

## Omitted Variables: Multiple Regression

- True model:  $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k + u$ , but we omit  $x_k$ .
- Write the regression of  $x_k$  on the other independent variables:  

$$x_k = \delta_0 + \delta_1 x_1 + \dots + \delta_{k-1} x_{k-1} + v.$$
- Substitute:

$$\begin{aligned} y &= \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_2(\delta_0 + \delta_1 x_1 + \dots + \delta_{k-1} x_{k-1} + v) + u \\ &= (\beta_0 + \beta_k \delta_0) + (\beta_1 + \beta_k \delta_1) x_1 + \dots + (\beta_{k-1} + \beta_k \delta_{k-1}) x_{k-1} + (\beta_k v + u) \end{aligned}$$

- Determine the signs of  $\beta_k$  and  $\delta_1$  to estimate the direction of bias on  $x_1$ .
  - It's easy to guess sign of  $\beta_k$ .
  - It's trickier for  $\delta_1$  because it comes from a multiple regression.
    - It depends on how  $x_1$  is correlated with other  $x$  values.

## Example

- $wage = \beta_0 + \beta_1 educ + \beta_2 exper + \beta_3 abil + u$
- We omit *abil* because it can't be measured.
- Woolridge suggests that *exper* is approximately uncorrelated with *educ* and *abil*.
- This assumption makes it easier to reason about the regression of *abil* on *educ* and *exper*.
  - All variation in *educ* is unique.
  - Coefficient on *educ* should be same as from simple regression.
  - Assume it is positive.
- If we further assume that *abil* is positively correlated with *wage*, we know that the omitted variable bias is positive on *educ*.