## PACQ Chapter 2 Solutions

## Arthur Chen

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## Exercise 2.1

Show that for  $\lambda'\beta$  estimable,

$$\frac{\lambda'\hat{\beta} - \lambda'\beta}{\sqrt{MSE\lambda'(X'X)^{-}\lambda}} \sim t(dfE)$$
 (1)

Find the form of an  $\alpha$  level test of  $H_0: \lambda'\beta = 0$  and the form for a  $(1-\alpha)100\%$  confidence interval for  $\lambda'\beta$ .

I will make the usual assumptions for the linear model. Rewrite the expression as the following:

$$\frac{\frac{\lambda'\hat{\beta} - \lambda'\beta}{\sigma\sqrt{\lambda'(X'X)^{-}\lambda)}}}{\sqrt{\frac{MSE}{\sigma^2}}} \tag{2}$$

In the numerator,

$$\lambda'\hat{\beta} \sim N(\lambda\beta, \sigma^2\lambda'(X'X)^-\lambda)$$

implies

$$\frac{\lambda'\hat{\beta} - \lambda'\beta}{\sigma\sqrt{\lambda'(X'X)^{-}\lambda)}} \sim N(0,1)$$

In the denominator,

$$(n-r)\frac{MSE}{\sigma^2} \sim X^2(n-r)$$

Since  $\lambda'\beta$  is estimable,  $\lambda'\hat{\beta}=\rho'MY$ , and since  $MSE=\frac{Y'(I-M)Y}{n-r}$ ,  $\rho'MY$  and Y'(I-M)Y are independent, given the usual linear model assumptions. Thus Equation 2 has a t(dfE) distribution, where dfE=n-r.

For the  $\alpha$  level test of  $H_0: \lambda'\beta = 0$ , the test rejects for large values of the test statistic in Equation 1. That is, we reject when

$$\frac{\lambda'\hat{\beta}}{\sqrt{MSE\lambda'(X'X)^{-}\lambda}} \ge t(1 - \alpha/2, n - r)$$

or

$$\frac{\lambda'\hat{\beta}}{\sqrt{MSE\lambda'(X'X)^{-}\lambda}} \le -t(1-\alpha/2, n-r)$$

For the  $(1-\alpha)100\%$  confidence interval, we do not reject if

$$-t(1-\alpha/2,n-r) < \frac{\lambda'\hat{\beta} - \lambda'\beta}{\sqrt{MSE\lambda'(X'X)^{-}\lambda}} < t(1-\alpha/2,n-r)$$

which is equivalent to

$$\lambda' \hat{\beta} - t(1 - \alpha/2, n - r) \sqrt{MSE\lambda'(X'X)^{-}\lambda} < \lambda' \beta < \lambda' \hat{\beta} + t(1 - \alpha/2, n - r) \sqrt{MSE\lambda'(X'X)^{-}\lambda}$$

## Exercise 2.2

Let  $y_{11}, y_{12} \dots y_{1r}$  be  $N(\mu_1, \sigma^2)$  and  $y_{21}, y_{22} \dots y_{2s}$  be  $N(\mu_2, \sigma^2)$  with all  $y_{ij}$ s independent. Write this as a linear model. Using the results of Chapter 2, find estimates of  $\mu_1, \mu_2, \mu_1 - \mu_2$ , and  $\sigma^2$ . Form a  $\alpha = 0.01$  test for  $H_0: \mu_1 = \mu_2$ . Form 95% confidence intervals for  $\mu_1 - \mu_2$  and  $\mu_1$ . What is the test for  $H_0: \mu_1 = \mu_2 + \Delta$ , where  $\Delta$  is some known fixed quantity? How do these results compare with the usual analysis for two independent samples with a common variance?

I will write this as a linear model in  $\mathbb{R}^{s+r}$ ,  $Y = X\mu + \epsilon$  with data, design, and coefficient matrices

$$Y = \begin{pmatrix} y_{11} \\ \vdots \\ y_{1r} \\ y_{21} \\ \vdots \\ y_{2s} \end{pmatrix}, X = \begin{pmatrix} J_{r,1} & 0 \\ 0 & Js, 1 \end{pmatrix}, \mu = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}$$

We will assume that  $\epsilon$  is distributed  $N(0, \sigma_2 I)$  as usual.

