

Sampling distributions,  
standard errors, and confidence intervals

# Overview

Review of sampling bias

Sampling bias and sampling distributions

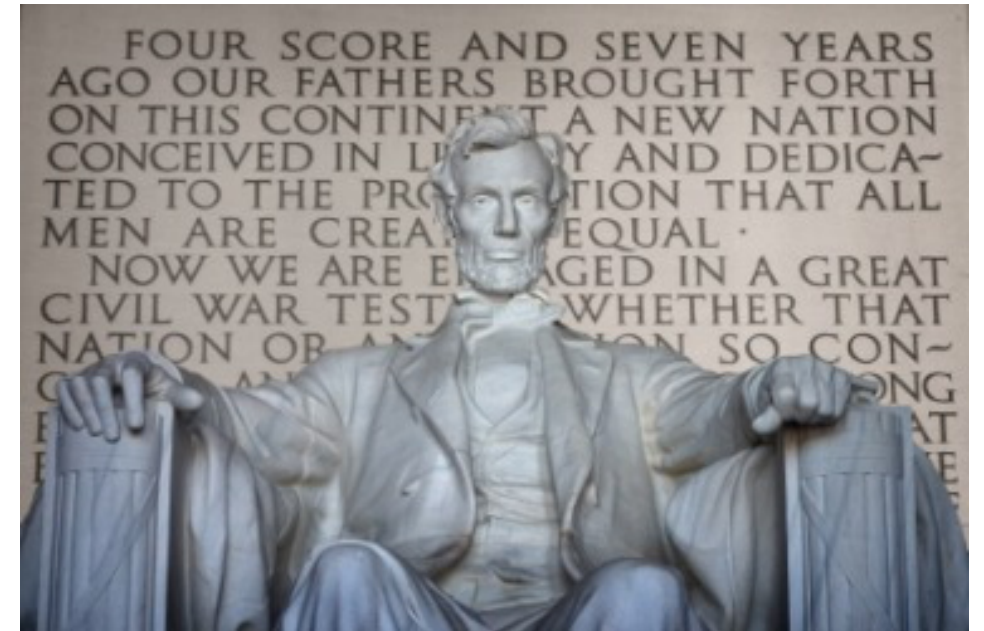
More on sampling distributions and the Standard Error

Point estimates and confidence intervals

# Review: sampling



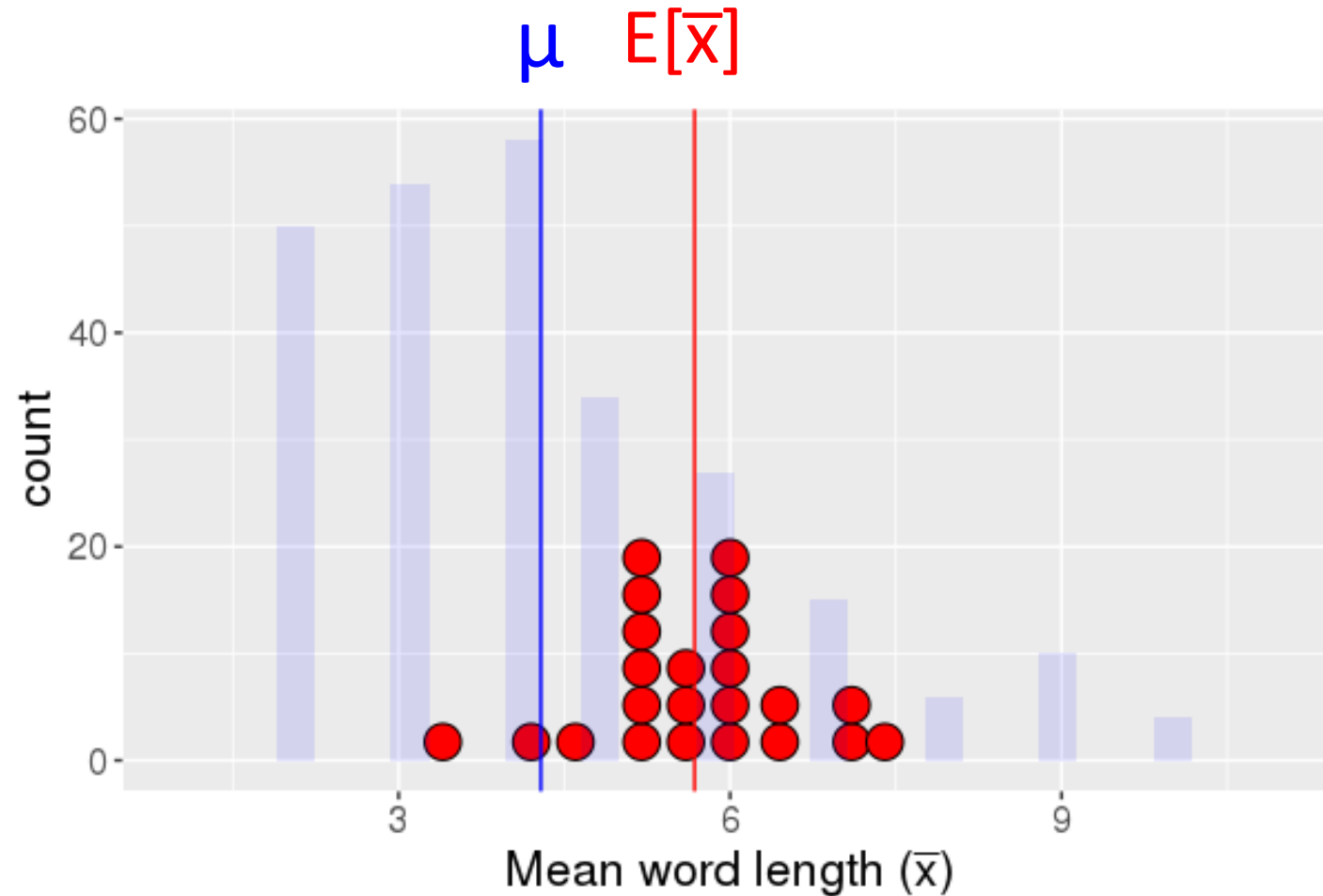
1	orange
2	red
3	green
4	white
5	white
6	white
7	white
8	white
9	red



Q: What symbol do we use to denote the sample size?

