7.2.1 (a) bias($\hat{\mu}_1$) = 0

The point estimate $\hat{\mu}_1$ is unbiased.

$$bias(\hat{\mu}_2) = 0$$

The point estimate $\hat{\mu}_2$ is unbiased.

$$\operatorname{bias}(\hat{\mu}_3) = 9 - \frac{\mu}{2}$$

(b) $Var(\hat{\mu}_1) = 6.2500$

$$Var(\hat{\mu}_2) = 9.0625$$

$$Var(\hat{\mu}_3) = 1.9444$$

The point estimate $\hat{\mu}_3$ has the smallest variance.

(c) $MSE(\hat{\mu}_1) = 6.2500$

$$MSE(\hat{\mu}_2) = 9.0625$$

$$MSE(\hat{\mu}_3) = 1.9444 + (9 - \frac{\mu}{2})^2$$

This is equal to 26.9444 when $\mu = 8$.

7.2.2 (a) $bias(\hat{\mu}_1) = 0$

$$bias(\hat{\mu}_2) = -0.217\mu$$

$$\operatorname{bias}(\hat{\mu}_3) = 2 - \frac{\mu}{4}$$

The point estimate $\hat{\mu}_1$ is unbiased.

(b) $Var(\hat{\mu}_1) = 4.444$

$$Var(\hat{\mu}_2) = 2.682$$

$$Var(\hat{\mu}_3) = 2.889$$

The point estimate $\hat{\mu}_2$ has the smallest variance.

(c) $MSE(\hat{\mu}_1) = 4.444$

$$MSE(\hat{\mu}_2) = 2.682 + 0.0469\mu^2$$

This is equal to 3.104 when $\mu = 3$.

$$MSE(\hat{\mu}_3) = 2.889 + (2 - \frac{\mu}{4})^2$$

This is equal to 4.452 when $\mu = 3$.

- 7.2.3 (a) $Var(\hat{\mu}_1) = 2.5$
 - (b) The value p = 0.6 produces the smallest variance which is $Var(\hat{\mu}) = 2.4$.
 - (c) The relative efficiency is $\frac{2.4}{2.5} = 0.96$.
- 7.2.4 (a) $Var(\hat{\mu}_1) = 2$
 - (b) The value p = 0.875 produces the smallest variance which is $Var(\hat{\mu}) = 0.875$.
 - (c) The relative efficiency is $\frac{0.875}{2} = 0.4375$.
- 7.2.5 (a) $a_1 + \ldots + a_n = 1$
 - (b) $a_1 = \ldots = a_n = \frac{1}{n}$
- 7.2.6 $MSE(\hat{\theta}_1) = 0.02 \ \theta^2 + (0.13 \ \theta)^2 = 0.0369 \ \theta^2$

$$\mathrm{MSE}(\hat{\theta}_2) = 0.07 \; \theta^2 + (0.05 \; \theta)^2 = 0.0725 \; \theta^2$$

$$MSE(\hat{\theta}_3) = 0.005 \ \theta^2 + (0.24 \ \theta)^2 = 0.0626 \ \theta^2$$

The point estimate $\hat{\theta}_1$ has the smallest mean square error.

7.2.7 bias($\hat{\mu}$) = $\frac{\mu_0 - \mu}{2}$

$$\mathrm{Var}(\hat{\mu}) = \frac{\sigma^2}{4}$$

$$MSE(\hat{\mu}) = \frac{\sigma^2}{4} + \frac{(\mu_0 - \mu)^2}{4}$$

$$\mathrm{MSE}(X) = \sigma^2$$

7.2.8 (a) bias(
$$\hat{p}$$
) = $-\frac{p}{11}$

(b)
$$Var(\hat{p}) = \frac{10 p (1-p)}{121}$$

(c)
$$MSE(\hat{p}) = \frac{10 p (1-p)}{121} + (\frac{p}{11})^2 = \frac{10p-9p^2}{121}$$

(d) bias
$$\left(\frac{X}{10}\right) = 0$$

$$\operatorname{Var}\left(\frac{X}{10}\right) = \frac{p(1-p)}{10}$$

$$MSE\left(\frac{X}{10}\right) = \frac{p(1-p)}{10}$$

7.2.9
$$\operatorname{Var}\left(\frac{X_1+X_2}{2}\right)$$

$$= \frac{\operatorname{Var}(X_1) + \operatorname{Var}(X_2)}{4}$$

$$= \frac{5.39^2 + 9.43^2}{4}$$

$$= 29.49$$

The standard deviation is $\sqrt{29.49} = 5.43$.

