

Note:

- 1 This is an open-book exam. But please finish it independently.
- 2 The TA will email you the exam 5 minutes before the exam.
- 3 Students should scan their answers and email back to [tzhou11@jhu.edu](mailto:tzhou11@jhu.edu) within 20 minutes after the exam.
- 4 The TA will be available for questions on Skype or Zoom during the exam.

**Directions:**

1. Any theorem in Hansen's book may be invoked without proof, but should be cited in your proofs.
2. Show all of your work and explain your reasoning: partial credit is given for partial solutions.

**QUESTIONS:**

1. (10 points) Take the model

$$y_i = \mathbf{x}'_{1i}\boldsymbol{\beta}_1 + \mathbf{x}'_{2i}\boldsymbol{\beta}_2 + e_i$$

$$\mathbb{E}[e_i|\mathbf{x}_i] = 0$$

$$\mathbb{E}[e_i^2|\mathbf{x}_i] = \sigma^2$$

where  $\mathbf{x}_i = (\mathbf{x}_{1i}, \mathbf{x}_{2i})$  with  $\mathbf{x}_{1i}$  is  $k_1 \times 1$  and  $\mathbf{x}_{2i}$  is  $k_2 \times 1$ . Consider the short regression

$$y_i = \mathbf{x}'_{1i}\hat{\boldsymbol{\beta}}_1 + \hat{e}_i$$

and define the error variance estimator

$$s^2 = \frac{1}{n-k} \sum_{i=1}^n \hat{e}_i^2$$

Find  $\mathbb{E}[s^2 | \mathbf{X}]$ .

2. (10 points) The model is

$$y_i = x_i\beta + e_i$$

$$\mathbb{E}[e_i|x_i] = 0$$

where  $x_i \in \mathbb{R}$ . Consider the two estimators

$$\hat{\beta} = \frac{\sum_i^n x_i y_i}{\sum_{i=1}^n x_i^2}$$

$$\tilde{\beta} = \frac{1}{n} \sum_{i=1}^n \frac{y_i}{x_i}.$$

- (a) Under the stated assumptions, are both estimators consistent for  $\beta$ ?
- (b) Are there conditions under which either estimators is efficient?  
(Hint: consider the *Cauchy-Schwarz inequality*)

3. (10 points) In the linear projection model  $y_i = \alpha + \mathbf{x}_i' \boldsymbol{\beta} + e_i$ , consider the restriction  $\boldsymbol{\beta} = \mathbf{0}$ .

- (a) Find the constrained least-squares (CLS) estimator of  $\alpha$  under the restriction  $\boldsymbol{\beta} = \mathbf{0}$ .
- (b) Find an expression for the efficient minimum distance estimator of  $\alpha$  under the restriction  $\boldsymbol{\beta} = \mathbf{0}$ .

4. (10 points) Take the model

$$y_i = x_i \beta_1 + x_i^2 \beta_2 + e_i$$

$$\mathbb{E}[e_i | x_i] = 0$$

where  $y_i$  is wages (dollars per hour) and  $x_i$  is age. Describe how you would test the hypothesis that the expected wage for a 40-year-old worker is \$20 an hour.

5. (10 points) You want to test  $\mathbb{H}_0 : \boldsymbol{\beta}_2 = \mathbf{0}$  against  $\mathbb{H}_1 : \boldsymbol{\beta}_2 \neq \mathbf{0}$  in the model

$$y_i = \mathbf{x}_{1i}' \boldsymbol{\beta}_1 + \mathbf{x}_{2i}' \boldsymbol{\beta}_2 + e_i$$

$$\mathbb{E}[\mathbf{x}_i e_i] = \mathbf{0}$$

You read a paper which estimates model

$$y_i = \mathbf{x}_{1i}' \hat{\boldsymbol{\gamma}}_1 + (\mathbf{x}_{2i} - \mathbf{x}_{1i})' \hat{\boldsymbol{\gamma}}_2 + \hat{e}_i$$

and reports a test of  $\mathbb{H}_0 : \boldsymbol{\gamma}_2 = \mathbf{0}$  against  $\mathbb{H}_1 : \boldsymbol{\gamma}_2 \neq \mathbf{0}$ . Is this related to the test you wanted to conduct?

6. (20 points) You have  $n$  i.i.d. observations  $(y_i, x_{1i}, x_{2i})$ , and consider two alternative regression models

$$(1) \quad \begin{aligned} y_i &= \mathbf{x}'_{1i} \boldsymbol{\beta}_1 + e_{1i} \\ \mathbb{E}[\mathbf{x}_{1i} e_{1i}] &= 0 \end{aligned}$$

$$(2) \quad \begin{aligned} y_i &= \mathbf{x}'_{2i} \boldsymbol{\beta}_2 + e_{2i} \\ \mathbb{E}[\mathbf{x}_{2i} e_{2i}] &= 0 \end{aligned}$$

where  $\mathbf{x}_{1i}$  and  $\mathbf{x}_{2i}$  have at least some different regressors. (For example, Eq. (1) is a wage regression on geographic variables and Eq. (2) is a wage regression on personal appearance measurements. ) You want to know if model 1 or model 2 fits the data better. Define  $\sigma_1^2 = \mathbb{E}[e_{1i}^2]$  and  $\sigma_2^2 = \mathbb{E}[e_{2i}^2]$ . You decide that the model with the smaller variance fits (e.g. , model 1 fits better if  $\sigma_1^2 < \sigma_2^2$ .) You decide to test for this by testing the hypothesis of equal fit  $\mathbb{H}_0 : \sigma_1^2 = \sigma_2^2$  against the alternative of unequal fit  $\mathbb{H}_1 : \sigma_1^2 \neq \sigma_2^2$ . For simplicity, suppose that  $e_{1i}$  and  $e_{2i}$  are observed.

