

$$E[\hat{\beta}_1] = \beta_1 + E_{\tilde{x}} \left[E \left[\frac{\sum_{i=1}^n (x_i - \bar{x}) u_i}{\sum_{i=1}^n (x_i - \bar{x})^2} \mid \tilde{x} \right] \right]$$

by Law of iterated expectations
 O by ASA #1

$$= \beta_1 + E_{\tilde{x}} \left[\frac{\sum_{i=1}^n (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2} E[u_i \mid \tilde{x}] \right]$$

$$= \beta_1 \quad \square$$

Note this is where
 biased estimators
 come from. When
 this $\neq 0$

Insert (*)

$$\frac{\hat{\beta}_n - \beta_n}{SE(\hat{\beta}_n)} \xrightarrow{d} N(0, 1)$$

why? CLT said if V_i is iid & has finite
 2nd moments,

$$\frac{\left(\frac{1}{n} \sum_{i=1}^n V_i - E[V_i] \right)}{SE\left(\frac{1}{n} \sum_{i=1}^n V_i \right)} \xrightarrow{d} N(0, 1)$$

So, to use this, we take

$V_i = (x_i - \bar{x}) u_i$ (iid & finite 2nd moment).

\Rightarrow By CLT

$$\frac{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}) u_i - E[(x_i - \bar{x}) u_i]}{SE\left(\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}) u_i \right)} \xrightarrow{d} N(0, 1)$$