

### 3.4 Exercises: Some Solutions (November 26, 2011)

**3.1.** Let  $\mathbf{U} = (\mathbf{1} : \mathbf{x} : \mathbf{y})$  be a  $100 \times 3$  matrix where  $\mathbf{x}$  and  $\mathbf{y}$  comprise the observed values of variables  $x$  and  $y$ , and assume that  $s_x^2 > 0$  and  $s_y^2 > 0$ .

- (a) Suppose that the columns of  $\mathbf{U}$  are orthogonal. Show that  $r_{xy} = 0$ .  
 (b) Confirm that if  $\text{rank}(\mathbf{U}) = 2$ , then  $r_{xy}^2 = 1$ .

**• SOLUTION TO EX. 3.1:**

(a)  $\mathbf{U}'\mathbf{U} = \text{diag}(n, \mathbf{x}'\mathbf{x}, \mathbf{y}'\mathbf{y}) \implies \mathbf{1}'\mathbf{x} = \mathbf{1}'\mathbf{y} = 0$  and so  $\mathbf{x}$  and  $\mathbf{y}$  are centered. Hence  $r_{xy} = \mathbf{x}'\mathbf{y} / \sqrt{\mathbf{x}'\mathbf{x} \cdot \mathbf{y}'\mathbf{y}}$ , which is 0 because  $\mathbf{x}'\mathbf{y} = 0$ .

(b)  $\text{rank}(\mathbf{U}) < 3 \implies$  there exists  $\mathbf{a} = (\alpha, \beta, \gamma)' \neq \mathbf{0}$  such that

$$\mathbf{U}\mathbf{a} = \alpha\mathbf{1} + \beta\mathbf{x} + \gamma\mathbf{y} = \mathbf{0}. \quad (*)$$

In (\*) necessarily  $\alpha \neq 0$  and  $\beta \neq 0$  because  $s_x^2 > 0$  and  $s_y^2 > 0$ . Hence

$$\mathbf{y} = -\frac{\alpha}{\gamma}\mathbf{1} - \frac{\beta}{\gamma}\mathbf{x} := \mathbf{a}\mathbf{1} + \mathbf{b}\mathbf{x},$$

i.e.,  $y_i = a + bx_i, i = 1, \dots, n$ , which trivially means that  $r_{xy}^2 = 1$ .  $\square$

**3.2.** Let  $\mathbf{x}$  and  $\mathbf{y}$  be  $n$ -dimensional variable vectors (comprising the observed values of  $x$  and  $y$ ). Show that

$$\text{cor}_d(\mathbf{a}\mathbf{1} + \mathbf{b}\mathbf{x}, \mathbf{c}\mathbf{1} + \mathbf{d}\mathbf{y}) = \begin{cases} \text{cor}_d(\mathbf{x}, \mathbf{y}) & \text{if } bd > 0, \\ -\text{cor}_d(\mathbf{x}, \mathbf{y}) & \text{if } bd < 0, \end{cases}$$

or in other notation,

$$\text{cor}_d[(\mathbf{1} : \mathbf{x}) \begin{pmatrix} a \\ b \end{pmatrix}, (\mathbf{1} : \mathbf{y}) \begin{pmatrix} c \\ d \end{pmatrix}] = \begin{cases} r_{xy} & \text{if } bd > 0, \\ -r_{xy} & \text{if } bd < 0. \end{cases}$$

**• SOLUTION TO EX. 3.2:**

Denoting  $\mathbf{u} = \mathbf{a}\mathbf{1} + \mathbf{b}\mathbf{x}$ ,  $\mathbf{v} = \mathbf{c}\mathbf{1} + \mathbf{d}\mathbf{y}$  and  $\mathbf{C}$  = the centering matrix, yields

$$\mathbf{u}'\mathbf{C}\mathbf{v} = bd\mathbf{x}'\mathbf{C}\mathbf{y}, \quad \mathbf{u}'\mathbf{C}\mathbf{u} = b^2\mathbf{x}'\mathbf{C}\mathbf{x}, \quad \mathbf{v}'\mathbf{C}\mathbf{v} = d^2\mathbf{y}'\mathbf{C}\mathbf{y},$$

$$\implies r_{uv} = \frac{bd\mathbf{x}'\mathbf{C}\mathbf{y}}{|b||d|\sqrt{\mathbf{x}'\mathbf{C}\mathbf{x} \cdot \mathbf{y}'\mathbf{C}\mathbf{y}}} = \frac{bd}{|bd|} r_{xy}. \quad \square$$

**3.3.** Consider the  $3 \times 3$  data matrix

$$\mathbf{U} = (\mathbf{1}_3 : \mathbf{x} : \mathbf{y}) = \begin{pmatrix} 1 & x_1 & y_1 \\ 1 & x_2 & y_2 \\ 1 & x_3 & y_3 \end{pmatrix}.$$

Confirm that the area of the triangle with vertices  $(x_i, y_i)$ ,  $i = 1, 2, 3$ , is  $\frac{1}{2}|\det(\mathbf{U})|^*$ . Show that  $\det(\mathbf{U}) = 0 \iff$  the three data points lie in the same line, and hence, assuming that  $x$  and  $y$  have nonzero variances, we have  $r_{xy}^2 = 1 \iff \det(\mathbf{U}) = 0$ .

