

**4.69** Let  $Y$  = magnitude of earthquake.  $Y$  is exponential with  $\beta = 2.4$ .

a.  $P(Y > 3) = \int_3^{\infty} \left(\frac{1}{2.4}\right) e^{-y/2.4} dy = -e^{-y/2.4} \Big|_3^{\infty} = e^{-3/2.4} = .2865$

b.  $P(2 < Y < 3) = \int_2^3 \left(\frac{1}{2.4}\right) e^{-y/2.4} dy = -e^{-y/2.4} \Big|_2^3 = .1481$

**4.71 a.** Let  $Y$  = demand for water.  $Y$  is exponential with  $\beta = 100$ .

$$P(Y > 200) = \int_{200}^{\infty} \left(\frac{1}{100}\right) e^{-y/100} dy = -e^{-y/100} \Big|_{200}^{\infty} = .1353$$

b. Let  $C$  = capacity.  $P(Y > c) = \int_c^{\infty} \left(\frac{1}{100}\right) e^{-y/100} dy = -e^{-y/100} \Big|_c^{\infty} = e^{-c/100} = .01$ .

so that  $c = -100 \ln(0.01) = 460.52$  cfs.

**4.73 a.**  $P(Y \leq 31) = \int_0^{31} \left(\frac{1}{44}\right) e^{-y/44} dy = 1 - e^{-31/44} = .5057$

b.  $V(Y) = \beta^2 = (44)^2 = 1936$

**4.74 a.**  $P(Y > 9) = \int_9^{\infty} \left(\frac{1}{3.6}\right) e^{-y/3.6} dy = -e^{-y/3.6} \Big|_9^{\infty} = e^{-9/3.6} = .0821$ .

b.  $P(Y > 9) = \int_9^{\infty} \left(\frac{1}{2.5}\right) e^{-y/2.5} dy = -e^{-y/2.5} \Big|_9^{\infty} = e^{-9/2.5} = .0273$ .

**4.87** From Theorem 4.8, a gamma-type random variable with  $\alpha = 3$  and  $\beta = 2$  has

$$E(Y) = \alpha\beta = 6 \text{ and } E(Y^2) = V(Y) + (E(Y))^2 = \alpha(\alpha + 1)\beta^2 = 3(4)(4) = 48. \text{ Hence}$$

$$E(L) = 30E(Y) + 2E(Y^2) = 30(6) + 2(48) = 276$$

Since

$$V(L) = E(L^2) - [E(L)]^2 = E(L^2) - 76,176 = E(900Y^2 + 120Y^3 + 4Y^4) - 76,176$$

we need the third and fourth moments about the origin.

$$E(Y^3) = \int_0^{\infty} \frac{y^5 e^{-y/2}}{\Gamma(3)2^3} dy = \frac{\Gamma(6)2^3}{\Gamma(3)} \int_0^{\infty} \frac{y^5 e^{-y/2}}{\Gamma(6)2^6} dy = \frac{5!2^3}{2!} = 480$$

$$E(Y^4) = \int_0^{\infty} \frac{y^6 e^{-y/2}}{\Gamma(3)2^3} dy = \frac{\Gamma(7)2^4}{\Gamma(3)} \int_0^{\infty} \frac{y^6 e^{-y/2}}{\Gamma(7)2^7} dy = \frac{6!2^4}{2!} = 5760$$

$$\text{Then } V(L) = 900(48) + 120(480) + 4(5760) - 76,176 = 47,664.$$

**4.88** Since  $Y$  has a gamma distribution with  $\alpha = 3$ ,  $\beta = \frac{1}{2}$ ,

$$E(Y) = \alpha\beta = \frac{3}{2} \quad \text{and} \quad V(Y) = \alpha\beta^2 = 3\left(\frac{1}{4}\right) = \frac{3}{4}$$

PROBLEM SET 7

4.104 a. Refer to Example 4.13 with  $\beta = \theta$ ,  $\alpha = 1$ . The mgf for  $y$  is  $m(t) = \frac{1}{(1-\beta t)^\alpha}$   
 $= \frac{1}{(1-\theta t)}$ .

b. Differentiating with respect to  $t$ ,

$$E(Y) = m'(0) = \frac{\theta}{(1-\theta t)^2} \Big|_{t=0} = \theta \quad \text{and} \quad E(Y^2) = m''(0) = \frac{2\theta^2}{(1-\theta t)^3} \Big|_{t=0} = 2\theta^2$$

so that  $V(Y) = 2\theta^2 - \theta^2 = \theta^2$ .

