

# Math 243

Least Squares Regression Line – Inv. 5.8 and 5.9

# Prediction

So far, we've

- *described* a dataset through graphs and numerical summaries,
- *tested* whether a parameter is a value ( $H_0$  vs.  $H_a$ ), and
- *estimated* a parameter (95% confidence interval)

Today, we'll

- *predict* the value of a quantitative variable based on the value of a second quantitative variable.

# Inv. 5.8: Footlength vs. Height

Given the length of a person's **foot**, *predict* their **height**.

Collect data:

# Inv. 5.8: Footlength vs. Height

Given the length of a person's **foot**, *predict* their **height**.

Collect data:

Observational units: 20 statistics students

Explanatory variable: foot length in cm

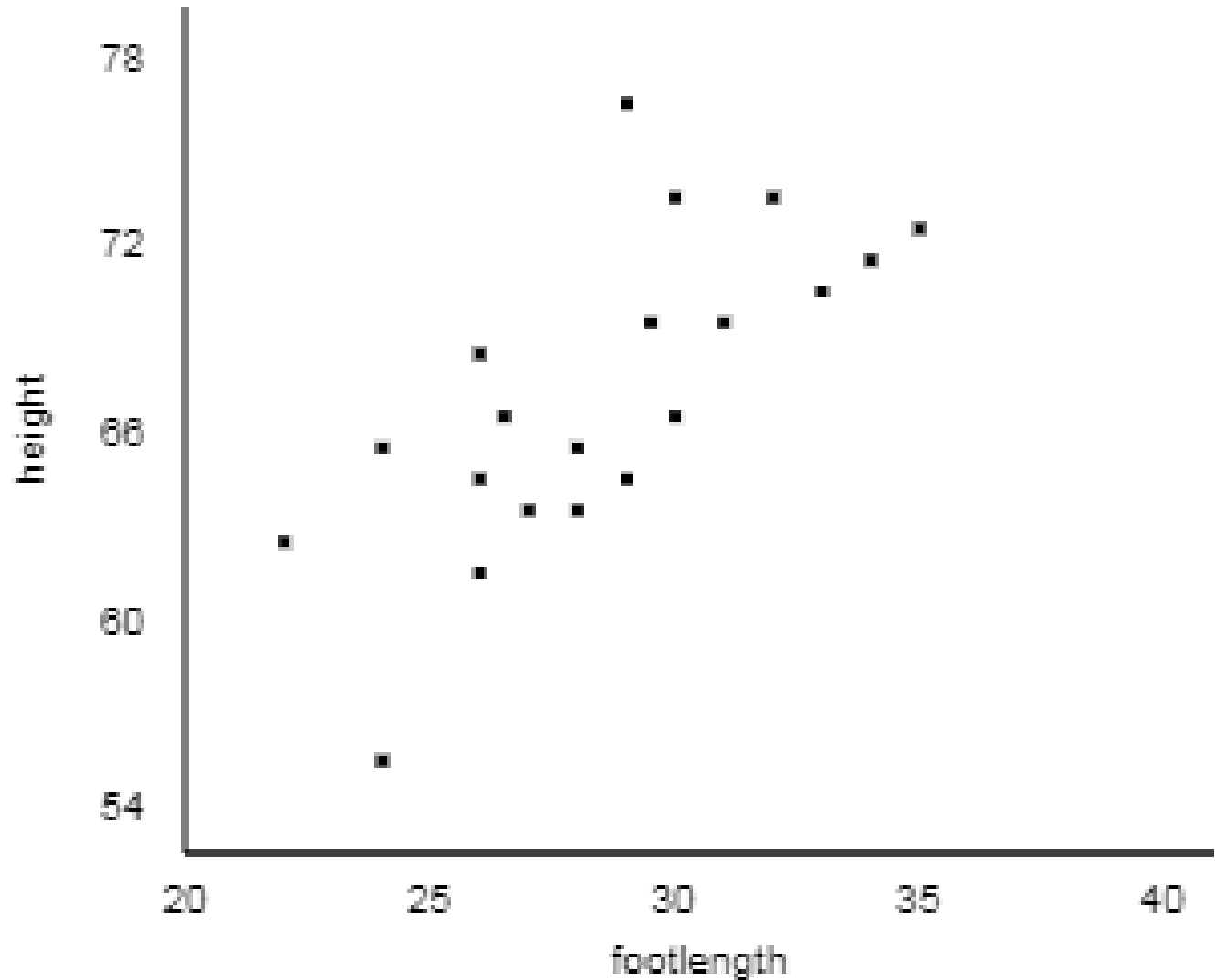
Response variable: height in inches

# Inv. 5.8: Footlength vs. Height

*Direction?*

*Linear?*

*Strength?*



# Inv. 5.8: Footlength vs. Height

*Direction?*

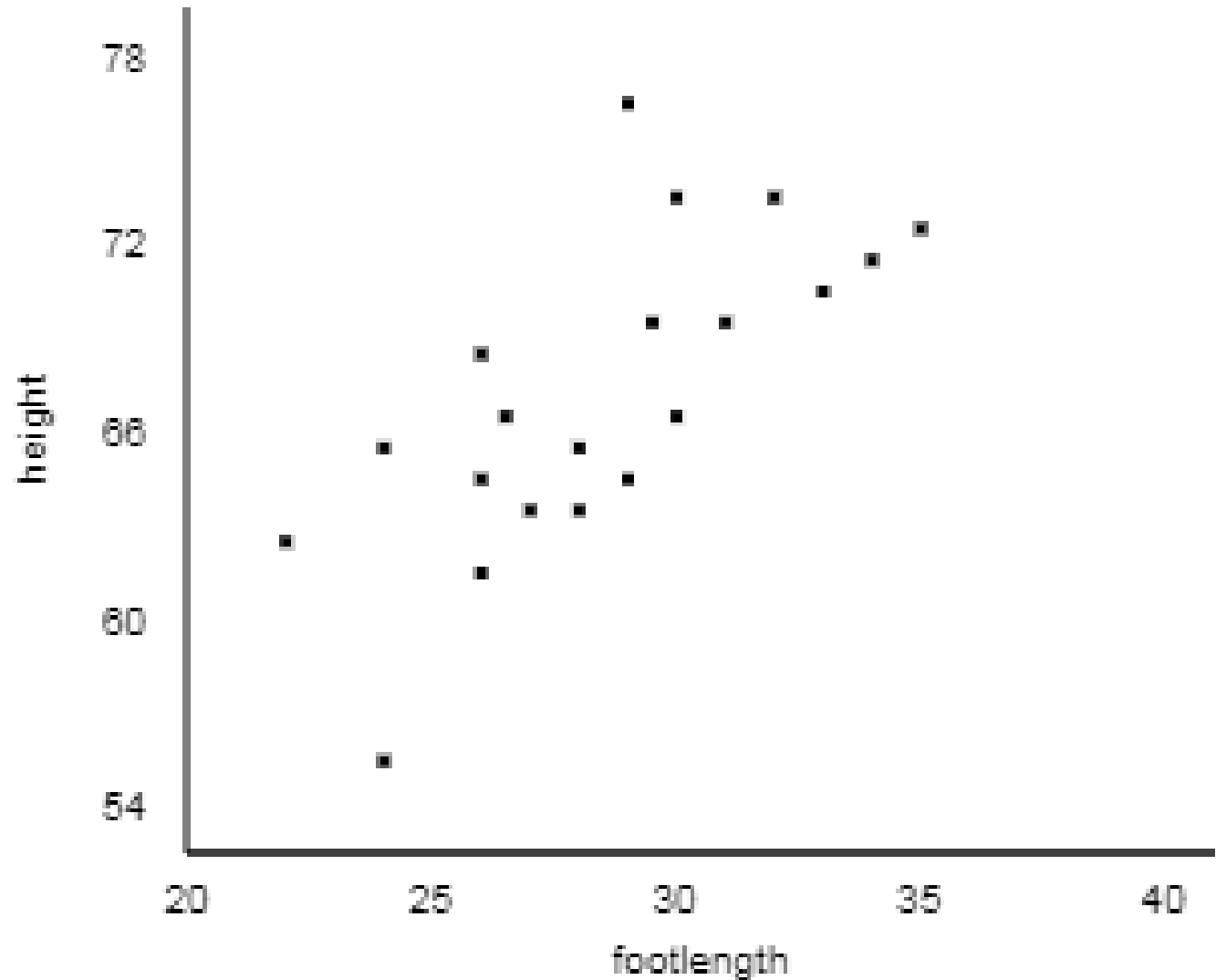
*positive*

*Linear?*

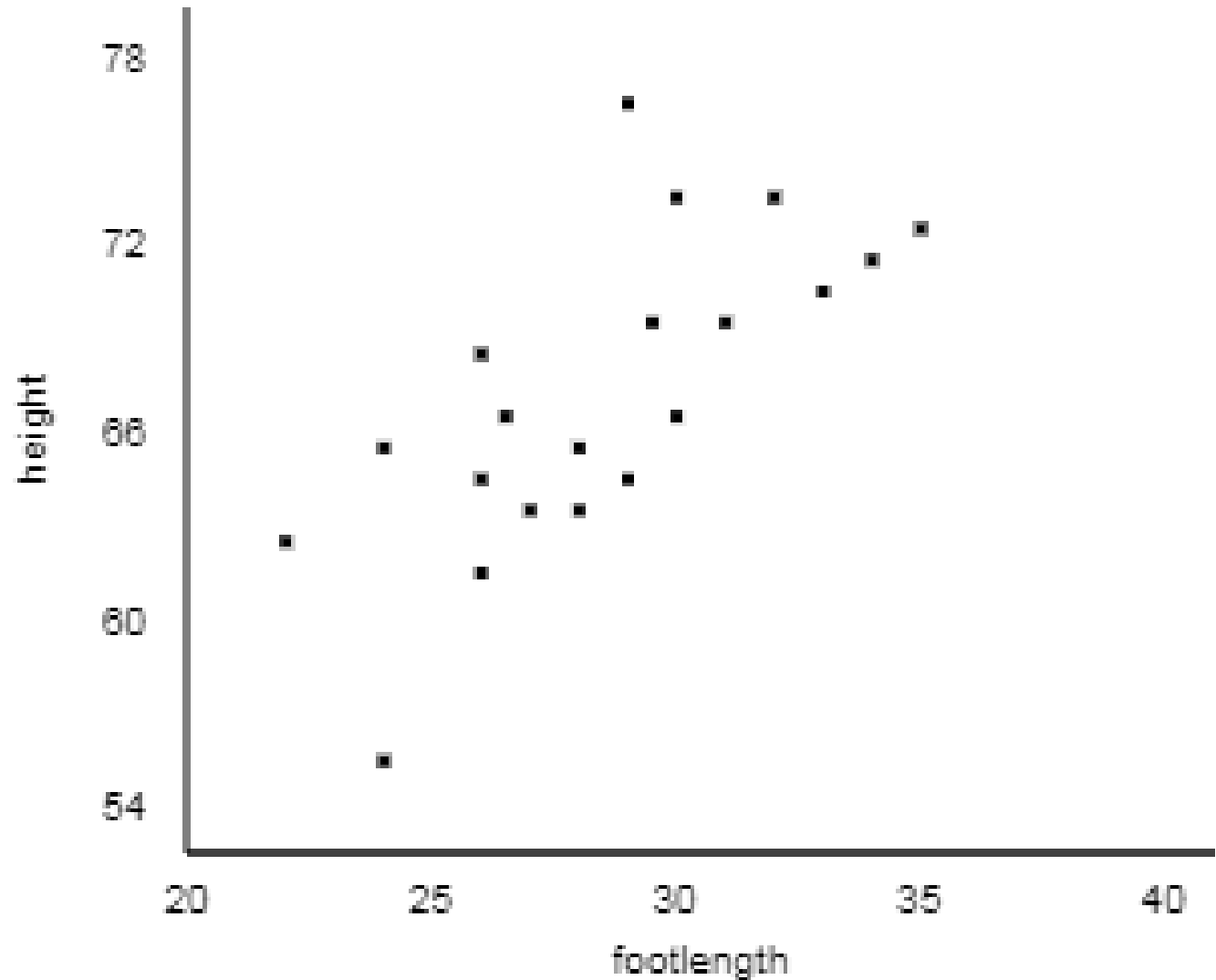
*Yes*

