# Solutions Manual for <br> Statistical Inference, Second Edition 

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"When I hear you give your reasons," I remarked, "the thing always appears to me to be so ridiculously simple that I could easily do it myself, though at each successive instance of your reasoning I am baffled until you explain your process."

Dr. Watson to Sherlock Holmes<br>A Scandal in Bohemia

### 0.1 Description

This solutions manual contains solutions for all odd numbered problems plus a large number of solutions for even numbered problems. Of the 624 exercises in Statistical Inference, Second Edition, this manual gives solutions for $484(78 \%)$ of them. There is an obtuse pattern as to which solutions were included in this manual. We assembled all of the solutions that we had from the first edition, and filled in so that all odd-numbered problems were done. In the passage from the first to the second edition, problems were shuffled with no attention paid to numbering (hence no attention paid to minimize the new effort), but rather we tried to put the problems in logical order.

A major change from the first edition is the use of the computer, both symbolically through Mathematica ${ }^{t m}$ and numerically using $R$. Some solutions are given as code in either of these languages. Mathematica ${ }^{t m}$ can be purchased from Wolfram Research, and $R$ is a free đownload from http://www.r-project.org/.

Here is a detailed listing of the solutions included.


Many people contributed to the assembly of this solutions manual. We again thank all of those who contributed solutions to the first edition - many problems have carried over into the second edition. Moreover, throughout the years a number of people have been in constant touch with us, contributing to both the presentations and solutions. We apologize in advance for those we forget to mention, and we especially thank Jay Beder, Yong Sung Joo, Michael Perlman, Rob Strawderman, and Tom Wehrly. Thank you all for your help.

And, as we said the first time around, although we have benefited greatly from the assistance and
comments of others in the assembly of this manual, we are responsible for its ultimate correctness. To this end, we have tried our best but, as a wise man once said, "You pays your money and you takes your chances."

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## Chapter 1

## Probability Theory

"If any little problem comes your way, I shall be happy, if I can, to give you a hint or two as to its solution."
1.1 a. Each sample point describes the result of the toss (H or T ) for each of the four tosses. So, for example THTT denotes T on 1 st, H on $2 \mathrm{nd}, \mathrm{T}$ on 3 rd and T on 4 th. There are $2^{4}=16$ such sample points.
b. The number of damaged leaves is a nonnegative integer. So we might use $S=\{0,1,2, \ldots\}$.
c. We might observe fractions of an hour. So we might use $S=\{t: t \geq 0\}$, that is, the half infinite interval $[0, \infty)$.
d. Suppose we weigh the rats in ounces. The weight must be greater than zero so we might use $S=(0, \infty)$. If we know no 10-day-old rat weighs more than 100 oz., we could use $S=(0,100]$.
e. If $n$ is the number of items in the shipment, then $S=\{0 / n, 1 / n, \ldots, 1\}$.
1.2 For each of these equalities, you must show containment in both directions.
a. $x \in A \backslash B \Leftrightarrow x \in A$ and $x \notin B \Leftrightarrow x \in A$ and $x \notin A \cap B \Leftrightarrow x \in A \backslash(A \cap B)$. Also, $x \in A$ and $x \notin B \Leftrightarrow x \in A$ and $x \in B^{c} \Leftrightarrow x \in A \cap B^{c}$
b. Suppose $x \in B$. Then either $x \in A$ or $x \in A^{c}$. If $x \in A$, then $x \in B \cap A$, and, hence $x \in(B \cap A) \cup\left(B \cap A^{c}\right)$. Thus $B \subset(B \cap A) \cup\left(B \cap A^{c}\right)$. Now suppose $x \in(B \cap A) \cup\left(B \cap A^{c}\right)$. Then either $x \in(B \cap A)$ or $x \in\left(B \cap A^{c}\right)$. If $x \in(B \cap A)$, then $x \in B$. If $x \in\left(B \cap A^{c}\right)$, then $x \in B$. Thus $(B \cap A) \cup\left(B \cap A^{c}\right) \subset B$. Since the containment goes both ways, we have $B=(B \cap A) \cup\left(B \cap A^{c}\right)$. (Note, a more straightforward argument for this part simply uses the Distributive Law to state that $\left.(B \cap A) \cup\left(B \cap A^{c}\right)=B \cap\left(A \cup A^{c}\right)=B \cap S=B.\right)$
c. Similar to part a).
d. From part b)
$A \cup B=A \cup\left[(B \cap A) \cup\left(B \cap A^{c}\right)\right]=A \cup(B \cap A) \cup A \cup\left(B \cap A^{c}\right)=A \cup\left[A \cup\left(B \cap A^{c}\right)\right]=$ $A \cup\left(B \cap A^{c}\right)$.
1.3 a. $x \in A \cup B \Leftrightarrow x \in A$ or $x \in B \Leftrightarrow x \in B \cup A$
$x \in A \cap B \Leftrightarrow x \in A$ and $x \in B \Leftrightarrow x \in B \cap A$.
b. $x \in A \cup(B \cup C) \Leftrightarrow x \in A$ or $x \in B \cup C \Leftrightarrow x \in A \cup B$ or $x \in C \Leftrightarrow x \in(A \cup B) \cup C$.
(It can similarly be shown that $A \cup(B \cup C)=(A \cup C) \cup B$.) $x \in A \cap(B \cap C) \Leftrightarrow x \in A$ and $x \in B$ and $x \in C \Leftrightarrow x \in(A \cap B) \cap C$.
c. $x \in(A \cup B)^{c} \Leftrightarrow x \notin A$ or $x \notin B \Leftrightarrow x \in A^{c}$ and $x \in B^{c} \Leftrightarrow x \in A^{c} \cap B^{c}$ $x \in(A \cap B)^{c} \Leftrightarrow x \notin A \cap B \Leftrightarrow x \notin A$ and $x \notin B \Leftrightarrow x \in A^{c}$ or $x \in B^{c} \Leftrightarrow x \in A^{c} \cup B^{c}$.
1.4 a. " $A$ or $B$ or both" is $A \cup B$. From Theorem 1.2 .9 b we have $P(A \cup B)=P(A)+P(B)-P(A \cap B)$.
b. " $A$ or $B$ but not both" is $\left(A \cap B^{c}\right) \cup\left(B \cap A^{c}\right)$. Thus we have

