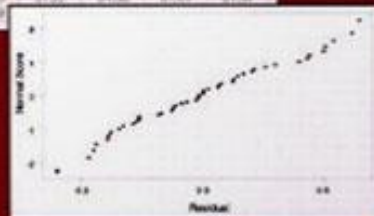


SOLUTIONS MANUAL

STATISTICS *and* DATA ANALYSIS *from Elementary to Intermediate*

Student	Student
51.28	51.40
51.43	51.75
51.24	51.21
51.46	51.27
52.25	51.84
51.27	51.44
52.31	51.50
51.37	52.09



AJIT C. TAMHANE
DOROTHY D. DUNLOP

INSTRUCTOR'S SOLUTIONS MANUAL

BRENT LOGAN

STATISTICS *and* DATA ANALYSIS

from Elementary to Intermediate

AJIT C. TAMHANE
DOROTHY D. DUNLOP

PRENTICE HALL, Upper Saddle River, NJ 07458

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Chapter 2 Solutions

Solutions to Section 2.1

2.1 (a)

Die Result	Coin(s) Result	Number of Outcomes
1	{(H),(T)}	2
2	{(H,H),..., (T,T)}	4
3	{(H,H,H),..., (T,T,T)}	8
4	{(H,H,H,H),..., (T,T,T,T)}	16
5	{(H,H,H,H,H),..., (T,T,T,T,T)}	32
6	{(H,H,H,H,H,H),..., (T,T,T,T,T,T)}	64

There are a total of 126 outcomes.

(b)

Die Result	Number of Heads	Number of Outcomes
1	{(0),(1)}	2
2	{(0),(1),(2)}	3
3	{(0),(1),..., (3)}	4
4	{(0),(1),..., (4)}	5
5	{(0),(1),..., (5)}	6
6	{(0),(1),..., (6)}	7

There are a total of 27 outcomes.

2.2 (a) $T \cap N^c$.

(b) $(T \cup R)^c$.

(c) $(T \cup R \cup N)^c$.

2.3 **Result 1:** Since A and A^c are mutually exclusive,

$$P(A) + P(A^c) = P(A \cup A^c) = P(S) = 1$$

and

$$P(A^c) = 1 - P(A).$$

Result 3:

$$P(A) = P((A \cap B) \cup (A \cap B^c)) = P(A \cap B) + P(A \cap B^c).$$

since $A \cap B$ and $A \cap B^c$ are mutually exclusive.

Result 2: (Uses Result 3)

$$P(A \cup B) = P((A \cap B) \cup (A \cap B^c) \cup (A^c \cap B)).$$

Since all of these are mutually exclusive,

$$\begin{aligned} P(A \cup B) &= P(A \cap B) + P((A \cap B^c) \cup (A^c \cap B)) \\ &= P(A \cap B) + P(A \cap B^c) + P(A^c \cap B) \\ &= P(A \cap B) + [P(A) - P(A \cap B)] + [P(B) - P(A \cap B)] \\ &= P(A) + P(B) - P(A \cap B). \end{aligned}$$

Result 4:

$$P(A \cap B^c) = P(A) - P(A \cap B) = P(A) - P(B).$$

Since

$$\begin{aligned} P(A \cap B^c) &= P(A) - P(B) \geq 0, \\ P(A) &\geq P(B). \end{aligned}$$

2.4

$$\begin{aligned} P(A \cap B) &= 1 - P((A \cap B)^c) \\ &= 1 - P(A^c \cup B^c) \\ &\geq 1 - [P(A^c) + P(B^c)] \\ &\geq 1 - [1 - P(A) + 1 - P(B)] \\ &\geq P(A) + P(B) - 1. \end{aligned}$$

2.5 (a) $\binom{52}{2} = 1326$.

(b) $\binom{4}{2} = 6$.

(c) $\binom{4}{1}\binom{48}{1} + \binom{4}{2} = 198$.

2.6 Out of the 54 numbers, 6 of them will be chosen for the lottery and 48 will not. So for the grand prize, the probability of winning is

$$\frac{\binom{6}{6}\binom{48}{0}}{\binom{54}{6}} = \frac{1}{25,827,165}.$$

For the second prize, the probability of winning is

$$\frac{\binom{6}{5}\binom{48}{1}}{\binom{54}{6}} = \frac{288}{25,827,165}.$$

For the third prize, the probability of winning is

$$\frac{\binom{6}{4}\binom{48}{2}}{\binom{54}{6}} = \frac{16,920}{25,827,165}.$$

2.7 The total number of moves is $m + n$. On each move, the path can either go right or up. So the number of paths between $(0,0)$ and (m,n) is the same as the number of ways to fill $m + n$ moves with m rights (and the rest ups), or $\binom{m+n}{m}$.

2.8 First split the n objects into two groups, one with n_1 and the other with n_2 . Then we can get r total objects by adding up the combinations selecting i from group 1 and $r - i$ from group 2, so that

$$\binom{n}{r} = \sum_i \binom{n_1}{i} \binom{n_2}{r-i}.$$

If $r < n_1$ then we can only select up to n_1 objects from the first group and the upper limit of the sum is n_1 . Otherwise, it is r , which yields $\min(n_1, r)$. Similarly, if $r > n_2$ then we can only select up to n_2 objects from the second group, so $r - i \leq n_2$ and the lower limit is $r - n_2$. Otherwise, we can select r objects yielding a lower limit on i of 0. So the final result is

$$\binom{n}{r} = \sum_{i=\max(0, r-n_2)}^{\min(n_1, r)} \binom{n_1}{i} \binom{n_2}{r-i}.$$

2.9

$$\begin{aligned}
\binom{n-1}{r-1} + \binom{n-1}{r} &= \frac{(n-1)!}{(r-1)!(n-r)!} + \frac{(n-1)!}{r!(n-r)!} \\
&= \frac{n!}{r!(n-r)!} \left(\frac{r}{n} + \frac{n-r}{n} \right) \\
&= \binom{n}{r}.
\end{aligned}$$

n	Coefficients						
0	1						
1	1		1				
2	1		2		1		
3	1	3		3		1	
4	1	4	6		4	1	
5	1	5	10	10		5	1

2.10 (a) $\binom{7}{2} = 21$.

(b) $\binom{7}{2}(2)^2(-3)^5 = -20,412$.

(c) $\frac{7!}{2!2!3!} = 210$.

2.11 (a)

$$\frac{\binom{12}{3}}{\binom{34}{3}} = \frac{220}{5984}.$$

(b)

$$\frac{\binom{2}{2}\binom{32}{1}}{\binom{34}{3}} = \frac{32}{5984}.$$

(c)

$$\frac{\binom{2}{2}\binom{32}{1}}{\binom{34}{3}} + \frac{\binom{12}{2}\binom{22}{1}}{\binom{34}{3}} + \frac{\binom{20}{2}\binom{14}{1}}{\binom{34}{3}} = \frac{4144}{5984}.$$

(d)

$$\frac{\binom{2}{2}\binom{32}{1}}{\binom{34}{3}} + \frac{\binom{12}{2}\binom{22}{1}}{\binom{34}{3}} + \frac{\binom{20}{2}\binom{14}{1}}{\binom{34}{3}} + \frac{\binom{12}{3}\binom{22}{0}}{\binom{34}{3}} + \frac{\binom{20}{3}\binom{14}{0}}{\binom{34}{3}} = \frac{5504}{5984}.$$

2.12 (a)

$$\frac{25}{30} \times \frac{5}{29} = 0.144.$$

(b)

$$\frac{25}{30} \times \frac{5}{29} + \frac{5}{30} \times \frac{4}{29} = 0.167.$$

(c)

$$\frac{5}{30} \times \frac{4}{29} \times \frac{3}{28} = 0.002.$$

2.13 (a)

$$P(T \cap N^c) = P(T) - P(T \cap N) = 0.77 - 0.45 = 0.32.$$

(b)

$$P((T \cup R)^c) = 1 - P(T \cup R) = 1 - P(T) - P(R) + P(T \cap R) = 1 - .77 - .47 + .29 = 0.05.$$

(c)

$$\begin{aligned} P((T \cup R \cup N)^c) &= 1 - P(N) - P(T \cup R) + P(N \cap (T \cup R)) \\ &= 1 - 0.63 - 0.95 + P((N \cap T) \cup (N \cap R)) \\ &= -0.58 + P(N \cap T) + P(N \cap R) - P(N \cap T \cap R) \\ &= -0.58 + 0.45 + 0.21 - 0.06 = 0.02. \end{aligned}$$

2.14 If we order the 12 kids by team, then there are $12!$ ways to assign the performance ranks. However, within each team, order is irrelevant, so we need to divide out the $3!$ ways of ordering the three kids per team. So the final number of ways of ranking the teams is

$$\frac{12!}{3!3!3!} = 369,600.$$

2.15 For all four suits to be represented, one suit must have 2 cards and the other suits must have 1 card each. There are $\binom{4}{1}$ ways to choose the suit with 2 cards, and there are $\binom{13}{2}$ ways to select the 2 cards from that suit. For the remaining suits, there are $\binom{13}{1}$ ways to select the 1 card from that suit. So the final probability is

$$\frac{\binom{4}{1} \binom{13}{2} \binom{13}{1} \binom{13}{1} \binom{13}{1}}{\binom{52}{5}} = \frac{685,464}{25,827,165}.$$

Solutions to Section 2.2

2.16

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{0.3}{0.6} = 0.5.$$

$$P(B|C) = \frac{P(B \cap C)}{P(C)} = \frac{0.3}{0.6} = 0.5.$$

$$P(A|B^c) = \frac{P(A \cap B^c)}{P(B^c)} = \frac{P(A) - P(A \cap B)}{1 - P(B)} = \frac{0.5 - 0.3}{1 - 0.6} = 0.5.$$

$$P(B^c|A \cap C) = \frac{P(A \cap B^c \cap C)}{P(A \cap C)} = \frac{P(A \cap C) - P(A \cap B \cap C)}{P(A \cap C)} = \frac{0.2 - 0.1}{0.2} = 0.5.$$

2.17 (a) For A and B to be mutually exclusive, $P(A \cap B) = 0$ or $P(A \cup B) = P(A) + P(B)$. Then

$$0.8 = 0.4 + p,$$

which means that $p = 0.4$.

(b) For A and B to be independent, $P(A \cap B) = P(A)P(B)$. Then

$$P(A \cap B) = P(A) + P(B) - P(A \cup B) = 0.4 + p - 0.8 = P(A)P(B) = 0.4p,$$

which means that $p = 2/3$.

