## CHAPTER 5

Solution to Problem 5.1. (a) We have  $\sigma_{M_n} = 1/\sqrt{n}$ , so in order that  $\sigma_{M_n} \leq 0.01$ , we must have  $n \geq 10,000$ .

(b) We want to have

$$\mathbf{P}(|M_n - h| \le 0.05) \ge 0.99.$$

Using the facts  $h = \mathbf{E}[M_n], \sigma_{M_n}^2 = 1/n$ , and the Chebyshev inequality, we have

$$\mathbf{P}(|M_n - h| \le 0.05) = \mathbf{P}(|M_n - \mathbf{E}[M_n]| \le 0.05)$$
  
= 1 -  $\mathbf{P}(|M_n - \mathbf{E}[M_n]| \ge 0.05)$   
 $\ge 1 - \frac{1/n}{(0.05)^2}.$ 

Thus, we must have

$$1 - \frac{1/n}{(0.05)^2} \ge 0.99,$$

which yields  $n \ge 40,000$ .

(c) Based on Example 5.3,  $\sigma_{X_i}^2 \leq (0.6)^2/4$ , so he should use 0.3 meters in place of 1.0 meters as the estimate of the standard deviation of the samples  $X_i$  in the calculations of parts (a) and (b). In the case of part (a), we have  $\sigma_{M_n} = 0.3/\sqrt{n}$ , so in order that  $\sigma_{M_n} \leq 0.01$ , we must have  $n \geq 900$ . In the case of part (b), we have  $\sigma_{M_n} = 0.3/\sqrt{n}$ , so in order that  $\sigma_{M_n} \leq 0.01$ , we must have  $n \geq 900$ . In the case of part (a), we must have  $n \geq 900$ . In the case of part (a), we must have  $\sigma_{M_n} = 0.3/\sqrt{n}$ , so in order that  $\sigma_{M_n} \leq 0.01$ , we must have  $n \geq 900$ . In the case of part (a), we must have

$$1 - \frac{0.09/n}{(0.05)^2} \ge 0.99,$$

which yields  $n \geq 3,600$ .

Solution to Problem 5.4. Proceeding as in Example 5.5, the best guarantee that can be obtained from the Chebyshev inequality is

$$\mathbf{P}(|M_n - f| \ge \epsilon) \le \frac{1}{4n\epsilon^2}.$$

(a) If  $\epsilon$  is reduced to half its original value, and in order to keep the bound  $1/(4n\epsilon^2)$  constant, the sample size *n* must be made four times larger.

(b) If the error probability  $\delta$  is to be reduced to  $\delta/2$ , while keeping  $\epsilon$  the same, the sample size has to be doubled.

**Solution to Problem 5.5.** In cases (a), (b), and (c), we show that  $Y_n$  converges to 0 in probability. In case (d), we show that  $Y_n$  converges to 1 in probability.

(a) For any  $\epsilon > 0$ , we have

$$\mathbf{P}\big(|Y_n| \ge \epsilon\big) = 0,$$

for all n with  $1/n < \epsilon$ , so  $\mathbf{P}(|Y_n| \ge \epsilon) \to 0$ .

(b) For all  $\epsilon \in (0, 1)$ , we have

$$\mathbf{P}(|Y_n| \ge \epsilon) = \mathbf{P}(|X_n|^n \ge \epsilon) = \mathbf{P}(X_n \ge \epsilon^{1/n}) + \mathbf{P}(X_n \le -\epsilon^{1/n}) = 1 - \epsilon^{1/n},$$

and the two terms in the right-hand side converge to 0, since  $\epsilon^{1/n} \to 1$ . (c) Since  $X_1, X_2, \ldots$  are independent random variables, we have

$$\mathbf{E}[Y_n] = \mathbf{E}[X_1] \cdots \mathbf{E}[X_n] = 0.$$

Also

$$\operatorname{var}(Y_n) = \mathbf{E}[Y_n^2] = \mathbf{E}[X_1^2] \cdots \mathbf{E}[X_n^2] = \operatorname{var}(X_1)^n = \left(\frac{4}{12}\right)^n$$

so  $var(Y_n) \to 0$ . Since all  $Y_n$  have 0 as a common mean, from Chebyshev's inequality it follows that  $Y_n$  converges to 0 in probability.

