

$$2.17 \quad (a) \int_0^m f(x)dx = m^3 = \frac{1}{2} \Rightarrow m = \left(\frac{1}{2}\right)^{\frac{1}{3}}$$

$$(b) f(-x) = f(x) \text{ 因此 } \int_{-\infty}^0 f(x)dx = \int_0^{\infty} f(x)dx = \frac{1}{2} \text{ 因此 } m = 0.$$

$$2.18 \quad E|x-a| = \int_{-\infty}^{\infty} |x-a|f(x)dx = \int_0^{\infty} (x-a)f(x)dx + \int_{-\infty}^0 (a-x)f(x)dx$$

$$\begin{aligned} E|x-a| - E|x-m| &= 2 \int_a^m xf(x)dx + \int_{-\infty}^a af(x)dx - \int_a^{\infty} af(x)dx \\ &= 2 \int_a^m xf(x)dx + a \left(1 - 2 \int_a^{\infty} f(x)dx \right) \\ &= 2 \int_a^m xf(x)dx + a \left(\int_m^{\infty} f(x)dx - 2 \int_a^{\infty} f(x)dx \right) \\ &= 2 \int_a^m (m-a)f(x)dx \quad \text{or} \quad 2 \int_m^a (a-m)f(x)dx \end{aligned}$$

$$\text{因为 } \int_a^m (m-a)f(x)dx \geq 0$$

$$\text{所以 } E|x-a| - E|x-m| \geq 0 \Rightarrow E|x-a| \geq E|x-m|$$

$$\text{即 } \min_a E|x-a| = E|x-m|.$$

$$2.27 \quad (a) f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

(b)

$$f(x) = \begin{cases} \frac{1}{\pi} \sin^2(x) & \text{if } x \in [-\pi, \pi] \\ 0 & \text{others} \end{cases}$$

(c) 假众数在 b 处且 $b \neq a$, 不妨假设 $a < b$, 由题意可知, 对 $a < x < b, f(a) < f(x) < f(b)$, 因为 a 是对称点, 所以存在点 $x' < a$ 满足 $a - x' = x - a$, 有 $f(x') = f(x)$ 此时, $f(x') > f(a)$ 且 $x' < a < b$ 与 b 是 $f(x)$ 的众数矛盾。故 $a = b$, 即 a 是 $f(x)$ 的众数。

(d) $x = 0$

$$\mathbf{3.13} \quad (\text{a}) P(X = x|\lambda) = \frac{e^{-\lambda}\lambda^x}{x!} \quad P(X = 0) = e^{-\lambda}$$

$$\text{所以} \quad P(X \geq 0) = 1 - e^{-\lambda}$$

$$P(X_T = x) = \frac{e^{-\lambda}}{1 - e^{-\lambda}} \frac{\lambda^x}{x!}$$

$$\text{所以} \quad EX_T = \sum_{x=1}^{\infty} xP(X_T = x) = \frac{\lambda e^{-\lambda}}{1 - e^{-\lambda}} \sum_{x=1}^{\infty} \frac{\lambda^{x-1}}{(x-1)!} = \frac{\lambda}{1 - e^{-\lambda}}$$

$$\begin{aligned} Var X_T &= EX_T^2 - (EX_T)^2 \\ &= \sum_{x=1}^{\infty} x^2 \frac{e^{-\lambda}\lambda^x}{(1 - e^{-\lambda})x!} - (EX_T)^2 \\ &= \frac{\lambda(\lambda + 1)}{1 - e^{-\lambda}} - \left(\frac{\lambda}{1 - e^{-\lambda}} \right)^2 \\ &= \frac{\lambda}{1 - e^{-\lambda}} \left(1 - \frac{\lambda e^{-\lambda}}{1 - e^{-\lambda}} \right) \end{aligned}$$

$$(\text{b}) \quad P(X = x) = \binom{r+x-1}{x} p^r (1-p)^x, \quad P(X > 0) = 1 - P(X = 0) = 1 - p^r$$

$$P(X_T = x) = \binom{r+x-1}{x} \frac{p^r}{1 - p^r} (1-p)^x$$

$$EX_T = \sum_{x=1}^{\infty} x \binom{r+x-1}{x} \frac{p^r}{1 - p^r} (1-p)^x = \frac{EX}{1 - p^r} = r \frac{1-p}{p(1-p^r)}$$

