

# Simple Linear Regression Beta Coefficient Variance Derivations

BIOS 6611

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Week 8

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## Formula Refreshers

## Formulas for $\hat{\beta}_0$ and $\hat{\beta}_1$

Our formulas for our beta coefficients are

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2} = \frac{S_{XY}}{S_{XX}}$$
$$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}$$

Our formulas for the standard errors of our beta coefficients are

$$SE(\hat{\beta}_0) = \sqrt{\sigma_{Y|X}^2 \left( \frac{1}{n} + \frac{\bar{X}^2}{\sum_{i=1}^n (X_i - \bar{X})^2} \right)}$$
$$SE(\hat{\beta}_1) = \sqrt{\frac{\sigma_{Y|X}^2}{\sum_{i=1}^n (X_i - \bar{X})^2}}$$

# Helpful Properties and Formula Refresher

$$\text{Var}(Y) = E(Y^2) - [E(Y)]^2 \implies E(Y^2) = \text{Var}(Y) + [E(Y)]^2$$

$$\text{Var}(cY) = c^2 \times \text{Var}(Y)$$

$$\text{SE}(cY) = |c| \times \text{SE}(Y)$$

$$\text{Cov}(aX + b, cY + d) = ac \times \text{Cov}(X, Y)$$

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$$

$$\text{Var}(X - Y) = \text{Var}(X) + \text{Var}(Y) - 2\text{Cov}(X, Y)$$

$$\text{Var}(Y) = \frac{\sum_{i=1}^n (Y_i - \bar{Y})^2}{n - 1}$$

$$\text{Cov}(X, Y) = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{n - 1}$$

$$r_{x,y} = \frac{\text{Cov}(X, Y)}{\text{SD}(X) \text{SD}(Y)} = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^n (X_i - \bar{X})^2 \sum_{i=1}^n (Y_i - \bar{Y})^2}}$$

## Calculating the Variances

## Deriving $Var(\hat{\beta}_1)$

Before deriving the variance of  $\hat{\beta}_1$ , let's do a little rearranging:

$$\begin{aligned}\hat{\beta}_1 &= \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2} \\ &= \frac{\sum_{i=1}^n (X_i - \bar{X}) Y_i - \bar{Y} \sum_{i=1}^n (X_i - \bar{X})}{\sum_{i=1}^n (X_i - \bar{X})^2} \\ &= \frac{\sum_{i=1}^n (X_i - \bar{X}) Y_i}{\sum_{i=1}^n (X_i - \bar{X})^2} \\ &= \frac{(X_1 - \bar{X}) Y_1 + \dots + (X_n - \bar{X}) Y_n}{\sum_{i=1}^n (X_i - \bar{X})^2}\end{aligned}$$

where  $\bar{Y} \sum_{i=1}^n (X_i - \bar{X}) = 0$  because

$$\sum_{i=1}^n (X_i - \bar{X}) = \sum_{i=1}^n X_i - \sum_{i=1}^n \bar{X} = \frac{n}{n} \times \sum_{i=1}^n X_i - n\bar{X} = n\bar{X} - n\bar{X} = 0$$

## Deriving $Var(\hat{\beta}_1)$

Now we are ready to derive our variance for the slope:

$$Var(\hat{\beta}_1) = Var \left[ \frac{(X_1 - \bar{X})Y_1 + \dots + (X_n - \bar{X})Y_n}{\sum_{i=1}^n (X_i - \bar{X})^2} \right]$$

Let's cut to our "whiteboard" to work through the math...



## Deriving $Var(\hat{\beta}_0)$

Now we are ready to derive our variance for the intercept:

$$Var(\hat{\beta}_0) = Var(\bar{Y} - \hat{\beta}_1 \bar{X})$$

Let's cut to our "whiteboard" to work through the math...

## Showing $\text{Cov}(\bar{Y}, \hat{\beta}_1 \bar{X}) = 0$

First we will note that  $E(Y_i Y_j) = E(Y_i) E(Y_j) \quad \forall i \neq j$  by independence and that  $\sum_{i=1}^n (X_i - \bar{X}) = 0$ .

Then we can note another representation of  $\hat{\beta}_1$ :

$$\begin{aligned}\hat{\beta}_1 &= \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2} \\ &= \frac{\sum_{i=1}^n [(X_i - \bar{X})Y_i] - \bar{Y} \sum_{i=1}^n (X_i - \bar{X})}{\sum_{i=1}^n (X_i - \bar{X})^2} \\ &= \frac{\sum_{i=1}^n (X_i - \bar{X})Y_i}{\sum_{i=1}^n (X_i - \bar{X})^2}\end{aligned}$$

Okay, we are ready to show (on the whiteboard):

$$\text{Cov}(\bar{Y}, \hat{\beta}_1 \bar{X}) = \text{Cov}\left(\frac{\sum_{i=1}^n Y_i}{n}, \frac{\sum_{i=1}^n (X_i - \bar{X}) Y_i}{\sum_{i=1}^n (X_i - \bar{X})^2} \times \bar{X}\right) = 0$$

## The Standard Error of the Mean

## Deriving $SE(\hat{\mu}_{Y|X_0})$

We can manipulate our predicted regression equation for this calculation:

$$\hat{\mu}_{Y|X_0} = \hat{\beta}_0 + \hat{\beta}_1 X_0 = (\bar{Y} - \hat{\beta}_1 \bar{X}) + \hat{\beta}_1 X_0 = \bar{Y} + \hat{\beta}_1 (X_0 - \bar{X})$$

Then we can calculate the variance as

$$\begin{aligned} \text{Var}(\hat{\mu}_{Y|X_0}) &= \text{Var}(\bar{Y} + \hat{\beta}_1 (X_0 - \bar{X})) \\ &= \text{Var}(\bar{Y}) + (X_0 - \bar{X})^2 \text{Var}(\hat{\beta}_1) + 2(X_0 - \bar{X}) \text{Cov}(\bar{Y}, \hat{\beta}_1) \\ &= \text{Var}(\bar{Y}) + (X_0 - \bar{X})^2 \frac{\hat{\sigma}_{Y|X}^2}{\sum_{i=1}^n (X_i - \bar{X})^2} \\ &= \frac{\hat{\sigma}_{Y|X}^2}{n} + \frac{\hat{\sigma}_{Y|X}^2}{n-1} \left( \frac{(X_0 - \bar{X})^2}{\hat{\sigma}_X^2} \right) \end{aligned}$$

because  $\hat{\sigma}_X^2 = \sum_{i=1}^n \frac{(X_i - \bar{X})^2}{n-1}$  and  $\text{Var}(\hat{\beta}_1) = \frac{\hat{\sigma}_{Y|X}^2}{\sum_{i=1}^n (X_i - \bar{X})^2}$ .