\* Notice: in the book we have erroneously  $\frac{1}{2} \det(\mathbf{U})$ .

**• SOLUTION TO EX. 3.3:**

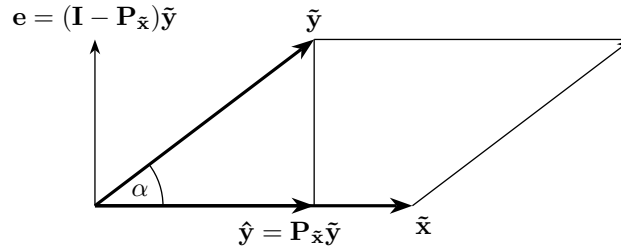
Subtracting the first row from the second and third row shows that

$$\begin{aligned} |\mathbf{U}| &= \begin{vmatrix} 1 & x_1 & y_1 \\ 1 & x_2 & y_2 \\ 1 & x_3 & y_3 \end{vmatrix} = \begin{vmatrix} 1 & x_1 & y_1 \\ 0 & x_2 - x_1 & y_2 - y_1 \\ 0 & x_3 - x_1 & y_3 - y_1 \end{vmatrix} = \begin{vmatrix} x_2 - x_1 & y_2 - y_1 \\ x_3 - x_1 & y_3 - y_1 \end{vmatrix} \\ &= (x_2 - x_1)(y_3 - y_1) - (y_2 - y_1)(x_3 - x_1). \end{aligned} \quad (1)$$

Consider a triangle  $OPG$ , say, in  $\mathbb{R}^2$ , formed by the points  $\mathbf{0} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ ,  $\mathbf{f} = \begin{pmatrix} a_1 \\ b_1 \end{pmatrix}$ , and  $\mathbf{g} = \begin{pmatrix} a_2 \\ b_2 \end{pmatrix}$ . We will next show that

$$\text{area}(OPG) = \frac{1}{2} |\det(\mathbf{L})| = \frac{1}{2} |a_1 b_2 - a_2 b_1|, \quad (2)$$

where  $\mathbf{L} = \begin{pmatrix} a_1 & b_1 \\ a_2 & b_2 \end{pmatrix} = \begin{pmatrix} \mathbf{f}' \\ \mathbf{g}' \end{pmatrix}$ . Proceeding as in Section 5.7, we get



**Figure 3.6** This is Figure 5.1. Replace here  $\tilde{\mathbf{x}}$  and  $\tilde{\mathbf{y}}$  with  $\mathbf{f}$  and  $\mathbf{g}$ , respectively.

$$\text{area}^2(OPG) = \frac{1}{4} \|\mathbf{f}\|^2 \|\mathbf{e}\|^2,$$

where  $\mathbf{e} = (\mathbf{I}_2 - \mathbf{P}_{\mathbf{f}})\mathbf{g}$ . Now we have

$$\begin{aligned}\|\mathbf{e}\|^2 &= \mathbf{g}'(\mathbf{I}_2 - \mathbf{P}_f)\mathbf{g} = \mathbf{g}'[\mathbf{I}_2 - \mathbf{f}(\mathbf{f}'\mathbf{f})^{-1}\mathbf{f}']\mathbf{g} \\ &= \mathbf{g}'\mathbf{g} - \frac{(\mathbf{f}'\mathbf{g})^2}{\mathbf{f}'\mathbf{f}} = \mathbf{g}'\mathbf{g}[1 - \cos^2(\mathbf{f}, \mathbf{g})],\end{aligned}$$

and hence

$$\begin{aligned}\text{area}^2(OFG) &= \frac{1}{4} \mathbf{f}'\mathbf{f} \left[ \mathbf{g}'\mathbf{g} - \frac{(\mathbf{f}'\mathbf{g})^2}{\mathbf{f}'\mathbf{f}} \right] \\ &= \frac{1}{4} [\mathbf{f}'\mathbf{f} \cdot \mathbf{g}'\mathbf{g} - (\mathbf{f}'\mathbf{g})^2] \\ &= \frac{1}{4} |\mathbf{L}\mathbf{L}'| = \frac{1}{4} |\mathbf{L}|^2,\end{aligned}$$

where  $\mathbf{L} = \begin{pmatrix} a_1 & b_1 \\ a_2 & b_2 \end{pmatrix} = \begin{pmatrix} \mathbf{f}' \\ \mathbf{g}' \end{pmatrix}$ . Thus (2) is proved.

