

HW 8.

$$\begin{aligned}
 (1). \quad EX &= \int_0^{+\infty} x \cdot f(x) dx \\
 &= \int_0^{+\infty} \frac{x^2}{\sigma^2} \exp\left\{-\frac{x^2}{2\sigma^2}\right\} dx \\
 &= \int_0^{+\infty} -x \cdot d e^{-\frac{x^2}{2\sigma^2}} \\
 &= -x \cdot e^{-\frac{x^2}{2\sigma^2}} \Big|_0^{+\infty} + \int_0^{+\infty} e^{-\frac{x^2}{2\sigma^2}} dx \\
 &= 0 + \frac{\sqrt{2\pi}\sigma}{2} = \frac{\sqrt{2\pi}\sigma}{2}
 \end{aligned}$$

$$\therefore \text{Var } X = EX^2 - (EX)^2 = 2\sigma^2 - \frac{\pi\sigma^2}{2} = \left(2 - \frac{\pi}{2}\right)\sigma^2$$

$$\begin{aligned}
 EX^2 &= \int_0^{+\infty} x^2 f(x) dx \\
 &= \int_0^{+\infty} \frac{x^3}{\sigma^2} \exp\left\{-\frac{x^2}{2\sigma^2}\right\} dx \\
 &= \int_0^{+\infty} -x^2 d e^{-\frac{x^2}{2\sigma^2}} \\
 &= -x^2 e^{-\frac{x^2}{2\sigma^2}} \Big|_0^{+\infty} + \int_0^{+\infty} 2x e^{-\frac{x^2}{2\sigma^2}} dx \\
 &= -2\sigma^2 e^{-\frac{x^2}{2\sigma^2}} \Big|_0^{+\infty} = 2\sigma^2
 \end{aligned}$$

$$\begin{aligned}
 (2) \quad EX &= \int_0^1 \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^\alpha (1-x)^{\beta-1} dx \\
 &= \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} B(\alpha+1, \beta) \\
 &= \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \frac{\Gamma(\alpha+1)\Gamma(\beta)}{\Gamma(\alpha+\beta+1)} = \frac{\alpha}{\alpha+\beta}
 \end{aligned}$$

$$\begin{aligned}
 \text{同理. } EX^2 &= \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \cdot \frac{\Gamma(\alpha+2)\Gamma(\beta)}{\Gamma(\alpha+\beta+2)} \\
 &= \frac{\alpha(\alpha+1)}{(\alpha+\beta)(\alpha+\beta+1)}
 \end{aligned}$$

$$\therefore \text{Var } X = EX^2 - (EX)^2 = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$$

注意. $\Gamma(n) = (n-1)!$

$$\Gamma(x) = \int_0^{+\infty} t^{x-1} e^{-t} dt$$

$$\Gamma(x+1) = x\Gamma(x).$$

$$B(\alpha, \beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx$$

$$B(x, y) = \frac{\Gamma(x)\Gamma(y)}{\Gamma(x+y)}.$$

$$\begin{aligned}
 B) \quad EX &= \int_0^{+\infty} \frac{kx}{\lambda} \left(\frac{x}{\lambda}\right)^{k-1} \exp\left\{-\left(\frac{x}{\lambda}\right)^k\right\} dx \\
 &= \int_0^{+\infty} x \cdot \frac{kx^{k-1}}{\lambda^k} \exp\left\{-\left(\frac{x}{\lambda}\right)^k\right\} dx \\
 &= \lambda \int_0^{+\infty} \frac{x}{\lambda} \exp\left\{-\left(\frac{x}{\lambda}\right)^k\right\} d\left(\frac{x}{\lambda}\right)^k \\
 y &= \left(\frac{x}{\lambda}\right)^k \\
 &= \lambda \int_0^{+\infty} y^{\frac{1}{k}} e^{-y} dy \\
 &= \lambda \Gamma\left(\frac{1}{k}+1\right) = \frac{\lambda}{k} \Gamma\left(\frac{1}{k}\right).
 \end{aligned}$$

$$\begin{aligned}
 EX^2 &= \int_0^{+\infty} x^2 \frac{kx^{k-1}}{\lambda^k} \exp\left\{-\left(\frac{x}{\lambda}\right)^k\right\} dx \\
 &= \lambda^2 \int_0^{+\infty} \left(\frac{x}{\lambda}\right)^2 \exp\left\{-\left(\frac{x}{\lambda}\right)^k\right\} d\left(\frac{x}{\lambda}\right)^k \\
 y &= \left(\frac{x}{\lambda}\right)^k \\
 &= \lambda^2 \int_0^{+\infty} y^{\frac{2}{k}} e^{-y} dy \\
 &= \lambda^2 \Gamma\left(\frac{2}{k}+1\right) = \frac{\lambda^2}{k} \Gamma\left(\frac{2}{k}\right)
 \end{aligned}$$

$$\text{Var } X = EX^2 - (EX)^2 = \frac{\lambda^2}{k} \Gamma\left(\frac{2}{k}\right) - \frac{\lambda^2}{k^2} \Gamma\left(\frac{1}{k}\right)^2$$

14. 求 pdf.

