

# Econ 103 – Statistics for Economists

## Chapter 2 feat. Z Scores and OLS

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# Survey Results

To be added soon.

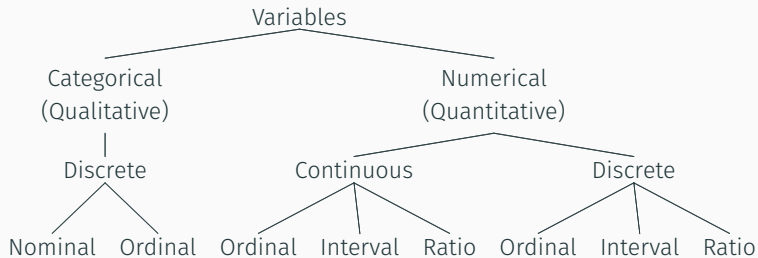
# Types of Variables

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# Discussion!

- What are the differences between the following variables?
  - “Age” and “gender”
  - “Gender” and “class standing”
  - “SAT score” and “job creation”

# A Few Definitions: A Taxonomy of Variables



# Definitions

- **Discrete:** Can be a countable number of values
- **Continuous:** Can take on any value

# Discussion!

Can you order the following from weakest to strongest?

Interval, Nominal, Ordinal, Ratio.

## From Weakest to Strongest

- **Nominal:** no order to the categories



## From Weakest to Strongest

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## From Weakest to Strongest

- **Nominal:** no order to the categories
- **Ordinal:** categories with natural order
- **Interval:** only differences meaningful, no natural zero
- **Ratio:** differences and ratios meaningful, natural zero

# Summary Statistics

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## 1. Measures of Central Tendency

- **Mean:** the average (“balance point”)

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

- **Median:** the middle observation (if data has even number of observations, take the mean of the middle two observations)

## 2. Percentiles

$P^{th}$  Percentile = Value in  $(P/100) \cdot (n + 1)^{th}$  Ordered Position

## An Example: $n = 12$

60 63 65 67 70 72 75 75 80 82 84 85

$$\begin{aligned} Q_1 &= \text{value in the } 0.25(n+1)^{\text{th}} \text{ ordered position} \\ &= \text{value in the } 3.25^{\text{th}} \text{ ordered position} \\ &= 0.75 * 65 + 0.25 * 67 \\ &= 65.5 \end{aligned}$$

# Definitions

## 3. Measures of Spread

- **Variance:** the spread from the mean

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

- **Standard Deviation:** another way to measure the spread

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

- **Range:** the distance between the highest and lowest value

$$Range = |x_{max} - x_{min}|$$

- **Interquartile Range (IQR):** the distance between the upper and lower quartiles

$$IQR = |x_{75\%} - x_{25\%}|$$



# Why Squares?

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

# Why Squares?

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

What's Wrong With This?

$$\begin{aligned} \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x}) &= \frac{1}{n-1} \left[ \sum_{i=1}^n x_i - \sum_{i=1}^n \bar{x} \right] = \frac{1}{n-1} \left[ \sum_{i=1}^n x_i - n\bar{x} \right] \\ &= \frac{1}{n-1} \left[ \sum_{i=1}^n x_i - n \cdot \frac{1}{n} \sum_{i=1}^n x_i \right] \\ &= \frac{1}{n-1} \left[ \sum_{i=1}^n x_i - \sum_{i=1}^n x_i \right] = 0 \end{aligned}$$

## 4. Measure of Symmetry

- **Skewness:** a measure of symmetry, positive values means the right tail is longer and vice versa

$$Skewness = \frac{1}{n} \frac{\sum_{i=1}^n (x_i - \bar{x})^3}{s^3}$$

# Skewness – A Measure of Symmetry

$$\text{Skewness} = \frac{1}{n} \frac{\sum_{i=1}^n (x_i - \bar{x})^3}{s^3}$$

**What do the values indicate?**

Zero  $\Rightarrow$  symmetry, positive right-skewed, negative left-skewed.

**Why cubed?**

To get the desired sign.

**Why divide by  $s^3$ ?**

So that skewness is unitless

**Rule of Thumb**

Typically (but not always), right-skewed  $\Rightarrow$  mean  $>$  median  
left-skewed  $\Rightarrow$  mean  $<$  median

## 5. Relationship between variables (to be covered later)

- **Covariance:** how two variables vary together
- **Correlation:** normalized version of covariance (can only range between -1 and +1)
- **Regression:** an estimation of how two variables are related
- We'll cover these in-depth later in the semester

# Charts

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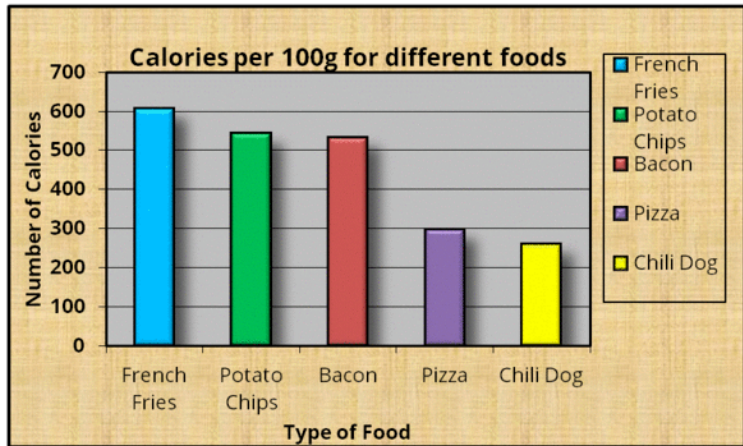
## Some Data Visualization Quotes