$$
\begin{array}{rlr}
P\left(\left(A \cap B^{c}\right) \cup\left(B \cap A^{c}\right)\right) & =P\left(A \cap B^{c}\right)+P\left(B \cap A^{c}\right) & \text { (disjoint union) } \\
& =[P(A)-P(A \cap B)]+[P(B)-P(A \cap B)] \quad \text { (Theorem1.2.9a) } \\
& =P(A)+P(B)-2 P(A \cap B) . & \tag{Theorem1.2.9a}
\end{array}
$$

c. "At least one of $A$ or $B$ " is $A \cup B$. So we get the same answer as in a).
d. "At most one of $A$ or $B$ " is $(A \cap B)^{c}$, and $P\left((A \cap B)^{c}\right)=1-P(A \cap B)$.
1.5 a. $A \cap B \cap C=\{$ a U.S. birth results in identical twins that are female $\}$
b. $P(A \cap B \cap C)=\frac{1}{90} \times \frac{1}{3} \times \frac{1}{2}$
1.6

$$
\begin{aligned}
& p_{0}=(1-u)(1-w), \quad p_{1}=u(1-w)+w(1-u), \quad p_{2}=u w \\
& p_{0}=p_{2} \Rightarrow u+w=1 \\
& p_{1}=p_{2} \Rightarrow u w=1 / 3
\end{aligned}
$$

These two equations imply $u(1-u)=1 / 3$, which has no solution in the real numbers. Thus, the probability assignment is not legitimate.
1.7 a.

$$
P(\text { scoring } i \text { points })= \begin{cases}1-\frac{\pi r^{2}}{A} & \text { if } i=0 \\ \frac{\pi r^{2}}{A}\left[\frac{(6-i)^{2}-(5-i)^{2}}{5^{2}}\right] & \text { if } i=1, \ldots, 5 .\end{cases}
$$

b.


