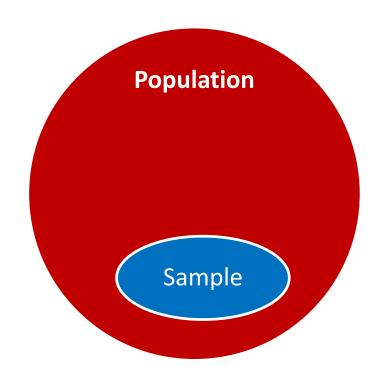
Lunch & Learn Intro Statistics

Heather Schroeder Nov. 14, 2016

Why do we do statistics?

- In one sentence To answer questions using facts.
- What do we want to know?
 - A population
 - We only have a sample



Overview

- Text:
 - Mind on Statistics 3rd Edition (2007)
- Object is to understand terms
- We won't calculate anything
 - Discuss interpretation and not calculation
- Broad overview of topics
- Use an example throughout presentation

Example Data

- ID, Gender, Height, NPR Listenership
- Categorical, Continuous

ID	Gender	Height (inches)	NPR?
1	F	73	Yes
2	M	70	Yes
•••	•••	•••	
99	F	74	No
100	M	67	What is NPR?

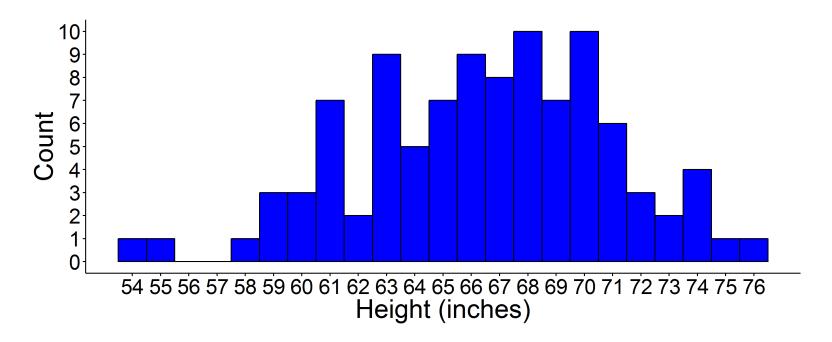
Describing Continuous Variables

- Mean
 - average
- Median
 - middle number when data is sorted
- Mode
 - Value seen MOst often in the data
- Histogram
 - Visual representation of data values
- Standard Deviation
 - Variability measure



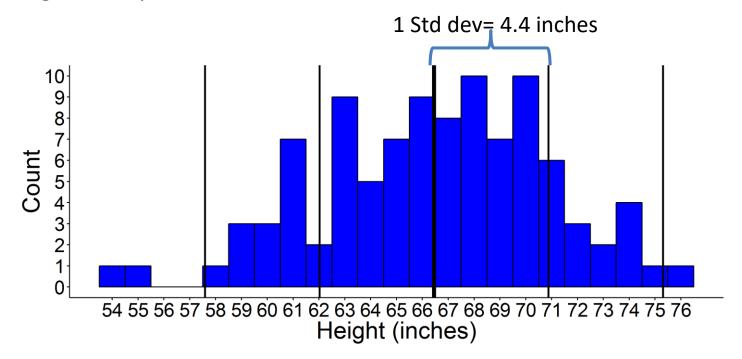
Mean, Median, Mode, Histogram

- Height example:
 - Mean= 66.5 in (5' 6.5")
 - Median= 67 in (5' 7")
 - Mode= 68 in (5' 8")



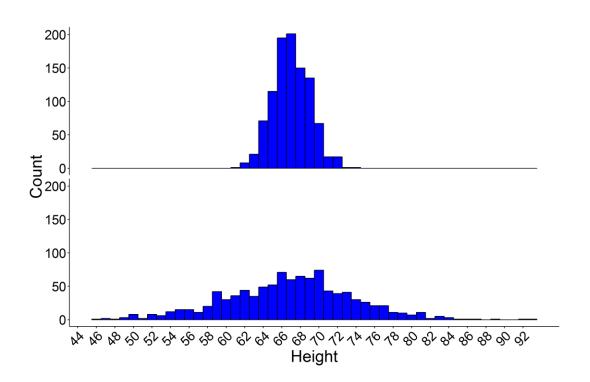
Standard Deviation

- Describes the spread of the data values
- Summarizing how far individual data values are from the mean
- If the data are normally distributed (height)
 - ~68% of the data is within +/- 1 standard deviation from the mean
 - ~95% of data points are captured within +/- 2 standard deviations from the mean
- Height example: Standard deviation= 4.4 inches