A: ***n***

# Bias and the Gettysburg address word length distribution

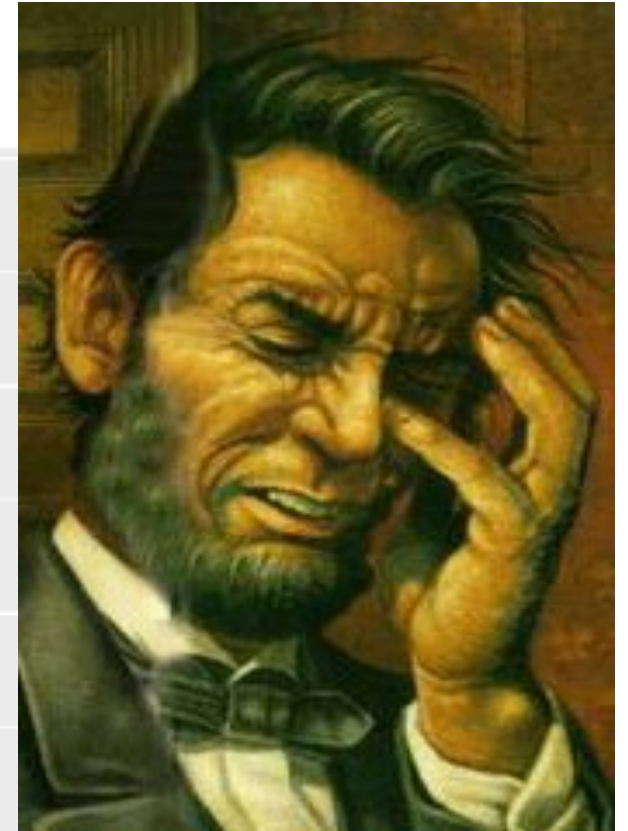
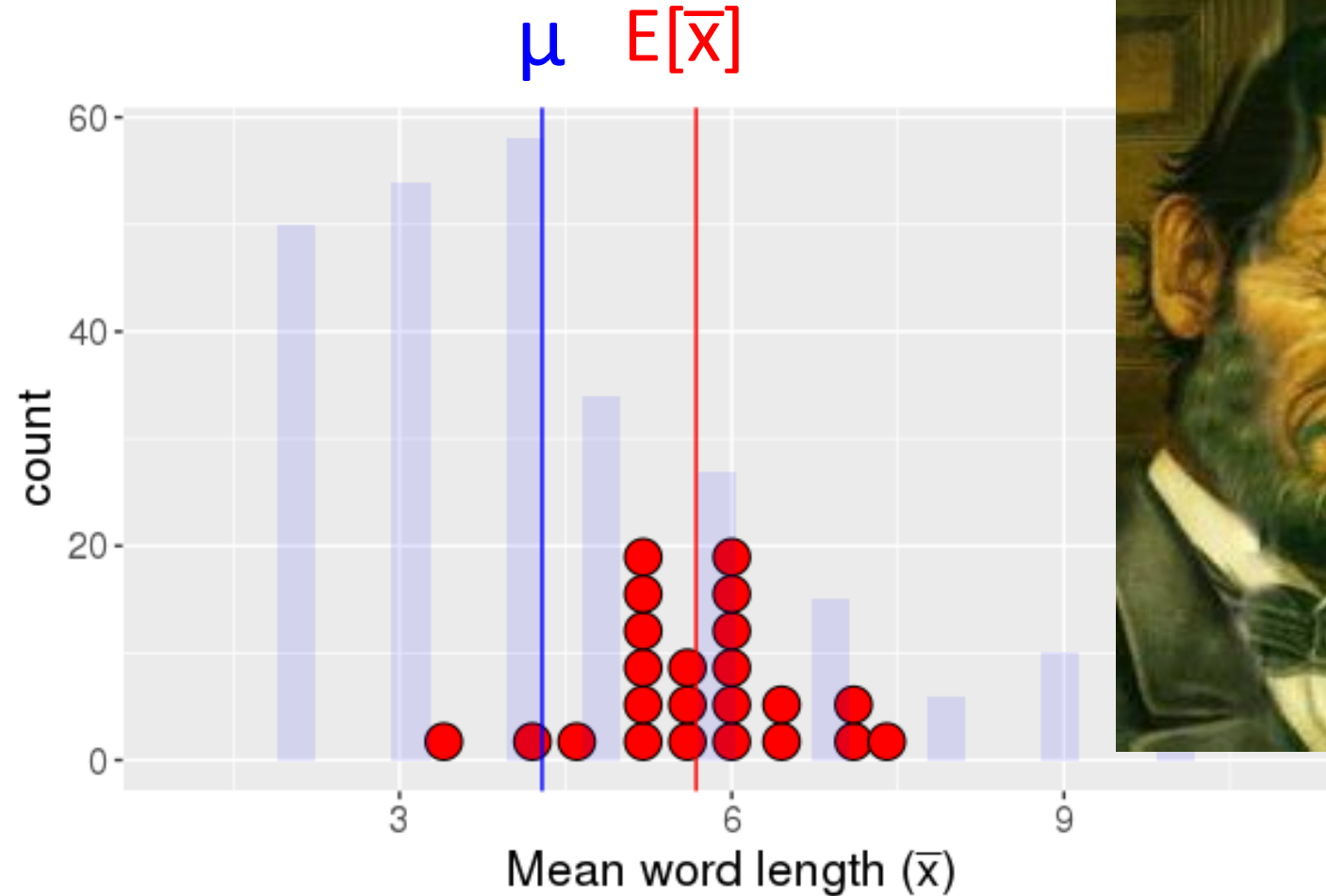


# Bias and the Gettysburg address word length distribution

**Bias** is when our average statistic does not equal the population parameter

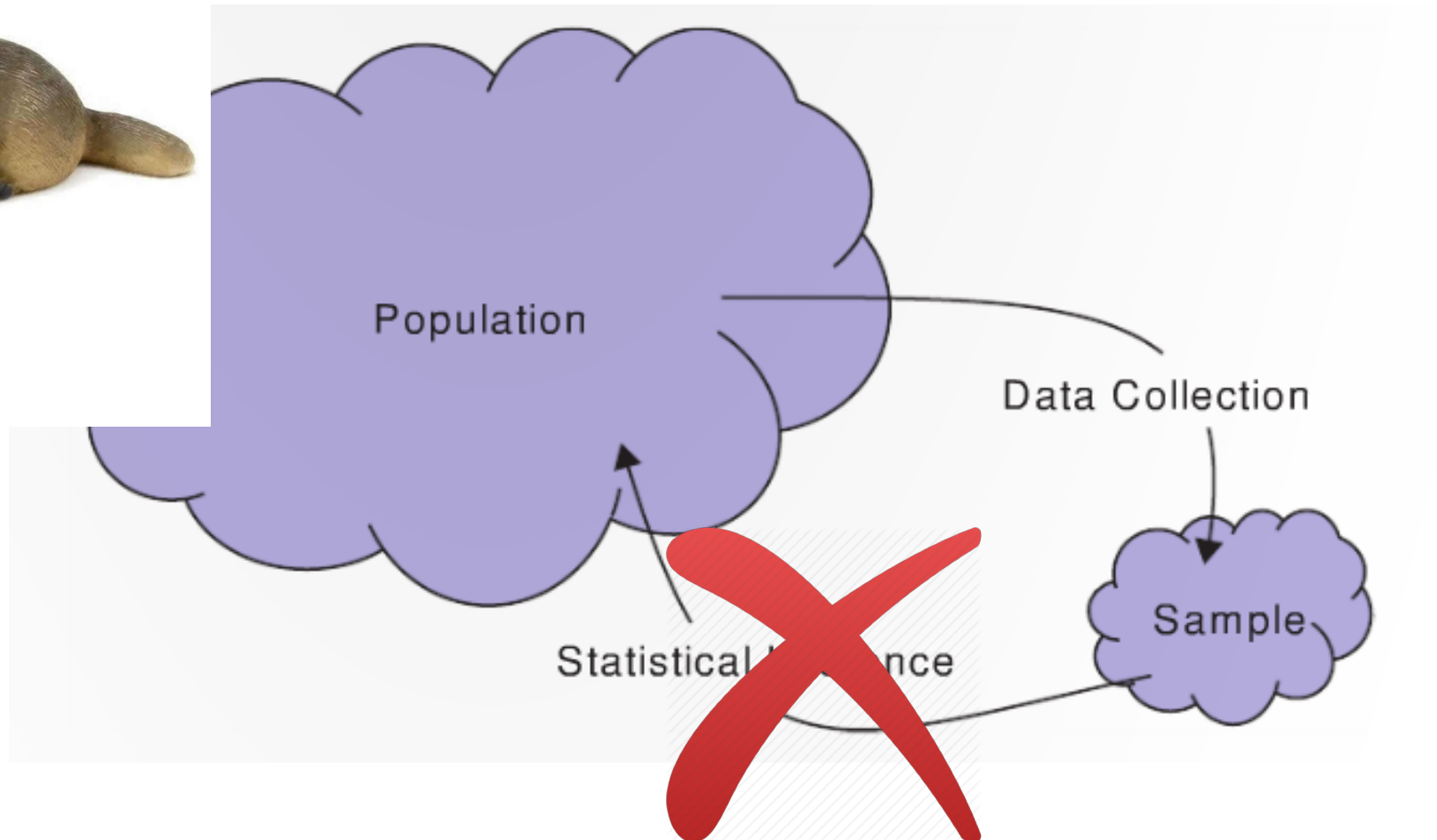
Here:

