

Sampling distributions,  
standard errors, and confidence intervals

# Overview

Review: Sampling distributions and the standard error

Exploring sampling distributions in R

Confidence intervals

# Announcement

Homework 4 has been posted!

It is due on Gradescope on **Sunday February 15<sup>th</sup>** at 11pm

- **Be sure to mark each question on Gradescope!**

The material this week is going to be a bit more conceptually challenging

**Please attend the practice sessions** and office hours to reinforce your understanding!

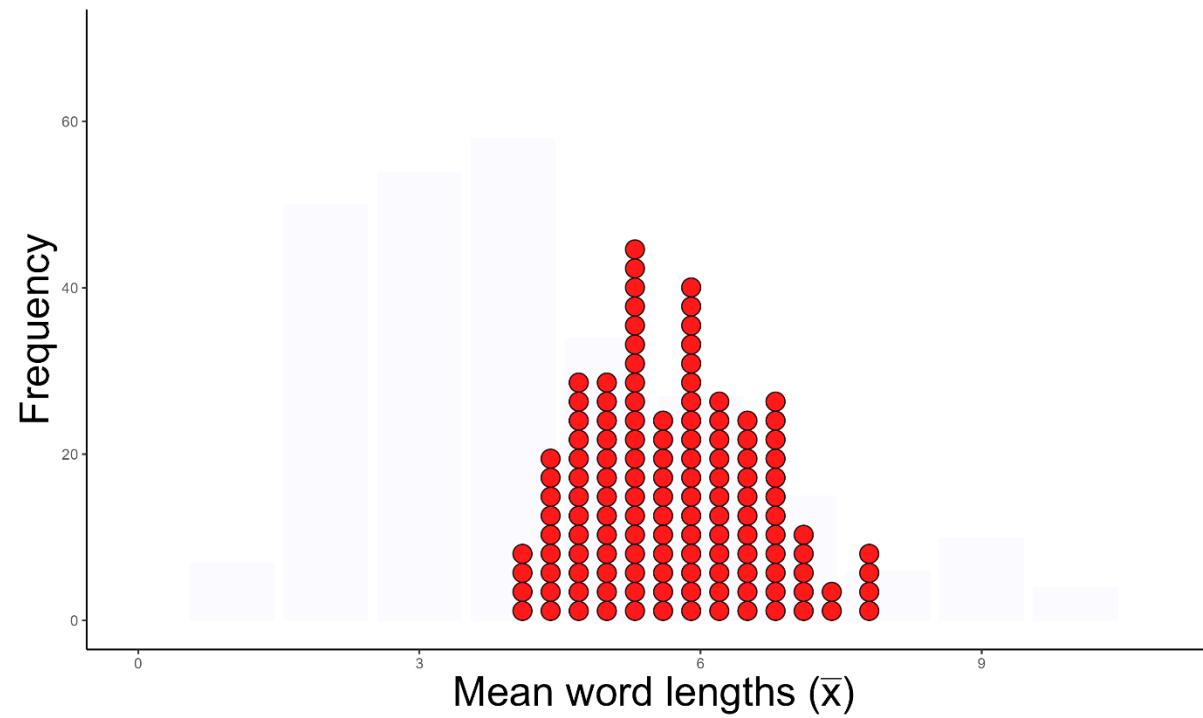
# Announcement: practice session locations

Day	Time	Location
Tuesday	4, 5pm	ESC 110 (21 Sachem Street )
Wednesday	4, 5pm	1105-B Kline tower
Thursday	4, 5pm	205 Kline tower
Friday	10, 11am	1105-B Kline tower

# Review: Sampling distributions

Recall for our distribution of Gettysburg word lengths...

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

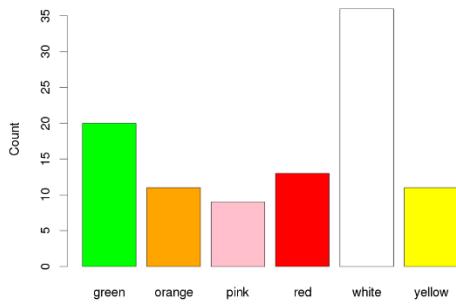


A: The mean length of 10 words ( $\bar{x}$ )  
i.e., each point in our distribution is a **statistic**!

