## UNIT2: TWO VARIABLE DATA

## WHAT IS OUR GOAL FOR UNIT 2?

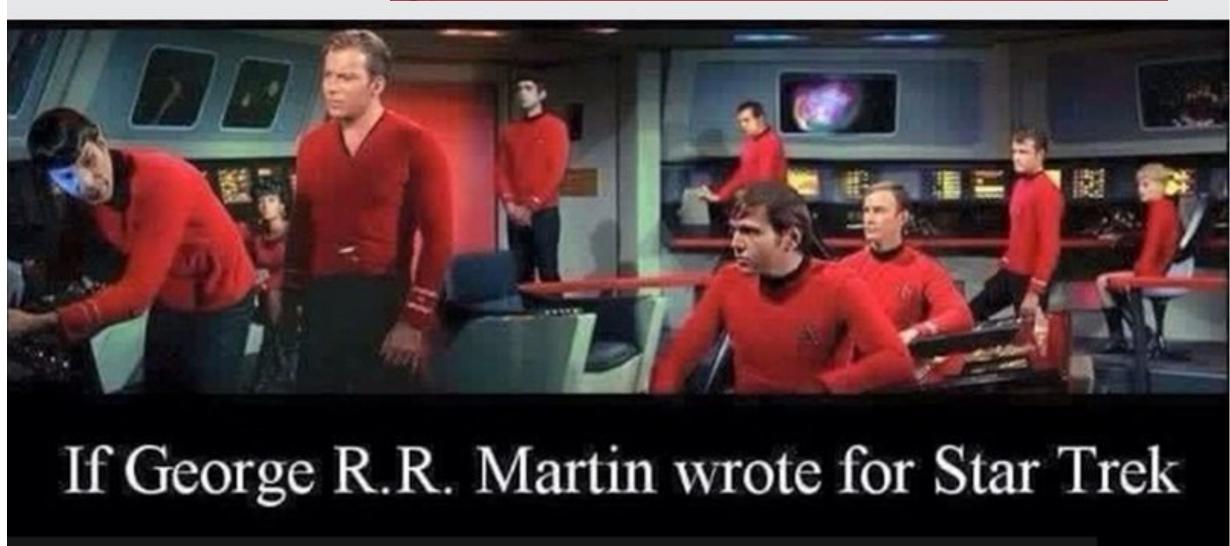
- Representing Relationships Between Bivariate
   Categorical Data
- Representing Relationships Between Bivariate
   Quantitative Data

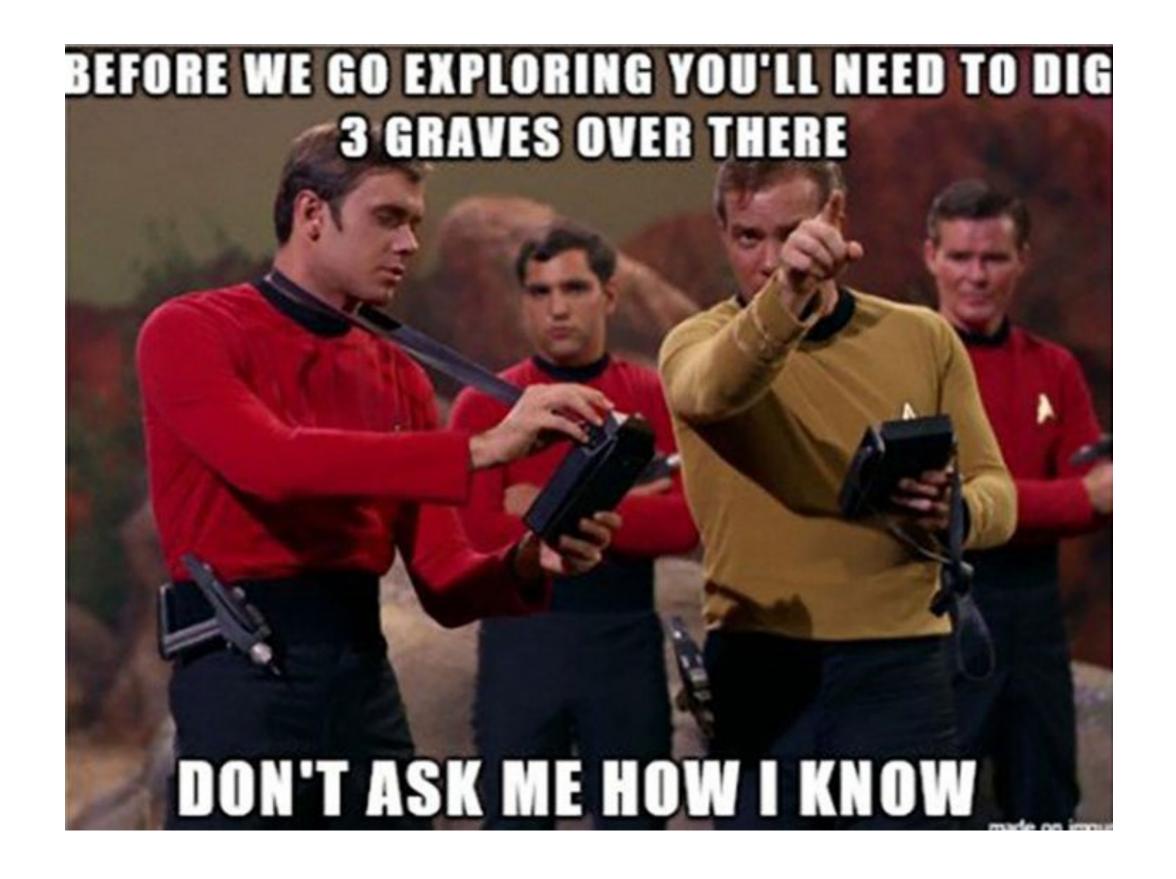
- Is there a relationship between two Categorical Variables?
- We will represent relationships using tables (same as treat example before), Graphs, and Statistics (numbers)

- Our Example: X: Shirt Colour 🔴 🔵 , Y: Status 🐹 😃



Matthew Barsalou published an article in Significance that studies this from a statistical perspective





- Our Example: X: Shirt Colour 🔴 🔵 , Y: Status 🐹 😃

Crew Member	Area	Shirt Color	Status
Isaiah	Operations, Engineering and Security	Red	DEAD 🐹
Atley	Command And Helm	Gold	DEAD 🐹
Johnnie	Science and Medical	Blue	Alive 😃
Datacat is composed of 130 crowmates			

Dataset is composed of 430 crewmates

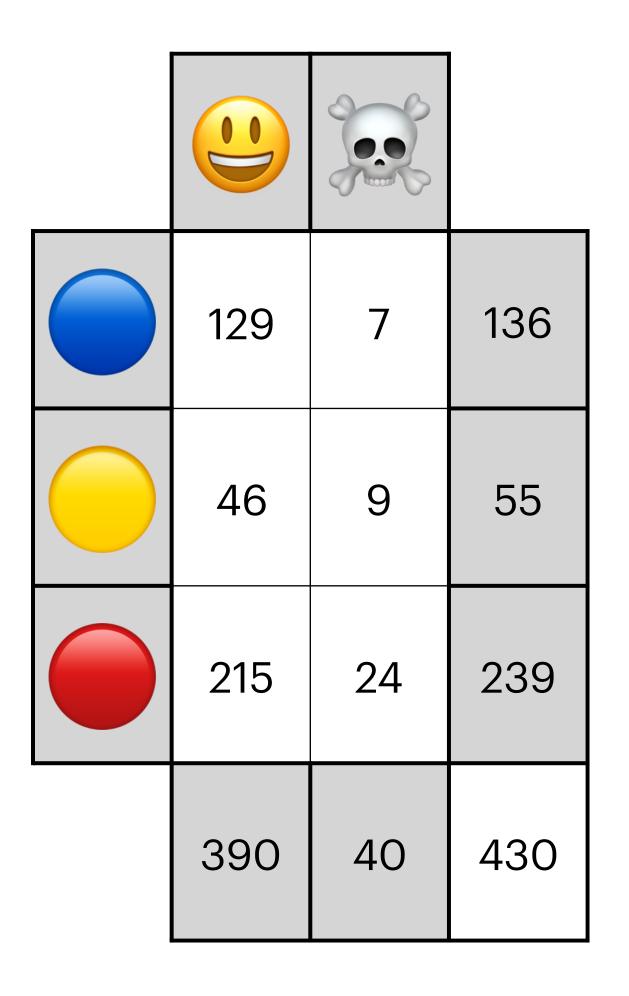
Enterprise NCC 1701 casualties from episodes aired between September 8, 1966 and June 03, 1969 based on casualty figures from Memory Alpha.

- First we tabulate data into a **contingency table** (also known as a two way table)

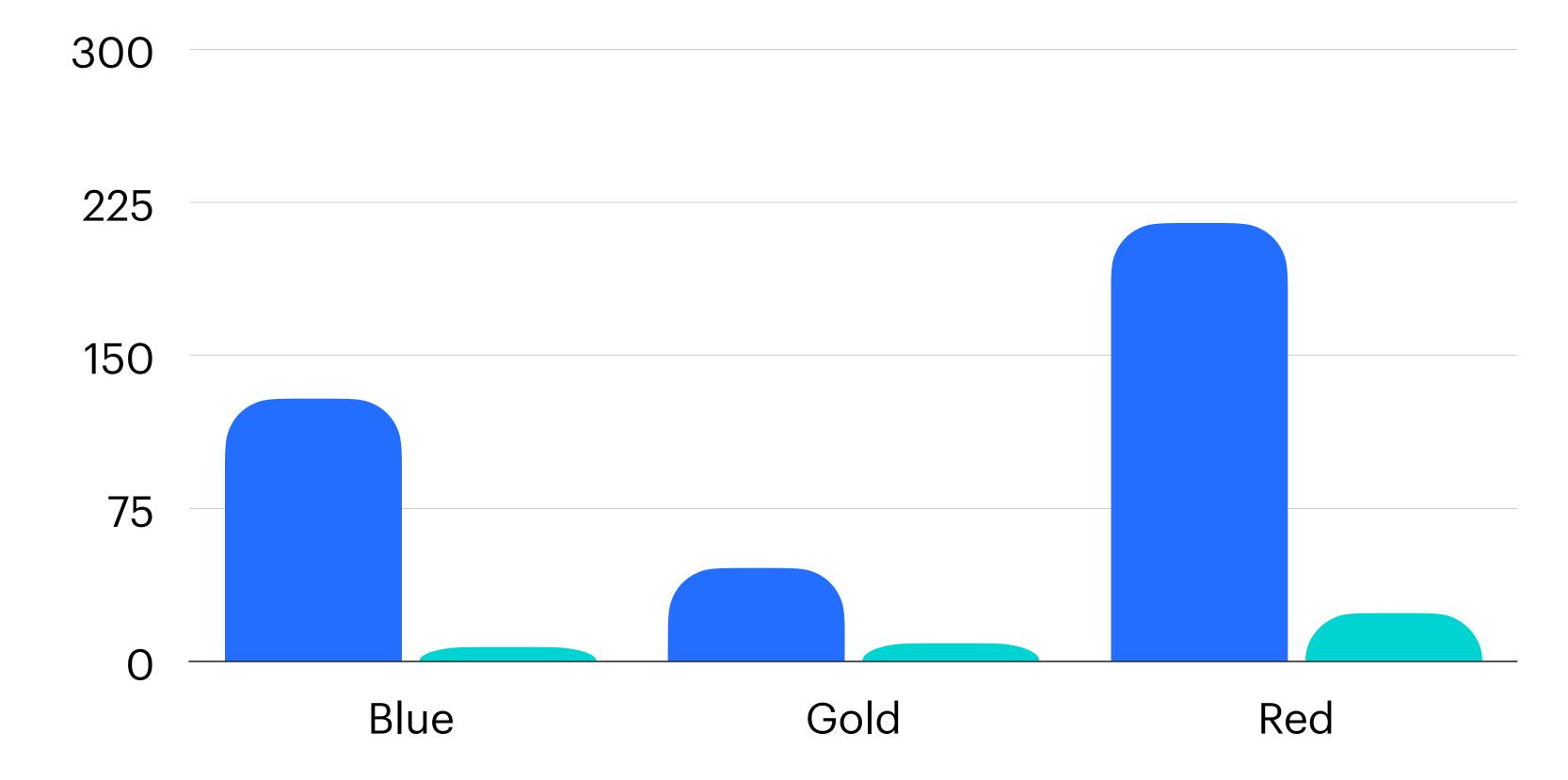
129	7	136
46	9	55
215	24	239
390	40	430

- Marginal Distribution
- Joint Distribution

- First we tabulate data into a **contingency table** (also known as a two way table)



It's hard to notice association when using frequencies



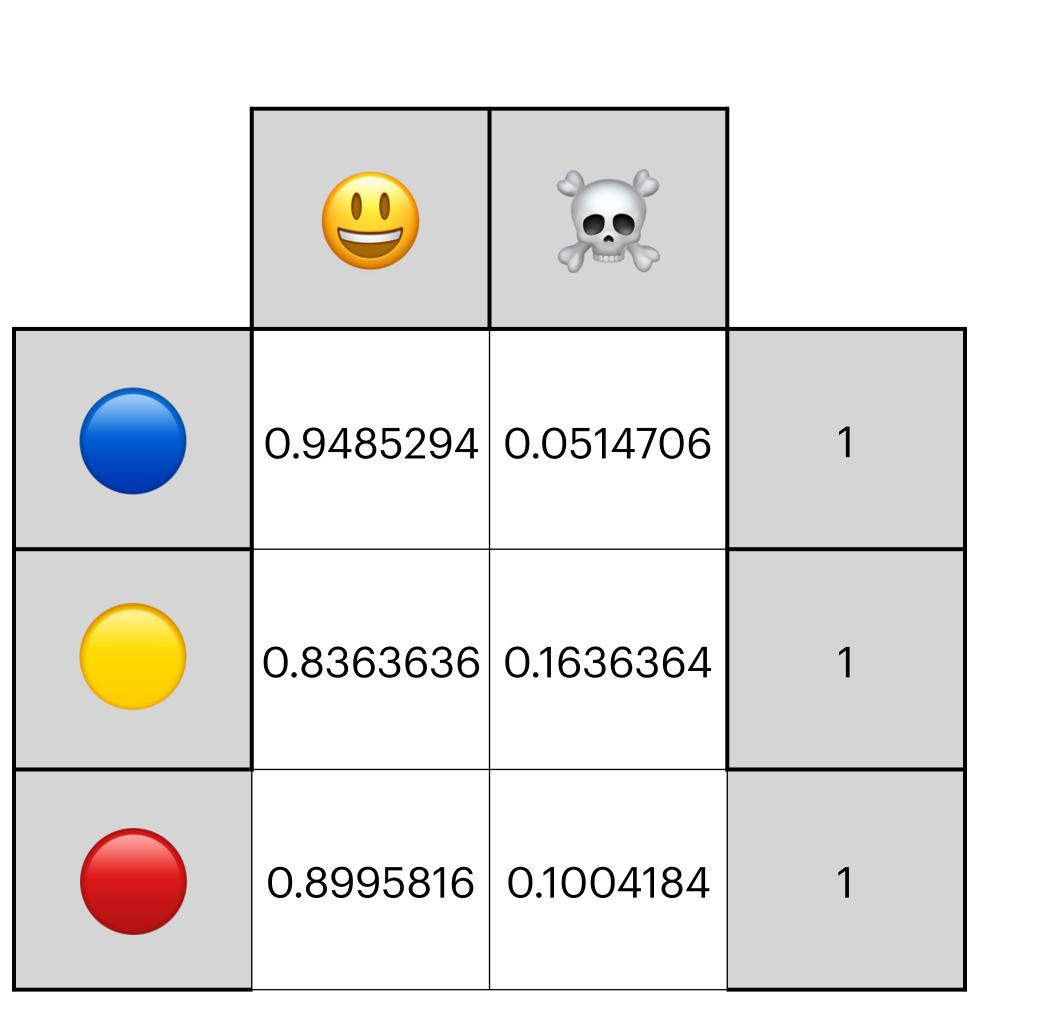
#### **Conditional Probability**

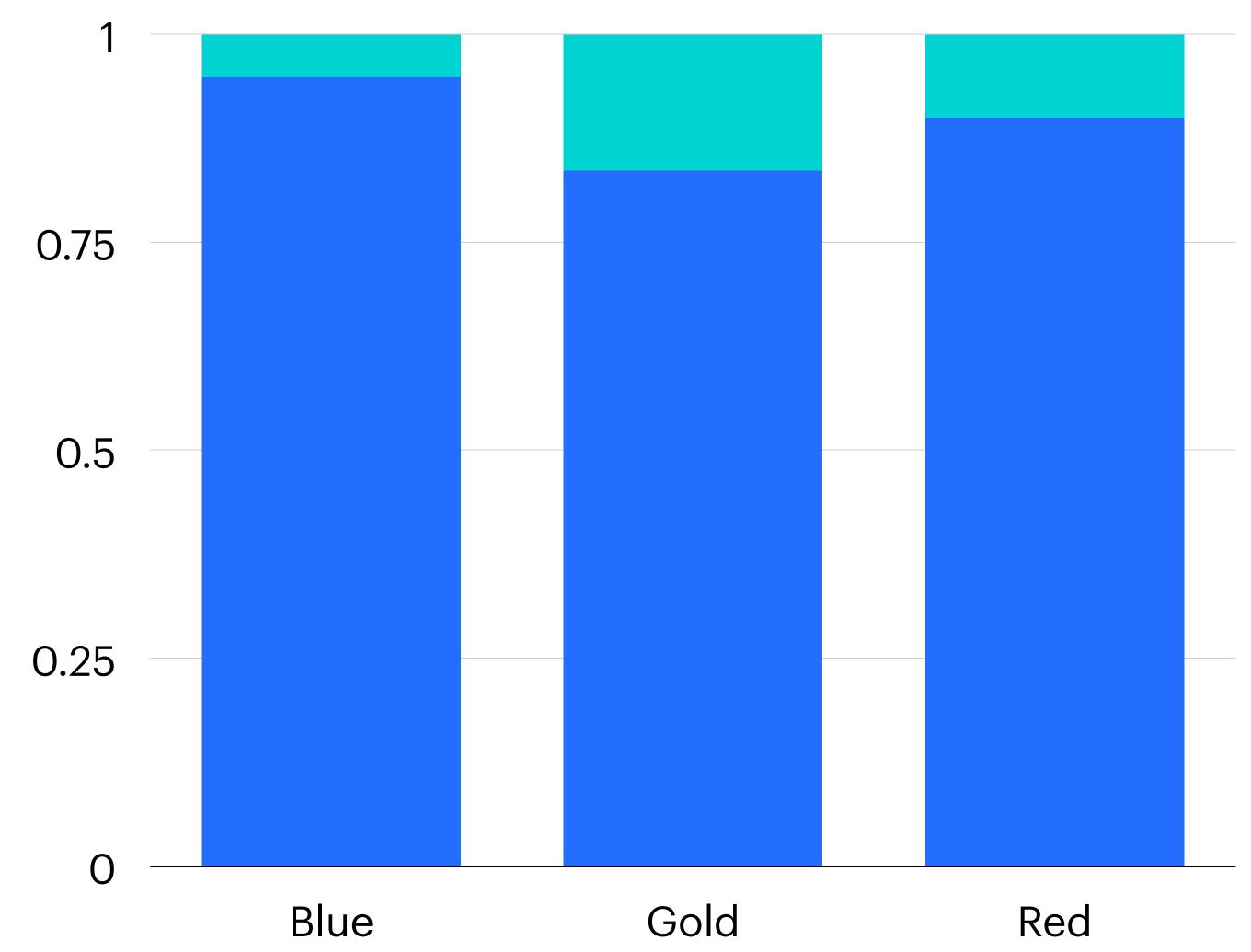
129	7	136
46	9	55
215	24	239
390	40	430

#### Questions

- 1. What is the probability of dying, given you are a Red Shirt?
- 2. What is the percentage of crew members that have red shirts and died?
- 3. What is the percentage of blue shirts who survived?
- 4. What is the probability of dying Given you are a Gold Shirt?