Let's go now back to the original problem, i.e., finding the area of the triangle  $TUV$ , say, in  $\mathbb{R}^2$ , formed by the points  $\mathbf{t} = \begin{pmatrix} x_1 \\ y_1 \end{pmatrix}$ ,  $\mathbf{u} = \begin{pmatrix} x_2 \\ y_2 \end{pmatrix}$ , and  $\mathbf{v} = \begin{pmatrix} x_3 \\ y_3 \end{pmatrix}$ . We can move this triangle to the origin by subtracting  $\mathbf{t}$  from  $\mathbf{u}$  and  $\mathbf{v}$ :  $\mathbf{u}_0 = \mathbf{u} - \mathbf{t}$ ,  $\mathbf{v}_0 = \mathbf{v} - \mathbf{t}$ , and thus we obtain the triangle  $OU_0V_0$ , whose area is obviously the absolute value of

$$\frac{1}{2} \begin{vmatrix} \mathbf{u}' - \mathbf{t}' \\ \mathbf{v}' - \mathbf{t}' \end{vmatrix} = \frac{1}{2} \begin{vmatrix} x_2 - x_1 & y_2 - y_1 \\ x_3 - x_1 & y_3 - y_1 \end{vmatrix}.$$

Comment: Very likely there exist proofs simpler than that above.

Consider then the last claim: Show that  $\det(\mathbf{U}) = 0 \iff$  the three data points lie in the same line, and hence, assuming that  $x$  and  $y$  have nonzero variances, we have  $r_{xy}^2 = 1 \iff \det(\mathbf{U}) = 0$ .

Obviously, in view of the nonzero variances,  $\det(\mathbf{U}) = 0 \iff \text{rk}(\mathbf{U}) = 2$ . Now  $\text{rk}(\mathbf{U}) = 2$  implies that  $\mathbf{y} = a\mathbf{1} + b\mathbf{x}$  for some  $a$  and  $b$  ( $b \neq 0$ ), i.e.,  $y_i = a + bx_i$ ,  $i = 1, 2, 3$ , and hence  $r_{xy}^2 = 1$ .  $\square$

**3.4.** Consider the variables  $x_1$  and  $x_2$  and their sum  $y = x_1 + x_2$ . Let the corresponding variable vectors be  $\mathbf{y}$ ,  $\mathbf{x}_1$ , and  $\mathbf{x}_2$ . Assume that  $\mathbf{x}_1$  and  $\mathbf{x}_2$  are centered and of equal length so that  $\mathbf{x}'_1\mathbf{x}_1 = \mathbf{x}'_2\mathbf{x}_2 = d^2$ . Show that

$$\text{var}_s(y) = \frac{1}{n-1} 2d^2(1 + r_{12}), \quad \text{cor}_s^2(x_1, y) = \frac{1}{2}(1 + r_{12}).$$

Moreover, let  $\mathbf{u} = \mathbf{x}_1 - \mathbf{P}_y\mathbf{x}_1$  and  $\mathbf{v} = \mathbf{x}_2 - \mathbf{P}_y\mathbf{x}_2$ . Show that  $\mathbf{u} = -\mathbf{v}$  and thereby  $r_{12 \cdot y} = \text{cor}_d(\mathbf{u}, \mathbf{v}) = -1$ .

**• SOLUTION TO EX. 3.4:**

In view of  $\mathbf{x}'_1\mathbf{x}_2 = d^2r_{12}$ , we have

$$\begin{aligned}
SS_y &= \mathbf{y}'\mathbf{y} = \mathbf{x}'_1\mathbf{x}_1 + \mathbf{x}'_2\mathbf{x}_2 + 2\mathbf{x}'_1\mathbf{x}_2 \\
&= 2d^2 + 2d^2r_{12} = 2d^2(1 + r_{12}), \\
SP_{x_1y} &= \mathbf{x}'_1\mathbf{y} = \mathbf{x}'_1(\mathbf{x}_1 + \mathbf{x}_2) = d^2 + d^2r_{12} = d^2(1 + r_{12}), \\
\text{cor}_s^2(x_1, y) &= \frac{[d^2(1 + r_{12})]^2}{d^2 \cdot 2d^2(1 + r_{12})} = \frac{1}{2}(1 + r_{12}).
\end{aligned}$$

We observe that  $\mathbf{u} = -\mathbf{v}$  because

$$\begin{aligned}
\mathbf{u} &= (\mathbf{I} - \mathbf{P}_y)\mathbf{x}_1 = (\mathbf{I} - \mathbf{P}_y)(\mathbf{y} - \mathbf{x}_2) \\
&= (\mathbf{I} - \mathbf{P}_y)\mathbf{y} - (\mathbf{I} - \mathbf{P}_y)\mathbf{x}_2 \\
&= -(\mathbf{I} - \mathbf{P}_y)\mathbf{x}_2 = -\mathbf{v}.
\end{aligned}$$

**3.5.** Let the variable vectors  $\mathbf{x}$  and  $\mathbf{y}$  be centered and of equal length.

- What is  $\text{cor}_s(x - y, x + y)$ ?
- Draw a figure from which you can conclude the result above.
- What about if  $\mathbf{x}'\mathbf{x} = \mathbf{y}'\mathbf{y} + 0.001$ ?