# Sampling distribution

A **sampling distribution** is the distribution of sample statistics computed from 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



$\pi_{\text{red}}$



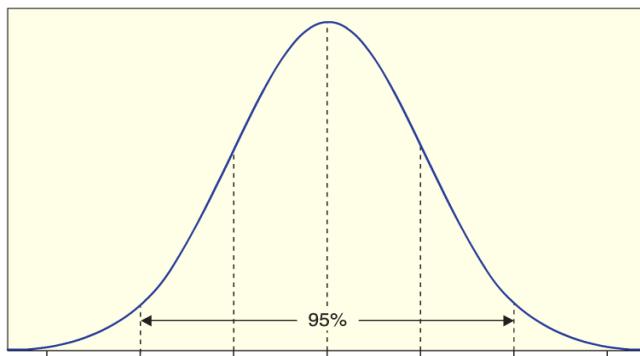
$n = 100$

$\hat{p}_{\text{red}}$

$\hat{p}_{\text{red}}$

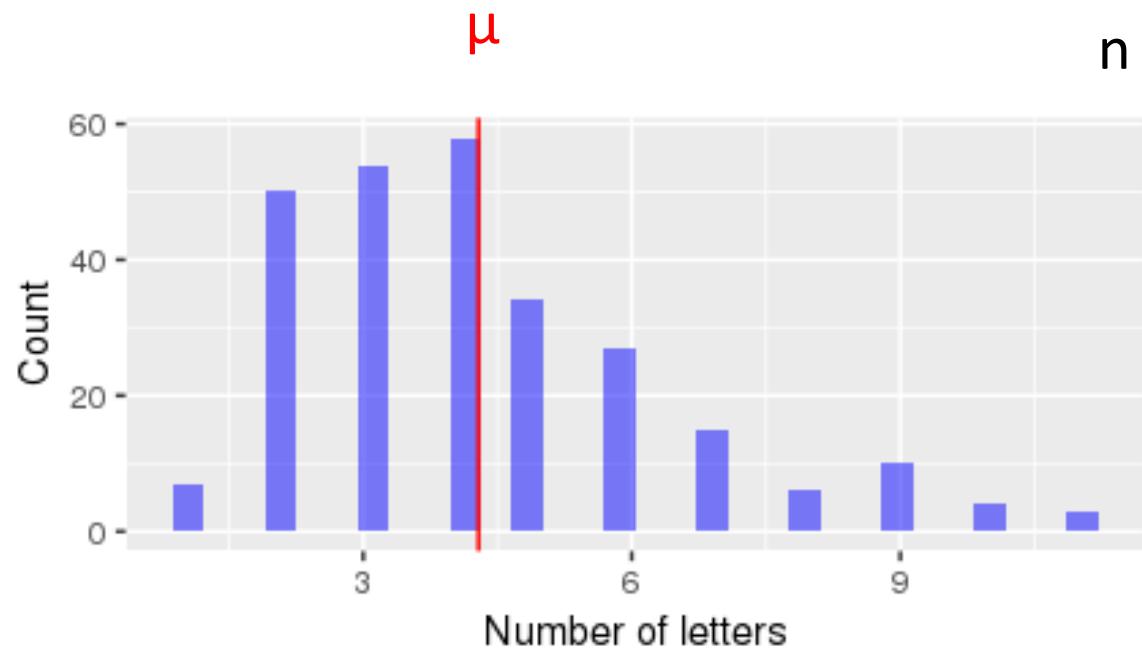
$\hat{p}_{\text{red}}$

Distribution  
of  $\hat{p}$ 's



Sampling distribution!