2.18 (a)

$$P(N|R) = \frac{P(N \cap R)}{P(R)} = \frac{0.21}{0.47} = 0.447.$$

(b)

$$P(R^c|T) = \frac{P(R^c \cap T)}{P(T)} = \frac{P(T) - P(R \cap T)}{P(T)} = \frac{0.77 - 0.29}{0.77} = 0.623.$$

(c)

$$P(T^c|N \cap R) = \frac{P(T^c \cap N \cap R)}{P(N \cap R)} = \frac{P(N \cap R) - P(N \cap R \cap T)}{P(N \cap R)} = \frac{0.21 - 0.06}{0.21} = 0.714.$$

2.19 (a) Let E_1 and F_1 denote whether the first return has an error or is flagged, respectively. Then

$$P(E_1 \cap F_1) = P(F_1|E_1)P(E_1) = 0.9 \times \frac{5}{30} = 0.15.$$

(b)

$$P(F_1) = P(F_1|E_1)P(E_1) + P(F_1|E_1^c)P(E_1^c) = 0.15 + 0.02 \times \frac{25}{30} = 0.167.$$

(c) Let E_2 and F_2 denote whether the second return has an error or is flagged, respectively. Then

$$P(E_2) = P(E_2|E_1)P(E_1) + P(E_2|E_1^c)P(E_1^c) = \frac{4}{29} \times \frac{5}{30} + \frac{5}{29} \times \frac{25}{30} = 0.167$$

and

$$P(F_2 \cap E_2) = P(F_2|E_2)P(E_2) = 0.9 \times 0.167 = 0.15.$$

2.20 Let A_i and P_i denote the events that an ace or a face card is drawn on the i th draw, respectively. Then

$$\begin{aligned} P(\text{ace before face card}) &= P(A_1) + P((A_1 \cup P_1)^c)P(A_2) + \dots \\ &= \frac{4}{52} + \left(\frac{36}{52}\right) \frac{4}{52} + \left(\frac{36}{52}\right)^2 \frac{4}{52} + \dots \\ &= \frac{4}{52} \sum_{i=0}^{\infty} \left(\frac{36}{52}\right)^i \\ &= \frac{4}{52} \left(\frac{1}{1 - \frac{36}{52}}\right) = 0.25. \end{aligned}$$

2.21

- (a) Let F and M denote the genetic contributions from the father and mother respectively. Then the probabilities that a father contributes A or a are, respectively,

$$P(F = A) = P(F = A|AA)P(AA) + P(F = A|Aa)P(Aa) = p_0 + q_0/2.$$

and

$$P(F = a) = P(F = a|aa)P(aa) + P(F = a|Aa)P(Aa) = r_0 + q_0/2.$$

Then the probability that the first generation is AA is

$$p_1 = P(F = A, M = A) = P(F = A)P(M = A) = [P(F = A)]^2 = (p_0 + q_0/2)^2.$$

Similarly, the probability that the first generation is Aa is

$$\begin{aligned} q_1 &= P(F = A, M = a) + P(F = a, M = A) \\ &= P(F = A)P(M = a) + P(F = a)P(M = A) \\ &= 2P(F = A)P(F = a) = 2(p_0 + q_0/2)(r_0 + q_0/2). \end{aligned}$$

Finally, the probability that the first generation is aa is

$$r_1 = P(F = a, M = a) = P(F = a)P(M = a) = [P(F = a)]^2 = (r_0 + q_0/2)^2.$$

- (b) Similar to (a),

$$p_2 = (p_1 + q_1/2)^2, \quad q_2 = 2(p_1 + q_1/2)(r_1 + q_1/2), \quad \text{and} \quad r_2 = (r_1 + q_1/2)^2.$$

Then

$$\begin{aligned} p_2 &= \left[(p_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right]^2 \\ &= [(p_0 + q_0/2)(p_0 + q_0/2 + r_0 + q_0/2)]^2 \\ &= (p_0 + q_0/2)^2 (1)^2, \\ q_2 &= 2 \left[(p_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right] \\ &\quad \times \left[(r_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right] \\ &= 2(p_0 + q_0/2)(r_0 + q_0/2), \\ r_2 &= \left[(r_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right]^2 \\ &= [(r_0 + q_0/2)(r_0 + q_0/2 + p_0 + q_0/2)]^2 \\ &= (r_0 + q_0/2)^2 (1)^2. \end{aligned}$$

For the recursive proof, assume that this set of equations is true for n . Then

$$\begin{aligned} p_{n+1} &= (p_n + q_n/2)^2 \\ &= \left[(p_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right]^2 \\ &= [(p_0 + q_0/2)(p_0 + q_0/2 + r_0 + q_0/2)]^2 \\ &= (p_0 + q_0/2)^2, \\ q_{n+1} &= 2(p_n + q_n/2)(r_n + q_n/2) \\ &= 2 \left[(p_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right] \end{aligned}$$

$$\begin{aligned}
& \times \left[(r_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right] \\
& = 2(p_0 + q_0/2)(r_0 + q_0/2), \\
r_{n+1} & = (r_n + q_n/2)^2 \\
& = \left[(r_0 + q_0/2)^2 + 2(p_0 + q_0/2)(r_0 + q_0/2)/2 \right]^2 \\
& = [(r_0 + q_0/2)(r_0 + q_0/2 + p_0 + q_0/2)]^2 \\
& = (r_0 + q_0/2)^2.
\end{aligned}$$

2.22 Place of residence and opinion on a tax increase are not independent, since

$$P(\text{Yes and City}) = \frac{100}{1000} = 0.1 \neq P(\text{Yes})P(\text{City}) = \frac{400}{1000} \times \frac{400}{1000} = 0.16.$$

2.23

$$\begin{aligned}
P(A_1 \cup A_2 \cup \dots \cup A_n) & = 1 - P((A_1 \cup A_2 \cup \dots \cup A_n)^c) \\
& = 1 - P(A_1^c \cap A_2^c \cap \dots \cap A_n^c) \\
& = 1 - P(A_1^c)P(A_2^c) \dots P(A_n^c) \\
& = 1 - (1 - p)^n.
\end{aligned}$$

Using $p = 0.9$, for $n = 2$ the reliability is 0.99, for $n = 3$, the reliability is 0.999, and for $n = 4$, the reliability is 0.9999. As n gets larger, the reliability approaches 1.

2.24 (a) The event that there is current from A to C is

$$(R_1 \cap R_2) \cup (R_3) \cup (R_4 \cap R_5).$$

(b) The probability that there is current from A to C is

$$\begin{aligned}
P(\text{Current}) & = P((R_1 \cap R_2) \cup (R_3) \cup (R_4 \cap R_5)) \\
& = P(R_1 \cap R_2) + P(R_3) + P(R_4 \cap R_5) \\
& \quad - P(R_1 \cap R_2 \cap R_3) - P(R_1 \cap R_2 \cap R_4 \cap R_5) - P(R_3 \cap R_4 \cap R_5) \\
& \quad + P(R_1 \cap R_2 \cap R_3 \cap R_4 \cap R_5) \\
& = (0.9)^2 + 0.9 + (0.9)^2 - (0.9)^3 - (0.9)^4 - (0.9)^3 + (0.9)^5 \\
& = 0.996.
\end{aligned}$$

2.25 (a) Let D be the event that an appliance is defective. Then

$$P(B \cap D) = P(D|B)P(B) = 0.08 \times 0.37 = 0.0296.$$

(b)

$$P(D) = P(A \cap D) + P(B \cap D) = P(D|A)P(A) + 0.0296 = 0.04 \times 0.63 + 0.0296 = 0.0548.$$

(c)

$$P(B|D) = \frac{P(B \cap D)}{P(D)} = \frac{0.0296}{0.0548} = 0.5401.$$

2.26

- (a) Let E be the event that a tax return contains an error, and let F be the event that a tax return is flagged. Then

$$P(E \cap F) = P(F|E)P(E) = 0.85 \times 0.15 = 0.1275.$$

- (b)

$$P(F) = P(E \cap F) + P(E^c \cap F) = 0.1275 + P(F|E^c)P(E^c) = 0.1275 + 0.05 \times 0.85 = 0.17.$$

- (c)

$$P(E^c|F^c) = \frac{P(E^c \cap F^c)}{P(F^c)} = \frac{P(F^c|E^c)P(E^c)}{1 - P(F)} = \frac{0.95 \times 0.85}{1 - 0.17} = 0.973.$$

- 2.27 (a) Let D and ND refer to the events where a person has or doesn't have the disease, respectively. Then

$$P(D|+) = \frac{P(+|D)P(D)}{P(+|D)P(D) + P(+|ND)P(ND)} = \frac{0.99 \times 0.1}{0.99 \times 0.1 + 0.02 \times 0.9} = 0.846.$$

- (b)

$$P(ND|-) = \frac{P(-|ND)P(ND)}{P(-|ND)P(ND) + P(-|D)P(D)} = \frac{0.98 \times 0.9}{0.98 \times 0.9 + 0.01 \times 0.1} = 0.999.$$

The diagnostic test appears pretty reliable, although it is less reliable in identifying true positives than true negatives.

- (c)

$$P(D|+) = \frac{P(+|D)P(D)}{P(+|D)P(D) + P(+|ND)P(ND)} = \frac{0.99 \times 0.001}{0.99 \times 0.001 + 0.02 \times 0.999} = 0.047.$$

- (d) For rare diseases, too many false positives would appear in the screening program, and it would not be very effective in identifying people with the disease.

Solutions to Section 2.3

- 2.28 (a) For this to be a p.m.f.,

$$\sum_x f(x) = c(1/2) + c(1/4) + c(1/8) + c(1/16) = 1,$$

$$\text{or } c = 16/15 = 1.067.$$

- (b) The c.d.f. is

$$F(x) = \begin{cases} 0 & \text{if } x < 1 \\ 1.067/2 = 0.533 & \text{if } 1 \leq x < 2 \\ 0.533 + 1.067/4 = 0.8 & \text{if } 2 \leq x < 3 \\ 0.8 + 1.067/8 = 0.933 & \text{if } 3 \leq x < 4 \\ 1 & \text{if } x \geq 4. \end{cases}$$

- 2.29 (a)

x	$f(x)$	$F(y)$, for $x \leq y < x + 1$
0	6/36	6/36
1	10/36	16/36
2	8/36	24/36
3	6/36	30/36
4	4/36	34/36
5	2/36	1

(b)

$$P(0 < x \leq 3) = F(3) - F(0) = 24/36.$$

$$P(1 \leq x < 3) = F(2) - F(0) = 18/36.$$

2.30 (a) For this to be a p.m.f.,

$$\sum_x f(x) = \sum_{x=1}^{\infty} \frac{1}{n(n+1)} = 1.$$

Since

$$\sum_{x=1}^n \frac{1}{n(n+1)} = \frac{n}{n+1},$$

then

$$\sum_{x=1}^{\infty} = \lim_{n \rightarrow \infty} \frac{n}{n+1} = 1,$$

and $f(x)$ is a p.m.f.

