

# Hypothesis Testing

## Regression

# Objectives

- 1 Perform a hypothesis test on the slope of the regression equation
- 2 Perform hypothesis tests on the linear correlation coefficient
- 3 Create a prediction interval for a regression equation

# Slope of a Line

Recall that the slope of a line tells us how much the  $y$ -coordinates change by increasing each  $x$ -coordinate by 1.

# Slope of a Line

Recall that the slope of a line tells us how much the  $y$ -coordinates change by increasing each  $x$ -coordinate by 1.

Previously, we looked at calculating regression equations in the form

$$\hat{y} = mx + b$$

which we can use to predict data values both inside and outside our dataset.

# Slope of a Line

Recall that the slope of a line tells us how much the  $y$ -coordinates change by increasing each  $x$ -coordinate by 1.

Previously, we looked at calculating regression equations in the form

$$\hat{y} = mx + b$$

which we can use to predict data values both inside and outside our dataset.

Also, recall that the coefficient of determination,  $r^2$ , tells us the prediction error decrease we get by using the linear regression equation, as opposed to the mean of the  $y$ -coordinates.

# Using the $y$ -Intercept

However, we might also be able to use the  $y$ -intercept of our linear regression equation as a predictor:

$$y = b$$

# Using the $y$ -Intercept

However, we might also be able to use the  $y$ -intercept of our linear regression equation as a predictor:

$$y = b$$

In this case, the slope of the equation,  $m$ , is 0.

# Using the $y$ -Intercept

However, we might also be able to use the  $y$ -intercept of our linear regression equation as a predictor:

$$y = b$$

In this case, the slope of the equation,  $m$ , is 0.

We can then test whether or not the linear regression model we obtained contributes to predicting values of  $y$ .



# Hypotheses Regarding Slope

For this hypothesis, testing, we are operating under the assumption that the following null hypothesis is true:

The slope of the regression equation does not help in predicting values of  $y$ .

# Hypotheses Regarding Slope

For this hypothesis, testing, we are operating under the assumption that the following null hypothesis is true:

The slope of the regression equation does not help in predicting values of  $y$ .

To put it mathematically:

$$H_0 : m = 0$$

# Hypotheses Regarding Slope

For this hypothesis, testing, we are operating under the assumption that the following null hypothesis is true:

The slope of the regression equation does not help in predicting values of  $y$ .

To put it mathematically:

$$H_0 : m = 0$$

Our alternative hypothesis will be one of the forms

Left-Tailed	Right-Tailed	Two-Tailed
$m < 0$	$m > 0$	$m \neq 0$

# Hypotheses Regarding Slope

For this hypothesis, testing, we are operating under the assumption that the following null hypothesis is true:

The slope of the regression equation does not help in predicting values of  $y$ .

To put it mathematically:

$$H_0 : m = 0$$

Our alternative hypothesis will be one of the forms

Left-Tailed	Right-Tailed	Two-Tailed
$m < 0$	$m > 0$	$m \neq 0$

$H_A$  states that  $x$  is not helpful in predicting  $y$ .

# Test Statistic

Since the population standard deviation of our regression slope is likely not to be known, we can estimate it by using the **estimated standard error of the least squares slope**.

# Test Statistic

Since the population standard deviation of our regression slope is likely not to be known, we can estimate it by using the **estimated standard error of the least squares slope**.

$$\frac{s}{\sqrt{SS_{xx}}}$$

where

$$SS_{xx} = \text{total sum of squares} = \sum x^2 - \frac{1}{n} \left( \sum x \right)^2$$

# Test Statistic

Since the population standard deviation of our regression slope is likely not to be known, we can estimate it by using the **estimated standard error of the least squares slope**.

$$\frac{s}{\sqrt{SS_{xx}}}$$

where

$$SS_{xx} = \text{total sum of squares} = \sum x^2 - \frac{1}{n} \left( \sum x \right)^2$$

Our test statistic,  $t$ , is thus

$$t = \frac{\hat{m}}{s/\sqrt{SS_{xx}}}$$

where  $\hat{m}$  is the slope of our regression equation.

# $p$ -Value

The area under the curve either to the left, right, or outside of our test statistic(s), i.e. the  $p$ -value, can be found by taking  $n - 2$  degrees of freedom with that test statistic.



# Confidence Intervals

We can construct confidence intervals as

$$\hat{m} \pm t_{\alpha/2} \left( s / \sqrt{SS_{xx}} \right)$$

# Confidence Intervals

We can construct confidence intervals as

$$\hat{m} \pm t_{\alpha/2} \left( s / \sqrt{SS_{xx}} \right)$$

Remember, for one-sided confidence intervals, use  $t_{\alpha}$  instead of  $t_{\alpha/2}$ .

# Residuals Review

Recall that the **residual** of a data point is the distance the  $y$ -coordinate of that data point is from the corresponding  $y$ -coordinate on the regression equation.

# Residuals Review

Recall that the **residual** of a data point is the distance the  $y$ -coordinate of that data point is from the corresponding  $y$ -coordinate on the regression equation.

We usually denote the residual of data point  $(x_i, y_i)$  as  $\epsilon_i$ , and it can be found by

$$\epsilon_i = y_{\text{observed}} - y_{\text{predicted}}$$

# Residuals Review

Recall that the **residual** of a data point is the distance the  $y$ -coordinate of that data point is from the corresponding  $y$ -coordinate on the regression equation.