# Gettysburg address word length sampling distribution



n = 10

10, 3, 3, 3, 4,  
3, 2, 6, 10, 5

$\bar{x} = 5$

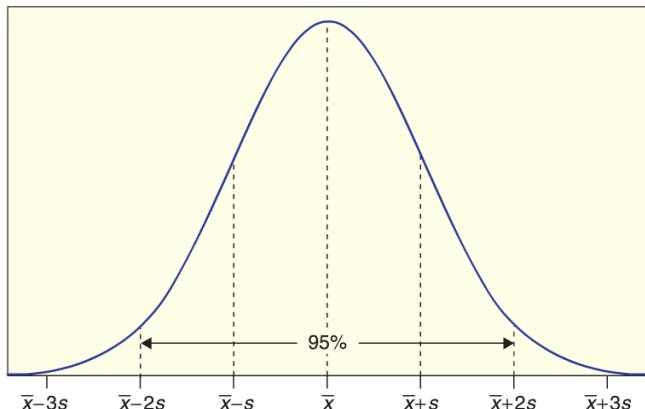
2, 6, 2, 6, 6,  
2, 5, 3, 2, 9

$\bar{x} = 4.3$

3, 9, 3, 4, 4,  
3, 6, 6, 2, 2

$\bar{x} = 4.2$

Distribution  
of  $\bar{x}$ 's



Sampling distribution!

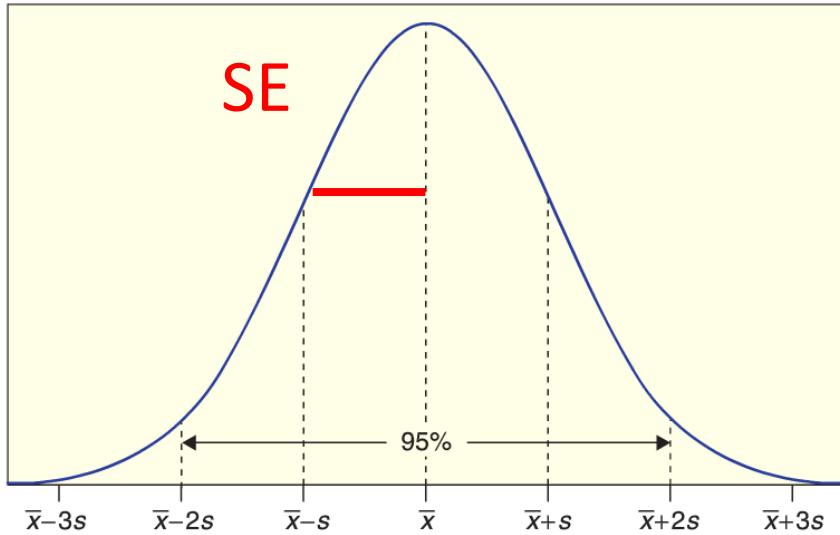
[Gettysburg sampling distribution app](#)

# The standard error

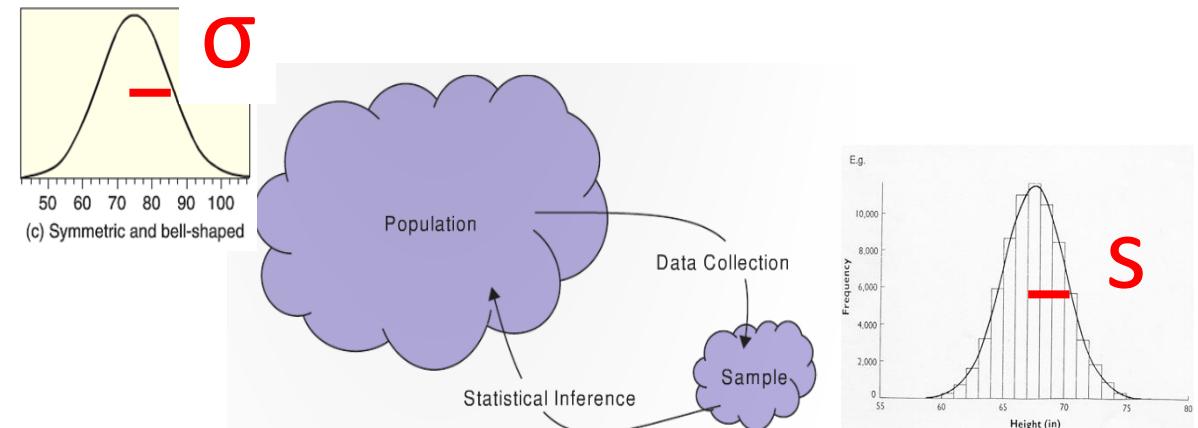
The **standard error** of a statistic, denoted  $SE$ , is the standard deviation of the sample statistic