1. “Overload, clutter, and confusion are not attributes of information, they are failures of design” –Edward Tufte
2. “...few people will appreciate the music if I just show them the notes. Most of us need to listen to the music to understand how beautiful it is. But often, that’s how we present statistics; we just show the notes we don’t play the music.” –Hans Rosling

<https://i.imgur.com/W4BKCVU.gif>

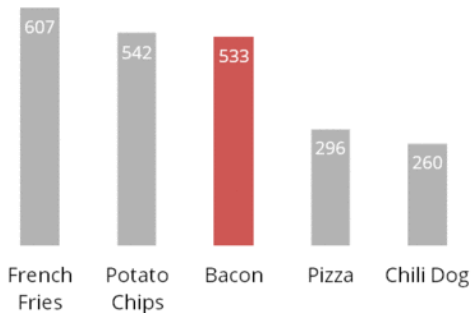


## Before

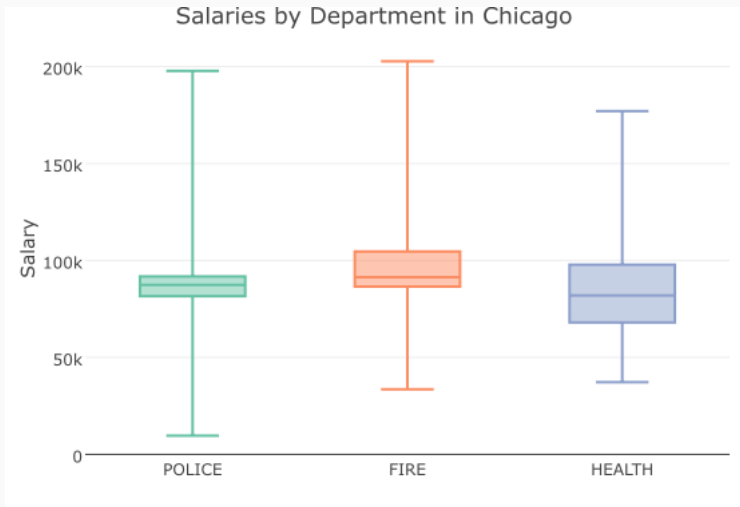


## After

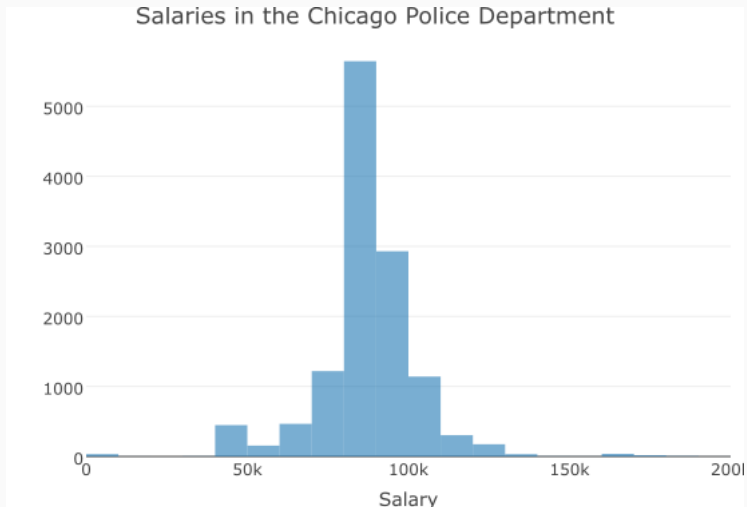
Calories per 100g



# Illustrating with Charts (Box and Whisker Chart)



## Illustrating with Charts (Histogram)

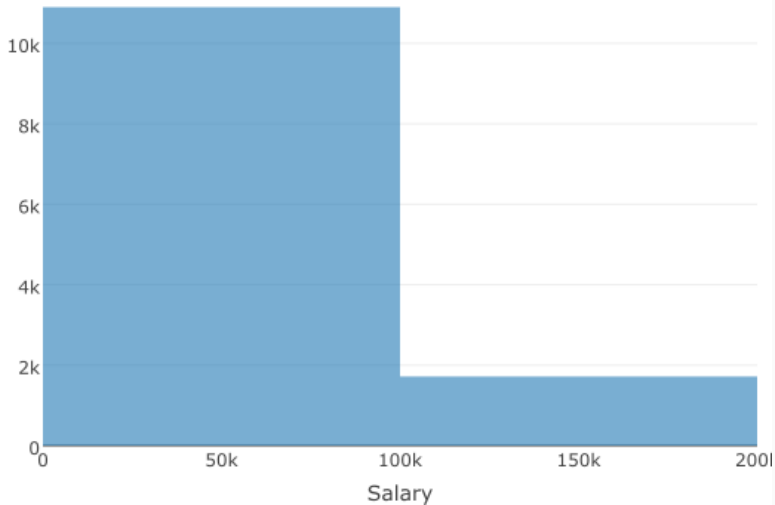


# Histograms are *Really* Important

1. Histograms show the frequency of different observations
2. **Important Choice:** How many bins?

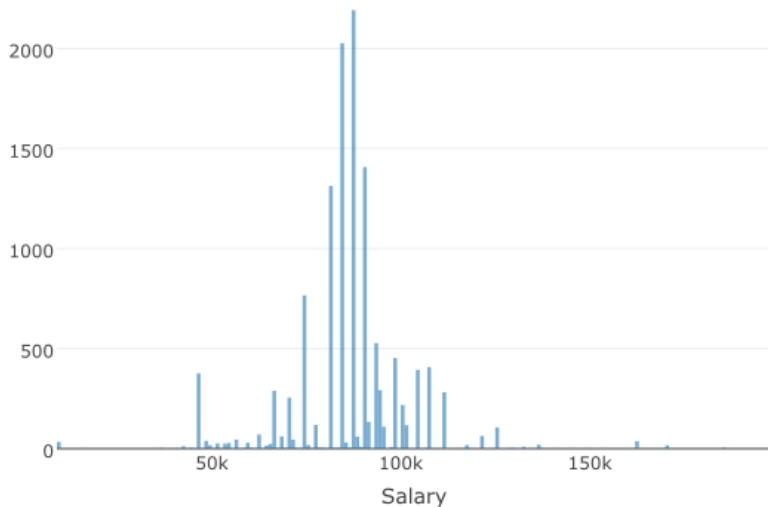
## Too Few Bins (Oversmoothing)

Salaries in the Chicago Police Department

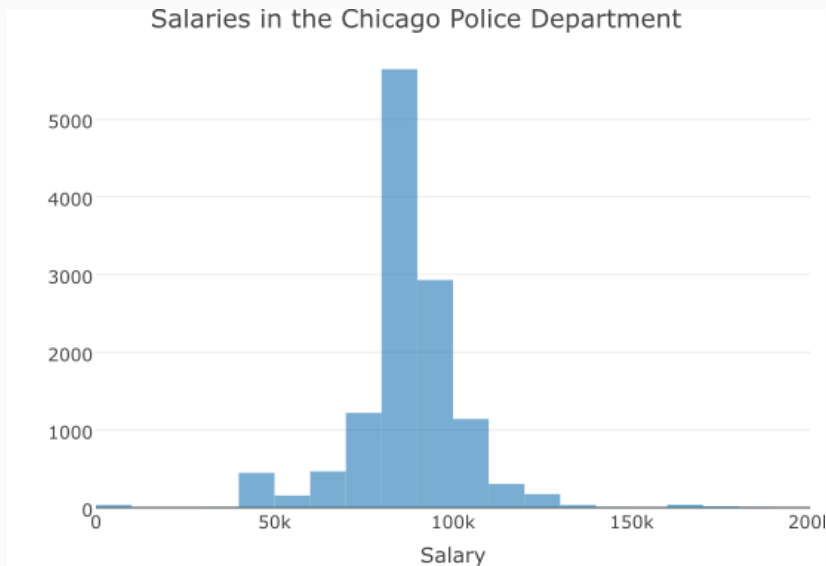


## Too Many Bins (Undersmoothing)

Salaries in the Chicago Police Department



## Just Right! (Usually around 20 bins or so)





# Questions to Ask Yourself about Each Summary Statistic

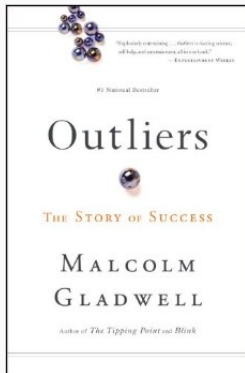
1. What does it measure?
2. What are its units compared to those of the data?
3. How do its units change if those of the data change?
4. What are the benefits and drawbacks of this statistic?