- Next we may find conditional relative frequencies





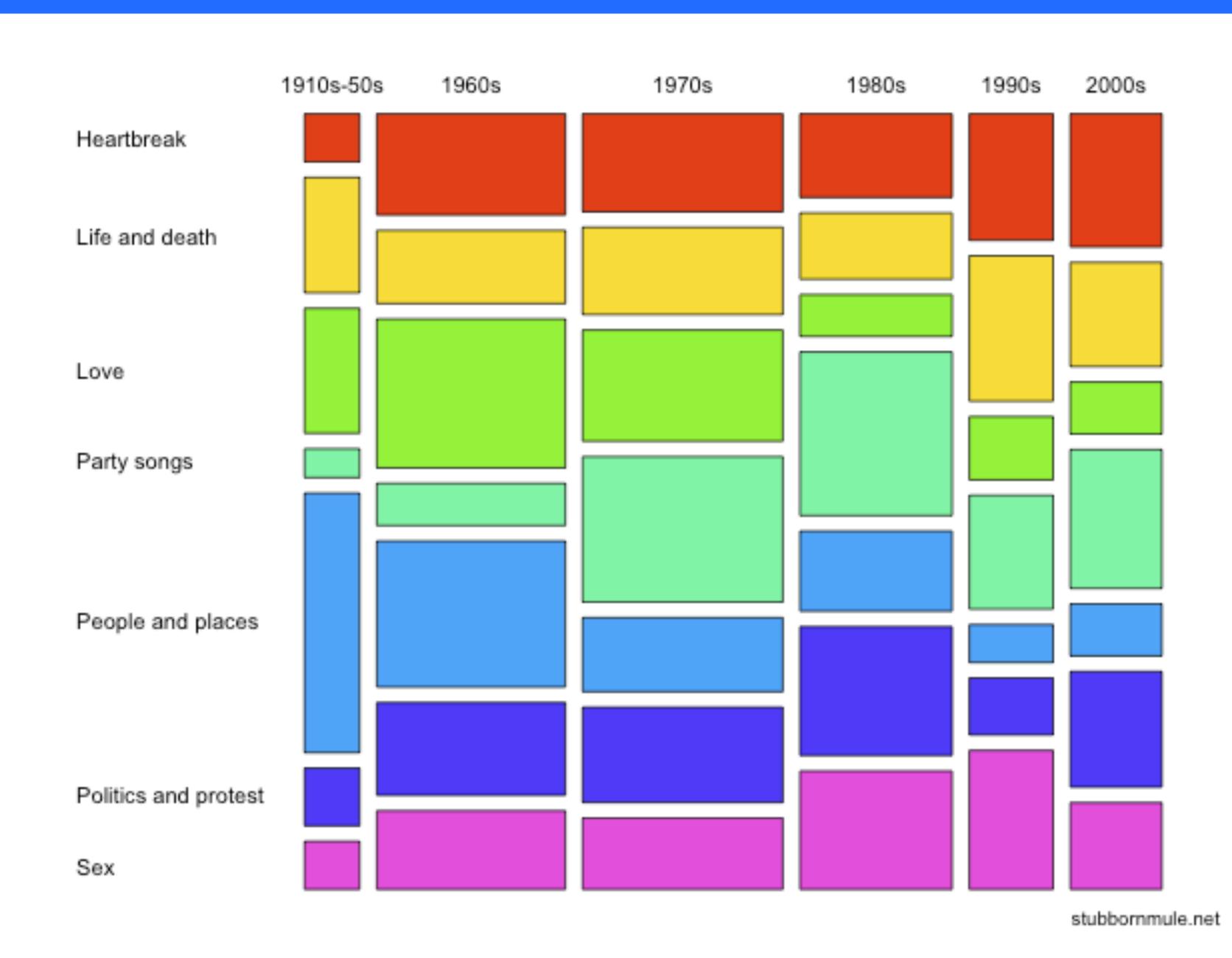
# Distribution of Conditional Relative Frequencies

0.9485294	0.0514706	1
0.8363636	0.1636364	1
0.8995816	0.1004184	1

If shirt colour is **Independent** of Status, then the probability of dying should be the same regardless of shirt colour.

**Chi-Square Tests** (For Later)

- Another type of graph used is
   Mosaic Plots.
   Widths describe how many observations fall in each category.
- Mosaic plot showing cross-sectional distribution through time of different musical themes in the Guardian's list of "1000 songs to hear before you die"



## Examples:

- Question 1, Page 107
- Question 2, Page 108
- Question 3, Page 112

Homework: Read Pages 97-104 Barron's, Quiz 6, Quiz 7

- We represent relationships between two numeric variables (sample data) using a **scatter plot**.
- When describing a relationship there are several things we must consider
  - Form
  - Direction
  - Strength

#### **EXAMPLE**

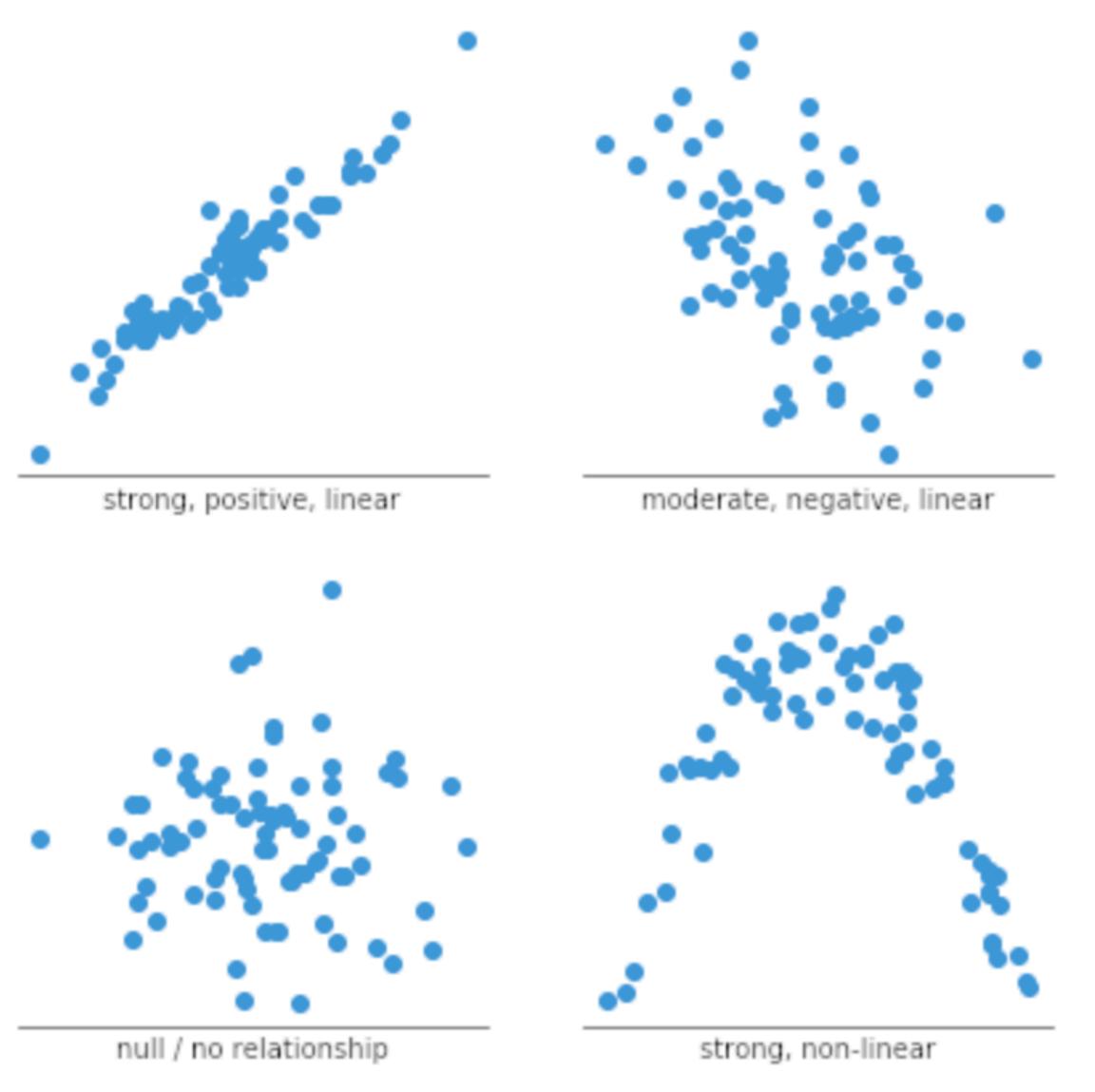
Is there a relationship between the amount of sugar (in grams) and the number of calories in movie-theatre candy? Here are the data from a sample of 12 types of candy.

Name	Sugar (g)	Calories
Butterfinger Minis	45	450
Junior Mints	107	570
M&M'S	62	480
Milk Duds	44	370
Peanut M&M'S	79	790
Raisinets	60	420
Reese's Pieces	61	580
Skittles	87	450
Sour Patch Kids	92	490
SweeTarts	136	680
Twizzlers	59	460
Whoppers	48	350

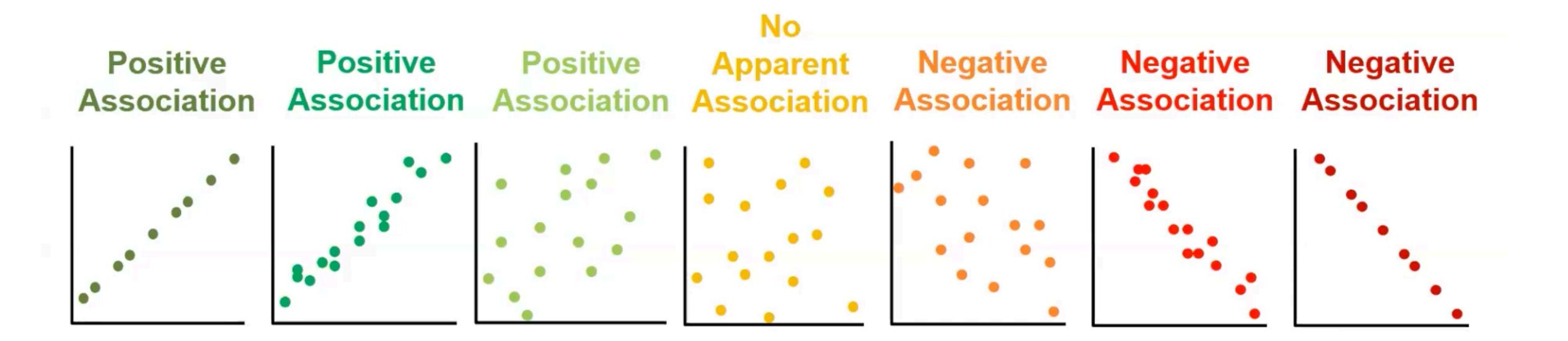
Using your TI-84, plot the data.

How Would You Describe the Relationship?

