

Homework 5

姓名: 方嘉聪 学号: 2200017849

Problem 1. 定义二维随机变量 X, Y . 证明: $\text{Corr}(X, Y) = \pm 1$ 当且仅当存在实数 $a \neq 0, b$, 使得 $\mathbb{P}(Y = aX + b) = 1$.

提示: 利用结论 (无需证明), 若随机变量 Z 满足 $\text{Var}(Z) = 0$, 则 $\mathbb{P}(Z = \mathbb{E}(Z)) = 1$. ◀

Solution. 分别证明两个方向. 记 X, Y 的均值分别为 μ_X, μ_Y , 标准差分别为 σ_X, σ_Y .

(1) 若 $\text{Corr}(X, Y) = \pm 1$. 这里我们证明 $\text{Corr}(X, Y) = 1$ 的情况, $\text{Corr}(X, Y) = -1$ 的情况同理. 由定义知

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} = 1. \implies \text{Cov}(X, Y) = \sigma_X \sigma_Y.$$

考虑 $a = \sigma_Y / \sigma_X \neq 0$, 我们来计算 $\text{Var}(Y - aX)$.

$$\text{Var}(Y - aX) = \sigma_Y^2 + a^2 \sigma_X^2 - 2a \text{Cov}(X, Y) = a^2 \sigma_X^2 - 2a \sigma_X \sigma_Y + \sigma_Y^2 = 0.$$

由结论知 $\mathbb{P}(Y - aX = \mathbb{E}(Y - aX)) = 1$. 令 $b = \mathbb{E}(Y - aX)$, 则 $\mathbb{P}(Y = aX + b) = 1$. 证毕.

注: 对于 $\text{Corr}(X, Y) = -1$ 的情况, 可以取 $a = -\sigma_Y / \sigma_X, b = \mathbb{E}(Y - aX)$.

(2) 若存在实数 $a \neq 0, b$, 使得 $\mathbb{P}(Y = aX + b) = 1$. 我们先来证明, 对于任意随机变量 Z 有

$$\mathbb{P}(Z = \mathbb{E}(Z)) = 1 \implies \text{Var}(Z) = 0.$$

证明: 注意到

$$\mathbb{P}(Z = \mathbb{E}(Z)) = 1 \implies \mathbb{P}((Z - \mathbb{E}(Z))^2 = 0) = 1 \implies \text{Var}(Z) = \mathbb{E}((Z - \mathbb{E}(Z))^2) = 0.$$

回到本题, 由于 $\mathbb{P}(Y = aX + b) = 1$, 那么 $\text{Var}(Y - aX) = 0$. 故

$$0 = \text{Var}(Y - aX) = \sigma_Y^2 + a^2 \sigma_X^2 - 2a \text{Cov}(X, Y)$$

那么 (利用均值不等式)

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} = \frac{1}{2a} \frac{\sigma_Y}{\sigma_X} + \frac{a}{2} \frac{\sigma_X}{\sigma_Y} \implies |\text{Corr}(X, Y)| \geq 1.$$

由于 $|\text{Corr}(X, Y)| \leq 1$, 那么 $\text{Corr}(X, Y) = \pm 1$. 证毕.

综上所述, $\text{Corr}(X, Y) = \pm 1$ 当且仅当存在实数 $a \neq 0, b$, 使得 $\mathbb{P}(Y = aX + b) = 1$. ◁

Problem 2. 对于 $\sigma_1 > 0, \sigma_2 > 0, -1 < \rho < 1$, 二维随机变量 $(U, V) \sim \mathcal{N}(0, 0, \sigma_1^2, \sigma_2^2, \rho)$. 本题中我们将计算 $\mathbb{E}(\text{ReLU}(U) \cdot \text{ReLU}(V))$. 其中 $\text{ReLU}(x) = \max\{0, x\}$.

