

**Solutions to Exercises 5.10.1 through 5.10.6.**

5.10.1. Let  $\mathbf{x} \in E_r$ ,  $\theta \in E_r$ ,  $c_0 \geq 0$ , and  $\mathfrak{F} > 0$ .

(a) *If.* Putting  $\mathbf{a} = \mathfrak{F}^{-1}(\mathbf{x} - \theta)$  into the right side gives  $((\mathbf{x} - \theta)^T \mathfrak{F}^{-1}(\mathbf{x} - \theta))^2 \leq c_0(\mathbf{x} - \theta)^T \mathfrak{F}^{-1}(\mathbf{x} - \theta)$ . If  $(\mathbf{x} - \theta)^T \mathfrak{F}^{-1}(\mathbf{x} - \theta) \neq 0$ , then this may be cancelled from both sides, resulting in the desired inequality. If  $(\mathbf{x} - \theta)^T \mathfrak{F}^{-1}(\mathbf{x} - \theta) = 0$ , the desired inequality is obviously true.

*Only if.* By Schwartz inequality,  $(\mathbf{a}^T(\mathbf{x} - \theta))^2 = ((\mathbf{a}^T \mathfrak{F}^{1/2})(\mathfrak{F}^{-1/2}(\mathbf{x} - \theta)))^2 \leq (\mathbf{a}^T \mathfrak{F} \mathbf{a})((\mathbf{x} - \theta)^T \mathfrak{F}^{-1}(\mathbf{x} - \theta))$ . Hence,  $(\mathbf{a}^T(\mathbf{x} - \theta))^2 \leq c_0(\mathbf{a}^T \mathfrak{F} \mathbf{a})$ .

(b) *If.* Since  $1^T(\mathbf{x} - \theta) = 0$ , we may put  $\mathbf{a} = (\mathbf{x} - \theta)$  into the right side and cancel.

*Only if.* By Schwartz,  $(\mathbf{a}^T(\mathbf{x} - \theta))^2 \leq (\mathbf{a}^T \mathbf{a})((\mathbf{x} - \theta)^T(\mathbf{x} - \theta))$  etc.

(c) *If.* Since  $1^T \mathfrak{F}^{-1}(\mathbf{x} - \theta) = 0$ , we may put  $\mathbf{a} = \mathfrak{F}^{-1}(\mathbf{x} - \theta)$  into the right side and cancel.

*Only if.* By Schwartz,  $(\mathbf{a}^T(\mathbf{x} - \theta))^2 \leq (\mathbf{a}^T \mathfrak{F} \mathbf{a})((\mathbf{x} - \theta)^T \mathfrak{F}^{-1}(\mathbf{x} - \theta))$  etc.

5.10.2. The confidence ellipsoid (5.133) is

$$\{\mu : \sum_{i=1}^I (\bar{X}_i - \bar{X}_{..} - \mu_i)^2 \leq (I-1)\hat{\sigma}^2 F_{(I-1), IJ-I-J+1; \alpha} / J\}$$

Since  $\sum_i (\bar{X}_i - \bar{X}_{..} - \mu_i) = 0$ , Exercise 1(b) states this is equivalent to the multiple comparison

$$\left(\sum_{i=1}^I a_i (\bar{X}_i - \bar{X}_{..} - \mu_i)\right)^2 \leq \left(\sum_{i=1}^I a_i^2\right) (I-1)\hat{\sigma}^2 F_{(I-1), IJ-I-J+1; \alpha} / J$$

for all vectors  $\mathbf{a}$  for which  $\sum_i a_i = 0$ . But since  $\sum_i a_i \bar{X}_{..} = 0$ , this reduces to (5.143).

5.10.3. (a) The UMP invariant confidence set of confidence coefficient  $1 - \alpha$ , given by (5.132), reduces to

$$\{\theta : \sum_i \sum_j (\bar{X}_i - \theta_i)^2 \leq I\hat{\sigma}^2 F_{I, N-I; \alpha}\}$$

where  $N = \sum_i n_i$ , and  $\hat{\sigma}^2 = (1/(N-I)) \sum \sum (X_{ij} - \bar{X}_i)^2$ . The summation over  $j$  in the confidence ellipsoid may be replaced by  $n_i$ , and the expression has the form of Exercise 1(a) with  $\mathfrak{F}^{-1}$  being the diagonal matrix with  $n_1, \dots, n_I$  down the diagonal. This exercise states that the confidence ellipsoid is equivalent to the multiple comparison

$$\left(\sum_i a_i (\bar{X}_i - \theta_i)\right)^2 \leq \left(\sum_i a_i^2 / n_i\right) I\hat{\sigma}^2 F_{I, N-I; \alpha}$$

for all vectors  $\mathbf{a}$ .

(b) The confidence ellipsoids reduce to

$$\{\mu : \sum_i n_i (\bar{X}_i - \bar{X}_{..} - \mu_i)^2 \leq (I-1)\hat{\sigma}^2 F_{I-1, N-I; \alpha}\}.$$

Here  $\sum_i (\bar{X}_i - \bar{X}_{..} - \mu_i) = 0$ , so that we may apply Exercise 1(c) to obtain the multiple comparison statement

$$\left(\sum_i a_i \bar{X}_i - \sum_i a_i \mu_i\right)^2 \leq \left(\sum_i a_i^2 / n_i\right) (I-1)\hat{\sigma}^2 F_{I-1, N-I; \alpha}.$$

for all  $\mathbf{a}$  such that  $\sum_i a_i = 0$ .

