Econometrics HW3

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1 7.28

1.1 Part A

	Edu	Exp	$\exp^2/100$	Constant
Coefficient	0.14431	0.042633	-0.095056	0.53089
Robust SE	0.011726	0.012422	0.033796	0.20005

1.2 Part B

The derivative of log wage with respect to education is β_1 and the derivative of log wage with respect to experience is $\beta_2 + \beta_3 exp/50$ so $\theta = \frac{\beta_1}{\beta_2 + \beta_3 exp/50}$. Therefore, for 10 experience, our estimate implied by our regressions is the following:

$$\hat{\theta} = \frac{\hat{\beta}_1}{\hat{\beta}_2 + \hat{\beta}_3 exp/50} \tag{1}$$

$$= \frac{0.1443}{0.0426 - 0.0951(10)/50} \tag{2}$$

$$=6.109$$
 (3)

1.3 Part C

We can find the asymptotic standard error as the square root of the asymptotic variance of the $\hat{\theta}$ estimator, which we can calculate through the delta method:

$$s(\hat{\theta}) = \sqrt{g'(\beta)'Vg'(\beta)},$$

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where V is the asymptotic covariance matrix of the non-intercept coefficients, and $g(\beta) = \frac{\hat{\beta_1}}{\hat{\beta_2} + \hat{\beta_3} exp/50}$. Then,

$$g'(\beta) = \begin{pmatrix} \frac{1}{\beta_2 + \beta_3 exp/50} \\ -\beta_1 \\ \frac{(\beta_2 + \beta_3 exp/50)^2}{-\beta_1 exp/50} \\ \frac{-\beta_1 exp/50}{(\beta_2 + \beta_3 exp/50)^2} \end{pmatrix}$$

We can calculate an estimate for $s(\hat{\theta})$ by plugging in OLS estimates of β and our robust standard error matrix we used in Part A. Our 90% CI is $[\hat{\theta} - 1.645s(\hat{\theta}), \hat{\theta} + 1.645s(\hat{\theta})]$.

1.4 Part D

Our computed $\hat{\theta}$, $s(\hat{\theta})$, and confidence interval are the following:

$$\hat{\theta} = 6.109$$
 $s(\hat{\theta}) = 1.6178$ $CI = [4.4912, 7.7269]$

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2 8.1

Let $\beta = [\beta_1, \beta_2]$ be the CLS estimator of $Y = X_1'\beta_1 + X_2'\beta_2 + e$ subject to the constraint that $\beta_2 = 0$. From definition (8.3),

$$\beta = \underset{\beta_2 = 0}{\arg\min} (Y - X_1 \beta_1 - X_2 \beta_2)' (Y - X_1 \beta_1 - X_2 \beta_2)$$

$$\Rightarrow \mathcal{L} = (Y - X_1 \beta_1 - X_2 \beta_2)' (Y - X_1 \beta_1 - X_2 \beta_2) + \lambda' (\beta_2 - 0)$$

$$\Rightarrow 0 = -2X_1' (Y - X_1 \beta_1 - X_2 \beta_2)$$

$$\Rightarrow X_1' Y = (X_1' X_1) \beta_1$$

$$\Rightarrow \beta_1 = (X_1' X_1)^{-1} X_1' Y.$$

3 8.3

$$\beta = \underset{\beta_1 = -\beta_2}{\min} (Y - X_1 \beta_1 - X_2 \beta_2)' (Y - X_1 \beta_1 - X_2 \beta_2)$$

$$\Rightarrow \mathcal{L} = (Y - X_1 \beta_1 - X_2 \beta_2)' (Y - X_1 \beta_1 - X_2 \beta_2) + \lambda' (\beta_2 + \beta_1)$$

$$\Rightarrow 0 = -2X_1' (Y - X_1 \beta_1 - X_2 \beta_2) + \lambda$$

$$\Rightarrow 0 = -2X_2' (Y - X_1 \beta_1 - X_2 \beta_2) + \lambda$$

$$\Rightarrow 0 = (X_1 - X_2)' (Y - X_1 \beta_1 + X_2 \beta_1)$$

$$\Rightarrow \beta_1 = -\beta_2 = ((X_1 - X_2)' (X_1 - X_2))^{-1} (X_1 - X_2)' Y$$

4 8.4(a)

Let Z = X

$$\alpha = \underset{\beta=0}{\arg\min} (Y - X\beta - \alpha)'(Y - X\beta - \alpha)$$

$$\Rightarrow \mathcal{L} = (Y - X\beta - \alpha)'(Y - X\beta - \alpha) + \lambda'(\beta)$$

$$\Rightarrow 0 = -\vec{1}(Y - X\beta - \alpha)$$

$$\Rightarrow \alpha = \frac{1}{n}\vec{1}'Y = \frac{1}{n}\sum_{i}Y_{i}$$

5 8.22

5.1 Part A

$$\tilde{\beta} = \underset{2\beta_2 = \beta_1}{\arg \min} (Y - X_1 \beta_1 - X_2 \beta_2)' (Y - X_1 \beta_1 - X_2 \beta_2)
\Rightarrow \mathcal{L} = (Y - X_1 \beta_1 - X_2 \beta_2)' (Y - X_1 \beta_1 - X_2 \beta_2) + \lambda' (2\beta_2 - \beta_1)
\Rightarrow 0 = -2X_1' (Y - X_1 \beta_1 - X_2 \beta_2) + \lambda
\Rightarrow 0 = -2X_2' (Y - X_1 \beta_1 - X_2 \beta_2) + 2\lambda
\Rightarrow 0 = (2X_1 + X_2)' (Y - X_1 2\beta_2 - X_2 \beta_2)
\Rightarrow \tilde{\beta}_2 = ((2X_1 + X_2)' (2X_1 + X_2))^{-1} (2X_1 + X_2)' Y
\Rightarrow \tilde{\beta}_1 = 2\tilde{\beta}_2$$

5.2 Part B

$$\begin{split} \sqrt{n}(\tilde{\beta}_2 - \beta_2) &= 2\sqrt{n}((2X_1 + X_2)'(2X_1 + X_2))^{-1}(2X_1 + X_2)'e \\ &= 2(\frac{1}{n}\sum_i (2X_{1,i} + X_{2,i})^2)^{-1}\frac{1}{\sqrt{n}}\sum_i (2X_{1,i} + X_{2,i})e_i \\ &\Rightarrow N\left(0, \frac{E[(2X_{1,i} + X_{2,i})^2e_i^2]}{E[(2X_{1,i} + X_{2,i})^2]^2}\right) \end{split}$$

- 6 9.1
- 7 9.2
- 8 9.4
- 9 9.7