$$E[\bar{x}] \neq \mu$$

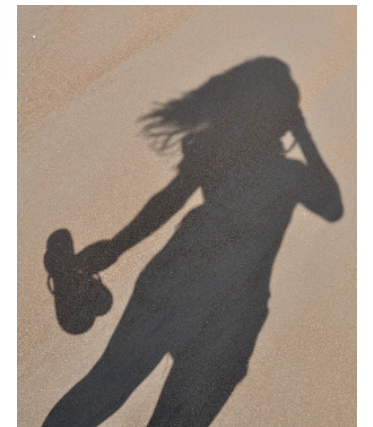


# Statistical bias

$\mu$



$\bar{x}$



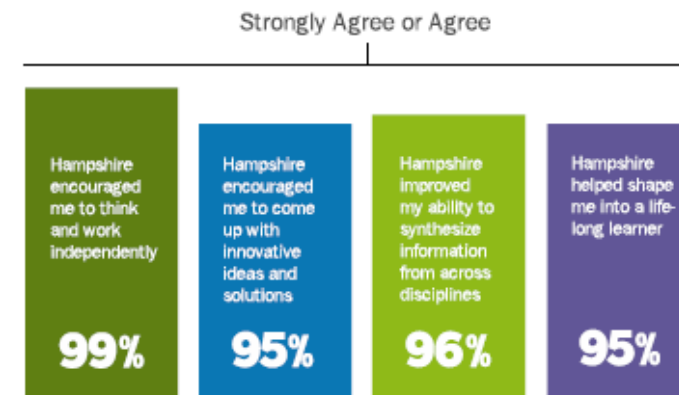
## Bias or No Bias?

As part of a strategic-planning process, in spring 2013 Hampshire College launched a survey of alums. Via email, the College **invited 8,160 alums to fill out an online questionnaire** administered by the campus's offices. **A total of 1,920 surveys were completed, yielding a response rate of 24%.**

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Note: The percentages in the data (below) are based on the number of responses received for each question.

**To what extent do you agree with the following statements?**



Please rate your student experience at Hampshire.



**65%** of our alumni earn advanced degrees within ten years of graduating.

**1 in 7** alumni holds a Ph.D. or other terminal degree.

Hampshire ranks in the **top 1%** of colleges nationwide in the % of grads that go on to earn doctorates.

**26%** of our graduates have started their own business or organization.

“

**Hampshire does a great job fostering the ability to ask good questions and to look at ideas with a critical lens.**

**Hampshire has encouraged me to be more engaged, socially aware and more of a critical thinker than my peers.**

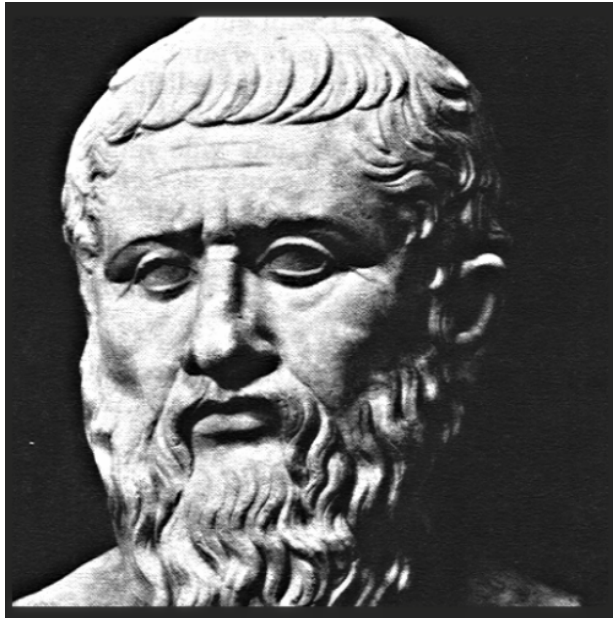
**I feel more able to adapt to a range of environments because Hampshire taught me skills and ideas rather than just knowledge.**

”



# Bias or No Bias?

$$\pi_{\text{replied}} \neq \pi_{\text{all}}$$



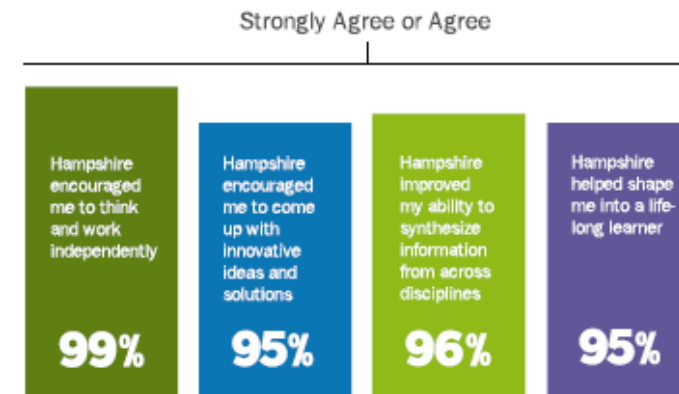
Sad Plato says:  
"There's no Truth in advertising"