7.3.1
$$\operatorname{Var}\left(\frac{X_1}{n_1}\right) = \frac{p(1-p)}{n_1}$$

$$\operatorname{Var}\left(\frac{X_2}{n_2}\right) = \frac{p(1-p)}{n_2}$$

The relative efficiency is the ratio of these two variances which is $\frac{n_1}{n_2}$.

7.3.2 (a)
$$P\left(\left|N\left(0, \frac{1}{10}\right)\right| \le 0.3\right) = 0.6572$$

(b)
$$P(|N(0, \frac{1}{30})| \le 0.3) = 0.8996$$

7.3.3 (a)
$$P\left(\left|N\left(0, \frac{7}{15}\right)\right| \le 0.4\right) = 0.4418$$

(b)
$$P\left(\left|N\left(0, \frac{7}{50}\right)\right| \le 0.4\right) = 0.7150$$

7.3.4 (a) Solving

$$P\left(5 \times \frac{\chi_{30}^2}{30} \le c\right) = P(\chi_{30}^2 \le 6c) = 0.90$$
 gives $c = 6.709$.

(b) Solving

$$P\left(5 \times \frac{\chi_{30}^2}{30} \le c\right) = P(\chi_{30}^2 \le 6c) = 0.95$$
 gives $c = 7.296$.

7.3.5 (a) Solving

$$P\left(32 \times \frac{\chi_{20}^2}{20} \le c\right) = P\left(\chi_{20}^2 \le \frac{5c}{8}\right) = 0.90$$
 gives $c = 45.46$.

(b) Solving

$$P\left(32 \times \frac{\chi_{20}^2}{20} \le c\right) = P\left(\chi_{20}^2 \le \frac{5c}{8}\right) = 0.95$$
 gives $c = 50.26$.

7.3.6 (a) Solving

$$P(|t_{15}| \le c) = 0.95$$

gives $c = t_{0.025,15} = 2.131$.

(b) Solving

$$P(|t_{15}| \le c) = 0.99$$

gives $c = t_{0.005,15} = 2.947$.

$$P\left(\frac{|t_{20}|}{\sqrt{21}} \le c\right) = 0.95$$

gives $c = \frac{t_{0.025,20}}{\sqrt{21}} = 0.4552$.

$$P\left(\frac{|t_{20}|}{\sqrt{21}} \le c\right) = 0.99$$

gives $c = \frac{t_{0.005,20}}{\sqrt{21}} = 0.6209$.

7.3.8
$$\hat{p} = \frac{234}{450} = 0.52$$

s.e.
$$(\hat{p}) = \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} = \sqrt{\frac{0.52 \times 0.48}{450}} = 0.0236$$

7.3.9
$$\hat{\mu} = \bar{x} = 974.3$$

s.e.
$$(\hat{\mu}) = \frac{s}{\sqrt{n}} = \sqrt{\frac{452.1}{35}} = 3.594$$

$$7.3.10 \quad \hat{p} = \frac{24}{120} = 0.2$$

s.e.
$$(\hat{p}) = \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} = \sqrt{\frac{0.2 \times 0.8}{120}} = 0.0365$$

7.3.11
$$\hat{p} = \frac{33}{150} = 0.22$$

s.e.
$$(\hat{p}) = \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} = \sqrt{\frac{0.22 \times 0.78}{150}} = 0.0338$$

7.3.12
$$\hat{p} = \frac{26}{80} = 0.325$$

s.e.
$$(\hat{p}) = \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} = \sqrt{\frac{0.325 \times 0.675}{80}} = 0.0524$$