设二维随机变量 $(X, Y) \sim \mathcal{N}(0, 0, 1, 1, \rho)$, 令二维随机变量 (R, Θ) 满足 $R \geq 0, \Theta \in [0, 2\pi]$, 且

$$\begin{cases} X = R \cdot (\sqrt{1 - \rho^2} \cdot \cos \Theta + \rho \cdot \sin \Theta) = R \cdot \sin(\arccos \rho + \Theta) \\ Y = R \sin \Theta \end{cases} \quad (1)$$

- (1) 令 $x = r \cdot (\sqrt{1-\rho^2} \cdot \cos \theta + \rho \cdot \sin \theta)$, $y = r \sin \theta$. 验证 $x^2 + y^2 - 2\rho xy = r^2(1-\rho^2)$.
 (2) 计算 R, Θ 的联合密度函数, R 和 Θ 的各自边际密度函数, 并判断 R 和 Θ 的独立性.
 (3) 计算 $\mathbb{E}(\text{ReLU}(X) \cdot \text{ReLU}(Y))$. 提示: 利用结论 (无需证明)

$$\int_0^{+\infty} x^3 e^{-x^2/2} dx = 2,$$

$$\int_0^{\pi - \arccos \rho} (\rho \cdot \sin^2 \theta + \sqrt{1-\rho^2} \sin \theta \cos \theta) d\theta = \frac{1}{2} (\rho(\pi - \arccos \rho) + \sqrt{1-\rho^2}).$$

- (4) 验证 $(\sigma_1 X, \sigma_2 Y)$ 与 (U, V) 具有相同的分布.
 (5) 计算 $\mathbb{E}(\text{ReLU}(U) \cdot \text{ReLU}(V))$.

Solution. (1) 直接带入验证即可.

$$\begin{aligned} x^2 + y^2 - 2\rho xy &= r^2 (\sqrt{1-\rho^2} \cos \theta + \rho \sin \theta)^2 + r^2 \sin^2 \theta - 2\rho r^2 (\sqrt{1-\rho^2} \cos \theta + \rho \sin \theta) \sin \theta \\ &= r^2(1-\rho^2) \cos^2 \theta + r^2 \rho^2 \sin^2 \theta + r^2 \sin^2 \theta - 2\rho r^2 \sin \theta \cos \theta \\ &= r^2(1-\rho^2). \end{aligned}$$

(2) 我们先来计算

$$\left| \frac{\partial(X, Y)}{\partial(R, \Theta)} \right| = \left| \begin{pmatrix} \sqrt{1-\rho^2} \cos \Theta + \rho \sin \Theta & R(-\sqrt{1-\rho^2} \sin \Theta + \rho \cos \Theta) \\ \sin \Theta & R \cos \Theta \end{pmatrix} \right| = R\sqrt{1-\rho^2}.$$

那么 R, Θ 的联合密度函数为

$$\begin{aligned} f_{R, \Theta}(r, \theta) &= f_{X, Y}(x(r, \theta), y(r, \theta)) \cdot \left| \frac{\partial(X, Y)}{\partial(R, \Theta)} \right| \\ &= \frac{1}{2\pi\sqrt{1-\rho^2}} \exp\left(-\frac{1}{2(1-\rho^2)}(x^2 + y^2 - 2\rho xy)\right) \cdot r\sqrt{1-\rho^2} \\ &= \frac{1}{2\pi} r \exp\left(-\frac{r^2}{2}\right). \quad \text{利用 (1) 中结论} \end{aligned}$$

那么 R 的边际密度函数为

$$f_R(r) = \int_0^{2\pi} f_{R, \Theta}(r, \theta) d\theta = r e^{-r^2/2} \int_0^{2\pi} \frac{1}{2\pi} d\theta = r e^{-r^2/2}.$$

而 Θ 的边际密度函数为

$$f_{\Theta}(\theta) = \int_0^{+\infty} f_{R, \Theta}(r, \theta) dr = \frac{1}{2\pi} \int_0^{+\infty} r e^{-r^2/2} dr = \frac{1}{2\pi}.$$

由于 $f_{R, \Theta}(r, \theta) = f_R(r)f_{\Theta}(\theta)$, 那么 R 和 Θ 是独立的.