Therefore,

$$
P(\text { scoring } i \text { points } \mid \text { board is hit })=\frac{(6-i)^{2}-(5-i)^{2}}{5^{2}} \quad i=1, \ldots, 5
$$

which is exactly the probability distribution of Example 1.2.7.
1.8 a. $P($ scoring exactly $i$ points $)=P($ inside circle $i)-P($ inside circle $i+1)$. Circle $i$ has radius $(6-i) r / 5$,
$P($ sscoring exactly $i$ points $)=\frac{\pi(6-i)^{2} r^{2}}{5^{2} \pi r^{2}}-\frac{\pi((6-(i+1)))^{2} r^{2}}{5^{2} \pi r^{2}}=\frac{(6-i)^{2}-(5-i)^{2}}{5^{2}}$.
b. Expanding the squares in part a) we find $P$ (scoring exactly $i$ points $)=\frac{11-2 i}{25}$, which is decreasing in $i$.
c. Let $P(i)=\frac{11-2 i}{25}$. Since $i \leq 5, P(i) \geq 0$ for all $i$. $P(S)=P$ (hitting the dartboard) $=1$ by definition. Lastly, $P(i \cup j)=$ area of $i$ ring + area of $j$ ring $=P(i)+P(j)$.
1.9 a. Suppose $x \in\left(\cup_{\alpha} A_{\alpha}\right)^{c}$, by the definition of complement $x \notin \cup_{\alpha} A_{\alpha}$, that is $x \notin A_{\alpha}$ for all $\alpha \in \Gamma$. Therefore $x \in A_{\alpha}^{c}$ for all $\alpha \in \Gamma$. Thus $x \in \cap_{\alpha} A_{\alpha}^{c}$ and, by the definition of intersection $x \in A_{\alpha}^{c}$ for all $\alpha \in \Gamma$. By the definition of complement $x \notin A_{\alpha}$ for all $\alpha \in \Gamma$. Therefore $x \notin \cup_{\alpha} A_{\alpha}$. Thus $x \in\left(\cup_{\alpha} A_{\alpha}\right)^{c}$.
b. Suppose $x \in\left(\cap_{\alpha} A_{\alpha}\right)^{c}$, by the definition of complement $x \notin\left(\cap_{\alpha} A_{\alpha}\right)$. Therefore $x \notin A_{\alpha}$ for some $\alpha \in \Gamma$. Therefore $x \in A_{\alpha}^{c}$ for some $\alpha \in \Gamma$. Thus $x \in \cup_{\alpha} A_{\alpha}^{c}$ and, by the definition of union, $x \in A_{\alpha}^{c}$ for some $\alpha \in \Gamma$. Therefore $x \notin A_{\alpha}$ for some $\alpha \in \Gamma$. Therefore $x \notin \cap_{\alpha} A_{\alpha}$. Thus $x \in\left(\cap_{\alpha} A_{\alpha}\right)^{c}$.
1.10 For $A_{1}, \ldots, A_{n}$

$$
\text { (i) }\left(\bigcup_{i=1}^{n} A_{i}\right)^{c}=\bigcap_{i=1}^{n} A_{i}^{c} \quad \text { (ii) }\left(\bigcap_{i=1}^{n} A_{i}\right)^{c}=\bigcup_{i=1}^{n} A_{i}^{c}
$$

