

Homework 8 - Answers.

1 (a) Problem 2:

(i) $H_0: \beta_3 = 0$

$H_A: \beta_3 > 0$

(ii) Note that $\frac{d \log(\text{salary})}{d \text{ros}} = \frac{1}{\text{salary}} \frac{d \text{salary}}{d \text{ros}}$

and $\frac{d \text{salary}}{\text{salary}} \approx \%$ change in salary. Hence, if

$d \text{ros} = 50$, and given that $\frac{d \log(\text{salary})}{d \text{ros}} = 0.00024$

$\%$ salary = $50 \times 0.00024 = 0.012$. Arguably a small effect on salary.

(iii) No effect, means $\beta_3 = 0$. The alternative is $H_A: \beta_3 > 0$. $n = 209$, $K+1 = 4$, hence degrees of freedom = 205. Choose $\alpha = 0.1$ (10%). This is a one-sided test, ($\beta_3 > 0$). Hence,

$$t_{1-\alpha, n-(K+1)}^* = t_{0.9, 205}^* = 1.2857 \quad (\text{Use } tinv \text{ in MATLAB})$$

$$\text{Now, } t = \frac{0.00024}{0.00054} = 0.444 < 1.2857 = t_{0.9, 205}^*$$

Hence, we reject $\beta_3 = 0$.

(iv) Yes. The fact that the test in (iii) rejects the H_0 suggests that "ros" should be included in the regression.

$$\begin{aligned} \text{Problem 8: (i) } V(\hat{\beta}_1 - 3\hat{\beta}_2) &= E\left([\hat{\beta}_1 - 3\hat{\beta}_2 - (\beta_1 - 3\beta_2)]^2\right) \\ &= E\left\{(\hat{\beta}_1 - \beta_1) - 3(\hat{\beta}_2 - \beta_2)\right\}^2 \\ &= E(\hat{\beta}_1 - \beta_1)^2 + 9E(\hat{\beta}_2 - \beta_2)^2 - 3E(\hat{\beta}_1 - \beta_1)(\hat{\beta}_2 - \beta_2) \\ &= V(\hat{\beta}_1) + 9V(\hat{\beta}_2) - 3\text{Cov}(\hat{\beta}_1, \hat{\beta}_2) \end{aligned}$$

The standard-error is $\sqrt{V(\hat{\beta}_1) + 9V(\hat{\beta}_2) - 3\text{Cov}(\hat{\beta}_1, \hat{\beta}_2)}$

(ii) Given that $H_0: \beta_1 - 3\beta_2 = 1$, we write

$R = [0 \ 1 \ -3 \ 0]$ and $r = 1$. Hence,

$$R\beta = r \Leftrightarrow [0 \ 1 \ -3 \ 0] \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} = 1. \text{ Then,}$$

$$\frac{(R\hat{\beta} - r)^T (R(X^T X)^{-1} R^T)^{-1} (R\hat{\beta} - r)}{\hat{\sigma}^2} \sim t_{1-\alpha/2, n-4}$$

(iii) Under H_0 , $\beta_1 = \theta_1 + 3\beta_2$. Hence,

$$\begin{aligned} y &= \beta_0 + [\theta_1 + 3\beta_2]x_1 + \beta_2 x_2 + \beta_3 x_3 + u \\ &= \beta_0 + \theta_1 x_1 + (3x_1 + x_2)\beta_2 + \beta_3 x_3 + u \end{aligned} \quad (1)$$

We can run the regression (1) and directly test if $\theta_1 = 1$ using a t -statistic.

