## Econometrics II: Assignment 3

Due: Thursday, March 6th

1 Binary Variables IV Estimator A researcher wants to study the effect of Malaria nets on child mortality in a developing country. She observes  $y_i$ , an indicator of infant death, and  $x_i$ , and indicator of whether the household in question had purchased a Malaria net. Consider estimation of the model

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

by OLS.

1.1 Show that in this particular setting,

$$\begin{array}{rcl} \hat{\beta} & = & \overline{y}_1 - \overline{y}_0 \\ \hat{\alpha} & = & \overline{y}_0 \\ \hat{y}_1 & = & \overline{y}_1 \\ \hat{y}_0 & = & \overline{y}_0, \end{array}$$

where  $\overline{y}_x = \frac{\sum_{i=1}^n y_i \cdot 1\{x_i = x\}}{\sum_{i=1}^n 1\{x_i = x\}}$  is the sample average of  $y_i$  for the subsample where  $x_i = x$ , and where  $\hat{y}_x$  is the fitted value for  $x_i = x$ .

- 1.2 Is  $\hat{\beta}$  a credible estimate for the causal effect of  $\beta$ ? The consideration I am after is *omitted variables*:  $x_i$  is likely correlated with other, unobserved traits of households.
- 1.3 Suppose an experiment had been conducted and Malaria nets had been randomly assigned to households. Suppose perfect compliance with the experiment, thus a household used a Malaria net if, and only if, it was randomized into *treatment*; the other households are *control*. Can you now estimate the causal effect of Malaria nets?
- 1.4 Suppose now more realistically that compliance is imperfect: Some households discard their malaria nets. And maybe there is a secondary market. (We continue to assume, however, that the true causal effect of a malaria net, if it were used, would be the same across households. This homogeneous treatment effect assumption is of course questionable.) However, we assume that having received a malaria net from the experimenter increases the chance that a household uses one. Letting the r.v.  $z_i$  denote receipt of a Malaria net and  $x_i$  use of a Malaria net, argue that you can estimate the causal effect of Malaria nets and express the estimator similarly to the simplification in 2.1.

2 Measurement Error Consider the model

$$Y^* = \beta_0 + \beta_1 X^* + \varepsilon,$$

where OLS assumptions hold with important exceptions that I am about to explain. (Assume throughout that moments exist as needed.) The substantive motivation for this exercise are different forms of measurement error. That is, you are invited to think of  $(Y^*, X^*)$  as "true" (and causally relevant) quantities but you observe some of them subject to additive measurement error.

- **2.1 Measurement Error in Outcome** You do not observe  $Y^*$  but  $Y = Y^* + \eta$ , where  $\eta$  is i.i.d. and independent from all other random variables with mean 0 and variance  $\sigma_{\eta}^2$ . Argue that you can still estimate  $\beta_1$  by OLS of Y on  $X^*$ . What is the estimator's asymptotic distribution?
- **2.2 Errors-in-Variables** You observe  $Y^*$ , however you do not observe  $X^*$  but  $X = X^* + \eta$ , where  $\eta$  is just as before. Argue that  $\beta_1$  can not be estimated by OLS of  $Y^*$  on X. Can you characterize the OLS estimator's bias?
- **2.3 Dual Measurements** As in the previous question but in addition, you observe  $\tilde{X} = X^* + \nu$  of  $X^*$ . Here,  $\nu$  is i.i.d. and independent from all other random variables with mean 0 and variance  $\sigma^2_{\nu}$ .

Argue that you can now estimate  $\beta_1$ . How? Provide an estimator and characterize the estimator's asymptotic distribution.

- 3. Empirical Exercise The data for this exercise are the "Card" data at https://www.ssc.wisc.edu/bhansen/econometrics/. They belong to the paper Card (1995) which is uploaded.
- **3.1** Please replicate the column 2SLS(a) in Table 12.1 and the final column of Table 12.2 in Hansen's textbook. Note that the variable experience has to be created as age-educ-6.
- **3.2** Add nearc2 ("grew up near a 2 year college") to the first stage/reduced form equation. Do results change appreciably?
- 3.3 Estimate the structural equation by TSLS but add instruments nearc4a,nearc4b,near4ca\*age76,near4ca\*age76squared/100 (the last two are generated interaction variables whose names should be self-explanatory). Do results change appreciably?

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