

**Economics 671**  
**Problem Set #5:**

(1) Consider a regression model with just an intercept parameter:

$$y_i = \mu + \epsilon_i, \quad E(\epsilon|X) = 0, \quad E(\epsilon\epsilon'|X) = \Omega,$$

where

$$\Omega = \begin{bmatrix} \sigma_1^2 & 0 & 0 & \cdots & 0 \\ 0 & \sigma_2^2 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & \sigma_n^2 \end{bmatrix}.$$

(1a) Given a random sample of  $n$  observations from this model, calculate the OLS estimator,  $\hat{\mu}$  and derive its variance.

(1b) Calculate the GLS estimator,  $\hat{\mu}_{GLS}$  and derive its variance,  $\text{Var}(\hat{\mu}_{GLS})$ .

(1c) Formally compare the variances in (1a) and (1b). When doing so, you might make note of the following inequality that orders arithmetic and harmonic means:

If  $a_1, \dots, a_n$  are positive real numbers, then

$$\bar{a}_n \geq \frac{n}{\sum_{i=1}^n a_i^{-1}}.$$

(2) Go back to the cars2 data set used in the previous problem set. Now, consider the regression:

$$Price_i = \beta_0 + \beta_1 NotReliable_i + \beta_2 RoadScore_i + \beta_3 MPG_i + \quad (1)$$

$$\beta_4 Zeroto60_i + \beta_5 German_i + \epsilon_i \quad (2)$$

where

$$E(\epsilon|X) = 0, \quad E(\epsilon_i^2|X) = \exp(\alpha_0 + \alpha_1 German_i + \alpha_2 RoadScore_i + \alpha_3 Zeroto60_i).$$

Using methods discussed in class, estimate this model using FGLS. In particular, do the following:

(a) Estimate (1) using OLS and obtain the variance-covariance matrix of the OLS estimator.

- (b) Estimate  $\hat{\alpha}$  and comment on the results. Note: Here you can ignore an issues with the intercept estimator  $\hat{\alpha}_0$  and not worry about the fact that the errors  $\epsilon_i$  have been replaced with estimated values  $\hat{\epsilon}_i$ .
- (c) Calculate the (asymptotic) variance-covariance matrix of the FGLS estimator and compare those results to what you obtained in part (a).

**(3)** Consider the simple regression model:

$$y_i = \beta_0 + \beta_1 x_i^* + \epsilon_i$$

where the “true” scalar  $x_i^*$  is never observed, but instead, we observe and use an error-ridden measure  $x_i$  where

$$x_i = \alpha_0 + \alpha_1 x_i^* + u_i.$$

Suppose we assume that the error of measurement  $u$  is independent of  $x^*$  and  $\epsilon$  and  $\epsilon$  is also independent of  $x^*$ . In this case, derive

$$\text{plim}(\hat{\theta}_1) \quad \text{and} \quad \text{plim}(\hat{\theta}_0)$$

obtained from the model

$$y_i = \theta_0 + \theta_1 x_i + \eta_i.$$

Comment on the role of  $\alpha_0$  on these results, and whether (and perhaps when) the standard attenuation result will hold.

Questions to think about (not to be turned in, nor will solutions be provided):

(4) Suppose you have a regression model where a single (i.e., a scalar) right-hand side variable, say  $x_1$ , is believed to be endogenous. Specifically, you seek to estimate the model

$$y = \beta_0 + \beta_1 x_1 + u, \quad E(u|x_1) \neq 0.$$

In addition, suppose you know the following:

- A scalar variable  $z_1$  is a valid instrument and is, in fact, independent of the regression error  $u$
- A scalar variable  $w$  is known to be an invalid instrument (i.e., it is known to be correlated with  $u$ ).
- A scalar variable  $z_2$ , which is potentially an instrument, is known to be independent of  $z_1$  (the valid instrument) and is also known to be correlated with  $w$  (the invalid instrument).

Do the above conditions imply that  $z_2$  is an invalid instrument? Prove or explain in detail why not.

(5:) Consider the *simple* regression model (i.e.,  $x_i$  is a scalar):

$$y_i = \beta_0 + \beta_1 x_i + \epsilon_i,$$

where  $x_i$  is regarded as potentially endogenous, meaning  $E(\epsilon|x) \neq 0$ .

Furthermore, there is a *BINARY* variable (dummy variable)  $z_i$ ,  $z_i \in \{0, 1\}$  that is correlated with  $x$  and satisfies  $E(\epsilon|z) = 0$ .

(5a:) Show that

$$\beta_1 = \frac{E(y|z=1) - E(y|z=0)}{E(x|z=1) - E(x|z=0)}.$$

(5b:) Based on the result in (5a), suggest a simple, consistent estimator of  $\beta_1$ .

(5c:) Now, consider the estimator (known as the instrumental variables estimator) of  $\beta = [\beta_0 \ \beta_1]'$ :

$$\hat{\beta}_{IV} = (Z'X)^{-1}Z'y,$$

where

$$Z \equiv \begin{bmatrix} 1 & z_1 \\ 1 & z_2 \\ \vdots & \vdots \\ 1 & z_n \end{bmatrix}, \quad X \equiv \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix}.$$

Show that the IV estimator of the *slope coefficient* ( $\beta_1$ ), from the above, reduces to a natural estimator of the expression in (5a).