7.3.13
$$\hat{\mu} = \bar{x} = 69.35$$

s.e.
$$(\hat{\mu}) = \frac{s}{\sqrt{n}} = \frac{17.59}{\sqrt{200}} = 1.244$$

7.3.14
$$\hat{\mu} = \bar{x} = 3.291$$

s.e.
$$(\hat{\mu}) = \frac{s}{\sqrt{n}} = \frac{3.794}{\sqrt{55}} = 0.512$$

7.3.15
$$\hat{\mu} = \bar{x} = 12.211$$

s.e.
$$(\hat{\mu}) = \frac{s}{\sqrt{n}} = \frac{2.629}{\sqrt{90}} = 0.277$$

7.3.16
$$\hat{\mu} = \bar{x} = 1.1106$$

s.e.
$$(\hat{\mu}) = \frac{s}{\sqrt{n}} = \frac{0.0530}{\sqrt{125}} = 0.00474$$

7.3.17
$$\hat{\mu} = \bar{x} = 0.23181$$

s.e.
$$(\hat{\mu}) = \frac{s}{\sqrt{n}} = \frac{0.07016}{\sqrt{75}} = 0.00810$$

7.3.18
$$\hat{\mu} = \bar{x} = 9.2294$$

s.e.
$$(\hat{\mu}) = \frac{s}{\sqrt{n}} = \frac{0.8423}{\sqrt{80}} = 0.0942$$

7.3.19 If a sample of size n = 100 is used, then the probability is

$$P(0.24 - 0.05 \le \hat{p} \le 0.24 + 0.05) = P(19 \le B(100, 0.24) \le 29).$$

Using a normal approximation this can be estimated as

$$\Phi\left(\frac{29+0.5-100\times0.24}{\sqrt{100\times0.24\times0.76}}\right) - \Phi\left(\frac{19-0.5-100\times0.24}{\sqrt{100\times0.24\times0.76}}\right)$$

$$= \Phi(1.288) - \Phi(-1.288) = 0.8022.$$

If a sample of size n=200 is used, then the probability is

$$P(38 \le B(200, 0.24) \le 58).$$

Using a normal approximation this can be estimated as

$$\Phi\left(\frac{58+0.5-200\times0.24}{\sqrt{200\times0.24\times0.76}}\right) - \Phi\left(\frac{38-0.5-200\times0.24}{\sqrt{200\times0.24\times0.76}}\right)$$

$$= \Phi(1.738) - \Phi(-1.738) = 0.9178.$$

7.3.20
$$P(173 \le \hat{\mu} \le 175) = P(173 \le \bar{X} \le 175)$$
 where

$$\bar{X} \sim N\left(174, \frac{2.8^2}{30}\right).$$

This is

$$\Phi\left(\frac{175-174}{\sqrt{2.8^2/30}}\right) - \Phi\left(\frac{173-174}{\sqrt{2.8^2/30}}\right)$$

$$= \Phi(1.956) - \Phi(-1.956) = 0.9496.$$

7.3.21
$$P(0.62 \le \hat{p} \le 0.64)$$

$$= P(300 \times 0.62 \le B(300, 0.63) \le 300 \times 0.64)$$

$$\simeq P(185.5 \le N(300 \times 0.63, 300 \times 0.63 \times 0.37) \le 192.5)$$

$$= P\left(\frac{185.5 - 189}{\sqrt{69.93}} \le N(0, 1) \le \frac{192.5 - 189}{\sqrt{69.93}}\right)$$

$$= \Phi(0.419) - \Phi(-0.419) = 0.324$$

7.3.22
$$P\left(109.9 \le N\left(110.0, \frac{0.4^2}{22}\right) \le 110.1\right)$$

$$= P\left(\frac{\sqrt{22}(109.9 - 110.0)}{0.4} \le N(0, 1) \le \frac{\sqrt{22}(110.1 - 110.0)}{0.4}\right)$$

$$= \Phi(1.173) - \Phi(-1.173) = 0.759$$

$$7.3.23 \quad \sqrt{\frac{0.126 \times 0.874}{360}} = 0.017$$

7.3.24
$$P\left(N\left(341, \frac{2^2}{20}\right) \le 341.5\right)$$

$$= P\left(N(0.1) \le \frac{\sqrt{20} \times (341.5 - 341)}{2}\right)$$

$$=\Phi(1.118)=0.547$$

7.3.25
$$P\left(\mu - 2 \le N\left(\mu, \frac{5.2^2}{18}\right) \le \mu + 2\right)$$

$$= P\left(\frac{-\sqrt{18}\times2}{5.2} \le N(0.1) \le \frac{\sqrt{18}\times2}{5.2}\right)$$

$$= \Phi(1.632) - \Phi(-1.632) = 0.103$$

7.3.26 The largest standard error is obtained when $\hat{p}=0.5$ and is equal to $\sqrt{\frac{0.5\times0.5}{1400}}=0.0134.$

