.

$$\begin{split} \mathbb{E}\left[\min\{X,Y\}\right] &= \sum_{n\geqslant 0} n \mathbb{P}\left(\min\{X,Y\} = n\right) \\ &= \sum_{n\geqslant 1} \mathbb{P}\left(\min\{X,Y\} \geqslant n\right) \\ &= \sum_{n\geqslant 1} \mathbb{P}\left(X \geqslant n\right) \mathbb{P}\left(Y \geqslant n\right) \end{split}$$

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(1)

$$\int_{-\infty}^{+\infty} |x| f_X(x) \mathrm{d}x = 2 \int_{0}^{+\infty} \frac{x}{\pi(x^2 + 1)} \mathrm{d}x = +\infty$$

广义积分  $\int_{-\infty}^{+\infty} x f_X(x) dx$  不绝对收敛, 因此  $\mathbb{E}[X]$  不存在.

(2)

$$\int_{-\infty}^{+\infty} |\arctan(x)| f_X(x) dx = 2 \int_0^{+\infty} \frac{\arctan(x)}{\pi (x^2 + 1)} dx$$

$$= \frac{2}{\pi} \int_0^{+\infty} \arctan(x) d \arctan(x)$$

$$= \frac{\arctan^2(x)}{\pi} \Big|_0^{+\infty}$$

$$= \frac{\pi}{4}$$

广义积分  $\int_{-\infty}^{+\infty}\arctan(x)f_X(x)\mathrm{d}x$  绝对收敛,因此  $\mathbb{E}\left[\arctan X\right]$  存在. 容易观察到积分函数是奇函数,因此有  $\mathbb{E}\left[\arctan X\right]=0$ .

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(1)

$$\mathbb{E}\left[e^{aX}\right] = \int_{-\infty}^{+\infty} e^{ax} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = e^{\frac{a^2}{2}} \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-a)^2}{2}} dx = e^{\frac{a^2}{2}}$$

(2)

$$\mathbb{E}\left[|X|\right] = 2\int_0^{+\infty} x \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{x}} dx = \frac{\sqrt{2}}{\sqrt{\pi}} \int_0^{+\infty} e^{-\frac{x^2}{2}} d\left(\frac{x^2}{2}\right) = -\frac{\sqrt{2}}{\sqrt{\pi}} e^{-u} \Big|_0^{+\infty} = \frac{\sqrt{2}}{\sqrt{\pi}} e^{-u} \Big|_0^{+\infty}$$

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(1)

$$\mathbb{E}[X_N] = \sum_{k=1}^{N} \mathbb{P}(X_N \ge k) = \sum_{k=1}^{N} \left(\frac{N-k+1}{N}\right)^n = \frac{1}{N^n} \sum_{k=1}^{N} k^n$$

(2)

$$\lim_{N \to \infty} \mathbb{E}\left[X_N/N\right] = \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^N \left(\frac{k}{N}\right)^n = \int_0^1 x^n dx = \frac{1}{n+1}$$

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记X表示需要支付的金额,x表示乙猜测的号码.

(1)

$$\mathbb{E}[X] = \frac{1}{15} \left[ (x-1)^2 + 2(x-2)^2 + 3(x-3)^2 + 4(x-4)^2 + 5(x-5)^2 \right] = x^2 - \frac{22}{3}x + 15$$
 当  $x$  取  $\frac{11}{3}$  时上式取到最小值  $\frac{14}{9}$ .

(2)

$$\mathbb{E}[X] = \frac{1}{15}[|x-1|+2|x-2|+3|x-3|+4|x-4|+5|x-5|]$$

当 x 取 4 时上式取到最小值 1.

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(0)

主要需要验证非负性与归一性.

当  $\max\{|x|,|y|\} \geqslant \pi$  时, g(x)g(y) = 0, 故  $f_{X,Y}(x,y) = \phi(x)\phi(y) > 0$ . 当  $|x|,|y| < \pi$  时,  $\phi(x)\phi(y) > \frac{\mathrm{e}^{-\pi^2}}{2\pi} > \frac{\mathrm{e}^{-\pi^2}}{2\pi}g(x)g(y)$ , 从而也有  $f_{X,Y}(x,y) = \phi(x)\phi(y) > 0$ .

易知  $\iint f_{X,Y}(x,y) dxdy = 1$ , 而  $\int_{\mathcal{X}} g(x) dx = 0$ , 因此  $\iint g(x)g(y) dxdy = 0$ , 故归一性满足.