**• SOLUTION TO EX. 3.5:**

Denoting  $u = x + y, v = x - y, \mathbf{x}'\mathbf{x} = a^2, \mathbf{y}'\mathbf{y} = b^2$ , we have

$$\begin{aligned}
SP_{uv} &= (\mathbf{x} + \mathbf{y})'(\mathbf{x} - \mathbf{y}) = \mathbf{x}'\mathbf{x} - \mathbf{y}'\mathbf{y} = a^2 - b^2, \\
SS_v &= (\mathbf{x} - \mathbf{y})'(\mathbf{x} - \mathbf{y}) = \mathbf{x}'\mathbf{x} + \mathbf{y}'\mathbf{y} - 2\mathbf{x}'\mathbf{y} = a^2 + b^2 - 2\mathbf{x}'\mathbf{y}, \\
SS_u &= (\mathbf{x} + \mathbf{y})'(\mathbf{x} + \mathbf{y}) = \mathbf{x}'\mathbf{x} + \mathbf{y}'\mathbf{y} + 2\mathbf{x}'\mathbf{y} = a^2 + b^2 + 2\mathbf{x}'\mathbf{y}, \\
r_{uv} &= \frac{a^2 - b^2}{\sqrt{(a^2 + b^2 - 2\mathbf{x}'\mathbf{y})(a^2 + b^2 + 2\mathbf{x}'\mathbf{y})}}.
\end{aligned}$$

**3.6.** Consider the variable vectors  $\mathbf{x}$  and  $\mathbf{y}$  which are centered and of unit length. Let  $\mathbf{u}$  be the residual when  $\mathbf{y}$  is explained by  $\mathbf{x}$  and  $\mathbf{v}$  be the residual when  $\mathbf{x}$  is explained by  $\mathbf{y}$ . What is the  $\text{cor}_d(\mathbf{u}, \mathbf{v})$ ? Draw a figure about the situation.

**• SOLUTION TO EX. 3.6:**

$$\begin{aligned}
\mathbf{u} &= (\mathbf{I} - \mathbf{P}_y)\mathbf{x} = (\mathbf{I} - \mathbf{y}\mathbf{y}')\mathbf{x} = \mathbf{x} - r_{xy}\mathbf{y}, \\
\mathbf{v} &= (\mathbf{I} - \mathbf{P}_x)\mathbf{y} = (\mathbf{I} - \mathbf{x}\mathbf{x}')\mathbf{y} = \mathbf{y} - r_{xy}\mathbf{x}, \\
SS_u &= \mathbf{x}'(\mathbf{I} - \mathbf{y}\mathbf{y}')\mathbf{x} = 1 - r_{xy}^2, \\
SS_v &= \mathbf{y}'(\mathbf{I} - \mathbf{x}\mathbf{x}')\mathbf{y} = 1 - r_{xy}^2, \\
SP_{uv} &= \mathbf{x}'(\mathbf{I} - \mathbf{y}\mathbf{y}')(\mathbf{I} - \mathbf{x}\mathbf{x}')\mathbf{y} \\
&= (\mathbf{x} - r_{xy}\mathbf{y})'(\mathbf{y} - r_{xy}\mathbf{x}) \\
&= r_{xy} - r_{xy} - r_{xy} + r_{xy}^3 = -r_{xy}(1 - r_{xy}^2).
\end{aligned}$$

Notice that  $\mathbf{u}$  and  $\mathbf{v}$  are centered since  $\mathbf{x}$  and  $\mathbf{y}$  are centered.  $\square$

**3.7.** Consider the variable vectors  $\mathbf{x}$ ,  $\mathbf{y}$  and  $\mathbf{z}$  which are centered and of unit length and whose correlation matrix is

$$\mathbf{R} = \begin{pmatrix} 1 & 0 & -1/\sqrt{2} \\ 0 & 1 & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} & 1 \end{pmatrix}.$$

Show that  $\mathbf{x} - \mathbf{y} + \sqrt{2}\mathbf{z} = \mathbf{0}$ .

