

CHAPTER 5

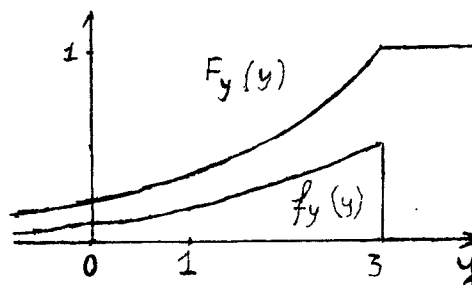
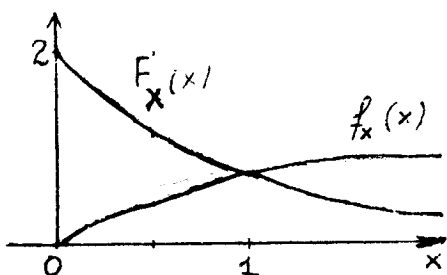
5-1 $\eta = 2\eta_x + 4 = 14$ $\sigma_y^2 = 4\sigma_x^2 = 16$

5-2 $\{y \leq y\} = \{-4x + 3 \leq y\} \{x \leq (y-3)/4\}$. Hence

$$F_y(y) = P \left\{ x \geq \frac{3-y}{4} \right\} = 1 - F_x \left(\frac{3-y}{4} \right) \quad f_y(y) = \frac{1}{4} f_x \left(\frac{3-y}{4} \right)$$

Since $F_x(x) = (1 - e^{-2x})U(x)$, this yields

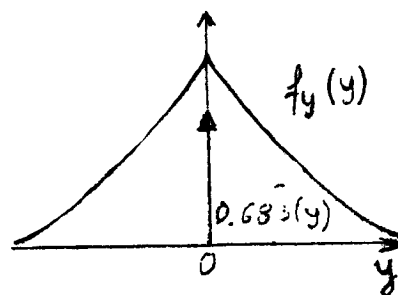
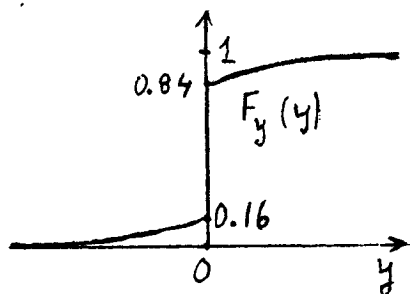
$$F_y(y) = e^{(y-3)/2} U \left(\frac{y-3}{2} \right) \quad f_y(y) = \frac{1}{2} e^{(y-3)/2} U \left(\frac{y-3}{2} \right)$$



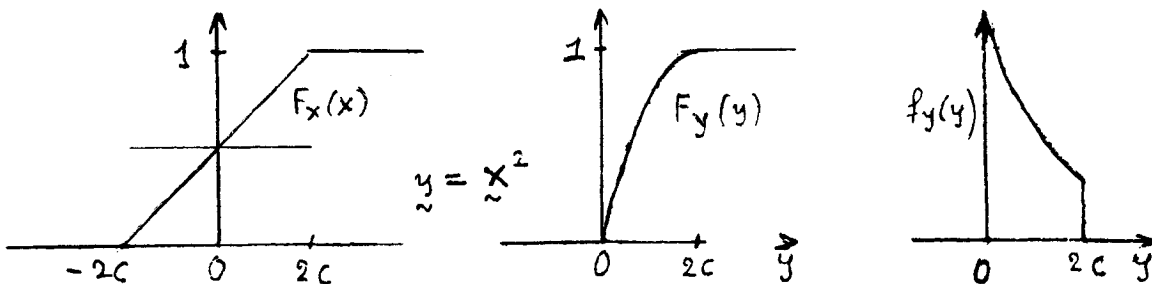
5-3 From Example 5-3 with $F_x = G(x/c)$:

$$f_y(y) = \begin{cases} G(y/c+1) & y \geq 0 \\ G(y/c-1) & y < 0 \end{cases}$$

$$f_y(y) = 0.68 \delta(y) + \frac{1}{c\sqrt{2\pi}} \left[e^{-(y+c)^2/2c^2} U(y) + e^{-(y-c)^2/2c^2} U(-y) \right]$$

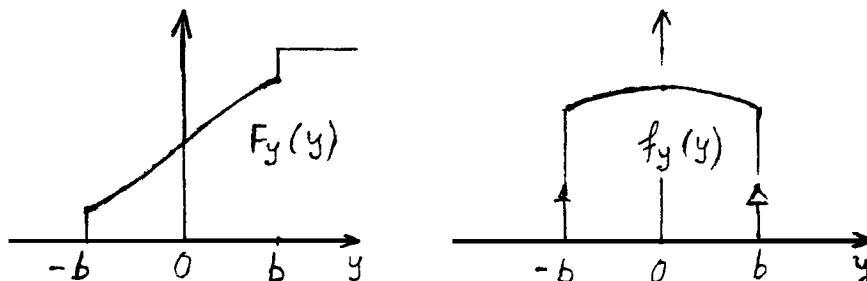


- 5-4 If $y = x^2$ and $F_x(x) = (x+2c)/4c$ for $|x| \leq 2c$, then (see Example 5-2) $F_y(y) = \sqrt{y}/2c$ and $f_y(y) = 1/4\sqrt{y}$ for $0 < y < 2c$.



- 5-5 From Example 5-4 with $F_x(x) = G(x/b)$: For $|x| \leq b$ $F_u(y) = G(y/b)$ and

$$f_y(y) = 0.16\delta(y+b) + \frac{1}{b\sqrt{2\pi}} e^{-y^2/2b^2} + 0.16\delta(y-b)$$



- 5-6 The equation $y = -\ln x$ has a single solution $x = e^{-y}$ for $y > 0$ and no solutions for $y < 0$. Furthermore, $g'(x) = -1/x = -e^y$. Hence

$$f_y(y) = \frac{f_x(e^{-y})}{e^y} \quad U(y) = e^{-y}U(y)$$

5-7 Clearly, $\underline{z} \leq z$ iff the number $\underline{n}(0,z)$ of the points in the interval $(0,z)$ is at least one.

