Week 2: OVB

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January 23, 2015

Anoucement: Today's Tasks

- OVB:
 - Regression Anatomy and OVB formula
 - Typical Usage of OVB formula
 - Bad controls
- Examples:
 - Estimating the payoff to attending a more selective college: An application of selection on observables and unobservables (Dale and Krueger (2002))
 - Selection on Observed and Unobserved Variables: Assessing the Effectiveness of Catholic Schools (Altonji, Elder, and Taber (2005))



Regression Review

- What kind of regressions do we want to run?
- ▶ By regression justifications #1 and #3 in MHE, a regression "gives" us a CEF
 - but most of the time, we only interested in one kind of CEF
 - we want a CEF that can be supported by a believable CIA, so that we will comparing like with like
 - this is about the same as runing a regression without OVB
- Questions about CIA?
 - recall from last section, CIA is hard to justified in observational studies

Regression Anatomy (the FWL Theorem)

- ▶ Consider a bivariate population regression: $Y_i = \alpha + \beta D_i + e_i$
 - ► take covariance in either side, $Cov(Y_i,D_i) = Cov(\alpha + \beta D_i + e_i,D_i) \implies$
- Consider a multivariate population regression:

$$Y_i = \alpha + \beta D_i + X_i' \gamma + e_i$$

- regression anatomy; how to get β ?
- ▶ Step 1 (optional): run a regression of Y_i on X_i and get the residual \tilde{Y}_i
- lacktriangle Step 2: run a regression of D_i on X_i and get the residual $\tilde{\mathrm{D}}_i$
- ▶ Step 3: run a bivariate regression of \tilde{Y}_i on \tilde{D}_i

▶ Thus
$$\beta = \frac{\operatorname{Cov}(\tilde{\mathbf{Y}}_i, \tilde{\mathbf{D}}_i)}{\operatorname{Var}(\tilde{\mathbf{D}}_i)}$$
, or $\beta = \frac{\operatorname{Cov}(\mathbf{Y}_i, \tilde{\mathbf{D}}_i)}{\operatorname{Var}(\tilde{\mathbf{D}}_i)}$ (Why?)

OVB Formula

- Consider a short and a long regression:
 - $Y_i = \alpha^s + \beta^s D_i + e_i^s$ $Y_i = \alpha^l + \beta^l D_i + \gamma^l X_i + e_i^l$
- Taking covaraince for the short regression and substituding the long regression gives a simple OVB formula

$$\beta^s = \frac{\text{Cov}(Y_i, D_i)}{\text{Var}(D_i)} \implies$$

$$\beta^s = \frac{\operatorname{Cov}(\alpha^l + \beta^l \operatorname{D}_i + \gamma^l \operatorname{X}_i + e_i^l, \operatorname{D}_i)}{\operatorname{Var}(\operatorname{D}_i)} \implies$$

$$\beta^s = \beta^l + \gamma^l \frac{\text{Cov}(X_i, D_i)}{\text{Var}(D_i)} = \beta^l + \gamma^l \pi_{XD}$$

- where π_{XD} is from the auxiliary regression $x_i = \phi + \pi_{\mathrm{XDD}} i + u_i$
- ▶ short = long + OVB = long + the effect of omitted \times the regression of omitted on include (what if more than one OV?)



Typical Usage of the OVB Formula

- Often times, omitted variables are variables we wish we had
- OVB formula is thus used to predict the likely direction of bias
 - what's the sign of ability bias in a wage-schooling regression?
 - ability is positively correlated with earning
 - theory predicts that correlation between ability and schooling can be ambiguous (why?)
- More exercises:

 - obisity_{ir} = $\alpha + \rho$ residential_sprawl_r + $X'_{ir}\gamma + u_i$
- Some time we want to estimate the extent of OVB
 - \blacktriangleright how large is the OVB if we want to explain away the entire estimate of your β^s
 - catholic school example