**• SOLUTION TO EX. 3.7:**

Denoting  $\mathbf{U} = (\mathbf{x} : \mathbf{y} : \mathbf{z})$  and  $\mathbf{a} = (1, -1, \sqrt{2})'$  we have  $\mathbf{R} = \mathbf{U}'\mathbf{U}$  and

$$\mathbf{U}'\mathbf{U}\mathbf{a} = \mathbf{R}\mathbf{a} = \begin{pmatrix} 1 & 0 & -1/\sqrt{2} \\ 0 & 1 & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ \sqrt{2} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

Now  $\mathbf{U}'\mathbf{U}\mathbf{a} = \mathbf{0}$  implies  $\mathbf{U}\mathbf{a} = \mathbf{0}$ .  $\square$

**3.8.** Consider the 17-dimensional variable vectors  $\mathbf{x} = (1, \dots, 1, 1, 4)'$  and  $\mathbf{y} = (-1, \dots, -1, -1 + 0.001, 4)'$ . Find  $\text{cor}_d(\mathbf{x}, \mathbf{y})$  and  $\cos(\mathbf{x}, \mathbf{y})$ .

**• SOLUTION TO EX. 3.8:**

$\mathbf{x}'\mathbf{x} = 16 + 16 = 32$ ,  $\mathbf{y}'\mathbf{y} = 31 + 0.999^2$ ,  $\mathbf{x}'\mathbf{y} = -16 + 0.001 + 16 = 0.001$ , and hence

$$\cos(\mathbf{x}, \mathbf{y}) \approx \frac{0.001}{32} = \frac{1}{3200}.$$

Correlation  $r_{xy} \approx 1$ , because 16 observations lie on the line  $y = \frac{5}{8}x - \frac{8}{3}$ ; the observation  $(1, -0.999)$  is a bit out of this line.  $\square$

**3.9.** Suppose that  $\mathbf{U} = (\mathbf{x} : \mathbf{y} : \mathbf{z}) \in \mathbb{R}^{100 \times 3}$  is a data matrix, where the  $x$ -values represent the results of throwing a dice and  $y$ -values are observations from the normal distribution  $N(0, 1)$ , and  $\mathbf{z} = \mathbf{x} + \mathbf{y}$ . What is your guess for the length of the variable vector  $\mathbf{z}$ ?

• **SOLUTION TO EX. 3.9:**

Theoretically,  $\mu_x = 0$ ,  $\mu_y = 3.5$ ,  $\sigma_x^2 = 1$ ,  $\sigma_y^2 = \frac{35}{12}$ , and hence

$$\begin{aligned} 1 &\approx s_x^2 \approx \frac{1}{99} \mathbf{x}'\mathbf{x} \implies \mathbf{x}'\mathbf{x} \approx 99, \\ \frac{35}{12} &\approx s_y^2 \approx \frac{1}{99} (\mathbf{y}'\mathbf{y} - 100 \cdot 3.5^2) \implies \mathbf{y}'\mathbf{y} \approx 99 \cdot \frac{35}{12} + 100 \cdot \frac{49}{4}, \\ \mathbf{z}'\mathbf{z} &\approx \mathbf{x}'\mathbf{x} + \mathbf{y}'\mathbf{y}. \end{aligned}$$

**3.10.** Consider a  $3 \times 3$  centering matrix

$$\mathbf{C} = \begin{pmatrix} 2/3 & -1/3 & -1/3 \\ -1/3 & 2/3 & -1/3 \\ -1/3 & -1/3 & 2/3 \end{pmatrix}.$$

Find the following:

- a basis for  $\mathcal{C}(\mathbf{C})$ ,
- orthonormal bases for  $\mathcal{C}(\mathbf{C})$  and  $\mathcal{C}(\mathbf{C})^\perp$ ,
- eigenvalues and orthonormal eigenvectors of  $\mathbf{C}$ ,
- $\mathbf{C}^+$ , and  $\mathbf{C}^-$  which has (i) rank 2, (ii) rank 3.

• **SOLUTION TO EX. 3.10:**

- (a)  $\mathcal{C}(\mathbf{C}) = \{\mathbf{y} \in \mathbb{R}^3 : \exists \mathbf{x} \text{ such that } \mathbf{y} = \mathbf{C}\mathbf{x}\}$ , i.e.,  $\mathcal{C}(\mathbf{C})$  is the set of all vectors which are centered or in other words, those vectors which satisfy  $\mathbf{x}'\mathbf{1}_3 = 0$ , so that  $\mathcal{C}(\mathbf{C}) = \mathcal{C}(\mathbf{1}_3)^\perp$ . The fact  $\text{rk}(\mathbf{C}) = 2$  can be concluded in many ways. For example, we always have

$$\text{rk}(\mathbf{I}_n - \mathbf{P}_\mathbf{A}) = n - \text{rk}(\mathbf{A}),$$

and so  $\text{rk}(\mathbf{C}) = \text{rk}(\mathbf{I}_3 - \mathbf{P}_\mathbf{1}) = 3 - \text{rk}(\mathbf{1}) = 2$ . Any two columns of  $\mathbf{C}$  create a basis for  $\mathcal{C}(\mathbf{C})$  as well as the columns of

$$\mathbf{A}_1 = \begin{pmatrix} 2 & -1 \\ -1 & 2 \\ -1 & -1 \end{pmatrix}, \quad \mathbf{A}_2 = \begin{pmatrix} 1 & 1 \\ -1 & 0 \\ 0 & -1 \end{pmatrix}.$$