7.3.27
$$P(X \ge 60) = e^{-0.02 \times 60} = 0.301$$

Let Y be the number of components that last longer than one hour.

$$P\left(0.301 - 0.05 \le \frac{Y}{110} \le 0.301 + 0.05\right)$$

$$= P(27.6 \le Y \le 38.6)$$

$$= P(28 \le B(110, 0.301) \le 38)$$

$$\simeq P(27.5 \le N(110 \times 0.301, 110 \times 0.301 \times 0.699) \le 38.5)$$

$$= P\left(\frac{27.5 - 33.11}{\sqrt{23.14}} \le N(0, 1) \le \frac{38.5 - 33.11}{\sqrt{23.14}}\right)$$

$$= \Phi(1.120) - \Phi(-1.166)$$

$$= 0.869 - 0.122 = 0.747$$

7.3.28 (a) $P(\mu - 0.5 \le \bar{X} \le \mu + 0.5)$ = $P\left(\mu - 0.5 \le N\left(\mu, \frac{0.82^2}{5}\right) \le \mu + 0.5\right)$

$$=\Phi\left(\frac{0.5\sqrt{5}}{0.82}\right) - \Phi\left(\frac{-0.5\sqrt{5}}{0.82}\right) = 0.827$$

(b)
$$P(\mu - 0.5 \le \bar{X} \le \mu + 0.5)$$

$$= P\left(\mu - 0.5 \leq N\left(\mu, \tfrac{0.82^2}{10}\right) \leq \mu + 0.5\right)$$

$$=\Phi\left(\frac{0.5\sqrt{10}}{0.82}\right)-\Phi\left(\frac{-0.5\sqrt{10}}{0.82}\right)=0.946$$

(c) In order for

$$P\left(\mu - 0.5 \le N\left(\mu, \frac{0.82^2}{n}\right) \le \mu + 0.5\right)$$

$$=\Phi\left(\frac{0.5\sqrt{n}}{0.82}\right) - \Phi\left(\frac{-0.5\sqrt{n}}{0.82}\right) \ge 0.99$$

it is necessary that

$$\frac{0.5\sqrt{n}}{0.82} \ge z_{0.005} = 2.576$$

which is satisfied for a sample size n of at least 18.

7.3.29 (a)
$$p = \frac{592}{3288} = 0.18$$

$$P(p - 0.1 \le \hat{p} \le p + 0.1)$$

$$= P\left(0.08 \le \frac{X}{20} \le 0.28\right)$$

$$= P(1.6 \le X \le 5.6)$$
 where $X \sim B(20, 0.18)$.

This probability is

$$\begin{split} &P(X=2) + P(X=3) + P(X=4) + P(X=5) \\ &= \binom{20}{2} \times 0.18^2 \times 0.82^{18} + \binom{20}{3} \times 0.18^3 \times 0.82^{17} \\ &+ \binom{20}{4} \times 0.18^4 \times 0.82^{16} + \binom{20}{5} \times 0.18^5 \times 0.82^{15} \\ &= 0.7626. \end{split}$$

(b) The probability that a sampled meter is operating outside the acceptable tolerance limits is now

$$p^* = \frac{184}{2012} = 0.09.$$

$$P(p - 0.1 \le \hat{p} \le p + 0.1)$$

$$= P\left(0.08 \le \frac{Y}{20} \le 0.28\right)$$

$$= P(1.6 \le Y \le 5.6)$$
where $Y \sim B(20, 0.09)$.

