

Econometrics 710  
Final Exam, Spring 2011  
Sample Answers

1.

$$\hat{\beta} = \beta + \left( \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i \mathbf{x}_i' \right)^{-1} \left( \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i e_i \right) \rightarrow_p \beta + \mathbf{Q}_{xx}^{-1} E(\mathbf{x}_i e_i)$$

Because the equation is just-identified,

$$\begin{aligned} \tilde{\beta} &= \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{x}_i' \right)^{-1} \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i y_i \right) \\ &= \beta + \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{x}_i' \right)^{-1} \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i e_i \right) \\ &\xrightarrow{p} \beta + \mathbf{Q}_{zx}^{-1} 0 = \beta \end{aligned}$$

Thus

$$\begin{aligned} \delta &= \text{plim}_{n \rightarrow \infty} (\hat{\beta} - \tilde{\beta}) \\ &= \beta + \beta + \mathbf{Q}_{xx}^{-1} E(\mathbf{x}_i e_i) - \beta \\ &= \mathbf{Q}_{xx}^{-1} E(\mathbf{x}_i e_i) \end{aligned}$$

2. Equation (3) means that  $\mathbf{x}_i$  is exogenous. Under this assumption,  $\delta = 0$

3. Differencing the above equations,

$$\begin{aligned} \hat{\beta} - \tilde{\beta} &= \left( \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i \mathbf{x}_i' \right)^{-1} \left( \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i e_i \right) - \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{x}_i' \right)^{-1} \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i e_i \right) \\ &= \left( \left( \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i \mathbf{x}_i' \right)^{-1} \quad - \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{x}_i' \right)^{-1} \right) \left( \frac{1}{n} \sum_{i=1}^n \begin{pmatrix} \mathbf{x}_i \\ \mathbf{z}_i \end{pmatrix} e_i \right). \end{aligned}$$

4. Since  $E(\mathbf{x}_i e_i) = 0$ ,  $E(\mathbf{z}_i e_i) = 0$  and

$$E \left( \begin{pmatrix} \mathbf{x}_i \\ \mathbf{z}_i \end{pmatrix} \begin{pmatrix} \mathbf{x}_i' & \mathbf{z}_i' \end{pmatrix} e_i^2 \right) = \begin{bmatrix} E(\mathbf{x}_i \mathbf{x}_i' e_i^2) & E(\mathbf{x}_i \mathbf{z}_i' e_i^2) \\ E(\mathbf{z}_i \mathbf{x}_i' e_i^2) & E(\mathbf{z}_i \mathbf{z}_i' e_i^2) \end{bmatrix} = \begin{bmatrix} \Omega_{xx} & \Omega_{xz} \\ \Omega_{zx} & \Omega_{zz} \end{bmatrix} = \Omega,$$

say, then by the CLT

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n \begin{pmatrix} \mathbf{x}_i \\ \mathbf{z}_i \end{pmatrix} e_i \rightarrow_d N(0, \Omega)$$

5.

$$\begin{aligned} \sqrt{n} (\hat{\beta} - \tilde{\beta}) &= \left( \left( \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i \mathbf{x}_i' \right)^{-1} \quad \left( \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{x}_i' \right)^{-1} \right) \left( \frac{1}{\sqrt{n}} \sum_{i=1}^n \begin{pmatrix} \mathbf{x}_i \\ \mathbf{z}_i \end{pmatrix} e_i \right) \\ &\rightarrow_d \left( \mathbf{Q}_{xx}^{-1} \quad -\mathbf{Q}_{zx}^{-1} \right) N(0, \Omega) \sim N(0, V) \end{aligned}$$

where

$$\begin{aligned} V &= \left( \mathbf{Q}_{xx}^{-1} \quad -\mathbf{Q}_{zx}^{-1} \right) \begin{bmatrix} \Omega_{xx} & \Omega_{xz} \\ \Omega_{zx} & \Omega_{zz} \end{bmatrix} \begin{pmatrix} \mathbf{Q}_{xx}^{-1} \\ -\mathbf{Q}_{zx}^{-1} \end{pmatrix} \\ &= \mathbf{Q}_{xx}^{-1} \Omega_{xx} \mathbf{Q}_{xx}^{-1} - \mathbf{Q}_{zx}^{-1} \Omega_{zx} \mathbf{Q}_{xx}^{-1} - \mathbf{Q}_{xx}^{-1} \Omega_{xz} \mathbf{Q}_{zx}^{-1} + \mathbf{Q}_{zx}^{-1} \Omega_{zz} \mathbf{Q}_{zx}^{-1} \end{aligned}$$

