

Assignment 5

12310520

芮煜涵

T1. X and Y are independent

$$\begin{aligned}
 P(Z=1) &= \int_0^\infty \int_0^y \lambda e^{-\lambda x} \mu e^{-\mu y} dx dy \\
 &= \int_0^\infty (-e^{-\lambda x}) \Big|_0^y \mu e^{-\mu y} dy \\
 &= \int_0^\infty (1 - e^{-\lambda y}) \mu e^{-\mu y} dy \\
 &= \int_0^\infty \mu e^{-\mu y} dy - \int_0^\infty \mu e^{-(\lambda+\mu)y} dy \\
 &= (\mu - e^{-\mu y}) \Big|_0^{+\infty} - \left(-\frac{\mu}{\lambda+\mu} e^{-(\lambda+\mu)y} \right) \Big|_0^{+\infty} \\
 &= 1 - \frac{\mu}{\lambda+\mu} = \frac{\lambda}{\lambda+\mu}
 \end{aligned}$$

$$P(Z=0) = 1 - P(Z=1) = \frac{\mu}{\lambda+\mu}$$

$$\Rightarrow f(z) = \begin{cases} \frac{\lambda}{\lambda+\mu}, & z=1 \\ \frac{\mu}{\lambda+\mu}, & z=0 \end{cases}$$

T2. X, Y are independent, identically distributed r.v.s

$$P(Z=1) = P(X=1) \cdot P(Y=1) = p^2$$

$$\begin{aligned}
 P(Z=2) &= P(X=1)P(Y=2) + P(X=2)P(Y=1) + P(X=2)(Y=2) \\
 &= (1-p)p^2 \cdot 2 + (1-p)^2 p^2
 \end{aligned}$$

$$\begin{aligned}
 P(Z=z) &= 2 \sum_{i=1}^{z-1} P(X=i) P(Y=z) + P(X=z) P(Y=z) \\
 &= 2(1-p)^{z-1} p \cdot p (1 + (1-p) + \dots + (1-p)^{z-2}) + (1-p)^{2z-2} p^2 \\
 &= 2(1-p)^{z-1} p^2 \frac{1 - (1-p)^{z-1}}{p} + (1-p)^{2z-2} p^2
 \end{aligned}$$

$$= 2p(1-p)^{8-1} - 2p(1-p)^{28-2} + (1-p)^{28-2} p^2$$

$$\Rightarrow P(\Sigma = \zeta) = 2p(1-p)^{8-1} - 2p(1-p)^{28-2} + p^2(1-p)^{28-2} \quad \zeta=1,2,\dots$$

T3. Let X, Y, Σ be the lifespan of three components.

$$f_X(t) = f_Y(t) = f_\Sigma(t) = \lambda e^{-\lambda t} \quad t > 0$$

$$\text{let } A = \min(X, Y), T = \min(A, \Sigma)$$

$$F_A(t) = 1 - (1 - F_X(t))(1 - F_Y(t)) = 1 - e^{-2\lambda t} \quad t > 0$$

$$F_T(t) = 1 - (1 - F_A(t))(1 - F_\Sigma(t)) = 1 - e^{-2\lambda t} e^{-\lambda t} \\ = 1 - e^{-3\lambda t} \quad t > 0$$

$$\Rightarrow f_T(t) = F'_T(t) = 3\lambda e^{-3\lambda t} \quad t > 0$$

T4. $\begin{pmatrix} X \\ Y \end{pmatrix} \sim N(\mu, \Sigma) \quad \mu = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \Sigma = \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}$

$$\text{let } \Sigma = X - Y$$

$$\Sigma \sim N(\mu_X - \mu_Y, 6X^2 - 2\rho 6X 6Y + 6Y^2)$$

$$\Rightarrow \Sigma \sim N(0, 2 - 2\rho)$$

$$f_\Sigma(\Sigma) = \frac{1}{\sqrt{2\pi(2-2\rho)}} e^{-\frac{\Sigma^2}{4-4\rho}} = \frac{1}{2\sqrt{\pi(1-\rho)}} e^{-\frac{\Sigma^2}{4-4\rho}}$$

$$(1) \text{ Cov}(X - Y, XY) = \text{Cov}(X, XY) - \text{Cov}(Y, XY)$$

$$= (E(X^2Y) - E(X)E(XY)) - (E(Y^2X) - E(Y)E(XY))$$

$$= E(X^2Y) - E(Y^2X)$$

X, Y are independent, identically distributed r.v.s

$$\Rightarrow E(X^2Y) = E(Y^2X)$$

$$\Rightarrow \text{Cov}(X-Y, XY) = 0$$

$$\rho_{x-y, XY} = \frac{\text{Cov}(X-Y, XY)}{\sqrt{\text{Var}(X-Y)\text{Var}(XY)}} = 0$$

T5. (1) $f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx$

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

$$= \frac{1}{2\pi b_x b_y \sqrt{1-p^2}} \cdot \exp\left(-\frac{y^2}{2b_y^2}\right) \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2(1-p^2)} \left(\frac{x}{b_x} - \frac{py}{b_y}\right)^2\right) dx$$

let $t = \frac{1}{\sqrt{1-p^2}} \left(\frac{x}{b_x} - \frac{py}{b_y}\right)$ $dx = dt \sqrt{1-p^2} b_x$

$$= \frac{1}{2\pi b_x b_y \sqrt{1-p^2}} \exp\left(-\frac{y^2}{2b_y^2}\right) \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2} t^2\right) \sqrt{1-p^2} b_x dt$$

$$= \frac{1}{2\pi b_y} \exp\left(-\frac{y^2}{2b_y^2}\right) \int_{-\infty}^{\infty} e^{-\frac{1}{2} t^2} dt$$

$$= \frac{1}{2\pi b_y} e^{-\frac{y^2}{2b_y^2}} \sqrt{2\pi} = \frac{1}{\sqrt{2\pi} b_y} e^{-\frac{y^2}{2b_y^2}}$$