(3) 注意到

$$\begin{aligned} \mathbb{E}(\text{ReLU}(X) \cdot \text{ReLU}(Y)) &= \iint_{x \geq 0 \wedge y \geq 0} xy \cdot f_{X, Y}(x, y) dx dy \\ (\text{使用(1)换元}) &= \iint_{\mathbf{D}} xy \cdot f_{X, Y}(x(r, \theta), y(r, \theta)) \cdot \left| \frac{\partial(X, Y)}{\partial(R, \Theta)} \right| dr d\theta \\ &= \iint_{\mathbf{D}} f_{R, \Theta}(r, \theta) \cdot r^2 (\sqrt{1-\rho^2} \sin \theta \cos \theta + \rho \sin^2 \theta) dr d\theta \end{aligned}$$

其中区域

$$\begin{aligned}\mathbf{D} &= \{(r, \theta) : r \geq 0, \sin \theta \geq 0, \sqrt{1 - \rho^2} \cos \theta + \rho \sin \theta \geq 0\} \\ &= \{(r, \theta) : r \geq 0, \theta \in [0, \pi - \arccos \rho]\}.\end{aligned}$$

那么我们有

$$\begin{aligned}\mathbb{E}(\text{ReLU}(X) \cdot \text{ReLU}(Y)) &= \frac{1}{2\pi} \int_0^{\pi - \arccos \rho} \int_0^{+\infty} r^3 e^{-r^2/2} \cdot (\sqrt{1 - \rho^2} \sin \theta \cos \theta + \rho \sin^2 \theta) \, dr \, d\theta \\ &= \frac{1}{2\pi} \int_0^{+\infty} r^3 e^{-r^2/2} \int_0^{\pi - \arccos \rho} (\rho \cdot \sin^2 \theta + \sqrt{1 - \rho^2} \sin \theta \cos \theta) \, d\theta \, dr \\ &= \frac{1}{2\pi} (\rho(\pi - \arccos \rho) + \sqrt{1 - \rho^2}).\end{aligned}$$

最后一个等式利用了提示中的结论.

(4) 记 $(W, T) = (\sigma_1 X, \sigma_2 Y)$, 那么

$$\begin{aligned}f_{W,T}(w, t) &= f_{X,Y}(x, y) \cdot \left| \frac{\partial(X, Y)}{\partial(W, T)} \right| \\ &= \frac{1}{\sigma_1 \sigma_2} \cdot \frac{1}{2\pi \sqrt{1 - \rho^2}} \exp \left(-\frac{1}{2(1 - \rho^2)} (x^2 + y^2 - 2\rho xy) \right) \\ &= \frac{1}{2\pi \sigma_1 \sigma_2 \sqrt{1 - \rho^2}} \exp \left\{ -\frac{1}{2(1 - \rho^2)} \left(\frac{x^2}{\sigma_1^2} + \frac{y^2}{\sigma_2^2} - \frac{2\rho xy}{\sigma_1 \sigma_2} \right) \right\}.\end{aligned}$$

故 $(\sigma_1 X, \sigma_2 Y) = (W, T) \sim \mathcal{N}(0, 0, \sigma_1^2, \sigma_2^2, \rho)$, 与 (U, V) 具有相同的分布.

(5) 由于 $(\sigma_1 X, \sigma_2 Y)$ 与 (U, V) 具有相同的分布, 那么

$$\begin{aligned}\mathbb{E}(\text{ReLU}(U) \cdot \text{ReLU}(V)) &= \mathbb{E}(\text{ReLU}(\sigma_1 X) \cdot \text{ReLU}(\sigma_2 Y)) = \sigma_1 \sigma_2 \mathbb{E}(\text{ReLU}(X) \cdot \text{ReLU}(Y)) \\ &= \frac{\sigma_1 \sigma_2}{2\pi} (\rho(\pi - \arccos \rho) + \sqrt{1 - \rho^2}).\end{aligned}$$

