

## Problem Set 4a (Due: March 30, 2020)

### Problem 1

Suppose that the data generating process is determined by the following Cobb-Douglas production function

$$\log Y_i = \alpha_0 + \alpha_1 \log L_{1i} + \alpha_2 \log L_{2i} + \alpha_3 \log K_i + u_i$$

where  $Y_i$  = output of firm  $i$ ,  $L_{1i}$  = production labor of firm  $i$ ,  $L_{2i}$  = non-production labor of firm  $i$ , and  $K_i$  = capital stock of firm  $i$ . Suppose you instead use OLS to estimate the following model:

$$\log Y_i = \beta_0 + \beta_1 \log L_{1i} + \beta_2 \log K_i + u_i$$

using cross-section data on many firms.

1. Will  $E[\hat{\beta}_1] = \alpha_1$  and  $E[\hat{\beta}_2] = \alpha_3$ ? Explain.
2. You think that the error term may be heteroskedastic so you use heteroskedasticity-robust standard errors for  $\hat{\beta}_0$ ,  $\hat{\beta}_1$ , and  $\hat{\beta}_2$ . Would they be valid for testing hypotheses about true values of  $\alpha_1$  and  $\alpha_3$ ? Explain.
3. How will your answers to parts 1 and 2 change if it is known that  $\alpha_2 = 0$ , i.e.  $L_2$  is an irrelevant input in the production function? Explain.
4. Assuming  $\alpha_2 = 0$ , explain how you would test for constant returns to scale.  
*Hint:* See SW Problem 7.9c

### Problem 2

Suppose that the data generating process is now determined by the following Cobb-Douglas production function:

$$\log Y_i = \beta_0 + \beta_1 \log L_{1i} + \beta_2 \log K_i + u_i \quad (1)$$

where  $Y_i$  = output of firm  $i$ ,  $L_{1i}$  = production labor of firm  $i$ , and  $K_i$  = capital stock of firm  $i$ . Suppose you instead use OLS to estimate the following model:

$$\log Y_i = \alpha_0 + \alpha_1 \log L_{1i} + \alpha_2 \log L_{2i} + \alpha_3 \log K_i + u_i \quad (2)$$

where  $L_{2i}$  = non-production labor of firm  $i$  using cross-section data on many firms.

1. Will  $E[\hat{\alpha}_1] = \beta_1$  and  $E[\hat{\alpha}_3] = \beta_2$ ? Explain.

- You think that the error term may be heteroskedastic so you use heteroskedasticity-robust standard errors for  $\hat{\alpha}_0$ ,  $\hat{\alpha}_1$ ,  $\hat{\alpha}_2$ , and  $\hat{\alpha}_3$ . Will the standard errors be valid for testing hypotheses about true values of  $\beta_0$ ,  $\beta_1$  and  $\beta_2$ ? Explain.
- Suppose you estimate both models (1) and (2) by OLS using cross-sectional data on **very** many firms. You test  $H_0 : \alpha_1 = 1$  against  $H_1 : \alpha_1 \neq 1$  using the t-statistics

$$t = \frac{\hat{\alpha}_1 - 1}{s.e.(\hat{\alpha}_1)}$$

and  $H_0 : \beta_1 = 1$  against  $H_1 : \beta_1 \neq 1$  using

$$t = \frac{\hat{\beta}_1 - 1}{s.e.(\hat{\beta}_1)}$$

and the decision rule “reject  $H_0$  if  $|t| > 1.96$ ” (since 1.96 is the 5% critical value). Suppose  $H_0$  is true. Are you more likely, less likely, or equally likely to reject  $H_0$  using the first  $t$  statistic (involving  $\hat{\alpha}_1$ ) than using the second  $t$  statistic (involving  $\hat{\beta}_1$ )? Explain.

- Consider the setup from part 3, but suppose that  $H_0$  is false. Are you more likely, less likely, or equally likely to reject  $H_0$  using the first  $t$  statistic (involving  $\hat{\alpha}_1$ ) than using the second  $t$  statistic (involving  $\hat{\beta}_1$ )? Explain.