(2) $f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = \frac{\frac{1}{2\pi b_x b_y \sqrt{1-p^2}} \exp\left(-\frac{1}{2(1-p^2)} \left[\frac{x^2}{b_x^2} - 2p \frac{xy}{b_x b_y} + \frac{y^2}{b_y^2}\right]\right)}{\frac{1}{\sqrt{2\pi} b_y} e^{-\frac{y^2}{2b_y^2}}}$

$$= \frac{1}{\sqrt{2\pi} \sqrt{1-p^2} b_x} \exp\left(-\frac{1}{2(1-p^2)b_x^2} \left[x - \frac{pb_x y}{b_y}\right]^2\right)$$

$$= \frac{1}{\sqrt{2\pi} \sqrt{1-p^2} 6x} \exp \left(-\frac{1}{2(1-p^2)} 6x^2 \left[x - \left(\frac{p_6x}{6Y} Y \right)^2 \right] \right)$$

$$\mu = \frac{p_6x}{6Y} Y \quad 6^2 = (1-p^2) 6x^2$$

$$Tb. \quad \begin{pmatrix} X \\ Y \end{pmatrix} \sim N(\mu, \Sigma) \quad \mu = \begin{pmatrix} 4500 \\ 5500 \end{pmatrix}$$

$$P_{6x, 6Y} = 0.65 \times 1500 \times 2000 = 1950000 \quad \Sigma = \begin{pmatrix} 1500^2 & 1950000 \\ 1950000 & 2000^2 \end{pmatrix}$$

$$X|Y=y \sim N\left(4500 + \frac{0.65 \times 1500}{2000} (6800 - 5500), (1 - 0.65^2) \times 1500^2\right)$$

$$\sim N(5133.75, 1299375)$$

$$P(X < 6800 | Y = 6800) = P\left(\frac{X - 5133.75}{\sqrt{1299375}} < \frac{6800 - 5133.75}{\sqrt{1299375}}\right)$$

$$= \Phi(1.46) = 0.9279$$

$$(2) \quad X - Y \mid X + Y = 12000 \quad A = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$$

$$A \begin{pmatrix} X \\ Y \end{pmatrix} \sim N(A\mu, A\Sigma A^T)$$

$$\sim N\left(\begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 4500 \\ 5500 \end{pmatrix}, \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1500^2 & 1950000 \\ 1950000 & 2000^2 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}\right)$$

$$\sim N\left(\begin{pmatrix} -1000 \\ 10000 \end{pmatrix}, \begin{pmatrix} 235 \times 10^4 & -175 \times 10^4 \\ -175 \times 10^4 & 1015 \times 10^4 \end{pmatrix}\right)$$

$$\text{let } M = X - Y, \quad N = X + Y \quad \mu_M = -1000, \quad \mu_N = 10000$$

$$6M = \sqrt{235} \times 100 \quad 6N = \sqrt{1015} \times 100 \quad \rho = \frac{-175}{\sqrt{235 \times 1015}}$$

$$M|N = 12000 \sim N\left(-1000 + \frac{p_6 M}{6N} \times (12000 - 10000), (1 - p^2) 6M^2\right)$$

$$\sim N(-1344.83, 2048275)$$

$$P(M > 0 | N=12000) = 1 - P(M \leq 0) = 1 - P\left(\frac{M + 1344.83}{\sqrt{2048275}} \leq \frac{1344.83}{\sqrt{2048275}}\right)$$

$$= 1 - \underline{\Phi}(0.94) = 1 - 0.8264 = 0.1736$$

T7. $X \sim \text{Poisson}(\lambda)$ then $P(X=k) = \frac{\lambda^k e^{-\lambda}}{k!}$

$$\mu = \lambda \quad \text{Var}(X) = \lambda \quad \text{approx}$$

\Rightarrow when $\lambda \rightarrow +\infty$ $X \sim N(\lambda, \lambda)$

$$P(X=k) = \frac{1}{\sqrt{2\pi\lambda}} e^{-\frac{(k-\lambda)^2}{2\lambda}}$$

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.stats import poisson, norm
4
5 # 定义不同的 λ 值
6 lambda_values = [5, 10, 20, 50]
7
8 # 创建图形
9 plt.figure(figsize=(12, 10))
10
11 for i, lam in enumerate(lambda_values):
12     # 泊松分布
13     x_poisson = np.arange(0, lam + 3 * np.sqrt(lam)) # 范围从 0 到 λ + 3σ
14     y_poisson = poisson.pmf(x_poisson, lam)
15
16     # 正态分布近似
17     x_normal = np.linspace(0, lam + 3 * np.sqrt(lam), 1000)
18     y_normal = norm.pdf(x_normal, lam, np.sqrt(lam))
19
20     # 创建子图
21     plt.subplot(2, 2, i + 1)
22     plt.bar(x_poisson, y_poisson, width=0.5, color='yellow', alpha=0.6, label=f'Poisson λ={lam}')
23     plt.plot(x_normal, y_normal, 'b-', label=f'Normal Approx. λ={lam}')
24     plt.title(f'Poisson and Normal Approximation (λ={lam})')
25     plt.xlabel('x')
26     plt.ylabel('Probability')
27     plt.legend()
28
29 # 显示图形
30 plt.tight_layout()
31 plt.show()

```