Hence,

$$F_z(z) = P(\underline{z} \leq z) = P(\underline{n}(0,z) > 0) = 1 - P(\underline{n}(0,z) = 0)$$

The probability p that a particular point is in the interval $(0,z)$ equals $z/100$. With $n = 200$, $k = 0$, and $p = z/100$, (3-21) yields $P(\underline{n}(0,z) = 0) = (1-p)^{200}$. Hence,

(a) $F_z(z) = 1 - \left(1 - \frac{z}{100}\right)^{100}$

(b) From (4-107) it follows that $F_z(z) \simeq 1 - e^{-2z}$ for $z \ll 100$.

5.8

$$Y = \sqrt{X} \Rightarrow x_1 = y^2$$

$$\frac{dy}{dx} = \frac{1}{2\sqrt{x}} = \frac{1}{2y}$$

Thus

$$f_Y(y) = \frac{1}{\left|\frac{dy}{dx}\right|} f_X(x_1) = 2y f_X(y^2)$$

$$\frac{2y}{\lambda} e^{-y^2/\lambda} = \begin{cases} \frac{y}{\sigma^2} e^{-y^2/2\sigma^2}, & y > 0 \\ 0, & \text{otherwise} \end{cases}$$

which represents Rayleigh density function (with $\lambda = 2\sigma^2$).

5-9 For both cases, $f_y(y) = 0$ for $y < 0$.

(a) If $y > 0$ and $|x| = y$, then $x_1 = y$, $x_2 = -y$. Hence

$$f_y(y) = [f_x(y) + f_x(-y)]U(y)$$

(b) If $y > 0$ and $e^{-x}U(x) = y$, then $x = -\ln y$.

Furthermore, $P(\underline{y} = 0) = P(\underline{x} \leq 0) = F_x(0)$. Hence

$$f_y(y) = F_x(0)\delta(y) + \frac{1}{y} f_x(-\ln y)U(y)$$

- 5-10 (a) If $y \geq 0$ and $(x-1)U(x-1) = y$, then $\{y \leq y\} = \{x \leq y+1\}$.
 If $y < 0$, then $\{y < y\} = \{\emptyset\}$

$$F_y(y) = F_x(1+y)U(y) = [1 - e^{-2(y+1)}]U(y)$$

$$f_y(y) = (1 - e^{-2})\delta(y) + 2e^{-2(y+1)}U(y)$$

- (b) If $y > 0$ and $y = x^2$, then $\{y \leq y\} = \{-\sqrt{y} \leq x \leq \sqrt{y}\}$

$$F_y(y) = F_x(\sqrt{y}) - F_x(-\sqrt{y}) = (1 - e^{-2\sqrt{y}})U(y)$$

$$f_y(y) = \frac{1}{\sqrt{y}} e^{-2\sqrt{y}}U(y)$$

- 5-11 If $y = \arctan x$, then $\frac{dy}{dx} = \frac{1}{1+x^2}$

$$f_y(y) = (1+x^2)f_x(\tan y) = \frac{1+x^2}{\pi(1+x^2)} = \frac{1}{\pi} \quad \frac{\pi}{2} < y < \frac{\pi}{2}$$

- 5-12 (a) If $y = x^3$ then $x = \sqrt[3]{y}$ for any y

$$f_y(y) = \frac{1}{3\sqrt[3]{y^2}} f_x(\sqrt[3]{y}) = \frac{1}{12\pi\sqrt[3]{y^2}}$$

for $|y| < 8\pi^3$ and zero otherwise

- (b) If $y = x^4$ and $y > 0$, then $x_1 = \sqrt[4]{y}$ $x_2 = -\sqrt[4]{y}$

$$f_y(y) = \frac{1}{4\sqrt[4]{y^3}} \left[f_x(\sqrt[4]{y}) + f_x(-\sqrt[4]{y}) \right] = \frac{1}{8\pi\sqrt[4]{y^3}}$$

for $0 < y < 16\pi^4$ and zero otherwise

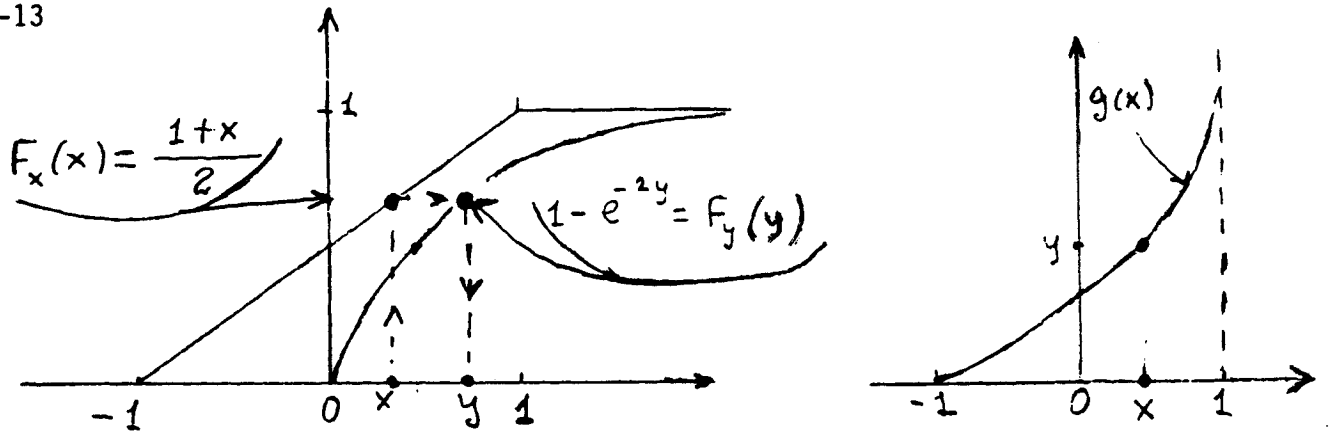
- (c) If $y = 2 \sin(3x + 40^\circ)$ and $|y| < 2$ then $x = x_i$ as shown.

$$\frac{dy}{dx} = \frac{1}{6\sqrt{1-y^2/4}}$$

In the interval $(-2\pi, 2\pi)$ there are 12 x_i 's. Hence

$$f_y(y) = \frac{1}{3\sqrt{4-y^2}} \sum_i f_x(x_i) = \frac{12}{12\pi\sqrt{4-y^2}} = \frac{1}{\pi\sqrt{4-y^2}}$$

for $|y| < 2$ and zero otherwise.



