Preface

This solution manual was prepared as an aid for instructors who will benefit by having solutions available. In addition to providing detailed answers to most of the problems in the book, this manual can help the instructor determine which of the problems are most appropriate for the class.

The vast majority of the problems have been solved with the help of available computer software (SAS, S-Plus, Minitab). A few of the problems have been solved with hand calculators. The reader should keep in mind that round-off errors can occur—particularly in those problems involving long chains of arithmetic calculations.

We would like to take this opportunity to acknowledge the contribution of many students, whose homework formed the basis for many of the solutions. In particular, we would like to thank Jorge Achcar, Sebastiao Amorim, W. K. Cheang, S. S. Cho, S. G. Chow, Charles Fleming, Stu Janis, Richard Jones, Tim Kramer, Dennis Murphy, Rich Raubertas, David Steinberg, T. J. Tien, Steve Verrill, Paul Whitney and Mike Wincek. Dianne Hall compiled most of the material needed to make this current solutions manual consistent with the sixth edition of the book.

The solutions are numbered in the same manner as the exercises in the book. Thus, for example, 9.6 refers to the 6th exercise of chapter 9.

We hope this manual is a useful aid for adopters of our Applied Multivariate Statistical Analysis, 6th edition, text. The authors have taken a little more active role in the preparation of the current solutions manual. However, it is inevitable that an error or two has slipped through so please bring remaining errors to our attention. Also, comments and suggestions are always welcome.

Richard A. Johnson
Dean W. Wichern
Chapter 1

1.1
\[
\begin{align*}
\bar{x}_1 &= 4.29 \\
\bar{x}_2 &= 6.29 \\
s_{11} &= 4.20 \\
s_{22} &= 3.56 \\
s_{12} &= 3.70 \\
\end{align*}
\]

1.2 a) Scatter Plot and Marginal Dot Plots

b) \( s_{12} \) is negative

c) \[
\begin{align*}
\bar{x}_1 &= 5.20 \\
\bar{x}_2 &= 12.48 \\
s_{11} &= 3.09 \\
s_{22} &= 5.27 \\
s_{12} &= -15.94 \\
r_{12} &= -.98 \\
\end{align*}
\]

Large \( x_1 \) occurs with small \( x_2 \) and vice versa.

d) \[
\begin{align*}
\bar{x} &= \begin{bmatrix} 5.20 \\ 12.48 \end{bmatrix} \\
S &= \begin{bmatrix} 3.09 & -15.94 \\ -15.94 & 5.27 \end{bmatrix} \\
R &= \begin{bmatrix} 1 & -.98 \\ -.98 & 1 \end{bmatrix}
\end{align*}
\]
1.3

\[ x_1 = \begin{bmatrix} 6 \\ 8 \\ 2 \end{bmatrix} \quad s_n = \begin{bmatrix} 6 & 4 & -1.4 \\ 8 & 1.2 \end{bmatrix} \] (symmetric) 2

\[ R = \begin{bmatrix} 1 & .577 & -.404 \\ 1 & .300 \end{bmatrix} \]

1.4

a) There is a positive correlation between \( x_1 \) and \( x_2 \). Since sample size is small, hard to be definitive about nature of marginal distributions. However, marginal distribution of \( x_1 \) appears to be skewed to the right. The marginal distribution of \( x_2 \) seems reasonably symmetric.

![Scatter Plot and Marginal Dot Plots](image)

b)

\[ \bar{x}_1 = 155.60 \quad \bar{x}_2 = 14.70 \quad s_{11} = 82.03 \quad s_{22} = 4.85 \]

\[ s_{12} = 273.26 \quad r_{12} = .69 \]

Large profits \((x_2)\) tend to be associated with large sales \((x_1)\); small profits with small sales.
1.5 a) There is negative correlation between $x_2$ and $x_3$ and negative correlation between $x_1$ and $x_3$. The marginal distribution of $x_1$ appears to be skewed to the right. The marginal distribution of $x_2$ seems reasonably symmetric. The marginal distribution of $x_3$ also appears to be skewed to the right.
The page contains mathematical expressions and tables. The expressions include mean ($\bar{x}$), standard deviation ($S_n$), and a matrix ($R$), with values such as $155.60$, $14.70$, $273.26$, and $-32018.36$.

