



**DATA 8**

Fall 2016

# Lecture 22, October 17

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## The Bootstrap

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# Announcements

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- The class did fine on the midterm. The score distribution, answers, regrading policy, and graded exams will be released by Tuesday evening.
  - Labs meet as usual this week.
  - No homework due this week.
  - Homework will be assigned on Friday.
  - We're being filmed today. If you'd prefer not to appear, try sitting near the back.
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# Percentiles

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- The data: a collection of numerical values
  - The  $p$ th percentile of the data is:
    - the smallest value in the collection
    - that is at least as large as
    - $p\%$  of all the values.
  - The median (50th percentile) of 4, 7, 9, 10, 15:
    - 9
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# percentile

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- Function:  
`percentile(p, array)`
  - `p` is between 0 and 100
  - Returns the  $p$ th percentile of the array
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# Inference: Estimation

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- How big is an unknown parameter?
- If you have a census (that is, the whole population):
  - Just calculate the parameter and you're done.
- If you don't have a census:
  - Take a random sample from the population.
  - Use a statistic as an **estimate** of the parameter.

(Demo)

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# Variability of the Estimate

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- One sample  One estimate
  - But the random sample could have come out differently.
  - Then the estimate would have been different.
  - Main question:
    - **How different could the estimate have been?**
  - The variability of the estimate tells us something about how accurate the estimate is.
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# Where to Get Another Sample?

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- One sample  One estimate
  - To get another value of the estimate, need another random sample.
  - Can't go back and sample again from the population:
    - No time, no money
  - Stuck?
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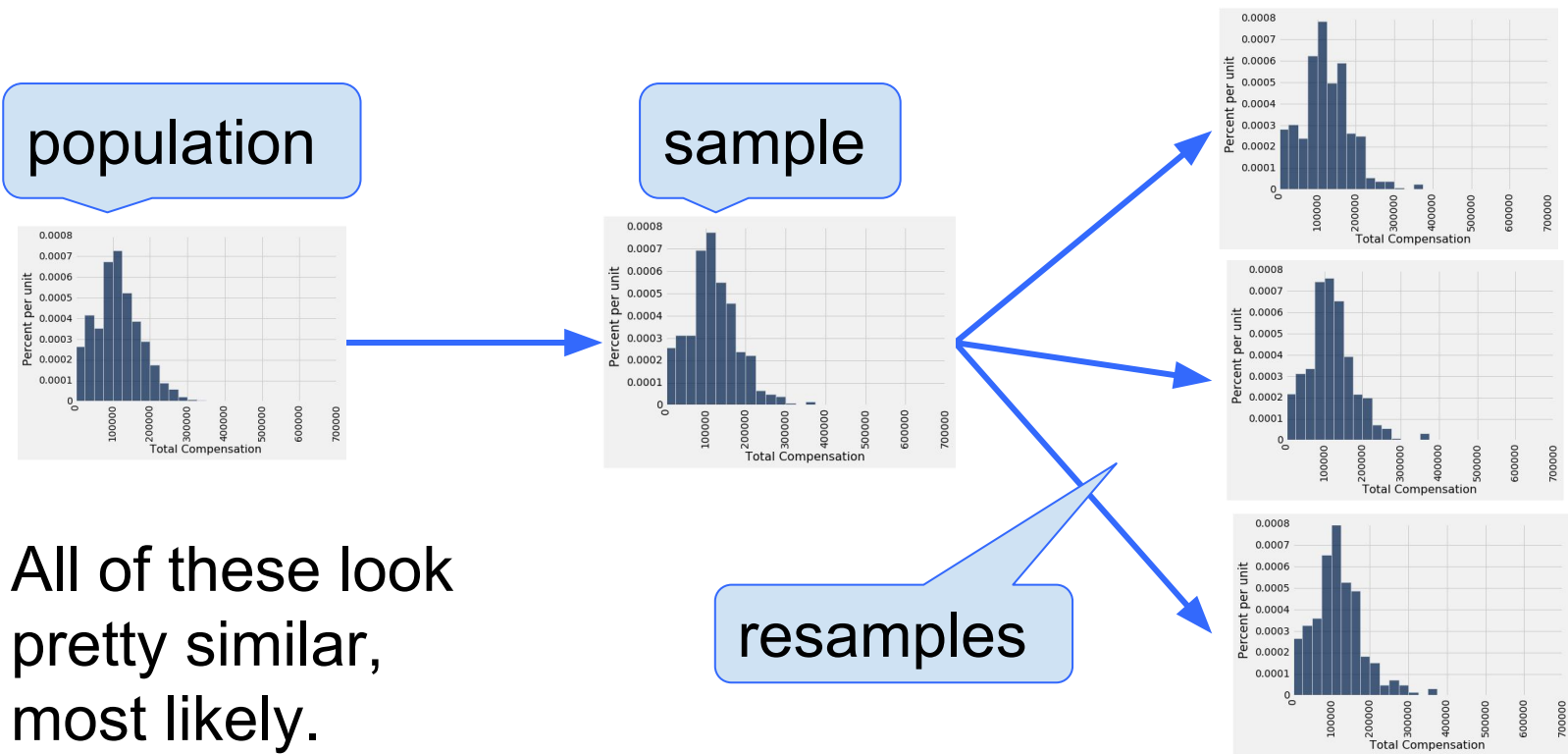
# The Bootstrap

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- Need another random sample that looks like the population
  - All that we have is the original sample
    - which is large and random.
    - It's a good bet that it resembles the population.
  - So **sample at random from the original sample!**
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# Why the Bootstrap Works



# Key to Resampling

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- From the original sample,
  - draw at random
  - **with** replacement
  - the **same number of times** as the original sample size.
- The size of the new sample has to be the same as the original one, so that the two estimates are comparable.

(Demo)

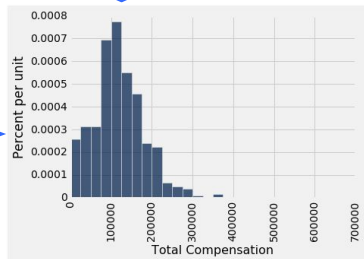
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# Inference Using the Bootstrap

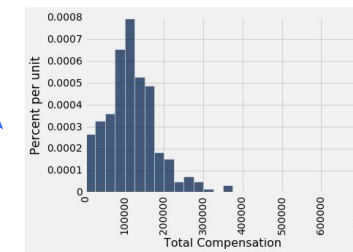
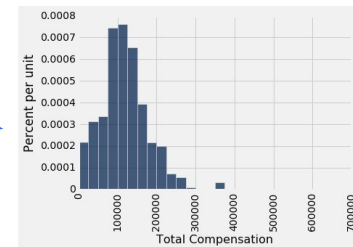
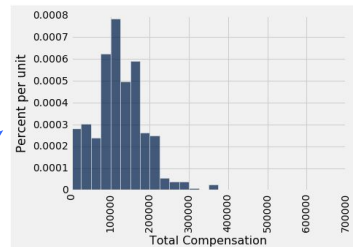
population



sample



resamples



All of these look pretty similar, most likely.