As in (5-43)

$$F_y[g(x)] = F_x(x)$$

$$\frac{1+x}{2} = 1 - e^{-2y}$$

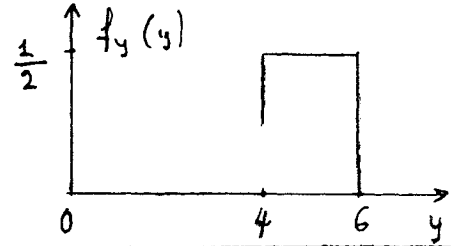
$$y = g(x) = -\frac{1}{2} \ln \frac{1-x}{2}$$

for $|x| < 1$. For $x \leq -1$, $g(x) = 0$; for $x \geq 1$, $g(x) = \infty$.

5-14 (a) $g(x) = 2F_x(x) + 4$ $g'(x) = 2f_x(x)$

If $4 < y < 6$ then $y = 2F_x(x) + 4$ has a unique solution x_1 and

$$f_y(y) = \frac{f_x(x_1)}{2f_x(x_1)} = \frac{1}{2}$$



(b) Similarly $g(x) = 2F_x(x) + 8$

5-15 (a) The RV x takes the values $k = 0, 1, \dots, 10$ and

$$P\{x = k\} = p_k = \binom{10}{k} \frac{1}{2^{10}} \quad 0 \leq k \leq 10$$

$F_x(x)$ is a staircase function with discontinuities at the points $x = k$ and jumps equal to p_k .

(b) The RV $y = (x - 3)^2$ takes the values $y = k^2$ for $k = 0, 1, \dots, 7$ and probabilities $P\{y = k^2\} = q_k$.

$k =$	0	1	2	3	4	5	6	7
$q_k =$	p_3	$p_2 + p_4$	$p_1 + p_5$	$p_0 + p_6$	p_7	p_8	p_9	p_{10}

5.16

 $X \sim \text{Beta}(\alpha, \beta)$ gives

$$f_X(x) = \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1}, \quad 0 < x < 1.$$

$$Y = 1 - X \Rightarrow x_1 = 1 - y, \quad \left| \frac{dy}{dx} \right| = 1$$

$$\Rightarrow F_Y(y) = \frac{1}{\left| \frac{dy}{dx} \right|} f_X(1-y) = \begin{cases} \frac{1}{B(\beta, \alpha)} y^{\beta-1} (1-y)^{\alpha-1}, & 0 < y < 1 \\ 0, & \text{otherwise.} \end{cases}$$

This gives

$$Y \sim \text{Beta}(\beta, \alpha).$$

5.17

 $X \sim \chi^2(n) \Rightarrow$

$$f_X(x) = \frac{1}{2^{n/2} \Gamma(n/2)} x^{n/2-1} e^{-x/2} U(x)$$

$$y = \sqrt{x} \Rightarrow x_1 = y^2$$

$$\frac{dy}{dx} = \frac{1}{2y}$$

Thus

$$f_Y(y) = 2y f_X(y^2) = \frac{y^{n-1}}{2^{n/2-1} \Gamma(n/2)} e^{-y^2/2} U(y)$$

and it represents the chi-distribution.

5.18

 $X \sim U(0, 1)$

$$Y = -2 \log X \Rightarrow x_1 = e^{-y/2}$$

$$\frac{dy}{dx} = -\frac{2}{x} = -2e^{y/2}$$

$$f_Y(y) = \frac{1}{\left| \frac{dy}{dx} \right|} f_X(x_1) = \frac{1}{2} e^{-y/2} U(y)$$

$$\sim \text{Exponential}(2) \equiv \chi^2(2)$$

5.19

$$f_X(x) = \lambda e^{-\lambda x} u(x)$$

$$Y = X^{1/\beta} \Rightarrow x_1 = y^\beta$$

$$\left| \frac{dy}{dx} \right| = \frac{1}{\beta} x^{1/\beta-1} = \frac{1}{\beta} y^{1-\beta}$$

$$f_Y(y) = \frac{1}{\left| \frac{dy}{dx} \right|} f_X(x_1) = \lambda \beta y^{\beta-1} e^{-\lambda y^\beta} U(y)$$

and it represents Weibull distribution

5-20 For $|y| < a$ the equation $y = a \sin \omega t$ has infinitely many solutions τ_i ; in each interval of length $2\pi/\omega$ there are two such solutions. Furthermore,
 $y'(t) = \omega \sqrt{a^2 - y^2}$

$$\tau_i = \frac{1}{\omega} \sin^{-1} \frac{y}{a} \quad \tau_{i+2} - \tau_i = \frac{2\pi}{\omega} \xrightarrow{\omega \rightarrow \infty} 0$$

Hence,

$$\frac{1}{\omega \sqrt{a^2 - y^2}} \sum_{i=-\infty}^{\infty} f_t(\tau_i) \xrightarrow{\omega \rightarrow \infty} \frac{1}{\sqrt{a^2 - y^2}} \frac{2}{2\pi} \int_{-\infty}^{\infty} f_t(\tau) d\tau = \frac{1}{\pi \sqrt{a^2 - y^2}}$$