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

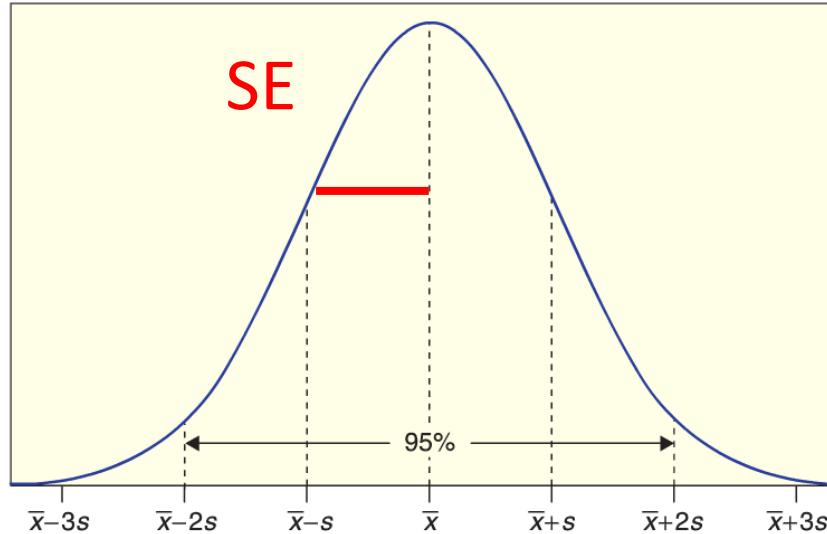
Note: **standard error  $SE$**  is different from  $\sigma$  and  $s$ !



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  $\mu$

# Sampling distributions in R!

# Let's create a sampling distribution in R

Load the SDS1000 library to make all SDS1000 functions available

```
library(SDS1000)
```

Get the Gettysburg population data

```
load("gettysburg.Rda")
```

```
word_lengths <- gettysburg$num_letters # lengths of the 268 words
```

# 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

The `do_it()` function (from the SDS1000 package) repeats a piece of code many times