6. Under (4),

$$\begin{bmatrix} \Omega_{xx} & \Omega_{xz} \\ \Omega_{zx} & \Omega_{zz} \end{bmatrix} = \sigma^2 \begin{bmatrix} \mathbf{Q}_{xx} & \mathbf{Q}_{xz} \\ \mathbf{Q}_{zx} & \mathbf{Q}_{zz} \end{bmatrix}$$

so

$$\begin{aligned} V &= \mathbf{Q}_{xx}^{-1} \Omega_{xx} \mathbf{Q}_{xx}^{-1} - \mathbf{Q}_{zx}^{-1} \Omega_{zx} \mathbf{Q}_{xx}^{-1} - \mathbf{Q}_{xx}^{-1} \Omega_{xz} \mathbf{Q}_{xz}^{-1} + \mathbf{Q}_{zx}^{-1} \Omega_{zz} \mathbf{Q}_{xz}^{-1} \\ &= \sigma^2 (\mathbf{Q}_{xx}^{-1} - \mathbf{Q}_{zx}^{-1} - \mathbf{Q}_{xx}^{-1} + \mathbf{Q}_{zx}^{-1} \mathbf{Q}_{zz} \mathbf{Q}_{xz}^{-1}) \\ &= \sigma^2 (\mathbf{Q}_{zx}^{-1} \mathbf{Q}_{zz} \mathbf{Q}_{xz}^{-1} - \mathbf{Q}_{xx}^{-1}) \end{aligned}$$

7.

$$\hat{V} = \hat{\sigma}^2 (\hat{\mathbf{Q}}_{zx}^{-1} \hat{\mathbf{Q}}_{zz} \hat{\mathbf{Q}}_{xz}^{-1} - \hat{\mathbf{Q}}_{xx}^{-1})$$

where  $\hat{\mathbf{Q}}_{xx} = \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i \mathbf{x}_i'$ ,  $\hat{\mathbf{Q}}_{xz} = \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i \mathbf{z}_i'$ ,  $\hat{\mathbf{Q}}_{zz} = \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i'$ ,  $\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n \hat{e}_i^2$ , and  $\hat{e}_i = y_i - \mathbf{x}_i' \hat{\beta}$ .

8. A test for  $H_0$  is

$$W_n = n (\hat{\beta} - \tilde{\beta})' \hat{V}^{-1} (\hat{\beta} - \tilde{\beta})$$

Let the distribution in question 5 be  $N \sim N(0, V)$ . Then the asymptotic distribution of  $W_n$  is

$$W_n \rightarrow_d N' V^{-1} N \sim \chi_k^2$$

[Another feasible test would be a GMM overidentification test. But there are some pitfalls in taking this approach. It is important to base the test on all of the moment equations. Thus you need to set

$$g_i(\beta) = \begin{pmatrix} \mathbf{x}_i \\ \mathbf{z}_i \end{pmatrix} (y_i - \mathbf{x}_i' \beta)$$

For example, if you set  $g_i(\beta) = \mathbf{x}_i (y_i - \mathbf{x}_i' \beta)$  and use the LS estimator  $\hat{\beta}$ , then  $\frac{1}{n} \sum_{i=1}^n g_i(\hat{\beta}) = 0$  identically so no test can be based on this moment. Or if you set  $g_i(\beta) = \mathbf{z}_i (y_i - \mathbf{x}_i' \beta)$  and use the 2SLS estimator  $\tilde{\beta}$  then  $\frac{1}{n} \sum_{i=1}^n g_i(\tilde{\beta}) = 0$  identically.]

9. Under exogeneity,  $W_n$  is asymptotically  $\chi_k^2$ . To test exogeneity, we compare  $W_n$  with the  $\chi_k^2$  distribution. If  $W_n$  is smaller than the 5% critical value, we do not reject the hypothesis of exogeneity. If  $W_n$  is larger than the critical value, we reject exogeneity in favor of endogeneity. The test works because under the alternative,  $\hat{\beta} - \tilde{\beta} \rightarrow_p \beta^* - \beta = \mathbf{Q}_{xx}^{-1} \delta \neq 0$ , so  $W_n \rightarrow_p \infty$ .

10. The asymptotic distribution implicitly assumed  $V = \sigma^2 (\mathbf{Q}_{zx}^{-1} \mathbf{Q}_{zz} \mathbf{Q}_{xz}^{-1} - \mathbf{Q}_{xx}^{-1}) > 0$ . This is true iff

$$\mathbf{Q}_{xx}^{-1} < \mathbf{Q}_{zx}^{-1} \mathbf{Q}_{zz} \mathbf{Q}_{xz}^{-1} >$$

or iff

$$\mathbf{Q}_{xx} > (\mathbf{Q}_{zx}^{-1} \mathbf{Q}_{zz} \mathbf{Q}_{xz}^{-1})^{-1} = \mathbf{Q}_{xz} \mathbf{Q}_{zz}^{-1} \mathbf{Q}_{zx}$$

or iff

$$\mathbf{Q}_{xx} - \mathbf{Q}_{xz} \mathbf{Q}_{zz}^{-1} \mathbf{Q}_{zx} > 0$$

This holds when  $V > 0$ , but not generally.