5-21 If $y > 0$ then

$$F_y(y|x \geq 0) = F_x(\sqrt{y}|x \geq 0) + F_x(-\sqrt{y}|x \geq 0) = F_x(\sqrt{y}|x \geq 0)$$

$$F_x(\sqrt{y}|x \geq 0) = \frac{P\{0 < x < \sqrt{y}\}}{P\{x \geq 0\}} = \frac{F_x(\sqrt{y}) - F_x(0)}{1 - F_x(0)}$$

$$f_y(y|x \geq 0) = \frac{d}{dy} F_y(\sqrt{y}|x \geq 0) = \frac{f_x(\sqrt{y})}{2\sqrt{y}[1 - F_x(0)]}$$

5-22 (a) $\eta_y = a \eta_x + b \quad \sigma_y^2 = E\{[a \underline{x} + b - (a \eta_x + b)]^2\}$

$$\sigma_y^2 = E\{a^2(\underline{x} - \eta_x)^2\} = a^2 \sigma_x^2$$

(b) $\underline{y} = \frac{\underline{x} - \eta_x}{\sigma_x} \quad E\{\underline{y}\} = 0 \quad \sigma_y^2 = \frac{\sigma_x^2}{\sigma_x^2} = 1$

5-23 If \underline{x} has a Rayleigh density, then [see (5-76)]

$$E\{\underline{x}^2\} = 2a^2 \quad E\{\underline{x}^4\} = 8a^4$$

If $\underline{y} = b + c\underline{x}^2$, then

$$E\{\underline{y}\} = b + 2a^2 c \quad E\{\underline{y}^2\} = b^2 + 4a^4 c + 8a^4 c^2$$

$$\sigma_y^2 = E\{\underline{y}^2\} - E^2\{\underline{y}\} = 4a^4 c^2$$

$$5-24 \quad y = 3x^2 \quad E\{x^2\} = \sigma_x^2 = 4 \quad E\{x^4\} = 3\sigma_x^4 = 48$$

$$E\{y\} = 12 \quad E\{y^2\} = 9 \times 48 = 432 \quad \sigma_y^2 = 432 - 144 = 288$$

If $y > 0$ then $3x^2 = y$ for $x = \pm\sqrt{y/3}$ $y' = 6x$

$$f_y(y) = \frac{24}{\sqrt{12y}} f_x\left(\sqrt{\frac{y}{3}}\right) = \frac{1}{\sqrt{24\pi y}} e^{-y/24} U(y)$$

5.25

$$X \sim B(n, p) \Rightarrow P(X = k) = \binom{n}{k} p^k q^{n-k}, \quad k = 0, 1, 2, \dots, n.$$

a)

$$\begin{aligned} E(X) &= \sum_{k=0}^n k P(X = k) = \sum_{k=1}^n k \frac{n!}{k!(n-k)!} p^k q^{n-k} \\ &= np \sum_{k=1}^n \frac{(n-1)!}{(k-1)!(n-k)!} p^{k-1} q^{n-k} \\ &= np (p+q)^{n-1} = np. \end{aligned}$$

b)

$$\begin{aligned} E[X(X-1)] &= \sum_{k=2}^n k(k-1) \frac{n!}{k!(n-k)!} p^k q^{n-k} \\ &= n(n-1)p^2 \sum_{k=2}^n \frac{(n-2)!}{(k-2)!(n-k)!} p^{k-2} q^{n-k} \\ &= n(n-1)p^2 (p+q)^{n-2} \\ &= n(n-1)p^2 \end{aligned}$$

c)

$$\begin{aligned} E[X(X-1)(X-2)] &= \sum_{k=3}^n k(k-1)(k-2) \frac{n!}{k!(n-k)!} p^k q^{n-k} \\ &= n(n-1)(n-2)p^3 \sum_{k=3}^n \frac{(n-3)!}{(k-3)!(n-k)!} p^{k-3} q^{n-k} \\ &= n(n-1)(n-2)p^3 (p+q)^{n-3} \\ &= n(n-1)(n-2)p^3 \end{aligned}$$

$$\begin{aligned} E(X^2) &= E(X(X-1)) + E(X) = n^2p^2 + npq \\ E(X^3) &= E(X(X-1)(X-2)) + 3E(X^2) - 2E(X) \\ &= n(n-1)(n-2)p^3 + 3(n^2p^2 + npq) - 2np \\ &= n^3p^3 + 3n^2p^2q + npq(q-p). \end{aligned}$$

5.26

$$X \sim P(\lambda) \Rightarrow P(X = k) = e^{-\lambda} \frac{\lambda^k}{k!}, \quad k = 0, 1, 2, \dots$$

a)

$$E(X) = \lambda, \quad \text{Var}(X) = \sigma_X^2 = \lambda$$

From Chebyshev's inequality (5-88)

$$P(|X - \mu| < \lambda) > 1 - \frac{\sigma^2}{\lambda^2} = 1 - \frac{1}{\lambda}$$

But

$$|X - \mu| < \lambda = |X - \lambda| < \lambda \Rightarrow 0 < X < 2\lambda$$

which gives

$$P(0 < X < 2\lambda) > 1 - \frac{1}{\lambda} = \frac{\lambda - 1}{\lambda}.$$

b)

$$\begin{aligned} E[X(X-1)] &= \sum_{k=2}^{\infty} k(k-1) e^{-\lambda} \frac{\lambda^k}{k!} \\ &= e^{-\lambda} \lambda^2 \sum_{k=2}^{\infty} \frac{\lambda^{k-2}}{(k-2)!} = e^{-\lambda} \lambda^2 e^{\lambda} = \lambda^2. \end{aligned}$$

$$\begin{aligned} E[X(X-1)(X-2)] &= \sum_{k=3}^{\infty} k(k-1)(k-2) e^{-\lambda} \frac{\lambda^k}{k!} \\ &= e^{-\lambda} \lambda^3 \sum_{k=3}^{\infty} \frac{\lambda^{k-3}}{(k-3)!} = \lambda^3. \end{aligned}$$