- (a) Construct an estimator  $\hat{\theta}$  of  $\theta = \sigma_1^2 - \sigma_2^2$ .
- (b) Find the asymptotic distribution of  $\sqrt{n}(\hat{\theta} - \theta)$  as  $n \rightarrow \infty$ .
- (c) Find an estimator of the asymptotic variance of  $\hat{\theta}$ .
- (d) Propose a test of asymptotic size  $\alpha$  of  $\mathbb{H}_0$  against  $\mathbb{H}_1$ .
- (e) Suppose the test accepts  $\mathbb{H}_0$ . Briefly, what is your interpretation?

7. (15 points) Consider the structural equation

$$(3) \quad y_i = \beta_0 + \beta_1 x_i + \beta_2 x_i^2 + e_i$$

with  $x_i$  treated as endogenous so that  $\mathbb{E}[x_i e_i] \neq 0$ . Assume  $y_i$  and  $x_i$  are scalar. Suppose we also have a scalar instrument  $z_i$  which satisfies

$$\mathbb{E}[e_i | z_i] = 0$$

so in particular  $\mathbb{E}[e_i] = 0$ ,  $\mathbb{E}[z_i e_i] = 0$  and  $\mathbb{E}[z_i^2 e_i] = 0$ .

- (a) Should  $x_i^2$  be treated as endogenous or exogenous?
- (b) Suppose we have a scalar instrument  $z_i$  which satisfies
- (4) 
$$x_i = \gamma_0 + \gamma_1 z_i + u_i$$

with  $u_i$  independent of  $z_i$  and mean zero.

Consider using  $(1, z_i, z_i^2)$  as instruments. Is this a sufficient number of instruments? (Would this be just-identified, over-identified, or under-identified)?

- (c) Write out the reduced form equation for  $x_i^2$ . Under what condition on the reduced form parameters in Eq. (4) are the parameters in Eq. (3) identified?

8. (15 points) Consider the model

(5) 
$$y_{1i} = \mathbf{x}_i' \boldsymbol{\beta} + e_i$$

$$\mathbf{x}_i = \boldsymbol{\Gamma}' \mathbf{z}_i + \mathbf{u}_i$$

(6) 
$$\mathbb{E}[\mathbf{z}_i e_i] = \mathbf{0}$$

(7) 
$$\mathbb{E}[\mathbf{z}_i \mathbf{u}_i'] = \mathbf{0}$$

with  $y_{1i}$  scalar and  $\mathbf{x}_i$  and  $\mathbf{z}_i$  each a  $k$  vector. You have a random sample  $(y_{1i}, \mathbf{x}_i, \mathbf{z}_i : i = 1, \dots, n)$ . Take the control function equation

$$e_i = \mathbf{u}_i' \boldsymbol{\gamma} + \varepsilon_i$$

(8) 
$$\mathbb{E}[\mathbf{u}_i \varepsilon_i] = \mathbf{0}$$

Inserting into the structural equation we find

$$y_{1i} = \mathbf{x}_i' \boldsymbol{\beta} + \mathbf{u}_i' \boldsymbol{\gamma} + \varepsilon_i.$$

If  $\mathbf{u}_i$  is observed, the control function estimator  $(\hat{\boldsymbol{\beta}}, \hat{\boldsymbol{\gamma}})$  is OLS estimation of this equation.

- (a) If  $\mathbf{u}_i$  is observed, show that  $(\hat{\boldsymbol{\beta}}, \hat{\boldsymbol{\gamma}})$  is consistent.
- (b) If  $\mathbf{x}_i$  is endogenous, write down the formula of the IV estimator  $\hat{\boldsymbol{\beta}}_{IV}$ .
- (c) Suppose now  $\mathbf{u}_i$  is not observed. Assume  $\mathbf{x}_i$  is *endogenous*, i.e.  $\mathbb{E}[\mathbf{x}_i e_i] \neq \mathbf{0}$  with assumptions in (6) and (7) still maintained. Mimicking the control function procedure stated in the question, explain how to conduct a control function approach to obtain a consistent estimator  $\hat{\boldsymbol{\beta}}_{CF}$ .
- (d) Show that  $\hat{\boldsymbol{\beta}}_{IV}$  in (b) and  $\hat{\boldsymbol{\beta}}_{CF}$  in (c) are numerically identical, i.e.  $\hat{\boldsymbol{\beta}}_{IV} = \hat{\boldsymbol{\beta}}_{CF}$ .