法1. 公式.

$$Y = \ln X \sim N(\mu, \sigma^2).$$

$$\therefore X = e^Y \triangleq g(Y) \quad Y = h(X) = \ln X.$$

$$h'(x) = \frac{1}{x}$$

$$\therefore p_X(x) = p_Y(h(x)) |h'(x)|$$

$$= \frac{1}{\sqrt{2\pi}\sigma x} \exp\left\{-\frac{(\ln x - \mu)^2}{2\sigma^2}\right\} \quad x > 0$$

$$EX = \int_0^{+\infty} x \cdot p(x) dx$$

$$= \int_0^{+\infty} \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(\ln x - \mu)^2}{2\sigma^2}\right\} dx$$

$$\text{令 } y = \frac{\ln x - \mu}{\sigma}, \quad x = e^{\sigma y + \mu}, \quad dx = \sigma e^{\sigma y + \mu} dy$$

$$\therefore EX = \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{+\infty} \sigma e^{\sigma y + \mu} \cdot e^{-\frac{y^2}{2}} dy$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{y^2 - 2\sigma y + \sigma^2}{2} + \frac{\sigma^2}{2} + \mu\right\} dy$$

$$= \left(\int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{(y-\sigma)^2}{2}\right\} dy \right) \exp\left\{\frac{\sigma^2}{2} + \mu\right\}$$

$$= e^{\frac{\sigma^2}{2} + \mu} \quad \text{正态密度 } N(\sigma, 1)$$

$$EX^2 = \int_0^{+\infty} x^2 p(x) dx$$

$$= \int_0^{+\infty} \frac{x}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(\ln x - \mu)^2}{2\sigma^2}\right\} dx$$

$$= \int_0^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left\{\sigma y + \mu + \sigma y + \mu - \frac{y^2}{2}\right\} dy$$

$$= \int_0^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{y^2 - 4\sigma y + 4\sigma^2}{2} + 2\sigma^2 + 2\mu\right\} dy$$

$$= \exp\{2\sigma^2 + 2\mu\} \int_0^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{(y-2\sigma)^2}{2}\right\} dy$$

$$= e^{2\sigma^2 + 2\mu}$$

$$\therefore \text{Var} X = EX^2 - (EX)^2 = e^{2\sigma^2 + 2\mu} - e^{\sigma^2 + 2\mu} = e^{\sigma^2 + 2\mu} (e^{\sigma^2} - 1).$$

18. (1) $0 < Y < 2$. $P(Y \leq y) = P(Y \leq y | X=1)P(X=1) + P(Y \leq y | X=2)P(X=2)$.

① $0 < y \leq 1$

$$P(Y \leq y) = y \cdot \frac{1}{2} + \frac{y}{2} \cdot \frac{1}{2} = \frac{3}{4}y$$

② $1 < y \leq 2$

$$P(Y \leq y) = 1 \cdot \frac{1}{2} + \frac{y}{2} \cdot \frac{1}{2} = \frac{y}{4} + \frac{1}{2}$$

$$F(y) = \begin{cases} 0 & y \leq 0 \\ \frac{3}{4}y & 0 < y \leq 1 \\ \frac{y}{4} + \frac{1}{2} & 1 < y \leq 2 \\ 1 & y > 2. \end{cases}$$

易错: $F(y) = \begin{cases} y & x=1 \\ \frac{y}{2} & x=2. \end{cases} \quad \times$

X 是随机变量.

法1. $p(y) = \begin{cases} \frac{3}{4} & 0 < y \leq 1 \\ \frac{1}{4} & 1 < y \leq 2 \\ 0 & \text{else.} \end{cases}$

$$EY = \int_{-\infty}^{+\infty} y \cdot p(y) dy = \int_0^1 \frac{3}{4}y dy + \int_1^2 \frac{1}{4}y dy = \frac{3}{8}y^2 \Big|_0^1 + \frac{y^2}{8} \Big|_1^2 = \frac{3}{4}.$$