### 1.5 b)

\[
\begin{bmatrix}
155.60 \\
14.70 \\
710.91
\end{bmatrix}
= 
\begin{bmatrix}
82.03 & 273.26 & -32018.36 \\
273.26 & 4.85 & -948.45 \\
-32018.36 & -948.45 & 461.90
\end{bmatrix}
\]

\[
R = 
\begin{bmatrix}
1 & .69 & -.85 \\
.69 & 1 & -.42 \\
-.85 & -.42 & 1
\end{bmatrix}
\]

### 1.6 a) Histograms

#### X

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#### X

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</table>
The pair $x_3, x_4$ exhibits a small to moderate positive correlation and so does the pair $x_3, x_5$. Most of the entries are small.
1.8 Using (1-12) \[ d(P, Q) = \sqrt{(-1-1)^2 + (-1-0)^2} = \sqrt{5} = 2.236 \]

Using (1-20) \[ d(P, Q) = \sqrt{\frac{1}{3}(-1-1)^2 + 2\left(\frac{1}{9}\right)(-1-1)(-1-0) + \frac{4}{27}(-1-0)^2} = \sqrt{\frac{52}{27}} = 1.388 \]

Using (1-20) the locus of points a constant squared distance 1 from \( Q = (1, 0) \) is given by the expression \[ \frac{1}{3}(x_{1}-1)^2 + \frac{2}{9}(x_{1}-1)x_{2} + \frac{4}{27}x_{2}^2 = 1. \]

To sketch the locus of points defined by this equation, we first obtain the coordinates of some points satisfying the equation:

\((-1, 1.5), (0, -1.5), (0, 3), (1, -2.6), (1, 2.6), (2, -3), (2, 1.5), (3, -1.5)\)

The resulting ellipse is:

\[\text{Diagram of an ellipse with points plotted.}\]

1.9 a) \[ s_{11} = 20.48 \]

\[ s_{22} = 6.19 \]

\[ s_{12} = 9.09 \]
1.10 a) This equation is of the form (1-19) with  a_{11} = 1, a_{12} = \frac{1}{2} and a_{22} = 4. Therefore this is a distance for correlated variables if it is non-negative for all values of x_1, x_2. But this follows easily if we write
\[ x_1^2 + 4x_2^2 + x_1x_2 = (x_1 + \frac{1}{2}x_2)^2 + \frac{15}{4}x_2^2 \geq 0. \]

b) In order for this expression to be a distance it has to be non-negative for all values x_1, x_2. Since, for (x_1, x_2) = (0,1) we have x_1^2 - 2x_2^2 = -2, we conclude that this is not a valid distance function.

d) \[(\bar{x}_1, \bar{x}_2) = (2.72, -3.55)\]
d(0,P) = 2.72 using (1-17).
e) d(0,P) = 2.72 using (1-19).

1.11 \[
d(P,Q) = \sqrt{4(x_1-y_1)^2 + 2(-1)(x_1-y_1)(x_2-y_2) + (x_2-y_2)^2} \\
= \sqrt{4(y_1-x_1)^2 + 2(-1)(y_1-x_1)(y_2-x_2) + (y_2-x_2)^2} = d(Q,P) \\
\]
Next, \[4(x_1-y_1)^2 - 2(x_1-y_1)(x_2-y_2) + (x_2-y_2)^2 = \]
\[= (x_1-y_1-x_2+y_2)^2 + 3(x_1-y_1)^2 \geq 0 \] so \(d(P,Q) \geq 0.\)

The second term is zero in this last expression only if \(x_1 = y_1\) and then the first is zero only if \(x_2 = y_2.\)
1.12  

a) If \( P = (-3,4) \) then \( d(0,P) = \max (|-3|,|4|) = 4 \)

b) The locus of points whose squared distance from \( (0,0) \) is 1 is

![Diagram of a square with vertices at (1,1), (-1,1), (-1,-1), (1,-1)]

c) The generalization to \( p \)-dimensions is given by \( d(0,P) = \max(|x_1|,|x_2|,...,|x_p|) \).