Small vs. Large Std. Deviation

- A small standard deviation: data points tend to be close to the mean
- A larger standard deviation: data points are more spread out over a range of values



Standard Error

- How good (accurate) is our estimate?
- Statistics (like mean) estimate trait of a population
- This estimate can have error
 - It can be (slightly) different from the true mean of the population
 - True mean of population = mean if we had measured every single member of the population
- We can estimate this error using a special kind of standard deviation: STANDARD ERROR
 - Standard Error is an estimate of the standard deviation of a statistic

Standard Error

- Roughly the average distance between our estimate and the true population parameter (mean)
- Standard Error = $\frac{Std.Deviation}{\sqrt{n}}$
- Takes sample size (n) into account
 - Small n -> larger error
 - Large n -> smaller error
 - Estimates get more accurate as n gets larger
- Different calculation for each kind of estimate (proportion, mean, difference between 2 means, etc)

Standard Error- Ex. Mean Height

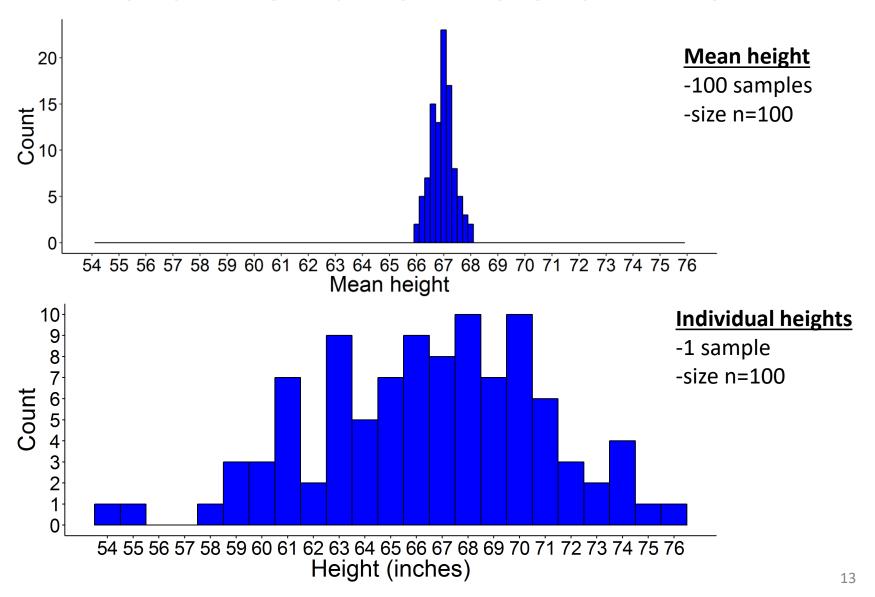
- Our example: Mean height=66.5 inches,
 n=100, Std. Dev = 4.4
- Stand error of mean height $\frac{\text{Std.Dev}}{\sqrt{n}} = \frac{4.4}{\sqrt{100}} = 0.44$ inches

Std. Deviation vs Std. Error

- How to remember difference between standard *deviation* and standard *error*:
 - Standard Deviation:
 - Spread of individual height measurements
 - No one has a "wrong" height -> not ERROR, a DEVIATION
 - Ex. Shaq is 7'1" (85 inches)
 - 4.2 std. deviations away from the mean
 - Standard Error:
 - Spread of mean heights
 - You CAN have error in your estimate



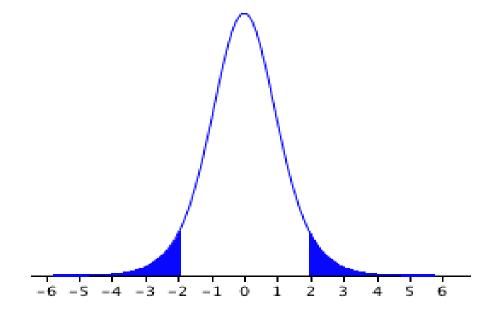
Std. Deviation vs Std. Error



- Is the mean (population) height for females 64 inches?
- We use our data (sample) to try to answer this question about the population
- Steps
 - 1. Get our estimate of mean female height
 - 2. Get a value for the t test statistic
 - 3. Get a p-value
 - 4. Use p-value to see if the difference has "statistical significance"