This probability is

$$\begin{split} &P(Y=2) + P(Y=3) + P(Y=4) + P(Y=5) \\ &= \binom{20}{2} \times 0.09^2 \times 0.91^{18} + \binom{20}{3} \times 0.09^3 \times 0.91^{17} \\ &+ \binom{20}{4} \times 0.09^4 \times 0.91^{16} + \binom{20}{5} \times 0.09^5 \times 0.91^{15} \\ &= 0.5416. \end{split}$$

- 7.3.30 D
- 7.3.31 D
- 7.3.32 D
- 7.3.33 B
- 7.3.34 D
- 7.3.35 A
- 7.4.1 $\hat{\lambda} = \bar{x} = 5.63$

s.e.
$$(\hat{\lambda}) = \sqrt{\frac{\hat{\lambda}}{n}} = \sqrt{\frac{5.63}{23}} = 0.495$$

7.4.2 Using the method of moments the point estimates \hat{a} and \hat{b} are the solutions to the equations

$$\frac{a}{a+b} = 0.782$$

and

$$\frac{ab}{(a+b)^2(a+b+1)} = 0.0083$$

which are $\hat{a} = 15.28$ and $\hat{b} = 4.26$.

7.4.3 Using the method of moments

$$E(X) = \frac{1}{\lambda} = \bar{x}$$

which gives $\hat{\lambda} = \frac{1}{\bar{x}}$.

The likelihood is

$$L(x_1, \dots, x_n, \lambda) = \lambda^n e^{-\lambda(x_1 + \dots + x_n)}$$

which is maximized at $\hat{\lambda} = \frac{1}{\bar{x}}$.

7.4.4 $\hat{p}_i = \frac{x_i}{n}$ for $1 \le i \le n$

7.4.5 Using the method of moments

$$E(X) = \frac{5}{\lambda} = \bar{x}$$

which gives $\hat{\lambda} = \frac{5}{\bar{x}}$.

The likelihood is

$$L(x_1, \dots, x_n, \lambda) = \left(\frac{1}{24}\right)^n \times \lambda^{5n} \times x_1^4 \times \dots \times x_n^4 \times e^{-\lambda(x_1 + \dots + x_n)}$$

which is maximized at $\hat{\lambda} = \frac{5}{\bar{x}}$.

7.7.1 bias(
$$\hat{\mu}_1$$
) = 5 - $\frac{\mu}{2}$

$$\operatorname{bias}(\hat{\mu}_2) = 0$$

$$\operatorname{Var}(\hat{\mu}_1) = \frac{1}{8}$$

$$\operatorname{Var}(\hat{\mu}_2) = \frac{1}{2}$$

$$MSE(\hat{\mu}_1) = \frac{1}{8} + (5 - \frac{\mu}{2})^2$$

$$MSE(\hat{\mu}_2) = \frac{1}{2}$$

7.7.2 (a) bias(
$$\hat{p}$$
) = $-\frac{p}{7}$

(b)
$$Var(\hat{p}) = \frac{3p(1-p)}{49}$$

(c)
$$MSE(\hat{p}) = \frac{3p(1-p)}{49} + (\frac{p}{7})^2 = \frac{3p-2p^2}{49}$$

(d)
$$MSE\left(\frac{X}{12}\right) = \frac{p(1-p)}{12}$$

7.7.3 (a)
$$F(t) = P(T \le t) = P(X_1 \le t) \times \ldots \times P(X_n \le t)$$

 $= \frac{t}{\theta} \times \ldots \times \frac{t}{\theta} = (\frac{t}{\theta})^n$
for $0 \le t \le \theta$

(b)
$$f(t) = \frac{dF(t)}{dt} = n \frac{t^{n-1}}{\theta^n}$$

for $0 \le t \le \theta$

(c) Notice that
$$E(T) = \int_0^\theta t f(t) dt = \frac{n}{n+1}\theta$$
 so that $E(\hat{\theta}) = \theta$.

(d) Notice that
$$E(T^2) = \int_0^\theta \ t^2 \ f(t) \ dt = \frac{n}{n+2} \theta^2$$
 so that
$$\operatorname{Var}(T) = \frac{n}{n+2} \theta^2 - \left(\frac{n}{n+1}\theta\right)^2$$

$$\operatorname{Var}(T) = \frac{n}{n+2}\theta^2 - \left(\frac{n}{n+1}\theta\right)^2$$
$$= \frac{n\theta^2}{(n+2)(n+1)^2}.$$

Consequently,

$$\operatorname{Var}(\hat{\theta}) = \frac{(n+1)^2}{n^2} \operatorname{Var}(T) = \frac{\theta^2}{n(n+2)}$$
 and

s.e.
$$(\hat{\theta}) = \frac{\hat{\theta}}{\sqrt{n(n+2)}}$$
.

