

Sampling Distributions

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Statistical Inference

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- ▶ In this activity, the population is end of semester grades of my previous Sta-209 students
 - ▶ I won't give you the population, but I'll let you take as many *random samples* of size $n = 10$ as you want
- ▶ Our short-term goal will be see what we can learn about a population by repeatedly taking random samples
 - ▶ Our long-term goal will be to apply this insight to situations involving only a single random sample

The Population Distribution

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- ▶ The *population distribution* contains *all* of the information about the variable of interest
 - ▶ In our example, we could view the population distribution using a table or bar chart of the end of semester letter grades
- ▶ In most situations, statisticians choose to focus on a single statistic that summarizes a single aspect of the population they are most interested in
 - ▶ With your group, decide upon a statistic that you're interested in from this population

Estimation

- ▶ Suppose you're interested in the proportion of A's in the population, denoted p_A
- ▶ How would you estimate p_A from a single random sample?
- ▶ How likely is it that your estimate is *exactly* p_A ?

Estimation

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- ▶ This estimate is *unlikely* to be exactly p_A , but for most samples it should be pretty close
 - ▶ Quantifying exactly how close \hat{p}_A is to p_A is a goal for what we'll do today
 - ▶ How might you approach this goal? (acknowledging that I'll never provide the true p_A)

Sampling Distribution Activity - Directions

This is the only time we'll use R in this class, but it is the software of choice for most statisticians, and you'll use it in future stats classes (if you choose to take them).

1. Open RStudio and type: `source("https://remiller1450.github.io/s209s20/funs.R")`
2. Enter `sample_grades()` to generate a random sample of student's end of semester grades
3. Find the proportion of A's in your sample and record it
4. Repeat steps 1-3 until you've recorded results from many different random samples

These values represent the distribution of possible sample proportions that *could occur* when taking a random sample of size $n = 10$ from this population. With your group, discuss why it is important to study this distribution.

Sampling Distribution Activity - Some Questions

1. Based off the **sampling distribution** (the dotplot on the board), what do you think p_A is?
2. Had you only collected a *single* random sample of size 10, what would you expect is the *most likely* value of \hat{p}_A for that sample?
3. How much variability is there across different samples?
4. Could we use this variability to come up with an **interval estimate** of p_A ?

Sampling Distribution Activity - Answers

1. Assuming the samples are *representative*, p_A is the center of the sampling distribution! This is because the sample statistic \hat{p}_A is **unbiased**
2. p_A is the center of the sampling distribution, so \hat{p}_A is most likely to be p_A !
3. We can assess the variability of the possible sample means that we could see by looking at the standard deviation of the sampling distribution, this is called the **standard error (SE)** since it describes an estimate
4. We could provide estimates of p_A that look like $\hat{p}_A \pm c * SE$. The 68-95-99 rule could help us choose c (at least for sampling distributions with the right shape)

Confidence Intervals

- ▶ Intervals of the form Estimate \pm *MOE*, where *MOE* is a carefully determined margin of error, are known as **confidence intervals**
- ▶ We will spend the next couple of weeks studying confidence intervals in greater detail
- ▶ For now, we'll see how a few different factors (like sample size and sampling bias) impact a *sampling distribution*

The Role of Sample Size

The sampling distribution depends upon:

1. The parameters of the population distribution
2. The size of the sample
3. How the sample was collected

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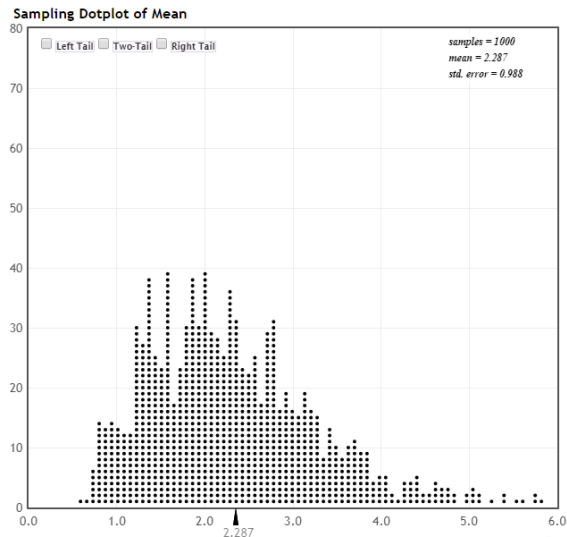
1. The parameters of the population distribution
 2. The size of the sample
 3. How the sample was collected
-
- ▶ We'll first investigate the role of sample size using *StatKey*, a free online companion to the Lock5 textbook: [StatKey Link](#)
 - ▶ We'll look at the "NFL Contracts" dataset that comes pre-loaded in StatKey

The Role of Sample Size - Directions

- ▶ Open StatKey at lock5stat.com/StatKey and navigate to “Sampling Distribution for a Mean”
- ▶ Select the “NFL Contracts” dataset in the top left (under the red StatKey logo)
- ▶ Describe the shape of the *population distribution*
- ▶ Describe the shape of the *sampling distribution* of samples of sizes $n = 10$, $n = 30$ and $n = 100$
- ▶ Record the *standard error* of each sampling distribution created above