Proof of $(i)$ : If $x \in\left(\cup A_{i}\right)^{c}$, then $x \notin \cup A_{i}$. That implies $x \notin A_{i}$ for any $i$, so $x \in A_{i}^{c}$ for every $i$ and $x \in \cap A_{i}$.
Proof of (ii): If $x \in\left(\cap A_{i}\right)^{c}$, then $x \notin \cap A_{i}$. That implies $x \in A_{i}^{c}$ for some $i$, so $x \in \cup A_{i}^{c}$
1.11 We must verify each of the three properties in Definition 1.2.1.
a. (1) The empty set $\emptyset \in\{\emptyset, S\}$. Thus $\emptyset \in \mathcal{B}$. (2) $\emptyset^{c}=S \in \mathcal{B}$ and $S^{c}=\emptyset \in \mathcal{B}$. (3) $\emptyset \cup S=S \in \mathcal{B}$.
b. (1) The empty set $\emptyset$ is a subset of any set, in particular, $\emptyset \subset S$. Thus $\emptyset \in \mathcal{B}$. (2) If $A \in \mathcal{B}$, then $A \subset S$. By the definition of complementation, $A^{c}$ is also a subset of $S$, and, hence, $A^{c} \in \mathcal{B}$. (3) If $A_{1}, A_{2}, \ldots \in \mathcal{B}$, then, for each $i, A_{i} \subset S$. By the definition of union, $\cup A_{i} \subset S$. Hence, $\cup A_{i} \in \mathcal{B}$.
c. Let $\mathcal{B}_{1}$ and $\mathcal{B}_{2}$ be the two sigma algebras. (1) $\emptyset \in \mathcal{B}_{1}$ and $\emptyset \in \mathcal{B}_{2}$ since $\mathcal{B}_{1}$ and $\mathcal{B}_{2}$ are sigma algebras. Thus $\emptyset \in \mathcal{B}_{1} \cap \mathcal{B}_{2}$. (2) If $A \in \mathcal{B}_{1} \cap \mathcal{B}_{2}$, then $A \in \mathcal{B}_{1}$ and $A \in \mathcal{B}_{2}$. Since $\mathcal{B}_{1}$ and $\mathcal{B}_{2}$ are both sigma algebra $A^{c} \in \mathcal{B}_{1}$ and $A^{c} \in \mathcal{B}_{2}$. Therefore $A^{c} \in \mathcal{B}_{1} \cap \mathcal{B}_{2}$. (3) If $A_{1}, A_{2}, \ldots \in \mathcal{B}_{1} \cap \mathcal{B}_{2}$, then $A_{1}, A_{2}, \ldots \in \mathcal{B}_{1}$ and $A_{1}, A_{2}, \ldots \in \mathcal{B}_{2}$. Therefore, since $\mathcal{B}_{1}$ and $\mathcal{B}_{2}$ are both sigma algebra, $\cup_{i=1}^{\infty} A_{i} \in \mathcal{B}_{1}$ and $\cup_{i=1}^{\infty} A_{i} \in \mathcal{B}_{2}$. Thus $\cup_{i=1}^{\infty} A_{i} \in \mathcal{B}_{1} \cap \mathcal{B}_{2}$.
1.12 First write

$$
\begin{array}{rlr}
P\left(\bigcup_{i=1}^{\infty} A_{i}\right) & =P\left(\bigcup_{i=1}^{n} A_{i} \cup \bigcup_{i=n+1}^{\infty} A_{i}\right) \\
& =P\left(\bigcup_{i=1}^{n} A_{i}\right)+P\left(\bigcup_{i=n+1}^{\infty} A_{i}\right) \quad & \left(A_{i}\right. \text { s are disjoint) } \\
& =\sum_{i=1}^{n} P\left(A_{i}\right)+P\left(\bigcup_{i=n+1}^{\infty} A_{i}\right) \quad \text { (finite additivity) }
\end{array}
$$

Now define $B_{k}=\bigcup_{i=k}^{\infty} A_{i}$. Note that $B_{k+1} \subset B_{k}$ and $B_{k} \rightarrow \phi$ as $k \rightarrow \infty$. (Otherwise the sum of the probabilities yould be infinite.) Thus

$$
P\left(\bigcup_{i=1}^{\infty} A_{i}\right)=\lim _{n \rightarrow \infty} P\left(\bigcup_{i=1}^{\infty} A_{i}\right)=\lim _{n \rightarrow \infty}\left[\sum_{i=1}^{n} P\left(A_{i}\right)+P\left(B_{n+1}\right)\right]=\sum_{i=1}^{\infty} P\left(A_{i}\right)
$$