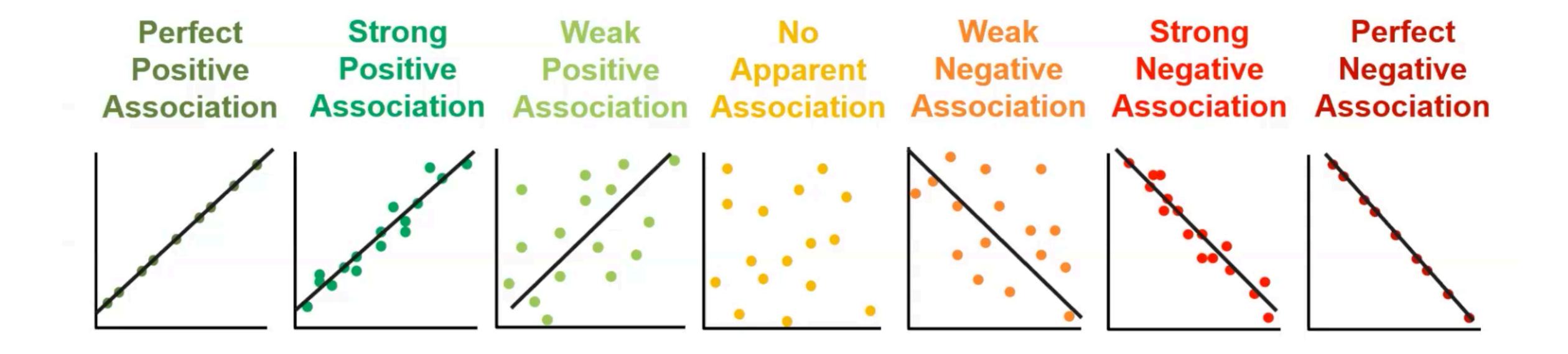
## Form of relationship



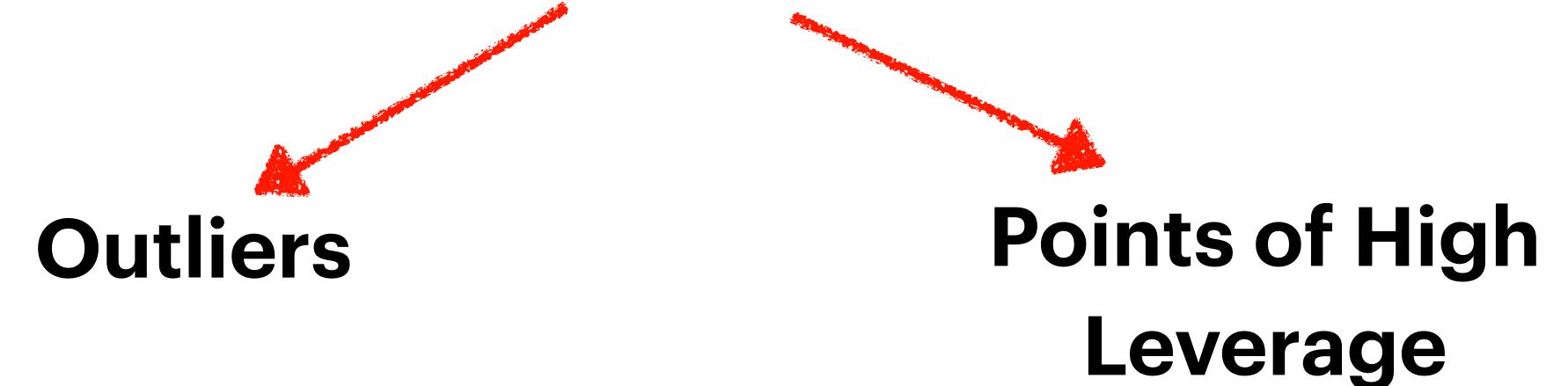
## Direction of relationship



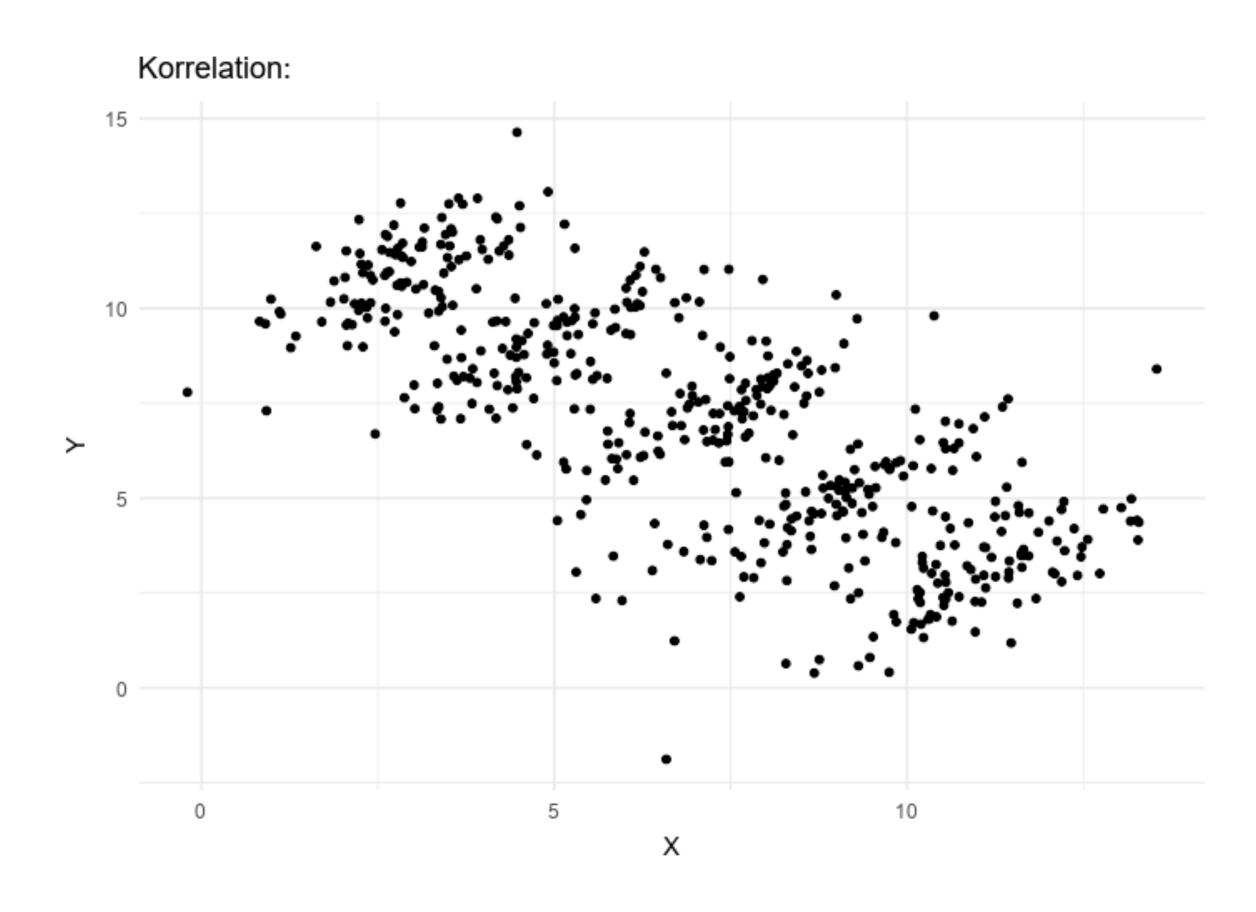
## Strength of relationship



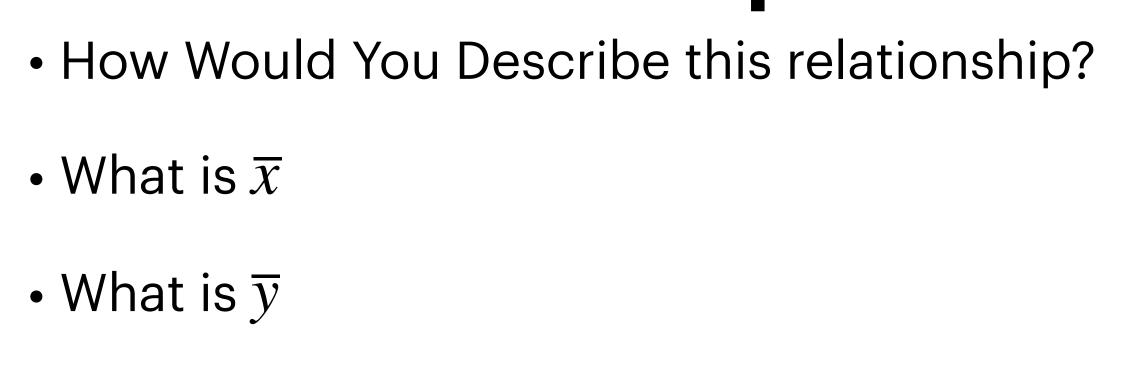
Influential Points of relationship

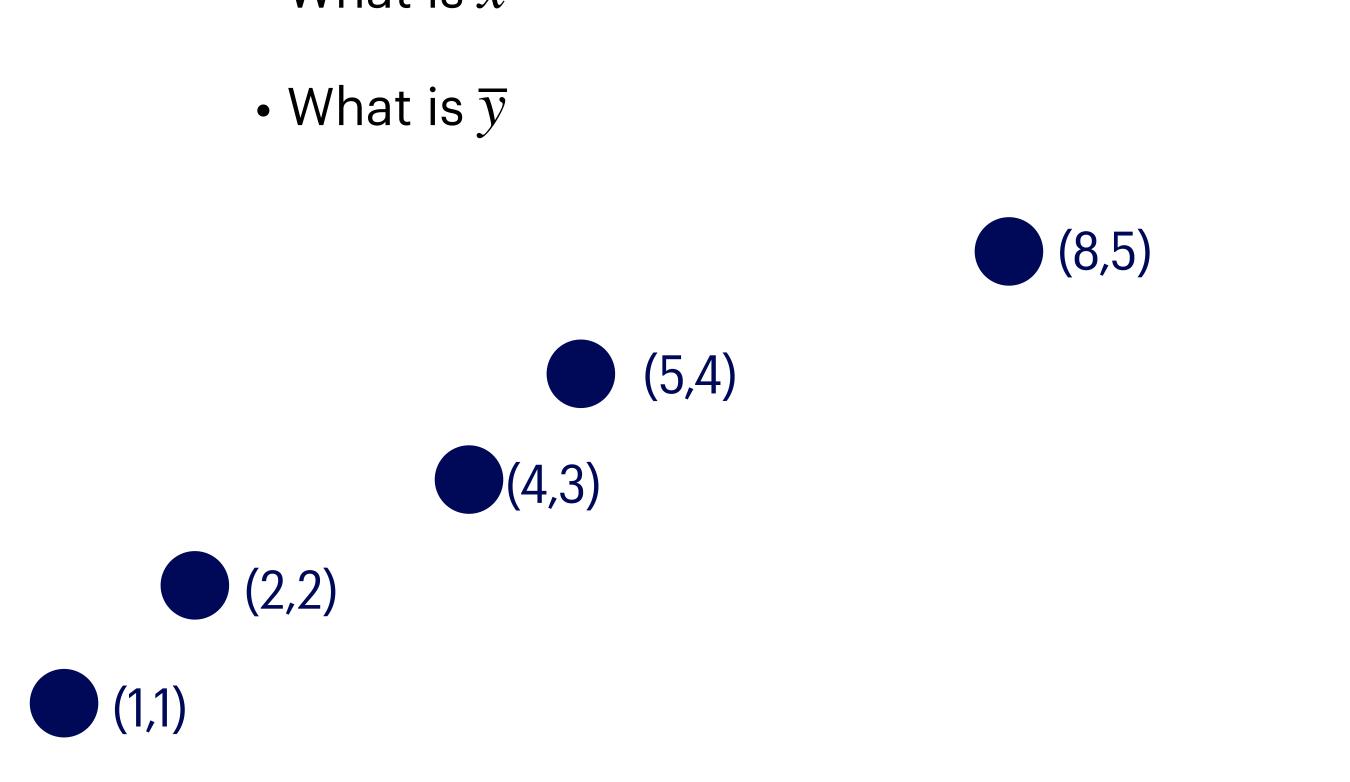


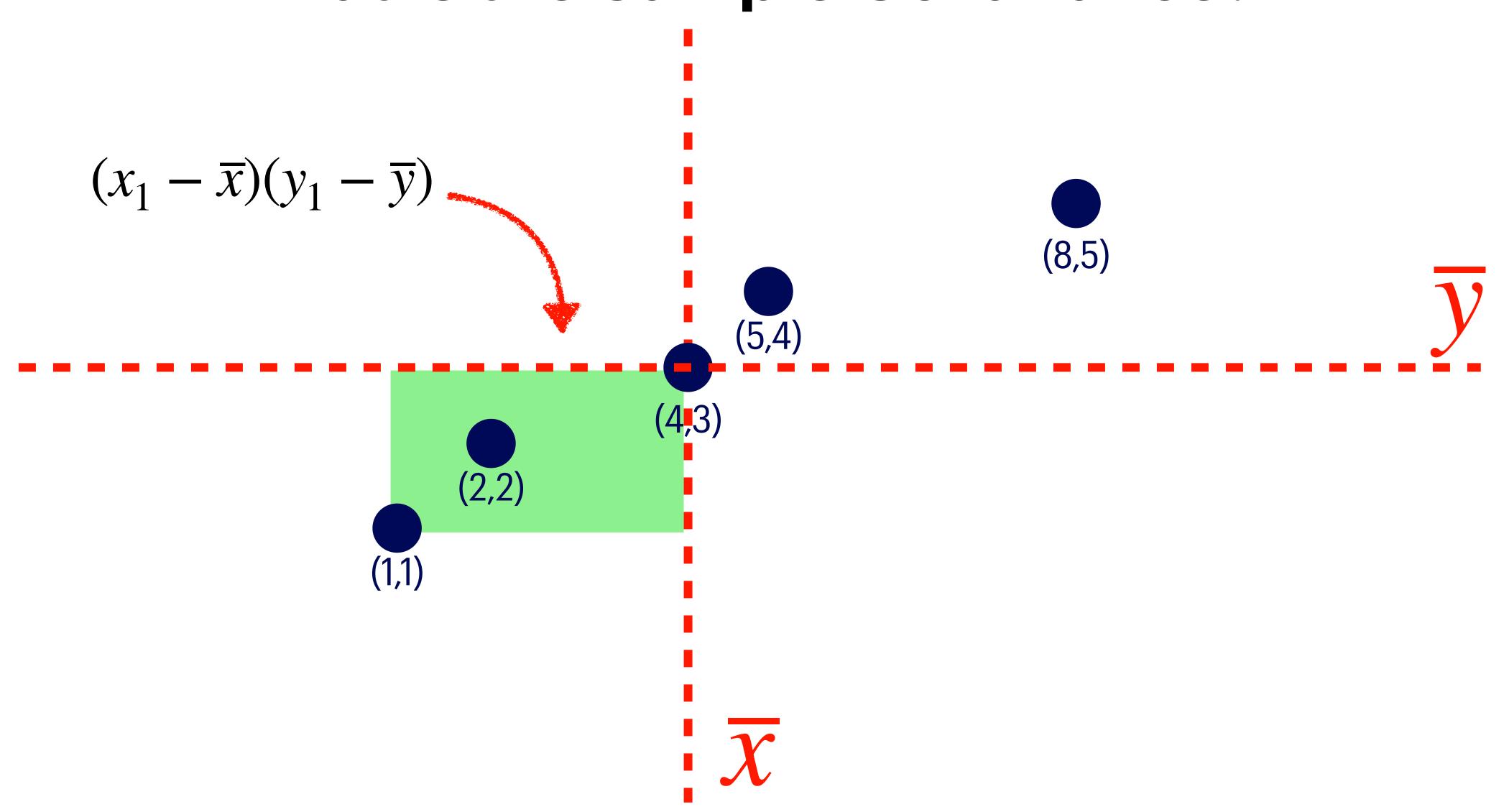
- Simpson's Paradox: There is an association within groups of data but the trend disappears, or reverses when groups are combined.
- **Example:** Q7 P.111

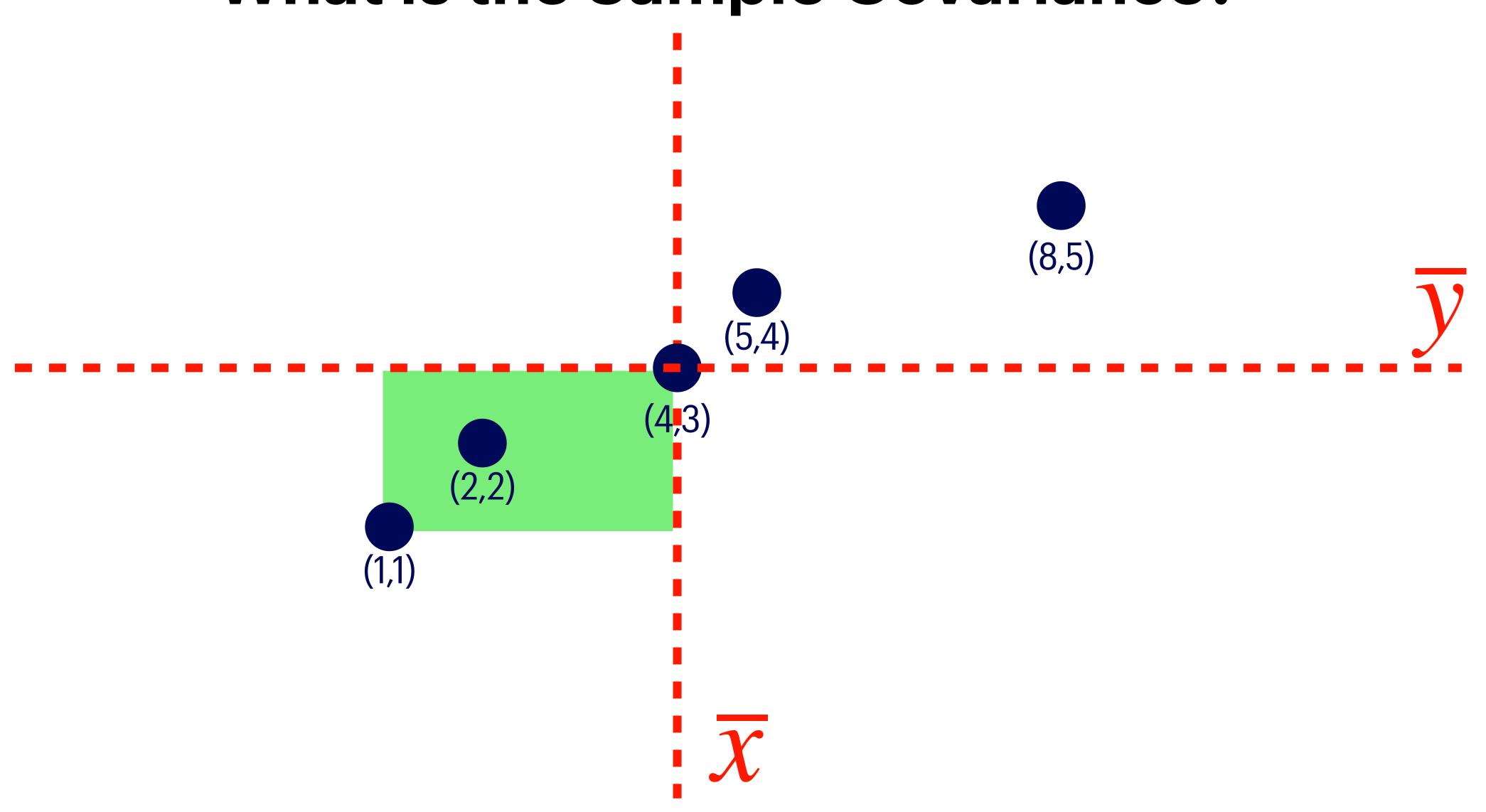


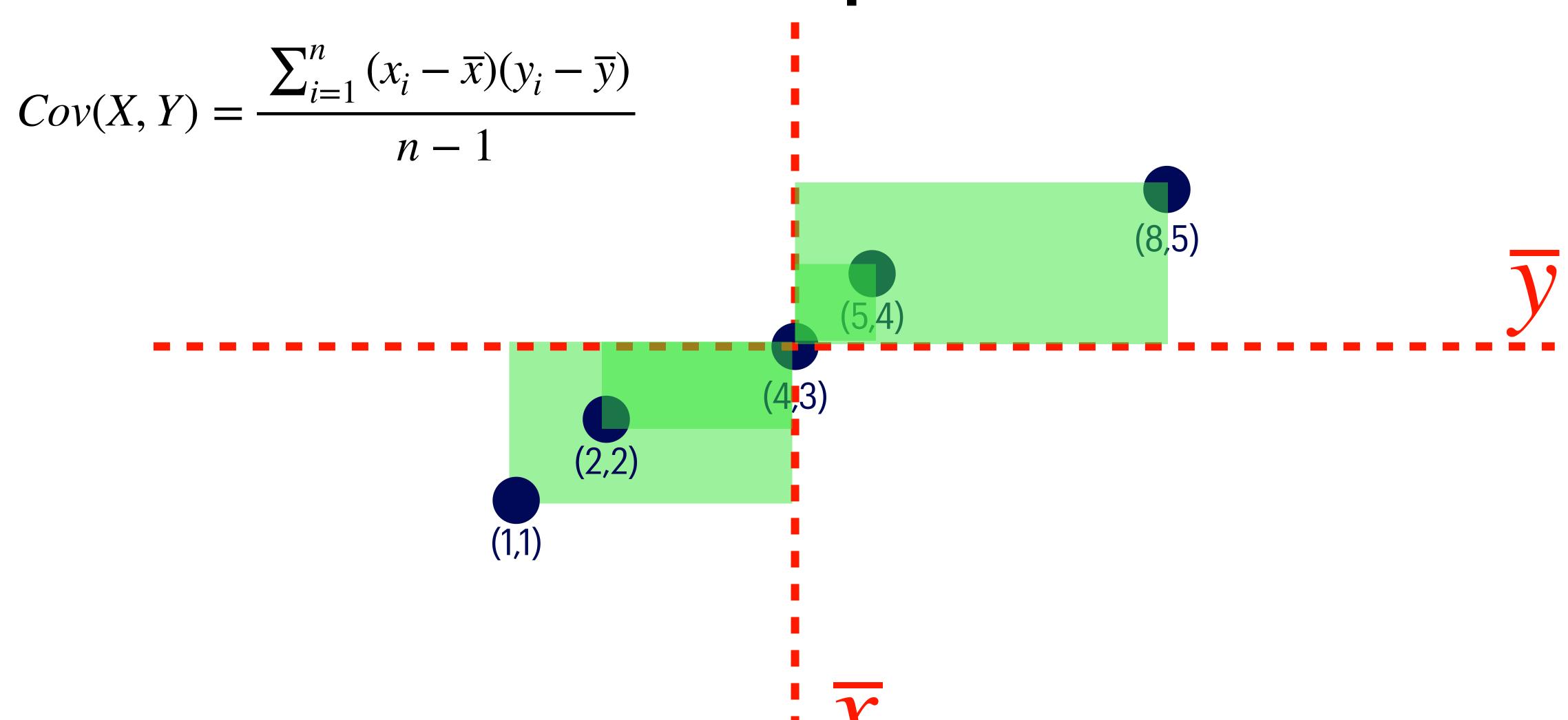
$$Cov(X, Y) = \frac{\sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})}{n-1}$$

