

# Simple Linear Regression

## Introduction

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# Topics

- Use simple linear regression to describe the relationship between a quantitative predictor and quantitative response variable.

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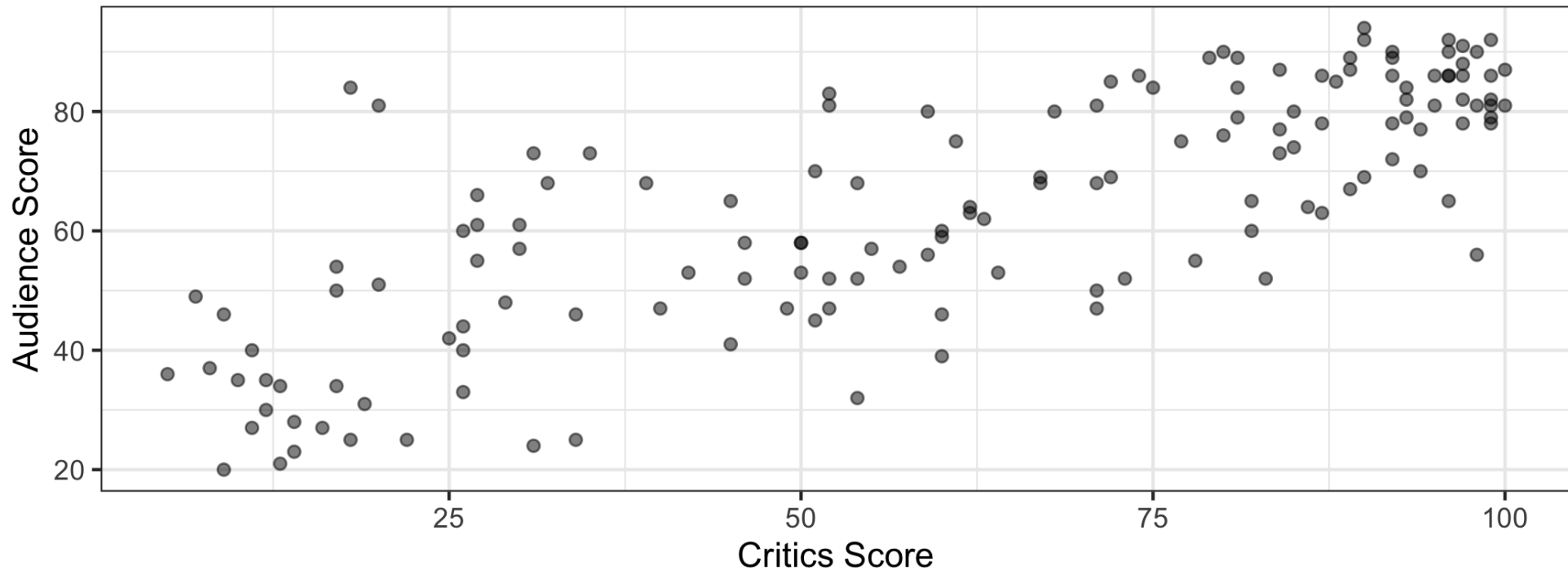
- Use simple linear regression to describe the relationship between a quantitative predictor and quantitative response variable.
- Estimate the slope and intercept of the regression line using the least squares method.

# Topics

- Use simple linear regression to describe the relationship between a quantitative predictor and quantitative response variable.
- Estimate the slope and intercept of the regression line using the least squares method.
- Interpret the slope and intercept of the regression line.