1.13 If $A$ and $B$ are disjoint, $P(A \cup B)=P(A)+P(B)=\frac{1}{3}+\frac{3}{4}=\frac{13}{12}$, which is impossible. More generally, if $A$ and $B$ are disjoint, then $A \subset B^{c}$ and $P(A) \leq P\left(B^{c}\right)$. But here $P(A)>P\left(B^{c}\right)$, so $A$ and $B$ cannot be disjoint.
1.14 If $S=\left\{s_{1}, \ldots, s_{n}\right\}$, then any subset of $S$ can be constructed by either including or excluding $s_{i}$, for each $i$. Thus there are $2^{n}$ possible choices.
1.15 Proof by induction. The proof for $k=2$ is given after Theorem 1.2.14. Assume true for $k$, that is, the entire job can be done in $n_{1} \times n_{2} \times \cdots \times n_{k}$ ways. For $k+1$, the $k+1$ th task can be done in $n_{k+1}$ ways, and for each one of these ways we can complete the job by performing
the remaining $k$ tasks. Thus for each of the $n_{k+1}$ we have $n_{1} \times n_{2} \times \cdots \times n_{k}$ ways of completing the job by the induction hypothesis. Thus, the number of ways we can do the job is $\underbrace{\left(1 \times\left(n_{1} \times n_{2} \times \cdots \times n_{k}\right)\right)+\cdots+\left(1 \times\left(n_{1} \times n_{2} \times \cdots \times n_{k}\right)\right)}_{n_{k+1} \text { terms }}=n_{1} \times n_{2} \times \cdots \times n_{k} \times n_{k+1}$.
1.16 a) $26^{3}$. b) $26^{3}+26^{2}$. c) $26^{4}+26^{3}+26^{2}$.
1.17 There are $\binom{n}{2}=n(n-1) / 2$ pieces on which the two numbers do not match. (Choose 2 out of $n$ numbers without replacement.) There are $n$ pieces on which the two numbers match. So the total number of different pieces is $n+n(n-1) / 2=n(n+1) / 2$.
1.18 The probability is $\frac{\binom{n}{2} n!}{n^{n}}=\frac{(n-1)(n-1)!}{2 n^{n-2}}$. There are many ways to obtain this. Here is one. The denominator is $n^{n}$ because this is the number of ways to place $n$ balls in $n$ cells. The numerator is the number of ways of placing the balls such that exactly one cell is empty. There are $n$ ways to specify the empty cell. There are $n-1$ ways of choosing the cell with two balls. There are $\binom{n}{2}$ ways of picking the 2 balls to go into this cell. And there are $(n-2)$ ! ways of placing the remaining $n-2$ balls into the $n-2$ cells, one ball in each cell. The product of these is the numerator $n(n-1)\binom{n}{2}(n-2)!=\binom{n}{2} n$ !.
1.19 a. $\binom{6}{4}=15$.
b. Think of the $n$ variables as $n$ bins. Differentiating with respect to one of the variables is equivalent to putting a ball in the bin. Thus there are $r$ unlabeled balls to be placed in $n$ unlabeled bins, and there are $\binom{n+r-1}{r}$ ways to do this.
1.20 A sample point specifies on which day ( 1 through 7) each of the 12 calls happens. Thus there are $7^{12}$ equally likely sample points. There are several different ways that the calls might be assigned so that there is at least one call each day. There might be 6 calls one day and 1 call each of the other days. Denote this by 6111111 . The number of sample points with this pattern is $7\binom{12}{6} 6$ !. There are 7 ways to specify the day with 6 calls. There are $\binom{12}{6}$ to specify which of the 12 calls are on this day. And there are 6 ! ways of assigning the remaining 6 calls to the remaining 6 days. We will now count another pattern. There might be 4 calls on one day, 2 calls on each of two days, and 1 call on each of the remaining four days. Denote this by 4221111. The number of sample points with this pattern is $7\binom{12}{4}\binom{6}{2}\binom{8}{2}\binom{6}{2} 4$ !. ( 7 ways to pick day with 4 calls, $\binom{12}{4}$ to pick the calls for that day, $\binom{6}{2}$ to pick two days with two calls, $\binom{8}{2}$ ways to pick two calls for lowered numbered day, $\binom{6}{2}$ ways to pick the two calls for higher numbered day, 4! ways to order remaining 4 calls.) Here is a list of all the possibilities and the counts of the sample points for each one.


The probability is the total number of sample points divided by $7^{12}$, which is $\frac{3,162,075,840}{7^{12}} \approx$ .2285 .
1.21 The probability is $\frac{\binom{n}{2 r} 2^{2 r}}{\binom{2 n}{2 r}}$. There are $\binom{2 n}{2 r}$ ways of choosing $2 r$ shoes from a total of $2 n$ shoes. Thus there are $\binom{2 n}{2 r}$ equally likely sample points. The numerator is the number of sample points for which there will be no matching pair. There are $\binom{n}{2 r}$ ways of choosing $2 r$ different shoes
styles. There are two ways of choosing within a given shoe style (left shoe or right shoe), which gives $2^{2 r}$ ways of arranging each one of the $\binom{n}{2 r}$ arrays. The product of this is the numerator $\binom{n}{2 r} 2^{2 r}$.
1.22
a) $\frac{\binom{31}{15}\binom{29}{15}\binom{31}{15}\binom{30}{15} \cdots\binom{31}{15}}{\binom{380}{180}}$
b) $\frac{\frac{336}{366} \frac{335}{365} \ldots \frac{316}{3366}}{\binom{356}{30}}$.
1.23

$$
\begin{aligned}
P(\text { same number of heads }) & =\sum_{x=0}^{n} P\left(1^{s t} \text { tosses } x, 2^{n d} \text { tosses } x\right) \\
& =\sum_{x=0}^{n}\left[\binom{n}{x}\left(\frac{1}{2}\right)^{x}\left(\frac{1}{2}\right)^{n-x}\right]^{2}=\left(\frac{1}{4}\right)^{n} \sum_{x=0}^{n}\binom{n}{x}^{2}
\end{aligned}
$$

