Unit 1: Introduction to data

2. Exploratory data analysis

Sta 101 - Spring 2015

Duke University, Department of Statistical Science

January 14, 2015

1. Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- Use side-by-side box plots to visualize relationships between numerical and categorical variables

- Sit in teams in class and lab going forward are you missing a team member?
- If you haven't yet done so, take the class survey
- ▶ Lab 1 due by your lab session next Tue one submission per team sufficient
- TA office hours:
 - SEC open 4pm 9pm Sunday Thursday
 - Sta 101 TA hours:
 - Christine Sun, 4-6pm
 - David Mon, 4-6pm
 - Anthony Mon. 7-9pm
 - Chris (Xinyi) Tues,
 5-7pm

- Radhika Tues, 7-9pm
- Mao Wed, 5-7pm
- Tori Wed, 7-9pm
- Fiamma Thur, 6-7pm
- Phillip TBA
- No class and no OH on Monday, review randomization test video before Wednesday's class
- Next week. No OH Tuesday. OH 3:30-5:30pm on Wednesday!

1. Housekeeping

2. Main ideas

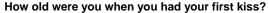
- 1. Always start your exploration with a visualization
- 2. When describing numerical distributions discuss shape, center, spread, and unusual observations
- 3. Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- 5. Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- 7. Use side-by-side box plots to visualize relationships between numerical and categorical variables

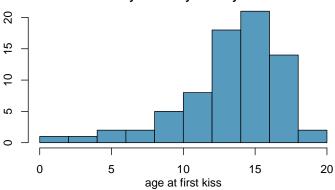
Housekeeping

2. Main ideas

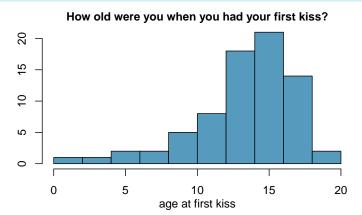
- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- 7. Use side-by-side box plots to visualize relationships between numerical and categorical variables

Do you see anything out of the ordinary?



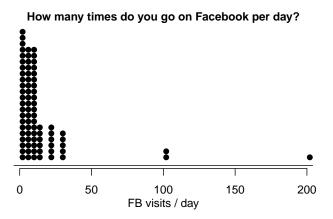


Do you see anything out of the ordinary?

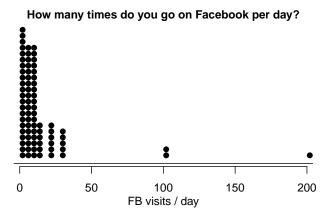


Some people reported very low ages, which might suggest the survey question wasn't clear: romantic kiss or any kiss?

How are people reporting lower vs. higher values of FB visits?

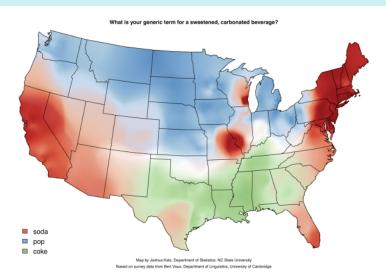


How are people reporting lower vs. higher values of FB visits?



Finer scale for lower numbers.

Describe the spatial distribution of preferred sweetened carbonated beverage drink.



What is missing in this visualization?



Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- 2. When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- Use side-by-side box plots to visualize relationships between numerical and categorical variables

Describing distributions of numerical variables

- Shape: skewness, modality
- Center: an estimate of a typical observation in the distribution (mean, median, mode, etc.)
 - Notation: μ : population mean, \bar{x} : sample mean
- Spread: measure of variability in the distribution (standard deviation, IQR, range, etc.)
- Unusual observations: observations that stand out from the rest of the data that may be suspected outliers

Which of these is most likely to have a roughly symmetric distribution?

- (a) salaries of a random sample of people from North Carolina
- (b) weights of adult females
- (c) scores on an well-designed exam
- (d) last digits of phone numbers

Which of these is most likely to have a roughly symmetric distribution?