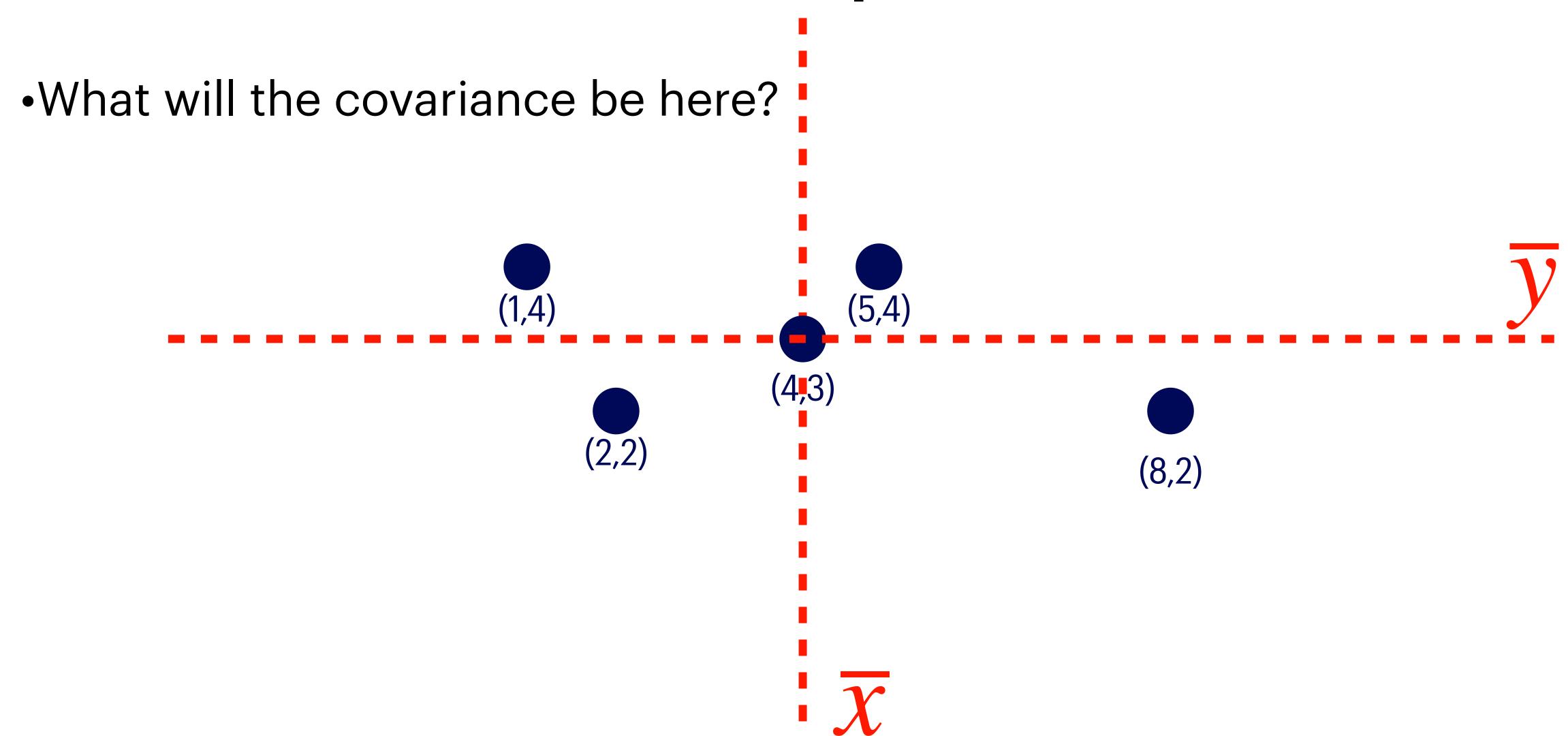


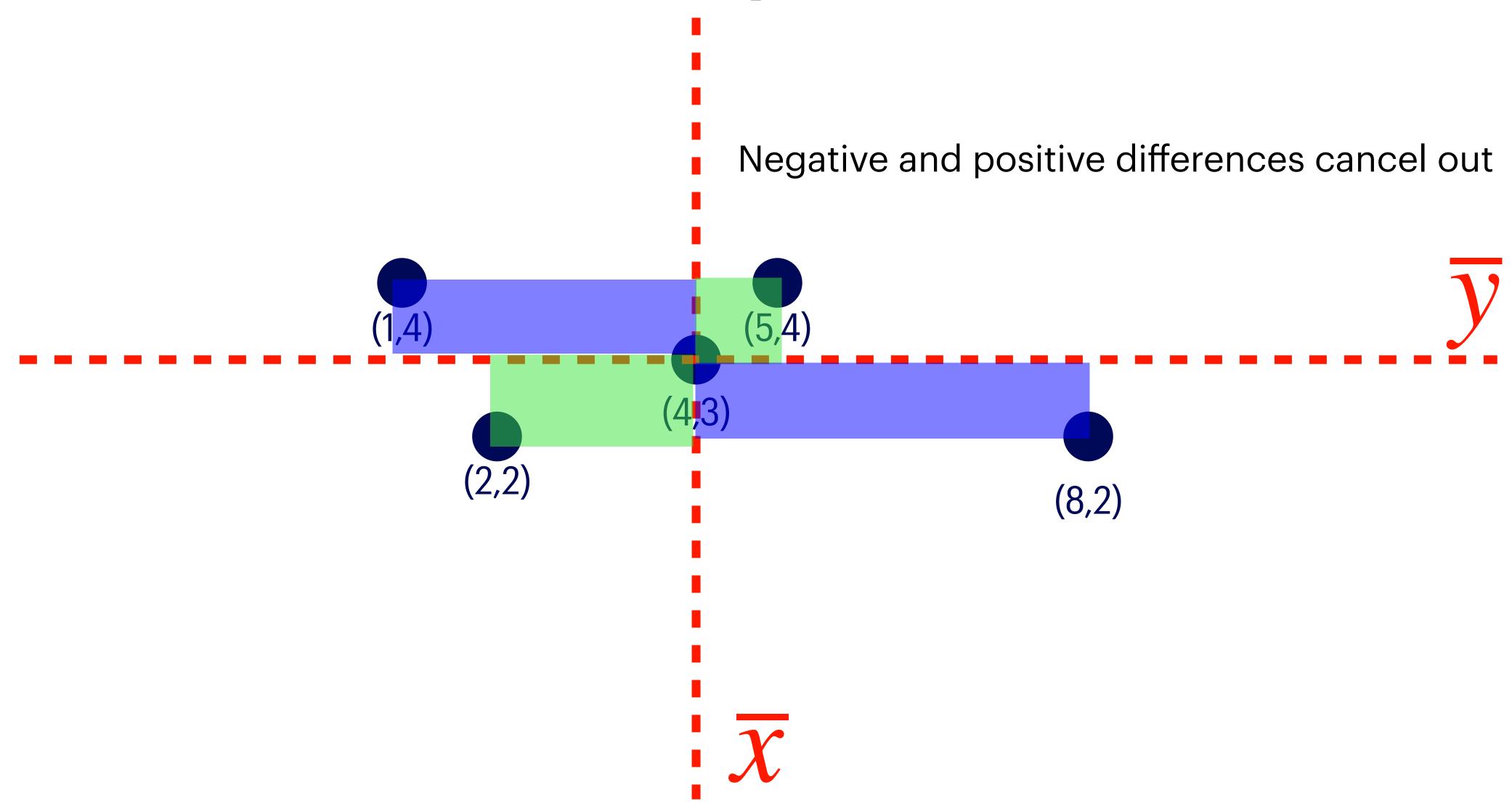


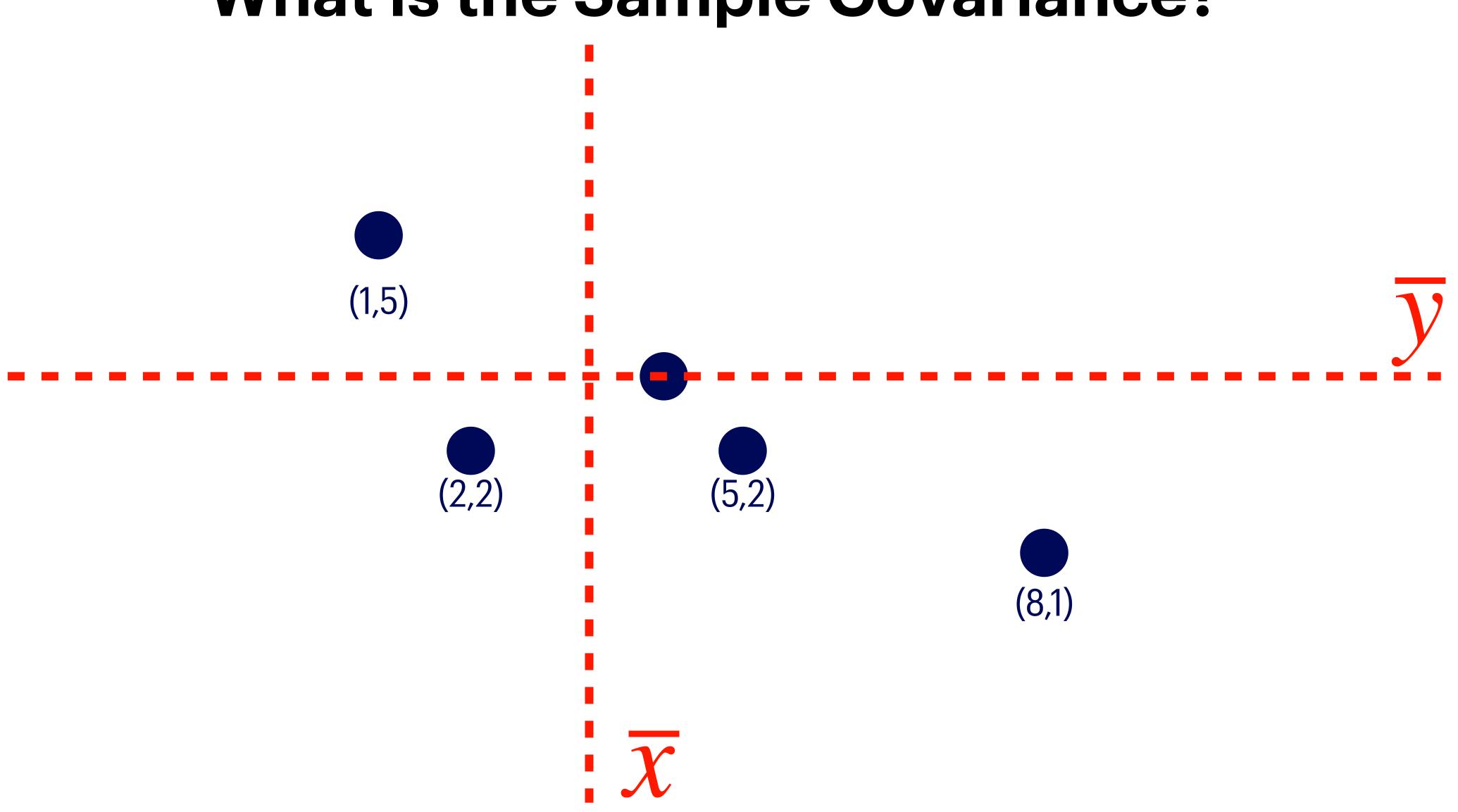


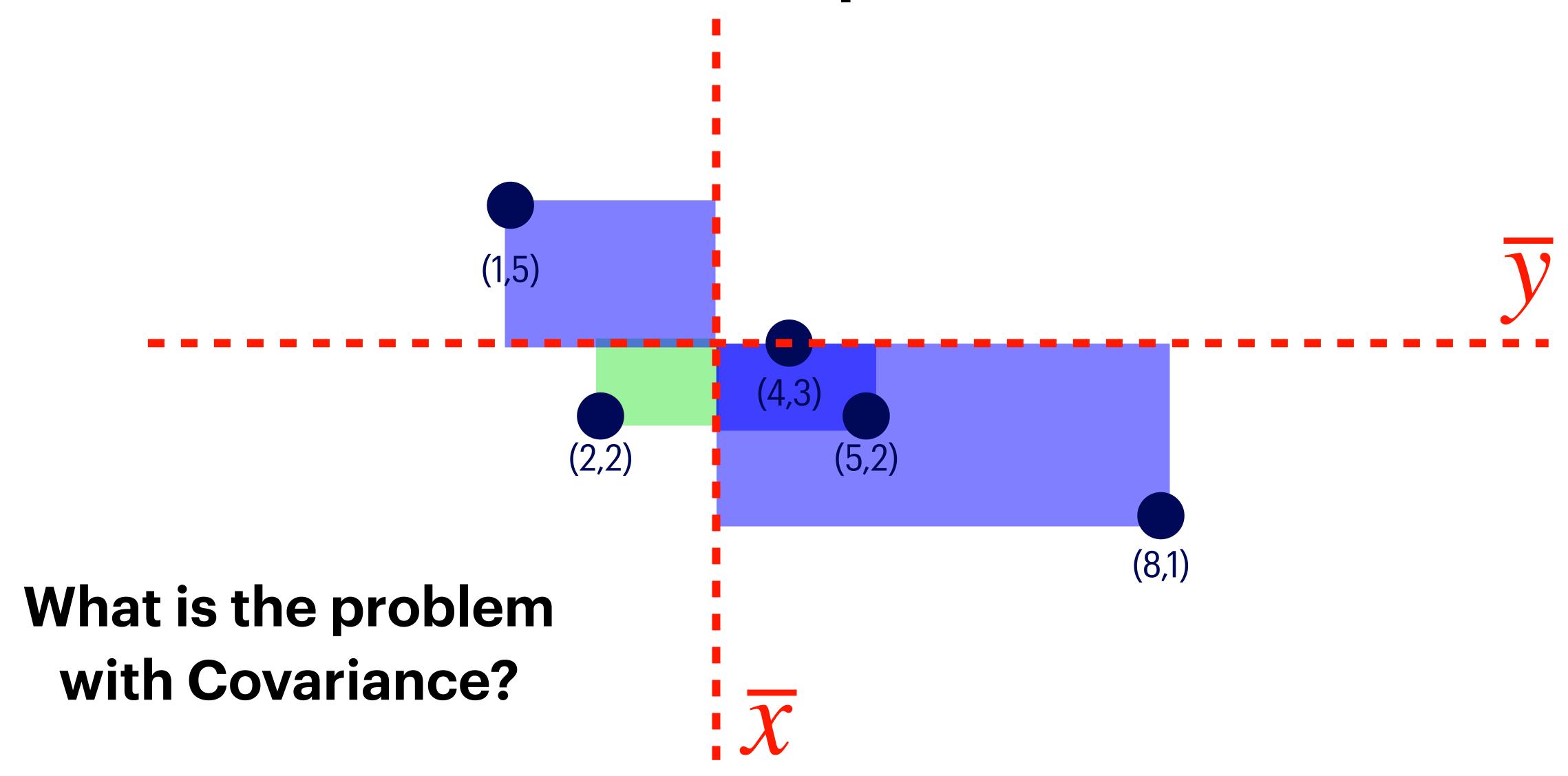












**More Examples:** http://digitalfirst.bfwpub.com/stats\_applet/stats\_applet\_5\_correg.html

## What is the Sample Correlation?

$$Cov(X, Y) = \frac{\sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})}{n-1}$$

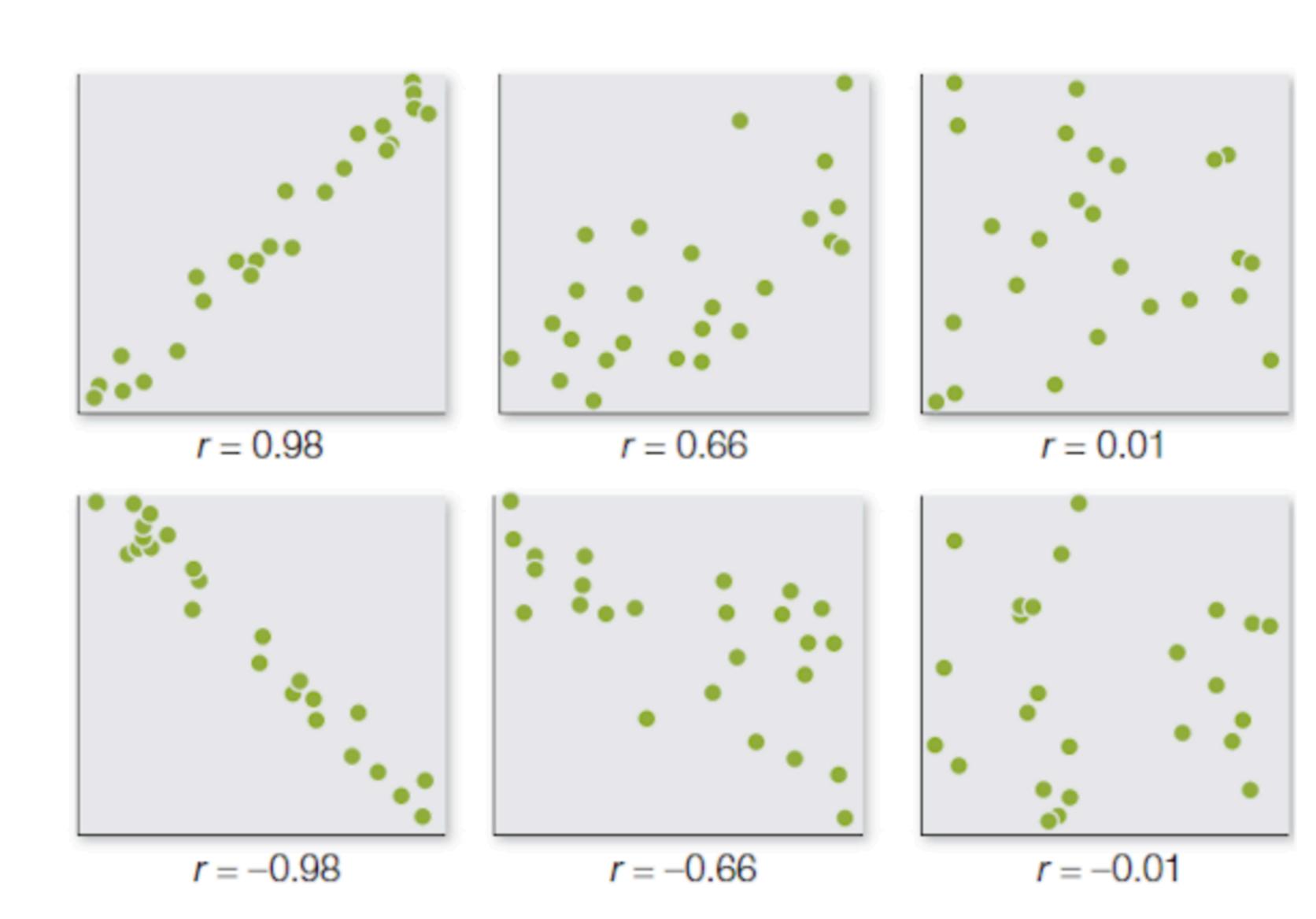
$$r = \frac{Cov(X, Y)}{s_{x}s_{y}} = \frac{1}{n-1} \sum_{i=1}^{n} \left(\frac{x_{i} - \overline{x}}{s_{x}}\right) \left(\frac{y_{i} - \overline{y}}{s_{y}}\right)$$

Note that r is a statistic

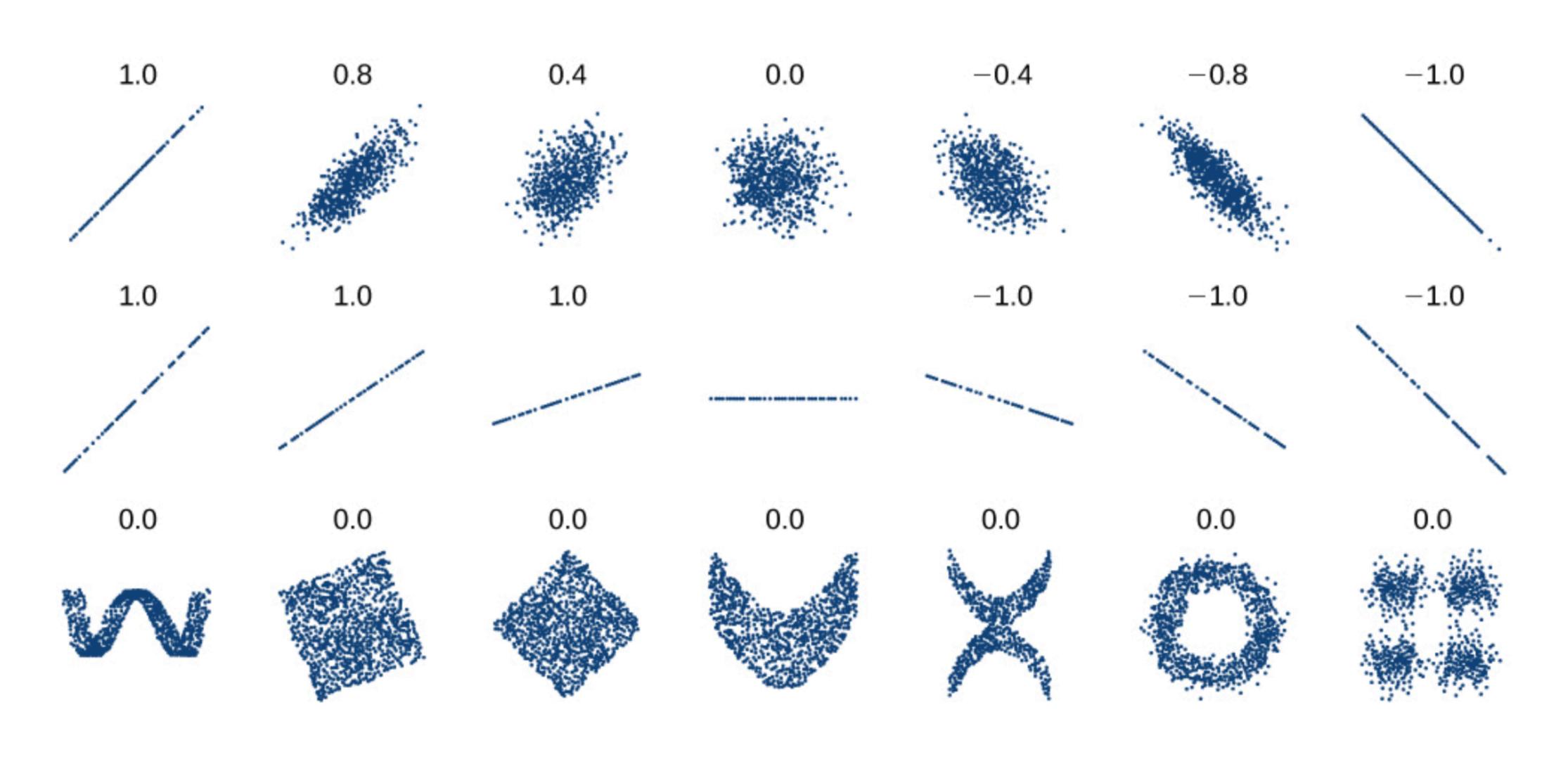
## What Does Correlation Measure?

- Direction
- Strength

Guess the Correlation
https://
www.rossmanchance.
com/applets/2021/
guesscorrelation/
GuessCorrelation.html

