

1. **Assessing Consistency.** *This problem highlights the differences between unbiasedness and consistency.*

Suppose you want to estimate the parameters of the model

$$y_i = \beta_1 + \frac{1}{\beta_2}x_i + \varepsilon_i$$

under the standard assumptions for OLS. You run OLS on the regression

$$y_i = \delta_1 + \delta_2x_i + \varepsilon_i$$

to obtain $\hat{\delta}_2$ and then use the estimator $\hat{\beta}_2 = \frac{1}{\hat{\delta}_2}$.

- Is $\hat{\beta}_2$ unbiased?
 - Is $\hat{\beta}_2$ consistent? If so, give a proof and state precisely the assumptions needed.
 - Derive the asymptotic distribution of $\hat{\beta}_2$.
 - Under the additional assumption that ε is normal, what is the maximum likelihood estimator of β_2 ?
2. **Assessing Consistency II.** *This problem asks you to assess consistency in the context of a dataset with missing observations but available proxies for the missing data. It's a nice example of how to check consistency and also offers some insight on how to think about data difficulties.*

You are estimating the relationship

$$y_i = \alpha + \beta x_i + \varepsilon_i$$

where the standard OLS assumptions are again satisfied. Unfortunately, data on x are missing for some observations; however, data on a proxy z for x are always available. The proxy satisfies $x = z + \eta$, where η is uncorrelated with both z and ε .

- Is OLS omitting observations where x is missing consistent?
 - Is OLS using x when it is available and z otherwise consistent?
 - Would the usual OLS standard error formula be correct for the estimator in (b)? What if $\beta = 0$?
 - Which of the above estimation procedures would you prefer for testing the null hypothesis $\beta = 0$ and why?
3. **Heteroscedasticity-Robust Standard Errors and Testing Procedures.** *This problem requires you to repeat part (a) – (b) of Question 6 in Problem Set 6 under the assumption of heteroscedasticity. How does heteroscedasticity affect the following estimates and hypothesis tests?*

- (a) Estimate a Cobb-Douglas production function of the form

$$\ln Q_i = \beta_1 + \beta_2 \ln L_i + \beta_3 \ln K_i + \varepsilon_i.$$

Report robust standard errors. Compare the results to the ones obtained using non-robust standard errors.

- (b) Test the hypothesis of constant returns to scale, using robust estimates of variance matrices. At what level could you reject? Compare your results to the results obtained using non-robust version.
4. **Small Sample Inference Beyond Normality.** Read Victor's "Finite Sample Inference Beyond Normality" handout (*finitesample.pdf*) on the Stellar site. Do the empirical exercise from Example 3. It's based on Temin (2005) which is also available on the course website.
5. **A Quick Review of the Course:** Consider the model, where $Y = X\beta + \epsilon$, where for each t , $\epsilon_t \sim \sigma(e_t - 1)$, where e_t is standard exponential variable such that $E[e_t] = 1$ and $Var[e_t] = 1$. Assume that X are independent of ϵ . Suppose that (x_t, ϵ_t) are i.i.d. across t .

- (a) Do Gauss-Markov assumptions hold for this model?
- (b) Consider the least squares estimator $\hat{\beta}$. Compute $E[\hat{\beta}|X]$ and $Var[\hat{\beta}|X]$. Is $\hat{\beta}$ normally distributed in finite samples, conditional on X ?
- (c) Carefully, but briefly, explain the label "BLUE". Is OLS BLUE in this set-up?
- (d) Consider estimating the following effect

$$E[y_t|x_t = x''] - E[y_t|x_t = x'] = (x'' - x')'\beta$$

Give an economic example where such an effect might be of interest. Is $(x'' - x')'\hat{\beta}$ BLUE (=Minimum Variance Linear Unbiased Estimator) for this effect? Why or why not?

- (e) (Bonus). Is OLS the BUE (best unbiased estimator) in this model? A brief answer suffices.
- (f) What is the large sample distribution of $\hat{\beta}$? Make any additional primitive assumptions you might need. [Note: high level assumptions will receive partial credit.]
- (g) Construct a consistent estimator for the large sample variance of $\hat{\beta}$. Prove its consistency by making any additional assumptions you need.
- (h) Suppose we want to test the null hypothesis $H_0 : \beta_j = 0$ vs $H_A : \beta_j < 0$. Construct a t-statistic for testing this hypothesis. Derive its limit distribution and describe how to select critical value for this test to maintain the level of significance equal to 5%.
- (i) Suppose the sample size $n = 6$. Do you expect the large sample distribution to be a good approximation to the exact distribution of the t-statistic in question h.. Discuss how to get the exact distribution of the t-statistic. How would you generate p-values (or critical values) for checking the hypothesis of question h. that would be valid even for $n = 6$?