- It returns a vector with the values created each time the code is repeated

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

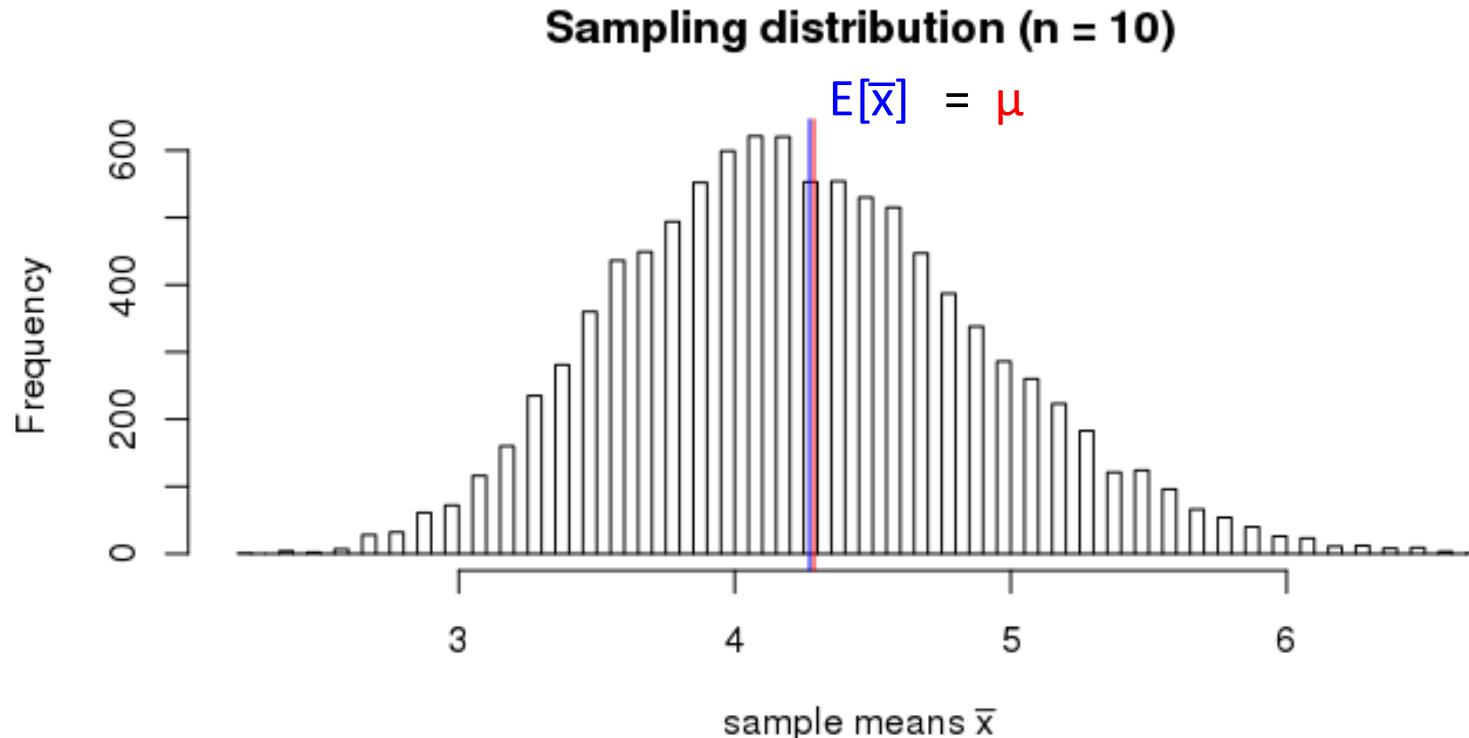
```
  2 + 3
```

```
}
```

# 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)  
SE <- sd(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*?

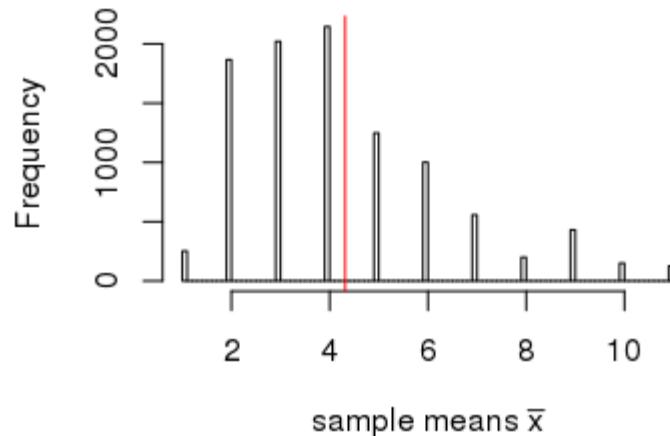
- Experiment for  $n = 1, 5, 20, 80$

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

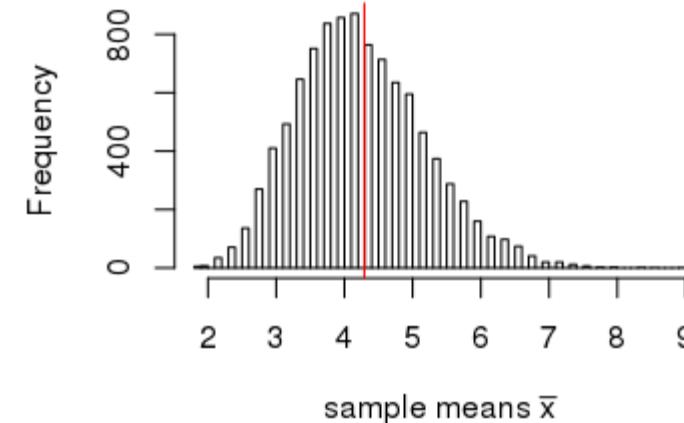
`hist(sample_means, breaks = 100)`

[Gettysburg sampling distribution app](#)