1.24 a.

$$
\begin{aligned}
& P(A \text { wins })=\sum_{i=1}^{\infty} P\left(A \text { wins on } i^{\text {th }} \text { toss }\right) \\
& =\frac{1}{2}+\left(\frac{1}{2}\right)^{2} \frac{1}{2}+\left(\frac{1}{2}\right)^{4}\left(\frac{1}{2}\right)+\cdots=\sum_{i=0}^{\infty}\left(\frac{1}{2}\right)^{2 i+1}=2 / 3 . \\
& \text { b. } P(A \text { wins })=p+(1-p)^{2} p+(1-p)^{4} p+\cdots=\sum_{i=0}^{\infty} p(1-p)^{2 i}=\frac{p}{1-(1-p)^{2}} \text {. }
\end{aligned}
$$

c. $\frac{d}{d p}\left(\frac{p}{1-(1-p)^{2}}\right)=\frac{p^{2}}{\left[1-(1-p)^{2}\right]^{2}}>0$. Thus the probability is increasing in $p$, and the minimum is at zero. Using L'Hôpital's rule we find $\lim _{p \rightarrow 0} \frac{p}{1-(1-p)^{2}}=1 / 2$.
1.25 Enumerating the sample space gives $S^{\prime}=\{(B, B),(B, G),(G, B),(G, G)\}$, with each outcome equally likely. Thus $P$ (at least one boy $)=3 / 4$ and $P$ (both are boys $)=1 / 4$, therefore

$$
P(\text { both are boys } \mid \text { at least one boy })=1 / 3
$$

An ambiguity may arise if order is not acknowledged, the space is $S^{\prime}=\{(B, B),(B, G),(G, G)\}$, with each outcome equally likely.
1.27 a. For $n$ odd the proof is straightforward. There are an even number of terms in the sum $(0,1, \cdots, n)$, and $\binom{n}{k}$ and $\binom{n}{n-k}$, which are equal, have opposite signs. Thus, all pairs cancel and the sum is zero. If $n$ is even, use the following identity, which is the basis of Pascal's triangle: For $k>0,\binom{n}{k}=\binom{n-1}{k}+\binom{n-1}{k-1}$. Then, for $n$ even

$$
\begin{aligned}
\sum_{k=0}^{n}(-1)^{k}\binom{n}{k} & =\binom{n}{0}+\sum_{k=1}^{n-1}(-1)^{k}\binom{n}{k}+\binom{n}{n} \\
& =\binom{n}{0}+\binom{n}{n}+\sum_{k=1}^{n-1}(-1)^{k}\left[\binom{n-1}{k}+\binom{n-1}{k-1}\right] \\
& =\binom{n}{0}+\binom{n}{n}-\binom{n-1}{0}-\binom{n-1}{n-1}=0
\end{aligned}
$$

b. Use the fact that for $k>0, k\binom{n}{k}=n\binom{n-1}{k-1}$ to write

$$
\sum_{k=1}^{n} k\binom{n}{k}=n \sum_{k=1}^{n}\binom{n-1}{k-1}=n \sum_{j=0}^{n-1}\binom{n-1}{j}=n 2^{n-1}
$$

c. $\sum_{k=1}^{n}(-1)^{k+1} k\binom{n}{k}=\sum_{k=1}^{n}(-1)^{k+1}\binom{n-1}{k-1}=n \sum_{j=0}^{n-1}(-1)^{j}\binom{n-1}{j}=0$ from part a).
1.28 The average of the two integrals is

$$
\begin{aligned}
{[(n \log n-n)+((n+1) \log (n+1)-n)] / 2 } & =[n \log n+(n+1) \log (n+1)] / 2-n \\
& \approx(n+1 / 2) \log n-n .
\end{aligned}
$$