Some of the information regarding items 2 and 3 is on the homework rather than in the slides because working it out for yourself is a good way to check your understanding.

# Outliers

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# What is an Outlier?



**Outlier:** A very unusual observation relative to the other observations in the dataset (i.e. very small or very big).

## Which Summary Stats are Sensitive to Outliers?

- Assume our data is 1, 2, 3, 4, 5. What are our summary stats (mean, median, variance, range, IQR)
- What will be affected if the data includes an outlier and becomes 1, 2, 3, 4, 4990?

## Which Summary Stats are Sensitive to Outliers?

- Mean changes from 3 to 1000

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- When Does the Median Change? IQR?

# Which Summary Stats are Sensitive to Outliers?

- Mean changes from 3 to 1000
- Median remains at 3
- Variance changes from 2.5 to 4,975,032
- Range changes from 4 to 4889
- IQR remains at 2
- When Does the Median Change? IQR?
  - Ranks would have to change.

# Summary of Sensitivity

## Variance

Essentially the average squared distance from the mean.  
Sensitive to both skewness and outliers.

## Standard Deviation

$\sqrt{\text{Variance}}$ , but more convenient since same units as data

## Range

Difference between largest and smallest observations. *Very* sensitive to outliers.

## Interquartile Range

Range of middle 50% of the data. Insensitive to outliers, skewness.