## Alumni Survey Results

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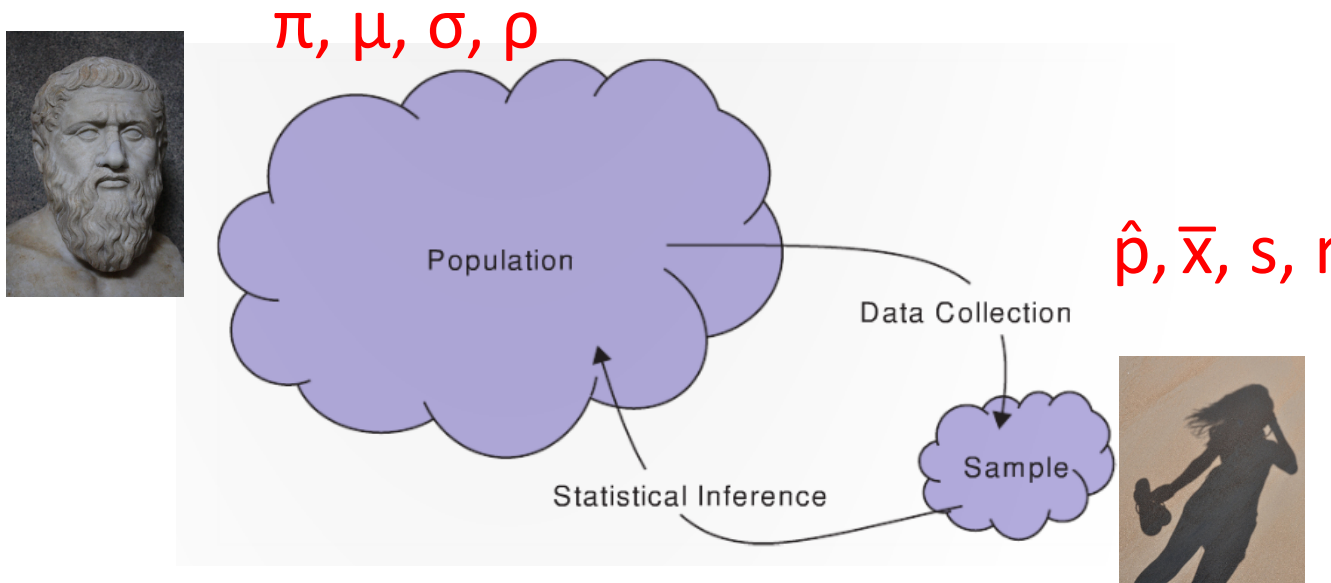
# Q: How can we prevent bias?

A: To prevent bias, use a **simple random sample**

- where each member in the population is equally likely to be in the sample

This allows for generalizations to the population!

Soup analogy!



# Q: How do we select a random sample?

Mechanically:

- Flip coins

- Pull balls from well mixed bins

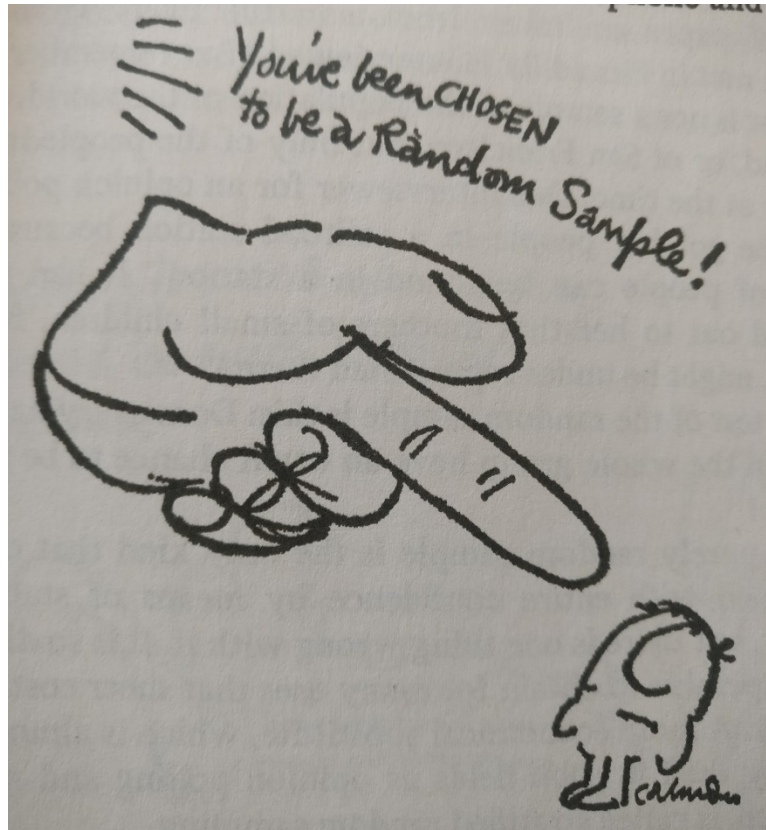
- Deal out shuffled cards, etc.

Use computer programs

Q: What computer program can we use?

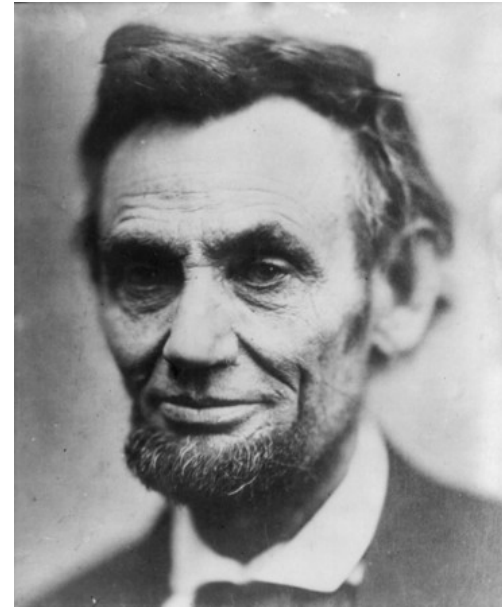
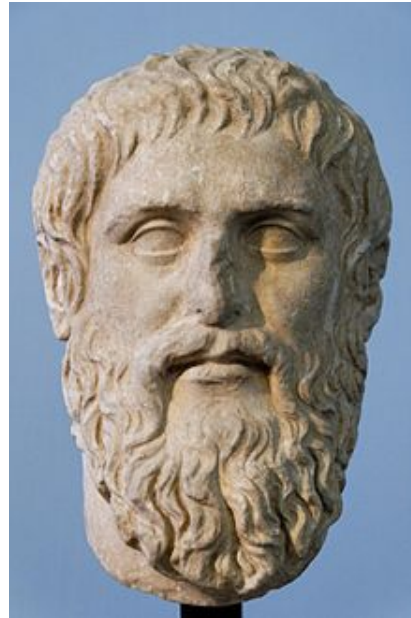