4.106 a. From Example 4.16,  $m_{Y-\mu}(t) = E[e^{t(Y-\mu)}] = e^{t^2\sigma^2/2}$ . Since

$$E[e^{t(Y-\mu)}] = E(e^{tY-\mu t}) = e^{-\mu t} E(e^{tY})$$

we have  $e^{t^2\sigma^2/2} = e^{-\mu t} E(e^{tY})$ , so the mgf for  $Y$  must be  $m_Y(t) = E(e^{tY})$   
 $= e^{\mu t + (t^2\sigma^2/2)}$ .

b. Differentiating with respect to  $t$ ,

$$EY = m'(0) = (\mu + t\sigma^2) e^{\mu t + (t^2\sigma^2/2)} \Big|_{t=0} = \mu$$

Then  $V(Y) = \mu^2 + \sigma^2 - \mu^2 = \sigma^2$ .

4.114 The interval must include 90% of all mileage on tires he sells. Using Tchebysheff's theorem, we must have

$$P(|Y - \mu| \leq k\sigma) \geq .90 = 1 - \frac{1}{k^2}.$$

Then  $k = \sqrt{\frac{1}{1-.90}} = \sqrt{10} = 3.1622$ . The necessary interval is then

$$|Y - 25,000| \leq 3.1622(4000) \quad \text{or} \quad 12,351 \leq Y \leq 37,649$$

4.115 It is necessary to have  $P(|Y - \mu| \leq 1) \geq .75$ . Hence,

$1 - \frac{1}{k^2} = .75$  and  $k = 2$ . According to Tchebysheff's inequality, then,  $1 = k\sigma$  and  
 $\sigma = \frac{1}{k} = \frac{1}{2}$ .

4.117 From Table A2.2, Appendix II, we find that when  $Y$  is uniform over  $(\theta_1, \theta_2)$

$$E(Y) = \frac{\theta_1 + \theta_2}{2} \quad V(Y) = \frac{(\theta_2 - \theta_1)^2}{12}$$

Thus,

$$2\sigma = 2\sqrt{V(Y)} = \frac{2(\theta_2 - \theta_1)}{\sqrt{12}} = \frac{\theta_2 - \theta_1}{\sqrt{3}}$$

The probability of interest is  $P(|Y - \mu| \leq 2\sigma) = P(\mu - 2\sigma \leq Y \leq \mu + 2\sigma)$ . Now

$$\theta_2 - \frac{\theta_1 + \theta_2}{2} = \frac{\theta_2 - \theta_1}{2} < \frac{\theta_2 - \theta_1}{\sqrt{3}} \quad \text{and hence} \quad \theta_2 < \frac{\theta_1 + \theta_2}{2} + \frac{\theta_2 - \theta_1}{\sqrt{3}} = \mu + 2\sigma$$

Similarly,

$$\frac{\theta_1 + \theta_2}{2} - \theta_1 = \frac{\theta_2 - \theta_1}{2} < \frac{\theta_2 - \theta_1}{\sqrt{3}} \quad \text{so that} \quad \theta_1 > \frac{\theta_1 + \theta_2}{2} - \frac{\theta_2 - \theta_1}{\sqrt{3}} = \mu - 2\sigma$$

But  $\theta_1$  and  $\theta_2$  are the upper and lower limits on  $Y$ . Hence

$$P(\mu - 2\sigma \leq Y \leq \mu + 2\sigma) = P\left[\left(\frac{\theta_1 + \theta_2}{2} - \frac{\theta_2 - \theta_1}{\sqrt{3}}\right) \leq Y \leq \left(\frac{\theta_1 + \theta_2}{2} + \frac{\theta_2 - \theta_1}{\sqrt{3}}\right)\right]$$

$$= P(\theta_1 \leq Y \leq \theta_2) = 1$$

Tchebysheff's theorem is satisfied, but the approximation suggested by the empirical rule is inaccurate. This is because the probability distribution for the uniform random variable is far from mound-shaped.

4.130 The 3000 light bulbs utilized by the manufacturing plant comprise the entire population (i.e., this is not a sample from the population) whose length of life is normally distributed with mean  $\mu = 500$ , and standard deviation  $\sigma = 50$ . The objective is to find a particular value,  $y_0$ , so that  $P(Y \leq y_0) = .01$ . That is, only 1% of the bulbs will burn out before they are replaced at time  $y_0$ . Then

$$P(Y \leq y_0) = P(Z \leq z_0) = .01 \quad \text{when} \quad z_0 = \frac{y_0 - 500}{50}$$