# Sample vs. Population

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# Essential Distinction: Sample vs. Population

For now, you can think of the population as a list of  $N$  objects:

Population:  $x_1, x_2, \dots, x_N$

from which we draw a sample of size  $n < N$  objects:

Sample:  $x_1, x_2, \dots, x_n$

## Important Point:

Later in the course we'll be more formal by considering *probability models* that represent the *act of sampling* from a population rather than thinking of a population as a list of objects. Once we do this we will no longer use the notation  $N$  as the population will be *conceptually infinite*.

# Essential Distinction: Parameter vs. Statistic

$N$  individuals in the Population,  $n$  individuals in the Sample:

	Parameter (Population)	Statistic (Sample)
Mean	$\mu = \frac{1}{N} \sum_{i=1}^N x_i$	$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$
Var.	$\sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$	$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$
S.D.	$\sigma = \sqrt{\sigma^2}$	$s = \sqrt{s^2}$

## Key Point

We use a **sample**  $x_1, \dots, x_n$  to calculate **statistics** (e.g.  $\bar{x}$ ,  $s^2$ ,  $s$ ) that serve as **estimates** of the corresponding population **parameters** (e.g.  $\mu$ ,  $\sigma^2$ ,  $\sigma$ ).

## Why Do Sample Variance and Std. Dev. Divide by $n - 1$ ?

$$\sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$$
$$s^2 = \frac{1}{n - 1} \sum_{i=1}^n (x_i - \bar{x})^2$$

There is an important reason for this, but explaining it requires some concepts we haven't learned yet.



## Some Intuition

- **Intuition 1:** If we only had one data point, what would be the sample variance? Would it even be defined?
- **Intuition 2:** We know that the deviations from the sample mean sum to zero (see discussion of why variance is squared). Hence, we only need to know  $n - 1$  of the deviations since the last one will be whatever it takes to make the sum of them equal to 0. Hence, it would be proper to divide by  $n - 1$  instead of  $n$

# Z Scores

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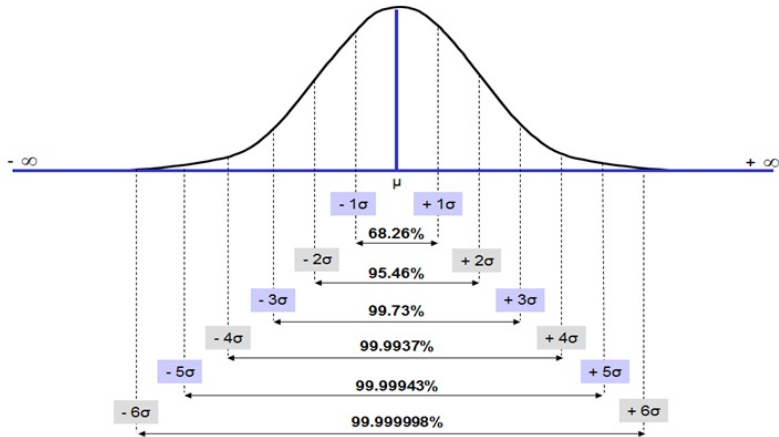
# Why Mean and Variance (and Std. Dev. )?

## Empirical Rule

For large populations that are approximately bell-shaped, standard deviation tells where most observations will be relative to the mean:

- $\approx 68\%$  of observations are in the interval  $\mu \pm \sigma$
- $\approx 95\%$  of observations are in the interval  $\mu \pm 2\sigma$
- Almost all of observations are in the interval  $\mu \pm 3\sigma$

# Standard Deviations



## Z-scores: How many standard deviations from the mean?

$$z_i = \frac{x_i - \bar{x}}{s}$$

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### Unitless

Allows comparison of variables with different units.

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### Detecting Outliers

Measures how “extreme” one observation is relative to the others.

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$$z_i = \frac{x_i - \bar{x}}{s}$$

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### Linear Transformation



What is the sample mean of the z-scores?

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$$\bar{z} = \frac{1}{n} \sum_{i=1}^n z_i = \frac{1}{n} \sum_{i=1}^n \frac{x_i - \bar{x}}{s} = \frac{1}{n \cdot s} \left[ \sum_{i=1}^n x_i - \sum_{i=1}^n \bar{x} \right]$$

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$$s_z^2 = \frac{1}{n-1} \sum_{i=1}^n (z_i - \bar{z})^2 = \frac{1}{n-1} \sum_{i=1}^n z_i^2 = \frac{1}{n-1} \sum_{i=1}^n \left( \frac{x_i - \bar{x}}{s_x} \right)^2$$

## What is the variance of the z-scores?

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So what is the *standard deviation* of the z-scores?



## Population Z-scores and the Empirical Rule: $\mu \pm 2\sigma$

If we knew the population mean  $\mu$  and standard deviation  $\sigma$  we could create a *population version* of a z-score. This leads to an important way of rewriting the Empirical Rule: **Bell-shaped**

population  $\Rightarrow$  approx. 95% of observations  $x_i$  satisfy

$$\mu - 2\sigma \leq x_i \leq \mu + 2\sigma$$

$$-2\sigma \leq x_i - \mu \leq 2\sigma$$

$$-2 \leq \frac{x_i - \mu}{\sigma} \leq 2$$

# Covariance and Correlation

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# Covariance and Correlation: Linear Dependence Measures