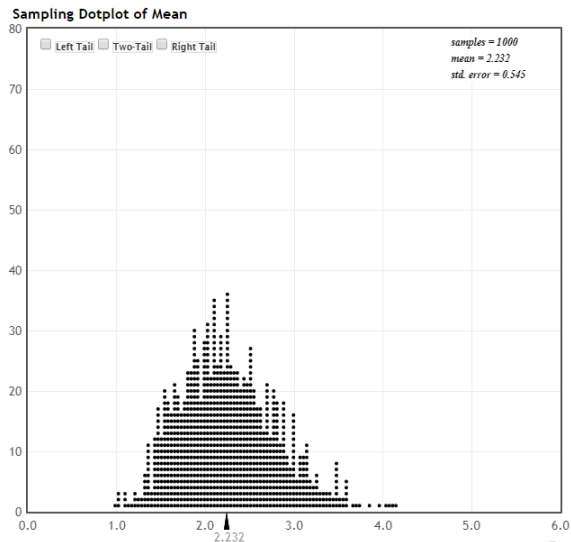
The Role of Sample Size - Results

Sampling distribution of \bar{x} for 1000 samples of size $n = 10$



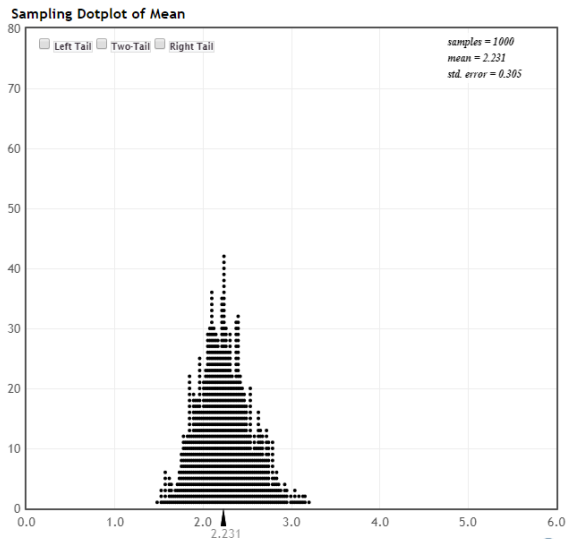
The Role of Sample Size - Results

Sampling distribution of \bar{x} for 1000 samples of size $n = 30$



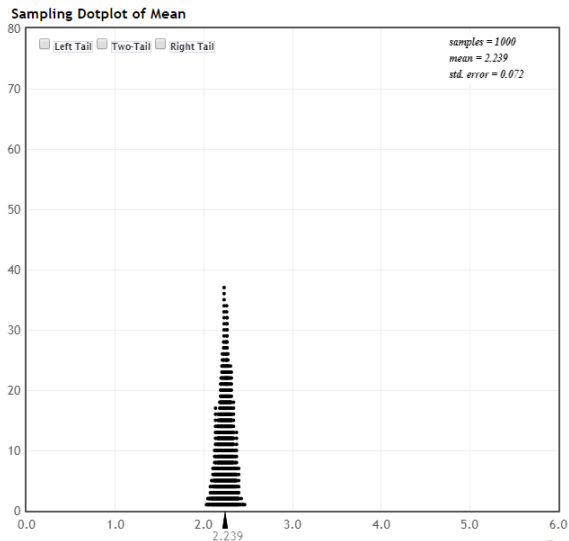
The Role of Sample Size - Results

Sampling distribution of \bar{x} for 1000 samples of size $n = 100$



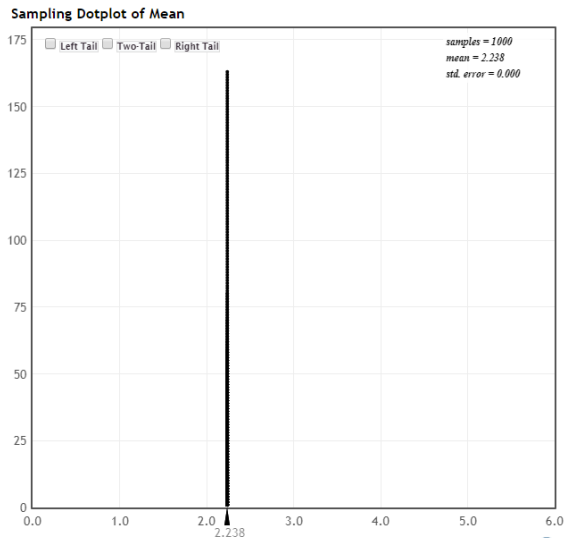
The Role of Sample Size - Results

Sampling distribution of \bar{x} for 1000 samples of size $n = 1000$



The Role of Sample Size

Sampling distribution of \bar{x} when the entire population is sampled



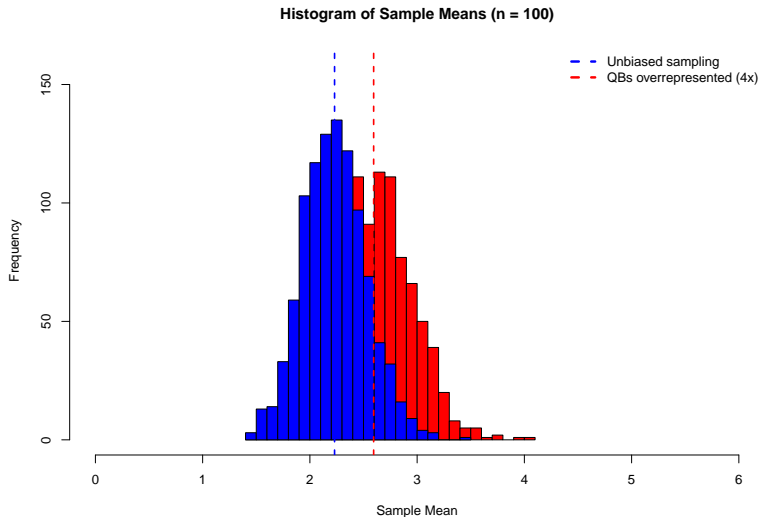
The Role of Sample Size - Conclusions

- ▶ As the size of our sample increases, the **standard error**, denoted SE , of our sample statistic decreases
- ▶ Standard error is the standard deviation of a sample statistic (ie: it describes variability in the sampling distribution)

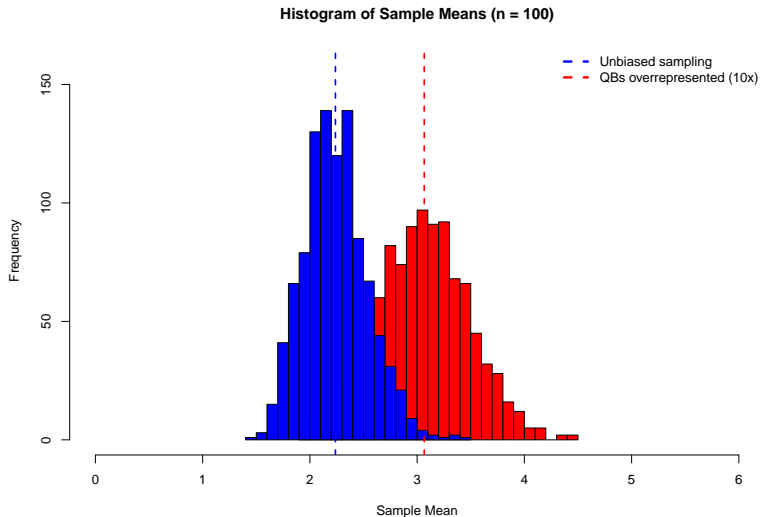
Sampling Bias

- ▶ Quarterbacks represent 4.3% of NFL players but tend to receive a disproportionate amount of media attention and are paid higher salaries than other positions