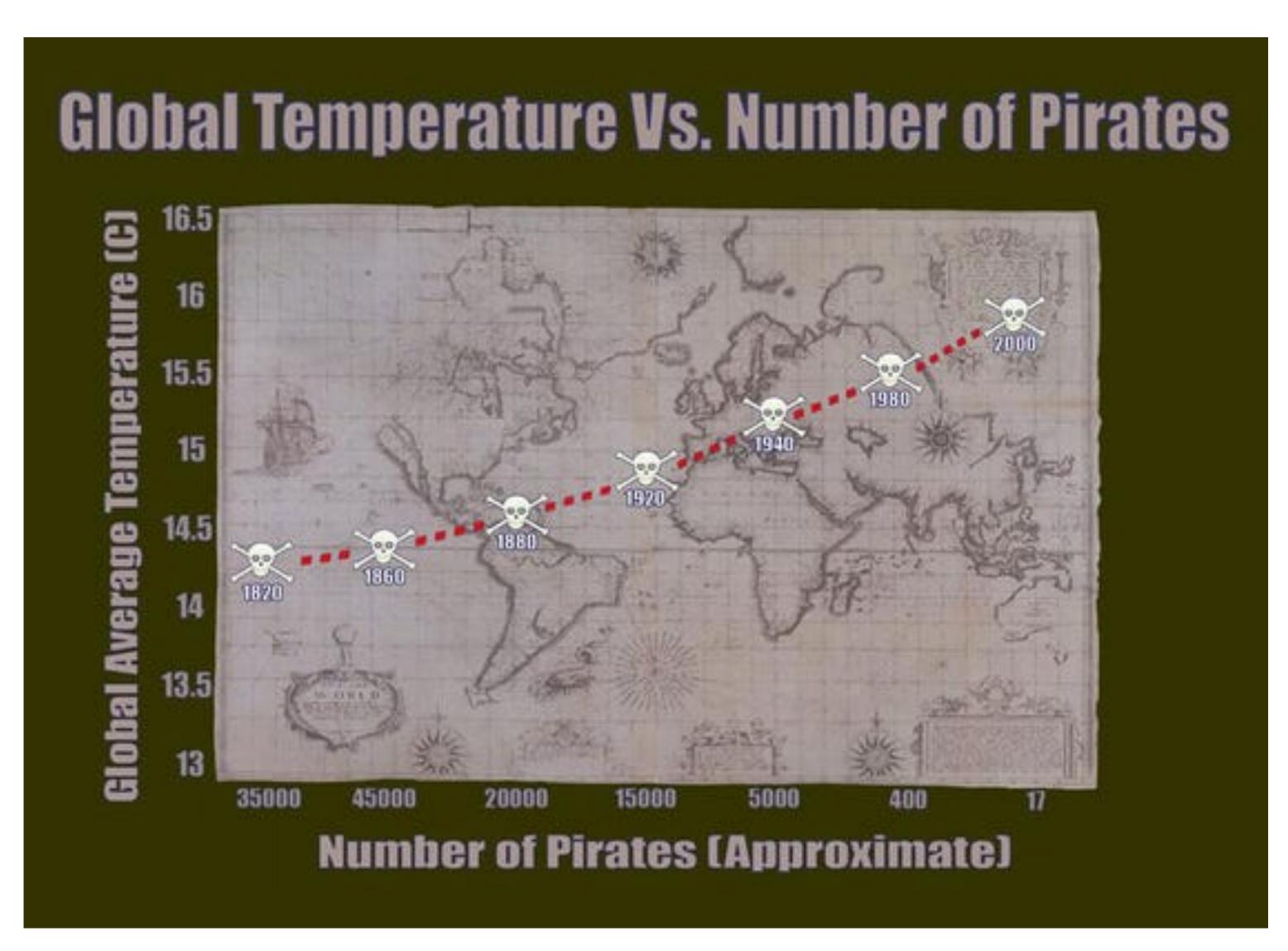


#### **ALWAYS PLOT THE DATA**

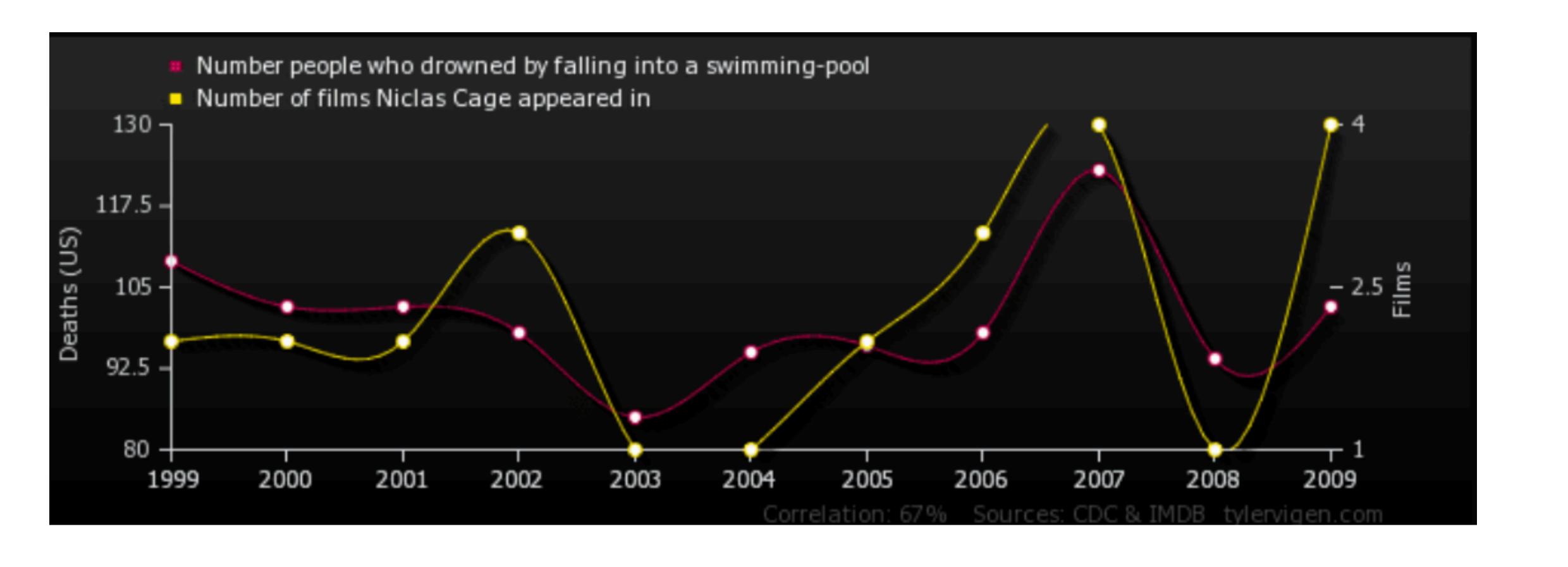


## What does correlation tell us about causation?

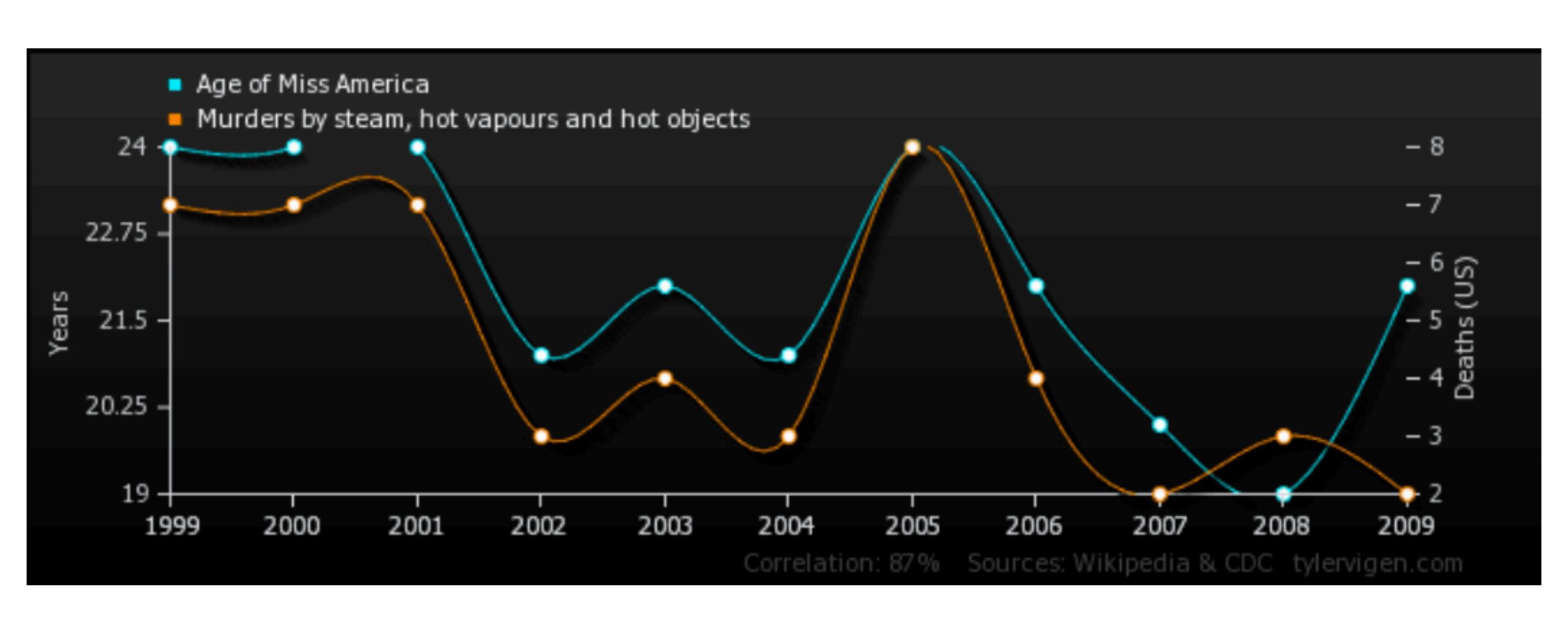
- Is a lack of pirates causing global warming?
- •Are Ice Cream
  Salesman
  responsible for
  increased drowning
  fatalities?



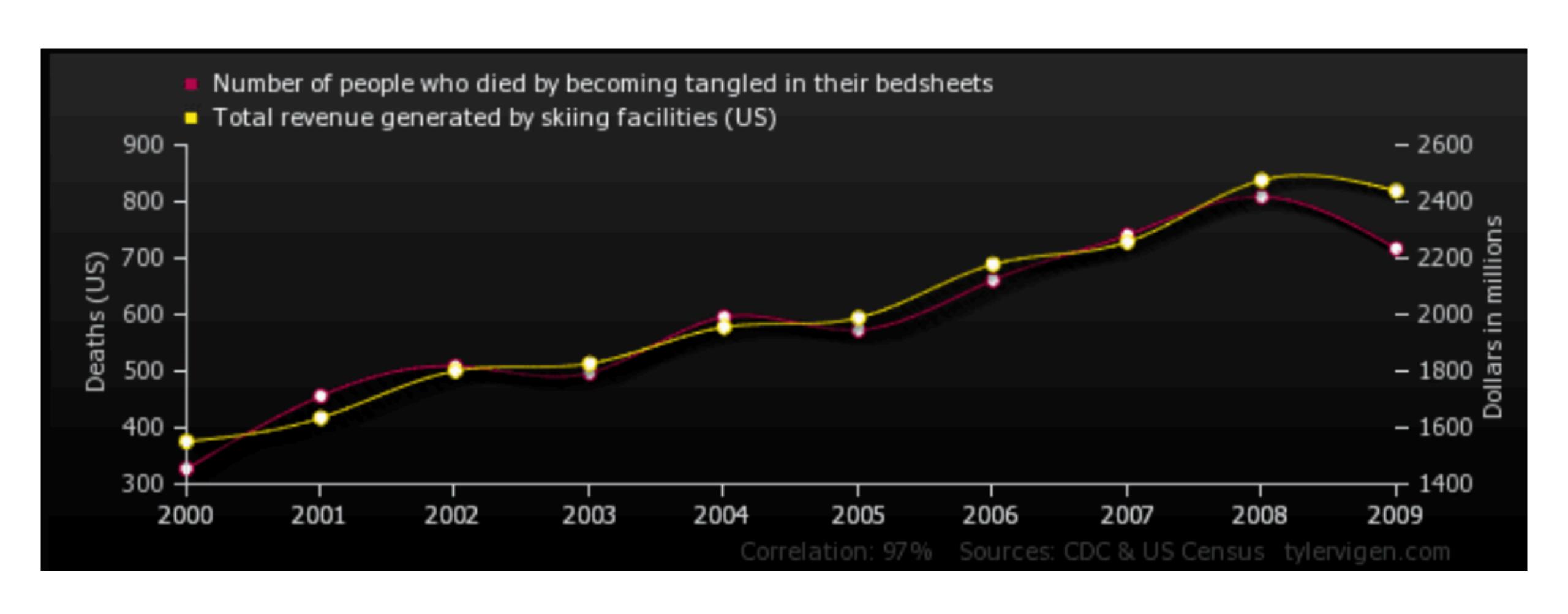
## **Correlation** ≠ Causation Examples



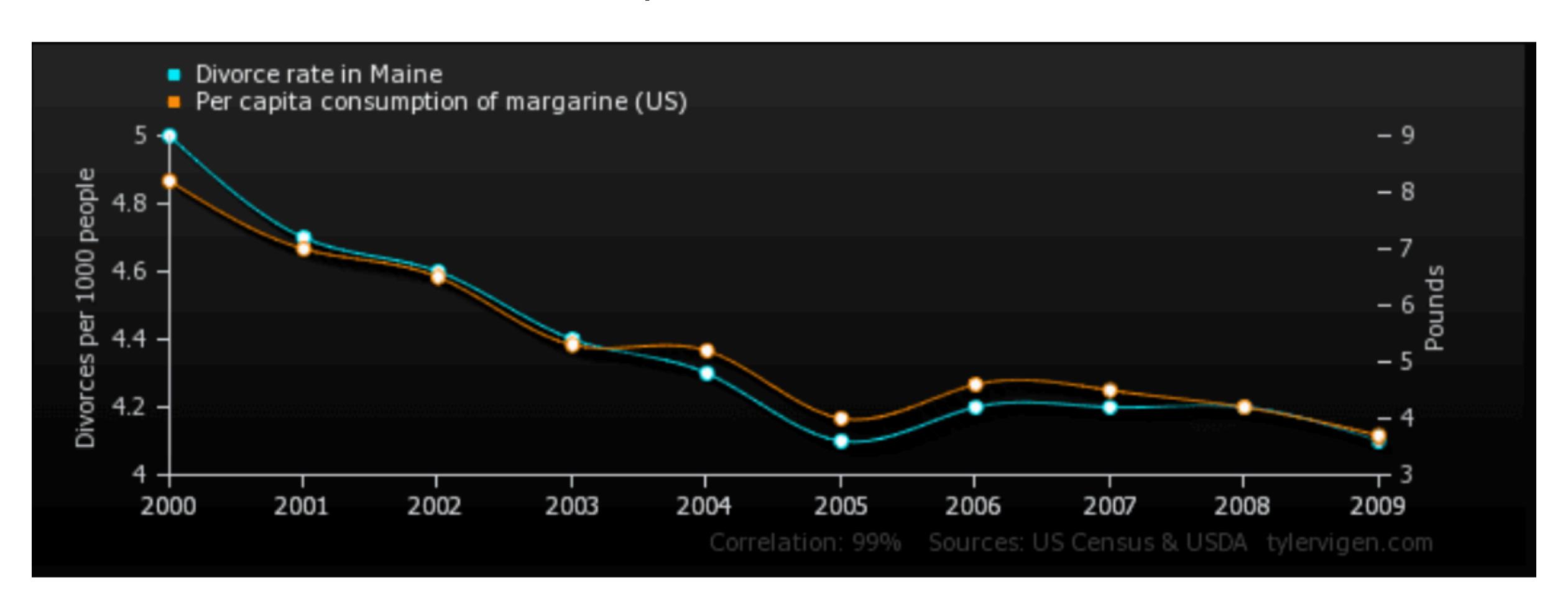
## **Correlation** $\neq$ **Causation Examples**



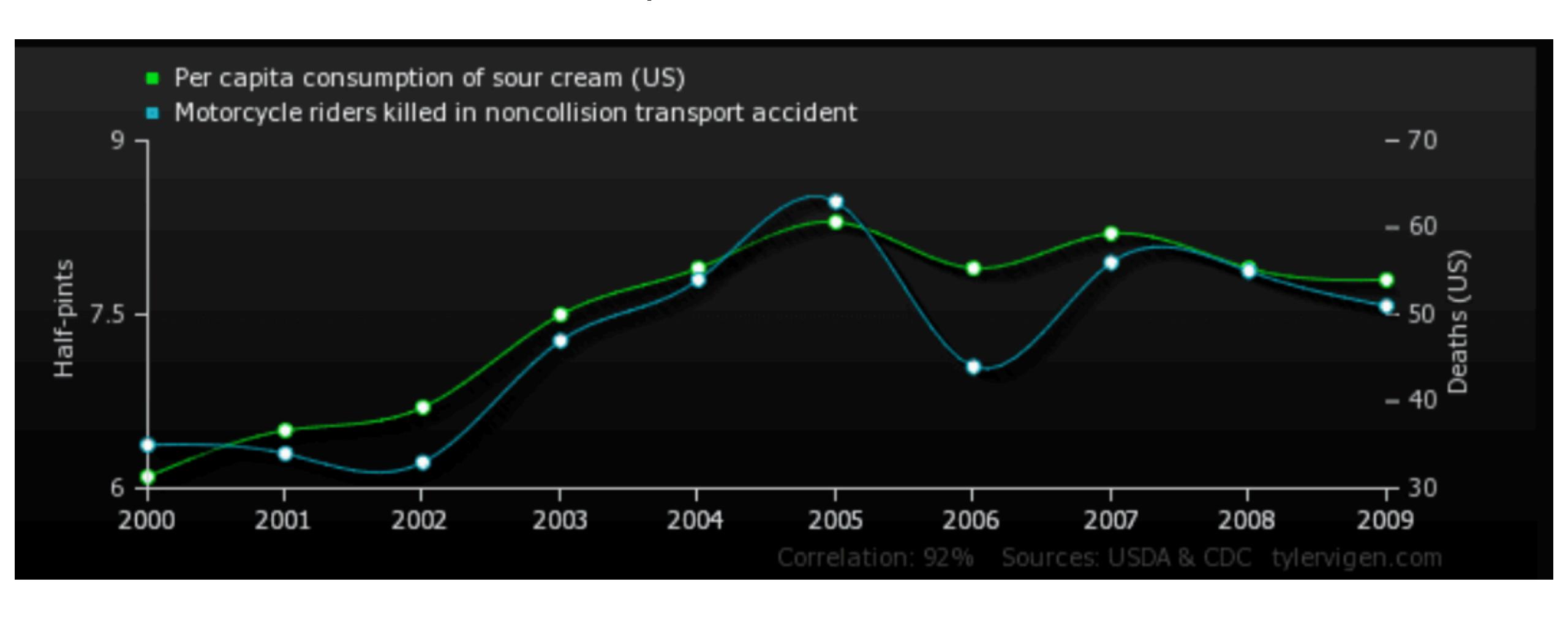
## **Correlation** $\neq$ **Causation Examples**



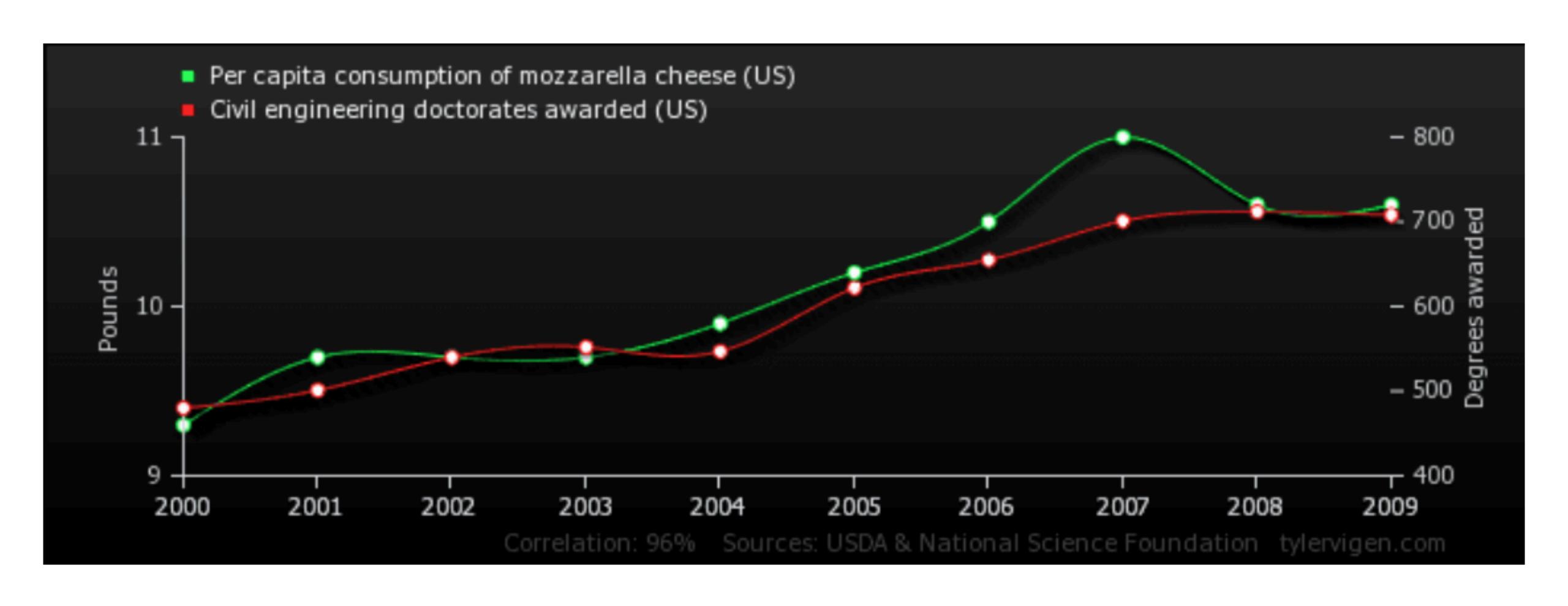
# **Correlation** $\neq$ **Causation Examples**



