

Applied Math 166 Problem Set 9 4/25/03

11.66 From Exercise 11.27, the complete model is $Y = \beta_0 + \beta_1 x + \epsilon$ with $SSE_c = 6.452$ (8 d.f.). The reduced model is $Y = \beta_0 + \epsilon$ with $\hat{\beta}_0 = \bar{Y}$ and $SSE_R = \sum (Y_i - \hat{Y})^2 = \sum (Y_i - \bar{Y})^2 = 99.149$ (9 d.f.). The test statistic for testing $H_0: \beta_1 = 0$ is

$$F = \frac{S^2_1}{S^2_2} = \frac{\frac{SSE_R - SSE_c}{9-8}}{\frac{SSE_c}{8}} = \frac{92.697}{6.452} = 114.94$$

The rejection region with $\alpha = .05$ is $F_{1,8} = 5.32$, and H_0 is rejected. There is evidence to indicate that $\beta_1 \neq 0$.

11.67 First, note that with $k = 1$, $SSE_R = S_{yy}$. Then, $g = 0$, and $SSE_c = S_{yy} - \hat{\beta}_1 S_{xy}$.

$$F = \frac{(SSE_R - SSE_c)(k-g)}{\frac{SSE_c}{n-(k+1)}} = \frac{S_{yy} - S_{yy} + \hat{\beta}_1 S_{xy}}{\frac{SSE_c}{n-2}} = \frac{\hat{\beta}_1 S_{xy}}{\frac{SSE_c}{n-2}} = \frac{\hat{\beta}_1 S_{xy}}{S^2_{xy}} = \frac{\hat{\beta}_1^2}{\frac{S^2_{xy}}{S^2_{xx}}} = t^2$$

In Exercise 11.66, $F = 114.94$, and in Exercise 11.27, $t = -10.72$. Thus, $t^2 = 1$ which equals F (except for roundoff error).

11.68a. Refer to Exercise 11.57. For the complete model, $Y = \beta_0 + \beta_1 x + \beta_2 x^2 + \epsilon$, $SSE_c = 245.69$ (see Exercise 11.60) with 7 degrees of freedom. For the reduced model, $Y = \beta_0 + \beta_1 x + \epsilon$,

$$SSE_R = \mathbf{Y}'\mathbf{Y} - \beta_1 \mathbf{X}'\mathbf{Y} = 10.208.67 - [29.99 \quad 1.3879] \begin{bmatrix} 299.9 \\ 458.3 \end{bmatrix} = 578.6$$

with 8 d.f. The test statistic for testing $H_0: \beta_2 = 0$ is

$$F = \frac{\frac{SSE_R - SSE_c}{8-7}}{\frac{SSE_c}{7}} = \frac{332.91}{245.69} = 9.49$$

The rejection region with $\alpha = .05$ is $F > F_{1,7} = 5.59$, and H_0 is rejected. There is evidence that $\beta_2 \neq 0$. These results do agree with the results of problem 11.60.

b. For the reduced model, $Y = \beta_0 + \epsilon$, $SSE_R = \sum (y_i - \bar{y})^2 = 1214.669$ with 9 d.f. Then

$$F = \frac{\frac{SSE_R - SSE_c}{9-1}}{\frac{SSE_c}{7}} = \frac{968.98}{245.69} = 13.80$$

The rejection region with $\alpha = .05$ is $F > F_{2,7} = 4.74$, and the null hypothesis, $H_0: \beta_1 = \beta_2 = 0$, is rejected.

11.69 The hypothesis of interest is $H_0: \beta_1 - \beta_4 = 0$. For the complete model, $Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \epsilon$, $SSE_c = 98.8125$ with 11 d.f. For the reduced model, the 2nd and 5th columns of the \mathbf{X} matrix are deleted, so that

$$(\mathbf{X}'\mathbf{X})^{-1} = \begin{bmatrix} \frac{1}{16} & 0 & 0 \\ 0 & \frac{1}{16} & 0 \\ 0 & 0 & \frac{1}{16} \end{bmatrix} \quad \mathbf{X}'\mathbf{Y} = \begin{bmatrix} 338 \\ -19.4 \\ -2.6 \end{bmatrix}$$

Then

$$SSE = \mathbf{Y}'\mathbf{Y} - \hat{\beta}'\mathbf{X}'\mathbf{Y} = 7446.52 - 7164.195 = 282.325$$

with 13 d.f. The test statistic is

$$F = \frac{\frac{SSE_R - SSE_c}{1-1}}{\frac{SSE_c}{11}} = 10.21$$

The rejection region with $\alpha = .05$ is $F > F_{2,11} = 3.98$, and H_0 is rejected. There is reason to believe that either T_1 or T_2 or both affect the yield.

11.70 We need to carry out an F test with test statistic

$$F = \frac{\frac{SSE_R - SSE_c}{24-6}}{\frac{SSE_c}{18}} = \frac{465.134 - 152.177}{18} = 12.34$$

Refer to Table 7. We have $12.34 > 5.92 = F_{0.05}$, thus the p -value $< .005$.

11.80 The F statistic is

$$F = \frac{\frac{SSE_1 - SSE_2}{2} }{\frac{SSE_2}{200-3}} = \frac{\frac{795.23 - 783.9}{2}}{\frac{783.9}{195}} = 1.41$$

The critical value is $F_{.05} = 3.00 > 1.41$, so we do not reject H_0 : salary is not dependent on sex.