## Two Samples of Numeric Data

$x_1, \dots, x_n$  and  $y_1, \dots, y_n$

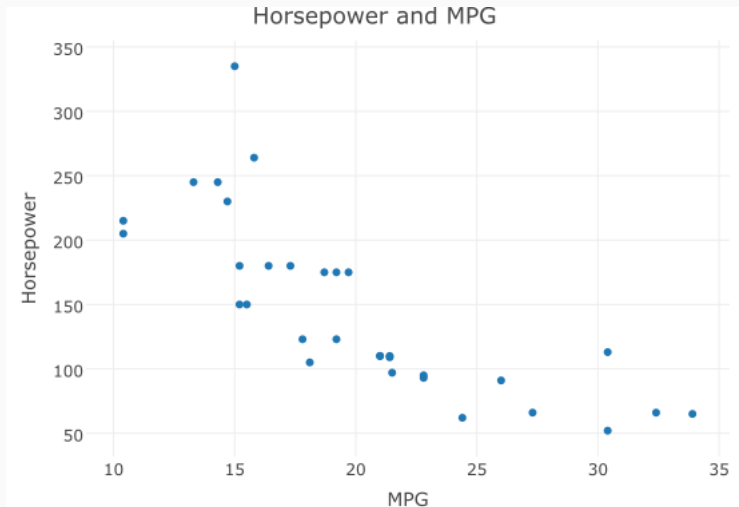
## Dependence

Do  $x$  and  $y$  both tend to be large (or small) at the same time?

## Key Point

Use the idea of centering and standardizing to decide what “big” or “small” means in this context.

# Are Engine Cylinders and Horsepower Related?



# Recall Formulas

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$$

$$s_x = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$$

$$s_y = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2}$$

# Covariance

$$s_{xy} = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

- Centers each observation around its mean and multiplies.
- Zero  $\Rightarrow$  no linear dependence
- Positive  $\Rightarrow$  positive linear dependence
- Negative  $\Rightarrow$  negative linear dependence
- Population parameter:  $\sigma_{xy}$
- Units?

# Correlation

$$r_{xy} = \frac{1}{n-1} \sum_{i=1}^n \left( \frac{x_i - \bar{x}}{s_x} \right) \left( \frac{y_i - \bar{y}}{s_y} \right) = \frac{s_{xy}}{s_x s_y}$$

- Centers *and* standardizes each observation
- Bounded between -1 and 1
- Zero  $\Rightarrow$  no linear dependence
- Positive  $\Rightarrow$  positive linear dependence
- Negative  $\Rightarrow$  negative linear dependence
- Population parameter:  $\rho_{xy}$
- Unitless

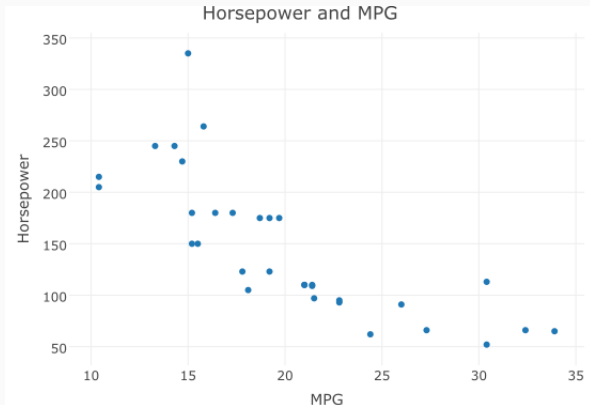
`guessthecorrelation.com`



# Introduction to Regression

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# Least Squares Regression – Predict Using a Line



- In order to fit a line through this, we need to estimate  $y = a + bx$
- How do we find  $a$  and  $b$ ?

## Finding $a$ and $b$

- Linear regression chooses the slope ( $b$ ) and intercept ( $a$ ) that minimize the sum of squared vertical deviations

$$\sum_{i=1}^n d_i^2 = \sum_{i=1}^n (y_i - a - bx_i)^2$$

- Why do we square the deviations?

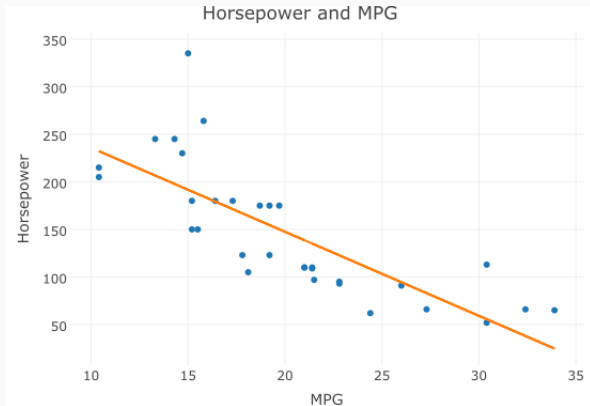
## Important Point About Notation

$$\underset{a,b}{\text{minimize}} \sum_{i=1}^n d_i^2 = \sum_{i=1}^n (y_i - a - bx_i)^2$$

$$\hat{y} = a + bx$$

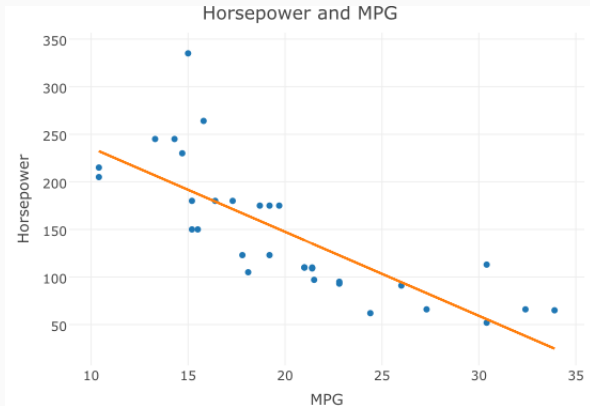
- $(x_i, y_i)_{i=1}^n$  are the **observed data**
- $\hat{y}$  is our **prediction** for a given value of  $x$
- Neither  $x$  nor  $\hat{y}$  needs to be in our dataset!