- (b)  $\mathbf{B} = \begin{pmatrix} -2/\sqrt{6} & 0 \\ 1/\sqrt{6} & 1/\sqrt{2} \\ 1/\sqrt{6} & -1/\sqrt{2} \end{pmatrix} = (\mathbf{b}_1 : \mathbf{b}_2)$  is an orthonormal basis for  $\mathcal{C}(\mathbf{C})$ .  
 $\mathcal{C}(\mathbf{C})^\perp = \mathcal{C}(\mathbf{1}_3) \implies \frac{1}{\sqrt{3}}\mathbf{1}_3$  is an orthonormal basis for  $\mathcal{C}(\mathbf{C})^\perp$ .
- (c) The eigenvalues of an idempotent matrix  $\mathbf{P}$ , say, are zeros and ones (why?) and the number of ones is  $\text{rk}(\mathbf{P})$ . Hence  $\text{ch}(\mathbf{C}) = \{1, 1, 0\}$ .
- $\mathbf{C}\mathbf{1} = 0 \cdot \mathbf{1}$ , and so  $\mathbf{1}$  is an eigenvector of  $\mathbf{C}$  w.r.t. eigenvalue 0, i.e.,  $(0, \mathbf{1})$  is an eigenpair for  $\mathbf{C}$ .
  - $\mathbf{C}\mathbf{t} = 1 \cdot \mathbf{t}$  iff  $\mathbf{t}$  is a centered vector. Hence any centered vector  $\mathbf{t}$  is an eigenvector of  $\mathbf{C}$  w.r.t. eigenvalue 1. For example, the columns of  $\mathbf{B}$  in (b) are orthonormal eigenvectors of  $\mathbf{C}$  w.r.t. eigenvalue 1. We then have the equation

$$\mathbf{C}\mathbf{B} = \mathbf{B}. \quad (1)$$

Postmultiplying (1) by an orthogonal  $\mathbf{Q}_{2 \times 2}$  yields

$$\mathbf{C}\mathbf{B}\mathbf{Q} = \mathbf{B}\mathbf{Q},$$

which shows that the columns of  $\mathbf{B}\mathbf{Q}$  are also orthonormal eigenvectors of  $\mathbf{C}$  w.r.t. eigenvalue 1. Recall that according to Theorem 18 (p. 357) the following holds for multiple eigenvalues:

- Consider the distinct eigenvalues of  $\mathbf{A}$ ,  $\lambda_{\{1\}} > \dots > \lambda_{\{s\}}$ , and let  $\mathbf{T}_{\{i\}}$  be an  $n \times m_i$  matrix consisting of the orthonormal eigenvectors corresponding to  $\lambda_{\{i\}}$ ;  $m_i$  is the multiplicity of  $\lambda_{\{i\}}$ . With this ordering,  $\mathbf{A}$  is unique and  $\mathbf{T}$  is unique up to postmultiplying by a blockdiagonal matrix  $\mathbf{U} = \text{blockdiag}(\mathbf{U}_1, \dots, \mathbf{U}_s)$ , where  $\mathbf{U}_i$  is an orthogonal  $m_i \times m_i$  matrix.
- (d)  $\mathbf{C} = \mathbf{C}^+$ , which is easy to establish;  $\text{rk}(\mathbf{C}) = 2$ , and so  $\mathbf{C}$  is a generalized inverse  $\mathbf{C}^-$  which has rank 2.

Denote

$$\mathbf{T} = (\mathbf{t}_1 : \mathbf{t}_2 : \mathbf{t}_3) = (\mathbf{T}_1 : \frac{1}{\sqrt{3}}\mathbf{1}_3),$$

where  $\mathbf{T}_1 = \mathbf{B}$ , as in (b). Then the vectors  $\mathbf{t}_i$  are the orthonormal eigenvectors of  $\mathbf{C}$  and  $\mathbf{C}$  has the eigenvalue decomposition

$$\mathbf{C} = \mathbf{T}\mathbf{\Lambda}\mathbf{T}' = \mathbf{T} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \mathbf{T}' = \mathbf{T}_1\mathbf{T}'_1 = \mathbf{t}_1\mathbf{t}'_1 + \mathbf{t}_2\mathbf{t}'_2.$$

In light of Section 19.5 (pp. 407–408),

$$\mathbf{T} \begin{pmatrix} 1 & 0 & a \\ 0 & 1 & b \\ c & d & e \end{pmatrix} \mathbf{T}' \in \{\mathbf{C}^-\} \quad \text{for all } a, b, c, d, e \in \mathbb{R}. \quad (2)$$

