## Problem Set 3: Solution

## Part One: Hand-Written Exercise

## 1. (a) Let

$$\hat{\sigma}_r^2 = \frac{1}{n - (k+1-q)} \sum_{1=1}^n \hat{e}_{i,r}^2 = \frac{SSR_r}{n - (k+1-q)}$$

$$\hat{\sigma}_{ur}^2 = \frac{1}{n - (k+1)} \sum_{1=1}^n \hat{e}_{i,ur}^2 = \frac{SSR_{ur}}{n - (k+1)},$$

where  $\hat{e}_{i,r}^2$  and  $\hat{e}_{i,ur}^2$  are the residuals from restricted and unrestricted models respectively. The fact that

$$\frac{(n-k-1+q)\hat{\sigma}_{r}^{2}}{\sigma^{2}} - \frac{(n-k-1)\hat{\sigma}_{ur}^{2}}{\sigma^{2}} \sim \chi^{2}(q)$$

implies

$$\frac{SSR_r - SSR_{ur}}{\sigma^2} \sim \chi^2(q).$$

Moreover, we know that

$$\frac{(n-k-1)\hat{\sigma}_{ur}^2}{\sigma^2} = \frac{SSR_{ur}}{\sigma^2} \sim \chi^2(n-k-1).$$

Finally, since  $(SSR_r - SSR_{ur})$  and  $SSR_{ur}$  are independent of each other, we have the following result:

$$\frac{(SSR_r - SSR_{ur})/q}{SSR_{ur}/(n-k-1)} = \frac{SSR_r - SSR_{ur}}{\sigma^2 q} / \frac{SSR_{ur}}{\sigma^2 (n-k-1)}$$
$$\sim \frac{\chi^2(q)/q}{\chi^2 (n-k-1)/(n-k-1)} \sim F(q, n-k-1). \quad \blacksquare$$

(b)

$$F = \frac{(SSR_r - SSR_{ur})/q}{SSR_{ur}/(n-k-1)} = \frac{\frac{(SSR_r - SSR_{ur})/q}{SST}/(n-k-1)}{\frac{SSR_{ur}}{SST}/(n-k-1)} = \frac{(1 - R_r^2 - 1 + R_{ur}^2)/q}{(1 - R_{ur}^2)/(n-k-1)} = \frac{(R_{ur}^2 - R_r^2)/q}{(1 - R_{ur}^2)/(n-k-1)}.$$

2. (a)

$$\mathbf{R} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 3 & -6 & 0 \\ 0 & -2 & 1 & 0 & 4 & 0 \\ 1 & 3 & 0 & 1 & 0 & 5 \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 4 \\ -1 \\ 7 \end{bmatrix}$$

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(b) Under the null hypothesis  $\mathbf{R}\boldsymbol{\beta} = \mathbf{c}$ , the test statistic F and its distribution is given by:

$$F = \frac{(\mathbf{R}\hat{\boldsymbol{\beta}} - \mathbf{c})' \left[ \mathbf{R} (\mathbf{X}'\mathbf{X})^{-1} \mathbf{R}' \right]^{-1} (\mathbf{R}\hat{\boldsymbol{\beta}} - \mathbf{c})}{6\hat{\sigma}^2} \sim F(6, 144).$$

3. (a)  $y_i = \alpha_0 + \alpha_1 D_{i1} + \alpha_2 D_{i2} + \beta_0 x_i + \beta_1 (x_i D_{i1}) + \beta_2 (x_i D_{i2}) + u_i$ 

(b) i. When  $D_{i1} = 1, D_{i2} = 0$ :  $y_i = (\alpha_0 + \alpha_1) + (\beta_0 + \beta_1)x_i + u_i$ 

ii. When  $D_{i1} = 0$ ,  $D_{i2} = 1$ :  $y_i = (\alpha_0 + \alpha_2) + (\beta_0 + \beta_2)x_i + u_i$ 

iii. When  $D_{i1} = 0, D_{i2} = 0$ :  $y_i = \alpha_0 + \beta_0 x_i + u_i$ 

4. (a) An 1% increase in "expendA" will lead to an  $0.01\beta_1$  unit increase for "voteA".

(b)  $H_0: \beta_1 + \beta_2 = 0.$ 

(c) Let  $\mathbf{R} = (0, 1, 1, 0)$ , and  $\hat{\boldsymbol{\beta}} = (\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2, \hat{\beta}_3)'$ . Then under the null hypothesis, our test statistic t and its distribution is then given by:

$$t = \frac{\mathbf{R}\hat{\boldsymbol{\beta}}}{\sqrt{\mathbf{R}\widehat{\mathrm{Var}}(\hat{\boldsymbol{\beta}})}\mathbf{R}'} = \frac{\mathbf{R}\hat{\boldsymbol{\beta}}}{\hat{\sigma}\sqrt{\mathbf{R}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{R}'}} \sim t(n-4)$$

(d)  $H_0: \beta_1 = \beta_2 = 0.$ 

(e) Let  $\mathbf{R} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$ , and  $\hat{\boldsymbol{\beta}} = (\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2, \hat{\beta}_3)'$ . Then under the null hypothesis, our test statistic F and its distribution is then given by:

$$F = \frac{(\mathbf{R}\hat{\boldsymbol{\beta}})' \left[ \mathbf{R} (\mathbf{X}'\mathbf{X})^{-1} \mathbf{R}' \right]^{-1} (\mathbf{R}\hat{\boldsymbol{\beta}})}{2\hat{\sigma}^2} \sim F(2, n-4).$$