5-27 Follows from (4-74)

$$E\{\underline{x}\} = \int_{-\infty}^{\infty} \underline{x} f(\underline{x}) d\underline{x} = \int_{-\infty}^{\infty} \underline{x} \sum_{\mathbf{1}} f(\underline{x} | A_{\mathbf{1}}) P(A_{\mathbf{1}}) d\underline{x}$$

because

$$E\{\underline{x} | A_{\mathbf{1}}\} = \int_{-\infty}^{\infty} \underline{x} f(\underline{x} | A_{\mathbf{1}}) d\underline{x}$$

5-28 From (5-89) with $\alpha = \sqrt{\eta}$:

$$P\{\underline{x} \geq \sqrt{\eta}\} \leq \eta / \sqrt{\eta} = \sqrt{\eta}$$

5-29 From (5-86) with $g(x) = x^3$ $g''(x) = 6x$:

$$E\{\underline{x^3}\} = \eta^3 + 6\eta \frac{\sigma^2}{2} = 1120$$

5-30 (a) If $y = x^3$, then $x = \sqrt[3]{y}$ $g'(x) = 3x^2 = 3\sqrt[3]{y^2}$

But $f_x(x) = 0.5$ for $10 < x < 12$, i.e., for $10^3 < y < 12^3$

and (5-16) yields

$$f_y(y) = \frac{0.5}{3\sqrt[3]{y^2}} \quad 10^3 < y < 12^3$$

and zero otherwise.

(b) 1.

$$E\{\underline{x^3}\} = 0.5 \int_{10}^{12} x^3 dx = 1342$$

2. With $g(x) = x^3$ $E\{\underline{x}\} = 11$ $\sigma_x^2 = 1/3$, (5-86) yields

$$E\{\underline{x^3}\} \approx 11^3 + 6 \times 11 \times \frac{1}{6} \approx 1342$$

5-31 With $g(x)=1/x$, $g''(x)=2/x^3$, $\eta=100$, and $\sigma=3$, (5-55) yields

$$E\left\{\frac{1}{\underline{x}}\right\} \approx \frac{1}{100} + \frac{9}{2} \times \frac{2}{100^3} = 0.010009$$

$$\frac{\partial |x-a|}{\partial a} = \begin{cases} 1 & x < a \\ -1 & x > a \end{cases} \quad \text{If } I(a) = E\{|x-a|\} \text{ then}$$

$$\begin{aligned} \frac{dI(a)}{da} &= E \frac{\partial |x-a|}{\partial a} = 1 P\{x < a\} - 1 P\{x > a\} \\ &= 2 F(a) - 1 \end{aligned}$$

$$\begin{aligned} \text{(a)} \quad I(a) &= I(m) + \int_m^a I'(\alpha) d\alpha = I(m) + \int_m^a [2 F(\alpha) - 1] d\alpha \\ &= E\{|x - m|\} - 2 \int_m^a x f(x) dx \end{aligned}$$

because

$$\int_m^a F(\alpha) d\alpha = a F(a) - m F(m) - \int_m^a x f(x) dx$$

$$F(m) = \frac{1}{2} \int_m^a f(x) dx = F(a) - F(m)$$

(b) $I(a) = E\{|x - a|\}$ is minimum if

$$I'(a) = 2F(a) - 1 = 0 \quad \text{i.e. if } F(a) = \frac{1}{2} \quad a = m$$

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

$$\eta = E\{x\} = \int_0^{\infty} x f(x) dx + \int_{-\infty}^0 x f(x) dx$$

$$\frac{E\{|x|+\eta\}}{2} = \int_0^{\infty} x f(x) dx = \frac{1}{\sigma\sqrt{2\pi}} \int_0^{\infty} x e^{-(x-\eta)^2/2\sigma^2} dx$$

$$\frac{1}{\sigma\sqrt{2\pi}} \int_0^{\infty} (x+\eta) e^{-(x-\eta)^2/2\sigma^2} dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{\eta}^{\infty} y e^{-y^2/2\sigma^2} dy = \frac{\sigma}{\sqrt{2\pi}} e^{-\eta^2/2\sigma^2}$$

$$\frac{1}{\sigma\sqrt{2\pi}} \int_0^{\infty} e^{-(x-\eta)^2/2\sigma^2} dx = G\left(\frac{\eta}{\sigma}\right)$$

Multiplying the last line by η and subtracting from the fourth line, we obtain

$$\frac{E\{|x|+\eta\}}{2} = \frac{\sigma}{\sqrt{2\pi}} e^{-\eta^2/2\sigma^2} + G\left(\frac{\eta}{\sigma}\right)$$

5-34 The proof is given in sec 14-3: [see (14-100)].

5-35 (a) Follows from (5-89) (b) $e^{sx} \geq e^{sA}$ iff $x \geq A$ for $s > 0$ and $x \leq A$ for $s < 0$.

5.36 See proof for Lyapunov inequality (Ch.5, Eq.(5-92).)

5-37 (a) If $\phi(\omega) = e^{-\alpha|\omega|}$ then [see (5-102)]

$$f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-\alpha|\omega|} e^{j\omega x} d\omega = \frac{1}{\pi} \int_0^{\infty} \cos \omega x e^{-\alpha\omega} d\omega = \frac{\alpha}{\pi(\alpha^2 + x^2)}$$

(b) If $f(x) = \frac{\alpha}{2} e^{-\alpha|x|}$, then [see (5-94)]

$$\phi(\omega) = \frac{\alpha}{2} \int_{-\infty}^{\infty} e^{-\alpha|x|} e^{-j\omega x} dx = \alpha \int_0^{\infty} e^{-\alpha x} \cos \omega x dx = \frac{\alpha^2}{\alpha^2 + \omega^2}$$