(1)

$$f_X(x) = \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dy = \phi(x) \int_{-\infty}^{+\infty} \phi(y) dy + \frac{e^{-\pi^2}}{2\pi} g(x) \int_{-\infty}^{+\infty} g(y) dy = \phi(x)$$

故 X 的边缘分布是正态分布. Y 显然是对称的, 故也是正态分布.

(2)

$$\rho_{X,Y} = \frac{\operatorname{Cov}(X,Y)}{\sqrt{\operatorname{Var}[X]\operatorname{Var}[Y]}} = \mathbb{E}\left[(X - \mathbb{E}[X])(Y - \mathbb{E}[Y])\right] = \mathbb{E}\left[XY\right]$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} xy \left(\phi(x)\phi(y) + \frac{\mathrm{e}^{-\pi^2}}{2}g(x)g(y)\right) \mathrm{d}x\mathrm{d}y$$

$$= \left[\int_{-\infty}^{+\infty} x\phi(x)\mathrm{d}x\right] \left[\int_{-\infty}^{+\infty} y\phi(y)\mathrm{d}y\right] + \frac{\mathrm{e}^{-\pi^2}}{2} \left[\int_{-\infty}^{+\infty} xg(x)\mathrm{d}x\right] \left[\int_{-\infty}^{+\infty} yg(y)\mathrm{d}y\right]$$

$$= 0$$

但  $f_{X,Y}(x,y) = f_X(x)f_Y(y)$  不恒成立, 因此 X,Y 不线性相关但不独立.

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(1)

$$\rho_{X_{1},X_{1}-X_{2}} = \frac{\operatorname{Cov}\left(X_{1},X_{1}-X_{2}\right)}{\sqrt{\operatorname{Var}\left[X_{1}\right]\operatorname{Var}\left[X_{1}-X_{2}\right]}} = \frac{\operatorname{Var}\left[X_{1}\right]-\operatorname{Cov}\left(X_{1},X_{2}\right)}{\sqrt{\operatorname{Var}\left[X_{1}\right]\left(\operatorname{Var}\left[X_{1}\right]+\operatorname{Var}\left[X_{2}\right]-\operatorname{Cov}\left(X_{1},X_{2}\right)\right)}} = \frac{1-\rho}{\sqrt{2-\rho}}$$

(2)

记  $Z = X_1 - X_2$ . 考虑 Z 的密度函数.

$$f_{Z}(z) = \int_{-\infty}^{+\infty} f_{X_{1},X_{2}}(x,x-z) dx$$

$$= \int_{-\infty}^{+\infty} \frac{1}{2\pi\sqrt{1-\rho^{2}}} \exp\left[-\frac{1}{2(1-\rho^{2})} \left(x^{2} - 2\rho x(x-z) + (x-z)^{2}\right)\right] dx$$

$$= \frac{1}{2\pi\sqrt{1-\rho^{2}}} \int_{-\infty}^{+\infty} \exp\left[-\frac{1}{2(1-\rho^{2})} \left(2(1-\rho)\left(x-\frac{z}{2}\right)^{2} + \frac{z^{2}(1+p)}{2}\right)\right] dx$$

$$= \frac{1}{2\pi\sqrt{1-\rho^{2}}} \exp\left(-\frac{z^{2}}{4(1-\rho)}\right) \int_{-\infty}^{+\infty} \exp\left(-\frac{\left(x-\frac{z}{2}\right)^{2}}{1+\rho}\right) dx$$

$$= \frac{1}{2\sqrt{\pi(1-\rho)}} \exp\left(-\frac{z^{2}}{4(1-\rho)}\right)$$

于是

$$\mathbb{E}\left[|Z|\right] = 2 \int_0^{+\infty} z f_Z(z) dz$$

$$= \frac{1}{\sqrt{\pi(1-\rho)}} \int_0^{+\infty} z \exp\left(-\frac{z^2}{4(1-\rho)}\right) dz$$

$$= \frac{2\sqrt{1-\rho}}{\sqrt{\pi}} \int_0^{+\infty} \exp\left(-\frac{z^2}{4(1-\rho)}\right) d\frac{z^2}{4(1-\rho)}$$

$$= \frac{2\sqrt{1-\rho}}{\sqrt{\pi}}$$

(3)