(b) From (a) the c.d.f. is

$$F(x) = \begin{cases} 0 & \text{if } x < 1 \\ \frac{i}{i+1} & \text{if } i \leq x < i + 1 \end{cases}$$

2.31 For x between n and N , the random variable $X = x$ when the largest chip is x and the remaining $n - 1$ chips are smaller than x . Out of the $x - 1$ chips smaller than x , we want to choose $n - 1$, so there are

$$\binom{x-1}{n-1}$$

ways to do this. The total number of ways to choose n chips is

$$\binom{N}{n},$$

so the p.m.f. is

$$P(X = x) = \frac{\binom{x-1}{n-1}}{\binom{N}{n}}$$

2.32 (a) For this to be a p.d.f.,

$$\int_x f(x) = \int_0^1 0.5 dx + \int_1^3 (0.5 + c(x-1)) dx = 1.$$

Since

$$\begin{aligned}\int_x f(x) &= .5x|_0^1 + .5x|_1^3 + \frac{c}{2}(x-1)^2|_1^3 \\ &= 1.5 + \frac{c}{2}(4-0) = 1.5 + 2c = 1,\end{aligned}$$

then c must be -0.25 .

(b) Using the value of c found above, the new p.d.f. for $1 \leq X < 3$ is

$$f(x) = 0.5 - 0.25(x-1) = 0.75 - 0.25x.$$

The c.d.f. for x between 0 and 1 is

$$F(x) = \int_0^x 0.5 dx = 0.5x.$$

At $x = 1$, $F(1) = 0.5$, so the c.d.f. for x between 1 and 3 is

$$F(x) = F(1) + \int_1^x (0.75 - 0.25x) dx = 0.5 + (0.75x - 0.125x^2)|_1^x = -.125x^2 + 0.75x - .125.$$

Then the final c.d.f. is given by

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ 0.5x & \text{if } 0 \leq x < 1 \\ -.125x^2 + .75x - .125 & \text{if } 1 \leq x < 3 \\ 1 & \text{if } x \geq 3. \end{cases}$$

2.33 (a) Continuous.

(b)

$$P(1 \leq X \leq 3) = F(3) - F(1) = 0.8 - 0.4 = 0.4.$$

(c)

$$P(X \geq 1) = 1 - F(1) = 1 - 0.4 = 0.6.$$

2.34 (a) Discrete.

(b)

$$P(1 \leq X < 2) = P(X = 1) = 0.8 - 0.4 = 0.4.$$

(c)

$$P(X \geq 1) = 1 - F(0) = 1 - 0.4 = 0.6.$$

Solutions to Section 2.4

2.35 (a) The p.d.f. is

$$f(x) = \begin{cases} \frac{1}{N} & \text{if } x = 1, 2, \dots, N \\ 0 & \text{otherwise.} \end{cases}$$

(b) The mean is

$$E(X) = \sum_x x f(x) = \sum_{x=1}^N x \times \frac{1}{N} = \frac{1}{N} \times \frac{N(N+1)}{2} = \frac{N+1}{2}.$$

The variance is

$$\begin{aligned} \text{Var}(X) &= \sum_x x^2 f(x) - \left(\frac{N+1}{2}\right)^2 \\ &= \frac{1}{N} \sum_{x=1}^N x^2 - \left(\frac{N+1}{2}\right)^2 \\ &= \frac{N(N+1)(2N+1)}{6N} - \left(\frac{N+1}{2}\right)^2 \\ &= \frac{N+1}{2} \left[\frac{2N+1}{3} - \frac{N+1}{2} \right] \\ &= \frac{(N+1)(N-1)}{12}. \end{aligned}$$

(c) For a single die, $N = 6$, so that $E(X) = 7/2 = 3.5$ and $\text{Var}(X) = (7)(5)/12 = 2.917$.

2.36 (a) The mean is

$$E(X) = \sum_x x f(x) = 0 \times 0.1 + 1 \times 0.2 + \dots + 8 \times 0.02 = 2.57.$$

The variance is

$$\text{Var}(X) = \sum_x x^2 f(x) - E(X)^2 = 0^2 \times 0.1 + 1^2 \times 0.2 + \dots + 8^2 \times 0.02 - 2.57^2 = 3.545.$$

(b) The skewness is

$$\beta_3 = \frac{E[(X - \mu)^3]}{\sigma^3} = \frac{(0 - 2.57)^3 \times 0.1 + \dots + (8 - 2.57)^3 \times 0.02}{(3.545)^{3/2}} = 0.948.$$

The distribution is positively skewed.

2.37 (a)

$$\begin{aligned} E(X) &= \sum_x x f(x) = \sum_{x=n}^N x \frac{\binom{x-1}{n-1}}{\binom{N}{n}} \\ &= \frac{1}{\binom{N}{n}} \sum_{x=n}^N \frac{x(x-1)!}{(n-1)!(x-n)!} \\ &= \frac{n}{\binom{N}{n}} \sum_{x=n}^N \binom{x}{n} \\ &= \frac{n}{\binom{N}{n}} \binom{N+1}{n+1} \\ &= \frac{n(n!(N-n)!)}{N!} \frac{(N+1)!}{(n+1)!(N-n)!} \\ &= \frac{n(N+1)}{n+1}. \end{aligned}$$

(b) We could estimate N by solving

$$X = \frac{n(N+1)}{n+1}$$

for N , which yields

$$\hat{N} = \frac{x(n+1)}{n} - 1.$$

2.38

$$E(X) = \sum_x xf(x) = \sum_{x=1}^{\infty} x \frac{1}{x(x+1)} = \sum_{x=1}^{\infty} \frac{1}{x+1} = \infty.$$

2.39 (a) The expected demand is

$$E(X) = \sum_x xf(x) = 1 \times 0.01 + 2 \times 0.04 + \dots + 10 \times 0.05 = 6.19.$$

(b) If $X \leq n$ then

$$\text{Profit} = 0.5X - 1(n - X) = 1.5X - n.$$

If $X > n$ then

$$\text{Profit} = 0.5n - 0.75(X - n) = 1.25n - 0.75X.$$

(c) The table of profits is below

Demand (x)	1	2	3	4	5	6	7	8	9	10
Probability ($f(x)$)	0.01	0.04	0.05	0.10	0.15	0.20	0.20	0.10	0.10	0.05
Profit ($n = 5$)	-3.5	-2	-0.5	1	2.5	1.75	1	0.25	-0.5	-1.25
Profit ($n = 6$)	-4.5	-3	-1.5	0	1.5	3	2.25	1.5	0.75	0
Profit ($n = 7$)	-5.5	-4	-2.5	-1	0.5	2	3.5	2.75	2	1.25

The expected profits are obtained by computing the weighted average of the profits, weighed by the probability of each occurring.

$$E(\text{Profit}) = \begin{cases} 0.798 & \text{if } n = 5 \\ 1.26 & \text{if } n = 6 \\ 1.205 & \text{if } n = 7. \end{cases}$$

$n = 6$ maximizes the expected profit.

2.40 (a) The c.d.f. is

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ \int_0^x \frac{1}{(1+z)^2} dz = \left. \frac{-1}{1+z} \right|_0^x = 1 - \frac{1}{1+x} & \text{if } x \geq 0 \end{cases}$$

(b) To find the p th quantile, set

$$p = F(x) = 1 - \frac{1}{1+x}$$

and solve for x . In this case,

$$x = \frac{1}{1-p} - 1$$

For $p = 0.5$,

$$x_{(0.5)} = \frac{1}{1-0.5} - 1 = 1.$$

(c)

$$E(\sqrt{X}) = \int_0^{\infty} \frac{\sqrt{x}}{(1+x)^2} dx$$

Let $x = \tan^2 \theta$, so that $dx = 2 \tan \theta \sec^2 \theta d\theta$. Then

$$\begin{aligned} E(\sqrt{X}) &= \int_0^{\pi/2} \frac{2 \tan^2 \theta \sec^2 \theta}{(\sec^2 \theta)^2} d\theta \\ &= 2 \int_0^{\pi/2} \sin^2 \theta d\theta \\ &= 2 \frac{\pi}{4} = \frac{\pi}{2}. \end{aligned}$$

2.41 (a) The c.d.f. is

$$F(x) = \begin{cases} 0 & \text{if } x < 1 \\ \int_1^x 2z^{-3} dz = -z^{-2}|_1^x = 1 - x^{-2} & \text{if } x \geq 1 \end{cases}$$

(b) To find the p th quantile, set

$$p = F(x) = 1 - x^{-2}$$

and solve for x . In this case,

$$x = \sqrt{\frac{1}{1-p}}$$

For $p = 0.5$,

$$x_{(0.5)} = \sqrt{\frac{1}{1-0.5}} = \sqrt{2} = 1.414.$$

(c) The mean is

$$E(X) = \int_1^{\infty} x \times 2x^{-3} dx = \int_1^{\infty} 2x^{-2} dx = -2x^{-1}|_1^{\infty} = 2.$$

Since

$$E(X^2) = \int_1^{\infty} x^2 \times 2x^{-3} dx = \int_1^{\infty} 2/x dx = 2 \ln x|_1^{\infty} = \infty.$$

the variance is also ∞ .

2.42 (a) For this to be a c.d.f.,

$$\int_0^1 f(x) = \int_0^1 cx(1-x) = [cx^2/2 - cx^3/3]_0^1 = c/6 = 1,$$

or $c = 6$.