Bad Controls: What Should We Control for

- ▶ Recall from CIA, a good control X_i allows us to say:
 - \blacktriangleright conditional on $X_i,$ the assignment of \mathbf{D}_i is as good as ramdomly assigned
 - $E(\mathbf{Y}_0 \mid \mathbf{D}_i, \mathbf{X}_i) = E(\mathbf{Y}_0 \mid \mathbf{X}_i)$
 - $\qquad \qquad \mathbf{E}\left(\eta_{i}\mid \mathbf{D}_{i}, \mathbf{X}_{i}\right) = \mathbf{E}\left(\eta_{i}\mid \mathbf{X}_{i}\right)$
- ▶ By the logic of OVB, it seems that we should control for all variables X_i that are correlated with both the outcome variable Y_i and the treatment D_i
 - \blacktriangleright but we also said that X_i should be pre-treatment variables
 - what happened if we controls for post-treatment variables (or intermediate outcomes)?

Bad Controls: Examples

- Example: suppose we want to know the impact of parents' occupations on kids' test scores: $Y_i = \alpha + \beta occu_i + e_i$
 - should we control for family incomes (potentially affected by occupations)?
 - ▶ is it a good control in terms of giving $E(\eta_i \mid \text{occu}, \text{income}) = E(\eta_i \mid \text{income})$?
 - does it make sense to compare two people with the same income (say 200K) but different occupations? (give an example)
- Example from MHE: should we control for occupations in a wage-schooling regression?
 - occupation is correlated with both the treatment (years of schooling) and outcome (earning)



Bad Controls: Arguments

- Intuitively, having more schooling allow one to work in a high-paying job, which is part of the return of schooling
 - controlling for occupation eliminates this benefit
 - the channel by which wages are increased is partly through occupations
- More importantly, controlling for occupations introduces selection bias since it changes the composition of treatment group and control group

Bad Controls: Composition Effects

Table 1:Bad controls and slection bias

| Person | Y_{0i} | Y_{1i} | W_{0i} | W_{1i} |
|--------|----------|----------|----------|----------|
| A | 1 | 1.5 | Blue | Blue |
| В | 2 | 2.5 | Blue | White |
| C | 3 | 3.5 | White | White |

► ATE: 0.5

Controlling for occupations: 0

The Question and Selection Problem

- Let's discuss Dale and Kruger's paper and review basic concepts in this week
- Question: "Does the 'quality' of the college that students attend influence their subsequent earnings?"
- Selection Problem:
 - ► Can naive comparisons between students from selective colleges and non-selective colleges answer the question?
 - What is the possible selection bias?
 - ▶ If we control for family income, gender, race, SAT scores, and high school ranking, will the selection bias dispear?
 - ▶ If not, what will drive the bias term and how will you predict the sign of the bias?



Solutions to the Selection Problem

- Choices made by school admission comittees and students in the application, screen, and match process might reveal useful information
 - when applying for schools, students reveal their potential ability by the choice of schools they apply to
 - admission comittees have the chance to read students' essays, letters of recommendation, etc
 - given application outcomes, the decision of attending a particular school again reveals unobserved characteristics
- Extend "selection on the observables" to "selection on the observables and unobservables"
 - "match-applicant model"
 - "self-revelation model"



Discussion

- ▶ Table 1: Information from Application Results
 - What does the application results tell about unobservables
- ► Table 3: The effect of college selectivity on earnings
 - Selection-on-observables and selection-on-unobservables
 - How does the matched/rejected results and students' application choices reveal unobserved characteristics

References I

Altonji, Joseph G, Todd E Elder, and Christopher R Taber. 2005. "Selection on Observed and Unobserved Variables: Assessing the Effectiveness of Catholic Schools." *Journal of Political Economy* 113 (1). ERIC: 151.

Dale, Stacy Berg, and Alan B Krueger. 2002. "Estimating the Payoff to Attending a More Selective College: An Application of Selection on Observables and Unobservables*." *The Quarterly Journal of Economics* 117 (4). MIT Press: 1491–1527.