# Questions about statistical bias?



# From now on we are going to assume no bias!

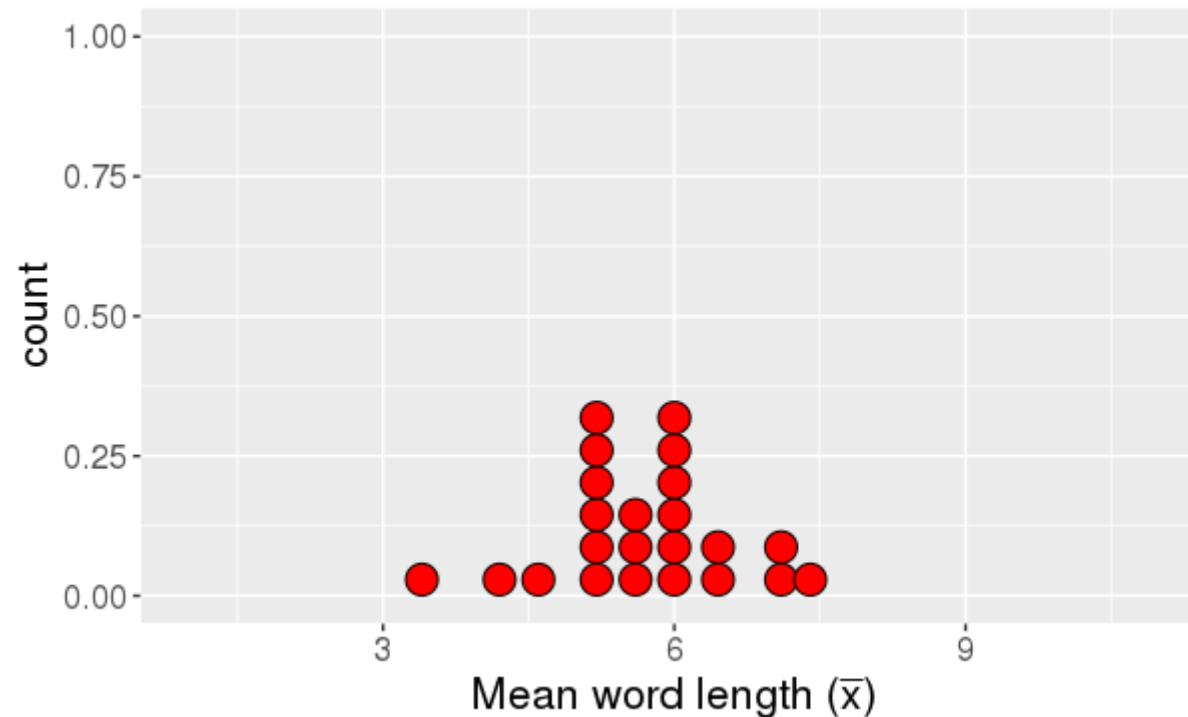
Happy Plato and Lincoln



statistics, on average, reflect the parameters

# For our distribution of Gettysburg word lengths...

Q: What does each case that is plotted correspond to?



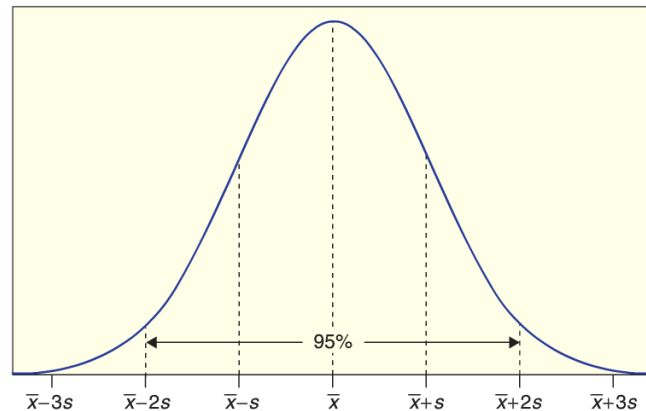
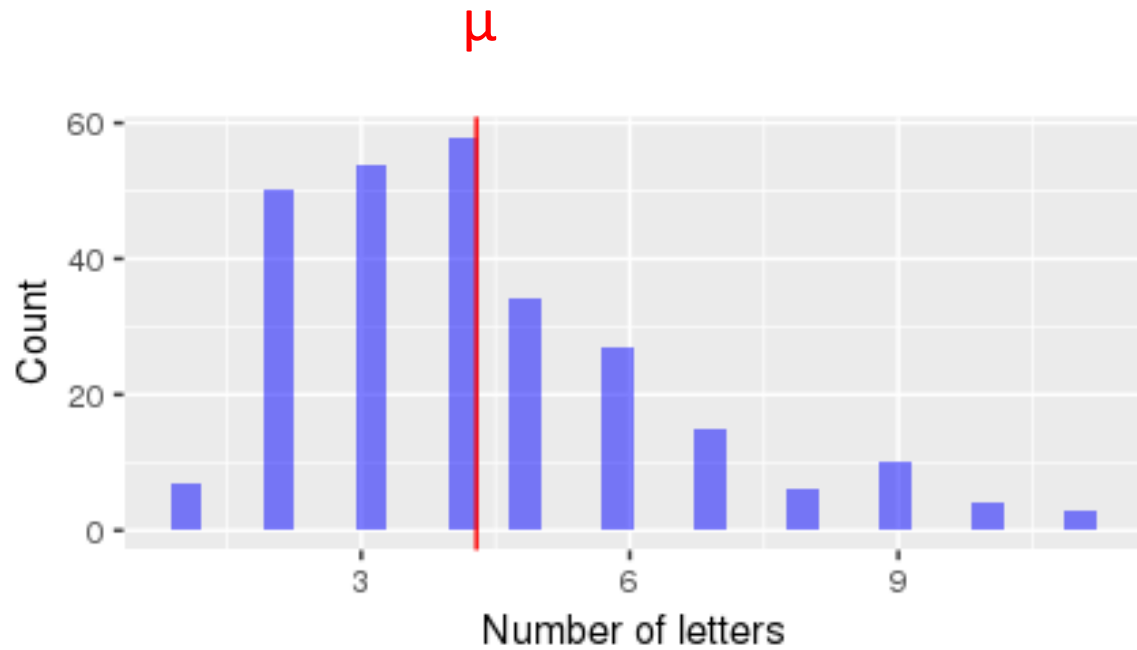
A: The mean length of 10 words ( $\bar{x}$ )  
i.e., each point in our **distribution** is a statistic!