(e)
$$\hat{\theta} = \frac{11}{10} \times 7.3 = 8.03$$

s.e. $(\hat{\theta}) = \frac{8.03}{\sqrt{10 \times 12}} = 0.733$

7.7.4 Recall that
$$f(x_i, \theta) = \frac{1}{\theta}$$
 for $0 \le x_i \le \theta$ (and $f(x_i, \theta) = 0$ elsewhere) so that the likelihood is $\frac{1}{\theta^n}$ as long as $x_i \le \theta$ for $1 \le i \le n$ and is equal to zero otherwise.

$$\operatorname{bias}(\hat{\theta}) = -\frac{\theta}{n+1}$$

7.7.5 Using the method of moments

$$E(X) = \frac{1}{p} = \bar{x}$$

which gives $\hat{p} = \frac{1}{\bar{x}}$.

The likelihood is

$$L(x_1, ..., x_n, \lambda) = p^n (1-p)^{x_1 + ... + x_n - n}$$

which is maximizized at $\hat{p} = \frac{1}{\bar{x}}$.

7.7.6
$$\hat{p} = \frac{35}{100} = 0.35$$

s.e.
$$(\hat{p}) = \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} = \sqrt{\frac{0.35 \times 0.65}{100}} = 0.0477$$

7.7.7
$$\hat{\mu} = \bar{x} = 17.79$$

s.e.
$$(\bar{x}) = \frac{s}{\sqrt{n}} = \frac{6.16}{\sqrt{24}} = 1.26$$

7.7.8
$$\hat{\mu} = \bar{x} = 1.633$$

s.e.
$$(\bar{x}) = \frac{s}{\sqrt{n}} = \frac{0.999}{\sqrt{30}} = 0.182$$

7.7.9
$$\hat{\mu} = \bar{x} = 69.618$$

s.e.
$$(\bar{x}) = \frac{s}{\sqrt{n}} = \frac{1.523}{\sqrt{60}} = 0.197$$

7.7.10
$$\hat{\mu} = \bar{x} = 32.042$$

s.e.
$$(\bar{x}) = \frac{s}{\sqrt{n}} = \frac{5.817}{\sqrt{40}} = 0.920$$

7.7.11
$$\operatorname{Var}(s_1^2) = \operatorname{Var}\left(\frac{\sigma^2 \chi_{n_1 - 1}^2}{n_1 - 1}\right)$$

$$= \left(\frac{\sigma^2}{n_1 - 1}\right)^2 \operatorname{Var}(\chi^2_{n_1 - 1})$$

$$= \left(\frac{\sigma^2}{n_1 - 1}\right)^2 2(n_1 - 1) = \frac{2\sigma^4}{n_1 - 1}$$

Similarly,
$$Var(s_2^2) = \frac{2\sigma^4}{n_2 - 1}$$
.

The ratio of these two variances is $\frac{n_1-1}{n_2-1}$.

7.7.12 The true proportion of "very satisfied" customers is

$$p = \frac{11842}{24839} = 0.4768.$$

The probability that the manager's estimate of the proportion of "very satisfied" customers is within 0.10 of p = 0.4768 is

$$P(0.4768 - 0.10 \le \hat{p} \le 0.4768 + 0.10)$$

$$= P(0.3768 \times 80 \le X \le 0.5768 \times 80)$$

$$= P(30.144 \le X \le 46.144) = P(31 \le X \le 46)$$

This probability is 0.9264.

where $X \sim B(80, 0.4768)$.

7.7.13 When a sample of size n = 15 is used

$$P(62.8 - 0.5 \le \hat{\mu} \le 62.8 + 0.5)$$

= $P(62.3 \le \bar{X} \le 63.3)$
where $\bar{X} \sim N(62.8, 3.9^2/15)$.