 $\mu_1$  is estimable. To show this, note that  $\lambda_1' = (1,0)$  implies that  $\lambda_1' \mu = \mu_1$ . For  $\rho_1' = (1,0...0)$ ,  $P_1'X\mu = \mu_1$ . Thus  $\hat{\mu}_1 = \rho_1'MY$  is the least squares estimate for  $\mu_1$ . Similarly, the least squares estimate of  $\mu_2$  is estimated with  $\rho_2' = (0...0,1)$ , and  $\mu_1 - \mu_2$  is estimable with  $(\rho_1 - \rho_2)'$ .  $\sigma^2$  is estimated as usual with the MSE, Y'(I-M)Y/(n-r(M)).

Calculating M, we have that

$$X'X = \begin{pmatrix} r & 0 \\ 0 & s \end{pmatrix}$$

implies

$$(X'X)^{-1} = \begin{pmatrix} 1/r & 0 \\ 0 & 1/s \end{pmatrix}$$

$$M = X(X'X)^{-1}X' = \begin{pmatrix} \frac{1}{r}J_r & 0 \\ 0 & \frac{1}{s}J_s \end{pmatrix}$$

$$MY = \begin{pmatrix} (\frac{1}{r}\sum_{i}^{r}y_{i1})J_{r,1} \\ (\frac{1}{s}\sum_{i}^{s}y_{i2})J_{s,1} \end{pmatrix} = \begin{pmatrix} \bar{Y}_{1}J_{r,1} \\ \bar{Y}_{2}J_{s,1} \end{pmatrix}$$

where  $\bar{Y}_1 = \frac{1}{r} \sum_{i=1}^r y_{1i}$ , and  $\bar{Y}_2$  similarly. Thus

$$\hat{\mu}_1 = \rho_1' M Y = \frac{1}{r} \sum_{i}^{r} y_{i1} \tag{3}$$

$$\hat{\mu}_2 = \rho_2' MY = \frac{1}{s} \sum_{i=1}^{s} y_{i2} \tag{4}$$

$$\widehat{\mu_1 - \mu_2} = (\rho_1 - \rho_2)' M Y = \hat{\mu}_1 - \hat{\mu}_2 \tag{5}$$

which match with the standard two-sample results. For  $\hat{\sigma^2} = \frac{Y'(I-M)Y}{n-r(M)} = \frac{Y'(I-M)Y}{n-2}$ , we have

$$Y'Y = \sum_{i=1}^{r} y_{1i}^{2} + \sum_{i=1}^{s} y_{2i}^{2}$$

$$Y'MY = \sum_{i=1}^{r} y_{1i}\bar{Y}_{1} + \sum_{i=1}^{s} y_{2i}\bar{Y}_{2}$$

$$Y'(I - M)Y = \sum_{i=1}^{r} y_{1i}(y_{1i} - \bar{Y}_{1}) + y_{2i}(y_{2i} - \bar{Y}_{2})$$

$$= \sum_{i=1}^{r} (y_{1i} - \bar{Y}_{1})^{2} + \sum_{i=1}^{s} (y_{2i} - \bar{Y}_{2})^{2}$$

and so the estimate of  $\sigma^2$ ,  $\hat{\sigma^2}=MSE=\frac{Y'(I-M)Y}{n-2}$  matches the standard two-sample results.

For a  $\alpha = .01$  level test for  $H_0: \mu_1 = \mu_2$ , we have that  $\lambda' = (1, -1)$  gives us a model of the form  $H_0: \lambda' \mu = 0$ . Thus by Exercise 2.1,

$$\frac{\lambda' \hat{\mu}}{\sqrt{MSE\lambda'(X'X)^{-1}\lambda}} = \frac{\hat{\mu}_2 - \hat{\mu}_2}{\sqrt{MSE(\frac{1}{r} + \frac{1}{s})}} \ge t(.995, n - 2)$$

forms a level  $\alpha = .01$  test for  $H_0$ .