5.38 a) On comparing Eq.(4-34) with Eq.(5-106), Example 5-29, we get

$$X \sim G(\alpha, \beta) \Rightarrow \phi_X(\omega) = (1 - j\beta\omega)^{-\alpha}$$

$$\phi'_X(\omega) = -\alpha(1 - j\beta\omega)^{-(\alpha+1)} (-j\beta)$$

so that

$$E(X) = \frac{1}{j} \phi'_X(0) = \alpha\beta.$$

Similarly

$$\phi''_X(\omega) = j\alpha\beta(\alpha+1)(1 - j\beta\omega)^{-(\alpha+2)} (j\beta)$$

and hence

$$E(X^2) = \frac{1}{j^2} \phi''_X(0) = \alpha\beta^2(\alpha+1).$$

Thus

$$\text{Var}(X) = E(X^2) - (E(X))^2 = \alpha\beta^2.$$

b)

$$X \sim \chi^2(n) \Rightarrow \alpha = \frac{n}{2}, \quad \beta = 2$$

in Gamma(α, β). This gives

$$\phi_X(\omega) = (1 - j2\omega)^{-n/2}$$

$$E(X) = n$$

$$\text{Var}(X) = 2n.$$

c)

$$X \sim B(n, p).$$

From Prob 5-25 (a)-(b)

$$E(X) = np$$

$$\text{Var}(X) = E(X(X-1)) + E(X) = npq.$$

and

$$\phi_X(\omega) = \sum_{k=0}^n e^{jk\omega} P(X=k)$$

$$= \sum_{k=0}^n \binom{n}{k} (pe^{j\omega})^k q^{n-k} = (pe^{j\omega} + q)^n.$$

d)

$X \sim N \text{ Binomial } (r, p)$.

From (4-64)

$$\begin{aligned}\phi_X(\omega) &= \sum_{k=0}^{\infty} e^{jk\omega} P(X = k) \\ &= \sum_{k=0}^{\infty} \binom{r+k-1}{k} p^r (qe^{j\omega})^k \\ &= p^r \sum_{k=0}^{\infty} \binom{-r}{k} (-qe^{j\omega})^k \\ &= p^r (1 - qe^{j\omega})^{-r}.\end{aligned}$$

5-39

$$\Gamma(z) = \sum_{k=0}^{\infty} p q^k z^k = \frac{p}{1 - qz} \quad q = 1-p$$

$$\Gamma'(z) = \frac{pq}{(1-qz)^2} \quad \Gamma'(1) = \frac{pq}{(1-q)^2} = \frac{p}{q} = \eta_x$$

$$\Gamma''(z) = \frac{2pq^2}{(1-qz)^3} \quad \Gamma''(1) = \frac{2q^2}{p^2} = m_2 - m_1$$

$$\sigma_x^2 = m_2 - m_1^2 = 2 \frac{q^2}{p^2} + m_1 - m_1^2 = \frac{q}{p}$$

5-40

$$\Gamma(z) = p^n \sum_{k=0}^{\infty} \binom{-n}{k} (-q)^k z^k = p^n (1 - qz)^{-n}$$

(binomial expansion with negative exponent)

$$\Gamma'(z) = \frac{n p^n q}{(1-qz)^{n+1}} \quad \Gamma'(1) = \frac{nq}{p} = \eta_x$$

$$\Gamma''(z) = \frac{n(n+1)p^n q^2}{(1-qz)^{n+2}} \quad \Gamma''(1) = \frac{n(n+1)q^2}{p^2} = m_2 - m_1$$

$$\sigma_x^2 = \Gamma''(1) + m_1 - m_1^2 = \frac{nq}{p}$$

5.41 We have

$$P(X = k) = \binom{k-1}{r-1} p^r q^{k-r}, \quad k = r, r+1, \dots$$

Let $k = n + r$ so that

$$\begin{aligned} P(X = n+r) &= \binom{n+r-1}{r-1} p^r q^n, \quad n = 0, 1, 2, \dots \\ &= \frac{(n+r-1)!}{n!(r-1)!} p^r (1-p)^n \\ &= \frac{1}{n!} \frac{(n+r-1)(n+r-2)\cdots(r)}{r^n} [r(1-p)]^n p^r \\ &= \frac{\lambda^n}{n!} \left\{ \left(1 + \frac{n-1}{r}\right) \left(1 + \frac{n-2}{r}\right) \cdots \right\} \left(1 - \frac{r(1-p)}{r}\right)^r \\ &= \frac{\lambda^n}{n!} \left\{ \prod_{k=1}^n \left(1 + \frac{n-k}{r}\right) \right\} \left(1 - \frac{\lambda}{r}\right)^r, \end{aligned}$$

where $\lambda = r(1-p)$. Thus

$$\begin{aligned} \lim_{r \rightarrow \infty} P(X = n+r) &= \frac{\lambda^n}{n!} \left\{ \lim_{r \rightarrow \infty} \prod_{k=1}^n \left(1 + \frac{n-k}{r}\right) \right\} \lim_{r \rightarrow \infty} \left(1 - \frac{\lambda}{r}\right)^r \\ &\rightarrow \frac{\lambda^n}{n!} e^{-\lambda} \sim P(\lambda). \end{aligned}$$

$$\begin{aligned}
 5-42 \quad E\{e^{sX}\} &= e^{s\eta} E\{e^{s(X-\eta)}\} = e^{s\eta} E\left\{\sum_{n=0}^{\infty} \frac{s^n}{n!} (X-\eta)^n\right\} \\
 &= e^{s\eta} \sum_{n=0}^{\infty} \frac{s^n}{n!} \mu_n
 \end{aligned}$$

5-43 If $\phi(\omega_1) = 0$, then [see also (9-176)]

$$\int_{-\infty}^{\infty} (1 - e^{j\omega_1 x}) f(x) dx = 0, \text{ hence, } f(x) = \sum_{n=-\infty}^{\infty} p_n \delta(x - \frac{2\pi n}{\omega_1})$$

5-44 (a) If $\eta = 0$, then $m_n = \mu_n$ $\lambda_1 = \eta = 0$

$$\phi(s) = \sum_{n=0}^{\infty} \frac{\mu_n}{n!} s^n \qquad \psi(s) = \sum_{n=2}^{\infty} \frac{\lambda_n}{n!} s^n$$

$$1 + \frac{\mu_2}{2!} s^2 + \frac{\mu_3}{3!} s^3 + \frac{\mu_4}{4!} s^4 + \dots = \exp\left\{\frac{\lambda_2}{2!} s^2 + \frac{\lambda_3}{3!} s^3 + \frac{\lambda_4}{4!} s^4 + \dots\right\}$$

