

CSCI 3022

# intro to data science with probability & statistics

Lecture 21  
April 4, 2018

The Bootstrap wrapup  
Intro to Regression *(maybe)*

# Stuff & Things

- HW5 due **this Friday**.

OH today 11-1

Fr 8-10 ~~10~~ 9:45

# Previously on CSCI 3022

- **Definition:** a bootstrapped resample is a set of  $n$  draws from the original dataset (drawn IID from  $X$ ), sampled *with replacement*.
- **Proposition:** a suitable estimate of the 95% confidence interval for the mean of the distribution  $X$  is given by  $[a, b]$ , where  $a$  and  $b$  are the 2.5 percentile and 97.5 percentile of the means of a large number of bootstrapped resamples.
- **In plain English:** resample your original data many times. Compute the mean for each resample. Compute the 2.5 and 97.5 percentiles of those means.

# Bootstrap: why we like it

- The bootstrap for a confidence interval around the mean is convenient, particularly when there are **not enough samples** to use the CLT.
- Of course, if we *can* use the CLT, we should. So why is the bootstrap so exciting?

# Bootstrap: why we like it

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## We can bootstrap CIs for things other than the mean!

- Median. ✓
- Standard Deviation. ✓
- Other statistical measures that we don't have a theory for.

# Bootstrap for the median

90%

✓

- Let's write **pseudocode** for how we would bootstrap a CI for the median:

1. Resample. Create  $M$  resampled datasets (with replacement)

2. For each resampled dataset, compute the median

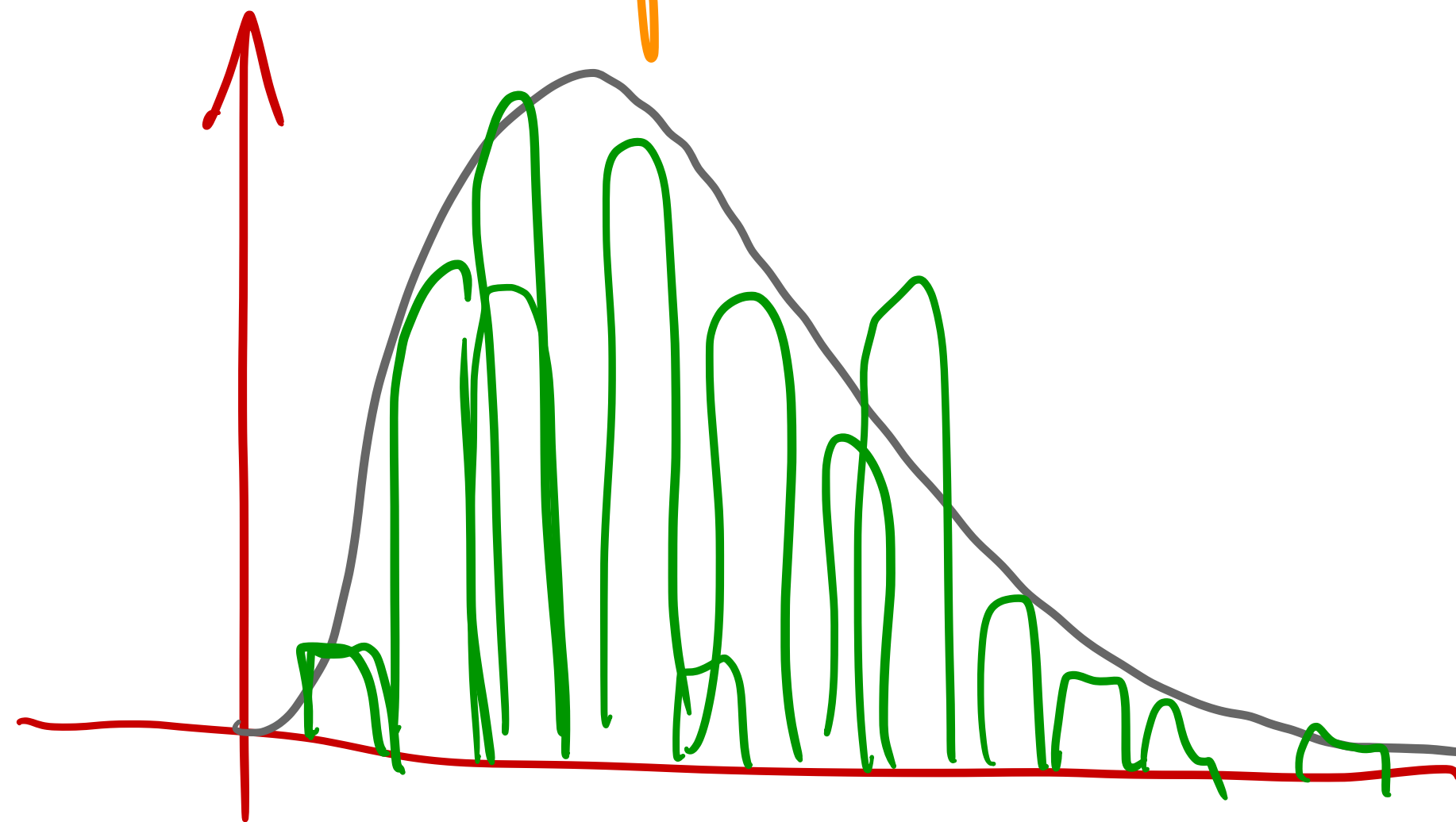
3. Take that distr. of medians, and compute  
5<sup>th</sup> percentile and 95<sup>th</sup> percentile

