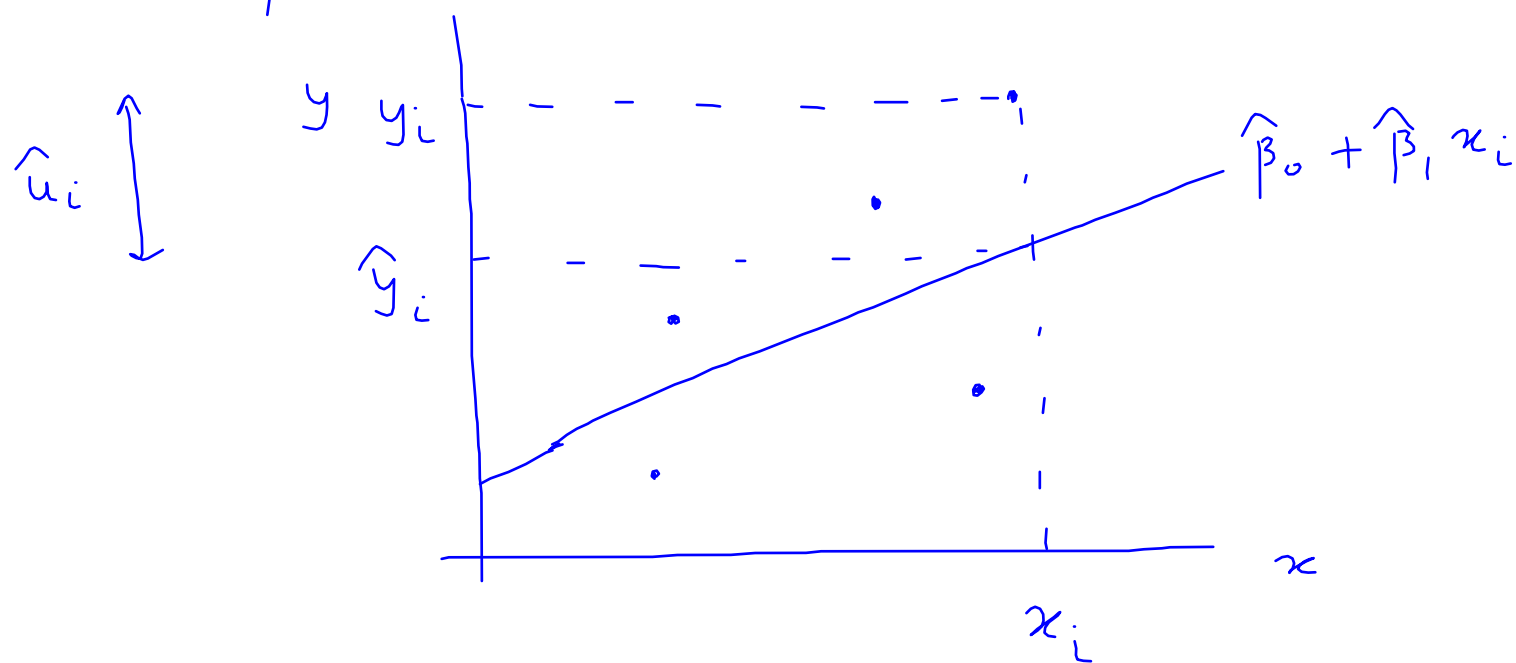


Estimated regression line:

$$\hat{\beta}_0 + \hat{\beta}_1 x_i = \hat{y}_i$$



For  $i$ th obs.  $y_i$ : dep. var. value

$\hat{y}_i$ : fitted value

$$\hat{u}_i \text{ (residual)} = y_i - \hat{y}_i$$

$\hat{\beta}_0$  and  $\hat{\beta}_1$  also minimize the sum of squared residuals (SSR)  $\rightarrow \sum_{i=1}^n \hat{u}_i^2$

Method  $\rightarrow$  also called ordinary least squares (OLS)

$$\hat{\beta}_1 = \frac{\text{how } x \text{ \& } y \text{ covary}}{\text{how } x \text{ varies}}$$