# Sampling distribution

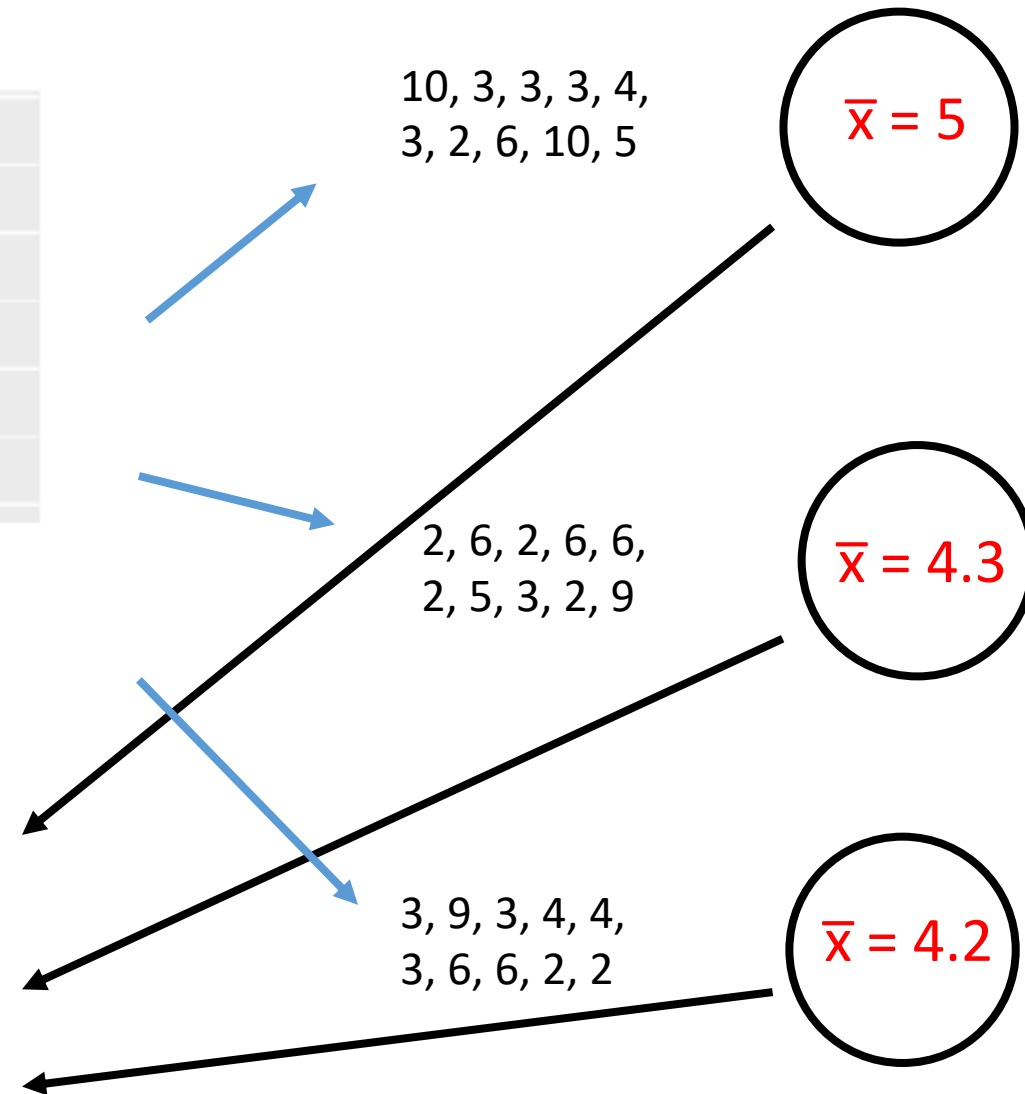
A **sampling distribution** is the distribution of sample statistics computed for different samples of the same size ( $n$ ) from the same population

A sampling distribution shows us how the sample statistic varies from sample to sample

# Gettysburg address word length sampling distribution



Sampling distribution!



[Gettysburg sampling distribution app](#)



# Let's create a sampling distribution in R

Log into Class workspace 2 – link is on Canvas

- Link is on Canvas
- > `library(ClassTools)`

Get the Gettysburg population data

```
> download_class_data("gettysburg.Rda")  
> load("gettysburg.Rda")  
> word_lengths <- gettysburg$num_letters
```

# Let's create a sampling distribution in R

We can use the `sample(data_vec, n)` to get a sample of length `n`:

```
> curr_sample <- sample(word_lengths, 10)
```

Q: How can we get  $\bar{x}$  from this sample in R?

```
> mean(curr_sample)
```

Q: How could we get a full sampling distribution?

- A: Repeat this many times to get an approximation of the sampling distribution
- If we store the  $\bar{x}$ 's in a vector, we can then plot the sampling distribution as a histogram