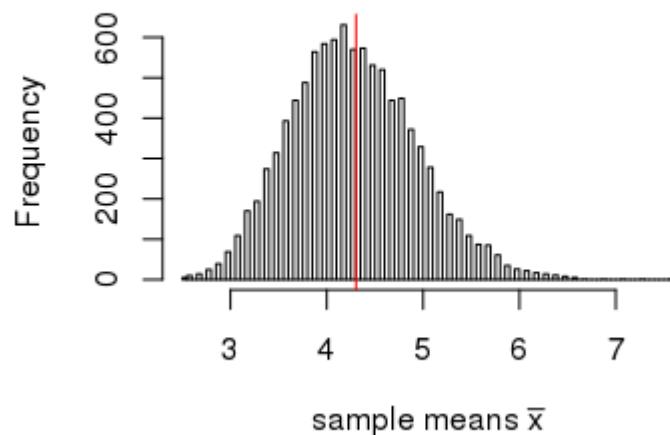
**Sampling distribution ( $n = 1$ )**



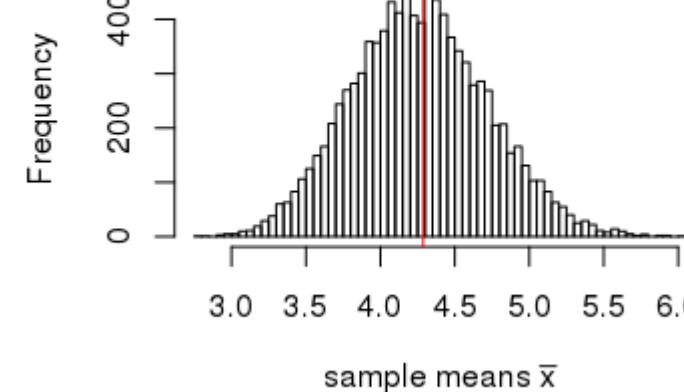
**Sampling distribution ( $n = 5$ )**



**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  $E[\bar{x}] = \mu$

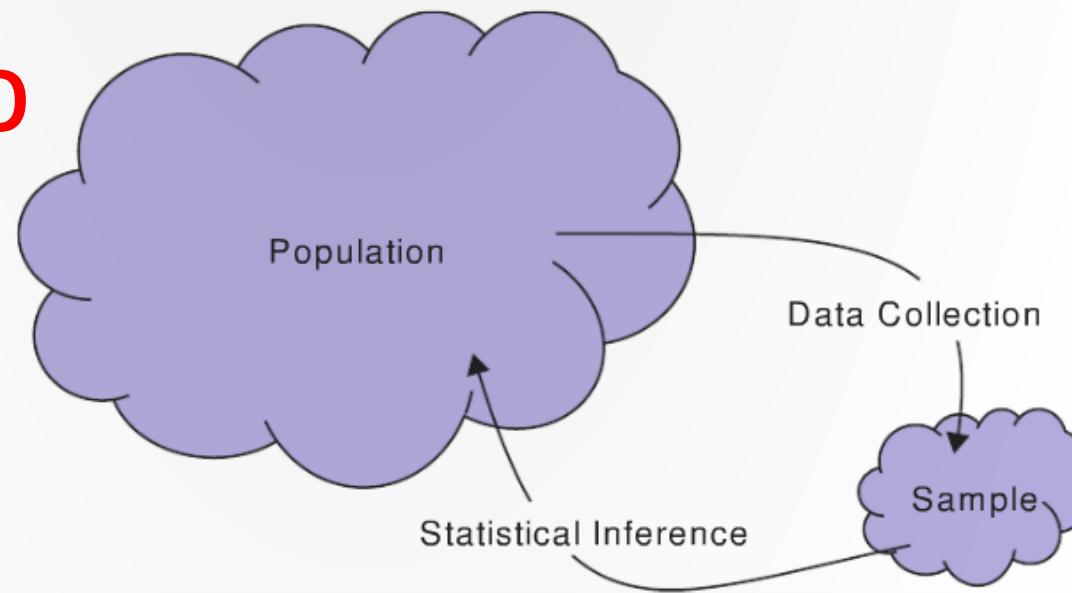
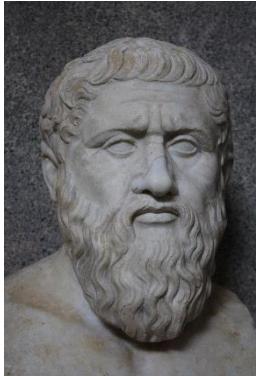
Point estimates and confidence intervals