◁

Problem 3. 在课上我们考虑了如下矩阵 $A \in \mathbb{R}^{n \times n}$: 对于任意 $1 \leq i, j \leq n$, $A_{i,j} \sim \mathcal{N}(0, 1)$, 且不同元素相互独立. 计算 $\mathbb{E}(\text{tr}(A^3))$ 和 $\mathbb{E}(\text{tr}(A^4))$. 提示: 考虑 $n = 1$ 的情况, 并参考作业三第六题. ◀

Solution. 我们先来计算 $\mathbb{E}(A_{i,j}^3)$ 与 $\mathbb{E}(A_{i,j}^4)$. 有对称性知 $\mathbb{E}(A_{i,j}^3) = 0$. 而由作业三第六题知 $X \sim \mathcal{N}(0, 1) \implies X^2 \sim \chi^2(1)$. 那么 $\mathbb{E}(A_{i,i}^2) = 1, \text{Var}(A_{i,i}^2) = 2$. 由此知

$$\mathbb{E}(A_{i,j}^4) = \mathbb{E}(A_{i,j}^2)^2 + \text{Var}(A_{i,j}^2) = 3.$$

记 $A_{(i,j)}^{(t)}$ 表示 A^t 的第 (i, j) 个元素. 那么我们有

$$A_{i,j}^{(2)} = \sum_{k=1}^n A_{i,k} A_{k,j} \implies A_{i,i}^{(3)} = \sum_{j=1}^n A_{i,j}^{(2)} A_{j,i} = \sum_{j=1}^n \sum_{k=1}^n A_{i,k} A_{k,j} A_{j,i}.$$

那么

$$\begin{aligned}\mathbb{E}(A_{i,i}^{(3)}) &= \mathbb{E} \left(\sum_{j=1}^n \sum_{k=1}^n A_{i,k} A_{k,j} A_{j,i} \right) = \sum_{j=1}^n \sum_{k=1}^n \mathbb{E}(A_{i,k} A_{k,j} A_{j,i}) = \mathbb{E}(A_{i,i}^3) = 0. \\ \mathbb{E}(\text{tr}(A^3)) &= \mathbb{E} \left(\sum_{i=1}^n A_{i,i}^{(3)} \right) = \sum_{i=1}^n \mathbb{E}(A_{i,i}^{(3)}) = 0.\end{aligned}$$

类似地, 我们有

$$A_{i,j}^{(4)} = \sum_{k=1}^n A_{i,k} A_{k,j}^{(3)} \implies A_{i,i}^{(4)} = \sum_{j=1}^n \sum_{k=1}^n \sum_{t=1}^n A_{i,k} A_{k,j} A_{j,t} A_{t,i}.$$

那么

$$\begin{aligned} \mathbb{E}(A_{i,i}^{(4)}) &= \mathbb{E} \left(\sum_{j=1}^n \sum_{k=1}^n \sum_{t=1}^n A_{i,k} A_{k,j} A_{j,t} A_{t,i} \right) = \sum_{j=1}^n \sum_{k=1}^n \sum_{t=1}^n \mathbb{E}(A_{i,k} A_{k,j} A_{j,t} A_{t,i}) \\ &= \sum_{j=1}^n \mathbb{E}(A_{i,j} A_{j,j} A_{j,j} A_{j,i}) = \mathbb{E}(A_{i,i}^4) + \sum_{j \neq i} \mathbb{E}(A_{i,j}^2) \mathbb{E}(A_{j,j}^2) = n + 2 \end{aligned}$$

故

$$\mathbb{E}(\text{tr}(A^4)) = \mathbb{E} \left(\sum_{i=1}^n A_{i,i}^{(4)} \right) = \sum_{i=1}^n \mathbb{E}(A_{i,i}^{(4)}) = n^2 + 2n.$$

综上, $\mathbb{E}(\text{tr}(A^3)) = 0, \mathbb{E}(\text{tr}(A^4)) = 3n$. ◁

Problem 4. 回答下列问题:

- (1) 令 X_1, X_2, \dots, X_n 为独立同分布随机变量, 且 $X_i \sim \mathcal{N}(0, 1)$. 令 $Y = \sum_{i=1}^n X_i^2$. 对于任意实数 $t \in [0, 1/4)$, 证明

$$\mathbb{E}(e^{t(Y-n)}) \leq e^{2t^2 n}.$$

提示: 首先考虑 $n = 1$ 的情况, 并参考作业三第六题, 以及作业一第三题的提示.