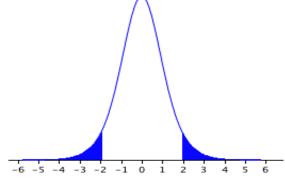
- 1. Estimate mean height from our sample
- 2. Get a value for the t test statistic
 - T-test statistic a standardized value of the difference between our estimate and the value we are comparing it to (we remove the units)
 - $-\left(\frac{Our\ estimate\ -Comparable}{Std.Error}\right)$
- 3. T-test statistic lets us look up a p-value
- 4. Draw a conclusion

- Test statistics have known distributions
 - t, z, chi square, F, etc.
 - What the distribution looks like depends on:
 - test statistic, sample size, assumptions



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P-value



- Use distribution of test statistic to
 - determine how likely it is we saw our result or an even more extreme result by random chance
 - "How likely" probability (range = 0 to 1)
 - If that probability is small enough, we call it...

Statistically significant!

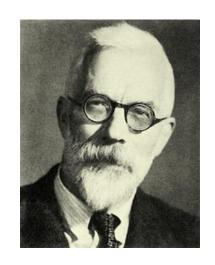
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Statistical Significance

- If what we are asking is true...
 - If the true mean female height is 64 inches ...
- A small p-value means it's not very likely that our result happened by chance
 - Therefore it must be statistically significant!
- A not so small p-value means our result isn't statistically significant.
 - Doesn't mean it's not practically significant (in real life), just that we don't have enough evidence to say it is statistically significant

Statistical Significance

- Unwritten standard:
 - 0.05 as a cut off for "small enough"
- Note: Sample size has a big effect on ability to detect statistical significance.
 - A large n makes it easier to have statistical significance
 - A small n makes it hard to have statistical significance



One sample t-test (Ex.)

Is the population mean female height 64 inches?

- 1. Estimate of mean female height from our sample
 - 65.6 inches (5' 6")
 - n=51
 - Std. Deviation= 5.1
 - Std. Error= 0.7206 $\left(\frac{Std.Dev.}{\sqrt{n}} = \frac{5.1}{\sqrt{51}}\right)$
- 2. Get a value for the test statistic
 - T-test statistic= 2.23 $\left(\frac{Our\ estimate\ -64}{Std.Error} = \frac{65.6-64}{0.72}\right)$
- 3. Look up a p-value
 - P-value= 0.03
- 4. Use p-value to see if there is "statistical significance"
 - Since the p-value is smaller than 0.05 it is "statistically significant"
 - We have evidence that the mean height for females is not 64 inches.

95% Confidence Interval (CI)

- Statistic (ex. Mean) is a point estimate
 - Probably not 100% correct
- Therefore also report a range of likely values
 - Calculated from the sample data
 - Likely to include the true population statistic
- 95% CI should contain the "true" mean 95% of the time
 - This method should work 95% of the time
 - If we took 100 different samples and calculated the CI for mean height each time we would capture the true mean 95% of the time
 - 95 of the 100 CI would contain the true mean
 - 5 would not

95% Confidence Interval (CI)

- Formula: Point estimate ± margin of error
 - Margin of error= Multiplier * Standard error
- Width of CI depends on:
 - Spread of the data (standard deviation)
 - n (Higher n = more confidence = narrower CI)
 - Level of confidence
 - Most people use 95% Cl

Confidence Interval (Ex.)

- CI for mean female height
 - (64.2, 67.1) inches
 - 65.6 ± 1.4 inches
- Based on our data it's almost certain that the mean height for females is between 64.2 inches and 67.1 inches
- Note: The interval does not include 64
 - This matches the result from the t-test!
 - T-test result: no evidence that the true mean is 64 inches
 - CI doesn't include 64 inches

Differences in Means

- Is mean height the same for males and females?
- Question: about the population
 - Answer: using our data
- We have two independent groups
 - Individuals from one population are not related in anyway to those in the other population
- How many groups do you have?
 - 2 Groups: Use two sample t-test
 - 3+ Groups: Use ANOVA

Two sample t-test

- Used when comparing two independent samples
- Are the two means equal?
 - Mean(Males) Mean(Females) = 0
- T distribution and t statistic
- $\frac{Sample\ statistic\ -comparison}{standard\ error} = \frac{Mean(M) Mean\ (F) 0}{standard\ error}$
 - Standard error pools the variance for each group

Difference in Means (Ex.)