# **Correlation** $\neq$ **Causation Examples**



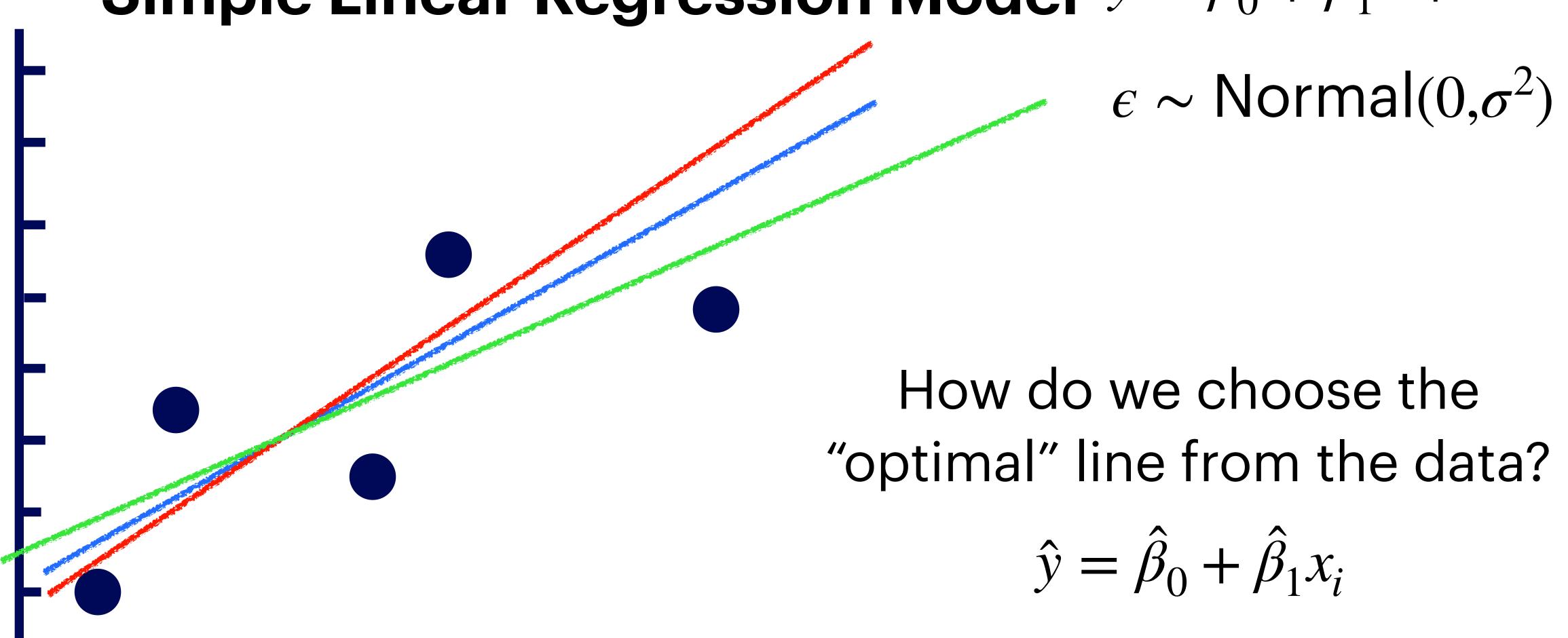
# **Correlation** $\neq$ **Causation Examples**



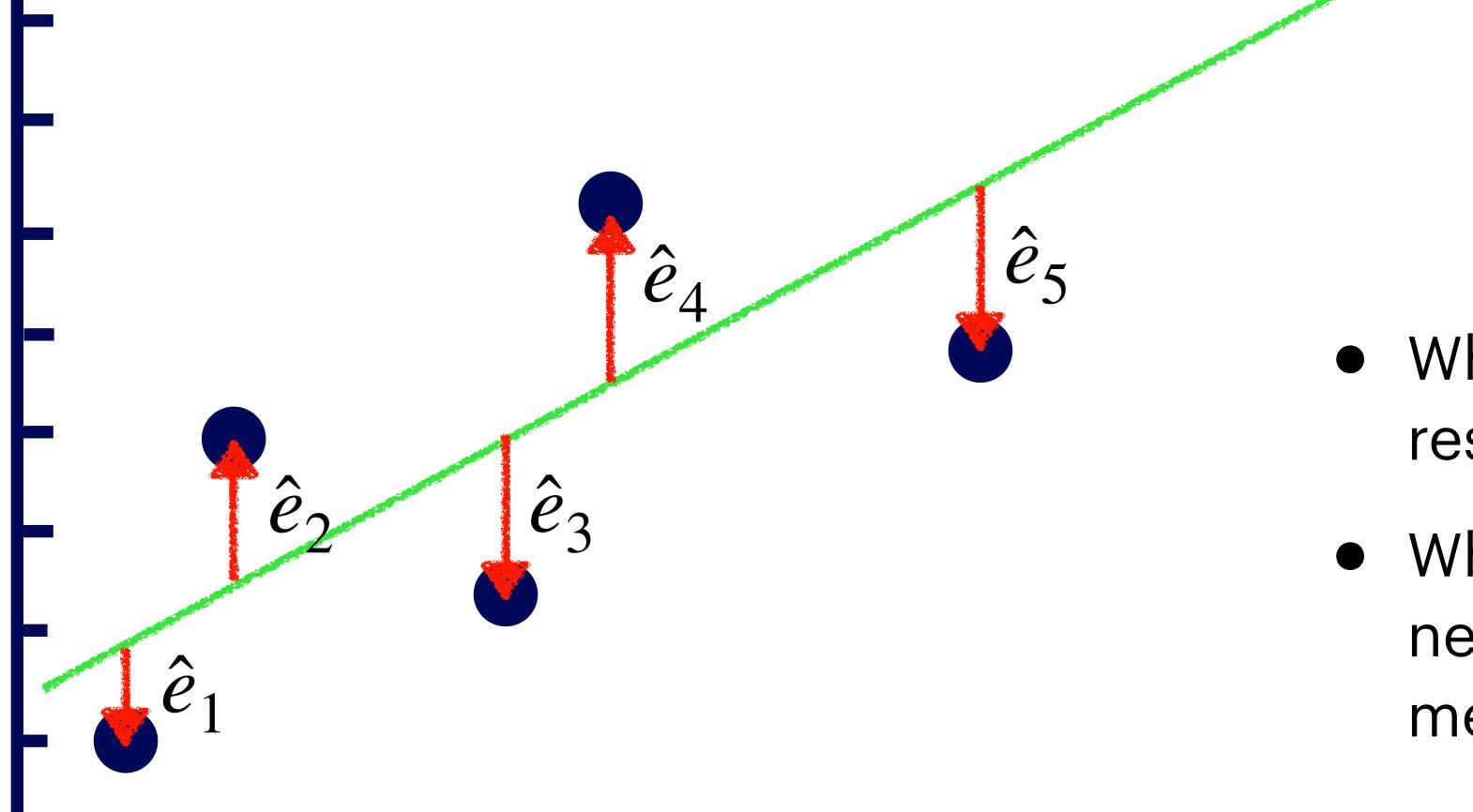
How do we "prove" something is a causal relationship?

**Experiments** will be discussed in more detail later (unit 3)

# Simple Linear Regression Model $y = \beta_0 + \beta_1 x + \epsilon$



# Simple Linear Regression Model $\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$

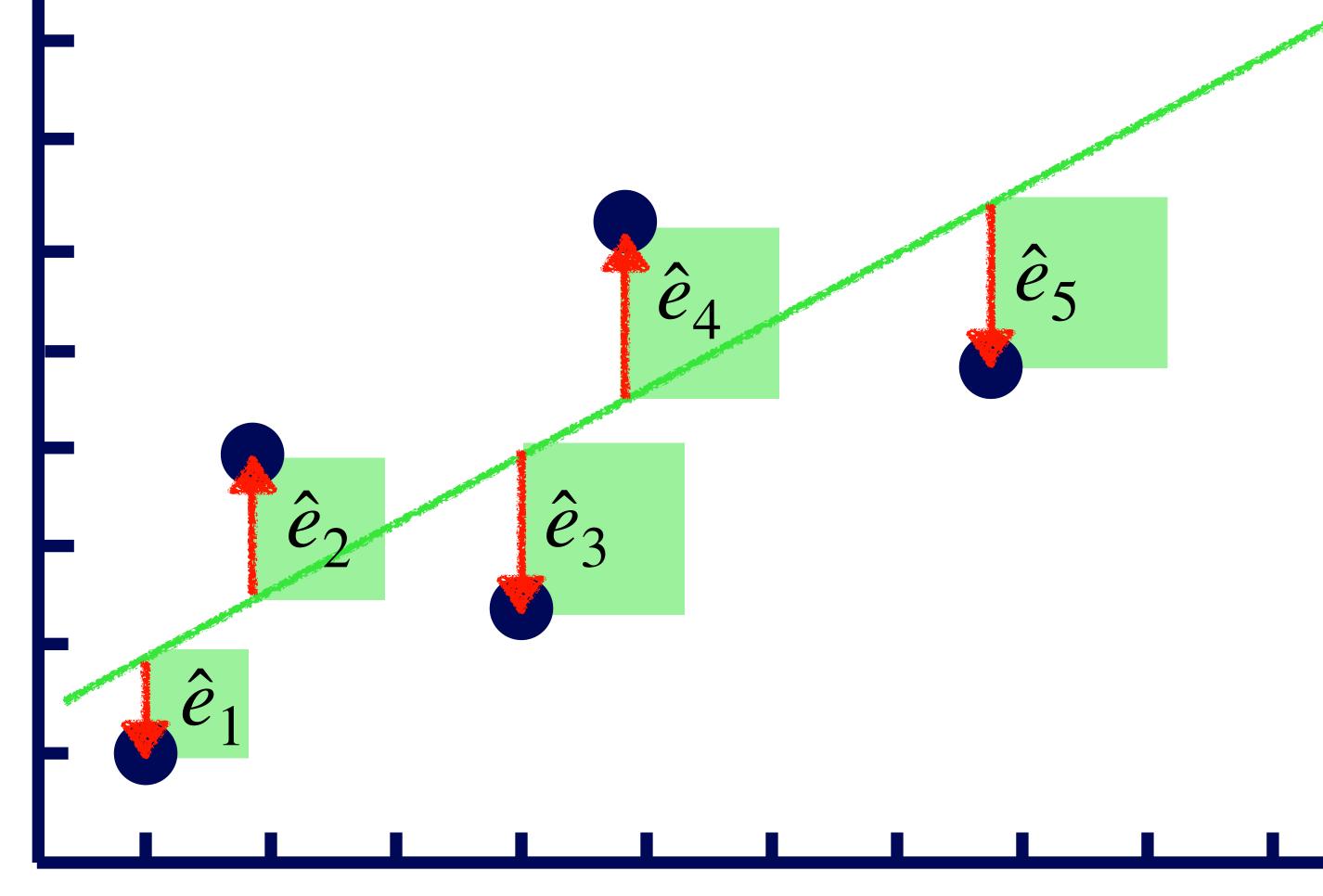


#### Residuals

$$\hat{e}_i = y_i - \hat{y}_i$$

- What does a positive residual mean?
- What does a negative Residual mean?

# Simple Linear Regression Model $\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$



# Sum of Square Residuals

$$\sum_{i=1}^{n} \hat{e}_i^2 = \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$

Want to minimize sum of square residuals w.r.t  $\hat{\beta}_0$ , and  $\hat{\beta}_1$  to get linear model

### The Derivation

$$0 = \frac{\partial}{\partial \hat{\beta}_0} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

$$0 = \frac{\partial}{\partial \hat{\beta}_1} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

https://www.desmos.com/calculator/lywhybetzt

### Formulas

$$\hat{\beta}_1 = r \frac{s_y}{s_x} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} \qquad b = r \frac{s_y}{s_x} \qquad b_1 = r \frac{s_y}{s_x}$$

$$\hat{\beta}_0 = \overline{y} - \hat{\beta}_1 \overline{x}$$
  $a = \overline{y} - b \overline{x}$   $b_0 = \overline{y} - b_1 \overline{x}$ 

$$\bar{y} = \hat{\beta}_0 + \hat{\beta}_1 \bar{x}$$

Don't forget that our line of best fit will always pass through  $(\overline{x}, \overline{y})$ 

$$\overline{y} = a + b\overline{x}$$

 $\hat{\beta}_0$  Represents the average value of "y" when "x" is zero. This is often meaningless

 $\hat{\beta}_1$  Represents the average increase in "y" for a **one unit** change in "x". Think Rise/One

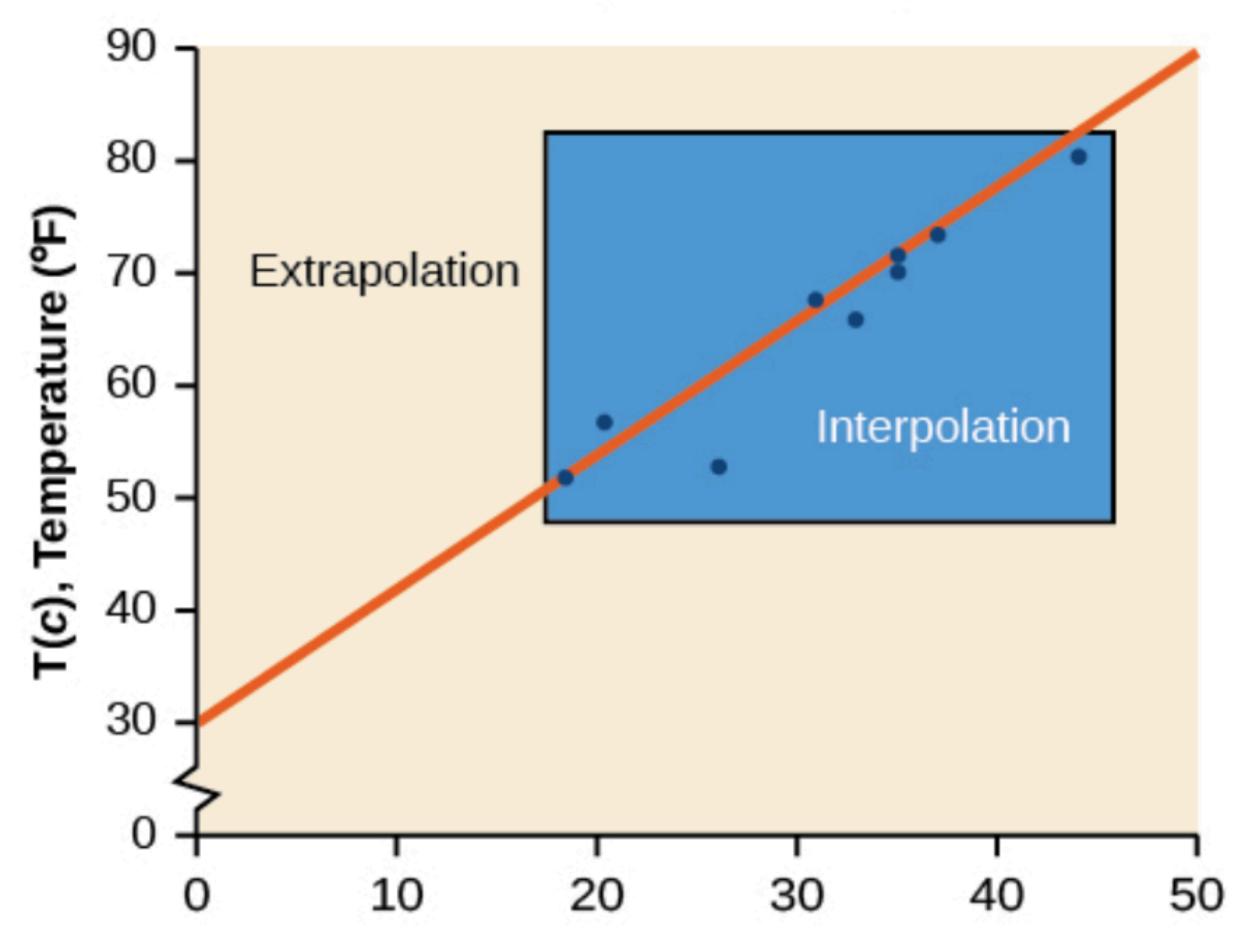
# Making Predictions: What does a prediction mean?

Average value of y given value of x. "Using our model we would predict an average temperature of Y for x Cricket Chirps in 15 seconds.

### What is extrapolation?

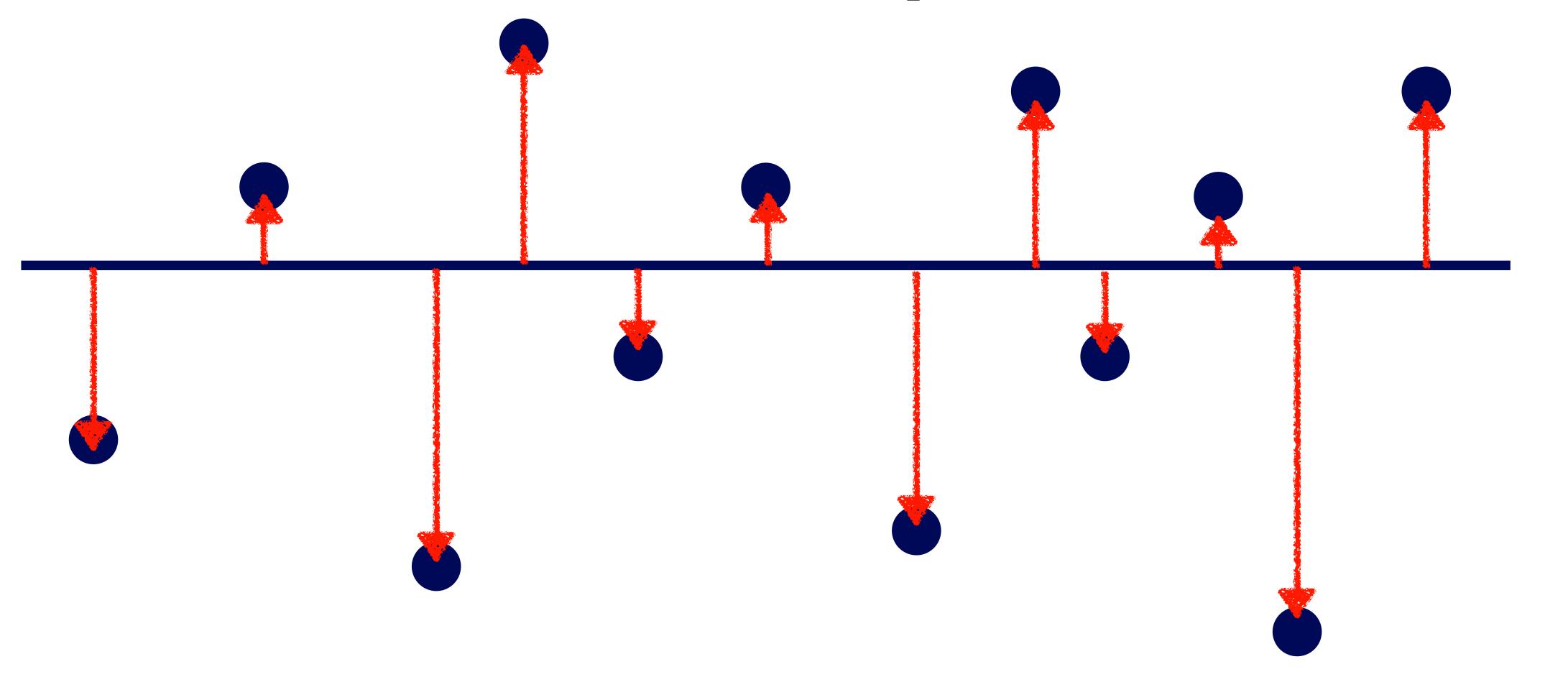
What does 0 cricket chirps in 15 seconds tell us?



