Fundamentals of

PROGRAM EVALUATION

JESSE LECY

We want to end up here. Need to work backwards.

This is the one formula you need to remember.



Standard Error of the Slope ≈

residual

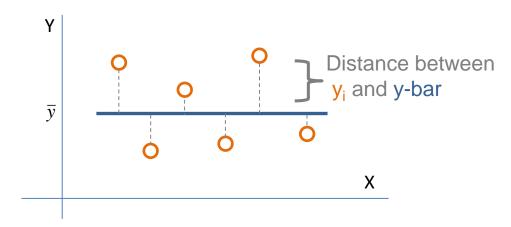
sample size · variance X

$$SE_{b_1} = \frac{sd_e}{\sqrt{(n-1) \cdot \text{var}(x)}}$$

THE VARIANCE CALCULATION

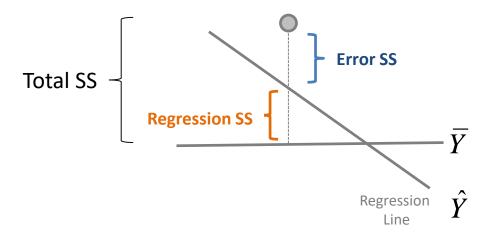
Distance between y_i and y-bar

$$var(y) = \frac{\sum (y_i - \overline{y})^2}{n - 1}$$



Variance: square the distances, add them up, divide by n-1

PARTITIONING THE VARIANCE OF Y

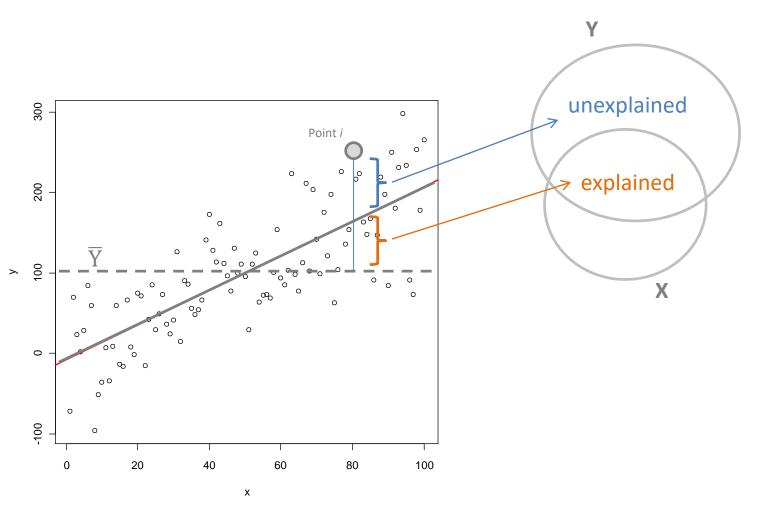


$$TotalSS = \sum (y_i - \overline{y})^2$$

$$Re gressionSS = \sum (\hat{y}_i - \overline{y})^2$$

ErrorSS
$$= \sum (y_i - \hat{y}_i)^2$$

$$TSS = RSS + ESS$$



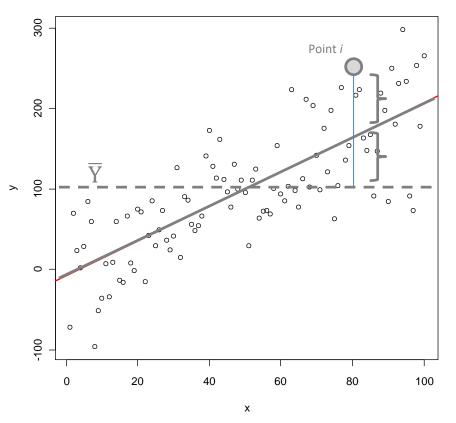
Unexplained: $Y_i - \hat{Y}_i = e_i$

Explained: $\hat{Y}_i - \overline{Y}$

Two parts of the variance of Y

The Venn diagram is a simplified representation of the regression model. In our regression, the explained portion of the variance of outcome will always be the distance from the mean to the predicted value of Y (which always falls on the regression line), and the unexplained portion is the distance between the regression line and the actual data point, also called the residual or the error e.