Expanding the exponential and equating powers of s , we obtain

$$\mu_2 = \lambda_2 \qquad \mu_3 = \lambda_3 \qquad \frac{\mu_4}{4!} = \frac{\lambda_4}{4!} + \frac{1}{2!} \left(\frac{\lambda_2}{2!}\right)^2$$

(b) If y is $N(0; \sigma_y^2)$ then

$$\psi_y(s) = \frac{\lambda_2}{2} s^2, \text{ hence, } \lambda_n = 0 \text{ for } n \geq 3$$

5-45

$$P\{\underline{y} = 0\} = P\{\underline{x} \leq 1\} = p_0 + p_1$$

$$P\{\underline{y} = k\} = P\{\underline{x} = k + 1\} = p_{k+1} \quad k \geq 1$$

$$\Gamma_y(z) = p_0 + p_1 + \sum_{k=1}^{\infty} p_{k+1} z^k = p_0 + z^{-1}[\Gamma_x(z) - p_0]$$

$$\eta_y = \sum_{k=1}^{\infty} k p_{k+1} = \sum_{r=1}^{\infty} r p_r - \sum_{r=1}^{\infty} p_r = \eta_x - 1 + p_0$$

$$E\{\underline{y}^2\} = \sum_{k=1}^{\infty} k^2 p_{k+1} = \sum_{r=1}^{\infty} (r-1)^2 p_r = E\{\underline{x}^2\} - 2\eta_x + 1 - p_0$$

5-46

$$0 \leq E \left\{ \left| \sum_{i=1}^n a_i e^{j\omega_i \underline{x}} \right|^2 \right\} = E \left\{ \sum_{i=1}^n \sum_{j=1}^n a_i a_j^* e^{j(\omega_i - \omega_j) \underline{x}} \right\}$$

$$= \sum_{i=1}^n \sum_{j=1}^n a_i a_j^* \phi(\omega_i - \omega_j)$$

5-47 From the assumptions it follows that

$$g'(-x) = -g'(x) \quad g''(x) \geq 0 \quad f(x-\eta) = f(\eta-x)$$

Hence, if $I(a) = E\{g(\underline{x}-a)\}$, then

$$I'(a) = - \int_{-\infty}^{\infty} g'(x-a) f(x) dx \quad I'(\eta) = 0$$

$$I''(a) = \int_{-\infty}^{\infty} g''(x-a) f(x) dx \geq 0 \quad \text{all } a$$

Hence, $I(a)$ is minimum for $a = \eta$.

5-48

$$f(x, v) = \frac{1}{\sqrt{2\pi v}} e^{-x^2/2v}$$

$$\sqrt{2\pi} \frac{\partial f}{\partial v} = \frac{-1 + x^2/v}{2v \sqrt{v}} e^{-x^2/2v}$$

$$\sqrt{2\pi} \frac{\partial^2 f}{\partial x^2} = \frac{-1 + x/v}{v \sqrt{v}} e^{-x^2/2v}$$

Hence

(see also (6-198) - (6-199))

$$\boxed{\frac{\partial f}{\partial v} = \frac{1}{2} \frac{\partial^2 f}{\partial x^2}} \quad (1)$$

- (a) Integrating by parts, using (1) and assuming that $g^{(k)}(x)f(x) \rightarrow 0$ as $|x| \rightarrow \infty$, $k = 0, 1, 2$, we obtain

$$\begin{aligned} E\{g''(x)\} &= \int_{-\infty}^{\infty} \frac{d^2 g}{dx^2} f dx = \int_{-\infty}^{\infty} g \frac{\partial^2 f}{\partial x^2} dx = 2 \int_{-\infty}^{\infty} g \frac{\partial f}{\partial v} dx \\ &= 2 \frac{d}{dv} \int_{-\infty}^{\infty} g f dx = 2 \frac{d}{dv} E\{g(x)\} \end{aligned}$$

- (b) The moments $\mu_n(u) = E\{x^n\}$ of \underline{x} depend on the variance v of \underline{x} and (i) yields

$$\mu_n'(v) = \frac{d}{dv} E\{x^n\} = \frac{1}{2} E\{n(n-1)x^{n-2}\} = \frac{n(n-1)}{2} \mu_{n-2}(v)$$

Furthermore, $\mu_n(0) = 0$ because, if $v = 0$, then $\underline{x} = 0$.

Hence

$$\mu_n(v) = \frac{n(n-1)}{2} \int_0^v \mu_{n-2}(\beta) d\beta$$

5-49 The function

$$\Gamma(e^{j\omega}) = E\{e^{jx\omega}\} = \sum_{k=0}^{\infty} p_k e^{jk\omega}$$

is periodic with period 2π and Fourier series coefficients $p_k = E\{x = k\}$.

5.50 The event $\{X = 1\}$ is given by the disjoint union "TH \cup HT". Similarly, the event " $X = k$ " is given by the union of the disjoint events (k "T"s followed by "H" or k "H"s followed by "T")

$$\text{"TT} \dots \text{TTH"} \cup \text{"HH} \dots \text{HHT"}, \quad k = 1, 2, \dots$$

Thus

$$\begin{aligned} P(X = k) &= P(\text{"TT} \dots \text{TTH"} \cup \text{"HH} \dots \text{HHT"}) \\ &= P(\text{TT} \dots \text{TH}) + P(\text{HH} \dots \text{HT}) = q^k p + p^k q, \quad k = 1, 2, \dots \end{aligned}$$

