

**YaleNUSCollege**

# YSC2239 Lecture 10

# Recap

- A/B testing
- Confidence Intervals
- Significant level (also called alpha level)

## Python command

- Percentile

# Today's class

- Central and Spread
  - Central limit theorem
  - Correlation
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- Reading: Chapter 14, 15

# Confidence Intervals For Testing

# Using a CI for Testing

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*What if we want to do a hypothesis test, but we can't simulate under the null?*

- Null hypothesis: **Population average =  $x$**
  - Alternative hypothesis: **Population average  $\neq x$**
  - Cutoff for P-value:  $p\%$
  - Method:
    - Construct a  $(100-p)\%$  confidence interval for the population average
    - If  $x$  is not in the interval, reject the null
    - If  $x$  is in the interval, can't reject the null
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# Center and Spread

# Questions

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- How can we quantify natural concepts like “center” and “variability”?
  - Why do many of the empirical distributions that we generate come out bell shaped?
  - How is sample size related to the accuracy of an estimate?
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Average



# The Average (or Mean)

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Data: 2, 3, 3, 9    **Average =  $(2+3+3+9)/4 = 4.25$**

- Need not be a value in the collection
- Need not be an integer even if the data are integers
- Somewhere between min and max, but not necessarily halfway in between
- Same units as the data
- Smoothing operator: collect all the contributions in one big pot, then split evenly

(Demo)

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# Comparing Mean and Median

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- **Mean:** Balance point of the histogram
  - **Median:** Half-way point of data; half the area of histogram is on either side of median
  - If the distribution is symmetric about a value, then that value is both the average and the median.
  - If the histogram is skewed, then the mean is pulled away from the median in the direction of the tail.
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# Standard Deviation

# Defining Variability

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**Plan A:** “biggest value - smallest value”

- Doesn't tell us much about the shape of the distribution

**Plan B:**

- Measure variability around the mean
- Need to figure out a way to quantify this

(Demo)

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# How Far from the Average?

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- Standard deviation (SD) measures roughly how far the data are from their average
  - $SD = \text{root mean square of deviations from average}$
  - SD has the same units as the data
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# Why Use the SD?

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There are two main reasons.

- **The first reason:**

No matter what the shape of the distribution,  
the bulk of the data are in the range “average  $\pm$  a few SDs”

- **The second reason:**

Coming up in the next lecture.

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# Standard Units

# Standard Units

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- How many SDs above average?
- $z = (\text{value} - \text{average})/\text{SD}$ 
  - Negative  $z$ : value below average
  - Positive  $z$ : value above average
  - $z = 0$ : value equal to average
- When values are in standard units: average = 0, SD = 1
- Chebyshev: At least 96% of the values of  $z$  are between -5 and 5 ( i.e.: average - 5\*SD , average + 5\*SD)

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(Demo)



# Discussion Question

Find whole numbers  
that are close to:

(a) the average age

(b) the SD of the ages

(Demo)

Age in Years	Age in Standard Units
27	-0.0392546
33	0.992496
28	0.132704
23	-0.727088
25	-0.383171
33	0.992496
23	-0.727088
25	-0.383171
30	0.476621
27	-0.0392546

... (1164 rows omitted)

# The SD and the Histogram

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- Usually, it's not easy to estimate the SD by looking at a histogram.
- But if the histogram has a bell shape, then you can.