- ▶ Suppose we sample in a way that makes QBs four times more likely to be sampled than other positions, how might this influence the sampling distribution (for estimates, \bar{x} , of mean the NFL salary)?
- ▶ What if QBs were ten times more likely to be sampled?

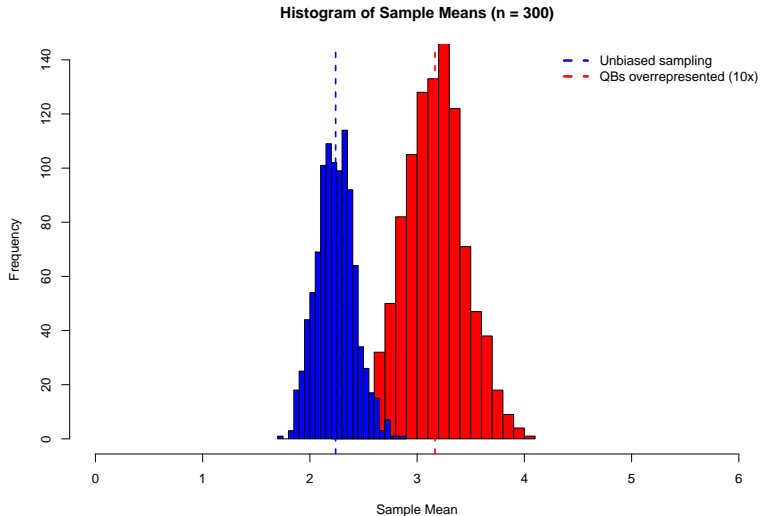
Sampling Bias



Sampling Bias



Sampling Bias



Sampling Distributions - Conclusions

- ▶ Larger samples tend to provide better estimates if the samples are representative
 - ▶ But larger sample size cannot fix sampling bias, it actually can exacerbate it
- ▶ Next we'll see how the sampling distribution can be used to construct *confidence intervals* and exactly how special it is for these intervals to be meaningful

Conclusion

Right now you should:

1. Understand the relationships between the **population distribution**, the **sample distribution**, and the **sampling distribution**
2. Be comfortable with the terminology of **parameters** and **statistics**
3. Understand, when we only have one sample, the sample statistic is our best guess at the population parameter
4. Understand the impact of bias and sample size (variability) on the sampling distribution

If you want more information:

- ▶ Read Ch 3.1