5.10.4. (a) The multiple comparison given by (5.140) is

$$\left(\sum_i a_i \bar{X}_i - \sum_i a_i \theta_i\right)^2 \leq \left(\sum_i a_i^2\right) (I-1)\hat{\sigma}^2 F_{I-1, IJ-I; \alpha} / J = \left(\sum_i a_i^2\right) 5F_{3, 20; \alpha}$$

for all  $\mathbf{a}$  such that  $\sum_i a_i = 0$ . When  $\alpha = .10$  and  $\sum_i a_i^2 = 2$ , the square root of the right side reduces to  $\sqrt{23.8} = 4.88 \dots$ . Only  $\bar{X}_1 - \bar{X}_3$  and  $\bar{X}_1 - \bar{X}_4$  are further apart than this. So the ordering is

$$\underline{\theta_1} < \underline{\theta_2} < \theta_3 < \theta_4$$

(b) When  $\alpha = .01$ , the square root of the right side is  $\sqrt{49.4} = 7.03$ , so this time all confidence intervals for the differences,  $\theta_i - \theta_j$ , contain the origin.

(c) To compute the  $F$ -test, we first compute  $\bar{X}_{..} = (0 + 4 + 5 + 7)/4 = 4$  and then  $\sum \sum (\bar{X}_i - \bar{X}_{..})^2 = 6[4^2 + 0^2 + 1^2 + 3^2] = 156$ . The  $F$ -statistic is  $F = (156/3)/10 = 5.2$ . This is greater than  $F_{3,20;.01} = 4.94$  so  $H_0$  is rejected.

(d) It may be seen that the worst case occurs when the  $a_i$  are proportional to the  $\bar{X}_i - \bar{X}_{..}$ , namely  $a_1 = -4$ ,  $a_2 = 0$ ,  $a_3 = 1$  and  $a_4 = 3$ . For this linear combination, we get the confidence region  $(3\theta_4 + \theta_3 - 4\theta_1 - 26)^2 \leq (\sum a_i^2)5F_{I-1, IJ-I; \alpha} = (26)24.2 = 629.2$ . This leads to the confidence interval  $.9 \leq 3\theta_4 + \theta_3 - 4\theta_1 \leq 51.1$ .

5.10.5. (a) The sum of squares to be minimized is  $S(\beta) = \sum (X_i - \beta_0 - \beta_1 z_i - \beta_2 z_i^2)^2$ . The formulas are somewhat simpler if we take  $\sum_i z_i^2 = n$  instead of  $\sum_i z_i^2 = 1$ . Let  $m_4 = (1/n) \sum_i z_i^4$  and assume  $m_4 > 1$ . Setting the three partial derivatives equal to zero, we have  $\hat{\beta}_0 + \hat{\beta}_2 = \bar{X}$ ,  $\hat{\beta}_1 = s_{xz}$  and  $\hat{\beta}_0 + m_4 \hat{\beta}_2 = s_{xz^2}$ . Solving simultaneously, we find

$$\hat{\beta}_1 = s_{xz}, \quad \hat{\beta}_2 = (s_{xz^2} - \bar{X})/(m_4 - 1), \quad \text{and} \quad \hat{\beta}_0 = \bar{X} - \hat{\beta}_2.$$

(b) With  $\beta_1$  and  $\beta_2$  fixed, the least squares estimate of  $\beta_0$  is  $\hat{\beta}_0 = \bar{X} - \beta_2$ . Hence the numerator sum of squares for the  $F$  is

$$\begin{aligned} & \sum (\hat{\beta}_0 + \hat{\beta}_1 z_i + \hat{\beta}_2 z_i^2 - \hat{\beta}_0 - \beta_1 z_i - \beta_2 z_i^2)^2 \\ &= \sum ((\hat{\beta}_1 - \beta_1)z_i + (\hat{\beta}_2 - \beta_2)(z_i^2 - 1))^2 \\ &= n(\hat{\beta}_1 - \beta_1)^2 + n(m_4 - 1)(\hat{\beta}_2 - \beta_2)^2 \end{aligned}$$

The confidence ellipsoid for  $(\beta_1, \beta_2)$  then becomes

$$(\hat{\beta}_1 - \beta_1)^2 + (m_4 - 1)(\hat{\beta}_2 - \beta_2)^2 \leq (2/n)\hat{\sigma}^2 F_{2, n-3; \alpha}$$

where  $\hat{\sigma}^2 = S(\hat{\beta})/(n - 3)$ .

(c) The multiple comparison statements become

$$(a_1 \beta_1 + a_2 \beta_2 - (a_1 \hat{\beta}_1 + a_2 \hat{\beta}_2))^2 \leq (a_1^2 + \frac{a_2^2}{m_4 - 1})(2/n)\hat{\sigma}^2 F_{2, n-3; \alpha}$$

for all  $(a_1, a_2)$ .

5.10.6. The  $F$ -test statement is

$$P\{\sum \sum (\bar{X}_i - \bar{X}_{..} - \mu_i)^2 + \sum \sum (\bar{X}_j - \bar{X}_{..} - \eta_j)^2 \leq c\hat{\sigma}^2\} = 1 - \alpha,$$

where  $c = (I+J-2)F_{I+J-2, (I-1)(J-1); \alpha}$  and  $\hat{\sigma}^2 = \frac{1}{(I-1)(J-1)} \sum \sum (X_{ij} - \bar{X}_i - \bar{X}_j + \bar{X}_{..})^2$ . By the argument given for the final example in the text, this probability statement is equivalent to the statement that the probability is  $1 - \alpha$  that

$$(\sum a_i \bar{X}_i + \sum b_j \bar{X}_j - \sum a_i \mu_i - \sum b_j \eta_j)^2 \leq c\hat{\sigma}^2 (\frac{1}{I} \sum a_i^2 + \frac{1}{J} \sum b_j^2)$$

for all  $\mathbf{a} \in E_I$  and all  $\mathbf{b} \in E_J$  such that  $\sum a_i = 0$  and  $\sum b_j = 0$ .