STAT 408/608 Homework 6 Solutions: Written Section

March 18, 2015

1.

$$Var(\hat{\beta}|X) = Var((X'X)^{-1}X'Y|X)$$

$$= (X'X)^{-1}X'Var(Y|X)X(X'X)^{-1}$$

$$= (X'X)^{-1}X'\sigma^{2}IX(X'X)^{-1}$$

$$= \sigma^{2}(X'X)^{-1}X'X(X'X)^{-1}$$

$$= \sigma^{2}(X'X)^{-1}$$

- 2. (a) Yes. Because number of pens in control group is different from the treatment groups. Therefore, I suggest WLS estimator in this experiment, where number of pens is used as a weight.
 - (b) Yes. Because the plot shows straight line will not be a good fit. I suggest a second order polynomial model.
 - (c) I will use one second order polynomial model for three group with the same intercept and without interaction. Because the plot does not show much cross relationship between dosage and source, I will ignore the interaction term. Also, when dosage=0, the weight gain for three sources are the same. Therefore, β_0 will be the baseline parameter with dosage 0 for 3 sources.
 - (d) Let:

iSource1 = 1 represent being methionine from Source 1; 0 otherwise;

iSource2 = 1 represent being methionine from Source 2 : 0 otherwise:

iSource3 = 1 represent being methionine from Source 3; 0 otherwise.

x represent dosage values, then:

 $x_1 = iSource1 \times x$ represent dosage value from Source 1;

 $x_2 = iSource2 \times x$ represent dosage value from Source 2;

 $x_3 = iSource3 \times x$ represent dosage value from Source 3;

The suggested model will be,

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_{11} x_1^2 + \beta_{22} x_2^2 + \beta_{33} x_3^2 + e$$

3. (a)
$$H_0: \frac{\mu_1}{2} + \frac{\mu_2}{2} - \mu_3 = 0$$
, then $A = \begin{bmatrix} 0.5 & 0.5 & -1 \end{bmatrix}_{r \times (p+1)}$.

$$F = \frac{(A\hat{\beta} - h)'(A(X'X)^{-1}A')^{-1}(A\hat{\beta} - h)/r}{SSE/(n - p - 1)}$$

$$= \frac{(A\begin{bmatrix} \hat{\mu}_1 \\ \hat{\mu}_2 \\ \hat{\mu}_3 \end{bmatrix})'(A\begin{bmatrix} 0.25 & 0 & 0 \\ 0 & 0.25 & 0 \\ 0 & 0 & 0.25 \end{bmatrix} A')^{-1}(A\begin{bmatrix} \hat{\mu}_1 \\ \hat{\mu}_2 \\ \hat{\mu}_3 \end{bmatrix})}{SSE/(12 - 2 - 1)}$$

$$= \frac{24(0.5\hat{\mu}_1 + 0.5\hat{\mu}_2 - \hat{\mu}_3)^2}{SSE}$$

where $A = [0.5 \ 0.5 \ -1]$.

- (b) Given $\hat{\mu}_1 = 5.6$, $\hat{\mu}_2 = 7.9$, $\hat{\mu}_3 = 6.1$, SSE = 12.8, $\alpha = 0.05$, $F = \frac{24(0.5 \times 5.6 + 0.5 \times 7.9 6.1)^2}{12.8} = 0.792 < F_{1,9} = 5.12$, which fails to reject H_0 . There is insufficient evidence to suggest the contrast $\frac{\mu_1}{2} + \frac{\mu_2}{2} \mu_3$ differ from 0.
- 4. (a) $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + e$ $beta_0$ is the mean response for group A. $beta_1$ is the mean difference between group B and A. $beta_2$ is the mean difference between group C and A. $beta_3$ is the mean difference between group D and A.
 - (b) The errors must be iid normal with mean 0 and constant variance.
 - (c) $\beta_1 = \mu_B \mu_A$. The confidence interval for β_1 :

$$\hat{\beta}_1 \pm t_{n-p-1}^* se(\hat{\beta}_1) = -11.5 \pm t_{196,0.025} \sqrt{19.45^2 \times 0.04}$$
$$= -11.5 \pm 1.97 \times 3.89$$
$$= (-19.17, -3.83)$$

(d) $A\hat{\beta} = \hat{\beta}_0 + \hat{\beta}_1$, where $A = \begin{bmatrix} 1 \ 1 \ 0 \ 0 \end{bmatrix}$. $Var(A\hat{\beta}) = AVar(\hat{\beta})A' = A\sigma^2(X'X)^{-1}A' = 0.02\hat{\sigma}^2$. The confidence interval for $\beta_0 + \beta_1$:

$$\hat{\beta}_0 + \hat{\beta}_1 \pm t_{n-p-1}^* se(\hat{\beta}_0 + \hat{\beta}_1) = 37.5 - 11.5 \pm t_{196,0.025} \sqrt{19.45^2 \times 0.02}$$
$$= 26 \pm 1.97 \times 2.75$$
$$= (20.58, 31.42)$$