## Prediction 28 MPG?



$$\widehat{hp} = 324.08 - 8.83mpg$$

# Prediction 28 MPG?



$$\widehat{hp} = 324.08 - 8.83mpg$$

$$76.84 = 324.08 - 8.83 * 28$$

# You Need to Know How To Derive This

Minimize the sum of squared vertical deviations from the line:

$$\min_{a,b} \sum_{i=1}^n (y_i - a - bx_i)^2$$

How should we proceed? Select all that apply.

- (a) Differentiate with respect to  $x$
- (b) Differentiate with respect to  $y$
- (c) Differentiate with respect to  $a$
- (d) Differentiate with respect to  $b$
- (e) You can't fool me! You can't solve this with calculus.

## Objective Function

$$\min_{a,b} \sum_{i=1}^n (y_i - a - bx_i)^2$$

FOC with respect to  $a$



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$$-2 \sum_{i=1}^n (y_i - a - bx_i) = 0$$

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$$\sum_{i=1}^n y_i - \sum_{i=1}^n a - b \sum_{i=1}^n x_i = 0$$

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$$\frac{1}{n} \sum_{i=1}^n y_i - \frac{na}{n} - \frac{b}{n} \sum_{i=1}^n x_i = 0$$

## Objective Function

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$$\frac{1}{n} \sum_{i=1}^n y_i - \frac{na}{n} - \frac{b}{n} \sum_{i=1}^n x_i = 0$$

$$\bar{y} - a - b\bar{x} = 0$$

## Regression Line Goes Through the Means!

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

Substitute  $a = \bar{y} - b\bar{x}$

$$\sum_{i=1}^n (y_i - a - bx_i)^2 =$$

Substitute  $a = \bar{y} - b\bar{x}$

$$\begin{aligned}\sum_{i=1}^n (y_i - a - bx_i)^2 &= \sum_{i=1}^n (y_i - \bar{y} + b\bar{x} - bx_i)^2 \\ &= \end{aligned}$$

Substitute  $a = \bar{y} - b\bar{x}$

$$\begin{aligned}\sum_{i=1}^n (y_i - a - bx_i)^2 &= \sum_{i=1}^n (y_i - \bar{y} + b\bar{x} - bx_i)^2 \\ &= \sum_{i=1}^n [(y_i - \bar{y}) - b(x_i - \bar{x})]^2\end{aligned}$$

FOC wrt  $b$



Substitute  $a = \bar{y} - b\bar{x}$

$$\begin{aligned}\sum_{i=1}^n (y_i - a - bx_i)^2 &= \sum_{i=1}^n (y_i - \bar{y} + b\bar{x} - bx_i)^2 \\ &= \sum_{i=1}^n [(y_i - \bar{y}) - b(x_i - \bar{x})]^2\end{aligned}$$

FOC wrt  $b$

$$-2 \sum_{i=1}^n [(y_i - \bar{y}) - b(x_i - \bar{x})] (x_i - \bar{x}) = 0$$

Substitute  $a = \bar{y} - b\bar{x}$

$$\begin{aligned}\sum_{i=1}^n (y_i - a - bx_i)^2 &= \sum_{i=1}^n (y_i - \bar{y} + b\bar{x} - bx_i)^2 \\ &= \sum_{i=1}^n [(y_i - \bar{y}) - b(x_i - \bar{x})]^2\end{aligned}$$

FOC wrt  $b$

$$\begin{aligned}-2 \sum_{i=1}^n [(y_i - \bar{y}) - b(x_i - \bar{x})] (x_i - \bar{x}) &= 0 \\ \sum_{i=1}^n (y_i - \bar{y}) (x_i - \bar{x}) - b \sum_{i=1}^n (x_i - \bar{x})^2 &= 0\end{aligned}$$

Substitute  $a = \bar{y} - b\bar{x}$

$$\begin{aligned}\sum_{i=1}^n (y_i - a - bx_i)^2 &= \sum_{i=1}^n (y_i - \bar{y} + b\bar{x} - bx_i)^2 \\ &= \sum_{i=1}^n [(y_i - \bar{y}) - b(x_i - \bar{x})]^2\end{aligned}$$

FOC wrt  $b$

$$-2 \sum_{i=1}^n [(y_i - \bar{y}) - b(x_i - \bar{x})] (x_i - \bar{x}) = 0$$

$$\sum_{i=1}^n (y_i - \bar{y}) (x_i - \bar{x}) - b \sum_{i=1}^n (x_i - \bar{x})^2 = 0$$

$$b = \frac{\sum_{i=1}^n (y_i - \bar{y}) (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

# Simple Linear Regression

## Problem

$$\min_{a,b} \sum_{i=1}^n (y_i - a - bx_i)^2$$

## Solution

$$b = \frac{\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$a = \bar{y} - b\bar{x}$$

## Relating Regression to Covariance and Correlation

$$b = \frac{\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2} = \frac{s_{xy}}{s_x^2}$$

$$r = \frac{s_{xy}}{s_x s_y} = b \frac{s_x}{s_y}$$

# Comparing Regression, Correlation and Covariance

## Units

Correlation is unitless, covariance and regression coefficients  $(a, b)$  are not. (What are the units of these?)

