DISCOVERING STATISTICS

Bootstrap

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DISCOVERING STATISTICS

## Chapter 2 Populations and samples

### Population

 The collection of units (alleles, cells, organisms, groups, species, etc.) to which we want to generalize a set of findings or a statistical model

### Sample

 A smaller (but hopefully representative) collection of units from a population used to determine truths about that population

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DISCOVERING STATISTICS

### From sample to population

### Sample

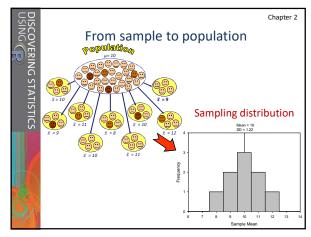
 Statistics we calculate (e.g. mean, sd, correlation coefficient, etc.) describe the sample from which they were calculated...

### Population

 $- \ \dots$  but are intended to describe the entire population

### • Sample to Population:

– How well does a given sample represent the population?



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# DISCOVERING STATISTICS USING

### From sample to population

Chapter 2

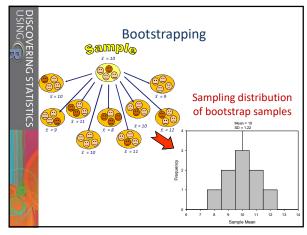
- However, it is often very impractical (if not impossible) to collect many different samples in the real world
  - Parametric tests overcome this "fundamental unknowability" of the sampling distribution by making the <u>assumption</u> that, if sample size is big enough, then the distribution of data is approximately normal
  - The significance value of test-statistics calculated from a sample, critically depends on the validity of this, and other, assumptions

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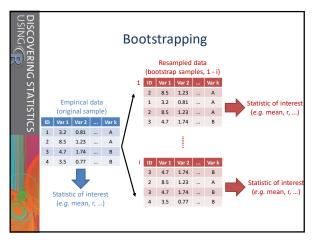


# From sample to population From re-sample to sample

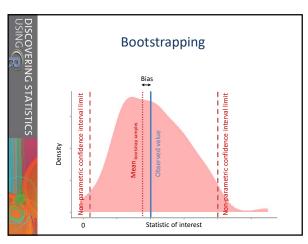
- However, it is often very impractical (if not impossible) to collect many different samples in the real world
  - Bootstrapping overcomes the "fundamental unknowability" of the true sampling distribution by generating many bootstrap samples from the observed data
  - It thus makes no assumptions about the distribution of data, but reconstructs the sampling distribution using the information contained in the collected data



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DISCOVERING STATISTICS
USING

### **Bootstrapping**

- Resampling (with replacement)
  - Useful 'trick', especially when:
    - The theoretical distribution of a statistic is unknown or complicated
    - Sample size is too small to allow parametric inference
- Confidence interval estimates
  - By default the "boot.ci()" function will try to calculate 5 non-parametric confidence intervals
  - Of these, the bias-corrected accelerated (BCa) intervals appear to be most robust across a wide range of data scenario's

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DISCOVERING STATISTICS

### Bootstrapping in R

- 3 steps
  - 1. Write a function to calculate a statistic of interest, *e.g.* the mean

stat<- function(variable, i) {
 mean(variable[i], na.rm= T)
}</pre>

- Plug this function into the 'boot()' function, and set the number of bootstrap samples you want to generate b.mean<- boot(dataframe\$column, stat, 10000)</li>
- 3. Look at outcome, and confidence intervals (e.g. 99%) b.mean

boot.ci(b.mean, 0.99)

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### Bootstrapping in R

- The output reports three values
  - original
  - value of the test statistic in the original sample
    - bias
      - difference between the original value and the average value across all bootstrap samples
    - std. error
      - the standard deviation of the sampling distribution of values across all bootstrap samples

# DISCOVERING STATISTICS USING (R)

### Bootstrapping in R

- Non-parametric confidence intervals
  - 5 different types
  - For the test statistic to be significant at the specified level (i.e. 95 or 99%), the interval should not include 0
- To visualise the bootstrap sampling distribution

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### Bootstrapping in R

- How to report findings...
  - Creativity was positively related to how well people performed in the annual World's Biggest Liar competition,  $\tau$  = .30 (n\_{participants}= 68, n\_{bootstraps}= 2000, 99%-Cl\_BCo= .01 .51)