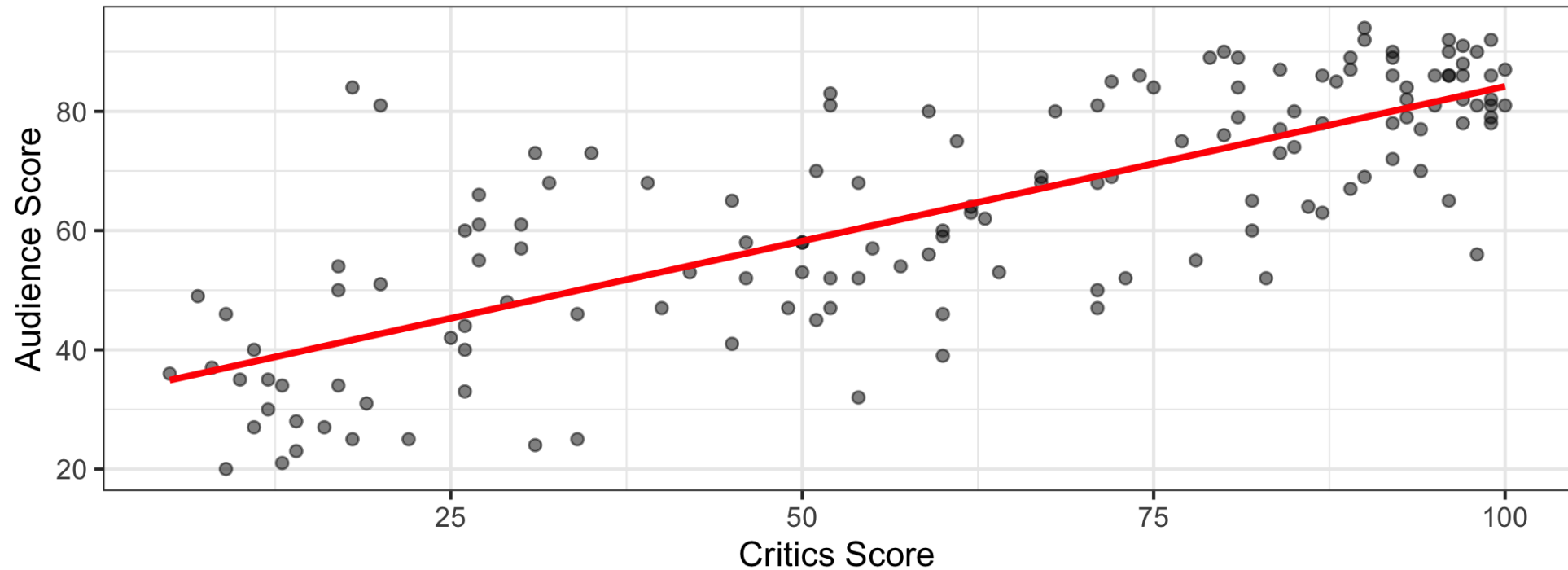
# Movie ratings data

The data set contains the "Tomatometer" score (**critics**) and audience score (**audience**) for 146 movies rated on rottentomatoes.com.



# Movie ratings data

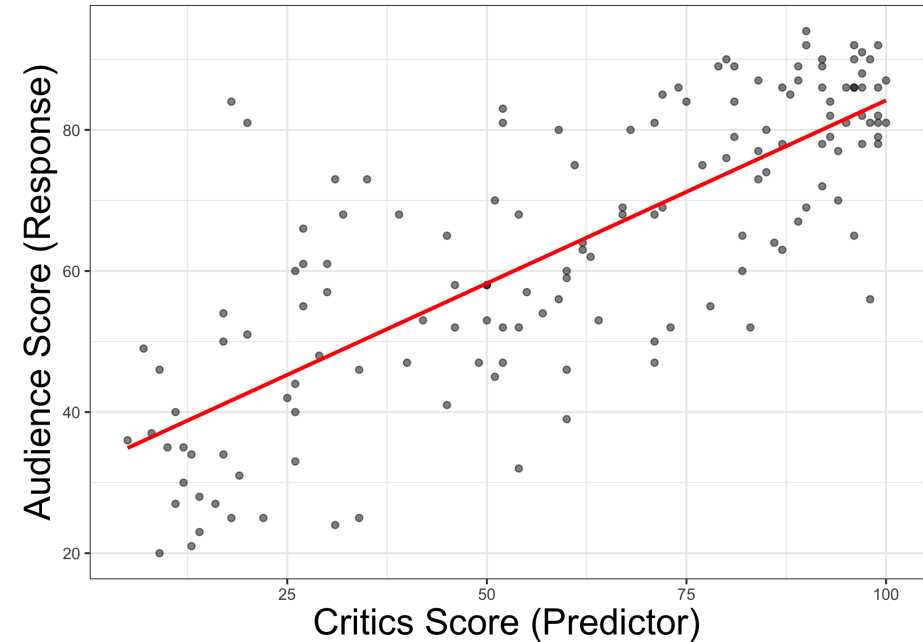
We want to fit a line to describe the relationship between the critics score and audience score.