(b) The c.d.f. is

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ \int_0^x 6z(1-z) dz = (3z^2 - 2z^3)|_0^x = (3x^2 - 2x^3) & \text{if } 0 \leq x < 1 \\ 1 & \text{if } x \geq 1 \end{cases}$$

(c) The mean is

$$E(X) = \int_0^1 x \times 6x(1-x) dx = \int_0^1 (6x^2 - 6x^3) dx = (2x^3 - 1.5x^4)_0^1 = 0.5.$$

Since

$$E(X^2) = \int_0^1 x^2 \times 6x(1-x) dx = \int_0^1 (6x^3 - 6x^4) dx = (1.5x^4 - 1.2x^5)_0^1 = 0.3.$$

the variance is

$$\text{Var}(X) = E(X^2) - E(X)^2 = 0.3 - 0.5^2 = 0.05.$$

2.43 Let $Y = aX + b$. Then

$$E(Y) = aE(X) + b$$

and

$$\begin{aligned} \text{Var}(Y) &= E[(Y - E(Y))^2] \\ &= E[(aX + b - (aE(X) + b))^2] \\ &= a^2 E[(X - E(X))^2] \\ &= a^2 \text{Var}(X) \end{aligned}$$

2.44 (a) For X with a finite range $0, 1, \dots, N$,

$$\begin{aligned} E(X) &= \sum_x x f(x) = \sum_{x=1}^N x(F(x) - F(x-1)) \\ &= 1(F(1) - F(0)) + 2(F(2) - F(1)) + \dots + N(F(N) - F(N-1)) \\ &= NF(N) - \sum_{x=0}^{N-1} F(x) \\ &= \sum_{x=0}^{N-1} [F(N) - F(x)] \\ &= \sum_{x=0}^{N-1} [1 - F(x)] \end{aligned}$$

As $N \rightarrow \infty$,

$$E(X) = \sum_{x=0}^{\infty} [1 - F(x)].$$

(b) For $F(x) = 1 - (1-p)^x$,

$$E(X) = \sum_{x=0}^{\infty} [1 - (1 - (1-p)^x)] = \sum_{x=0}^{\infty} (1-p)^x = \frac{1}{1 - (1-p)} = 1/p.$$

2.45

(a) For X with a finite range $[0, A]$,

$$E(X) = \int_0^A xf(x) dx.$$

Let $u = x, dv = f(x)dx$. Then $du = dx$ and $v = F(x)$, and integration by parts yields

$$E(X) = xF(x)|_0^A - \int_0^A F(x) dx = A - \int_0^A F(x) dx = \int_0^A (1 - F(x)) dx.$$

As $A \rightarrow \infty$,

$$E(X) = \int_0^\infty (1 - F(x)) dx.$$

(b) For $F(x) = 1 - e^{-\lambda x}$,

$$E(X) = \int_0^\infty [1 - (1 - e^{-\lambda x})] dx = \int_0^\infty e^{-\lambda x} dx = \frac{-1}{\lambda} e^{-\lambda x} \Big|_0^\infty = 1/\lambda.$$

2.46 (a) Let Y_i be Bernoulli(p). Then

$$\psi_{Y_i}(t) = E(e^{tY_i}) = \sum_{y=0}^1 e^{ty} f(y) = pe^t + q.$$

Then

$$\psi_X(t) = \psi_{Y_1}(t) \cdots \psi_{Y_n}(t) = [\psi_{Y_i}(t)]^n = (pe^t + q)^n.$$

(b) The first derivative of the moment generating function is

$$\psi'_X(t) = n(pe^t + q)^{n-1}pe^t.$$

Then

$$E(X) = \psi'_X(0) = n(p+q)^{n-1}p = np.$$

The second derivative is

$$\psi''_X(t) = n(n-1)(pe^t + q)^{n-2}p^2e^{2t} + n(pe^t + q)^{n-1}pe^t.$$

Then

$$E(X^2) = \psi''_X(0) = n(n-1)(p+q)^{n-2}p + n(p+q)^{n-1}p = np[(n-1)p + 1],$$

and

$$\text{Var}(X) = E(X^2) - E(X)^2 = np[(n-1)p + 1] - (np)^2 = np(1-p).$$

2.47 (a)

$$\begin{aligned} \psi_X(t) &= E(e^{tX}) \\ &= \sum_{x=0}^{\infty} e^{tx} \frac{\lambda^x e^{-\lambda}}{x!} \\ &= \sum_{x=0}^{\infty} \frac{(\lambda e^t)^x e^{-\lambda}}{x!} \\ &= \frac{e^{-\lambda}}{e^{-\lambda e^t}} \sum_{x=0}^{\infty} \frac{(\lambda e^t)^x e^{-(\lambda e^t)}}{x!} \end{aligned}$$

Since this sum is the sum of a Poisson(λe^t) p.m.f. from 0 to ∞ , it reduces to 1 and

$$\psi_X(t) = e^{\lambda(e^t-1)}.$$

The first derivative of the moment generating function is

$$\psi'_X(t) = (\lambda e^t)e^{\lambda(e^t-1)}.$$

Then

$$E(X) = \psi'_X(0) = \lambda.$$

The second derivative is

$$\psi''_X(t) = (\lambda^2 e^{2t})e^{\lambda(e^t-1)} + (\lambda e^t)e^{\lambda(e^t-1)}.$$

Then

$$E(X^2) = \psi''_X(0) = \lambda^2 + \lambda,$$

and

$$\text{Var}(X) = E(X^2) - E(X)^2 = \lambda^2 + \lambda - (\lambda)^2 = \lambda.$$

(b) Let $Y = \sum_i X_i$. Then

$$\psi_Y(t) = \psi_{X_1}(t) \cdots \psi_{X_n}(t) = e^{\lambda_1(e^t-1)} \cdots e^{\lambda_n(e^t-1)} = e^{(e^t-1) \sum_i \lambda_i}.$$

This is the moment generating function of a Poisson ($\sum_i \lambda_i$) random variable.

Solutions to Section 2.5

2.48 (a)

$$\begin{aligned} \text{Cov}(aX + b, cY + d) &= E[(aX + b - (aE(X) + b))(cY + d - (cE(Y) + d))] \\ &= acE[(X - E(X))(Y - E(Y))] \\ &= ac\text{Cov}(X, Y). \end{aligned}$$

(b)

$$\begin{aligned} \text{Var}(X \pm Y) &= E\{[(X \pm Y) - (E(X) \pm E(Y))]^2\} \\ &= E\{[(X - E(X)) \pm (Y - E(Y))]^2\} \\ &= E[(X - E(X))^2] \pm 2E[(X - E(X))(Y - E(Y))] + E[(Y - E(Y))^2] \\ &= \text{Var}(X) + \text{Var}(Y) \pm 2\text{Cov}(X, Y). \end{aligned}$$

(c) If X and Y are independent, then

$$\begin{aligned} E(XY) &= \int_{x,y} xyf(x,y) dx dy \\ &= \int_x \int_y xyf(x)f(y) dx dy \\ &= \int_y yf(y) dy \int_x xf(x) dx = E(X)E(Y) \end{aligned}$$

and

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = 0.$$

2.49

$$\begin{aligned}
\text{Cov}(X, Y) &= \text{Cov}(X_1 + X_2, X_1 - X_2) \\
&= E[(X_1 + X_2 - (E(X_1) + E(X_2)))(X_1 - X_2 - (E(X_1) - E(X_2)))] \\
&= E[((X_1 - E(X_1)) + (X_2 - E(X_2)))((X_1 - E(X_1)) - (X_2 - E(X_2)))] \\
&= E[(X_1 - E(X_1))^2 - (X_2 - E(X_2))^2] \\
&= V(X_1) - V(X_2) = 0.
\end{aligned}$$

So the correlation ρ is also 0.

2.50

$$\begin{aligned}
\text{Cov}(Y_1, Y_2) &= \text{Cov}(X_0 - X_1, X_0 - X_2) \\
&= \text{Var}(X_0) - \text{Cov}(X_0, X_1) - \text{Cov}(X_0, X_2) + \text{Cov}(X_1, X_2) \\
&= \sigma_0^2.
\end{aligned}$$

So the correlation is

$$\rho = \frac{\sigma_0^2}{\sqrt{\sigma_0^2 + \sigma_1^2} \sqrt{\sigma_0^2 + \sigma_2^2}}.$$

2.51

$$\begin{aligned}
\text{Cov}(\bar{X}, X_i - \bar{X}) &= \text{Cov}\left(\frac{1}{n} \sum_j X_j, X_i - \frac{1}{n} \sum_j X_j\right) \\
&= \text{Cov}\left(\frac{1}{n} \sum_j X_j, \frac{n-1}{n} X_i - \frac{1}{n} \sum_{j \neq i} X_j\right) \\
&= \frac{1}{n^2} \left[\text{Cov}\left(\sum_j X_j, (n-1)X_i\right) - \text{Cov}\left(\sum_j X_j, \sum_{j \neq i} X_j\right) \right] \\
&= \frac{1}{n^2} \left[(n-1)\text{Var}(X_i) - \sum_{j \neq i} \text{Var}(X_j) \right] \\
&= \frac{1}{n^2} [(n-1)\sigma^2 - (n-1)\sigma^2] \\
&= 0.
\end{aligned}$$

2.52 (a)

(x, y)	t	$f(x, y)$
(0,0)	0	0.3
(0,1)	1	0.18
(0,2)	2	0.12
(1,0)	1	0.15
(1,1)	2	0.09
(1,2)	3	0.06
(2,0)	2	0.05
(2,1)	3	0.03
(2,2)	4	0.02

(b)

t	$f(t)$
0	0.3
1	0.33
2	0.26
3	0.09
4	0.02

(c)

$$E(T) = \sum_t tf(t) = 0 \times 0.3 + 1 \times 0.33 + \dots + 4 \times 0.02 = 1.2.$$

$$\text{Var}(T) = \sum_t t^2 f(t) - E(T)^2 = 0^2 \times 0.3 + 1^2 \times 0.33 + \dots + 4^2 \times 0.02 - (1.2)^2 = 1.06.$$

(d) If heart and lung problems were positively correlated, $E(T)$ would be slightly decreased. Since 0, 2, and 4 (when $X = Y$) would be more frequent and most of the probability is located at $X = 0$, a positive correlation would increase the likelihood that Y was also low, and decrease the mean. The variance would be increased because, with the correlation between X and Y , extreme values like 0 or 4 would be more likely.

