

## Exercise B

In the linear model  $IE[y_i] = \beta_1 + \beta_2 x_i$ , suppose that instead of observing  $x_i$  we observe  $x_i^* = x_i + u_i$ , where  $u_i$  is independent of  $x_i$  and  $var(u_i) = \sigma_u^2$ , with  $IE[u_i] = \mu_u$ . Recall that

$$y_i = \beta_1 + \beta_2 x_i + \varepsilon_i \quad (\text{By assumption})$$

Add and subtract  $\beta_2 u_i$   $\rightarrow$

$$\begin{aligned} &= \beta_1 + \beta_2 u_i - \beta_2 u_i + \beta_2 x_i + \varepsilon_i \\ &= \beta_1 + \beta_2 (x_i - u_i) + \beta_2 u_i + \varepsilon_i \end{aligned}$$

Add and subtract  $\beta_2 \mu_u$   $\rightarrow$

$$\begin{aligned} &= \beta_1 + \beta_2 \mu_u + \beta_2 x_i^* + \underbrace{\beta_2 (u_i - \mu_u)}_{\varepsilon_i^*} + \varepsilon_i \end{aligned}$$

$$y_i^* = \beta_1^* + \beta_2^* x_i^* + \varepsilon_i^*.$$

$\varepsilon_i^* \sim \mathcal{N}_*$ , with

$$\begin{aligned} IE[\varepsilon_i^*] &= \beta_2 IE[u_i - \mu_u] + IE[\varepsilon_i] = 0 \\ var(\varepsilon_i^*) &= \beta_2^2 \sigma_u^2 + \sigma^2 = \sigma_*^2 \end{aligned}$$

This is a linear model with covariate  $x_i^*$  and errors  $\varepsilon_i^*$ . The estimator

$$\hat{\beta}_2^* = \frac{cov(x_i^*, y)}{var(x_i^*)} = \frac{\frac{1}{n} \sum_{i=1}^n (x_i^* - \bar{x}^*)(y_i - \bar{y})}{\frac{1}{n} \sum_{i=1}^n (x_i^* - \bar{x}^*)^2}$$

is unbiased and consistent for  $\beta_2^* = \beta_2$ , even under a "perturbed" covariate.

$$IE[\hat{\beta}_2^*] = \beta_2^* = \beta_2.$$

(This follows from the equation above, but you can prove it directly).

However, the estimator for  $\beta_1$  is problematic

$$IE[\hat{\beta}_1^*] = IE[\bar{y} - \bar{x}^* \hat{\beta}_2^*] = \boxed{\beta_1^* \neq \beta_1}$$

because  $\beta_1^* = \beta_1 + \beta_2 \mu_u$ .

Therefore the estimator is inconsistent for  $\beta_2$  unless

1.  $\beta_2 = 0$  ( $\beta_{12}^* = \beta_1 + 0 = \beta_1$ ). The covariate is irrelevant a mom-interesting situation.
2.  $\mu_0 = 0$  ( $\beta_{12}^* = \beta_1 + 0 = \beta_1$ ). The measurement error is unbiased.