

objective : estimate β_0 and β_1

2 assⁿs : $E(u) = 0$

$$E(u|x) = E(u)$$

$$y = \beta_0 + \beta_1 x + \underbrace{u}_{\text{if } E(u) = 150}$$

$\beta_0 + 150$
 $u - 150$

$$\Rightarrow E(u|x) = 0$$

$$\text{corr.}(x, u) = 0$$

$$E(xu) = 0$$

Deriving OLS Estimates

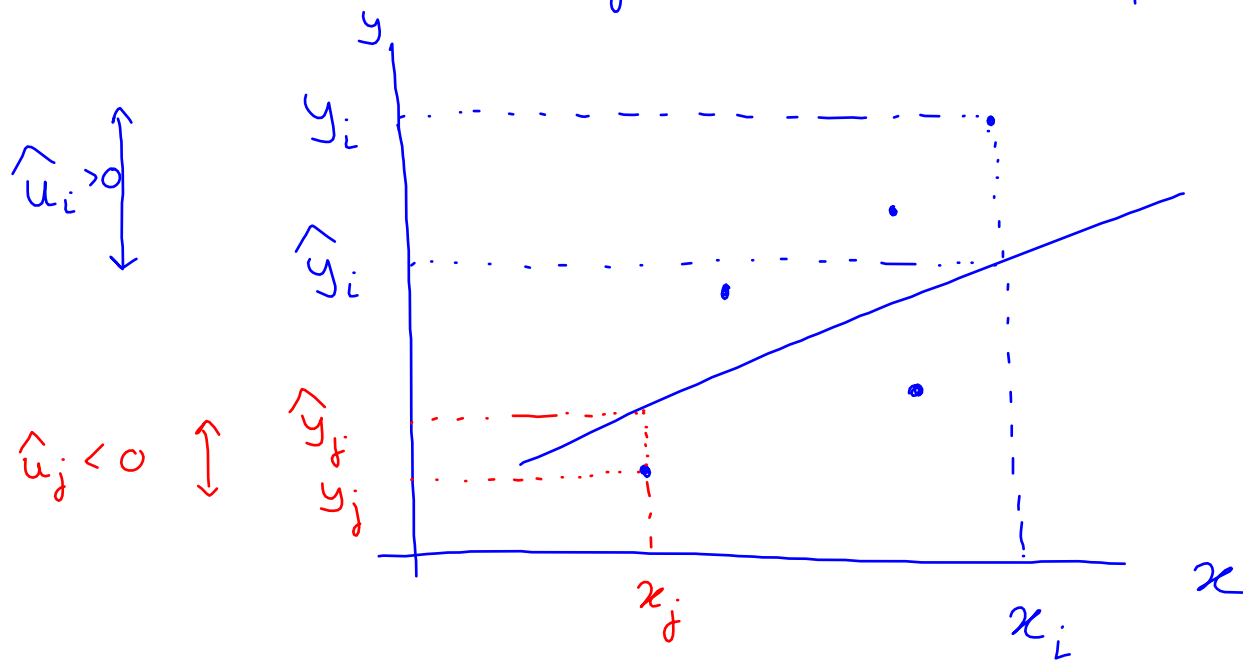
Sample analogs : $\hat{\beta}_0$ and $\hat{\beta}_1$ such that

$$\frac{1}{n} \sum_i (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) = 0$$

n: sample size
i: obs. i
1, 2, ..., n

$$\frac{1}{n} \sum_i x_i (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) = 0$$

Estimated regression line : $\hat{\beta}_0 + \hat{\beta}_1 x_i = \hat{y}_i$



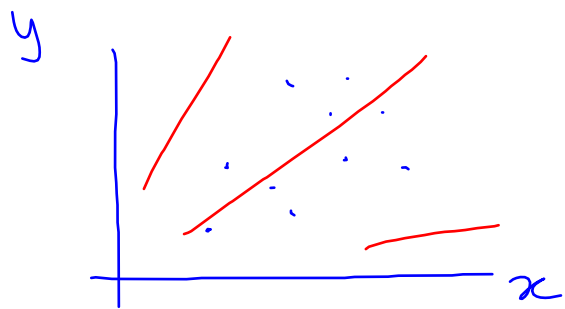
For i th obs. : y_i : value of dep. var.

\hat{y}_i : fitted value

\hat{u}_i : residual

$$\hat{u}_i = y_i - \hat{y}_i$$

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



Method \rightarrow ordinary least squares (OLS)

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

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

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

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

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

Properties of OLS

Sum / avg. of OLS residuals

$$\sum_i \hat{u}_i = 0$$

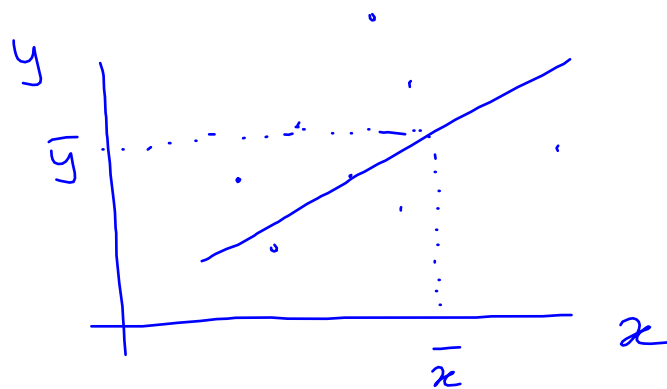
Sample correlⁿ b/w x and \hat{u}

$$\sum_i x_i \hat{u}_i = 0$$

Sample correlⁿ b/w \hat{y} and \hat{u}

$$\sum_i \hat{y}_i \hat{u}_i = 0$$

(\bar{x}, \bar{y}) : on the OLS reg. line



For each obs. i

$$y_i = \hat{y}_i + \hat{u}_i$$

Total variation

explained
variation

residual
variation