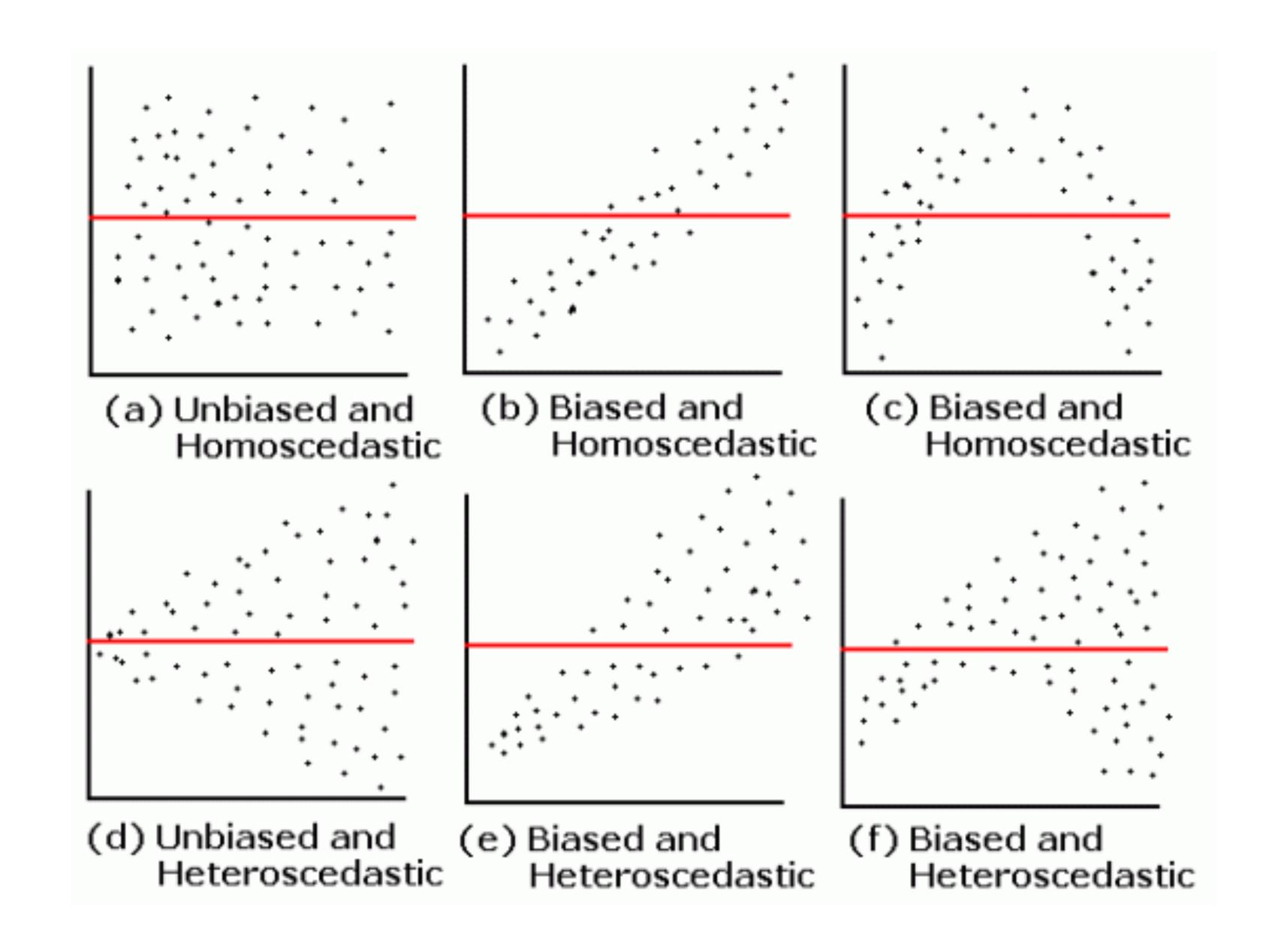


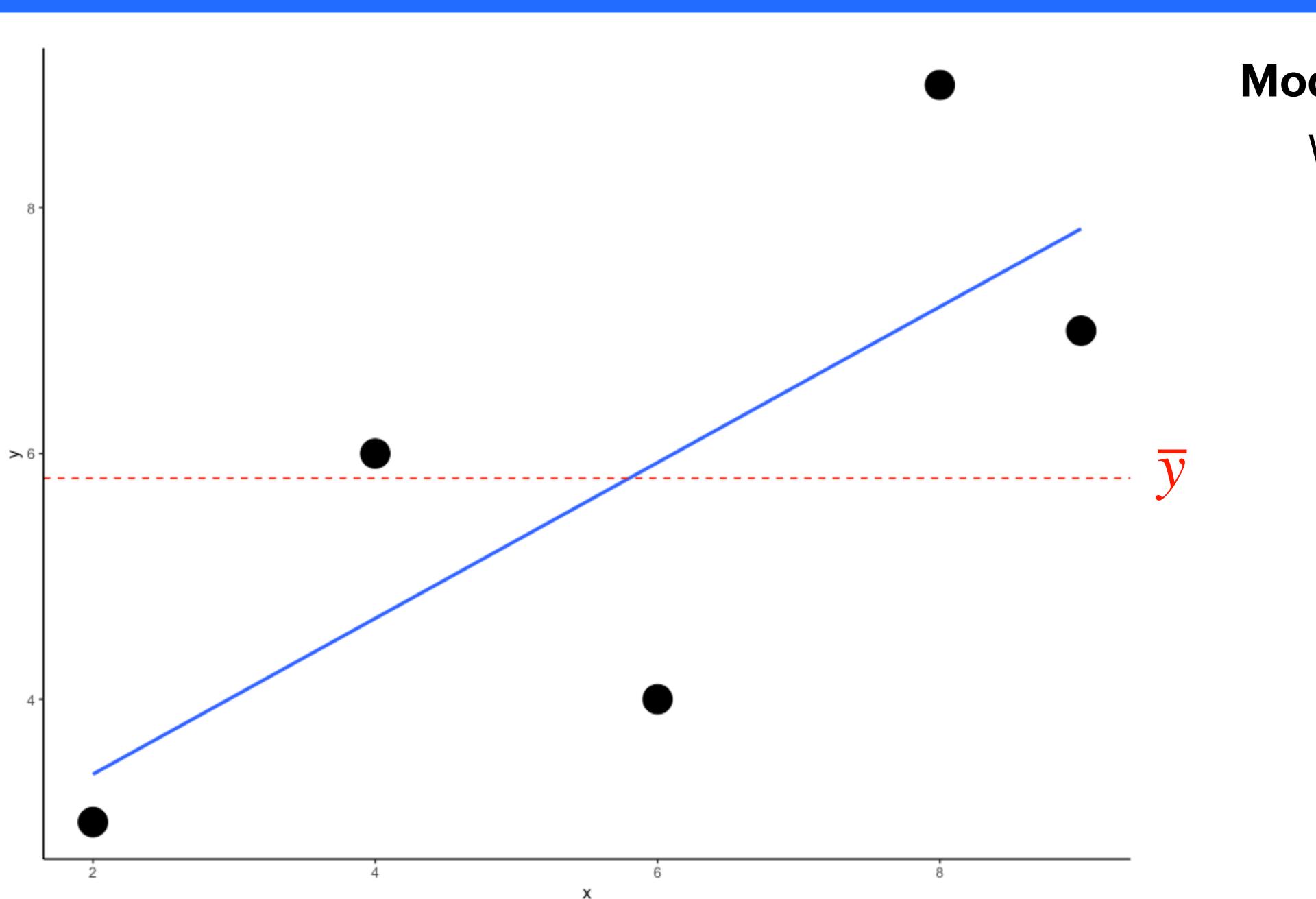
Model Validation: What is a residual plot?

Recall Our Model:  $y = \beta + \beta_1 x + \epsilon$   $\epsilon \sim \text{Normal}(0, \sigma^2)$ 



### RESIDUAL PLOTS





#### **Model Validation:**

What is  $r^2$ ?

$$SST = \sum_{i=1}^{n} (y - \overline{y})^2$$

$$SSE = \sum_{i=1}^{n} (y - \hat{y})^2$$

$$SSR = \sum_{i=1}^{n} (\hat{y} - \overline{y})^2$$



$$y - \overline{y}$$



$$SST = \sum_{i=1}^{n} (y - \overline{y})^{2} \qquad SSR = \sum_{i=1}^{n} (\hat{y} - \overline{y})^{2} \qquad SSE = \sum_{i=1}^{n} (y - \hat{y})^{2}$$

$$r^2 = \frac{SSR}{SST} = 1 - \frac{SSE}{SST}$$

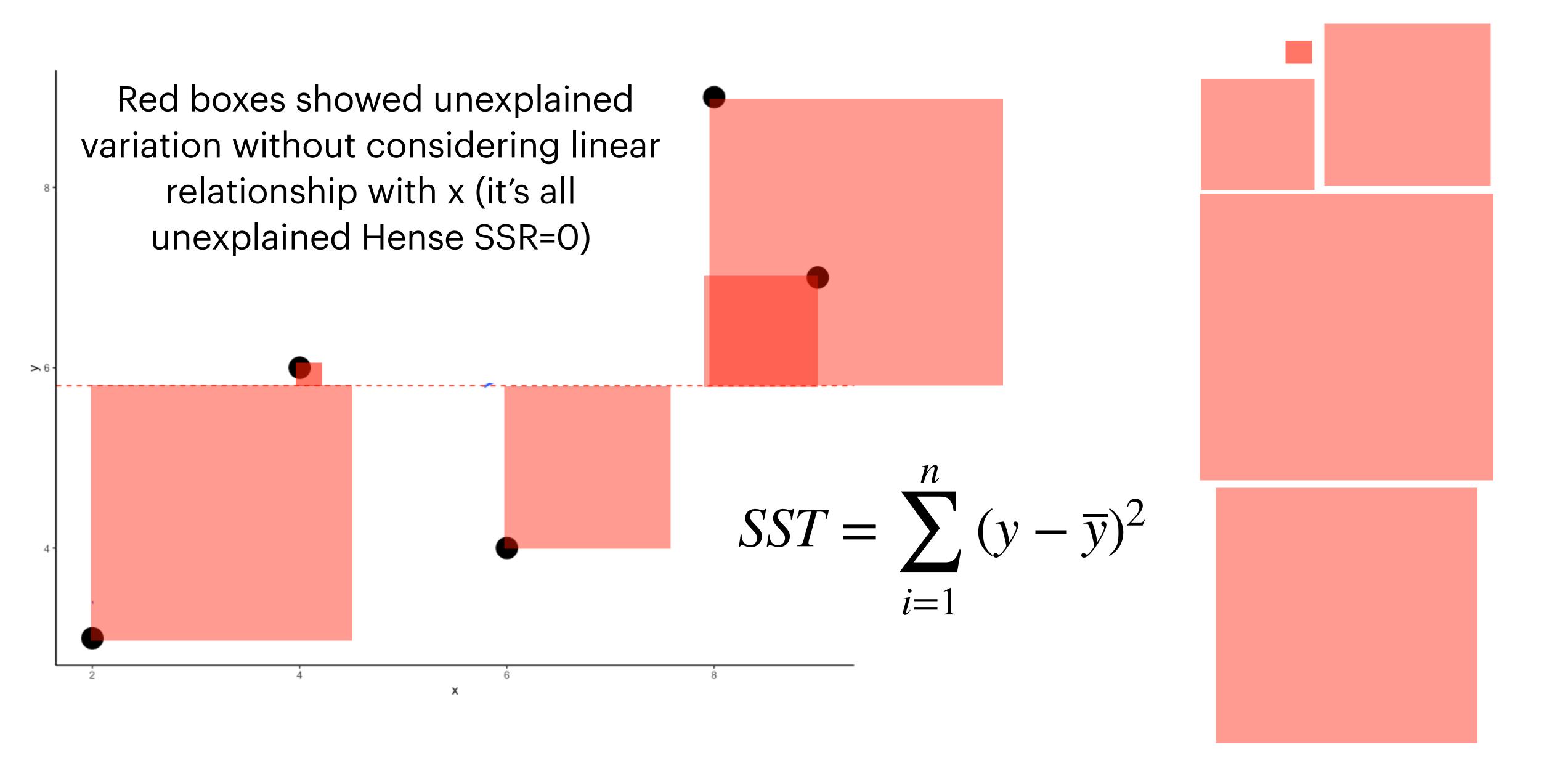
$$\hat{y} - \overline{y}$$

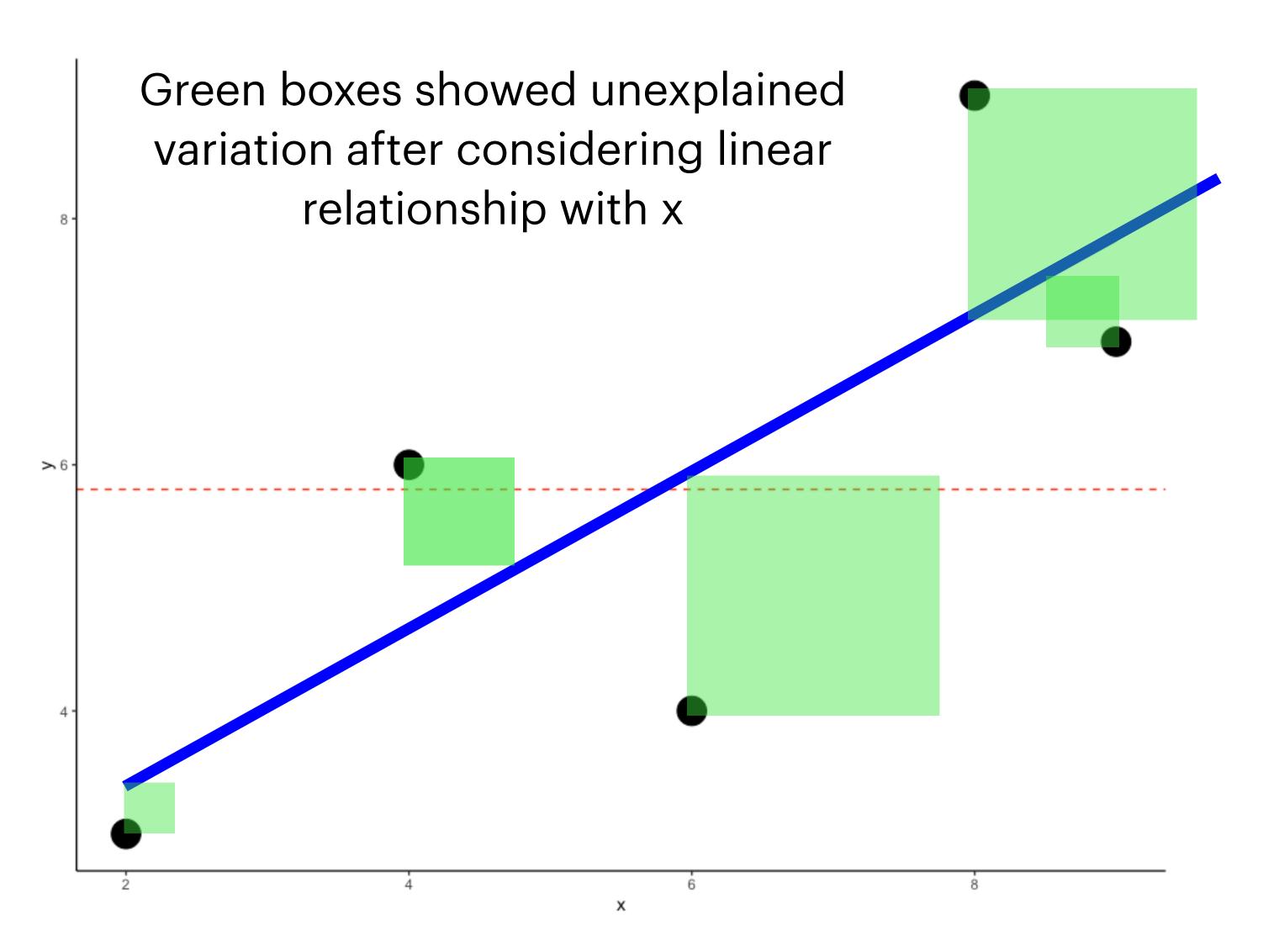
 $y - \bar{y}$ 

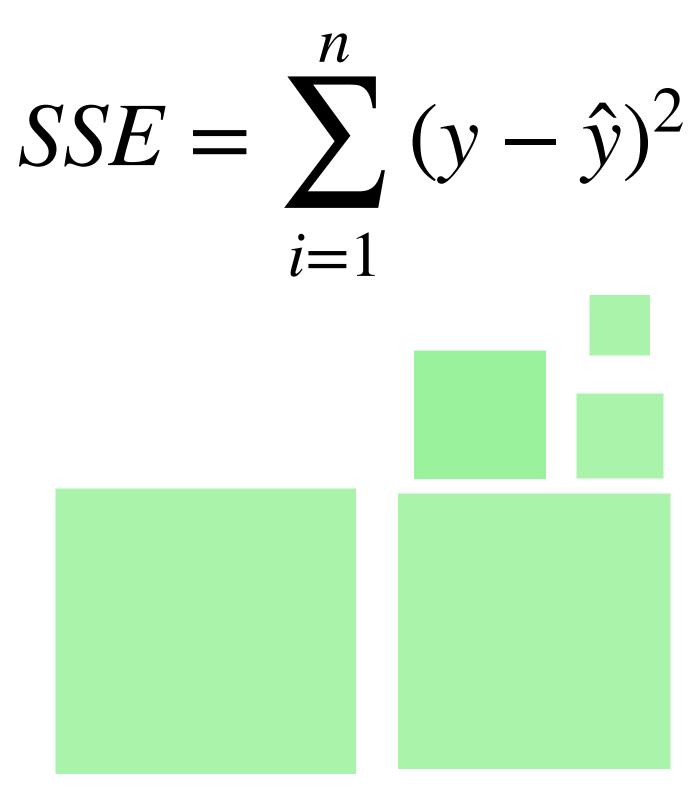
$$SST = \sum_{i=1}^{n} (y - \overline{y})^{2} \qquad SSR = \sum_{i=1}^{n} (\hat{y} - \overline{y})^{2} \qquad SSE = \sum_{i=1}^{n} (y - \hat{y})^{2}$$

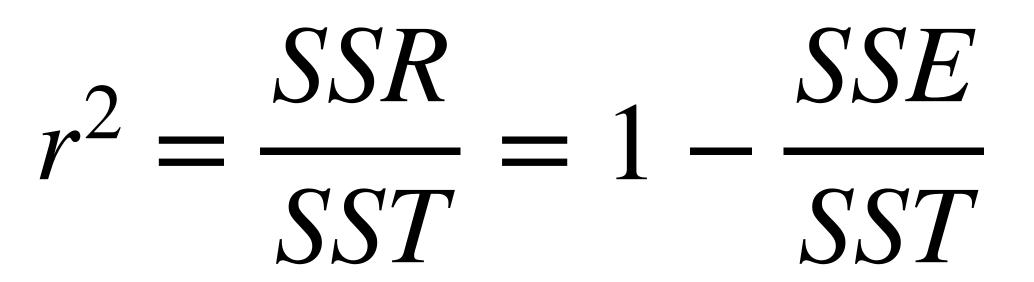
$$r^2 = \frac{SSR}{SST} = 1 - \frac{SSE}{SST}$$

 $r^2$  describes the percentage of variation in "y" that can be explained by "y's" linear relationship with "x"









When we consider the linear relationship the unexplained variance is reduced by  $\frac{SSE}{SST}$  percent. The percentage "explained" by the model is  $\frac{SSR}{SST}$ 

$$SSE = \sum_{i=1}^{n} (y - \hat{y})^2$$

**Standard Error** of the regression Line tells us the average residual length, in other words the average amount our model over/under predicts.

$$S = \sqrt{\frac{SSE}{n-2}}$$

Not expected to calculate by hand

#### BIVARIATE CATEGORICAL DATA

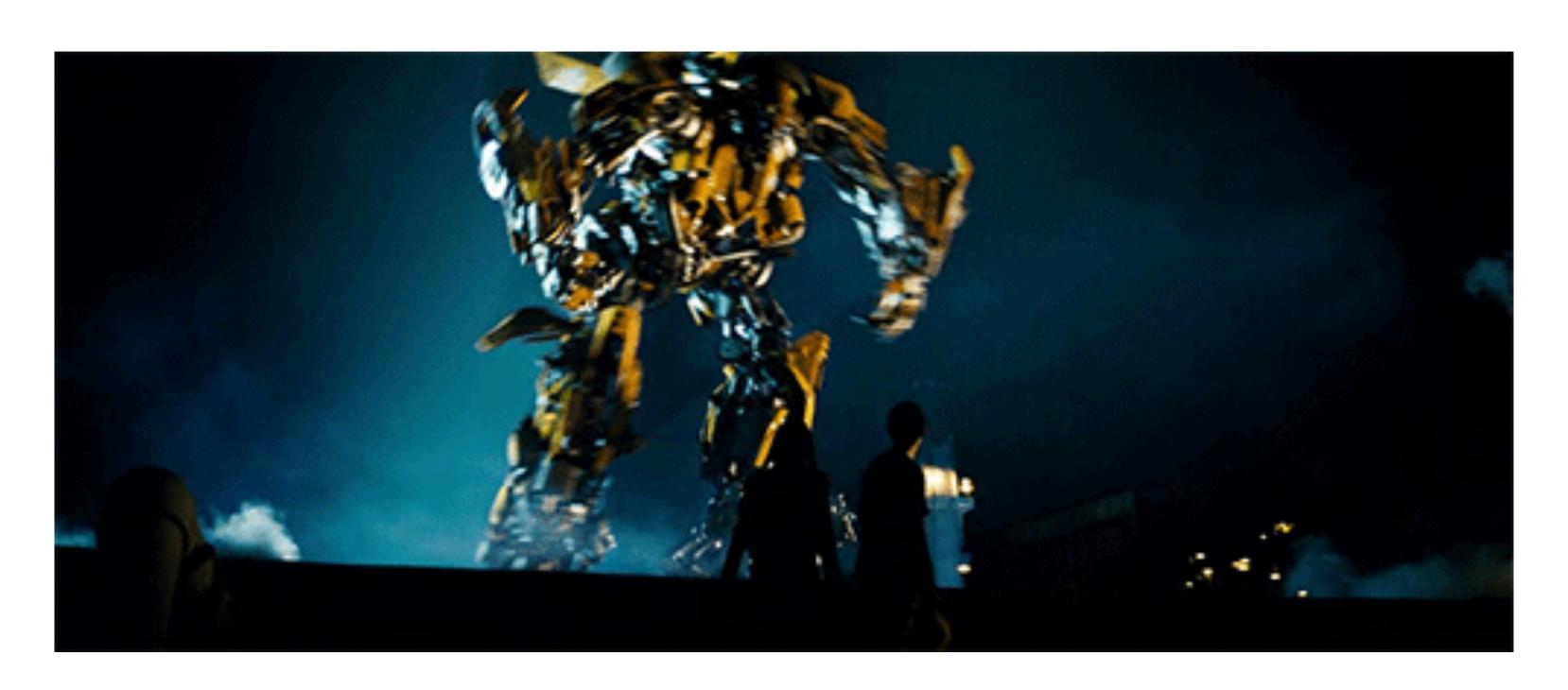
# Examples:

- Deep Thoughts Unit 2 Q1-Q4
- Question 1 Page 143

Homework: Read Pages 113-130 Barron's, Quiz 8, Quiz 9

### Transformations

Scatter Plot is Non-Linear



Linear Model Appropriate

# There are many different transformations we might use

Method	Transform	Regression equation	Predicted value (ŷ)
Standard linear regression	None	$y = b_0 + b_1 x$	$\hat{y} = b_0 + b_1 x$
Exponential model	DV = log(y)	$log(y) = b_0 + b_1x$	$\hat{y} = 10^{b_0 + b_1 x}$
Quadratic model	DV = sqrt(y)	$sqrt(y) = b_0 + b_1x$	$\hat{y} = (b_0 + b_1 x)^2$
Reciprocal model	DV = 1/y	$1/y = b_0 + b_1x$	$\hat{y} = 1 / (b_0 + b_1 x)$
Logarithmic model	IV = log(x)	$y=b_0+b_1\log(x)$	$\hat{y} = b_0 + b_1 \log(x)$
Power model	DV = log(y) $IV = log(x)$	$log(y)=$ $b_0 + b_1 log(x)$	$\hat{y} = 10^{b_0 + b_1 \log(x)}$

Example: the length of a year for a planet, based on its distance from the sun. Here are the data:

Distance (millions of miles)	Year (# of Earth-years)	
36	0.24	
67	0.61	
93	1	
142	1.88	
484	11.86	
887	29.46	
1784	84.07	
2796	164.82	
3666	247.68	

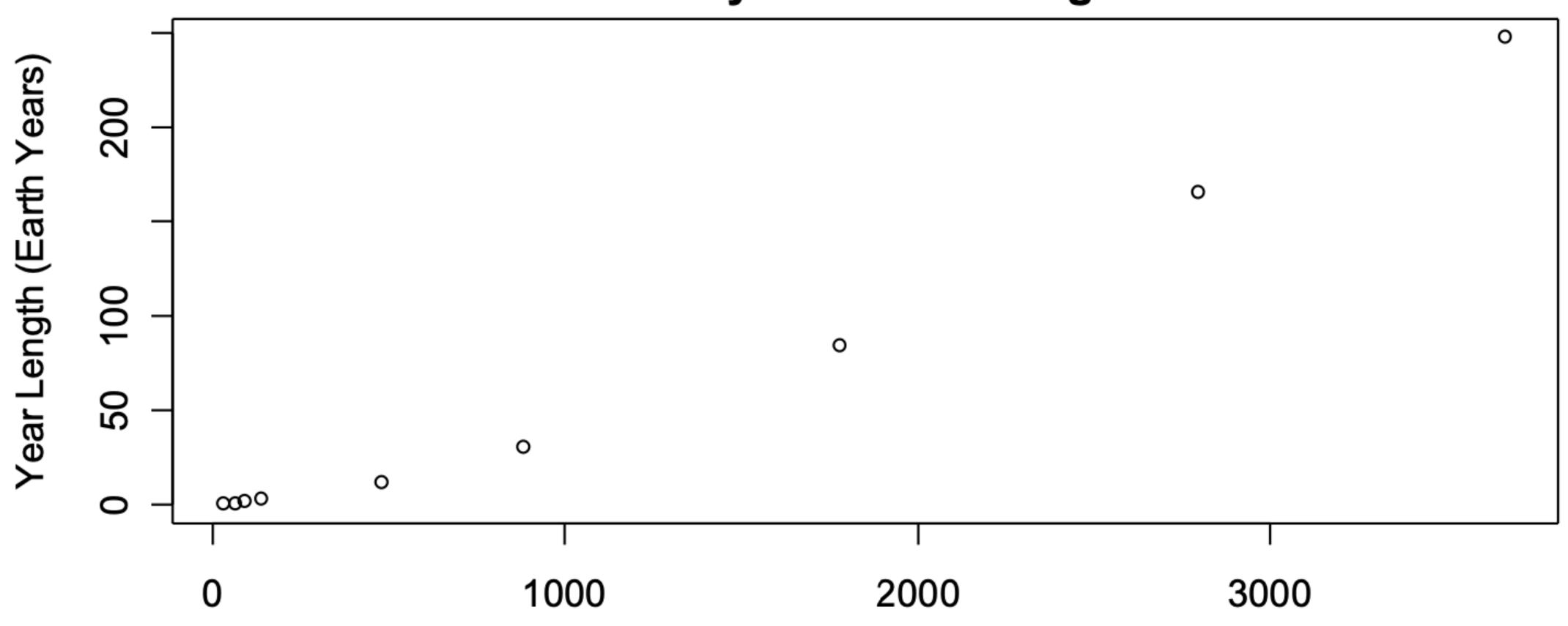
# 1. Let's run a simple linear regression.

What is  $r^2$ ?

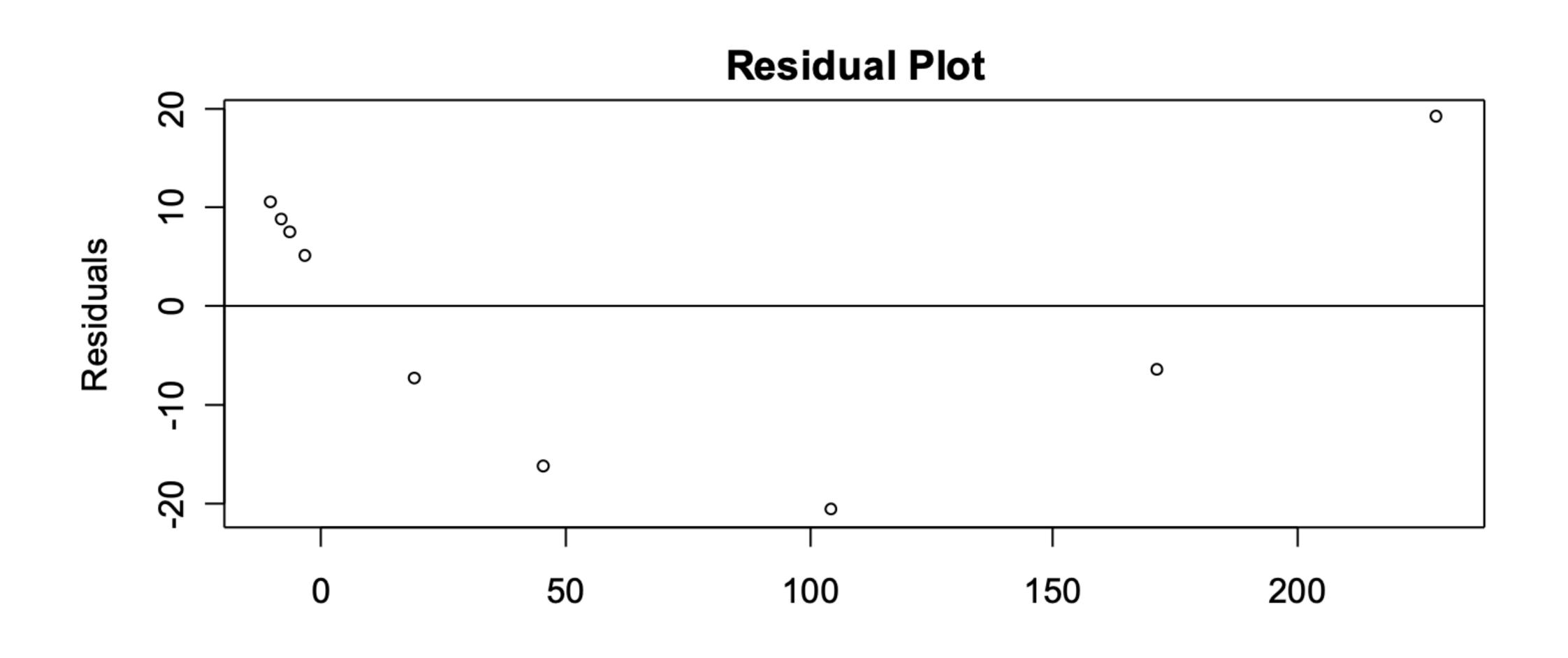
Is the Model Appropriate?

### Scatter Plot Looks non-linear





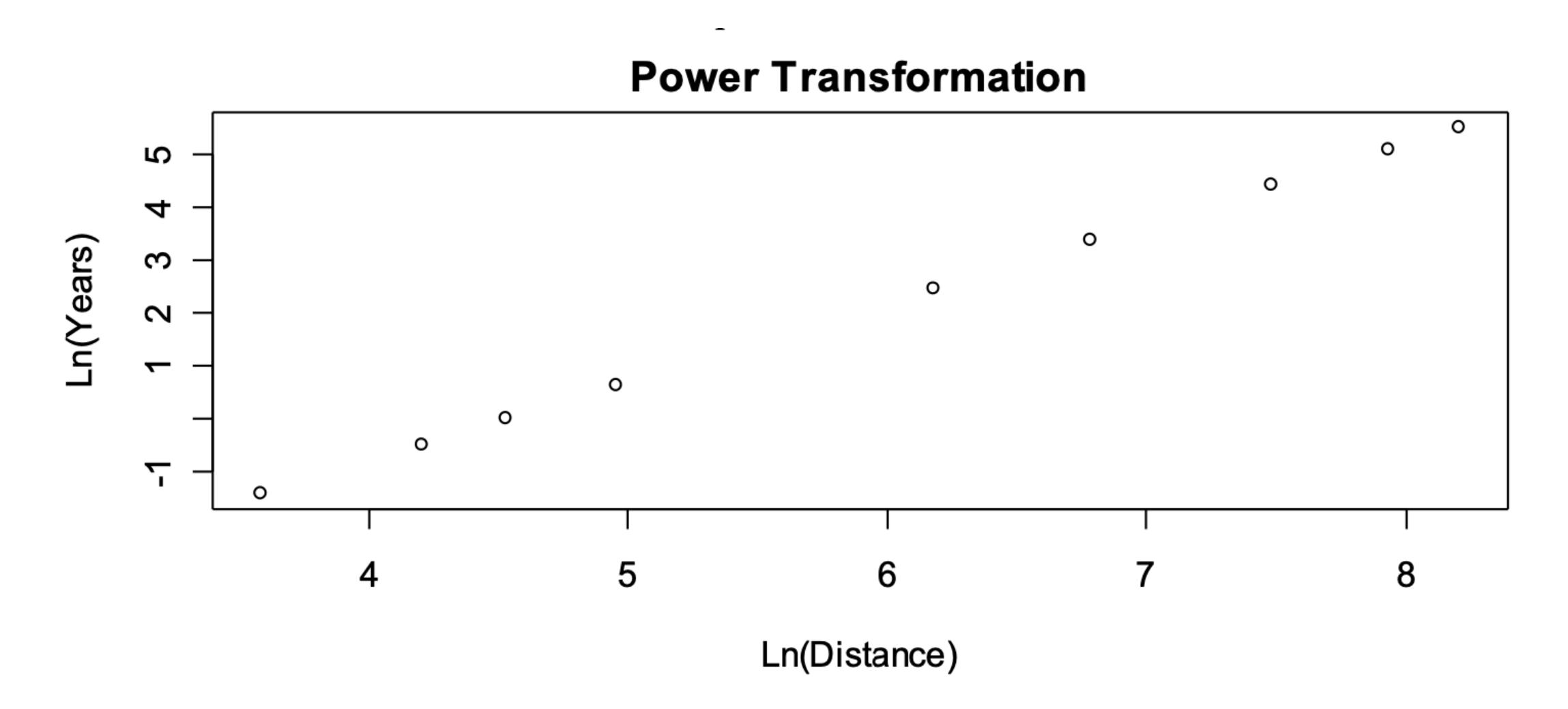
### Residual plot makes non-linear pattern even more clear



- 1. Let's run a simple linear regression.
- 2. Problem: EW, that's not linear. Lets apply a power transformation

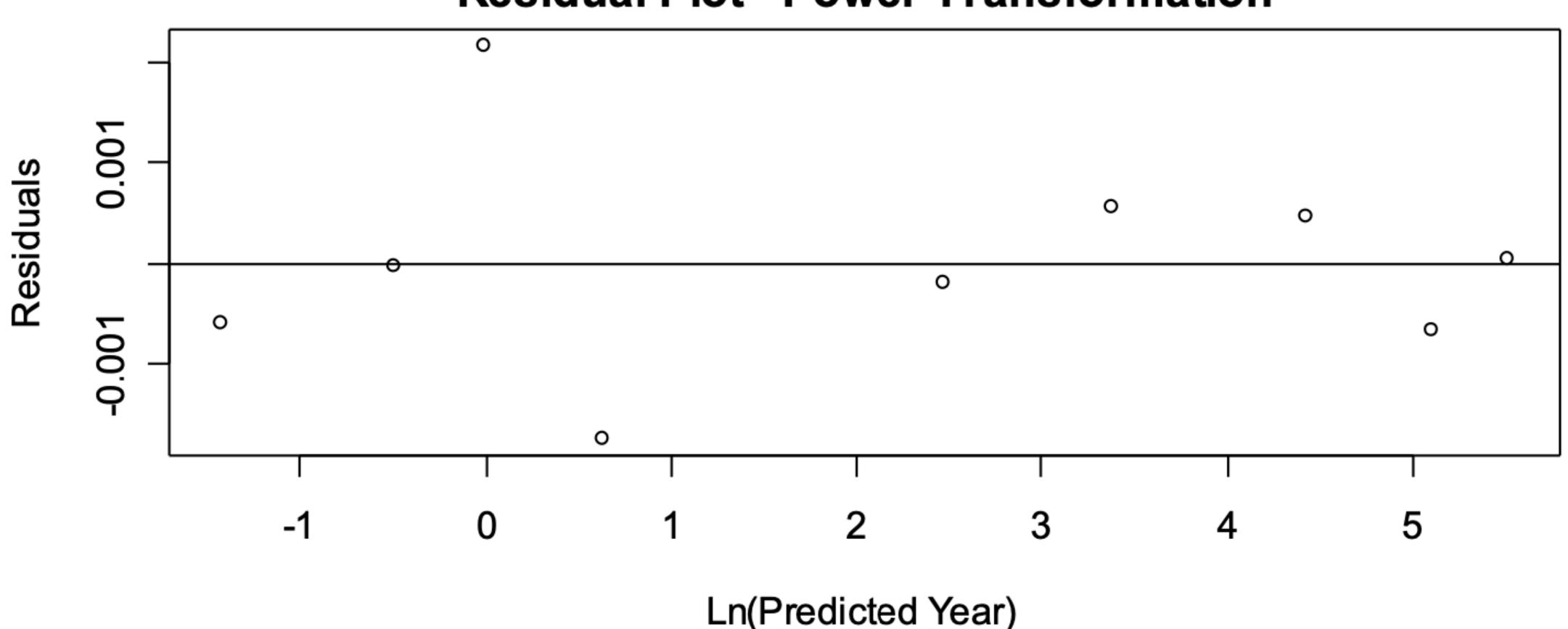


### New Scatter Plot Looks much more linear



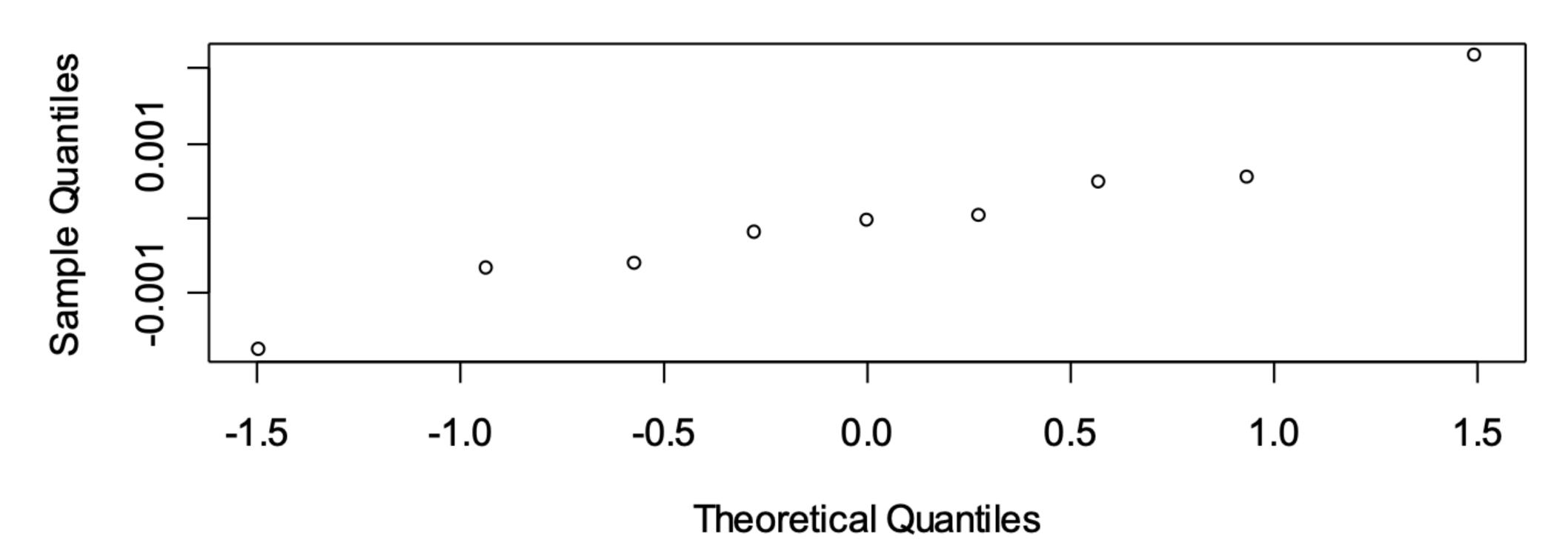
# Residual plot improves significantly





# Normality in residuals isn't bad either!

#### **Normal Probability Plot**



- 1. Let's run a simple linear regression.
- 2. Problem: EW, that's not linear. Lets apply a power transformation
- 3. Run simple linear regression with transformed data.
- **4. Model:**  $ln(\hat{y}) = -6.8046 + 1.5008 \cdot ln(x)$

Model:  $ln(\hat{y}) = -6.8046 + 1.5008 \cdot ln(x)$ 

Let's use this model to predict the year length of a planet that doesn't exist. The halfway point between Mars and Jupiter is around 313 million miles from Sol. What will this model predict for a year length if a planet occupied this position?

Model:  $ln(\hat{y}) = -6.8046 + 1.5008 \cdot ln(x)$ 

$$ln(\hat{y}) = -6.8046 + 1.5008 \cdot ln(313)$$

$$ln(\hat{y}) = 1.8192$$

$$\hat{y} = e^{1.8192} = 6.167$$

# What you need to know

- Recognize the need for a transformation
- Justify a transformations appropriateness

## **Examples:**

- Barron's pg. 130 Example 2.26
- Deep Thoughts Q5-Q6