Standard Error in Regression



$$Y_i - \hat{Y}_i = e_i$$
 $Y_i - \bar{Y}$

$$SSE = \sum e_i^2$$

Sum of Squared Error Terms

$$\hat{\sigma}_{\varepsilon}^2 = \frac{SSE}{n-2}$$

Variance of the residual

The standard error of the slope is one of the most important

standard error. As a result, the size of the standard error will be

proportional to the amount of unexplained variance (plus a couple of other considerations to be covered later).

$$SE_{b_1} = \sqrt{\frac{\hat{\sigma}_{\varepsilon}^2}{\sum (x_i - \bar{x})^2}}$$

Standard error of the slope

STANDARD ERROR IN REGRESSION

$$SE_{\bar{x}} = \frac{S_x}{\sqrt{n}}$$

The size of the standard error of the mean is driven by the variance of the variable, and the sample size.

$$\operatorname{var}(x) = \frac{\sum (x_i - \overline{x})^2}{n - 1}$$
 \Rightarrow

$$(n-1) \cdot \text{var}(x) = \sum (x_i - \overline{x})^2$$

We can write the formula for the standard error of the slope in a couple of ways. I prefer the top because it is explicit about sample size and var(x).

$$SE_{b_1} = \frac{s_{\varepsilon}}{\sqrt{(n-1) \cdot \text{var}(x)}}$$



$$SE_{b_1} = \frac{s_{\varepsilon}}{\sqrt{\sum (x_i - \overline{x})^2}}$$

Similarly, the standard error of the slope is a function of the variance of the residual (the amount of unexplained variance in the outcome), the sample size (n-1), AND the variance of the explanatory variable.

THE INTUITIVE STANDARD ERROR

$$SE_{b_1} = \sqrt{\frac{\sigma_{\varepsilon}^2}{\sum (x_i - \bar{x})^2}}$$

NOTE:

$$\operatorname{var}(x) = \frac{\sum (x_i - \bar{x})^2}{n - 1} \implies (n - 1) \cdot \operatorname{var}(x) = \sum (x_i - \bar{x})^2$$

THUS

$$SE_{b_1} = \sqrt{\frac{\sigma_{\varepsilon}^2}{(n-1)\operatorname{var}(x)}}$$

THEREFORE:

$$SE_{b_1} \approx \frac{residual_y}{sample \ size \cdot var(x)}$$

$$SE_{b_1} = \sqrt{\frac{\hat{\sigma}_{\varepsilon}^2}{\sum (x_i - \bar{x})^2}}$$

Standard error of the Slope

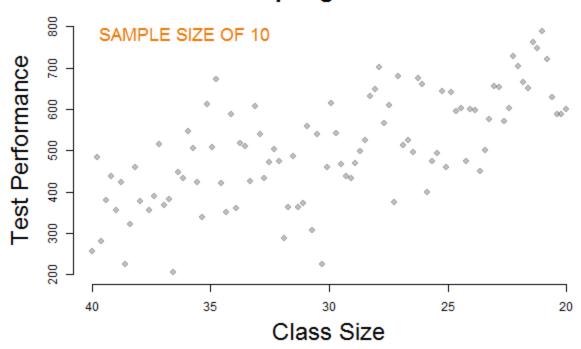
Don't get too caught up with the math. The formula for the standard error of a regression coefficient is actually quite simple when you break it down. There are three moving parts – three things that can affect the size of the standard error. The portion of unexplained variance of the dependent variable (the residual), the sample size of the regression, and the amount of variance in the variable X associated with the regression slope.

Standard Error of the Slope $\approx \frac{\text{residual}}{\text{sample size} \cdot \text{variance X}}$