# Terminology

The **response,  $Y$** , is the variable describing the outcome of interest.

The **predictor,  $X$** , is the variable we use to help understand the variability in the response.





# Regression model

A regression model is a function that describes the relationship between the response,  $Y$ , and the predictor,  $X$ .

$$Y = \mathbf{Model} + \text{Error}$$

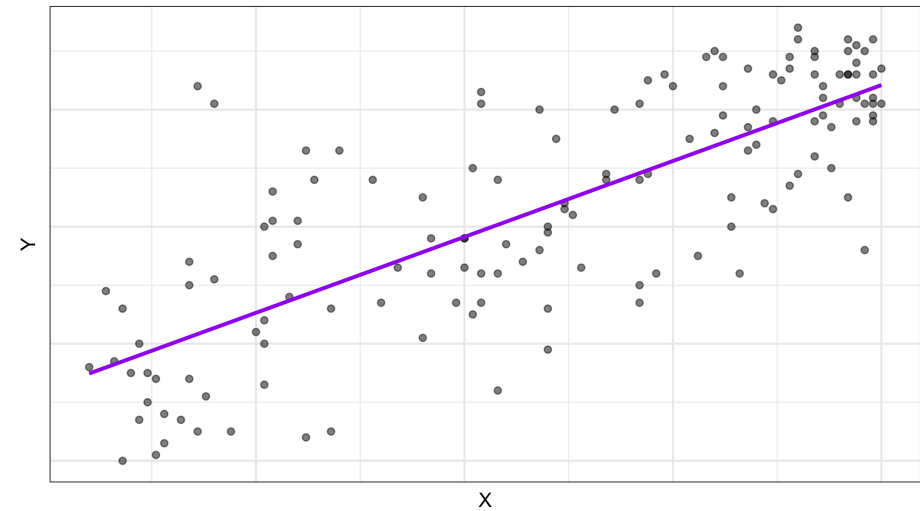
$$= \mathbf{f(X)} + \epsilon$$

$$= \mu_{Y|X} + \epsilon$$

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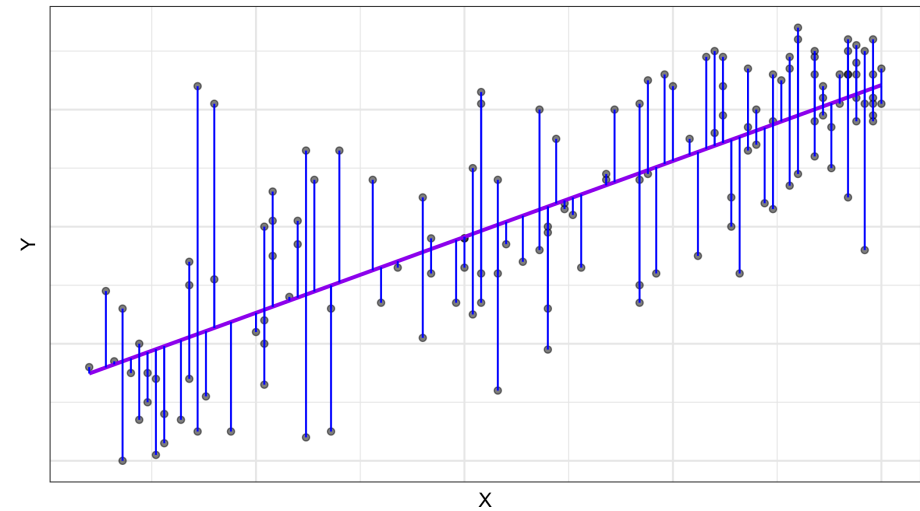
$$= \mu_{Y|X} + \epsilon$$



$$Y = \text{Model} + \text{Error}$$

$$= \mathbf{f(X)} + \epsilon$$

$$= \mu_{Y|X} + \epsilon$$



# Simple linear regression

When we have a quantitative response,  $Y$ , and a single quantitative predictor,  $X$ , we can use a **simple linear regression** model to describe the relationship between  $Y$  and  $X$ .