- Is mean height the same for males and females?
 - 1. Our estimates
 - Mean Height males: 67.5 inches (n=49)
 - Mean Height females: 65.6 inches (n=51)
 - Difference: 1.9 inches
 - 2. Get a value for the two sample t-test statistic
 - 2.16
 - 3. Get a p-value
 - P-value= 0.03
 - 4. Use p-value to see if there is "statistical significance"
 - The p-value is smaller than 0.05 it's "statistically significant"
 - Therefore, we have evidence that the mean heights for males and females are different

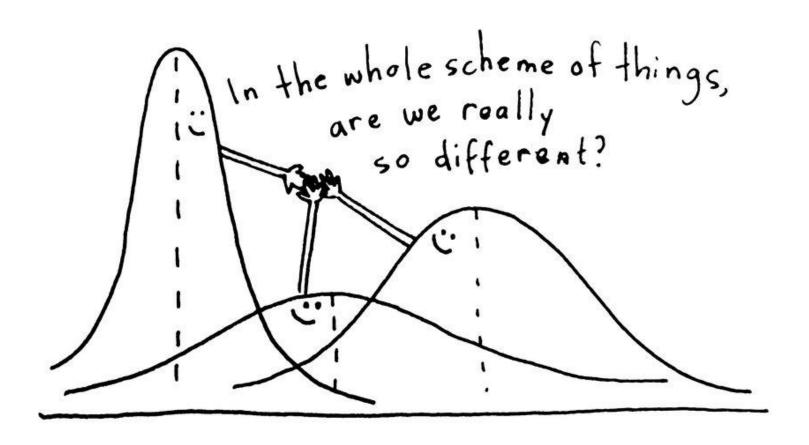
Difference in 3+ Means

- Is mean height the same for NPR listener groups?
 - Listen, Don't listen, What is NPR?
- Question: about the population
 - Answer: using our data
- We have independent groups
 - Individuals from one population are not related in anyway to those in the other population
- How many groups do you have?
 - 2 Groups: Use two sample t-test
 - 3+ Groups: Use ANOVA
- Our example has 3 groups: we'll use ANOVA

ANOVA (ANalysis Of VAriance)

- Extension of the two sample t-test
- Can handle 3+ groups
- Test if all the means are the same value
 - Or at least one different
- F distribution and F statistic
- F statistic= $\frac{Variation\ among\ sample\ means}{Natural\ variation\ within\ groups}$
 - Denominator pools variance info from each group

ANOVA (ANalysis Of VAriance)



ANOVA (Ex.)

- Is mean height the same for NPR listener groups?
 - 1. Our estimates
 - Mean height "Yes": 66.4 inches (n=32)
 - Mean height "No": 67.3 inches (n=33)
 - Mean height "What is NPR?": 65.8 inches (n=35)
 - 2. Get a value for the ANOVA F-test statistic
 - 0.98
 - 3. Get a p-value
 - P-value= 0.37
 - 4. Use p-value to see if there is "statistical significance"
 - The p-value is bigger than 0.05. It is not statistically significant
 - Therefore, we don't have evidence to say there is a difference in mean heights between NPR listener groups

Other kinds of data

- We could have used similar methods with proportions
 - Formulas change, same basic methods
 - Plan analysis
 - 2. Calculate estimates from sample
 - 3. Get a value for appropriate test statistic
 - 4. Get a p-value
 - Use p-value to see if there is "statistical significance" and interpret results

"Unwritten Standard"

96,

- Sir R. A. Fischer
 - Biologist and statistician
 - ANOVA and experimental design
- <u>Statistical Methods for Research Workers</u> (1925), he states
- ... it is convenient to draw the line at about the level at which we can say: "Either there is something in the treatment, or a coincidence has occurred such as does not occur more than once in twenty trials."... If one in twenty does not seem high enough odds, we may, if we prefer it, draw the line at one in fifty (the 2 per cent point), or one in a hundred (the 1 per cent point). Personally, the writer prefers to set a low standard of significance at the 5 per cent point, and ignore entirely all results which fail to reach this level. A scientific fact should be regarded as experimentally established only if a properly designed experiment rarely fails to give this level of significance.
- http://www.p-value.info/2013/01/whats-significance-of-005-significance_6.html