$$= \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

;  $\bar{x}, \bar{y}$  :  
sample  
means

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

## Properties of OLS

Sum / avg. of OLS residuals

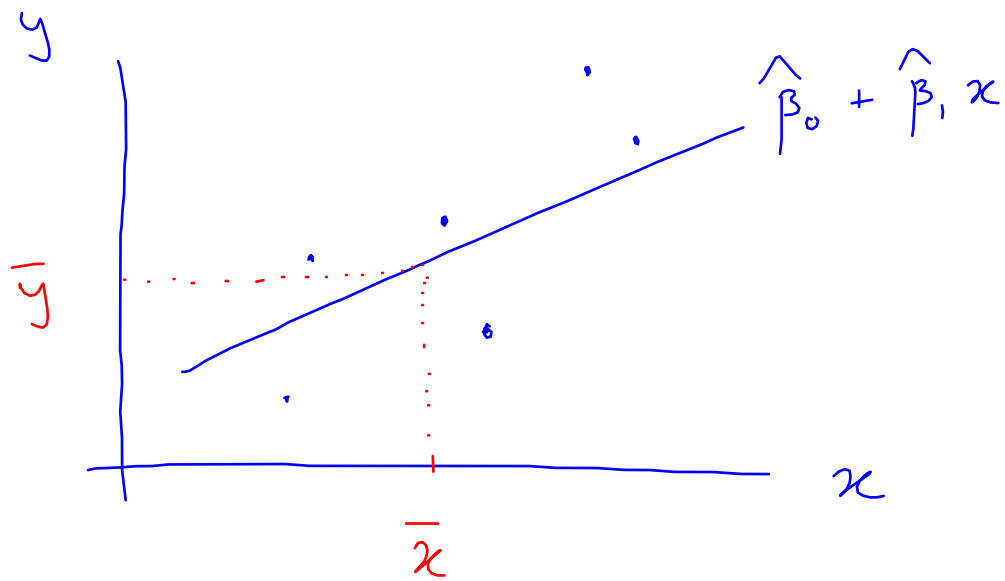
$$\sum_{i=1}^n \hat{u}_i = 0$$

Sample correlation b/w  $x$  and  $\hat{u}$

$$\sum_{i=1}^n x_i \hat{u}_i = 0$$

Sample correlation b/w  $\hat{y}$  and  $\hat{u}$

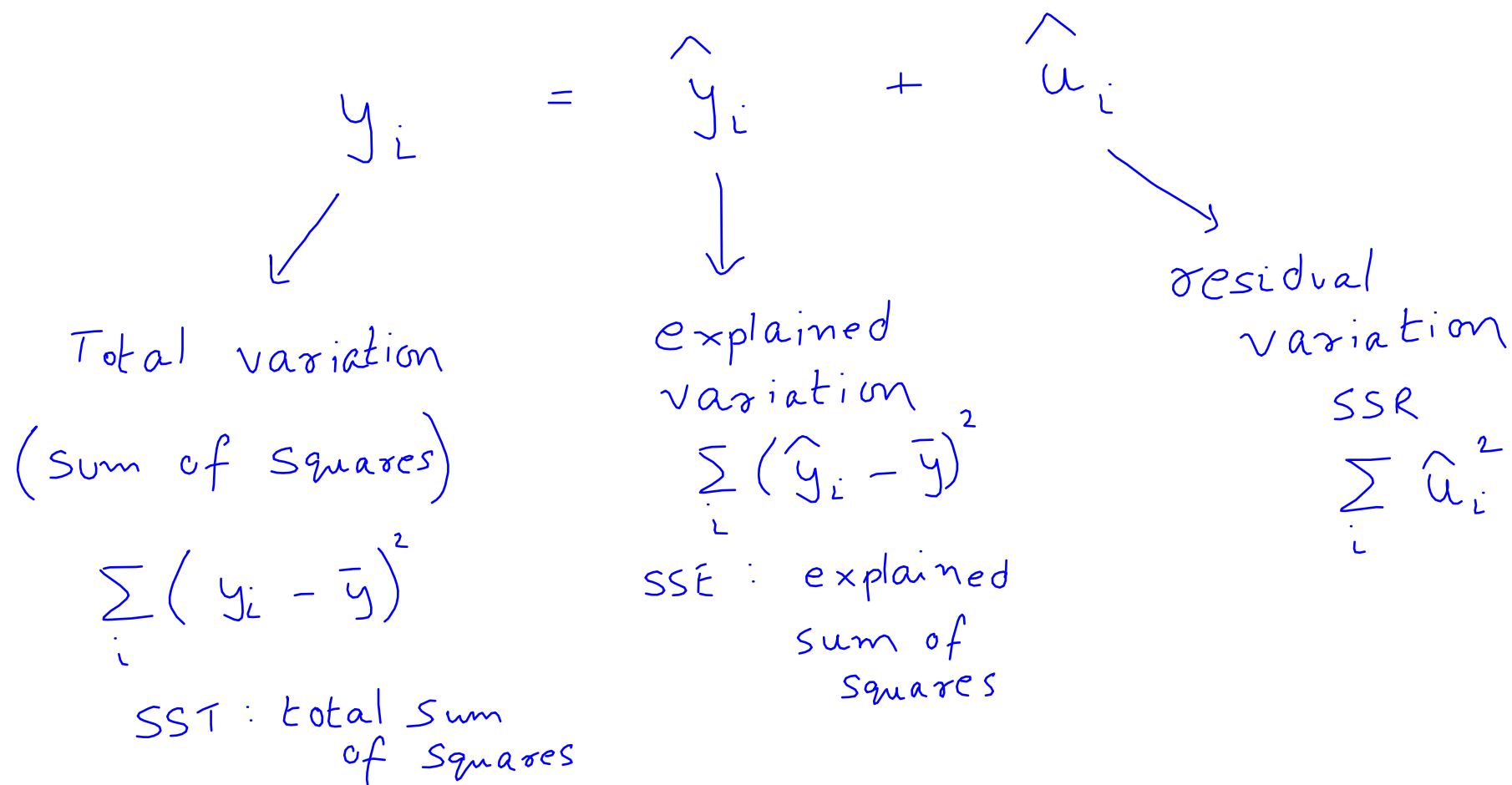
$$\sum_{i=1}^n \hat{y}_i \hat{u}_i = 0$$



$(\bar{x}, \bar{y}) \rightarrow$  on the OLS regression line

$$\bar{y} = \hat{\beta}_0 + \hat{\beta}_1 \bar{x}$$

For each obs.



$$R^2 = \frac{SSE}{SST} = 1 - \frac{SSR}{SST}$$

$$0 \leq R^2 \leq 1$$

also = square of correl<sup>n</sup> b/w  $y$  and  $\hat{y}$

High  $R^2$ : not the ultimate objective