$$CI: \left[ \frac{100 - \alpha}{2} \%, 100 - \frac{100 - \alpha}{2} \% \right]$$

# Bootstrap for the variance

- Let's write **pseudocode** for how we would bootstrap a CI for the variance:

No. See prev. slide.



# The Non-Parametric Bootstrap

- In the literature—your book, the Wikipedia, etc—you may read about a “non-parametric bootstrap.” What is this?



# The Non-Parametric Bootstrap

- In the literature—your book, the Wikipedia, etc—you may read about a “non-parametric bootstrap.” What is this?
- Let’s decode this word, “non-parametric”
- **Definition:** *parametric statistics* assumes that sample data comes from a population that follows a probability distribution based on a fixed set of parameters.
- Can you name some **examples** of distributions with parameters?

Urchin Eating

$\text{Pois}(\lambda)$

$N(\mu, \sigma^2)$

$\text{Bin}(n, p)$

- Can you name a *non*-parametric distribution we’ve talked about in class?

Let  $X$  be a r.v. such that  $P(X=-1)=0.2$ ,  $P(X=0)=0.5$ ,  $P(X=1)=0.3$

# The Parametric Bootstrap

Replace with  
your preferred  
distr. and statistic

- We call the bootstrap discussed in class today the non-parametric bootstrap because it doesn't assume any parametric distribution. What you resample is what you get.
- **Definition:** the parametric bootstrap estimates a CI for a desired property in two steps: (1) repeatedly estimate the parameter(s) of the known distribution, and then (2) compute a CI for the desired property by sampling from the known known distribution using the parameters that you inferred.

1. Create  $M$  bootstrapped datasets.

2. Assume that the data came from  $\text{Pois}(\lambda)$ , and estimate  $\lambda$  for each of the  $M$  datasets.

3. Use those parameters, and for each one, compute the median.

4. Compute the CI from those medians.

# The Parametric Bootstrap

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- **Definition:** the parametric bootstrap estimates a CI for a desired property in two steps: (1) repeatedly estimate the parameter(s) of the known distribution, and then (2) compute a CI for the desired property by sampling from the known known distribution using the parameters that you inferred.
- **Why?** The parametric bootstrap can be shown to do a better job than the non-parametric bootstrap in various scenarios.
- Why not use the parametric bootstrap all the time?
  1. Might not be getting data from a parametric distrib.
  2. Might not know what the parametric distrib. is!

# Let's notebook it up!