# The SD and Bell-Shaped Curves

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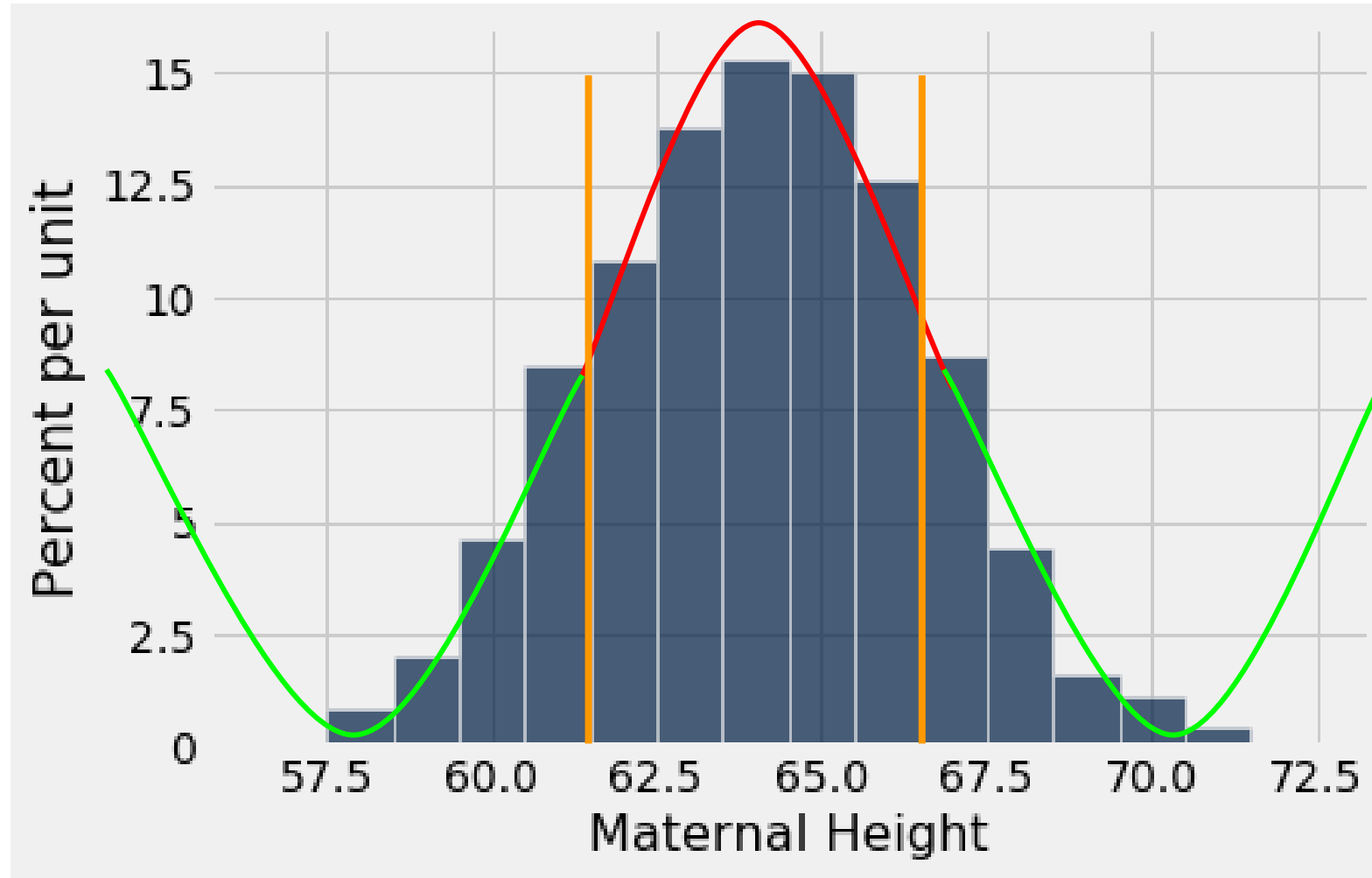
If a histogram is bell-shaped, then

- the average is at the center
- the SD is the distance between the average and the points of inflection on either side