# 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 a statistic from a sample as a **point estimate** for a population parameter

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

40% of Americans approve of Trump's job performance according to a poll of 2,464 people

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



([outdated graphic](#))

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: Gallup poll

40% 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 37% to 43%

i.e., if they sampled all voters, the true population proportion ( $\pi$ ) would 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

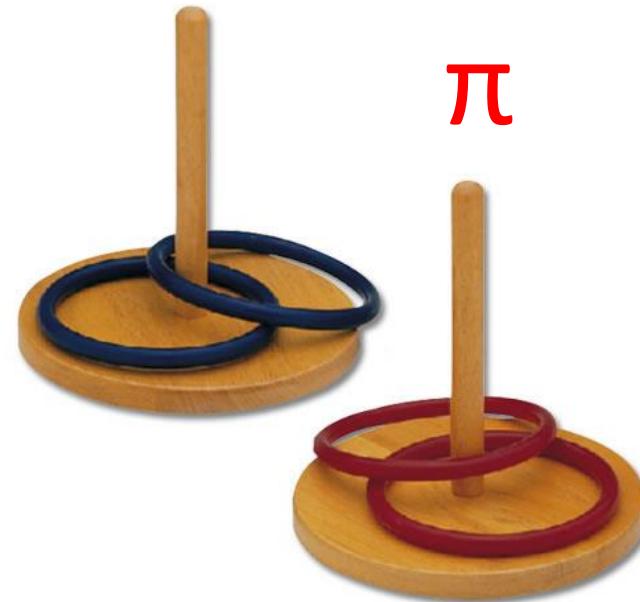
- i.e., if the estimation were repeated many times, the interval will have the parameter x% of the time

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

Think ring toss...