*Strength?*

*$r = 0.711$*

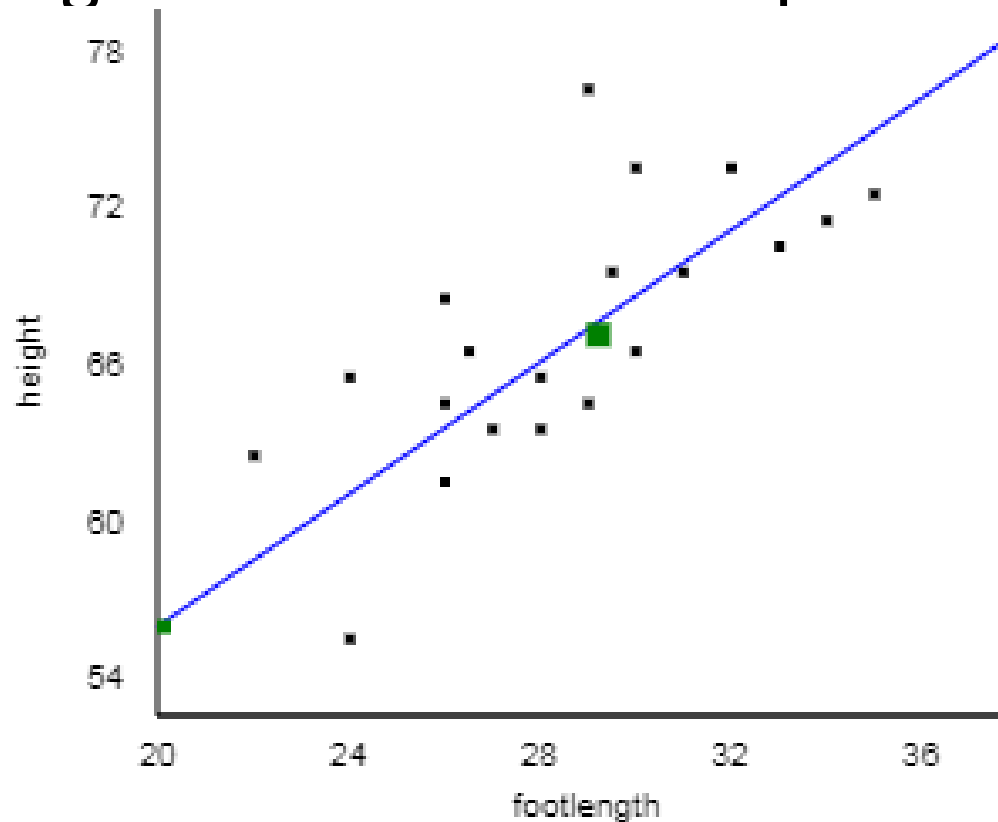


Suppose you come across a footprint that is 28 cm long. How tall do you **predict** the maker was?



# Prediction from a line

By drawing a **line** through the points, I can consistently predict the value of the response variable for a given value of the explanatory variable.