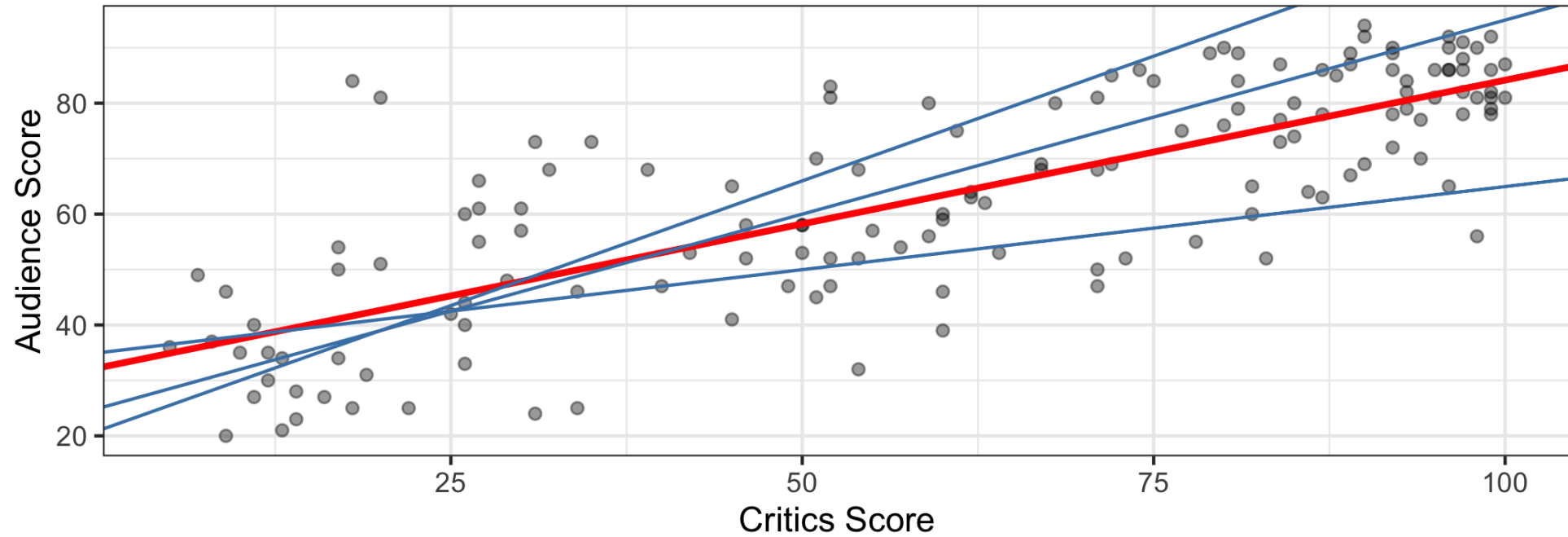
$$Y = \beta_0 + \beta_1 X + \epsilon$$

$\beta_1$  : Slope       $\beta_0$  : Intercept

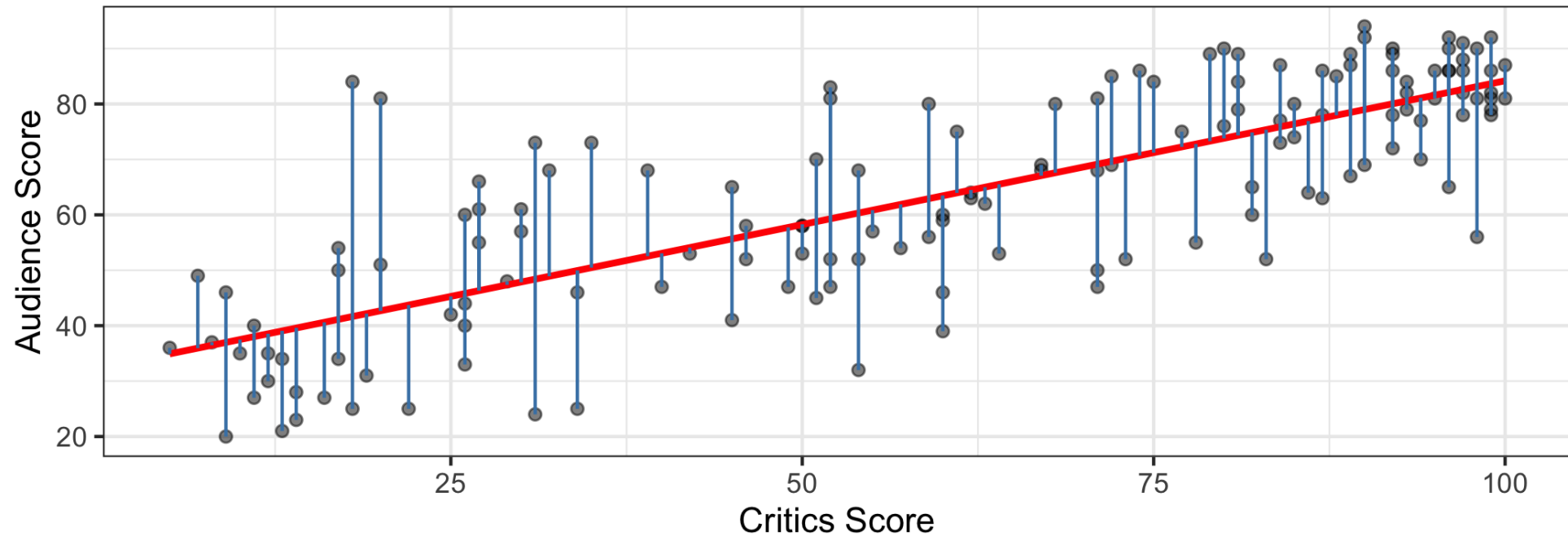
$$\hat{Y} = \hat{\beta}_0 + \hat{\beta}_1 X$$

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How do we choose values for  $\hat{\beta}_1$  and  $\hat{\beta}_0$ ?



# Residuals



$$\text{residual} = \text{observed} - \text{predicted} = y - \hat{y}$$



# Least squares line

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- The residual for the  $i^{th}$  observation is

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- The **sum of squared residuals** is

$$e_1^2 + e_2^2 + \cdots + e_n^2$$

- The **least squares line** is the one that minimizes the sum of squared residuals

# Estimating the slope

$$\hat{\beta}_1 = r \frac{s_Y}{s_X}$$

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$$s_Y = 20.024$$

$$r = 0.781$$

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$$r = 0.781$$

$$\begin{aligned}\hat{\beta}_1 &= 0.781 \times \frac{20.024}{30.169} \\ &= \mathbf{0.518}\end{aligned}$$

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



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$$\bar{x} = 60.850$$

$$\bar{y} = 63.877$$

$$\hat{\beta}_1 = 0.518$$

# Estimating the intercept

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

$$\bar{x} = 60.850$$

$$\bar{y} = 63.877$$

$$\hat{\beta}_1 = 0.518$$

$$\begin{aligned}\hat{\beta}_0 &= 63.877 - 0.518 \times 60.850 \\ &= \mathbf{32.296}\end{aligned}$$

# Interpreting slope & intercept

$$\hat{\text{audience}} = 32.296 + 0.518 \times \text{critics}$$

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# Interpreting slope & intercept

$$\hat{\text{audience}} = 32.296 + 0.518 \times \text{critics}$$

**Slope:** For every one point increase in the critics score, we expect the audience score to increase by 0.518 points, on average.

**Intercept:** If the critics score is 0 points, we expect the audience score to be 32.296 points.

# Does it make sense to interpret the intercept?

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✓ Interpret the intercept if

- the predictor can feasibly take values equal to or near zero.
- there are values near zero in the data.

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- the predictor can feasibly take values equal to or near zero.
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⬮ Otherwise, don't interpret the intercept!



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- Used simple linear regression to describe the relationship between a quantitative predictor and quantitative response variable.
- Used the least squares method to estimate the slope and intercept.
- We interpreted the slope and intercept.
  - **Slope**: For every one unit increase in  $x$ , we expect  $y$  to change by  $\hat{\beta}_1$  units, on average.
  - **Intercept**: If  $x$  is 0, then we expect  $y$  to be  $\hat{\beta}_0$  units.