2.53 (a)

$$E(X) = -200 \times 0.2 + 400 \times 0.8 = 280.$$

$$\sigma_X = \sqrt{(-200 - 280)^2 \times 0.2 + (400 - 280)^2 \times 0.8} = 240.$$

(b)

$$E(Y) = -100 \times 0.1 + 300 \times 0.9 = 260.$$

$$\sigma_Y = \sqrt{(-100 - 260)^2 \times 0.1 + (300 - 260)^2 \times 0.9} = 120.$$

(c)

u	$f(u)$	v	$f(v)$
-300	0.02	-500	0.18
100	0.18	-100	0.02
300	0.08	100	0.72
700	0.72	500	0.08

(d)

$$E(U) = -300 \times 0.02 + \dots + 700 \times 0.72 = 540.$$

$$\sigma_U = \sqrt{(-300)^2 \times 0.02 + \dots + (700)^2 \times 0.72 - (540)^2} = 268.328.$$

$$E(V) = -500 \times 0.18 + \dots + 300 \times 0.08 = 20.$$

$$\sigma_V = \sqrt{(-500)^2 \times 0.18 + \dots + (300)^2 \times 0.08 - (20)^2} = 268.328.$$

(e)

$$E(U) = 280 + 260 = 540.$$

$$E(V) = 280 - 260 = 20.$$

$$\sigma_U = \sigma_V = \sqrt{(240)^2 + (120)^2} = 268.328.$$

2.54 (a)

x	$f(x)$	y	$f(y)$
100	0.5	0	0.25
250	0.5	100	0.25
		200	0.5

X and Y are not independent because

$$P(X = 100, Y = 0) = 0.2 \neq P(X = 100)P(Y = 0) = 0.5 \times 0.25 = 0.125.$$

(b)

y	$f(y x = 100)$	$f(y x = 250)$
0	0.4	0.1
100	0.2	0.3
200	0.4	0.6

(c)

$$E(XY) = \sum_{x,y} xyf(x,y) = 100 \times 100 \times \frac{1500}{15000} + \dots + 250 \times 200 \times \frac{4500}{15000} = 23750,$$

$$E(X) = 175,$$

and

$$E(Y) = 0 \times 0.25 + 100 \times 0.25 + 200 \times 0.5 = 125.$$

Then

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = 23750 - (175)(125) = 1875.$$

2.55 (a)

$$f_X(x) = \int_x^1 8xy \, dy = \frac{8xy^2}{2} \Big|_x^1 = 4x - 4x^3.$$

$$f_Y(y) = \int_0^y 8xy \, dx = \frac{8yx^2}{2} \Big|_0^y = 4y^3.$$

Since

$$f(x,y) \neq f_X(x)f_Y(y)$$

X and Y are not independent.

(b)

$$f(y|x) = \frac{f(x,y)}{f_X(x)} = \frac{8xy}{4x(1-x^2)} = \frac{2y}{1-x^2}.$$

(c)

$$\begin{aligned} E(XY) &= \int_0^1 \int_x^1 xy \, dy \, dx \\ &= \int_0^1 8x^2 \int_x^1 y^2 \, dy \, dx \\ &= \int_0^1 8x^2 y^3 / 3 \Big|_x^1 \, dx \\ &= \frac{8}{3} \int_0^1 (x^2 - x^5) \, dx \\ &= \frac{8}{3} (x^3/3 - x^6/6) \Big|_0^1 \\ &= 4/9 = 0.444. \end{aligned}$$

$$E(X) = \int_0^1 x(4x - 4x^3) \, dx = (4x^3/3 - 4x^5/5) \Big|_0^1 = 8/15 = 0.533.$$

$$E(Y) = \int_0^1 y(4y^3) \, dy = 4y^5/5 \Big|_0^1 = 4/5 = 0.8.$$

Then

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = 0.444 - (0.533)(0.8) = 0.017.$$

Solutions to Section 2.6

2.56 (a) Since

$$E(X_i) = 0$$

and

$$\text{Var}(X_i) = (-1)^2 \times 0.5 + (1)^2 \times 0.5 = 1,$$

then

$$E(Y) = 10E(X_i) = 0.$$

and

$$\text{Var}(Y) = 10\text{Var}(X_i) = 10.$$

(b) An upper bound is

$$P(|Y| \geq 4) \leq \frac{\sigma_Y^2}{4^2} = \frac{10}{16} = 0.625.$$

(c) An upper bound is

$$P(|\bar{X}| \geq 4) \leq \frac{\sigma_{\bar{X}}^2}{4^2} = \frac{1/10}{16} = 0.006.$$

2.57 (a)

$$P(|\bar{X} - \mu| \leq c) = 1 - P(|\bar{X} - \mu| \geq c) \geq 1 - \frac{\sigma_{\bar{X}}^2}{c^2} = 1 - \frac{1}{nc^2}.$$

(b) To assure a probability of at least 0.95, i.e.

$$1 - \frac{1}{nc^2} = 0.95$$

then

$$n = \frac{1}{0.05c^2} = \frac{20}{c^2}.$$

For $c = 0.1$, n must be 2000. For $c = 1.0$, n must be 20. For $c = 2.0$, n must be 5.

2.58

$$P(15 \leq X \leq 25) = P(|X - 20| \leq 5) = 1 - P(|X - 20| \geq 5) \geq 1 - \frac{4.4^2}{5^2} = 0.2256.$$

Solutions to Section 2.7

- 2.59** (a) Let X be the number of occupied lanes, then X is binomial with $n = 10$ and $p = 0.75$.
Then

$$P(X \leq 9) = 1 - P(X = 10) = 1 - \binom{10}{10} (0.75)^{10} (1 - 0.75)^0 = 0.944.$$

(b)

$$E(X) = np = 10 \times 0.75 = 7.5$$

and

$$\sigma_X = \sqrt{np(1-p)} = \sqrt{10 \times 0.75 \times (1 - 0.75)} = 1.369.$$

- 2.60** (a) Let X be the number of times a participant will be selected, then X is binomial with $n = 12$ and $p = 0.1$. Then

$$E(X) = np = 12 \times 0.1 = 1.2.$$

(b)

$$P(X = 2) = \binom{12}{2} (0.1)^2 (1 - 0.1)^{10} = 0.230.$$

(c) Now $p = 0.2$, so that

$$P(X = 2) = \binom{12}{2} (0.2)^2 (1 - 0.2)^{10} = 0.283.$$

- 2.61** (a)

$$P(X \leq 2) = \frac{\binom{10}{0} \binom{30}{4-0}}{\binom{40}{4}} + \frac{\binom{10}{1} \binom{30}{4-1}}{\binom{40}{4}} + \frac{\binom{10}{2} \binom{30}{4-2}}{\binom{40}{4}} = 0.300 + 0.444 + 0.214 = 0.958.$$

(b)

$$\begin{aligned} P(X \leq 2) &\approx \binom{4}{0} (0.25)^0 (1 - 0.25)^4 + \binom{4}{1} (0.25)^1 (1 - 0.25)^3 + \binom{4}{2} (0.25)^2 (1 - 0.25)^2 \\ &= 0.316 + 0.422 + 0.211 = 0.949. \end{aligned}$$

The approximation is fairly accurate.

- 2.62** (a) X has a hypergeometric distribution with $N = 50$, $M = 2$, and $n = 3$. Then

$$f(x) = \frac{\binom{2}{x} \binom{48}{3-x}}{\binom{50}{3}} \text{ for } x = 0, 1, 2.$$

(b)

$$P(X = 0) = \frac{\binom{2}{0} \binom{48}{3-0}}{\binom{50}{3}} = 0.882.$$

(c) Using $p = M/N = 2/50 = 0.04$,

$$P(X = 0) \approx \binom{3}{0} (0.04)^0 (1 - 0.04)^3 = 0.885.$$

The approximation is very accurate.

2.63 (a) Let X be the number of emissions in a week. X has a Poisson distribution with $\lambda = 0.25$. Then

$$P(X \geq 1) = 1 - P(X = 0) = 1 - \frac{e^{-0.25} (0.25)^0}{0!} = 0.221.$$

(b) Let Y be the number of emissions in a year. Y has a Poisson distribution with $\lambda = 0.25 \times 52 = 13$. Then

$$P(Y \geq 1) = 1 - P(Y = 0) = 1 - \frac{e^{-13} (13)^0}{0!} = 1.000.$$

2.64 (a) X has a Binomial distribution with $n = 200$ and $p = 1/20 = 0.05$. Then

$$P(X \geq 5) = 1 - P(X \leq 4) = 1 - \sum_{i=0}^4 \binom{200}{i} (0.05)^i (1 - 0.05)^{200-i}$$

(b) The Poisson distribution fits because an error on a page is a rare event, and the number of pages in the manuscript is large. Using $\lambda = 0.05 \times 200 = 10$, and the table of cumulative Poisson probabilities,

$$P(X \geq 5) = 1 - P(X \leq 4) = 1 - \sum_{i=0}^4 \frac{e^{-10} 10^i}{i!} = 1 - 0.029 = 0.971.$$

2.65 (a) Using the table of cumulative Poisson probabilities,

$$P(X > 10) = 1 - \sum_{i=0}^{10} \frac{e^{-5} 5^i}{i!} = 1 - 0.986 = 0.014.$$

(b) The distribution needs to be renormalized by the probability that $X \leq 10$, so the p.m.f. is

$$f(x) = \frac{e^{-5} 5^x}{x!} \frac{1}{0.986}.$$

Now the total probability sums to 1. Then

$$E(X) = \sum_{x=0}^{10} x f(x) = 4.910.$$

2.66 (a) $E(X_1) = np_1 = 2$, $E(X_2) = np_2 = 4$, $E(X_3) = np_3 = 6$, and $E(X_4) = np_4 = 8$.

