## Where is the Bias in OLS Estimates?

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

Recall that the estimate of  $\boldsymbol{\beta}$  is defined as

(1) 
$$\hat{\beta} = \sum_{i}^{n} (x_{i} - \bar{x})(y_{i} - \bar{y}) / \sum_{i}^{n} (x_{i} - \bar{x})^{2}$$

We can reduce the algebra slightly

substitute: 
$$a = \sum_{i}^{n} (x_{i} - \bar{x})^{2}$$
  
$$\sum_{i}^{n} (x_{i} - \bar{x})(y_{i} - \bar{y}) = \sum_{i}^{n} (x_{i} - \bar{x}) y_{i}$$

Substitute these values into (1)

(2) 
$$\hat{\beta} = (1/a) \sum_{i}^{n} (x_i - \bar{x}) y_i$$

now substitute  $\alpha + \beta x_i + \varepsilon_i$  for  $y_i$  and expand

(3) 
$$\hat{\beta} = (1/a) \sum_{i}^{n} (x_{i} - \bar{x}) (\alpha + \beta x_{i} + \epsilon_{i}) =$$
  
(1/a)  $\sum_{i}^{n} (x_{i} - \bar{x}) \alpha + (1/a) \sum_{i}^{n} \beta(x_{i} - \bar{x}) x_{i} + (1/a) \sum_{i}^{n} (x_{i} - \bar{x}) + \epsilon_{i}$ 

Some simple algebra tricks

Recall: 
$$\sum_{i}^{n} (x_i - \overline{x}) = 0$$
 so  $(1/a) \sum_{i}^{n} (x_i - \overline{x})\alpha = 0$ 

also

$$\sum_{i}^{n} (x_{i} - \bar{x})x_{i} = \sum_{i}^{n} (x_{i} - \bar{x})(x_{i} - \bar{x}) = \sum_{i}^{n} (x_{i} - \bar{x})^{2}$$
so
$$(1/a) \sum_{i}^{n} \beta(x_{i} - \bar{x})^{2} = \beta \left[\sum_{i}^{n} (x_{i} - \bar{x})^{2}\right] / \left[\sum_{i}^{n} (x_{i} - \bar{x})^{2}\right] = \beta$$

(4) 
$$\hat{\beta} = \beta + (1/a) \sum_{i}^{n} (x_i - \bar{x}) \epsilon_i$$
  
so  
 $\hat{\beta} = \beta + (1/a) \sum_{i}^{n} (x_i - \bar{x}) (\epsilon_i - \bar{\epsilon})$ 

Recall the definition of covariance

$$\hat{\sigma}_{x,y} = \hat{Cov}(x,y) = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

Divide top and bottom of right-hand-side terms in previous equation by (n-1)

$$\hat{\beta} = \beta + \frac{[1/(n-1)]\sum_{i=1}^{n} (x_i - \bar{x})(\epsilon_i - \bar{\epsilon})}{[1/(n-1)]\sum_{i=1}^{n} (x_i - \bar{x})^2}$$

Notice that:

$$\hat{\sigma}_{xe} = [1/(n-1)] \sum_{i=1}^{n} (x_i - \overline{x}) (\epsilon_i - \overline{\epsilon})$$
$$\hat{\sigma}_{x}^2 = [1/(n-1)] \sum_{i=1}^{n} (x_i - \overline{x})^2$$

Substitute these values into previous equation for  $\beta$ :

$$\hat{\boldsymbol{\beta}} = \boldsymbol{\beta} + \frac{\hat{\boldsymbol{\sigma}}_{xe}}{\hat{\boldsymbol{\sigma}}_{x}^{2}}$$

Notice that if the correlation between x and  $\epsilon_i$  is zero, the estimate of  $\beta$  equals the true value. In contrast, if x and  $\epsilon_i$  are correlated, then the estimate will systematically differ from the true value (hence it is biased)