In particular,

$$\mathbf{G} = \mathbf{T} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3\alpha \end{pmatrix} \mathbf{T}' = \mathbf{t}_1 \mathbf{t}'_1 + \mathbf{t}_2 \mathbf{t}'_2 + \alpha \mathbf{1}_3 \mathbf{1}'_3 \in \{\mathbf{C}^-\},$$

i.e.,

$$\mathbf{G} = \mathbf{t}_1 \mathbf{t}'_1 + \mathbf{t}_2 \mathbf{t}'_2 + \alpha \mathbf{1}_3 \mathbf{1}'_3 = \mathbf{C} + \alpha \mathbf{1}_3 \mathbf{1}'_3 \in \{\mathbf{C}^-\} \quad \text{for all } \alpha \in \mathbb{R}.$$

If  $\alpha \neq 0$  then  $\text{rk}(\mathbf{G}) = 3$ , and  $\mathbf{G}$  is a positive definite generalized inverse of  $\mathbf{C}$ . Of course there are many other positive definite generalized inverses for  $\mathbf{C}$ .

Let us next consider the conditions under which  $\mathbf{G}$  in (2) is a symmetric nonnegative definite generalized inverse of  $\mathbf{C}$ ; denote this set as

$$\{\mathbf{C}_{\text{nnd}}^-\}.$$

We observe that

$$\mathbf{G} = \mathbf{T} \begin{pmatrix} 1 & 0 & f_1 \\ 0 & 1 & f_2 \\ f_1 & f_2 & \delta \end{pmatrix} \mathbf{T}' = \mathbf{T} \begin{pmatrix} \mathbf{I}_2 & \mathbf{f} \\ \mathbf{f}' & \delta \end{pmatrix} \mathbf{T}' \in \text{NND}_3$$

if and only if (why?)

$$\mathbf{K} := \begin{pmatrix} \mathbf{I}_2 & \mathbf{f} \\ \mathbf{f}' & \delta \end{pmatrix} \in \text{NND}_3,$$

which in view of (14.7) and (14.8) (p. 306) holds if and only if

$$\mathbf{f}'\mathbf{f} \leq \delta. \tag{3}$$

For notational convenience, we can replace  $\mathbf{K}$  with

$$\mathbf{L} = \begin{pmatrix} \mathbf{I}_2 & \sqrt{3}\mathbf{f} \\ \sqrt{3}\mathbf{f}' & 3\delta \end{pmatrix},$$

which is nonnegative definite iff (3) holds.

Two side-questions:

(a) Is the symmetry of

$$\begin{pmatrix} 1 & 0 & a \\ 0 & 1 & b \\ c & d & e \end{pmatrix} := \mathbf{N}$$

necessary for the symmetry of  $\mathbf{TNT}'$ ? (Yes.)

(b) Is the nonnegative definiteness of  $\mathbf{K}$  necessary for the nonnegative definiteness of  $\mathbf{TKT}'$ ? (Yes.)



Let's calculate the matrix  $\mathbf{G} = \mathbf{T}\mathbf{L}\mathbf{T}'$ :

$$\begin{aligned}\mathbf{G} &= \mathbf{T}\mathbf{L}\mathbf{T}' = \mathbf{T} \begin{pmatrix} \mathbf{I}_2 & \sqrt{3}\mathbf{f} \\ \sqrt{3}\mathbf{f}' & 3\delta \end{pmatrix} \mathbf{T}' \\ &= \mathbf{T}_1\mathbf{T}'_1 + \mathbf{t}_3\sqrt{3}\mathbf{f}'\mathbf{T}'_1 + \mathbf{T}_1\sqrt{3}\mathbf{f}\mathbf{t}'_3 + \delta 3\mathbf{t}_3\mathbf{t}'_3 \\ &= \mathbf{C} + \mathbf{1}\mathbf{f}'\mathbf{T}'_1 + \mathbf{T}_1\mathbf{f}\mathbf{1}' + \delta \mathbf{1}\mathbf{1}'.\end{aligned}$$

Now any matrix of the form

$$\mathbf{G} = \mathbf{C} + \mathbf{1}\mathbf{f}'\mathbf{T}'_1 + \mathbf{T}_1\mathbf{f}\mathbf{1}' + \delta \mathbf{1}\mathbf{1}' \quad (4)$$

is a nonnegative definite generalized inverse of  $\mathbf{C}$  for any  $\mathbf{f} \in \mathbb{R}^2$  and  $\delta \in \mathbb{R}$  which satisfy

$$\mathbf{f}'\mathbf{f} \leq \delta. \quad (5)$$

We stop here for a while and advise the reader to have a look at the Exercise 15.10 (p. 342) and the references therein.