11.81 Let

$\mathbf{Y}' = [Y_1 \ Y_2 \ Y_3 \ \dots \ Y_n]$ and $\mathbf{1}' = [1 \ 1 \ 1 \ \dots \ 1]$
be two $1 \times n$ vectors. Then we can write

$$\bar{Y} = \left[\frac{1}{n} \ \frac{1}{n} \ \dots \ \frac{1}{n} \right] \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \\ \vdots \\ Y_n \end{bmatrix} = \frac{1}{n} \mathbf{1}' \mathbf{Y}$$

In matrix form, the equation of interest is $Y = \mathbf{x}'\hat{\beta}$, where

$$\mathbf{x}' = [1 \ x_1 \ x_2 \ \dots \ x_k]$$

and

$$\hat{\beta}' = [\hat{\beta}_0 \ \hat{\beta}_1 \ \dots \ \hat{\beta}_k].$$

Suppose $Y = \bar{Y}$. Then

$$\bar{Y} = \mathbf{x}'\hat{\beta} = \mathbf{x}'(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y}$$

implying

$$\frac{1}{n} \mathbf{1}' \mathbf{Y} \mathbf{Y}' = \mathbf{x}'(\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}' \mathbf{Y} \mathbf{Y}'$$

Which must then imply

$$\frac{1}{n} \mathbf{1}' = \mathbf{x}'(\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}'$$

or

$$\frac{1}{n} \mathbf{1}' \mathbf{X} = \mathbf{x}'(\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}' \mathbf{X}$$

or

$$\frac{1}{n} \mathbf{1}' \mathbf{X} = \mathbf{x}'.$$

That is

$$\mathbf{x}' = [1 \ \bar{x}_1 \ \bar{x}_2 \ \dots \ \bar{x}_k].$$

That is, the point $(\bar{x}_1, \bar{x}_2, \dots, \bar{x}_k, \bar{Y})$ satisfies the equation $Y = \mathbf{x}'\hat{\beta}$, so that the least squares prediction line must pass through this point.

11.82a. Use the coding

$$X_1^* = \frac{X_1 - 65}{15}$$

and

$$X_2^* = \frac{X_2 - 200}{100}$$

Then $X_1 = 50, 80$ correspond to $X_1^* = -1, 1$, while $X_2 = 100, 200, 300$ correspond to $X_2^* = -1, 0, 1$. Now

$$Y = \begin{bmatrix} 21 \\ 23 \\ 26 \\ 22 \\ 23 \\ 28 \end{bmatrix} \quad X^* = \begin{array}{c} \begin{matrix} x_0 & x_1 & x_2 & x_2^2 \end{matrix} \\ \begin{bmatrix} 1 & -1 & -1 & 1 \\ 1 & -1 & 0 & 0 \\ 1 & -1 & 1 & 1 \\ 1 & 1 & -1 & 1 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{array} \quad X^{*'} X^* = \begin{bmatrix} 6 & 0 & 0 & 4 \\ 0 & 6 & 0 & 0 \\ 0 & 0 & 4 & 0 \\ 4 & 0 & 0 & 4 \end{bmatrix}$$

$$(X^{*'} X^*)^{-1} = \begin{bmatrix} \frac{1}{2} & 0 & 0 & -\frac{1}{2} \\ 0 & \frac{1}{6} & 0 & 0 \\ 0 & 0 & \frac{1}{4} & 0 \\ -\frac{1}{2} & 0 & 0 & \frac{3}{4} \end{bmatrix} \quad X^{*'} Y = \begin{bmatrix} 143 \\ 3 \\ 11 \\ 97 \end{bmatrix} \quad \hat{\beta}^* = \begin{bmatrix} 23 \\ .5 \\ 2.75 \\ 1.25 \end{bmatrix}$$

Finally, the least squares equation is

$$\hat{y} = 23 + .5 \left(\frac{x_1 - 65}{15} \right) + 2.75 \left(\frac{x_2 - 200}{100} \right) + 1.25 \left(\frac{x_2 - 200}{100} \right)^2$$

$$= 20.33 + .0333x_1 - .0225x_2 + .000125x_2^2$$

- b. The hypothesis of interest, $H_0: \beta_3 = 0$, is equivalent to a test of $H_0: \beta_3^* = 0$ since $\beta_3 = \left(\frac{1}{100} \right)^2 \beta_3^*$. Using the coded calculations, then, we have

$$SSE = Y'Y - \hat{\beta}^{*'} X^{*'} Y = 3443 - 3442 = 1$$

and

$$s^2 = \frac{SSE}{n-4} = .5.$$

The test statistic is

$$t = \frac{\hat{\beta}_3^*}{\sqrt{c_{33}s^2}} = \frac{1.25}{\sqrt{\left(\frac{1}{2}\right)\left(\frac{1}{2}\right)}} = 2.042$$

The rejection region with $\alpha = .05$ is $|t| > 4.303$, and H_0 is not rejected. The quadratic temperature effect is not significant.

- c. In order to test $H_0: \beta_2 = \beta_3 = 0$, or, equivalently, $H_0: \beta_2^* = \beta_3^* = 0$, complete and reduced models are fitted. For the complete model, $SSE_2 = 1$ with 2 d.f. [from part b]. For the reduced model, columns 3 and 4 of the X^* matrix are omitted and

$$(X^{*'} X^*)^{-1} = \begin{bmatrix} \frac{1}{6} & 0 \\ 0 & \frac{1}{6} \end{bmatrix}$$

$$X^{*'} Y = \begin{bmatrix} 143 \\ 3 \end{bmatrix}$$

so that

$$SSE_1 = 3443 - [23.8333 \quad .5] \begin{bmatrix} 143 \\ 3 \end{bmatrix} = 33.33.$$

with 4 d.f. The test statistic is then

$$F = \frac{\frac{SSE_1 - SSE_2}{2}}{\frac{SSE_2}{2}} = \frac{33.33 - 1}{2} = 32.33$$

The rejection region with $\alpha = .05$ is $F > F_{2,2} = 19.00$, and the null hypothesis is rejected. Temperature does affect yield.