$$\begin{aligned} Var X_T &= EX_T^2 - (EX_T)^2 = \frac{EX^2}{1 - p^r} - (EX_T)^2 \\ &= \frac{Var X + (EX)^2}{1 - p^r} - (EX_T)^2 \\ &= \frac{r(1-p) + r^2(1-p)^2}{p^2(1-p^r)} - \left[\frac{r(1-p)}{p(1-p^r)} \right]^2 \end{aligned}$$

$$\mathbf{3.24} \quad (\text{a}) \quad f(x) = \frac{1}{\beta} e^{-x/\beta} \quad , \quad Y = X^{1/\gamma} = g(X) \quad , \quad g^{-1}(y) = y^\gamma$$

$$f_Y(y) = f_X(g^{-1}(y)) \frac{d}{dy} g^{-1}(y) = \frac{1}{\beta} e^{-y^\gamma/\beta} \gamma y^{\gamma-1}$$

$$\begin{aligned} EY &= \int_0^\infty \frac{\gamma}{\beta} y^\gamma e^{-y^\gamma/\beta} dy = \int_0^\infty y e^{-y^\gamma/\beta} d(y^\gamma/\beta) \\ &= \int_0^\infty (\beta t)^{1/\gamma} e^{-t} dt = \beta^{1/\gamma} \Gamma(1 + \frac{1}{\gamma}) \\ &= \frac{\beta^{1/\gamma}}{\gamma} \Gamma(\frac{1}{\gamma}) \end{aligned}$$

$$\begin{aligned} VarY &= EY^2 - (EY)^2 \\ &= \int_0^\infty y^2 e^{-y^\gamma/\beta} d(\frac{y^\gamma}{\beta}) - (EY)^2 \\ &= \int_0^\infty (\beta t)^{2/\gamma} e^{-t} dt - (EY)^2 \\ &= \beta^{2/\gamma} [\Gamma(\frac{2}{\gamma} + 1) - \Gamma^2(\frac{1}{\gamma} + 1)] \end{aligned}$$

$$(\text{b}) \quad f_X(x) = \frac{1}{\beta} e^{-x/\beta} \quad , \quad Y = (2X/\beta)^{1/2} = g(x) \quad , \quad g^{-1}(y) = \frac{\beta y^2}{2}$$

$$f_Y(y) = f_X(g^{-1}(y)) \frac{d}{dy} g^{-1}(y) = y e^{-y^2/2}$$

$$EY = \int_0^\infty y f_Y(y) dy = \int_0^\infty y^2 e^{-y^2/2} dy = \sqrt{\frac{\pi}{2}}$$

$$VarY = EY^2 - (EY)^2 = \int_0^\infty y^3 e^{-y^2/2} dy - \frac{\pi}{2} = \int_0^\infty 2t e^{-t} dt - \frac{\pi}{2}$$

$$= 2 - \frac{\pi}{2}$$

$$(\text{c}) \quad f_X(x) = \frac{1}{\Gamma(a)b^a} x^{a-1} e^{-x/b} \quad , \quad Y = \frac{1}{X}, g^{-1}(y) = \frac{1}{y}$$

$$f_Y(y) = \frac{1}{\Gamma(a)b^a y^{a+1}} e^{-\frac{1}{by}}$$

$$EY = \int_0^\infty y f_Y(y) dy = \int_0^\infty \frac{1}{x} \frac{1}{\Gamma(a)b^a} x^{a-1} e^{-x/b} dx$$

$$= \int_0^\infty \frac{1}{\Gamma(a)b^a} x^{a-2} e^{-x/b} dx = \frac{\Gamma(a-1)}{\Gamma(a)b}$$

$$= \frac{1}{(a-1)b}$$

$$\begin{aligned}
EY^2 &= \int_0^\infty y^2 f_Y(y) dy = \int_0^\infty \frac{1}{x^2 \Gamma(a) b^a} x^{a-1} e^{-x/b} dx \\
&= \int_0^\infty \frac{1}{\Gamma(a) b^a} x^{a-3} e^{-x/b} dx = \frac{\Gamma(a-2)}{\Gamma(a) b^2} \\
&= \frac{1}{(a-1)(a-2)b^2}
\end{aligned}$$