- (2) 对于任意 $0 \leq \Delta < 1$, 证明

$$\mathbb{P}(Y \geq (1 + \Delta)n) \leq e^{-n\Delta^2/8}.$$

提示: 根据 $0 \leq \Delta < 1$, 选择合适的 t 使得 $t \in [0, 1/4)$, 并利用马尔可夫不等式.

- (3) 对于任意 $0 \leq \Delta < 1$, 证明

$$\mathbb{P}(Y \leq (1 - \Delta)n) \leq e^{-n\Delta^2/8}.$$

Solution. (1) 在作业三第六题中我们已经计算了, 对于 $X \sim \mathcal{N}(0, 1)$, 有

$$\mathbb{E}(e^{tX^2}) = \frac{1}{\sqrt{1-2t}}, \quad \text{for } t \in (-\infty, 1/2).$$

注意到 $\{X_i\}_{i=1}^n \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, 1)$, 那么

$$\mathbb{E}(e^{t(Y-n)}) = e^{-nt} \mathbb{E}(e^{t \sum_{i=1}^n X_i^2}) = e^{-nt} \prod_{i=1}^n \mathbb{E}(e^{tX_i^2}) = e^{-nt} \frac{1}{(1-2t)^{n/2}}$$

对于 $t \in (-1/4, 1/4)$, 等价于需证明

$$\begin{aligned} \frac{e^{-nt}}{(1-2t)^{n/2}} \leq e^{2t^2n} &\iff (1-2t)^{n/2} \geq e^{-2t^2n - tn} \\ &\iff \ln(1-2t) \geq -(2t)^2 - 2t \end{aligned}$$

令 $h(t) = \ln(1-2t) + 4t^2 + 2t$, 而

$$h'(t) = \frac{4t(1-4t)}{1-2t} \implies h(t) \text{ 在 } (-1/4, 0) \text{ 上单调递减, 在 } (0, 1/4) \text{ 上单调递增.}$$

那么 $h(t) \geq h(0) = 0$, 证毕.

(2) 取 $t = \Delta/4 \in [0, 1/4)$, 那么

$$\begin{aligned} \mathbb{P}(Y \geq (1+\Delta)n) &= \mathbb{P}\left(e^{tY} \geq e^{t(1+\Delta)n}\right) = \mathbb{P}\left(e^{t(Y-n)} \geq e^{t\Delta n}\right) \\ (\text{Markov's inequality}) &\leq \frac{\mathbb{E}[e^{t(Y-n)}]}{e^{tn\Delta}} \leq \frac{e^{2t^2n}}{e^{tn\Delta}} \\ (t = \Delta/4) &= e^{-n\Delta^2/8}. \end{aligned}$$

证毕.

(3) 类似 (2) 中的证明, 取 $t = \Delta/4 \in [0, 1/4)$, 那么:

$$\begin{aligned} \mathbb{P}(Y \leq (1-\Delta)n) &= \mathbb{P}\left(e^{tY} \leq e^{t(1-\Delta)n}\right) = \mathbb{P}\left(e^{-t(Y-n)} \geq e^{t\Delta n}\right) \\ (\text{Markov's inequality}) &\leq \frac{\mathbb{E}[e^{-t(Y-n)}]}{e^{tn\Delta}} \\ (-t = -\frac{\Delta}{4} \in (-\frac{1}{4}, 0], \text{ Use (1)}) &\leq \frac{e^{2t^2n}}{e^{tn\Delta}} = e^{-n\Delta^2/8}. \end{aligned}$$

证毕.

◁