## Symmetry

Correlation and covariance are symmetric, regression isn't. (Switching  $x$  and  $y$  axes changes the slope and intercept.)

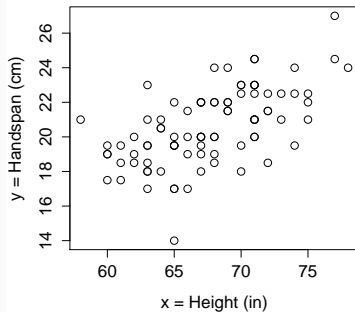
## On the Homework

Regression with z-scores rather than raw data gives

$$a = 0, b = r_{xy}$$

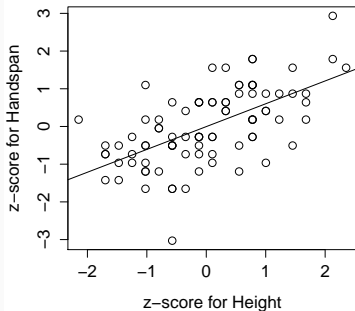
$$s_{xy} = 6, \quad s_x = 5, \quad s_y = 2, \quad \bar{x} = 68, \quad \bar{y} = 21$$

What is the sample correlation between height ( $x$ ) and handspan ( $y$ )?



$$s_{xy} = 6, \quad s_x = 5, \quad s_y = 2, \quad \bar{x} = 68, \quad \bar{y} = 21$$

What is the sample correlation between height (x) and handspan (y)?



$$r = \frac{s_{xy}}{s_x s_y} = \frac{6}{5 \times 2} = 0.6$$

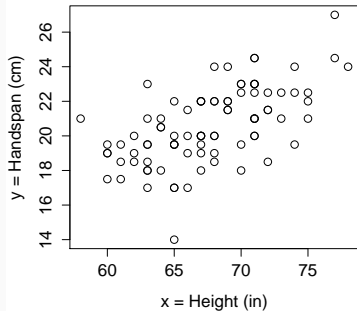


$$s_{xy} = 6, \quad s_x = 5, \quad s_y = 2, \quad \bar{x} = 68, \quad \bar{y} = 21$$

What is the value of  $b$  for the regression:

$$\hat{y} = a + bx$$

where  $x$  is height and  $y$  is handspan?

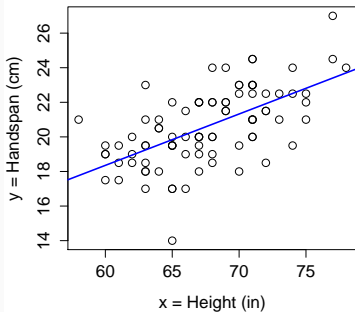


$$s_{xy} = 6, \quad s_x = 5, \quad s_y = 2, \quad \bar{x} = 68, \quad \bar{y} = 21$$

What is the value of  $b$  for the regression:

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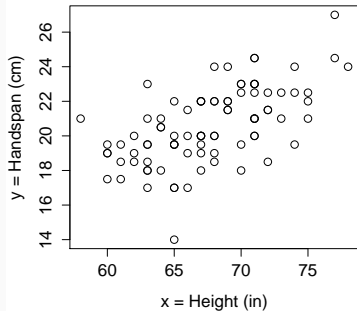
$$b = \frac{s_{xy}}{s_x^2} = \frac{6}{5^2} = 6/25 = 0.24$$

$$s_{xy} = 6, \quad s_x = 5, \quad s_y = 2, \quad \bar{x} = 68, \quad \bar{y} = 21$$

What is the value of  $a$  for the regression:

$$\hat{y} = a + bx$$

where  $x$  is height and  $y$  is handspan?  
(prev. slide  $b = 0.24$ )

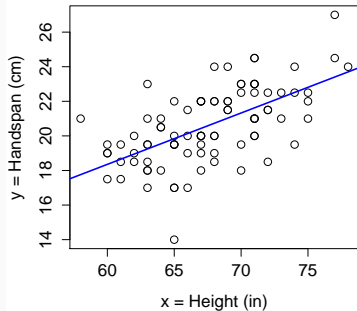


$$s_{xy} = 6, \quad s_x = 5, \quad s_y = 2, \quad \bar{x} = 68, \quad \bar{y} = 21$$

What is the value of  $a$  for the regression:

$$\hat{y} = a + bx$$

where  $x$  is height and  $y$  is handspan?  
(prev. slide  $b = 0.24$ )



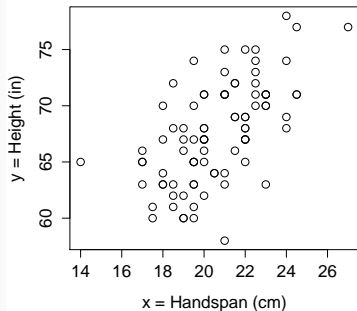
$$a = \bar{y} - b\bar{x} = 21 - 0.24 \times 68 = 4.68$$

$$s_{xy} = 6, \quad s_y = 5, \quad s_x = 2, \quad \bar{y} = 68, \quad \bar{x} = 21$$

What is the value of  $b$  for the regression:

$$\hat{y} = a + bx$$

where  $x$  is handspan and  $y$  is height?