Let $d_{n}=\log n!-[(n+1 / 2) \log n-n]$, and we want to show that $\lim _{n \rightarrow \infty} m d_{n}=c$, a constant. This would complete the problem, since the desired limit is the exponential of this one. This is accomplished in an indirect way, by working with differences, which avoids dealing with the factorial. Note that

$$
d_{n}-d_{n+1}=\left(n+\frac{1}{2}\right) \log \left(1+\frac{1}{n}\right)-1
$$

Differentiation will show that $\left(\left(n+\frac{1}{2}\right)\right) \log \left(\left(1+\frac{1}{n}\right)\right)$ is increasing in $n$, and has minimum value $(3 / 2) \log 2=1.04$ at $n=1$. Thus $d_{n}-d_{n+1}>0$. Next recall the Taylor expansion of $\log (1+x)=x-x^{2} / 2+x^{3} / 3-x^{4} / 4+\cdots$. The first three terms provide an upper bound on $\log (1+x)$, as the remaining adjacent pairs are negative. Hence

$$
0<d_{n} d_{n+1}<\left(n+\frac{1}{2}\right)\left(\frac{1}{n} \frac{1}{2 n^{2}}+\frac{1}{3 n^{3}}\right)-1=\frac{1}{12 n^{2}}+\frac{1}{6 n^{3}}
$$

It therefore follows, by the comparison test, that the series $\sum_{1}^{\infty} d_{n}-d_{n+1}$ converges. Moreover, the partial sums must approach a limit. Hence, since the sum telescopes,

$$
\lim _{N \rightarrow \infty} \sum_{1}^{N} d_{n}-d_{n+1}=\lim _{N \rightarrow \infty} d_{1}-d_{N+1}=c .
$$

Thus $\lim _{n \rightarrow \infty} d_{n}=d_{1}-c$, a constant.
1.29 a

| Unordered | Ordered |
| :---: | :---: |
| a. $\{4,4,12,12\}$ | (4,4,12,12), (4,12,12,4), (4,12,4,12) |
|  | (12,4,12,4), (12,4,4,12), (12,12,4,4) |
| Unordered | Ordered |
|  | (2,9,9,12), (2,9,12,9), (2,12,9,9), (9,2,9,12) |
| \{2,9,9,12\} | (9,2,12,9), (9,9,2,12), (9,9,12,2), (9,12,2,9) |
|  | $(9,12,9,2),(12,2,9,9),(12,9,2,9),(12,9,9,2)$ |

b. Same as
c. There are $6^{6}$ ordered samples with replacement from $\{1,2,7,8,14,20\}$. The number of ordered samples that would result in $\{2,7,7,8,14,14\}$ is $\frac{6!}{2!2!1!1!}=180$ (See Example 1.2.20). Thus the probability is $\frac{180}{6^{6}}$.
d. If the $k$ objects were distinguishable then there would be $k$ ! possible ordered arrangements. Since we haye $k_{1}, \ldots, k_{m}$ different groups of indistinguishable objects, once the positions of the objects are fixed in the ordered arrangement permutations within objects of the same group won't change the ordered arrangement. There are $k_{1}!k_{2}!\cdots k_{m}$ ! of such permutations for each ordered component. Thus there would be $\frac{k!}{k_{1}!k_{2}!\cdots k_{m}!}$ different ordered components.
e. Think of the $m$ distinct numbers as $m$ bins. Selecting a sample of size $k$, with replacement, is the same as putting $k$ balls in the $m$ bins. This is $(\underset{k}{k+m-1})$, which is the number of distinct bootstrap samples. Note that, to create all of the bootstrap samples, we do not need to know what the original sample was. We only need to know the sample size and the distinct values.
1.31 a . The number of ordered samples drawn with replacement from the set $\left\{x_{1}, \ldots, x_{n}\right\}$ is $n^{n}$. The number of ordered samples that make up the unordered sample $\left\{x_{1}, \ldots, x_{n}\right\}$ is $n$ !. Therefore the outcome with average $\frac{x_{1}+x_{2}+\cdots+x_{n}}{n}$ that is obtained by the unordered sample $\left\{x_{1}, \ldots, x_{n}\right\}$
has probability $\frac{n!}{n^{n}}$. Any other unordered outcome from $\left\{x_{1}, \ldots, x_{n}\right\}$, distinct from the unordered sample $\left\{x_{1}, \ldots, x_{n}\right\}$, will contain m different numbers repeated $k_{1}, \ldots, k_{m}$ times where $k_{1}+k_{2}+\cdots+k_{m}=n$ with at least one of the $k_{i}$ 's satisfying $2 \leq k_{i} \leq n$. The probability of obtaining the corresponding average of such outcome is

$$
\frac{n!}{k_{1}!k_{2}!\cdots k_{m}!n^{n}}<\frac{n!}{n^{n}}, \text { since } k_{1}!k_{2}!\cdots k_{m}!>1
$$