This probability is equal to

$$\begin{split} &\Phi\left(\frac{63.3 - 62.8}{\sqrt{3.9^2/15}}\right) - \Phi\left(\frac{62.3 - 62.8}{\sqrt{3.9^2/15}}\right) \\ &= \Phi(0.4965) - \Phi(-0.4965) = 0.3804. \end{split}$$

When a sample of size n = 40 is used

$$P(62.8 - 0.5 \le \hat{\mu} \le 62.8 + 0.5)$$

= $P(62.3 \le \bar{X} \le 63.3)$
where $\bar{X} \sim N(62.8, 3.9^2/40)$.

This probability is equal to

$$\begin{split} &\Phi\left(\frac{63.3 - 62.8}{\sqrt{3.9^2/40}}\right) - \Phi\left(\frac{62.3 - 62.8}{\sqrt{3.9^2/40}}\right) \\ &= \Phi(0.8108) - \Phi(-0.8108) = 0.5826. \end{split}$$

7.7.14
$$\hat{\mu} = \bar{x} = 25.318$$

s.e. $(\bar{x}) = \frac{s}{\sqrt{n}} = \frac{0.226}{\sqrt{44}} = 0.0341$

The upper quartile of the distribution of soil compressibilities can be estimated by the upper sample quartile 25.50.

- 7.7.15 Probability theory
- 7.7.16 Probability theory

7.7.17
$$\hat{p} = \frac{39}{220} = 0.177$$

s.e. $(\hat{p}) = \sqrt{\frac{0.177 \times 0.823}{220}} = 0.026$

7.7.18 Let X be the number of cases where the treatment was effective.

$$\begin{split} &P\left(0.68-0.05 \leq \frac{X}{140} \leq 0.68+0.05\right) \\ &= P(88.2 \leq X \leq 102.2) \\ &= P(89 \leq B(140, 0.68) \leq 102) \\ &\simeq P(88.5 \leq N(140 \times 0.68, 140 \times 0.68 \times 0.32) \leq 102.5) \\ &= P\left(\frac{88.5-95.2}{5.519} \leq N(0, 1) \leq \frac{102.5-95.2}{5.519}\right) \\ &= \Phi(1.268) - \Phi(-1.268) = 0.80 \end{split}$$

7.7.19 (a)
$$\hat{\mu} = \bar{x} = 70.58$$

(b)
$$\frac{s}{\sqrt{n}} = \frac{12.81}{\sqrt{12}} = 3.70$$

(c)
$$\frac{67+70}{2} = 68.5$$

- 7.7.20 Statistical inference
- 7.7.21 Statistical inference
- 7.7.22 (a) True
 - (b) True
 - (c) True
 - (d) True

7.7.23
$$P(722 \le \bar{X} \le 724)$$

= $P(722 \le N(723, \frac{3^2}{11}) \le 724)$
= $P(\frac{-1 \times \sqrt{11}}{3} \le N(0, 1) \le \frac{1 \times \sqrt{11}}{3})$
= $\Phi(1.106) - \Phi(-1.106) = 0.73$

7.7.24 (a)
$$P(\mu - 20.0 \le \bar{X} \le \mu + 20.0)$$

 $= P(\mu - 20.0 \le N(\mu, \frac{40.0^2}{10}) \le \mu + 20.0)$
 $= P(\frac{-20.0 \times \sqrt{10}}{40.0} \le N(0, 1) \le \frac{20.0 \times \sqrt{10}}{40.0})$
 $= \Phi(1.58) - \Phi(-1.58) = 0.89$

(b)
$$P(\mu - 20.0 \le \bar{X} \le \mu + 20.0)$$

 $= P(\mu - 20.0 \le N(\mu, \frac{40.0^2}{20}) \le \mu + 20.0)$
 $= P(\frac{-20.0 \times \sqrt{20}}{40.0} \le N(0, 1) \le \frac{20.0 \times \sqrt{20}}{40.0})$
 $= \Phi(2.24) - \Phi(-2.24) = 0.97$

7.7.25
$$\hat{p}_A = \frac{852}{1962} = 0.434$$

$$\text{s.e.}(\hat{p}_A) = \sqrt{\frac{0.434 \times (1 - 0.434)}{1962}} = 0.011$$

- 7.7.34 A
- 7.7.35 A
- 7.7.36 D
- 7.7.37 A
- 7.7.38 A