法2. $EY = E[E(Y|X)]$

条件期望. $E[Y|X=1] = \frac{1}{2} \quad E[Y|X=2] = 1$

$$EY = E[Y|X=1]P(X=1) + E[Y|X=2]P(X=2) = \frac{3}{4}.$$

$$21. 1) X \sim \text{Poi}(\lambda) \Rightarrow P(X=k) = \frac{\lambda^k}{k!} e^{-\lambda}.$$

$$Y \sim \text{Poi}(\mu).$$

由 Poisson 分布再生性, $X+Y \sim \text{Poi}(\lambda+\mu).$

$$\begin{aligned} P(X=k|X+Y=m) &= \frac{P(X=k, X+Y=m)}{P(X+Y=m)} = \frac{P(X=k)P(Y=m-k)}{P(X+Y=m)} \\ &= \frac{\frac{\lambda^k}{k!} e^{-\lambda} \cdot \frac{\mu^{m-k}}{(m-k)!} e^{-\mu}}{\frac{(\lambda+\mu)^m}{m!} e^{-(\lambda+\mu)}} = \frac{m!}{k!(m-k)!} \frac{\lambda^k \mu^{m-k}}{(\lambda+\mu)^m} \\ &= \binom{m}{k} \left(\frac{\lambda}{\lambda+\mu}\right)^k \left(\frac{\mu}{\lambda+\mu}\right)^{m-k} \sim B(m, \frac{\lambda}{\lambda+\mu}) \end{aligned}$$

$$\therefore E[X|X+Y=m] = m \cdot \frac{\lambda}{\lambda+\mu} = \frac{\lambda m}{\lambda+\mu}.$$

硬算.

$$E[X|X+Y=m] = \sum_{k=0}^m k \cdot P(X=k|X+Y=m) = \sum_{k=1}^m k P(X=k|X+Y=m)$$

$$\begin{aligned} &= \sum_{k=1}^m k \cdot \frac{m!}{k!(m-k)!} \frac{\lambda^k \mu^{m-k}}{(\lambda+\mu)^m} = \frac{\lambda}{(\lambda+\mu)^m} \sum_{k=1}^m \frac{m(m-1)!}{(k-1)!(m-k)!} \lambda^{k-1} \mu^{m-k} \\ &= \frac{\lambda m}{(\lambda+\mu)^m} \sum_{k=0}^{m-1} \frac{(m-1)!}{k!(m-1-k)!} \lambda^k \mu^{m-1-k} = \frac{\lambda m}{(\lambda+\mu)^m} (\lambda+\mu)^{m-1} = \frac{\lambda m}{\lambda+\mu} \end{aligned}$$

$$2) X, Y \sim B(n, p)$$

$$X+Y \sim B(2n, p)$$

$$\begin{aligned} \therefore P(X=k|X+Y=m) &= \frac{P(X=k)P(Y=m-k)}{P(X+Y=m)} \\ &= \frac{\binom{n}{k} p^k (1-p)^{n-k} \binom{n}{m-k} p^{m-k} (1-p)^{n-m+k}}{\binom{2n}{m} p^m (1-p)^{2n-m}} = \frac{\binom{n}{k} \binom{n}{m-k}}{\binom{2n}{m}} \end{aligned}$$

$\therefore X|X+Y=m \sim H(m, n, 2n)$ 超几何分布 Hypergeometric Distribution.

$$E[X|X+Y=m] = m \cdot \frac{n}{2n} = \frac{m}{2}$$

硬算. $E[X|X+Y=m] = \sum_{k=0}^m k \cdot \frac{\binom{n}{k} \binom{n}{m-k}}{\binom{2n}{m}} = \frac{1}{\binom{2n}{m}} \sum_{k=1}^m k \cdot \frac{n!}{k!(n-k)!} \frac{n!}{(m-k)!(n-m+k)!}$

$$= \frac{n}{\binom{2n}{m}} \sum_{k=1}^m \frac{(n-1)!}{(k-1)!(n-k)!} \frac{n!}{(m-k)!(n-m+k)!} \stackrel{i=k-1}{=} \frac{n}{\binom{2n}{m}} \sum_{i=0}^{m-1} \frac{(n-1)!}{i!(n-i-1)!} \frac{n!}{(m-i-1)!(n-m+i+1)!}$$

$$= \frac{n}{\binom{2n}{m}} \sum_{i=0}^{m-1} \binom{n-1}{i} \binom{n}{m-i-1} = \frac{n}{\binom{2n}{m}} \binom{2n-1}{m-1} = \frac{m}{2}$$

注. 有再生性的分布

① $B(n, p)$

② $\text{Poi}(\lambda)$

③ $N(\mu, \sigma^2)$.