$$SE_{b_1} = \frac{sd_e}{\sqrt{(n-1) \cdot \text{var}(x)}}$$

STANDARD ERROR AS A SAMPLING STATISTIC

Sampling Process

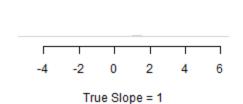


900 Test Performance 500 400 300

Class Size

Repeated Samples

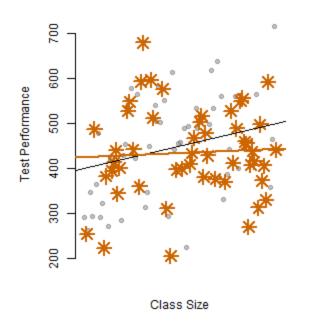
Sampling Distribution

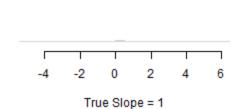


SAMPLE SIZE = 10

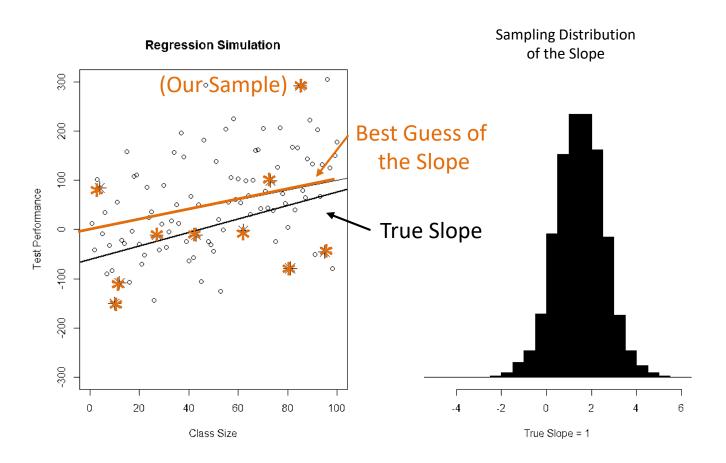
Repeated Samples

Sampling Distribution





SAMPLE SIZE = 50



$$SE_{b_1} = \sqrt{\frac{\sigma_{\varepsilon}^2}{\sum (x_i - \bar{x})^2}}$$

"AVERAGE ERROR"
(OF THE SLOPE ESTIMATE)

TRANSLATING CONCEPTS

Standard Error of the Mean

Standard Error of the Slope

$$SE_{\bar{x}} = \frac{S_x}{\sqrt{n}}$$

$$SE_{b_1} = \frac{S_{\varepsilon}}{\sqrt{(n-1) \cdot \text{var}(x)}}$$

Χ

← Source of Variance →

Υ

Increase sample size

← Reduction of St. Error →

- (1) Increase sample size
- (2) Explain more variance of Y (i.e. add controls)
- (3) Increase variance of X

STANDARD DEVIATION VERSUS STANDARD ERROR

The standard deviation is, how far the data is from the mean, on average.

~

The standard error is, how far our best guest is from 'the truth', on average.

The 'truth' means different things depending upon what kind of standard error you are calculating.

THE ROAD MAP

Of the Mean:

Of the Slope:

Sampling Variance:



$$\sigma_x^2 = \frac{\sum (x_i - \bar{x})^2}{n - 1}$$

(for x)

$$\sigma_{\varepsilon}^2 = \frac{SSE}{n-2} = \frac{\sum e_i^2}{n-2}$$

(using the residual)

Standard Deviation:



$$\sigma_x = \sqrt{\sigma_x^2}$$

$$\sigma_{\varepsilon} = \sqrt{\sigma_{\varepsilon}^2}$$

Standard Error:

$$SE_{\bar{x}} = \frac{\sigma_x}{\sqrt{n}}$$

$$SE_{b_1} = \sqrt{\frac{\sigma_{\varepsilon}^2}{\sum (x_i - \bar{x})^2}}$$

Confidence Interval

$$\mu = \overline{x} \pm t \cdot SE_{\overline{x}}$$

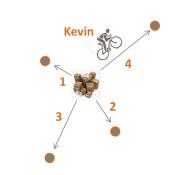
(of the mean)

$$\beta_1 = b_1 \pm t \cdot SE_{b_1}$$

(of the slope)

USEFUL METAPHORS

Variance



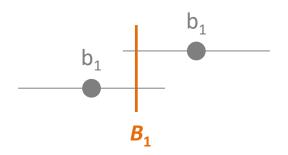
Standard Deviation



Standard Error



Confidence Interval



What should be clear in my mind?

- 1. We split the variance of Y into explained and unexplained portions with a trick, inserting the regression line y-hat.
- 2. The standard error of the slope is derived from the unexplained portion of Y, the residual.