(Demo)

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# Point of Inflection



# The Normal Distribution

# The Standard Normal Curve

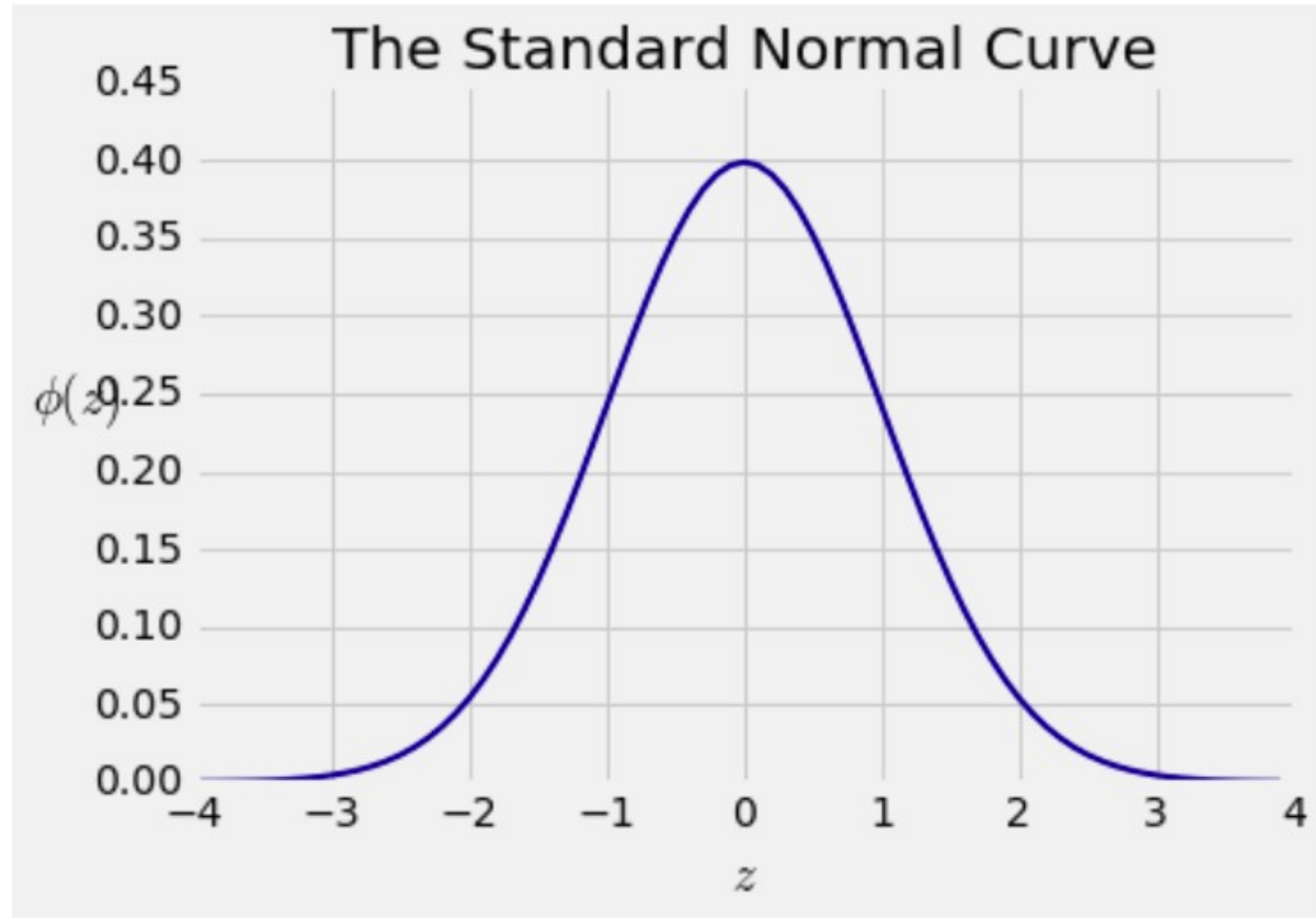
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A beautiful formula that we won't use at all:

$$\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}, \quad -\infty < z < \infty$$

# Bell Curve

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# Normal Proportions



# How Big are Most of the Values?

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*No matter what the shape of the distribution,*  
the bulk of the data are in the range “average  $\pm$  a few SDs”

*If a histogram is bell-shaped,* then

- Almost all of the data are in the range  
“average  $\pm$  3 SDs”

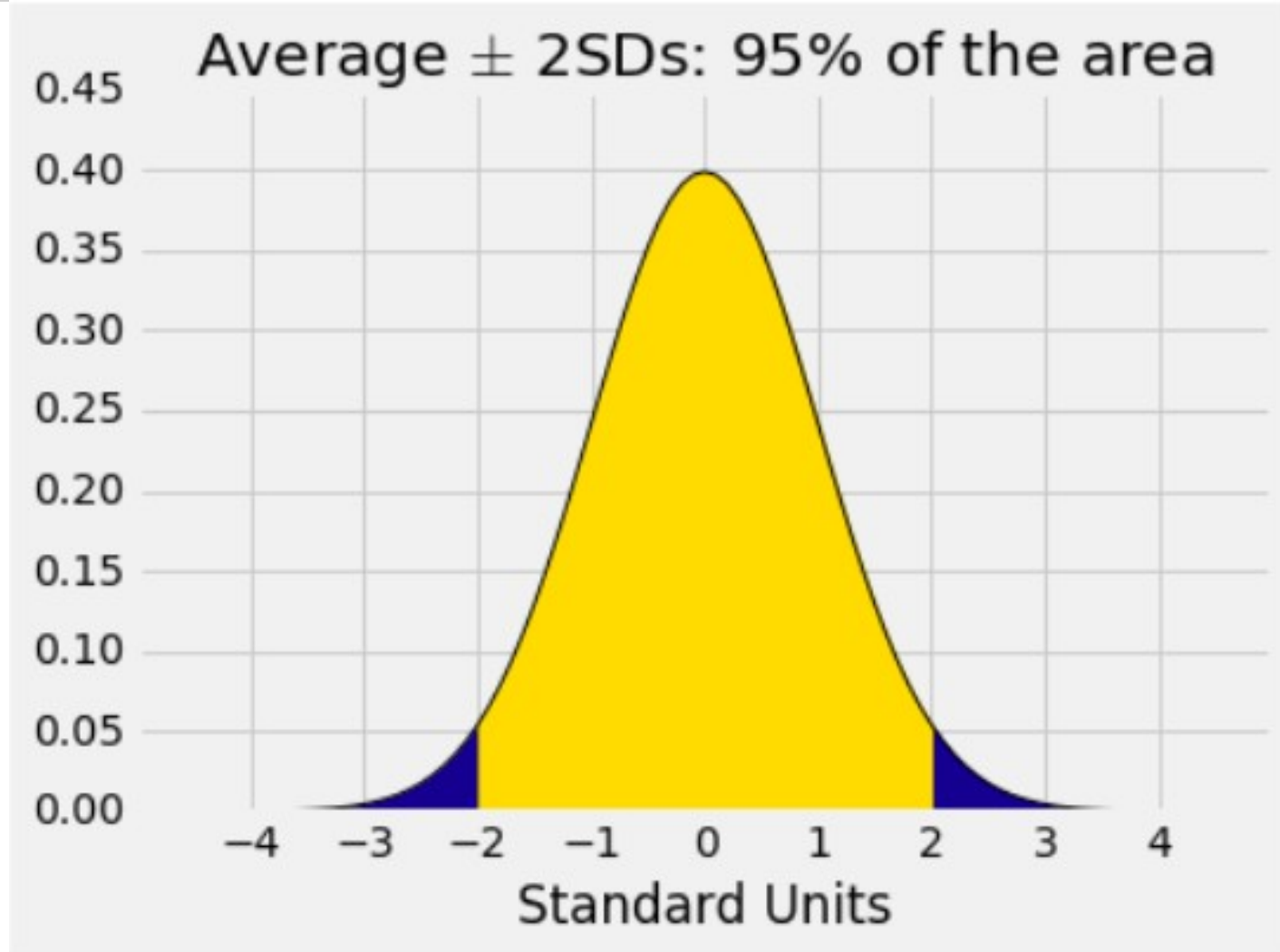
# Bounds and Normal Approximations

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<b>Percent in Range</b>	<b>All Distributions</b>	<b>Normal Distribution</b>
average $\pm$ 1 SD	at least 0%	about 68%
average $\pm$ 2 SDs	at least 75%	about 95%
average $\pm$ 3 SDs	at least 88.888...%	about 99.73%

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# A “Central” Area



# Central Limit Theorem

# Sample Averages

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- The Central Limit Theorem describes how the normal distribution (a bell-shaped curve) is connected to random sample averages.
  - We care about sample averages because they estimate population averages.
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# Central Limit Theorem

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If the sample is

- large, and
- drawn at random with replacement,

Then, *regardless of the distribution of the population,*

**the probability distribution of the sample sum  
(or the sample average) is roughly normal**

(Demo)

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# Distribution of the Sample Average

# Why is There a Distribution?

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- You have only one random sample, and it has only one average.
  - But **the sample could have come out differently**.
  - And then the sample average might have been different.
  - So there are many possible sample averages.
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# Distribution of the Sample Average

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- Imagine all possible random samples of the same size as yours. There are lots of them.
- Each of these samples has an average.
- The **distribution of the sample average** is the distribution of the averages of all the possible samples.

(Demo)

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# Specifying the Distribution

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Suppose the random sample is large.

- We have seen that the distribution of the sample average is roughly bell shaped.
  - Important questions remain:
    - Where is the center of that bell curve?
    - How wide is that bell curve?
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Center of the Distribution

# The Population Average

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The distribution of the sample average is roughly a bell curve centered at the population average.

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# Variability of the Sample Average

# Why Is This Important?

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- Along with the center, the spread helps identify exactly which normal curve is the distribution of the sample average.
- The variability of the sample average helps us measure how accurate the sample average is as an estimate of the population average.
- If we want a specified level of accuracy, understanding the variability of the sample average helps us work out how large our sample has to be.

(Demo)

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# Variability of the Sample Average

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- The distribution of all possible sample averages of a given size is called the *distribution of the sample average*.
- We approximate it by an empirical distribution.
- By the CLT, it's roughly normal:
  - Center = the population average
  - $SD = (\text{population SD}) / \sqrt{\text{sample size}}$

(Demo)

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# Discussion Question

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A city has 500,000 households. The annual incomes of these households have an average of \$65,000 and an SD of \$45,000. The distribution of the incomes [\[pick one and explain\]](#):

- (a) is roughly normal because the number of households is large.
  - (b) is not close to normal.
  - (c) may be close to normal, or not; we can't tell from the information given.
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# Correlation Coefficient

# Definition of $r$

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**Correlation Coefficient ( $r$ ) =**

average of	product of	x in standard units	and	y in standard units
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Measures how clustered the scatter is around a straight line

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# The Correlation Coefficient $r$

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- Measures **linear** association
- Based on standard units
- $-1 \leq r \leq 1$ 
  - $r = 1$ : scatter is perfect straight line sloping up
  - $r = -1$ : scatter is perfect straight line sloping down
- $r = 0$ : No linear association; *uncorrelated*

(Demo)

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# Watch Out For ...

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- Nonlinearity
- Outliers
- Correlation does not imply causations (  
<https://www.tylervigen.com/spurious-correlations>)

# To-do

- Lab 5
- Assignment 5