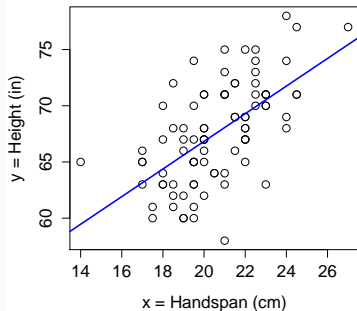


$$s_{xy} = 6, \quad s_y = 5, \quad s_x = 2, \quad \bar{y} = 68, \quad \bar{x} = 21$$

What is the value of  $b$  for the regression:

$$\hat{y} = a + bx$$

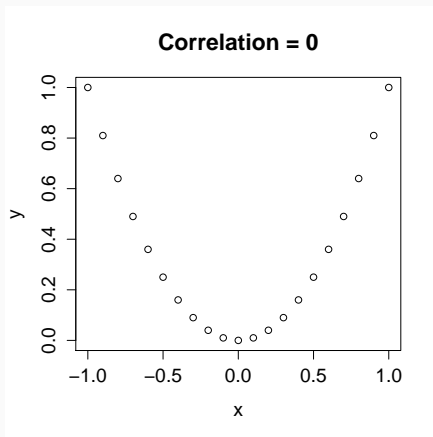
where  $x$  is handspan and  $y$  is height?



$$b = \frac{s_{xy}}{s_x^2} = 6/2^2 = 1.5$$

## EXTREMELY IMPORTANT

- Regression, Covariance and Correlation: linear association.
- Linear association  $\neq$  causation.
- Linear is not the only kind of association!

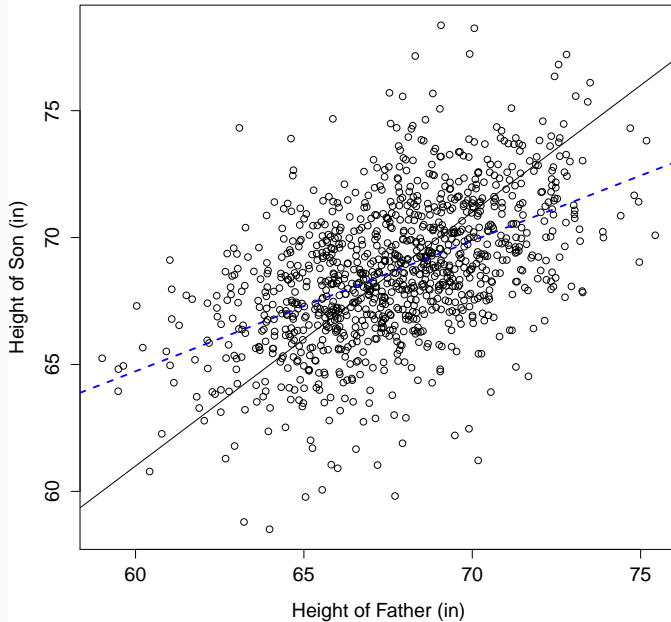


# Regression to the Mean and the Regression Fallacy

Please read Chapter 17 of “Thinking Fast and Slow” by Daniel Kahnemann which I have posted on Piazza. This reading is fair game on an exam or quiz.



## Pearson Dataset



# Regression to the Mean

Skill and Luck / Genes and Random Environmental Factors

Unless  $r_{xy} = 1$ , There Is Regression to the Mean

$$\frac{\hat{y} - \bar{y}}{s_y} = r_{xy} \frac{x - \bar{x}}{s_x}$$

Least-squares Prediction  $\hat{y}$  closer to  $\bar{y}$  than  $x$  is to  $\bar{x}$

You will derive the above formula in this week's homework.

## Next Up: Basic Probability

Please do the following before our next class:

1. Complete the “Odd Questions” quiz posted on Piazza **OddQuestions.pdf** – we’ll be discussing these in class.
2. If you’re rusty on permutations, combinations, etc. from High School math, read this review <http://ditraglia.com/Econ103Public/ClassicalProbability.pdf>

# Review

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# Essential Distinction: Parameter vs. Statistic

$N$  individuals in the Population,  $n$  individuals in the Sample:

	Parameter (Population)	Statistic (Sample)
Mean	$\mu_x = \frac{1}{N} \sum_{i=1}^N x_i$	$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$
Var.	$\sigma_x^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$	$s_x^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$
S.D.	$\sigma_x = \sqrt{\sigma_x^2}$	$s_x = \sqrt{s_x^2}$
Cov.	$\sigma_{xy} = \frac{\sum_{i=1}^N (x_i - \mu_x)(y_i - \mu_y)}{N}$	$s_{xy} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{n-1}$
Corr.	$\rho = \frac{\sigma_{xy}}{\sigma_x \sigma_y}$	$r = \frac{s_{xy}}{s_x s_y}$

## Related Reading

- Wonnacott: Chapter 2, Section 4-5 A+B, Section 5-3, Appendix 2-2, and Appendix 2-5
- How to Lie with Statistics: Chapters 2, 5, and 6