

10-31 作业

14. 令 $Y = \ln X$, 所以对 $\forall x > 0$,

$$P(X \leq x) = P(\ln X \leq \ln x) = P(Y \leq \ln x) = \Phi\left(\frac{\ln x - \mu}{\sigma}\right),$$

所以

$$p(x) = \begin{cases} \frac{1}{\sqrt{2\pi}\sigma x} e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}}, & x > 0, \\ 0, & \text{其他.} \end{cases}$$

$$\begin{aligned} EX &= \int_0^\infty x \frac{1}{\sqrt{2\pi}\sigma x} e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}} dx = \int e^{t\sigma + \mu} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \\ &= e^{\frac{\mu^2}{2} + \mu} \int \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2 - 2t\sigma + \sigma^2}{2}} dt = e^{\frac{\mu^2}{2} + \mu} \\ EX^2 &= \int_0^\infty x^2 \frac{1}{\sqrt{2\pi}\sigma x} e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}} dx = \int_{-\infty}^\infty e^{2(t\sigma) + \mu} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \\ &= e^{2\mu + 2\sigma^2} \int_{-\infty}^\infty \frac{1}{\sqrt{2\pi}} e^{-\frac{(t - 2\sigma)^2}{2}} dt = e^{2\mu + 2\sigma^2} \\ \Rightarrow \quad Var(X) &= EX^2 - (EX)^2 = e^{\sigma^2 + 2\mu} (e^{\sigma^2} - 1) \end{aligned}$$

17. 已知 X 的密度函数, 有

$$\begin{aligned} E[\min\{|X|, 1\}] &= \int_{|x|>1} f(x) dx + \int_{|x|\leq 1} |x| \cdot f(x) dx \\ &= 2 \int_1^\infty \frac{1}{\pi(1+x^2)} dx + 2 \int_0^1 \frac{x}{\pi(1+x^2)} dx \quad (\text{对称性}) \\ &= \frac{2 \arctan x}{\pi} \Big|_1^\infty + \int_0^1 \frac{d(1+x^2)}{\pi(1+x^2)} = \frac{1}{2} + \frac{1}{\pi} \ln(1+x^2) \Big|_0^1 \\ &= \frac{1}{2} + \frac{\ln 2}{\pi} \end{aligned}$$

19. $(X, Y) \sim N(1, 1, 0.5, 0.5, 0.5)$, 则 (X, Y) 的联合密度为

$$f(x, y) = \frac{2}{\sqrt{3}\pi} \exp \left\{ -\frac{2}{3} [2(x-1)^2 - 2(x-1)(y-1) + 2(y-1)^2] \right\}.$$

(1) $Z = |X - Y|$, 可以先求 $X - Y$ 的密度. 令 $\begin{cases} S = X - Y \\ T = Y \end{cases}$ 则 $\begin{cases} X = S + T \\ Y = T \end{cases}$,

对应的雅可比行列式为

$$J = \begin{vmatrix} 1 & 1 \\ 0 & 1 \end{vmatrix} = 1.$$

所以 (S, T) 的联合密度为

$$\begin{aligned} f_{ST}(s, t) &= f(s+t, t) \cdot |J| \\ &= \frac{2}{\sqrt{3}\pi} \exp \left\{ -\frac{2}{3} [2(s+t-1)^2 - 2(s+t-1)(t-1) + 2(t-1)^2] \right\} \\ &= \frac{2}{\sqrt{3}\pi} \exp \left\{ -\frac{4}{3}s^2 - \frac{4}{3}s(t-1) - \frac{4}{3}(t-1)^2 \right\}, -\infty < s, t < \infty. \end{aligned}$$

则 U 的边际密度为

$$\begin{aligned} f_S(s) &= \int_{-\infty}^{\infty} \frac{2}{\sqrt{3}\pi} \exp \left\{ -\frac{4}{3}s^2 - \frac{4}{3}s(t-1) - \frac{4}{3}(t-1)^2 \right\} dt \\ &= \int_{-\infty}^{\infty} \frac{2}{\sqrt{3}\pi} \exp \left\{ -\frac{4}{3} \left[\left(t + \frac{s-2}{2} \right)^2 + \frac{3}{4}s^2 \right] \right\} dt \\ &= \frac{1}{\sqrt{\pi}} e^{-s^2} \int_{-\infty}^{\infty} \frac{2}{\sqrt{3}\pi} \exp \left\{ -\frac{(t + \frac{s-2}{2})^2}{3/4} \right\} dt \\ &= \frac{1}{\sqrt{\pi}} e^{-s^2}, \quad -\infty < s < \infty \end{aligned}$$