考虑  $\mathbb{E}[\max\{X_1, X_2\}] = \mathbb{E}[X_2] + \mathbb{E}[\max\{0, X_1 - X_2\}],$  显然  $\mathbb{E}[X_2] = 0$ , 而

$$\mathbb{E}\left[\max\{0, X_1 - X_2\}\right] = \int_0^{+\infty} z f_Z(z) dz = \frac{\sqrt{1 - \rho}}{\sqrt{\pi}}$$

因此  $\mathbb{E}\left[\max\{X_1, X_2\}\right] = \frac{\sqrt{1-\rho}}{\sqrt{\pi}}.$ 

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用  $X_i, Y_i$   $(1 \le i \le n)$  分别表示第 i 次摸球时是否摸到了白球/黑球, 则显然有  $X = \sum_{i=1}^n X_i, Y = \sum_{i=1}^n Y_i$  由于  $\forall i, \mathbb{E}[X_i] = p, \mathbb{E}[Y_i] = q$ , 故  $\mathbb{E}[X] = np, \mathbb{E}[Y] = nq$ .

简单分析可知 
$$\mathbb{E}[X_iY_j] = \begin{cases} 0, & i=j \\ pq, & i \neq j \end{cases}$$
,  $\mathbb{E}[X_iX_j] = \begin{cases} p, & i=j \\ p^2, & i \neq j \end{cases}$ ,  $\mathbb{E}[Y_iY_j] = \begin{cases} q, & i=j \\ q^2, & i \neq j \end{cases}$ .

(1)

$$Cov (X, Y) = \mathbb{E} [XY] - \mathbb{E} [X] \mathbb{E} [Y]$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{n} \mathbb{E} [X_i Y_j] - \mathbb{E} [X] \mathbb{E} [Y]$$

$$= n(n-1)pq - n^2 pq$$

$$= -npq$$

(2)

$$\begin{aligned} & \text{Var}\left[X\right] = \mathbb{E}\left[X^{2}\right] - \mathbb{E}\left[X\right]^{2} = \sum_{i=1}^{n} \sum_{j=1}^{n} \mathbb{E}\left[X_{i}X_{j}\right] - \mathbb{E}\left[X\right]^{2} \\ & = np + (n^{2} - n)p^{2} - (np)^{2} = np(1 - p) \\ & \text{Var}\left[Y\right] = nq(1 - q) \\ & \rho_{X,Y} = \frac{\text{Cov}\left(X,Y\right)}{\sqrt{\text{Var}\left[X\right] \text{Var}\left[Y\right]}} = \frac{-npq}{\sqrt{n^{2}p(1 - p)q(1q)}} = -\sqrt{\frac{pq}{(1 - p)(1 - q)}} \end{aligned}$$

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(1)

 $\mathbb{E}[X] = \mathbb{E}[X_1 + X_2 + \cdots + X_n]$  即为从 N 个球中抽出总计 n 个, 其数码和的期望. 因此有

$$\mathbb{E}[X] = \sum_{k} k \binom{a}{k} \binom{b}{n-k} / \binom{N}{n}$$

$$= \sum_{k} a \binom{a-1}{k-1} \binom{b}{n-k} / \binom{N}{n}$$

$$= a \binom{N-1}{n-1} / \binom{N}{n}$$

$$= \frac{an}{N}$$

(2)(3)

不难验证  $\mathbb{E}[X_i] = \frac{a}{N}$  (因为 **(1)** 的结论对所有 n 均成立) 以及  $\mathbb{E}[X_i X_j] = \begin{cases} \frac{a}{N}, & i = j \\ \frac{a(a-1)}{N(N-1)}, & i \neq j \end{cases}$ 

$$\operatorname{Var}[X] = \mathbb{E}[X^{2}] - \mathbb{E}[X]^{2} = \sum_{i=1}^{n} \sum_{j=1}^{n} \mathbb{E}[X_{i}X_{j}] - \mathbb{E}[X]^{2}$$
$$= n\frac{a}{N} + (n^{2} - n)\frac{a(a-1)}{N(N-1)} - \frac{n^{2}a^{2}}{N^{2}}$$
$$= \frac{na(N-n)(N-a)}{N^{2}(N-1)}$$

$$\operatorname{Cov}\left(X_{i}, X_{j}\right) = \mathbb{E}\left[X_{i} X_{j}\right] - \mathbb{E}\left[X_{i}\right] \mathbb{E}\left[X_{j}\right] = \begin{cases} \frac{a(N-a)}{N^{2}}, & i = j\\ \frac{-a(N-a)}{N^{2}(N-1)}, & i \neq j \end{cases}$$

特别的, 当 n = N 时, 有 Var[X] = 0.