(b) Using $\text{Var}(X_i) = np_i(1 - p_i)$ and $\text{Cov}(X_i, X_j) = -np_i p_j$, the variance-covariance matrix is

$$\Sigma = \begin{bmatrix} 1.8 & -0.4 & -0.6 & -0.8 \\ -0.4 & 3.2 & -1.2 & -1.6 \\ -0.6 & -1.2 & 4.2 & -2.4 \\ -0.8 & -1.6 & -2.4 & 4.8 \end{bmatrix}.$$

(c) Using

$$\rho_{ij} = \frac{\text{Cov}(X_i, X_j)}{\sigma_i \sigma_j}$$

the correlation matrix is

$$\rho = \begin{bmatrix} 1 & -0.167 & -0.218 & -0.272 \\ -0.167 & 1 & -0.327 & -0.408 \\ -0.218 & -0.327 & 1 & -0.535 \\ -0.272 & -0.408 & -0.535 & 1 \end{bmatrix}.$$

(d)

$$P(X_1 = 2, X_2 = 4, X_3 = 6, X_4 = 8) = \frac{20!}{2!4!6!8!} (0.1)^2 (0.2)^4 (0.3)^6 (0.4)^8 = 0.013.$$

2.67 (a) Let X_1 , X_2 , and X_3 be the numbers of low, middle, and upper income people surveyed out of 4, respectively. Then X_1 , X_2 , and X_3 have a multinomial distribution with $n = 4$, $p_1 = 0.3$, $p_2 = 0.45$, and $p_3 = 0.25$. Then

$$P(X_1 = 1, X_2 = 1, X_3 = 2) = \frac{4!}{1!1!2!} (0.3)^1 (0.45)^1 (0.25)^2 = 0.101.$$

(b)

$$P(X_1 = 0, X_2 = 4, X_3 = 0) = \frac{4!}{0!4!0!} (0.3)^0 (0.45)^4 (0.25)^0 = 0.041.$$

(c) The number of people with high income, X_3 , is marginally binomial with $p = 0.25$. Then

$$P(X_3 = 0) = \binom{4}{0} (0.25)^0 (1 - 0.25)^4.$$

2.68

$$\begin{aligned} P(N_1 = n_1, N_2 = n_2, \dots, N_c = n_c | N = n) &= \frac{f(n_1, n_2, \dots, n_c)}{f_N(n)} \\ &= \frac{\frac{e^{-\lambda_1} (\lambda_1)^{n_1}}{n_1!} \dots \frac{e^{-\lambda_c} (\lambda_c)^{n_c}}{n_c!}}{\frac{e^{-\lambda} (\lambda)^n}{n!}} \\ &= \frac{n!}{n_1! n_2! \dots n_c!} \left(\frac{\lambda_1}{\lambda}\right)^{n_1} \left(\frac{\lambda_2}{\lambda}\right)^{n_2} \dots \left(\frac{\lambda_c}{\lambda}\right)^{n_c} \\ &= \frac{n!}{n_1! \dots n_c!} p_1^{n_1} \dots p_c^{n_c}. \end{aligned}$$

2.69

- (a) Out of the first $x - 1$ trials, we need $r - 1$ successes, and then the final trial must be a success. The number of ways to do this is

$$\binom{x-1}{r-1}.$$

Then since there are r successes and $x - r$ failures,

$$f(x) = \binom{x-1}{r-1} p^r (1-p)^{x-r}, \text{ for } x = r, r+1, \dots$$

- (b) Let X_i be i.i.d. geometric random variables with success probability p . Then we obtain r successes by letting $X = X_1 + \dots + X_r$. Then

$$E(X) = \sum_{i=1}^r E(X_i) = r \times \frac{1}{p}$$

and

$$\text{Var}(X) = \sum_{i=1}^r V(X_i) = r \times \frac{1-p}{p^2}.$$

- 2.70** (a) Let E denote the event that the Eastern Conference team wins the series. Then

$$P(E) = P(E, 4 \text{ games}) + P(E, 5 \text{ games}) + P(E, 6 \text{ games}) + P(E, 7 \text{ games}).$$

For the Eastern Conference team to win in j games, it must have won 3 out of the first $j - 1$ games, as well as the last one. So

$$P(E, j \text{ games}) = \binom{j-1}{3} p^4 q^{j-4}$$

and

$$P(E) = \binom{3}{3} p^4 q^0 + \binom{4}{3} p^4 q^1 + \binom{5}{3} p^4 q^2 + \binom{6}{3} p^4 q^3.$$

- (b) Let W denote the event that the Western Conference team wins the series. Then the formula for the probability that the Western Conference team wins in j games is similar to that of the Eastern Conference, but with p and q switched. Then

$$\begin{aligned} P(\text{Series ends in } j \text{ games}) &= P(E, j \text{ games}) + P(W, j \text{ games}) \\ &= \binom{j-1}{3} p^4 q^{j-4} + \binom{j-1}{3} q^4 p^{j-4} \\ &= \binom{j-1}{3} [p^4 q^{j-4} + q^4 p^{j-4}]. \end{aligned}$$

Solutions to Section 2.8

- 2.71** (a)

$$P(90 \leq X \leq 100) = \frac{100 - 90}{100 - 70} = 0.333.$$

(b) To find the 90th percentile, set

$$P(b \leq X \leq 100) = \frac{100 - b}{100 - 70} = 0.1.$$

Solving for b gives $b = 97$.

2.72 (a) If city A is at mile 0 and city C is at mile 75, then city B is at mile 25. Then the probability that the car is towed more than 10 miles is

$$P(10 \leq X \leq 15) + P(35 \leq X \leq 65) = \frac{5}{75} + \frac{30}{75} = 0.467.$$

(b) The probability that the car is towed more than 20 miles is

$$P(45 \leq X \leq 55) = \frac{10}{75} = 0.133.$$

(c)

$$\begin{aligned} P(\text{Towed} > 10 \text{ miles} | X > 20) &= \frac{P(\text{Towed} > 10 \text{ miles} \cap X > 20)}{P(X > 20)} \\ &= \frac{P(35 \leq X \leq 65)}{P(20 \leq X \leq 75)} \\ &= \frac{30/75}{55/75} = 0.545. \end{aligned}$$

2.73 (a) To find the p th quantile, set

$$F(x) = 1 - e^{-0.1x} = p$$

and solve for x , which yields

$$x = -10 \ln(1 - p).$$

For $p = 0.5$, the median is 6.931. For $p = 0.75$, the 75th percentile is 13.863.

(b) There are no jobs in 15 minutes if the arrival time of the first job is above 15 minutes, so

$$P(\text{No jobs in 15 minutes}) = P(X > 15) = 1 - F(15) = e^{-0.1(15)} = 0.223.$$

2.74 (a) If the mean time to failure is 10,000 hours, then $\lambda = 1/10000$. To find the median, set

$$F(x) = 1 - e^{-x/10000} = 0.5$$

and solve for x , which yields

$$x = -10000 \ln(1 - 0.5) = 6931.472.$$

(b)

$$P(X \geq 1000) = 1 - F(1000) = e^{-1000/10000} = 0.905.$$

(c) Because of the memoryless property of the exponential distribution,

$$P(X \geq 2000 | X \geq 1000) = P(X \geq 1000) = 0.905.$$

2.75

Let X_i be the failure time of the i th bulb. Then X_i is exponential with failure rate $\lambda = 1/10000$. Since $T = X_1 + X_2 + \dots + X_5$ is the sum of 5 exponential distributions with failure rate $\lambda = 1/10000$, T has a gamma distribution with the same failure rate $\lambda = 1/10000$ and $r = 5$. Then

$$E(T) = \frac{r}{\lambda} = \frac{5}{1/10000} = 50,000$$

and

$$\text{Var}(T) = \frac{r}{\lambda^2} = \frac{5}{(1/10000)^2} = 500,000,000.$$

2.76 Let X be the beta random variable under study. For $E(X)$ to be $3/4$,

$$\frac{a}{a+b} = \frac{3}{4}$$

or $a = 3b$. For $\text{Var}(X)$ to be $3/32$,

$$\frac{ab}{(a+b)^2(a+b+1)} = \frac{3b^2}{(4b)^2(4b+1)} = \frac{3}{32}$$

or

$$3 \times 32 = 3 \times 16 \times (4b+1)$$

or $b = 1/4$. Then solving for a yields $a = 3/4$.

2.77

$$E(X) = \frac{a}{a+b} = \frac{3}{3+1} = 0.75.$$

$$\text{Var}(X) = \frac{ab}{(a+b)^2(a+b+1)} = \frac{3 \times 1}{(3+1)^2(3+1+1)} = 0.0375.$$

Then solving

$$E(X) = \frac{E(T) - 70}{30}$$

for $E(T)$ yields

$$E(T) = 30E(X) + 70 = 92.5.$$

Similarly, solving

$$\text{Var}(X) = \frac{\text{Var}(T)}{30^2}$$

for $\text{Var}(T)$ yields

$$\text{Var}(T) = 30^2 \text{Var}(X) = 33.75.$$

Solutions to Section 2.9

2.78 (a)

$$P(Z \leq 1.68) = 0.9535,$$

$$P(Z > 0.75) = 1 - 0.7734 = 0.2266,$$

$$P(Z \leq -2.42) = 0.0078,$$

$$P(Z > -1) = 1 - 0.1587 = 0.8413.$$

(b)

$$\begin{aligned}P(1 \leq Z \leq 2) &= 0.9772 - 0.8413 = 0.1359, \\P(-2 \leq Z \leq -1) &= 0.1359, \\P(-1.5 \leq Z \leq 1.5) &= 0.9332 - 0.0668 = 0.8664, \\P(-1 \leq Z \leq 2) &= 0.9772 - 0.1587 = 0.8185.\end{aligned}$$

(c)

$$\begin{aligned}P(Z \leq z_{.1}) &= 1 - 0.1 = 0.9, \\P(Z > -z_{.05}) &= 1 - 0.05 = 0.95, \\P(z_{.25} \leq Z \leq z_{.01}) &= 0.25 - 0.01 = 0.24, \\P(-z_{.25} \leq Z \leq z_{.01}) &= 0.75 - 0.01 = 0.74.\end{aligned}$$

2.79 (a)

$$z_{.3} = 0.525, \quad z_{.15} = 1.04, \quad \text{and} \quad z_{.075} = 1.44.$$

(b) Since $x_p = \mu + z_p\sigma$,

$$\begin{aligned}x_{.3} &= 4 + (0.525) \times 3 = 5.575, \\x_{.15} &= 4 + (1.04) \times 3 = 7.12, \\x_{.075} &= 4 + (1.44) \times 3 = 8.32.\end{aligned}$$

2.80 (a) Let X be the weight of coffee in a can. Then

$$P(X < 16) = P\left(Z = \frac{X - \mu}{\sigma} < \frac{16 - 16.1}{0.5}\right) = P(Z < -0.2) = 0.4207.$$

(b)

$$\begin{aligned}P(16 < X < 16.5) &= P\left(\frac{16 - 16.1}{0.5} < Z = \frac{X - \mu}{\sigma} < \frac{16.5 - 16.1}{0.5}\right) \\&= P(-0.2 < Z < 0.8) = 0.7881 - 0.4207 = 0.3674.\end{aligned}$$

(c) To find the 10th percentile, set

$$P\left(Z \leq \frac{x_{.9} - \mu}{\sigma}\right) = 0.1$$

or

$$x_{.9} = \mu + z_{.9}\sigma = 16.1 + (-1.28) \times 0.5 = 15.46.$$

2.81 (a) W has a normal distribution.