We usually denote the residual of data point  $(x_i, y_i)$  as  $\epsilon_i$ , and it can be found by

$$\epsilon_i = y_{\text{observed}} - y_{\text{predicted}}$$

When performing hypothesis testing regarding the slope of the regression equation, we operate under the following assumptions:

# You Know What Happens When You Assume, Right?

**Assumptions:**

# You Know What Happens When You Assume, Right?

## **Assumptions:**

- The mean of all values of  $\epsilon$  is 0.

# You Know What Happens When You Assume, Right?

## **Assumptions:**

- The mean of all values of  $\epsilon$  is 0.
- The variance of all values of  $\epsilon$  is constant for all values of  $x$ .



# You Know What Happens When You Assume, Right?

## **Assumptions:**

- The mean of all values of  $\epsilon$  is 0.
- The variance of all values of  $\epsilon$  is constant for all values of  $x$ .
- The distribution of  $\epsilon$  is normal.

# You Know What Happens When You Assume, Right?

## **Assumptions:**

- The mean of all values of  $\epsilon$  is 0.
- The variance of all values of  $\epsilon$  is constant for all values of  $x$ .
- The distribution of  $\epsilon$  is normal.
- Values of  $\epsilon$  associated with any two values of  $y$  are independent.

## Example 1

The table below lists the total study times  $x$ , in hours, and the score on a statistics exam  $y$  for 12 students.

(a) Find the least squares regression equation for the data and interpret the slope.

<b>Time</b>	1.9	1.3	1.7	1.2	1.2	2.7	1.7	2.7	2.0	1.8	1.4	2.4
<b>Score</b>	84	81	86	82	77	97	93	96	89	89	88	95

## Example 1

The table below lists the total study times  $x$ , in hours, and the score on a statistics exam  $y$  for 12 students.

(a) Find the least squares regression equation for the data and interpret the slope.

<b>Time</b>	1.9	1.3	1.7	1.2	1.2	2.7	1.7	2.7	2.0	1.8	1.4	2.4
<b>Score</b>	84	81	86	82	77	97	93	96	89	89	88	95

$$\hat{y} = 10.3158x + 69.1711$$

## Example 1

The table below lists the total study times  $x$ , in hours, and the score on a statistics exam  $y$  for 12 students.

(a) Find the least squares regression equation for the data and interpret the slope.

<b>Time</b>	1.9	1.3	1.7	1.2	1.2	2.7	1.7	2.7	2.0	1.8	1.4	2.4
<b>Score</b>	84	81	86	82	77	97	93	96	89	89	88	95

$$\hat{y} = 10.3158x + 69.1711$$

$$\text{Score} = 10.3158(\text{Time}) + 69.1711$$

## Example 1

The table below lists the total study times  $x$ , in hours, and the score on a statistics exam  $y$  for 12 students.

(a) Find the least squares regression equation for the data and interpret the slope.

<b>Time</b>	1.9	1.3	1.7	1.2	1.2	2.7	1.7	2.7	2.0	1.8	1.4	2.4
<b>Score</b>	84	81	86	82	77	97	93	96	89	89	88	95

$$\hat{y} = 10.3158x + 69.1711$$

$$\text{Score} = 10.3158(\text{Time}) + 69.1711$$

For every increased hour in study time, the predicted score increases by about 10.3

## Example 1

(b) At the 5% significance level, test the claim whether study time is useful as a predictor of test score.

## Example 1

(b) At the 5% significance level, test the claim whether study time is useful as a predictor of test score.

$H_0 : m = 0$  (study time is not useful)



## Example 1

(b) At the 5% significance level, test the claim whether study time is useful as a predictor of test score.

$H_0 : m = 0$  (study time is not useful)

$H_A : m \neq 0$  (study time is useful)

## Example 1

(b) At the 5% significance level, test the claim whether study time is useful as a predictor of test score.

$H_0 : m = 0$  (study time is not useful)

$H_A : m \neq 0$  (study time is useful)

$t = 5.5365$  (critical values:  $\pm 2.2281$ )

p-value: 0.0002 ( $\alpha = 0.05$ )

95.0% confidence interval for population slope: (6.1643, 14.4673)

## Example 1

(b) At the 5% significance level, test the claim whether study time is useful as a predictor of test score.

$H_0 : m = 0$  (study time is not useful)

$H_A : m \neq 0$  (study time is useful)

$t = 5.5365$  (critical values:  $\pm 2.2281$ )

p-value: 0.0002 ( $\alpha = 0.05$ )

95.0% confidence interval for population slope: (6.1643, 14.4673)

**Reject the null hypothesis**

## Example 1

(b) At the 5% significance level, test the claim whether study time is useful as a predictor of test score.

$H_0 : m = 0$  (study time is not useful)

$H_A : m \neq 0$  (study time is useful)

$t = 5.5365$  (critical values:  $\pm 2.2281$ )

p-value: 0.0002 ( $\alpha = 0.05$ )

95.0% confidence interval for population slope: (6.1643, 14.4673)

### Reject the null hypothesis

*At the 5% significance level, we have sufficient evidence to reject the claim that study time is not useful in predicting the score on this statistics exam.*

# Objectives

- 1 Perform a hypothesis test on the slope of the regression equation
- 2 Perform hypothesis tests on the linear correlation coefficient
- 3 Create a prediction interval for a regression equation

# Objectives

- 1 Perform a hypothesis test on the slope of the regression equation
- 2 Perform hypothesis tests on the linear correlation coefficient
- 3 Create a prediction interval for a regression equation