对于 $Z = |S|$, $P(Z \leq z) = P(-z \leq S \leq z) = 2 \int_0^z \frac{1}{\sqrt{\pi}} e^{-s^2} ds$, ($z > 0$), 可得 Z 的密度函数为

$$f_Z(z) = \frac{2}{\sqrt{\pi}} e^{-z^2}, z > 0.$$

期望为

$$E(Z) = \int_0^{\infty} z \cdot \frac{2}{\sqrt{\pi}} e^{-z^2} dz = -\frac{1}{\sqrt{\pi}} \int_0^{\infty} e^{-z^2} d(-z^2) = \frac{1}{\sqrt{\pi}}.$$

(2) $U = \max(X, Y)$, $V = \min(X, Y)$, 则有 $\begin{cases} U + V = X + Y \\ U - V = |X - Y| \end{cases}$ 由期望的性质有:

$$\begin{cases} E(U + V) = E(X + Y) \\ E(U - V) = E|X - Y| \end{cases} \Rightarrow \begin{cases} E(U) + E(V) = E(X) + E(Y) = 2 \\ E(U) - E(V) = EZ = \frac{1}{\sqrt{\pi}} \end{cases}$$

有

$$\begin{cases} E(U) = 1 + \frac{1}{2\sqrt{\pi}} \\ E(V) = 1 - \frac{1}{2\sqrt{\pi}} \end{cases}$$

20. 由题意易知, 对于一球, 在未落入第 1 个盒子的条件下, 落入第 2 个盒子的概率为 $\frac{p_2}{1-p_1}$. 由此可得, 给定 X_1 时, $X_2 | X_1 = k \sim B\left(m-k, \frac{p_2}{1-p_1}\right)$, 所以

$$E(X_2 | X_1 = k) = (m-k) \frac{p_2}{1-p_1},$$

$$\text{Var}(X_2 | X_1 = k) = (m-k) \frac{p_2}{1-p_1} \left(1 - \frac{p_2}{1-p_1}\right)$$

(2) 由题意易知: $X_k \sim b(m, p_k)$, 所以

$$E(X_1 + X_2) = EX_1 + EX_2 = m(p_1 + p_2).$$

且可将 $X_1 + \dots + X_k$ 看做落入前 k 个盒子的总球数, 每个小球独立地以概率 p_k 落入第 k 个盒子, 则 $X_1 + \dots + X_k \sim b(m, p_1 + \dots + p_k)$, $k = 1, \dots, n$, 所以

$$\text{Var}(X_1 + \dots + X_k) = m \left(\sum_{i=1}^k p_i \right) \left(1 - \sum_{i=1}^k p_i \right)$$

27. (1) $X_1 \sim b(n, p)$, 所以

$$M_{X_1}(t) = E(e^{tX_1}) = \sum_{k=1}^n e^{tk} \binom{n}{k} p^k (1-p)^{n-k} = \sum_{k=1}^n \binom{n}{k} (pe^t)^k (1-p)^{n-k} = (pe^t + 1 - p)^n$$

(2) $X_2 \sim P(\lambda)$, 所以

$$M_{X_2}(t) = E(e^{tX_2}) = \sum_{k=0}^{\infty} e^{tk} \frac{\lambda^k}{k!} e^{-\lambda} = \sum_{k=0}^{\infty} \frac{(\lambda e^t)^k}{k!} e^{-\lambda} = e^{-\lambda} e^{\lambda e^t} = e^{\lambda(e^t - 1)}$$

(3) $X_3 \sim EXP(\lambda)$, 所以

$$M_{X_3}(t) = E(e^{tX_3}) = \int_0^{\infty} e^{tx_3} \lambda e^{-\lambda x_3} dx_3 = \int_0^{\infty} \lambda e^{-(\lambda-t)x_3} dx_3 = \frac{\lambda}{\lambda-t} (t < \lambda).$$

(4) $X_4 \sim N(\mu, \sigma^2)$, 所以

$$\begin{aligned} M_{X_4}(t) &= E(e^{tX_4}) = \int e^{tx_4} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x_4-\mu)^2}{2\sigma^2}} dx_4 = \int e^{ty\sigma+t\mu} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{y^2}{2}} \sigma dy \\ &= e^{\frac{t^2\sigma^2}{2}+t\mu} \int \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2-2ty\sigma+t^2\sigma^2}{2}} dy \\ &= e^{\frac{t^2\sigma^2}{2}+t\mu} \end{aligned}$$