1.13 Place the facility at C-3.
Strong positive correlation. No obvious "unusual" observations.

b) **Multiple-sclerosis group.**

\[
\begin{pmatrix}
42.07 \\
179.64 \\
12.31 \\
236.62 \\
13.16
\end{pmatrix}
\]

\[
\begin{pmatrix}
116.91 & 61.78 & -20.10 & 61.13 & -27.65 \\
812.72 & 218.35 & 865.32 & 90.48 \\
305.94 & 221.93 & 286.60 \\
1146.38 & 82.53 \\
(symmetric)
\end{pmatrix}
\]
\[
\begin{bmatrix}
1 & .200 & -.106 & .167 & -.139 \\
1 & .438 & .896 & .173 \\
1 & .375 & .892 \\
1 & .133 \\
\end{bmatrix}
\]

\(R = \text{(symmetric)}\)

Non multiple-sclerosis group.

\[
\begin{bmatrix}
37.99 \\
147.21 \\
1.56 \\
195.57 \\
1.62 \\
\end{bmatrix}
\]

\(x = \text{(symmetric)}\)

\[
\begin{bmatrix}
273.61 & 95.08 & 5.28 & 101.67 & 3.20 \\
110.13 & 1.84 & 103.28 & 2.15 \\
1.78 & 2.22 & .49 \\
183.04 & 2.35 \\
\end{bmatrix}
\]

\(S_n = \text{(symmetric)}\)

\[
\begin{bmatrix}
1 & .548 & .239 & .454 & .127 \\
1 & .132 & .727 & .134 \\
1 & .123 & .244 \\
\end{bmatrix}
\]

\(R = \text{(symmetric)}\)
1.15 a) Scatterplot of $x_2$ and $x_3$.

b)

\[
\mathbf{x} = \begin{pmatrix}
3.54 \\
1.81 \\
2.14 \\
2.21 \\
2.58 \\
1.27
\end{pmatrix}
\]
The largest correlation is between appetite and amount of food eaten. Both activity and appetite have moderate positive correlations with symptoms. Also, appetite and activity have a moderate positive correlation.
1.16
There are significant positive correlations among all variables. The lowest correlation is 0.4420 between Dominant humerus and Ulna, and the highest correlation is 0.89365 between Dominant humerus and Humerus.

\[ \mathbf{X} = \begin{pmatrix} 0.8438 \\ 0.8183 \\ 1.7927 \\ 1.7348 \\ 0.7044 \\ 0.6938 \end{pmatrix}, \quad \mathbf{R} = \begin{pmatrix} 1.0000 & 0.85181 & 0.69146 & 0.66826 & 0.74369 & 0.67789 \\ 0.85181 & 1.00000 & 0.61192 & 0.74909 & 0.74218 & 0.80980 \\ 0.69146 & 0.61192 & 1.00000 & 0.89365 & 0.55222 & 0.44020 \\ 1.7927 & 0.74909 & 0.89365 & 1.00000 & 0.62555 & 0.61882 \\ 1.7348 & 0.74218 & 0.55222 & 0.62555 & 1.00000 & 0.72889 \\ 0.7044 & 0.67789 & 0.80980 & 0.44020 & 0.61882 & 1.00000 \end{pmatrix} \]

\[ \mathbf{S}_n = \begin{pmatrix} 0.0124815 \\ 0.0099633 \\ 0.0214560 \\ 0.0192822 \\ 0.0087559 \\ 0.0076395 \end{pmatrix} \]

1.17
There are large positive correlations among all variables. Particularly large correlations occur between running events that are "similar", for example, the 100m and 200m dashes, and the 1500m and 3000m runs.

\[ \mathbf{x} = \begin{pmatrix} 11.36 \\ 23.12 \\ 51.99 \\ 2.02 \\ 4.19 \\ 9.08 \end{pmatrix}, \quad \mathbf{S}_n = \begin{pmatrix} 4.254 & 10.193 & 28.368 & 1.197 & 3.474 & 10.508 \end{pmatrix} \]

\[ \mathbf{R} = \begin{pmatrix} 1.000 & 0.941 & 0.871 & 0.809 & 0.782 & 0.728 & 0.669 \\ 0.941 & 1.000 & 0.909 & 0.820 & 0.801 & 0.732 & 0.680 \\ 0.871 & 0.909 & 1.000 & 0.806 & 0.720 & 0.674 & 0.677 \\ 0.809 & 0.820 & 0.806 & 1.000 & 0.905 & 0.867 & 0.854 \\ 0.782 & 0.801 & 0.720 & 0.905 & 1.000 & 0.973 & 0.791 \\ 0.728 & 0.732 & 0.674 & 0.867 & 0.973 & 1.000 & 0.799 \\ 0.669 & 0.680 & 0.677 & 0.854 & 0.791 & 0.799 & 1.000 \end{pmatrix} \]
There are positive correlations among all variables. Notice the correlations decrease as the distances between pairs of running events increase (see the first column of the correlation matrix $R$). The correlation matrix for running events measured in meters per second is very similar to the correlation matrix for the running event times given in Exercise 1.17.