$$VarY = EY^2 - (EY)^2 = \frac{1}{(a-1)^2(a-2)b^2}$$

$$(d) f_X(x) = \frac{x}{\sqrt{\pi}\beta} x^{1/2} e^{-x/\beta}, \quad Y = (x/\beta)^{1/2}, \quad g^{-1}(y) = \beta y^2$$

$$f_Y(y) = \frac{4}{\sqrt{\pi}} y^2 e^{-y^2}$$

$$EY = \frac{2}{\sqrt{\pi}} \int_0^\infty y^2 e^{-y^2} d(y^2) = \frac{2}{\sqrt{\pi}}$$

$$EY^2 = \int_0^\infty \frac{2}{\sqrt{\pi}} y^3 e^{-y^2} d(y^2) = \int_0^\infty \frac{2}{\sqrt{\pi}} t^{2/3} e^{-t} dt = \frac{x}{\sqrt{\pi}} \Gamma\left(\frac{5}{2}\right) = \frac{3}{2}$$

$$VarY = EY^2 - (EY)^2 = \frac{9}{4} - \frac{4}{\pi}$$

$$(e) f_X(x) = e^{-x}, \quad Y = \alpha - \gamma \log(X), \quad g^{-1}(y) = e^{\frac{\alpha-y}{\gamma}}$$

$$f_Y(y) = \frac{1}{\gamma} \exp(-e^{\frac{\alpha-y}{\gamma}}) e^{\frac{\alpha-y}{\gamma}}$$

$$EY = \int_{-\infty}^\infty y f_Y(y) dy = \int_0^\infty (\alpha - \gamma \log(x)) e^{-x} dx = \alpha - \gamma \int_0^\infty \log(x) e^{-x} dx$$

$$\text{因为 } \log(x) = \left. \frac{dx^t}{dt} \right|_{t=0}$$

$$\int_0^\infty \log(x) e^{-x} dx = \left. \frac{d}{dt} \left(\int_0^\infty x^t e^{-x} dx \right) \right|_{t=0} = \Gamma'(t+1)|_{t=0} = \Gamma'(1)$$

$$\text{并且 } \Gamma'(z) = \Gamma(z)\phi(z), \quad \phi(1) = -c \text{ (c 是欧拉常数)} \Rightarrow \Gamma'(1) = -c$$

$$EY = \alpha + \gamma c$$

$$EY^2 = \int_{-\infty}^\infty y^2 f_Y(y) dy = \int_0^\infty (\alpha - \gamma \log(x))^2 e^{-x} dx$$

$$\int_0^\infty \log^2(x) e^{-x} dx = 2 \left. \frac{d^2}{dt^2} \left(\int_0^\infty x^t e^{-x} dx \right) \right|_{t=0} = \Gamma''(1)$$

$$\Gamma''(z) = \Gamma'(z)\phi(z) + \Gamma(z)\phi'(z)$$

$$\phi'(1) = \frac{\pi^2}{6} \Rightarrow \Gamma''(1) = c^2 + \frac{\pi^2}{6} \Rightarrow$$

$$EY^2 = \int_0^\infty (\alpha - \gamma \log(x))^2 e^{-x} dx = \alpha^2 + 2\alpha\gamma c + \gamma^2(c^2 + \frac{\pi^2}{6}) \text{ (c 是欧拉常数)}$$

$$VarY = EY^2 - (EY)^2 = \frac{\gamma^2 \pi^2}{6}$$

$$\mathbf{3.25} \quad P(t \leq T \leq t + \delta) = P(t \leq t + \delta) - P(T \leq t) = F_T(t + \delta) - F_T(t)$$

$$\begin{aligned} h_T(t) &= \lim_{\delta \rightarrow \infty} \frac{F_T(t + \delta) - F_T(t)}{\delta} \frac{1}{F_T(t)} \\ &= \frac{f_T(t)}{1 - F_T(t)} = -\log \left(1 - F_T(t) \right) \end{aligned}$$

$$\mathbf{3.26} \quad (\text{a}) f_T(t) = \frac{1}{\beta - t/\beta}, \quad F_T(t) = 1 - e^{-t/\beta}$$