Therefore the outcome with average $\frac{x_{1}+x_{2}+\cdots+x_{n}}{n}$ is the most likely.
b. Stirling's approximation is that, as $n \rightarrow \infty, n!\approx \sqrt{2 \pi} n^{n+(1 / 2)} e^{-n}$, and thus

$$
\left(\frac{n!}{n^{n}}\right) /\left(\frac{\sqrt{2 n \pi}}{e^{n}}\right)=\frac{n!e^{n}}{n^{n} \sqrt{2 n \pi}}=\frac{\sqrt{2 \pi} n^{n+(1 / 2)} e^{-n} e^{n}}{n^{n} \sqrt{2 n \pi}}=1 .
$$

c. Since we are drawing with replacement from the set $\left\{x_{1}, \ldots, x_{n}\right\}$, the probability of choosing any $x_{i}$ is $\frac{1}{n}$. Therefore the probability of obtaining an ordered sample of size $n$ without $x_{i}$ is $\left(1-\frac{1}{n}\right)^{n}$. To prove that $\lim _{n \rightarrow \infty}\left(1-\frac{1}{n}\right)^{n}=e^{-1}$, calculate the limit of the log. That is

$$
\lim _{n \rightarrow \infty} n \log \left(1-\frac{1}{n}\right)=\lim _{n \rightarrow \infty} \frac{\log \left(1-\frac{1}{n}\right)}{1 / n}
$$

L'Hôpital's rule shows that the limit is -1 , establishing the result. See also Lemma 2.3.14.
1.32 This is most easily seen by doing each possibility. Let $P(i)=$ probability that the candidate hired on the $i$ th trial is best. Then

$$
P(1)=\frac{1}{N}, \quad P(2)=\frac{1}{N-1}, \quad \cdots, P(i)=\frac{1}{N-i+1}, \quad \ldots \quad, P(N)=1
$$

1.33 Using Bayes rule

$$
P(M \mid C B)=\frac{P(C B \mid M) P(M)}{P(C B \mid M) P(M)+P(C B \mid F) P(F)}=\frac{.05 \times \frac{1}{2}}{.05 \times \frac{1}{2}+.0025 \times \frac{1}{2}}=.9524
$$

1.34 a.
$P$ (Brown Hair)
$P($ Brown HairyLitter 1$) P($ Litter 1$)+P($ Brown Hair $\mid$ Litter 2$) P($ Litter 2$)$

$$
\left(\frac{2}{3}\right)\left(\frac{1}{2}\right)+\left(\frac{3}{5}\right)\left(\frac{1}{2}\right)=\frac{19}{30} .
$$

## Use Bayes

Theorem
$P($ Litter 1|Brown Hair $)=\frac{P(B H \mid L 1) P(L 1)}{P(B H \mid L 1) P(L 1)+P(B H \mid L 2) P(L 2}=\frac{\left(\frac{2}{3}\right)\left(\frac{1}{2}\right)}{\frac{19}{30}}=\frac{10}{19}$.
1.35 Clearly $P(\cdot \mid B) \geq 0$, and $P(S \mid B)=1$. If $A_{1}, A_{2}, \ldots$ are disjoint, then

$$
\begin{aligned}
P\left(\bigcup_{i=1}^{\infty} A_{i} \mid B\right) & =\frac{P\left(\bigcup_{i=1}^{\infty} A_{i} \cap B\right)}{P(B)}=\frac{P\left(\bigcup_{i=1}^{\infty}\left(A_{i} \cap B\right)\right)}{P(B)} \\
& =\frac{\sum_{i=1}^{\infty} P\left(A_{i} \cap B\right)}{P(B)}=\sum_{i=1}^{\infty} P\left(A_{i} \mid B\right) .
\end{aligned}
$$