# The do\_it() function

```
do_it(100) * {
```

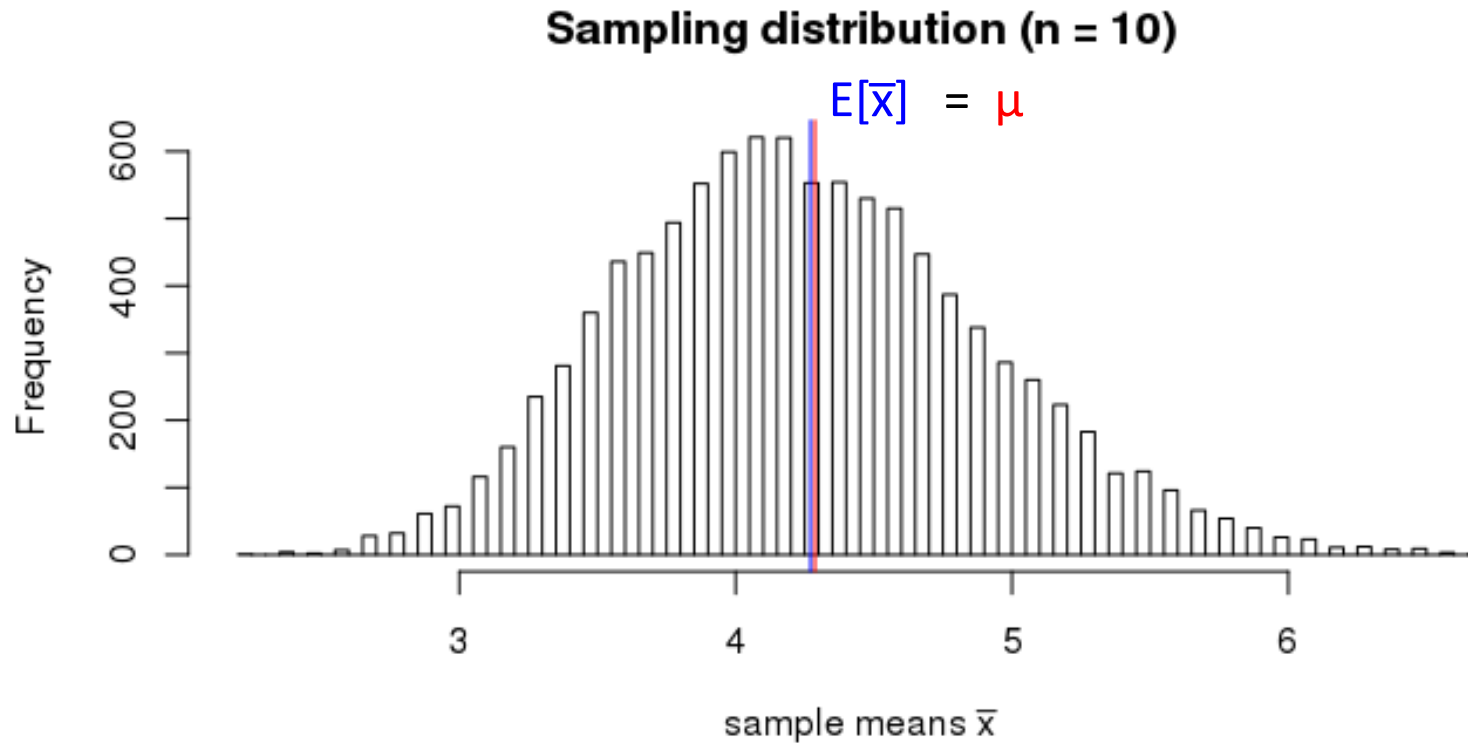
```
    2 + 3
```

```
}
```

# Let's create a sampling distribution in R

```
sampling_dist <- do_it(10000) * {  
  
    curr_sample <- sample(word_lengths, 10)  
    mean(curr_sample)  
  
}  
  
hist(sampling_dist)
```

# Sampling distribution in R



`mean(sampling_dist)`

`mean(word_lengths)`    # these are the same so no bias

# Changing the sample size $n$

What happens to the sampling distribution as we change  $n$ ?

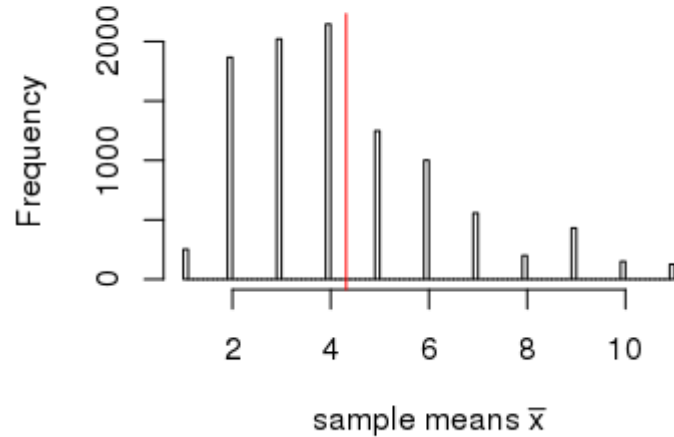
- Experiment for  $n = 1, 5, 10, 20$

```
sampling_dist <- do_it(10000) * {  
    curr_sample <- sample(word_lengths, 20)  
    mean(curr_sample)  
}
```

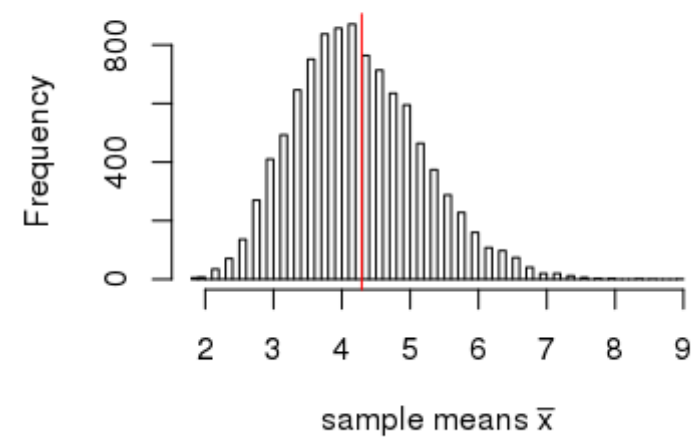
```
hist(sample_means, nclass = 100)
```

[Gettysburg sampling distribution app](#)

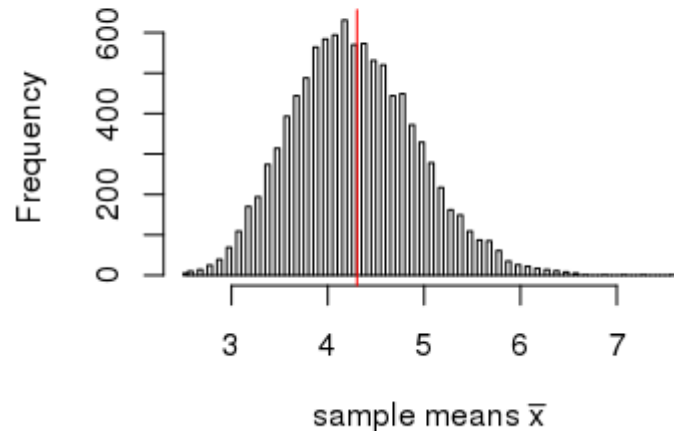
**Sampling distribution ( $n = 1$ )**



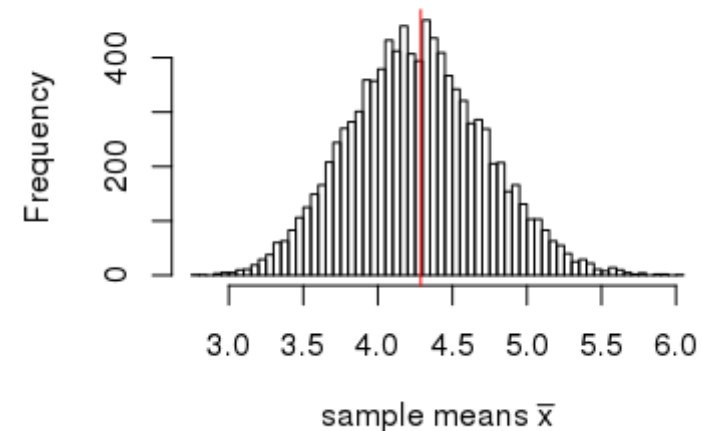
**Sampling distribution ( $n = 5$ )**



**Sampling distribution ( $n = 10$ )**



**Sampling distribution ( $n = 20$ )**



x-axis range 9 vs. 6

As the sample size  $n$  increases

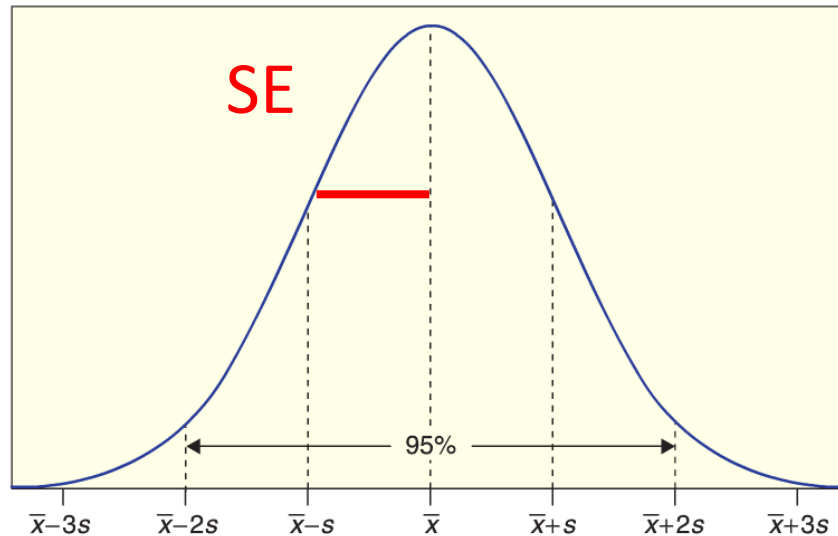
1. The sampling distribution becomes more like a normal distribution
2. The sampling distribution points ( $\bar{x}$ 's) become more concentrated around the mean  $E[\bar{x}] = \mu$