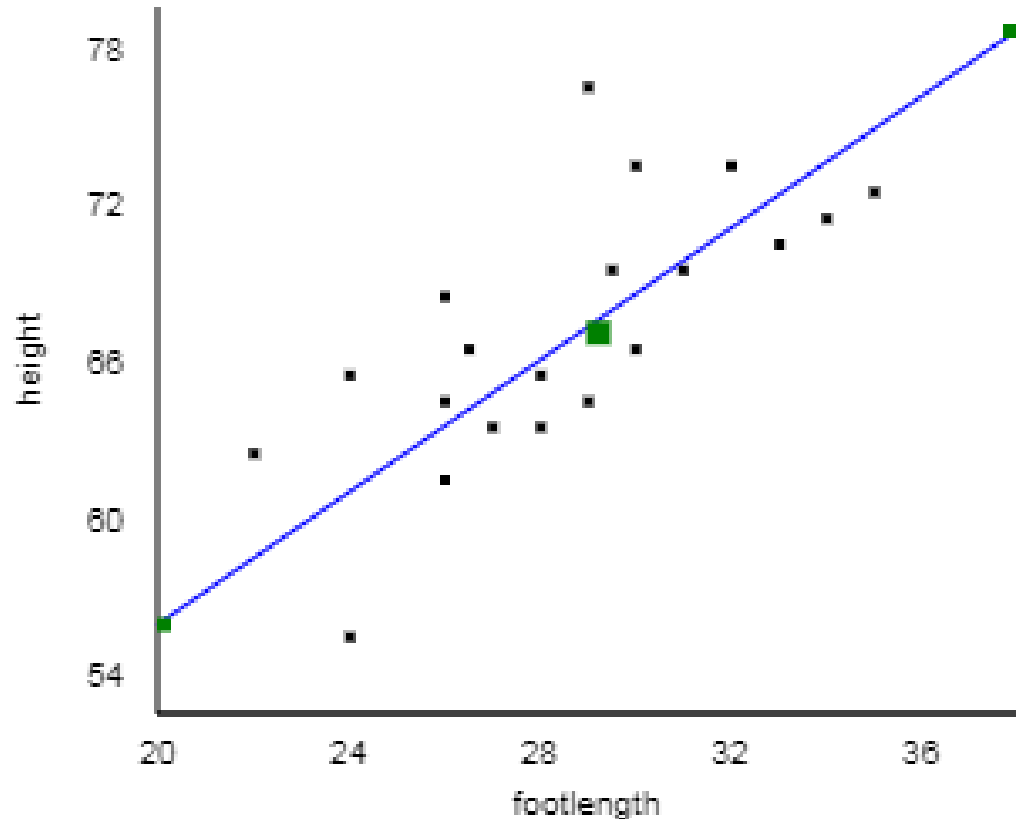




# Prediction from a line

By agreeing on *how to draw* a **line** through the points, **we** can consistently predict the value of the response variable for a given value of the explanatory variable.

Idea: Choose the line that minimizes the distances from the points to the prediction line



# In HW 9, you'll be asked to find the equation of the line you would use.

## Analyzing Two Quantitative Variables

Sample data:

(Explanatory, Response)

26.0	62
26.5	67
28.0	66
28.0	64
26.0	69
35.0	73
30.0	74
31.0	70
29.0	65
34.0	72

Use Data Revert Clear

n = 20

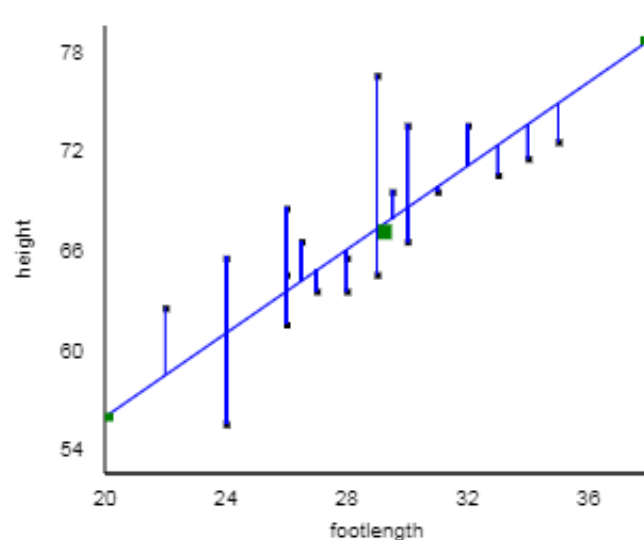
Show Movable Line:

height<sup>^</sup> = 31.18 + 1.26 x footlength

Show Residuals:

SAE=58.24

Show Squared Residuals:



Show Regression Line:

# Some terminology for choosing the “best” line

A **residual** is the difference between the *predicted* value and the *observed* value

The *sum of the absolute residuals* is denoted **SAE**

The *sum of squared residuals* is denoted **SSE**

# Choosing the “best” line

- Could choose the line that minimizes either SAE or SSE.
- Historically, people have chosen the line that minimizes SSE because it is possible to compute without computers: this line is called the **“least squares line”** or **“regression line”**