$\pi$

**A parameter exists in the world**



**We toss intervals at it**

- These are our confidence intervals

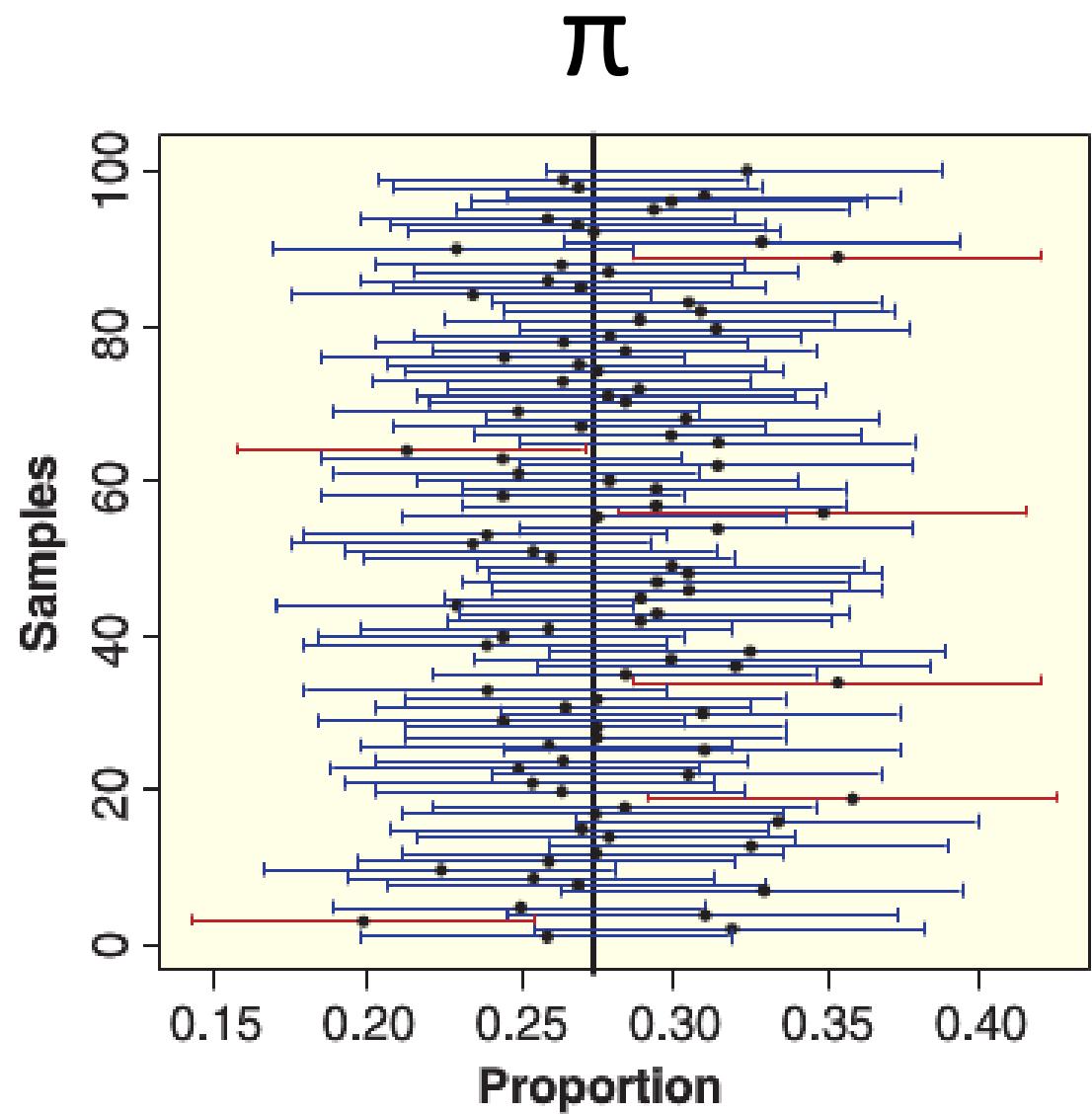
**95% of those intervals capture the parameter**

- (95% is our confidence level)

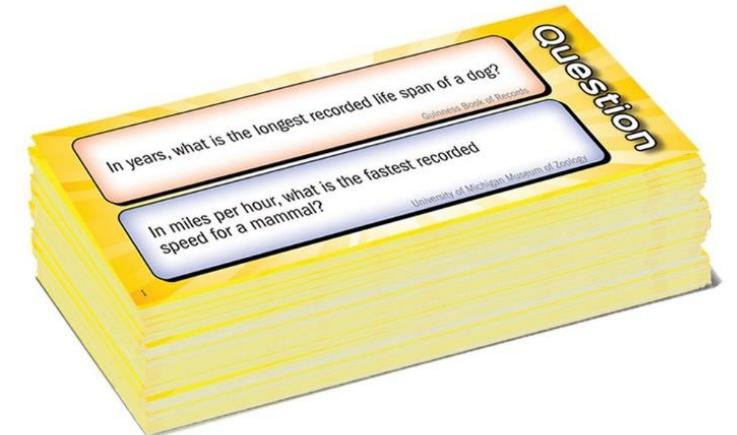
# Confidence Intervals

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

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



# Wits and Wagers: 90% confidence interval estimator



I will ask 10 questions that have numeric answers

Please come up with a range of values that contains the true value for 9 out of the 10 questions

- I.e., be a 90% confidence interval estimator

Write your range of estimates for each question as two numbers

- E.g., [10.2 to 50.7]