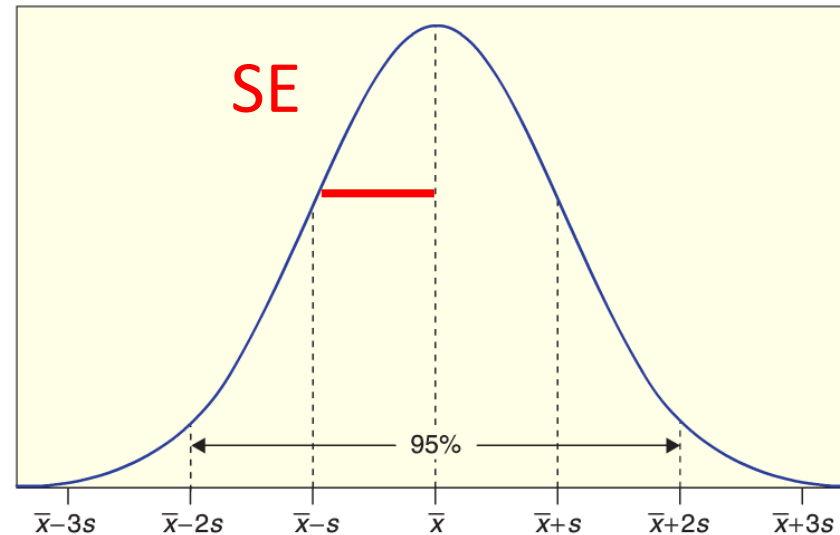
# The standard error

The **standard error** of a statistic, denoted SE, is the standard deviation of the sample statistic

- i.e., SE is the standard deviation of the *sampling distribution*



# What does the size of a standard error tell us?



Q: If we have a large SE, would we believe a given statistic is a good estimate for the parameter?

- E.g., would we believe a particular  $\bar{x}$  is a good estimate for  $\mu$ ?

A: A large SE means our statistic (point estimate) could be far from the parameter

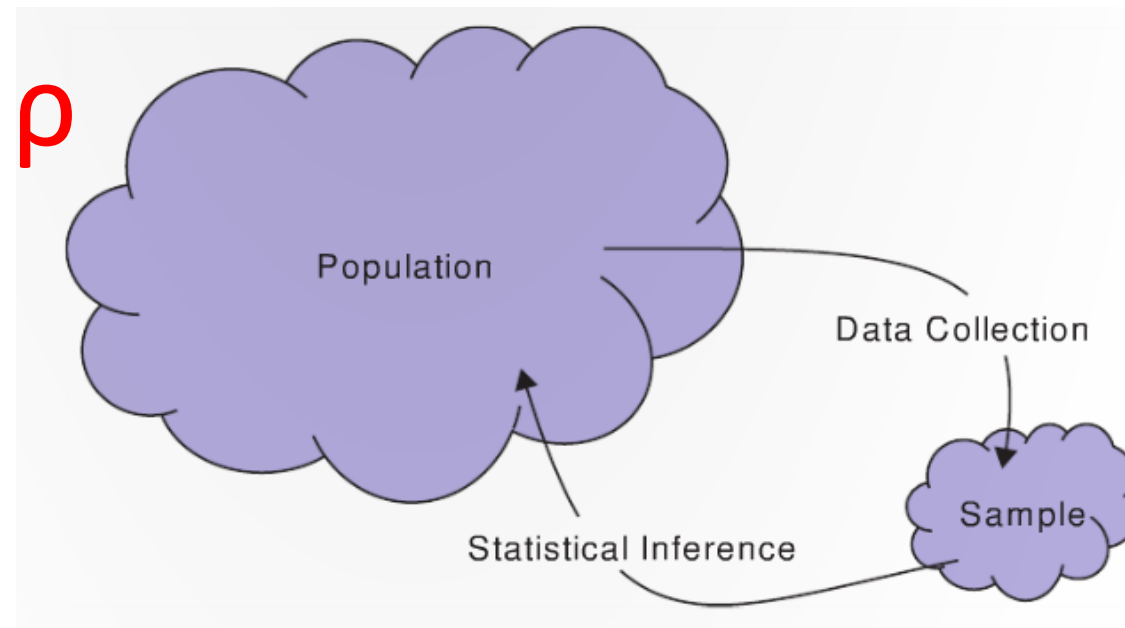
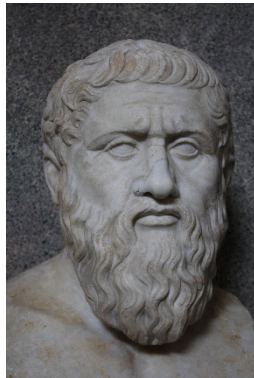
- E.g.,  $\bar{x}$  could be far from  $\mu$

# Back to the big picture: Inference

## Statistical inference is...?

the process of drawing conclusions about the entire population based on information in a sample

$\pi, \mu, \sigma, \rho$



$\hat{p}, \bar{x}, s, r$



# Point Estimate

We use the statistics from a sample as a **point estimate** for a population parameter

- $\bar{x}$  is a point estimate for...?  $\mu$

49% of American approve of Trump's job performance according to a recent Gallup poll

Q: What are  $\pi$  and  $\hat{p}$  here?

Q: Is  $\hat{p}$  a good estimate for  $\pi$  in this case?

A: We can't tell from the information given

# Interval estimate based on a margin of error

An **interval estimate** give a range of plausible values for a population parameter.

One common form of an interval estimate is:

*Point estimate  $\pm$  margin of error*

Where the **margin of error** is a number that reflects the precision of the sample statistic as a point estimate for this parameter

# Example: Fox news poll

49% of American approve of Trump's job performance, plus or minus 3%

How do we interpret this?

Says that the population parameter ( $\pi$ ) lies somewhere between 46% to 52%

i.e., if they sampled all voters the true population proportion ( $\pi$ ) would be likely be in this range

# Confidence Intervals

A **confidence interval** is an interval computed by a method that will contain the *parameter* a specified percent of times

- i.e., if the estimation were repeated many times, the interval will have the parameter x% of the time

The **confidence level** is the percent of all intervals that contain the parameter



# Think ring toss...

Parameter exists in the ideal world

We toss intervals at it

95% of those intervals capture the parameter



# Confidence Intervals

For a **confidence level** of 95%...

95% of the **confidence intervals** will have the parameter in them