(b)

$$E(W) = E(U) - E(V) = 160 - 120 = 40$$

and

$$\text{Var}(W) = \text{Var}(X) + \text{Var}(Y) = 30^2 + 25^2 = 1525.$$

(c)

$$\begin{aligned}P(U - V > 50) &= P(W > 50) = P\left(Z = \frac{W - \mu}{\sigma} > \frac{50 - 40}{\sqrt{1525}}\right) \\ &= P(Z > 0.256) = 1 - 0.6020 = 0.398.\end{aligned}$$

2.82 (a) $X - Y$ has a normal distribution with

$$E(X - Y) = 0.526 - 0.525 = 0.001 \text{ and}$$

$$\sigma_{X-Y} = \sqrt{(3)^2 + (4)^2} = 5 \times 10^{-4}.$$

(b)

$$P(X - Y > 0) = P\left(Z = \frac{X - Y - \mu}{\sigma} > \frac{0 - 0.001}{5 \times 10^{-4}}\right) = P(Z > -2) = 1 - 0.0227 = 0.9773.$$

(c) Since the number of pairs that fit together has a binomial distribution with $n = 10$ and $p = 0.9772$, this probability is

$$\binom{10}{9}(0.9772)^9(0.0228)^1 + \binom{10}{10}(0.9772)^{10}(0.0228)^0 = 0.185 + 0.794 = 0.979.$$

2.83 (a) \bar{X} has a normal distribution with mean $\mu = 90$ and SD $= \sigma/\sqrt{n} = 20/\sqrt{25} = 4$.

(b)

$$P(\bar{X} > 100) = P\left(Z = \frac{\bar{X} - \mu_{\bar{X}}}{\sigma_{\bar{X}}} > \frac{100 - 90}{4}\right) = P(Z > 2.5) = 0.0062.$$

(c) To find the 90th percentile, set

$$P\left(Z \leq \frac{\bar{x}_{0.10} - \mu_{\bar{X}}}{\sigma_{\bar{X}}}\right) = 0.90$$

or

$$\bar{x}_{0.10} = \mu + z_{0.10}\sigma = 90 + (1.28) \times 4 = 95.12.$$

Solutions to Section 2.10

2.84

$$\begin{aligned}F_Y(y) &= P(Y \leq y) = P\left(\ln \frac{X}{1-X} \leq y\right) \\ &= P\left(X \leq \frac{e^y}{1+e^y}\right) \\ &= F_X\left(\frac{e^y}{1+e^y}\right) \\ &= \frac{e^y}{1+e^y}.\end{aligned}$$

Then

$$f_Y(y) = F'_Y(y) = \frac{e^y}{1+e^y} - \frac{e^{2y}}{(1+e^y)^2} = \frac{e^y + e^{2y} - e^{2y}}{(1+e^y)^2} = \frac{e^y}{(1+e^y)^2}.$$

2.85

$$\begin{aligned}
F_Y(y) &= P(Y \leq y) = P(\tan(X) \leq y) \\
&= P(X \leq \arctan(y)) \\
&= \frac{\arctan(y) - (-\pi/2)}{\pi/2 - (-\pi/2)} \\
&= \frac{\arctan(y) + \pi/2}{\pi}.
\end{aligned}$$

Then

$$f_Y(y) = F_Y'(y) = \frac{1}{\pi(1+y^2)}.$$

2.86 The c.d.f of X is

$$\int_1^x \lambda x^{-(\lambda+1)} dx = -x^{-\lambda} \Big|_1^x = 1 - x^{-\lambda}.$$

Then the c.d.f. of Y is

$$\begin{aligned}
F_Y(y) &= P(Y \leq y) = P(\ln(X) \leq y) \\
&= P(X \leq e^y) \\
&= 1 - (e^y)^{-\lambda} = 1 - e^{-\lambda y}.
\end{aligned}$$

So Y has an exponential distribution with rate parameter λ .

2.87 (a) Solving

$$Y_1 = X_1/X_2 \text{ and } Y_2 = X_2$$

for X_1 and X_2 yields

$$X_1 = Y_1 Y_2 \text{ and } X_2 = Y_2.$$

Then the Jacobian of the transformation is

$$J = \begin{vmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{vmatrix} = \begin{vmatrix} y_2 & y_1 \\ 0 & 1 \end{vmatrix} = y_2.$$

Finally, the p.d.f. is

$$\begin{aligned}
g(y_1, y_2) &= f(\psi_1(y_1, y_2), \psi_2(y_1, y_2)) |J| \\
&= \frac{1}{2\pi} \exp \left\{ -\frac{(y_1 y_2)^2}{2} - \frac{y_2^2}{2} \right\} y_2 \\
&= \frac{y_2}{2\pi} \exp \left\{ -\frac{y_2^2}{2} (1 + y_1^2) \right\}.
\end{aligned}$$

(b)

$$\begin{aligned}
f(y_1) &= \int_{-\infty}^{\infty} \frac{y_2}{2\pi} \exp \left\{ -\frac{y_2^2}{2} (1 + y_1^2) \right\} dy_2 \\
&= 2 \int_0^{\infty} \frac{y_2}{2\pi} \exp \left\{ -\frac{y_2^2}{2} (1 + y_1^2) \right\} dy_2 \\
&= \frac{1}{\pi} \left(\frac{-1}{1 + y_1^2} \right) \exp \left\{ -\frac{y_2^2}{2} (1 + y_1^2) \right\} \Big|_0^{\infty} \\
&= \frac{1}{\pi(1 + y_1^2)} \text{ for } -\infty < y_1 < \infty.
\end{aligned}$$

2.88

Solving

$$Y_1 = \frac{1}{2}(X_1 - X_2) \text{ and } Y_2 = \frac{1}{2}(X_1 + X_2)$$

for X_1 and X_2 yields

$$X_1 = Y_1 + Y_2 \text{ and } X_2 = -Y_1 + Y_2.$$

Then the Jacobian of the transformation is

$$J = \begin{vmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{vmatrix} = \begin{vmatrix} 1 & 1 \\ -1 & 1 \end{vmatrix} = 2.$$

Finally, the p.d.f. is

$$\begin{aligned} g(y_1, y_2) &= f(\psi_1(y_1, y_2), \psi_2(y_1, y_2))|J| \\ &= \frac{1}{2}e^{-(y_1+y_2)/2} \frac{1}{2}e^{-(-y_1+y_2)/2} \times 2 \\ &= \frac{1}{2}e^{-y_2/2}. \end{aligned}$$

This p.d.f. is valid only when $x_1 > 0$ (or $y_1 > -y_2$) and $x_2 > 0$ (or $y_1 < y_2$). So the final domain is $-y_2 < y_1 < y_2$. Then the marginal distribution of Y_2 is

$$f(y_2) = \int_{-y_2}^{y_2} \frac{1}{2}e^{-y_2} dy_1 = \frac{1}{2}y_1 e^{-y_2} \Big|_{-y_2}^{y_2} = y_2 e^{-y_2}, \quad y_2 > 0.$$

Similarly, the marginal distribution of Y_1 is

$$\begin{aligned} f(y_1) &= \int_{|y_1|}^{\infty} \frac{1}{2}e^{-y_2} dy_2 \\ &= \frac{-1}{2}e^{-y_2} \Big|_{|y_1|}^{\infty} \\ &= \frac{1}{2}e^{-|y_1|} \text{ for } -\infty < y_1 < \infty. \end{aligned}$$

Since $f(y_1, y_2) \neq f(y_1)f(y_2)$ they are not independent.

Solutions to Chapter 2 Advanced Exercises

- 2.89** (a) Since there are $n - 1$ husband/wife pairs left after the i th man is paired with his wife, there are $(n - 1)!$ ways to permute the wives (keeping a fixed ordering of the husbands). Then

$$P(A_i) = \frac{(n - 1)!}{n!}.$$

Similarly, after the i th and j th men are paired with their wives, there are $n - 2$ husband/wife pairs left, with $(n - 2)!$ ways to permute them, yielding

$$P(A_i \cap A_j) = \frac{(n - 2)!}{n!}.$$

This reasoning applies for any number of men paired with their wives.

(b) From the inclusion-exclusion formula,

$$\begin{aligned} P(A_1 \cap \dots \cap A_n) &= \sum_i P(A_i) - \sum_{i \neq j} P(A_i \cap A_j) + \dots + (-1)^{n-1} P(A_1 \cap \dots \cap A_n) \\ &= n \frac{(n-1)!}{n!} - \binom{n}{2} \frac{(n-2)!}{n!} + \dots + (-1)^{n-1} \frac{1}{n!} \\ &= 1 - \frac{1}{2!} + \dots + (-1)^{n-1} \frac{1}{n!}. \end{aligned}$$

(c)

$$P(A_1 \cap \dots \cap A_n) = \sum_{i=1}^n \frac{(-1)^{i-1}}{i!}.$$

The Taylor Series expansion for e^x is

$$e^x = 1 + x + \frac{x^2}{2!} + \dots + \frac{x^n}{n!} + \dots$$

Letting $x = -1$ gives

$$e^{-1} = 1 - 1 + \frac{1}{2!} - \frac{1}{3!} + \dots + \frac{(-1)^n}{n!} + \dots$$

and

$$1 - e^{-1} = 1 - 1 + \frac{1}{2!} - \frac{1}{3!} + \dots + \frac{(-1)^{n-1}}{n!} + \dots = \lim_{n \rightarrow \infty} P(A_1 \cap \dots \cap A_n).$$

2.90 (a) For the left matchbox to be emptied on the $(2n - m + 1)$ st trial, out of the first $2n - m$ trials, the left pocket must have been chosen exactly n times. There are

$$\binom{2n - m}{n}$$

ways this could happen, each with probability

$$(1/2)^n (1/2)^{n-m}.$$

Then on the final trial, the left matchbox must be chosen (with probability $1/2$), so the final probability is

$$\binom{2n - m}{n} (1/2)^n (1/2)^{n-m} \times (1/2) = \binom{2n - m}{n} (1/2)^{n+1} (1/2)^{n-m}.$$