\[
\begin{bmatrix}
8.81 \\
8.66 \\
7.71 \\
6.60 \\
5.99 \\
5.54 \\
4.62 \\
\end{bmatrix}
\begin{bmatrix}
0.91 & 0.96 & 0.97 & 0.65 & 0.82 & 0.92 & 0.81 \\
0.96 & 0.115 & 0.114 & 0.075 & 0.096 & 0.105 & 0.093 \\
0.97 & 0.114 & 0.138 & 0.081 & 0.095 & 0.108 & 0.102 \\
0.065 & 0.075 & 0.081 & 0.074 & 0.086 & 0.100 & 0.094 \\
0.082 & 0.096 & 0.095 & 0.086 & 0.124 & 0.144 & 0.118 \\
0.092 & 0.105 & 0.108 & 0.100 & 0.144 & 0.177 & 0.147 \\
0.081 & 0.093 & 0.102 & 0.094 & 0.118 & 0.147 & 0.167 \\
\end{bmatrix}
\]

\[
\begin{bmatrix}
1.000 & 0.938 & 0.866 & 0.797 & 0.776 & 0.729 & 0.660 \\
0.938 & 1.000 & 0.906 & 0.816 & 0.806 & 0.741 & 0.675 \\
0.866 & 0.906 & 1.000 & 0.804 & 0.731 & 0.694 & 0.672 \\
0.797 & 0.816 & 0.804 & 1.000 & 0.906 & 0.875 & 0.852 \\
0.776 & 0.806 & 0.731 & 0.906 & 1.000 & 0.972 & 0.824 \\
0.729 & 0.741 & 0.694 & 0.875 & 0.972 & 1.000 & 0.854 \\
0.660 & 0.675 & 0.672 & 0.852 & 0.824 & 0.854 & 1.000 \\
\end{bmatrix}
\]
(a) The plot looks like a cigar shape, but bent. Some observations in the lower left hand part could be outliers. From the highlighted plot in (b) (actually non-bankrupt group not highlighted), there is one outlier in the nonbankrupt group, which is apparently located in the bankrupt group, besides the strung out pattern to the right.

(b) The dotted line in the plot would be an orientation for the classification.
(a) There are two outliers in the upper right and lower right corners of the plot.

(b) Only the points in the gasoline group are highlighted. The observation in the upper right is the outlier. As indicated in the plot, there is an orientation to classify into two groups.
Possible outliers are indicated.
1.23 b) A visual clustering using Chernoff faces is given below.

Cluster 1
- Total

Cluster 2
- Raisin Bran

Cluster 3
- Life

Cluster 4
- Wheaties
- Special K

Cluster 5
- Rice Krispies
- Sugar Corn Pops
- Grape Nuts
- Sugar Snacks
- Puffed Rice
- Super Sugar Crisp
Cluster 1

Cluster 2

Cluster 3
We have clustered these faces in the same manner as those in Example 1.12. Note, however, other groupings are equally plausible. For instance, utilities 9 and 18 might be switched from Cluster 2 to Cluster 3 and so forth.
We illustrate one cluster of "stars". The remaining stars (not shown) can be grouped in 3 or 4 additional clusters.
1.26 Bull data

(a) XBAR

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<td>145.35</td>
<td>16628.94</td>
</tr>
</tbody>
</table>

![Graphs showing relationships between different variables](image-url)
1.27

(a) Correlation $r = .173$

(b) Great Smoky is unusual park. Correlation with this park removed is $r = .391$. This single point has reasonably large effect on correlation reducing the positive correlation by more than half when added to the national park data set.

(c) The correlation coefficient is a dimensionless measure of association. The correlation in (b) would not change if size were measured in square miles instead of acres.