(d) We have for all  $\epsilon \in (0, 1)$ , using the independence of  $X_1, X_2, \ldots$ ,

$$\mathbf{P}(|Y_n - 1| \ge \epsilon) = \mathbf{P}(\max\{X_1, \dots, X_n\} \ge 1 + \epsilon) + \mathbf{P}(\max\{X_1, \dots, X_n\} \le 1 - \epsilon)$$
  
=  $\mathbf{P}(X_1 \le 1 - \epsilon, \dots, X_n \le 1 - \epsilon)$   
=  $(\mathbf{P}(X_1 \le 1 - \epsilon))^n$   
=  $(1 - \frac{\epsilon}{2})^n$ .

Hence  $\mathbf{P}(|Y_n - 1| \ge \epsilon) \to 0.$ 

Solution to Problem 5.8. Let S be the number of times that the result was odd, which is a binomial random variable, with parameters n = 100 and p = 0.5, so that  $\mathbf{E}[X] = 100 \cdot 0.5 = 50$  and  $\sigma_S = \sqrt{100 \cdot 0.5 \cdot 0.5} = \sqrt{25} = 5$ . Using the normal approximation to the binomial, we find

$$\mathbf{P}(S > 55) = \mathbf{P}\left(\frac{S - 50}{5} > \frac{55 - 50}{5}\right) \approx 1 - \Phi(1) = 1 - 0.8413 = 0.1587.$$

A better approximation can be obtained by using the de Moivre-Laplace approximation, which yields

$$\mathbf{P}(S > 55) = \mathbf{P}(S \ge 55.5) = \mathbf{P}\left(\frac{S - 50}{5} > \frac{55.5 - 50}{5}\right)$$
$$\approx 1 - \Phi(1.1) = 1 - 0.8643 = 0.1357.$$

**Solution to Problem 5.9.** (a) Let S be the number of crash-free days, which is a binomial random variable with parameters n = 50 and p = 0.95, so that  $\mathbf{E}[X] = 50 \cdot 0.95 = 47.5$  and  $\sigma_S = \sqrt{50 \cdot 0.95 \cdot 0.05} = 1.54$ . Using the normal approximation to the binomial, we find

$$\mathbf{P}(S \ge 45) = \mathbf{P}\left(\frac{S - 47.5}{1.54} \ge \frac{45 - 47.5}{1.54}\right) \approx 1 - \Phi(-1.62) = \Phi(1.62) = 0.9474.$$

A better approximation can be obtained by using the de Moivre-Laplace approximation, which yields

$$\mathbf{P}(S \ge 45) = \mathbf{P}(S > 44.5) = \mathbf{P}\left(\frac{S - 47.5}{1.54} \ge \frac{44.5 - 47.5}{1.54}\right)$$
$$\approx 1 - \Phi(-1.95) = \Phi(1.95) = 0.9744.$$

(b) The random variable S is binomial with parameter p = 0.95. However, the random variable 50 - S (the number of crashes) is also binomial with parameter p = 0.05. Since the Poisson approximation is exact in the limit of small p and large n, it will give more accurate results if applied to 50 - S. We will therefore approximate 50 - S by a Poisson random variable with parameter  $\lambda = 50 \cdot 0.05 = 2.5$ . Thus,

$$\mathbf{P}(S \ge 45) = \mathbf{P}(50 - S \le 5)$$
$$= \sum_{k=0}^{5} \mathbf{P}(n - S = k)$$
$$= \sum_{k=0}^{5} e^{-\lambda} \frac{\lambda^{k}}{k!}$$
$$= 0.958.$$

It is instructive to compare with the exact probability which is

$$\sum_{k=0}^{5} \binom{50}{k} 0.05^k \cdot 0.95^{50-k} = 0.962.$$

Thus, the Poisson approximation is closer. This is consistent with the intuition that the normal approximation to the binomial works well when p is close to 0.5 or n is very large, which is not the case here. On the other hand, the calculations based on the normal approximation are generally less tedious.