$$h_T(t) = \frac{1}{\beta}$$

$$(\text{b}) f_T(t) = \frac{\gamma}{\beta} t^{\gamma-1} e^{-t^\gamma/\beta}, \quad F_T(t) = \int_0^t f_T(s) ds = 1 - e^{-t^\gamma/\beta}$$

$$h_T(t) = \frac{\gamma}{\beta} t^{\gamma-1}$$

$$(\text{c}) F_T(t) = \frac{1}{1 + e^{-(t-\mu)/\beta}}, \quad f_T(t) = F'_T(t) = \frac{e^{-(t-\mu)/\beta}}{\beta(1 + e^{-(t-\mu)/\beta})^2}$$

$$h_T(t) = \frac{1}{\beta(1 + e^{-(t-\mu)/\beta})}$$

$$\mathbf{3.28} \quad (\text{a}) f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{\mu^2}{2\sigma^2}} e^{-\frac{x^2}{\sigma^2} + \frac{\mu x}{\sigma^2}}$$

$$h(x) = 1, \quad c = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{\mu^2}{2\sigma^2}}, \quad \omega_1 = \frac{-1}{2\sigma^2}, \quad t_1 = x^2, \quad \omega_2 = \frac{\mu}{\sigma^2}, \quad t_2 = x$$

$$(\text{b}) f(x) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta} = \frac{1}{\Gamma(\alpha)\beta^\alpha} e^{(\alpha-1)\log x - x/\beta}$$

$$h(x) = 1, \quad c(\alpha, \beta) = \frac{1}{\Gamma(\alpha)\beta^\alpha}, \quad \omega_1 = \alpha - 1, \quad t_1 = \log x, \quad \omega_2 = -\frac{1}{\beta}, \quad t_2 = x$$

$$(\text{c}) f(x) = \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} = \frac{1}{B(\alpha, \beta)} e^{(\alpha-1)\log x + (\beta-1)\log(1-x)}$$

$$h(x) = 1, \quad c(\alpha, \beta) = \frac{1}{B(\alpha, \beta)}, \quad \omega_1 = \alpha - 1, \quad t_1 = \log x, \quad \omega_2 = \beta - 1,$$

$$t_2 = \log(1-x)$$

$$(\text{d}) f(x) = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-\lambda}}{x!} e^{\lambda \log x}$$

$$h(x) = \frac{1}{x!}, \quad c(\lambda) = e^{-\lambda}, \quad \omega_1 = \lambda, \quad t_1 = \log x$$

$$(\text{e}) f(x) = \binom{r+x-1}{x} p^t (1-p)^x = \binom{r+x-1}{x} p^r e^{(1-p)\log x}$$

$$h(x) = 1 \binom{r+x-1}{x}, \quad c(p) = p^r, \quad \omega_1 = 1-p, \quad t_1 = \log x$$

$$\mathbf{3.41} \quad (\text{a}) f(x|\mu) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

设 $f_1(x) = f(x|\mu_1)$, $f_2(x) = f(x|\mu_2)$, 且 $\mu_1 < \mu_2$

$$F_1(x) = \int_{-\infty}^x f_1(t)dt, \quad F_2(x) = \int_{-\infty}^x f_2(t)dt$$

因为 $f_2(x) = f_1(x - (\mu_2 - \mu_1))$

$$\text{所以 } F_2(x) = \int_{-\infty}^x f_1(t - (\mu_2 - \mu_1))dt = F_1(x) - \int_{x-(\mu_2-\mu_1)}^x f_1(t)dt$$

因为 $f_1(t) > 0$, 所以 $F_2(x) < F_1(x)$ (对所有的 x)

即 $F(x|\mu)$ 关于 μ 随机递增

$$(\text{b}) f(x|\beta) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta}$$

仍记 $f_1(x) = f(x|\beta_1)$, $f_2(x) = f(x|\beta_2)$, 且 $\beta_1 < \beta_2$

$$F_1(x) = \int_{-\infty}^x f_1(t)dt, \quad F_2(x) = \int_{-\infty}^x f_2(t)dt$$

因为 $f_2(t) = \frac{\beta_1}{\beta_2} f_1\left(\frac{\beta_1}{\beta_2}t\right)$

$$\text{所以 } F_2(x) = \int_0^x f_1\left(\frac{\beta_1}{\beta_2}t\right) \frac{\beta_1}{\beta_2} dt = \int_0^{\frac{\beta_1}{\beta_2}x} f_1(t)dt$$