From Table 4, Appendix III, the value of  $z$  corresponding to an area (in the left tail of the distribution) of .01 is  $z_0 = -2.327$  (see Figure 4.23). Solving for  $y_0$  corresponding to  $z_0 = -2.327$ , we obtain

$$-2.327 = \frac{y_0 - 500}{50}$$

$$-116.35 = y_0 - 500 \quad \text{or} \quad y_0 = 383.65$$

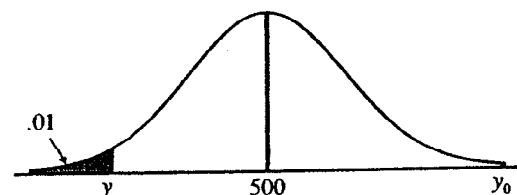


Figure 4.23

4.133a. The variable factor of  $f(y)$  is that of a gamma density with  $\alpha = 2$  and  $\beta = \frac{1}{2}$ . Hence

$$c = \frac{1}{\Gamma(\alpha)\beta^\alpha} = \frac{1}{\Gamma(2)(\frac{1}{2})^2} = \frac{4}{1!} = 4$$

b. Since  $Y$  has a gamma distribution with  $\alpha = 2$ ,  $\beta = \frac{1}{2}$ ,

$$E(Y) = \alpha\beta = 1 \quad V(Y) = \alpha\beta^2 = 2\left(\frac{1}{4}\right) = \frac{1}{2}$$

c. Recall that the moment-generating function of a gamma random variable is

$$m(t) = \frac{1}{(1 - \beta t)^\alpha} \quad \text{and in this case} \quad m(t) = \frac{1}{[1 - (\frac{1}{2})t]^2} = \left(1 - \frac{t}{2}\right)^{-2}$$

4.136 The probability distribution for  $n$ , the number of arrivals in time  $(0, t)$ , is Poisson with mean  $\lambda t$ , so that

$$P[n \text{ arrivals in } (0, t)] = \frac{(\lambda t)^n e^{-\lambda t}}{n!}$$

Let  $T$  be the length of time until the first arrival, and consider the distribution function for  $T$ .

$$F(t) = P(T \leq t) = 1 - P(T > t) = 1 - P[n = 0 \text{ in } (0, t)] = 1 - \frac{(\lambda t)^0 e^{-\lambda t}}{0!} = 1 - e^{-\lambda t}$$

Since  $T$  is a continuous random variable and  $F(t)$  is its distribution function, the density function for  $T$  may be found, using Definition 4.3, to be

$$f(t) = \frac{d}{dt} F(t) = \lambda e^{-\lambda t} \quad \text{for } t > 0$$

Note that the first derivative exists and is everywhere continuous, as is  $F(t)$  itself.

Hence Definitions 4.2 and 4.3 are satisfied. Moreover,  $f(t)$  is the density of an Exponential random variable with mean  $1/\lambda$ .

4.137 Let  $Y$  be the time between the arrival of two calls, measured in hours. We require  $P(Y > \frac{1}{4})$ . Since  $\lambda t = 10$  and  $t = 1$  (hour),  $\frac{1}{\lambda} = \frac{1}{10}$ , and

$$f(y) = \frac{1}{1} e^{-y/1} = 10e^{-10y}$$

and

$$P(Y > \frac{1}{4}) = \int_{1/4}^{\infty} 10e^{-10y} dy = -e^{-10y} \Big|_{1/4}^{\infty} = e^{-2.5} = .082$$

4.152 For  $m = 2$ ,

$$E(Y) = \int_0^{\infty} \frac{2y^2 e^{-y^2/\alpha}}{\alpha} dy$$

Let  $z = y^2$ . Then  $dz = 2y dy$  and

$$E(Y) = \int_0^{\infty} \frac{\sqrt{z} e^{-z/\alpha}}{\alpha} dz = \int_0^{\infty} \frac{z^{1/2} e^{-z/\alpha}}{\alpha} dz$$

which, when the proper constant is added, will be the integral of the density function of a gamma random variable with parameters  $\frac{3}{2}$  and  $\alpha$ . Then

$$E(Y) = \frac{\alpha^{3/2} \Gamma(\frac{3}{2})}{\alpha} \int_0^{\infty} \frac{z^{1/2} e^{-z/\alpha}}{\Gamma(\frac{3}{2}) \alpha^{3/2}} dz = \frac{\alpha^{3/2} \Gamma(\frac{3}{2})}{\alpha} = \alpha^{1/2} \Gamma(\frac{3}{2}) = \alpha^{1/2} \left(\frac{1}{2}\right) \Gamma\left(\frac{1}{2}\right) = \frac{(\alpha\pi)^{1/2}}{2}$$

where  $\Gamma(\frac{1}{2}) = \sqrt{\pi}$  is shown in exercise 4.162.

Again using the transformation  $z = y^2$ , we find

$$E(Y^2) = \int_0^{\infty} \frac{2y^3 e^{-y^2/\alpha}}{\alpha} dy = \int_0^{\infty} \frac{z e^{-z/\alpha}}{\alpha} dz = \frac{\Gamma(2)\alpha^2}{\alpha} \int_0^{\infty} \frac{z e^{-z/\alpha}}{\Gamma(2)\alpha^2} dz = \alpha$$

so that  $V(Y) = \alpha - [\alpha^{1/2} \Gamma(\frac{3}{2})]^2 = \alpha \{1 - [\Gamma(\frac{3}{2})]^2\} = \alpha [1 - \frac{\pi}{4}]$

See Exercise 4.162.