22.

$$1) P(Y \leq y) = \sum_{k=0}^2 P(Y \leq y | X=k) P(X=k)$$

$$= \frac{1}{3} (\Phi(y) + \Phi(y-1) + \Phi(y-2))$$

$$\therefore F(y) = \frac{1}{3} [\Phi(y) + \Phi(y-1) + \Phi(y-2)]$$

$$f_Y(y) = \frac{1}{3\sqrt{2\pi}} (e^{-\frac{y^2}{2}} + e^{-\frac{(y-1)^2}{2}} + e^{-\frac{(y-2)^2}{2}})$$

$$EY = \int_{-\infty}^{+\infty} y \cdot f_Y(y) dy = \frac{1}{3} \left[\int_{-\infty}^{+\infty} y \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy + \int_{-\infty}^{+\infty} y \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-1)^2}{2}} dy + \int_{-\infty}^{+\infty} y \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-2)^2}{2}} dy \right]$$

$$= \frac{1}{3} (0 + 1 + 2) = 1$$

或条件期望 $EY = E[E(Y|X)]$

$$12) F_{X+Y}(z) = P(X+Y \leq z) = \sum_{k=0}^2 P(X+Y \leq z | X=k) P(X=k)$$

$$\star = \frac{1}{3} \sum_{k=0}^2 P(Y \leq z-k | X=k) \neq \frac{1}{3} \sum_{k=0}^2 P(Y \leq z-k). \quad X \text{ 与 } Y \text{ 不独立}$$

$$= \frac{1}{3} [P(Y \leq z | X=0) + P(\underbrace{Y \leq z-1}_{Y|X \sim N(1,1)} | X=1) + P(Y \leq z-2 | X=2)]$$

$$= \frac{1}{3} [\Phi(z) + \Phi(z-1) + \Phi(z-2)] = \frac{1}{3} [\Phi(z) + \Phi(z-2) + \Phi(z-4)]$$

$$13) \text{cov}(X, Y) = EXY - EX \cdot EY$$

$$EXY = E[E(XY|X)]$$

$$= E[XY | X=0] P(X=0) + E[XY | X=1] P(X=1) + E[XY | X=2] P(X=2)$$

$$= 0 \cdot \frac{1}{3} + 1 \cdot \frac{1}{3} + 2 \cdot 2 \cdot \frac{1}{3} = \frac{5}{3}$$

$$EX=1 \quad EY=1$$

$$\therefore \text{cov}(X, Y) = \frac{2}{3}$$

$$25. X_1, X_2 \text{ iid} \sim \text{Exp}(\lambda) \quad \lambda=2.$$

$$f_1(x) = f_2(x) = \begin{cases} \lambda e^{-\lambda x} & x > 0, \\ 0 & x \leq 0 \end{cases}$$

$$F_1(x) = F_2(x) = 1 - e^{-\lambda x} \quad x > 0.$$

$$\text{记 } Y_1 = \min\{X_1, X_2\} \quad Y_2 = \max\{X_1, X_2\}$$

$$F_{Y_1}(z) = P(\min\{X_1, X_2\} \leq z)$$

$$= 1 - P(\min\{X_1, X_2\} > z)$$

$$= 1 - e^{-\lambda z} \cdot e^{-\lambda z}$$

$$= 1 - e^{-2\lambda z}$$

$$f_{Y_1}(z) = 2\lambda e^{-2\lambda z} \sim \text{Exp}(2\lambda).$$

$$\therefore EY_1 = \frac{1}{2\lambda} = \frac{1}{4}$$

$$\underline{\min\{x_1, x_2\} + \max\{x_1, x_2\} = x_1 + x_2.}$$

$$\begin{aligned}\therefore E\max\{x_1, x_2\} &= EX_1 + EX_2 - EY_1 \\ &= \frac{1}{2} + \frac{1}{2} - \frac{1}{4} = \frac{3}{4}\end{aligned}$$

或求 $\max\{x_1, x_2\}$ cdf \rightarrow pdf.

$$\begin{aligned}P(Y_2 \leq z) &= P(X_1 \leq z) \cdot P(X_2 \leq z) \\ &= (1 - e^{-\lambda z})^2 \\ &= e^{-2\lambda z} - 2e^{-\lambda z} + 1 = F_{Y_2}(z)\end{aligned}$$

$$f_{Y_2}(z) = -2\lambda e^{-2\lambda z} + 2\lambda e^{-\lambda z}$$

$$EY_2 = \int_0^{+\infty} x(-2\lambda e^{-2\lambda x} + 2\lambda e^{-\lambda x}) dx$$

$$= 2 \int_0^{+\infty} x \cdot \lambda e^{-\lambda x} dx - \int_0^{+\infty} x \cdot 2\lambda e^{-2\lambda x} dx$$

$$= 2 \cdot \frac{1}{\lambda} - \frac{1}{2\lambda} = 1 - \frac{1}{4} = \frac{3}{4}.$$

方法 2.

$$\begin{aligned}\underline{\min\{x_1, x_2\} = \frac{x_1 + x_2}{2} - \frac{|x_1 - x_2|}{2}} \\ \underline{\max\{x_1, x_2\} = \frac{x_1 + x_2}{2} + \frac{|x_1 - x_2|}{2}}\end{aligned}$$