而 $\frac{\beta_1}{\beta_2}x < x$, $f_1(x) > 0$ (对 $x \neq 0, x \neq \infty$)

$F_2(x) \leq F_1(x)$ 对所有 x , $F_2(x) < F_1(x)$ 对某些 x

即 $f(x|\beta)$ 关于 β 随机递增

3.42 (a) 设标准函数为 $f_0(x)$

$$f_1(x) = f_0(x - \mu_1), \quad f_2(x) = f_0(x - \mu_2), \quad \mu_1 < \mu_2$$

$$F_2(x) = \int_0^x f_2(t)dt = \int_0^x f_1(t - (\mu_2 - \mu_1))dt = F_1(x) - \int_{x-(\mu_2-\mu_1)}^x f_1(t)dt$$

因为 $f_1(t) \geq 0$ (对所有 x), 且 $f_1(t) \geq 0$ (对某些 x)

$F_2(x) \leq F_1(x)$ (对所有 x), 且 $F_2(x) < F_1(x)$ (对某些 x)

$f_0(x)$ 的位置函数族关于位置随机递增

(b) 设标准函数为 $f_0(x)$, $x \in [0, \infty)$

$$f_1(x) = \frac{1}{\sigma_1} f_0\left(\frac{x}{\sigma_1}\right) \quad f_2(x) = \frac{1}{\sigma_2} f_0\left(\frac{x}{\sigma_2}\right), \quad \sigma_2 > \sigma_1$$

$$F_2(x) = \int_0^x \frac{\sigma_1}{\sigma_2} f_1\left(\frac{\sigma_1}{\sigma_2} t\right) dt = \int_0^{\frac{\sigma_1 x}{\sigma_2}} f_1(t) dt = F_1(x) - \int_{\frac{\sigma_1 x}{\sigma_2}}^x f_1(t) dt$$

因为 $f_1(t) \geq 0$ (对所有 x), 且 $f_1(t) \geq 0$ (对某些 x)

$F_2(x) \leq F_1(x)$ (对所有 x), 且 $F_2(x) \leq F_1(x)$ (对某些 x)

$f_0(x)$ 的尺度函数族关于尺度随机递增

Proof of Theorem 3.4.2. (1) 首先 $\int f(x|\theta) dx = 1$ 并且有

$$\begin{aligned} \frac{\partial}{\partial \theta_j} f(x|\theta) &= h(x) \frac{\partial c(\theta)}{\partial \theta_j} \exp\left(\sum_i \omega_i t_i\right) + h(x) c(\theta) \exp\left(\sum_i \omega_i t_i\right) \left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right) \\ &= \frac{\partial \log c(\theta)}{\partial \theta_j} f(x|\theta) + f(x|\theta) \left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right) \end{aligned}$$

将上式两边积分 \Rightarrow

$$\begin{aligned} \frac{\partial \log c(\theta)}{\partial \theta_j} + \int f(x|\theta) \left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right) dx &= 0 \Rightarrow \\ E\left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right) &= -\frac{\partial}{\partial \theta_j} \log c(\theta) \end{aligned}$$

$$\begin{aligned} (2) \frac{\partial^2}{\partial \theta_j^2} \log c(\theta) &= -\int \left[f(x|\theta) \left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right)^2 + f(x|\theta) \left(\frac{\partial^2 \omega_i t_i}{\partial \theta_j^2}\right) - f(x|\theta) \left(\frac{\partial}{\partial \theta_j} \log c(\theta)\right)^2 \right] dx \\ &= -E\left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right)^2 - E\left(\frac{\partial^2 \omega_i t_i}{\partial \theta_j^2}\right) + \left(\frac{\partial}{\partial \theta_j} \log c(\theta)\right)^2 \Rightarrow \\ Var\left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right) &= E\left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right)^2 - \left(E \sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right)^2 \\ &= E\left(\sum_i \frac{\partial \omega_i t_i}{\partial \theta_j}\right)^2 - \frac{\partial^2}{\partial \theta_j^2} \log c(\theta) \end{aligned}$$

□