**Solution to Problem 5.10.** (a) Let  $S_n = X_1 + \cdots + X_n$  be the total number of gadgets produced in n days. Note that the mean, variance, and standard deviation of  $S_n$  is 5n, 9n, and  $3\sqrt{n}$ , respectively. Thus,

$$\mathbf{P}(S_{100} < 440) = \mathbf{P}(S_{100} \le 439.5)$$
  
=  $\mathbf{P}\left(\frac{S_{100} - 500}{30} < \frac{439.5 - 500}{30}\right)$   
 $\approx \Phi\left(\frac{439.5 - 500}{30}\right)$   
=  $\Phi(-2.02)$   
=  $1 - \Phi(2.02)$   
=  $1 - \Phi(2.02)$   
=  $1 - 0.9783$   
=  $0.0217.$ 

(b) The requirement  $\mathbf{P}(S_n \ge 200 + 5n) \le 0.05$  translates to

$$\mathbf{P}\left(\frac{S_n - 5n}{3\sqrt{n}} \ge \frac{200}{3\sqrt{n}}\right) \le 0.05,$$

or, using a normal approximation,

$$1 - \Phi\left(\frac{200}{3\sqrt{n}}\right) \le 0.05,$$

and

$$\Phi\left(\frac{200}{3\sqrt{n}}\right) \ge 0.95.$$

From the normal tables, we obtain  $\Phi(1.65) \approx 0.95$ , and therefore,

$$\frac{200}{3\sqrt{n}} \ge 1.65,$$

which finally yields  $n \leq 1632$ .

(c) The event  $N \ge 220$  (it takes at least 220 days to exceed 1000 gadgets) is the same as the event  $S_{219} \le 1000$  (no more than 1000 gadgets produced in the first 219 days). Thus,  $\mathbf{P}(N \ge 200) = \mathbf{P}(G = \le 1000)$ 

$$\mathbf{P}(N \ge 220) = \mathbf{P}(S_{219} \le 1000)$$
  
=  $\mathbf{P}\left(\frac{S_{219} - 5 \cdot 219}{3\sqrt{219}} \le \frac{1000 - 5 \cdot 219}{3\sqrt{219}}\right)$   
=  $1 - \Phi(2.14)$   
=  $1 - 0.9838$   
=  $0.0162$ .

Solution to Problem 5.11. Note that W is the sample mean of 16 independent identically distributed random variables of the form  $X_i - Y_i$ , and a normal approximation is appropriate. The random variables  $X_i - Y_i$  have zero mean, and variance equal to 2/12. Therefore, the mean of W is zero, and its variance is (2/12)/16 = 1/96. Thus,

$$\mathbf{P}(|W| < 0.001) = \mathbf{P}\left(\frac{|W|}{\sqrt{1/96}} < \frac{0.001}{\sqrt{1/96}}\right) \approx \Phi(0.001\sqrt{96}) - \Phi(-0.001\sqrt{96})$$
$$= 2\Phi(0.001\sqrt{96}) - 1 = 2\Phi(0.0098) - 1 \approx 2 \cdot 0.504 - 1 = 0.008.$$

Let us also point out a somewhat different approach that bypasses the need for the normal table. Let Z be a normal random variable with zero mean and standard deviation equal to  $1/\sqrt{96}$ . The standard deviation of Z, which is about 0.1, is much larger than 0.001. Thus, within the interval [-0.001, 0.001], the PDF of Z is approximately constant. Using the formula  $\mathbf{P}(z - \delta \leq Z \leq z + \delta) \approx f_Z(z) \cdot 2\delta$ , with z = 0 and  $\delta = 0.001$ , we obtain

$$\mathbf{P}(|W| < 0.001) \approx \mathbf{P}(-0.001 \le Z \le 0.001) \approx f_Z(0) \cdot 0.002 = \frac{0.002}{\sqrt{2\pi}(1/\sqrt{96})} = 0.0078.$$