# The Least Squares Line

## Analyzing Two Quantitative Variables

Sample data:

(Explanatory, Response)	
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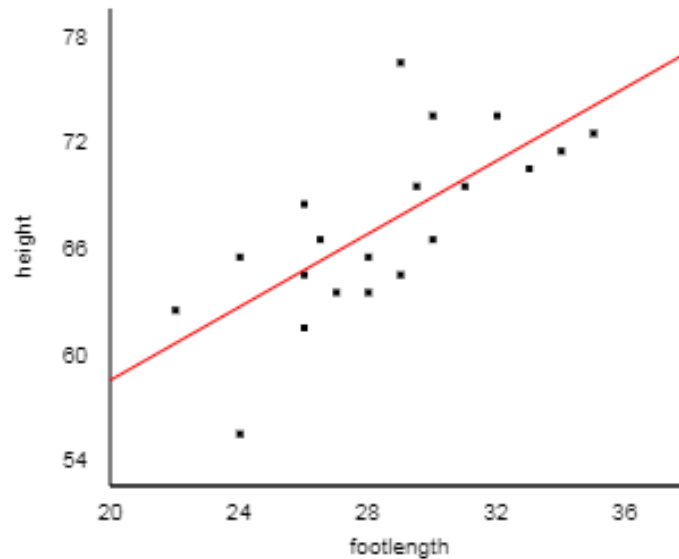
Use Data Revert Clear

n = 20

Show Movable Line:

Show Regression Line:

$$\text{height}^{\wedge} = 38.30 + 1.03 \times \text{footlength}$$



# Equation of a line

Recall the equation of a line from algebra:

$$y=mx+b$$

Example:  $y=3x+1$

What does the 3 mean?

What does the 1 mean?

# Equation of a line

Recall the equation of a line from algebra:

$$y=mx+b$$

Example:  $y=3x+1$

What does the **3** mean?

*If  $x$  increases by 1 unit then  $y$  increases by **3** units*

What does the **1** mean?

*When  $x = 0$ ,  $y = 1$ .*

# Equation of a line

Recall the equation of a line from algebra:

$$y=mx+b$$

Example:  $y=-3x+1$

What does the -3 mean?

*If x increases by 1 unit then y **decreases** by 3 units*



# Equation of a “least squares line”

$$\hat{y} = b_0 + b_1x$$

Here,

- $\hat{y}$  is the **predicted value** of the response variable when the value of the explanatory variable is  $x$
- $b_1$  is the **regression slope**,
- $b_0$  is the **regression intercept**

The “least squares line” from Inv. 5.8

$$\widehat{height} = 38.1 + 1.03footlength$$

Here,

- $\widehat{height}$  is the **predicted height** for a given footlength
- 1.03 is the **regression slope**,
- 38.1 is the **regression intercept**

Interpreting the “least squares line” from Inv. 5.8

$$\widehat{height} = 38.3 + 1.03 \text{ footlength}$$

Slope:

*If the footlength increases by 1 cm then the predicted height increases by 1.03 inches.*

Intercept:

*The predicted height is 38.3 inches when the footlength is 0.*

Using the “least squares line” from Inv. 5.8

$$\widehat{height} = 38.3 + 1.03 \text{ footlength}$$

Predict the height of someone whose footlength is 28 cm:

$$\widehat{height} = 38.3 + 1.03(28) = 67.14$$

# Coefficient of determination, $R^2$

- Provides a measure of how useful the least squares line is

$R^2$  = percent error of the SSE of prediction line  $\bar{y}$   
and the SSE of the least squares line

where SSE is the sum of the squared residuals

# Interpretation of $R^2$ in Inv. 5.8

$R^2$  = percent of variability of the **response variable  $y$**  that is ***explained*** by the least squares line with the **explanatory variable  $x$**

$R^2 = 50.6\%$  so...

The least square line with **footlength** ***explains*** 50.6% of the variability in **height**.

# Interpretation of $R^2$

$R^2$  = percent of variability of the response variable  $y$  that is explained by the least squares line with the explanatory variable  $x$

## Notes:

- $R^2$  is always between 0% and 100%
- $R^2 = r^2$ , where  $r$  is the correlation coefficient.