Also

$$\begin{aligned} E(X) &= \sum_{k=1}^{\infty} kP(X = k) \\ &= \sum_{k=1}^{\infty} kq^k p + \sum_{k=1}^{\infty} kp^k q = pq \left\{ \sum_{k=1}^{\infty} kq^{k-1} + \sum_{k=1}^{\infty} kp^{k-1} \right\} \\ &= pq \left\{ \frac{\partial}{\partial q} \sum_{k=1}^{\infty} q^k + \frac{\partial}{\partial p} \sum_{k=1}^{\infty} p^k \right\} = pq \left\{ \frac{\partial}{\partial q} \left(\frac{q}{1-q} \right) + \frac{\partial}{\partial p} \left(\frac{p}{1-p} \right) \right\} \\ &= pq \left\{ \frac{1}{p^2} + \frac{1}{q^2} \right\} = \frac{p}{q} + \frac{q}{p}. \end{aligned}$$

5.51 (a) When samples are drawn with replacement, probability of each item being defective is given by

$$p = \frac{M}{N} < 1 \quad (\text{constant})$$

and

$$q = 1 - p = \frac{N - M}{N} < 1$$

represents the constant probability that the chosen item is not defective. In that case (with replacement), there are $\binom{n}{k}$ possible ways of arranging k defective items among n chosen items, and each such arrangement has probability $p^k q^{n-k}$. This gives

$$P(X = k) = \binom{n}{k} p^k q^{n-k}, \quad k = 0, 1, 2, \dots, n$$

which represents the Binomial distribution.

(b) If the samples are drawn without replacement, there are $\binom{M}{k}$ possible ways of choosing k defective item from a total of M defective items, and $\binom{N-M}{n-k}$ possible ways of choosing $n-k$ “good” items from $(N-M)$ “good” items independently. This gives

$$\binom{M}{k} \binom{N-M}{n-k}$$

to be the total number of ways of selecting k defective items and $n-k$ “good” items from a subsample of M and $N-M$ items respectively (favorable ways). But there are a total of $\binom{N}{n}$ ways of selecting n items among N items. This gives

$$P(X = k) = \frac{\binom{M}{k} \binom{N-M}{n-k}}{\binom{N}{n}},$$

since $0 \leq k \leq M$ and $n-k \leq N-M$, $n-k \geq 0$, i.e. $0 \leq k \leq M$, $k \leq n$, $k \geq n + M - N$.

(c) From (b)

$$\begin{aligned} P(X = k) &= \frac{M!}{k!(M-k)!} \frac{(N-M)!}{(n-k)!(N-M-n+k)!} \frac{n!(N-n)!}{N!} \\ &= \binom{n}{k} \frac{M(M-1)\cdots(M-k+1)}{N(N-1)\cdots(N-k+1)} \frac{(N-M)(N-M-1)\cdots(N-M-n+k+1)}{(N-k)(N-k-1)\cdots(N-n+1)} \frac{1}{\binom{N}{n}} \\ &\simeq \binom{n}{k} \left(\frac{M}{N}\right)^k \left(\frac{N-M}{N}\right)^{n-k} = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = 0, 1, 2, \dots, n \end{aligned}$$

since $N \rightarrow \infty$, $M \rightarrow \infty$ such that $M/N \rightarrow p$, and $n \ll N$. Thus

$$P(X = k) \rightarrow \text{Binomial}(n, p = M/N)$$

under the above conditions.

5.52 (a) Refer to discussions in problem 5.51 (a) if sampling is done with replacement, then

$$p = \frac{n}{n+m}$$

represents the probability of selecting a white marble on any trial. The event " $X = k$ " is given by " $r - 1$ white marbles among the first $k - 1$ trials" followed by "a white marble at the k^{th} trial". But from problem 5.51 (a), the event $r - 1$ white marbles among the first $k - 1$ trials has a binomial distribution whose probability is given by $\binom{k-1}{r-1} p^{r-1} q^{k-r}$. Thus

$$P(X = k) = \binom{k-1}{r-1} p^{r-1} q^{k-r} p = \binom{k-1}{r-1} p^r q^{k-r}, \quad k = r, r+1, \dots$$

which represents the Negative-binomial distribution

(b) If sampling is done with replacement, then the favorable ways of choosing the white balls are given by:

(i) $\binom{k-1}{r-1}$ ways of selecting $r - 1$ white balls among the first $k - 1$ trials/balls.

(ii) One way of selecting (the r^{th}) white ball at the k^{th} trial

(iii) $\binom{m+n-k}{n-r}$ ways of selecting the remaining $n - r$ white balls among the remaining $m + n - k$ balls.

This gives $\binom{k-1}{r-1} \cdot 1 \cdot \binom{m+n-k}{n-r}$ to be the total number of favorable ways of selecting the white balls. Since there are $n + m$ balls there are a total of $\binom{n+m}{n}$ ways of selecting n white balls. This gives

$$P(X = k) = \binom{k-1}{r-1} \frac{\binom{m+n-k}{n-r}}{\binom{n+m}{n}}, \quad k = r, r+1, \dots$$

(c) From (b)

$$\begin{aligned} P(X = k) &= \binom{k-1}{r-1} \frac{(m+n-k)!}{(n-r)!(m-k+r)!} \frac{n!m!}{(m+n)!} \\ &= \binom{k-1}{r-1} \left(\frac{n}{m+n}\right) \left(\frac{n-1}{m+n-1}\right) \cdots \left(\frac{n-r+1}{m+n-r+1}\right) \left(\frac{m!(m+n-k)!}{(m+n-r)!(m-k+r)!}\right) \\ &\simeq \binom{k-1}{r-1} \left(\frac{n}{m+n}\right)^r \left(\frac{m}{m+n-r}\right) \left(\frac{m-1}{m+n-r-1}\right) \cdots \left(\frac{m-k+r+1}{m+n-k+1}\right) \\ &\simeq \binom{k-1}{r-1} \left(\frac{n}{m+n}\right)^r \left(\frac{m}{m+n}\right)^{k-r} \text{ as } m+n \rightarrow \infty \\ &= \binom{k-1}{r-1} p^r q^{k-r}, \quad k = r, r+1, \dots, \quad q = 1-p \end{aligned}$$

$$\sim NB(r, p = n/(n+m)).$$