After a short break, let's go back to business. According to Exercise 15.10 it *seems* that the set of nnd matrices given in (4) equals the set of nnd matrices of the form

$$\mathbf{V} = \mathbf{I}_3 + \mathbf{a}\mathbf{1}' + \mathbf{1}\mathbf{a}' := \mathbf{I}_3 + \mathbf{W}, \quad (6)$$

where  $\mathbf{a} \in \mathbb{R}^3$  is an arbitrary vector subject to the condition that  $\mathbf{W}$  is nonnegative definite. Actually in Exercise 15.10 we are dealing with positive definiteness and hence seems is *seems* above. When is  $\mathbf{V}$  in (6) nnd? We observe that

$$\text{ch}(\mathbf{V}) = \{1 + \text{ch}(\mathbf{a}\mathbf{1}' + \mathbf{1}\mathbf{a}')\}. \quad (7)$$

Now consider the equation

$$\mathbf{W}\mathbf{q} = (\mathbf{a}\mathbf{1}' + \mathbf{1}\mathbf{a}')\mathbf{q} = 0\mathbf{q} = \mathbf{0}$$

i.e.,

$$\mathbf{a}(\mathbf{1}'\mathbf{q}) + \mathbf{1}(\mathbf{a}'\mathbf{q}) = \mathbf{0}.$$

If  $\text{rk}(\mathbf{1} : \mathbf{a}) = 2$ , then  $\mathbf{q} \in \mathcal{C}(\mathbf{1} : \mathbf{a})^\perp$  and thereby 0 is an eigenvalue of  $\mathbf{W}$  of multiplicity 1 (=  $n - 2$ ) with  $\mathbf{q}$  being the corresponding eigenvector. The remaining two eigenvalues of  $\mathbf{W} = \mathbf{a}\mathbf{1}' + \mathbf{1}\mathbf{a}'$  are  $\mathbf{1}'\mathbf{a} \pm \sqrt{n\mathbf{a}'\mathbf{a}}$ . How to prove this? We leave this open and refer to the references; in particular, Farebrother (1987, Cor. 1). In any event,  $\mathbf{V}$  is nnd iff

$$1 + \mathbf{1}'\mathbf{a} - \sqrt{n\mathbf{a}'\mathbf{a}} \geq 0, \quad \text{i.e.,} \quad 1 \geq \sqrt{n\mathbf{a}'\mathbf{a}} - \mathbf{1}'\mathbf{a}. \quad (8)$$

If  $\text{rk}(\mathbf{1} : \mathbf{a}) = 1$ , i.e.,  $\mathbf{a} = c\mathbf{1}$  for some nonzero  $c \in \mathbb{R}$ , then  $\mathbf{V} = \mathbf{I}_3 + 2c\mathbf{1}\mathbf{1}'$  whose eigenvalues are  $\{1 + \text{ch}(2c\mathbf{1}\mathbf{1}')\} = \{1, 1, 6c\}$ .

It's time to stop here. There are some related considerations in the solution of Exercise 15.10. We may mention that Chaganty & Vaish (1997, Cor. 2.1) characterized the class of all nnd generalized inverses of the centering matrix  $\mathbf{C} \in \mathbb{R}^{n \times n}$  as

$$\mathbf{V} = \mathbf{C} + \mathbf{1}\mathbf{b}' + \mathbf{b}\mathbf{1}' - \bar{b}\mathbf{1}\mathbf{1}', \quad (9)$$

where  $\mathbf{b} \in \mathbb{R}^n$  is such that

$$\mathbf{b}'\mathbf{C}\mathbf{b} \leq \bar{b}, \quad \text{and } \bar{b} = \mathbf{1}'\mathbf{b}/n.$$

Some references; those not appearing in the Tricks References, written in full:

- Chaganty, N. Rao & Vaish, A. K. (1997). An invariance property of common statistical tests. *Linear Algebra and its Applications*, 264, 421–437.
- Farebrother, R. W. (1987). Three theorems with applications to Euclidean distance matrices. *Linear Algebra and its Applications*, 95, 11–16.
- Jensen, D. R. (1996). Structured dispersion and validity in linear inference. *Linear Algebra and its Applications*, 249, 189–196.
- Jensen, D. R. & Srinivasan, S. S. (2004). Matrix equivalence classes with applications. *Linear Algebra and its Applications*, 388, 249–260.
- Mathew (1985).
- Sharpe, G. E. & Styan, G. P. H. (1965). Circuit duality and the general network inverse. *IEEE Trans. Circuit Theory CT-12*, 22–27.
- Styan & Subak-Sharpe (1997). □

**3.11.** Suppose that the variable vectors  $\mathbf{x}$ ,  $\mathbf{y}$  and  $\mathbf{z}$  are centered and of unit length. Show that corresponding to (3.8) (p. 93),

$$r_{xy \cdot z} = 0 \iff \mathbf{y} \in \mathcal{C}(\mathbf{Q}_z \mathbf{x})^\perp = \mathcal{C}(\mathbf{x} : \mathbf{z})^\perp \boxplus \mathcal{C}(\mathbf{z}),$$

where  $\mathbf{Q}_z = \mathbf{I}_n - \mathbf{P}_z$ . See also Exercise 8.7 (p. 185).