1 (b). Problem 2: First, note that if $y^T = (y_1 \dots y_n)$ we can define

$\text{co } y$. Let $C = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & c_1 & 0 & \dots & 0 \\ \vdots & & c_2 & & \vdots \\ \vdots & & & \ddots & \vdots \\ 0 & \dots & 0 & & c_r \end{bmatrix}$; and note that

$$C^{-1} = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1/c_1 & & & \\ & & 1/c_2 & & \\ & & & \ddots & \\ 0 & & & & 1/c_r \end{bmatrix} \quad \text{Also, if } X = \begin{bmatrix} 1 & x_{11} & \dots & x_{1k} \\ 1 & x_{21} & \dots & x_{2k} \\ \vdots & \vdots & & \vdots \\ 1 & x_{n1} & \dots & x_{nk} \end{bmatrix}$$

$$\text{Then, } XC = \begin{bmatrix} 1 & c_1 x_{11} & \dots & c_r x_{1r} \\ \vdots & c_1 x_{21} & \dots & c_r x_{2r} \\ \vdots & \vdots & & \vdots \\ \vdots & c_1 x_{n1} & \dots & c_r x_{nr} \end{bmatrix}$$

$$\begin{aligned} \text{Hence, } \tilde{\beta} &= ((XC)^T XC)^{-1} (XC)^T \text{co } y \\ &= (C^T X^T X C)^{-1} C^T X^T \text{co } y \\ &= C^{-1} (X^T X)^{-1} (C^T)^{-1} C^T X^T \text{co } y \\ &= C^{-1} (X^T X)^{-1} X^T y \quad \text{since } C \text{ is a scalar} \end{aligned}$$

$$= \begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \vdots \\ \hat{\beta}_K \end{bmatrix} \begin{bmatrix} c_0 \\ c_0/c_1 & \dots \\ \vdots \\ c_0/c_K \end{bmatrix}$$

$$= \begin{bmatrix} c_0 \hat{\beta}_0 \\ c_0 \hat{\beta}_1 \\ \vdots \\ c_0 \hat{\beta}_K \end{bmatrix}$$

1(b) Problem 4:

$$(i) \frac{d \log(\text{wage})}{d \text{Educ}} = \beta_1 + \beta_2 \text{pareduc}$$

Here, the percent change on wages due to a change of year of education is a linear function of the parents years of education. If $\beta_2 > 0$, this implies that % wage gains from increasing the years of education is larger for those with more educated parents. It is unclear what might justify such assumption, or for that matter the reverse, i.e., $\beta_2 < 0$.

$$(ii) \frac{d \log(\text{wages})}{d \text{Educ}} = 0.047 + 0.00078 \text{educ} * \text{pareduc}$$

For two individuals with the same level of education, and one with college educated parents and, the other with high school educated parents, we have

$$\frac{d \log(\text{wages})}{d \text{Educ}} = 0.047 + 0.00078 \text{Educ} \cdot 24 \quad (\text{HSE})$$

$$\frac{d \log(\text{wages})}{d \text{Educ}} = 0.047 + 0.00078 \text{Educ} \cdot 32 \quad (\text{CE})$$

$$\begin{aligned} (\text{CE}) - (\text{HSE}) &= [0.02496 - 0.01872] \text{Educ} \\ &= 0.00624 \text{Educ} \rightarrow \text{difference on returns from education.} \end{aligned}$$

(iii)

$$\frac{\partial \log(\text{wages})}{\partial \text{Educ}} = 0.097 - 0.0016 \text{ parents education}$$

The impact of another year of education depends negatively on parents education.

$H_0: \beta_3 = 0$; $H_A: \beta_3 \neq 0$. Degrees of freedom = $722 - 6 = 716$

At $\alpha = 0.05$, we have $t = -0.0016/0.0012 = -1.33$ and

$t_{716, 0.975}^* = 1.9633$. Hence, since $-1.9633 < -1.33$, we accept H_0 that $\beta_3 = 0$.

2. (a) Here $R = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$, $r = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

$$F_{2, 522}^* = 3.013, \quad F = 111.8. \quad \text{Reject } H_0.$$

(b) Here $R = [0 \ 0 \ 0 \ 1]$, $r = [0]$

$$F_{1, 522}^* = 3.8593, \quad F = 37.99, \quad \text{Reject } H_0.$$

(c) Here $H_0: \beta_1 - \beta_2 - \beta_3 = 0$, so

$$R = [0 \ 1 \ -1 \ -1], \quad r = [0].$$

$$F_{1, 522}^* = 3.8593, \quad F = 28.83. \quad \text{Reject } H_0.$$