(b) The probability that either is found empty with m remaining in the other is just the sum of the probabilities for when the left is emptied and when the right is emptied. Since these are identical, this probability is

$$2 \binom{2n - m}{n} (1/2)^{n+1} (1/2)^{n-m} = \binom{2n - m}{n} (1/2)^{2n-m}.$$

2.91 (a)

$$\begin{aligned}
 p_i &= P(\text{A ruined} | i \text{ dollars}) \\
 &= P(\text{A ruined} | i \text{ dollars, loses next round})P(\text{loses next round}) \\
 &\quad + P(\text{A ruined} | i \text{ dollars, wins next round})P(\text{wins next round}) \\
 &= P(\text{A ruined} | i - 1 \text{ dollars})(1/2) + P(\text{A ruined} | i + 1 \text{ dollars})(1/2) \\
 &= \frac{1}{2}p_{i-1} + \frac{1}{2}p_{i+1}.
 \end{aligned}$$

so

$$p_i/2 - p_{i-1}/2 = p_{i+1}/2 - p_i/2$$

and

$$p_i - p_{i-1} = p_{i+1} - p_i = \delta.$$

(b) The recursive equations are $p_1 = p_0 + \delta = 1 + \delta$, $p_2 = p_1 + \delta = 1 + 2\delta$, and so forth, so that $p_i = 1 + i\delta$. Then at stage $a + b$,

$$0 = p_{a+b} = 1 + (a + b)\delta$$

or $\delta = -1/(a + b)$. Then

$$p_a = 1 + a\delta = 1 - \frac{a}{a + b} = \frac{b}{a + b}.$$

(c)

$$\begin{aligned}
 p_i &= P(\text{B ruined} | i \text{ dollars}) \\
 &= P(\text{B ruined} | i \text{ dollars, loses next round})P(\text{loses next round}) \\
 &\quad + P(\text{B ruined} | i \text{ dollars, wins next round})P(\text{wins next round}) \\
 &= P(\text{B ruined} | i - 1 \text{ dollars})(1/2) + P(\text{B ruined} | i + 1 \text{ dollars})(1/2) \\
 &= \frac{1}{2}p_{i-1} + \frac{1}{2}p_{i+1}.
 \end{aligned}$$

so

$$p_i - p_{i-1} = p_{i+1} - p_i = \delta.$$

The recursive equations are $p_1 = p_0 + \delta = 1 + \delta$, $p_2 = p_1 + \delta = 1 + 2\delta$, and so forth, so that $p_i = 1 + i\delta$. Then at stage $a + b$,

$$0 = p_{a+b} = 1 + (a + b)\delta$$

or $\delta = -1/(a + b)$. Then

$$p_b = 1 + b\delta = 1 - \frac{b}{a + b} = \frac{a}{a + b}.$$

Since $p_a + p_b = 1$, there is no probability that the game will go on forever.

2.92 (a)

$$f(x) = f(x|Y = 1)P(Y = 1) + f(x|Y = 2)P(Y = 2) = p_1f_1(x) + p_2f_2(x).$$

(b)

$$\begin{aligned} E(X) &= \int_x x f(x) dx \\ &= \int_x x(p_1 f_1(x) + p_2 f_2(x)) dx \\ &= p_1 \int_x x f_1(x) dx + p_2 \int_x x f_2(x) dx \\ &= p_1 \mu_1 + p_2 \mu_2. \end{aligned}$$

(c)

$$\begin{aligned} E(X^2) &= \int_x x^2 f(x) dx \\ &= \int_x x^2(p_1 f_1(x) + p_2 f_2(x)) dx \\ &= p_1 \int_x x^2 f_1(x) dx + p_2 \int_x x^2 f_2(x) dx \\ &= p_1(\sigma_1^2 + \mu_1^2) + p_2(\sigma_2^2 + \mu_2^2). \end{aligned}$$

Then

$$\sigma^2 = p_1 \sigma_1^2 + p_2 \sigma_2^2 + p_1 \mu_1^2 + p_2 \mu_2^2 - (p_1 \mu_1 + p_2 \mu_2)^2.$$

2.93 (a) Using

$$E(X) = E[E(X|Y)],$$

and

$$E(X^2) = \text{Var}(X) + [E(X)]^2,$$

then

$$\begin{aligned} \text{Var}(X) &= E(X^2) - [E(X)]^2 \\ &= E[E(X^2|Y)] - [E(E(X|Y))]^2 \\ &= E[\text{Var}(X|Y) + [E(X|Y)]^2] - \{E[E(X|Y)^2] - \text{Var}[E(X|Y)]\} \\ &= E[\text{Var}(X|Y)] + \text{Var}[E(X|Y)]. \end{aligned}$$

(b) Let $1\{\cdot\}$ denote an indicator variable which takes on the value of 1 if the condition inside the braces is true. Since $E(1\{\cdot\}) = P(\cdot)$, the probability of the condition,

$$\begin{aligned} \text{Var}(X) &= E[\text{Var}(X|Y)] + \text{Var}[E(X|Y)] \\ &= E[\sigma_1^2 1\{Y=1\} + \sigma_2^2 1\{Y=2\}] + \text{Var}[\mu_1 1\{Y=1\} + \mu_2 1\{Y=2\}] \\ &= \sigma_1^2 p_1 + \sigma_2^2 p_2 + \mu_1^2 p_1(1-p_1) + \mu_2^2 p_2(1-p_2) + 2\mu_1 \mu_2 \text{Cov}(1\{Y=1\}1\{Y=2\}) \\ &= p_1 \sigma_1^2 + p_2 \sigma_2^2 + p_1 \mu_1^2 + p_2 \mu_2^2 - (\mu_1 p_1)^2 - (\mu_2 p_2)^2 + 2\mu_1 \mu_2 p_1 p_2 \\ &= p_1 \sigma_1^2 + p_2 \sigma_2^2 + p_1 \mu_1^2 + p_2 \mu_2^2 - (p_1 \mu_1 + p_2 \mu_2)^2. \end{aligned}$$

2.94 (a)

$$\begin{aligned} \{N_t = n\} &\Leftrightarrow n \text{ events have occurred by time } t \\ &\Leftrightarrow \text{the total time for } n \text{ events is } \leq t \text{ and the total time for } n+1 \text{ events is } > t \\ &\Leftrightarrow \{X \leq t, X + T_{n+1} > t\} \end{aligned}$$

(b)

$$\begin{aligned}P(N_t = n) &= P(X \leq t, X + T_{n+1} > t) \\&= \int_0^t P(X \in [x, x + dx])P(T_{n+1} > t - x) dx \\&= \int_0^t \frac{\lambda^n x^{n-1} e^{-\lambda x}}{\Gamma(n)} e^{-\lambda(t-x)} dx \\&= \int_0^t \frac{\lambda^n x^{n-1} e^{-\lambda t}}{\Gamma(n)} dx \\&= \frac{\lambda^n e^{-\lambda t}}{\Gamma(n)} \frac{x^n}{n} \Big|_0^t \\&= \frac{(\lambda t)^n e^{-\lambda t}}{n!},\end{aligned}$$

since $\Gamma(n) = (n - 1)!$. So N_t is Poisson(λt).

2.95 (a)

$$\begin{aligned}P(X \leq t) &= \int_0^t f_X(x) dx \\&= \int_0^t \frac{\lambda^n x^{n-1} e^{-\lambda x}}{\Gamma(n)} dx \\&= \frac{\lambda^n}{\Gamma(n)} \int_0^t x^{n-1} e^{-\lambda x} dx.\end{aligned}$$

Also, $X \leq t$ if and only if the number of events at time t is $\geq n$, so

$$P(X \leq t) = P(N_t \geq n) = \sum_{i=n}^{\infty} \frac{e^{-\lambda t} (\lambda t)^i}{i!}.$$

(b) Using the table of Poisson cumulative probabilities,

$$P(N_{10} \geq 5) = 1 - \sum_{i=0}^4 \frac{e^{-5} 5^i}{i!} = 1 - 0.440 = 0.560.$$

2.96 (a)

$$\begin{aligned}\psi_X(t|N) &= E(e^{tX}|N) = E(e^{t(Y_1+Y_2+\dots+Y_N)}|N) \\&= E(e^{tY_1}) \dots E(e^{tY_N}) \text{ because independent} \\&= E(e^{tY_i})^N \text{ because identically distributed} \\&= (pe^{t(1)} + qe^{t(0)})^N \\&= (pe^t + q)^N.\end{aligned}$$

(b)

$$\begin{aligned} E(e^{tX}) &= E[E(e^{tX} | N)] \\ &= E[(pe^t + q)^N] \\ &= \sum_{N=0}^{\infty} (pe^t + q)^N \frac{e^{-\lambda} \lambda^N}{N!} \\ &= \frac{e^{-\lambda}}{e^{-\lambda(pe^t + q)}} \sum_{N=0}^{\infty} \frac{e^{-\lambda(pe^t + q)} [\lambda(pe^t + q)]^N}{N!} \\ &= \exp\{-\lambda + \lambda pe^t + \lambda(1 - p)\} \times 1 \\ &= \exp\{-\lambda p(e^t - 1)\} \end{aligned}$$

Note that the sum reduces to 1 because it is the sum of the Poisson p.m.f. over the entire range. This unconditional m.g.f. is the m.g.f. of a Poisson(λp) random variable.

2.97 (a) Since the m.g.f. of X is

$$M_X(t) = E(e^{tX}) = E(e^{t \ln T}) = E(T^t),$$

then

$$E(T) = M_X(1) = e^{\mu(1) + \frac{\sigma^2}{2}(1)^2} = e^{\mu + \sigma^2/2}$$

and

$$E(T^2) = M_X(2) = e^{\mu(2) + \frac{\sigma^2}{2}(2)^2} = e^{2\mu + 2\sigma^2}.$$

The variance is then

$$\text{Var}(T) = E(T^2) - E(T)^2 = e^{2\mu + 2\sigma^2} - e^{2\mu + \sigma^2} = e^{2\mu + \sigma^2} (e^{\sigma^2} - 1).$$

(b) Since

$$Y = \log_e T = \log_e 10 \times \log_1 0T = \log_e 10 \times X,$$

then

$$Y \sim N(\log_e 10 \times 4, (\log_e 10)^2 \times (0.5)^2).$$

Then by the results of (a),

$$E(T) = \exp\{\mu_Y + \frac{1}{2}\sigma_Y^2\} = \exp\{4 \log_e 10 + \frac{1}{2}(0.5 \log_e 10)^2\} = 19,400.$$