

The OLS estimators for β_0 and β_1

$$\hat{\beta}_1 = \frac{\sum_{i=1}^N (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^N (x_i - \bar{x})^2}$$

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

Work through proof. **This is a proof you should know.**

Or

$$\hat{\beta}_1 = \frac{\sum_{i=1}^N x_i y_i - \frac{1}{N} \sum_{i=1}^N x_i \sum_{i=1}^N y_i}{\sum_{i=1}^N x_i^2 - \frac{1}{N} \left(\sum_{i=1}^N x_i \right)^2}$$

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$$\min_{\hat{\beta}_0, \hat{\beta}_1} \sum_{i=1}^N \hat{u}_i^2 = \sum_{i=1}^N (y_i - \hat{y}_i)^2 = \sum_{i=1}^N (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2$$

F.O.C.

$$\frac{\partial (\cdot)}{\partial \hat{\beta}_0} = \sum_{i=1}^N 2(y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)(-1) = -2 \sum_{i=1}^N (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) = 0$$

Aside:

$$\frac{1}{N} \sum_{i=1}^N y_i = \bar{y} \Rightarrow \sum_{i=1}^N y_i = N\bar{y}$$

$$\text{likewise, } \sum_{i=1}^N x_i = N\bar{x}$$

$$\sum_{i=1}^N (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) = 0$$

$$\left(\sum_{i=1}^N y_i \right) - N\hat{\beta}_0 - \hat{\beta}_1 \sum_{i=1}^N x_i = 0$$

$$N\bar{y} - N\hat{\beta}_0 - \hat{\beta}_1 N\bar{x} = 0$$

$$\bar{y} - \hat{\beta}_0 - \hat{\beta}_1 \bar{x} = 0 \Rightarrow \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

See next page for $\hat{\beta}_1$

$$\frac{\partial L}{\partial \hat{\beta}_1} = \sum_{i=1}^N 2(y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)(-x_i) = 0$$

$$= -2 \sum_{i=1}^N x_i (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) = 0$$

$$\sum_{i=1}^N x_i (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) = 0$$

$$\sum_{i=1}^N x_i y_i - \hat{\beta}_0 \sum_{i=1}^N x_i - \hat{\beta}_1 \sum_{i=1}^N x_i^2 = 0$$

$$\sum_{i=1}^N x_i y_i = \hat{\beta}_0 \sum_{i=1}^N x_i + \hat{\beta}_1 \sum_{i=1}^N x_i^2$$

$$\uparrow$$

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

$$\sum_{i=1}^N x_i y_i = (\bar{y} - \hat{\beta}_1 \bar{x}) \sum_{i=1}^N x_i + \hat{\beta}_1 \sum_{i=1}^N x_i^2$$

$$\sum_{i=1}^N x_i y_i = \bar{y} \sum_{i=1}^N x_i - \hat{\beta}_1 \bar{x} \sum_{i=1}^N x_i + \hat{\beta}_1 \sum_{i=1}^N x_i^2$$

Solve for $\hat{\beta}_1$:

$$\sum_{i=1}^N x_i y_i - \bar{y} \sum_{i=1}^N x_i = \hat{\beta}_1 \left[\sum_{i=1}^N x_i^2 - \bar{x} \sum_{i=1}^N x_i \right]$$

$$\hat{\beta}_1 = \frac{\sum_{i=1}^N x_i y_i - \bar{y} \sum_{i=1}^N x_i}{\sum_{i=1}^N x_i^2 - \bar{x} \sum_{i=1}^N x_i}$$

$$= \frac{\sum_{i=1}^N x_i y_i - \frac{1}{N} \sum_{i=1}^N x_i \sum_{i=1}^N y_i}{\sum_{i=1}^N x_i^2 - \frac{1}{N} \left(\sum_{i=1}^N x_i \right)^2}$$

$$\bar{y} = \frac{1}{N} \sum_{i=1}^N y_i$$

$$\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i$$

Can be simplified to:

$$\hat{\beta}_1 = \frac{\sum_{i=1}^N (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^N (x_i - \bar{x})^2}$$

See (1) on next page
for $\sum x_i y_i - \bar{y} \sum x_i = \sum (x_i - \bar{x})(y_i - \bar{y})$
See (2) on next page for
 $\sum x_i^2 - \bar{x} \sum x_i = \sum (x_i - \bar{x})^2$

Intermediate steps for showing the equivalence of our two formulas for $\hat{\beta}_0$ and $\hat{\beta}_1$

1.

$$\begin{aligned}\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x}) &= \sum_{i=1}^n (y_i x_i - \bar{x} y_i - \bar{y} x_i + \bar{y} \bar{x}) = \sum_{i=1}^n y_i x_i - \bar{x} \sum_{i=1}^n y_i - \bar{y} \sum_{i=1}^n x_i + n \bar{y} \bar{x} \\ &= \sum_{i=1}^n y_i x_i - n \bar{x} \bar{y} - \bar{y} \sum_{i=1}^n x_i + n \bar{y} \bar{x} = \sum_{i=1}^n y_i x_i - \bar{y} \sum_{i=1}^n x_i\end{aligned}$$

2.

$$\begin{aligned}\sum_{i=1}^n (x_i - \bar{x})^2 &= \sum_{i=1}^n (x_i^2 - 2\bar{x}x_i + \bar{x}\bar{x})^2 = \sum_{i=1}^n x_i^2 - 2\bar{x} \sum_{i=1}^n x_i + n \bar{x} \bar{x} \\ &= \sum_{i=1}^n x_i^2 - 2n\bar{x}^2 + n\bar{x}^2 = \sum_{i=1}^n x_i^2 - \bar{x} \sum_{i=1}^n x_i\end{aligned}$$