- (a) salaries of a random sample of people from North Carolina
- (b) weights of adult females
- (c) scores on an well-designed exam
- (d) last digits of phone numbers

How do the mean and median of the following two datasets compare?

Dataset 1: 30, 50, 70, 90 Dataset 2: 30, 50, 70, 1000

- (a) $\bar{\mathbf{x}}_1 = \bar{\mathbf{x}}_2$, $\mathrm{median}_1 = \mathrm{median}_2$
- (b) $\bar{\mathbf{x}}_1 < \bar{\mathbf{x}}_2$, median₁ = median₂
- (c) $\bar{\mathbf{x}}_1 < \bar{\mathbf{x}}_2$, median₁ < median₂
- (d) $\bar{x}_1 > \bar{x}_2$, median₁ < median₂
- (e) $\bar{\mathbf{x}}_1 > \bar{\mathbf{x}}_2$, median₁ = median₂

How do the mean and median of the following two datasets compare?

Dataset 1: 30, 50, 70, 90 Dataset 2: 30, 50, 70, 1000

- (a) $\bar{\mathbf{x}}_1 = \bar{\mathbf{x}}_2$, median₁ = median₂
- (b) $\bar{\mathbf{x}}_1 < \bar{\mathbf{x}}_2$, median₁ = median₂
- (c) $\bar{\mathbf{x}}_1 < \bar{\mathbf{x}}_2$, median₁ < median₂
- (d) $\bar{x}_1 > \bar{x}_2$, median₁ < median₂
- (e) $\bar{x}_1 > \bar{x}_2$, median₁ = median₂

- Most commonly used measure of variability is the standard deviation, which roughly measures the average deviation from the mean
 - Notation: σ : population standard deviation, s: sample standard deviation
- Calculating the standard deviation, for a population (rarely, if ever) and for a sample:

$$\sigma = \sqrt{\frac{\sum_{i=1}^N (x_i - \mu)^2}{n}} \qquad s = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}}$$

▶ Square of the standard deviation is called the *variance*.

Lose a "degree of freedom" for using an estimate (the sample mean, \bar{x}), in estimating the sample variance/standard deviation.

Lose a "degree of freedom" for using an estimate (the sample mean, \bar{x}), in estimating the sample variance/standard deviation.

Why do we use the squared deviation in the calculation of variance?

Lose a "degree of freedom" for using an estimate (the sample mean, \bar{x}), in estimating the sample variance/standard deviation.

Why do we use the squared deviation in the calculation of variance?

- ➤ To get rid of negatives so that observations equally distant from the mean are weighed equally.
- To weigh larger deviations more heavily.

True / False: The range is always at least as large as the IQR for a given dataset.

- (a) Yes
- (b) No

True / False: The range is always at least as large as the IQR for a given dataset.

- (a) Yes
- (b) No

Range = max - min, IQR = Q3 - Q1

True / False: The range is always at least as large as the IQR for a given dataset.

- (a) Yes
- (b) No

Range = max - min, IQR = Q3 - Q1

Is the range or the IQR more robust to outliers?

True / False: The range is always at least as large as the IQR for a given dataset.

- (a) Yes
- (b) No

Range = max - min, IQR = Q3 - Q1

Is the range or the IQR more robust to outliers?

IQR

Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- 3. Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- 7. Use side-by-side box plots to visualize relationships between numerical and categorical variables

Robust statistics

- Mean and standard deviation are easily affected by extreme observations since the value of each data point contributes to their calculation.
- Median and IQR are more robust.
- ➤ Therefore we choose median&IQR (over mean&SD) when describing skewed distributions.

Application exercise: 1.2 Distributions of numerical variables

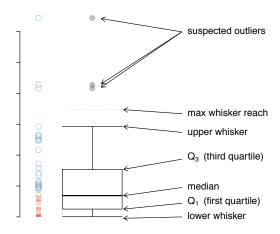
See the course website for instructions.

Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- 7. Use side-by-side box plots to visualize relationships between numerical and categorical variables

A box plot visualizes the median, the quartiles, and suspected outliers. An *outlier* is defined as an observation more than 1.5×IQR away from the quartiles.



Application exercise: 1.3 Boxplots

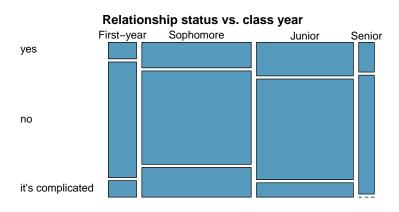
See the course website for instructions.

Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- 5. Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- 7. Use side-by-side box plots to visualize relationships between numerical and categorical variables

What do the widths of the bars represent? What about the heights of the boxes? Is there a relationship between class year and relationship status? What other tools could we use to summarize these data?



Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- 3. Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables

6. Be aware of Simpson's paradox

Use side-by-side box plots to visualize relationships between numerical and categorical variables

Race and death-penalty sentences in Florida murder cases

A 1991 study by Radelet and Pierce on race and death-penalty (DP) sentences gives the following table:

Defendant's race	DP	No DP	Total	% DP
Caucasian	53	430	483	
African American	15	176	191	
Total	68	606	674	

Adapted from Subsection 2.3.2 of A. Agresti (2002), Categorical Data Analysis, 2nd ed., and

http://math.stackexchange.com/questions/83756/examples-of-simpsons-paradox.

Race and death-penalty sentences in Florida murder cases

A 1991 study by Radelet and Pierce on race and death-penalty (DP) sentences gives the following table:

Defendant's race	DP	No DP	Total	% DP
Caucasian	53	430	483	11%
African American	15	176	191	
Total	68	606	674	

Adapted from Subsection 2.3.2 of A. Agresti (2002), Categorical Data Analysis, 2nd ed., and

http://math.stackexchange.com/questions/83756/examples-of-simpsons-paradox.

Race and death-penalty sentences in Florida murder cases

A 1991 study by Radelet and Pierce on race and death-penalty (DP) sentences gives the following table:

Defendant's race	DP	No DP	Total	% DP
Caucasian	53	430	483	11%
African American	15	176	191	7.9%
Total	68	606	674	

Adapted from Subsection 2.3.2 of A. Agresti (2002), Categorical Data Analysis, 2nd ed., and

http://math.stackexchange.com/questions/83756/examples-of-simpsons-paradox.

Race and death-penalty sentences in Florida murder cases

A 1991 study by Radelet and Pierce on race and death-penalty (DP) sentences gives the following table:

Defendant's race	DP	No DP	Total	% DP
Caucasian	53	430	483	11%
African American	15	176	191	7.9%
Total	68	606	674	

Who is more likely to get the death penalty?

Adapted from Subsection 2.3.2 of A. Agresti (2002), Categorical Data Analysis, 2nd ed., and

http://math.stackexchange.com/questions/83756/examples-of-simpsons-paradox.

Victim's race	Defendant's race	DP	No DP	Total	% DP
Caucasian	Caucasian	53	414	467	
Caucasian	African American	11	37	48	
African American	Caucasian	0	16	16	
African American	African American	4	139	143	
Total		68	606	674	

Victim's race	Defendant's race	DP	No DP	Total	% DP
Caucasian	Caucasian	53	414	467	11.3%
Caucasian	African American	11	37	48	
African American	Caucasian	0	16	16	
African American	African American	4	139	143	
Total		68	606	674	

Victim's race	Defendant's race	DP	No DP	Total	% DP
Caucasian	Caucasian	53	414	467	11.3%
Caucasian	African American	11	37	48	22.9%
African American	Caucasian	0	16	16	
African American	African American	4	139	143	
Total		68	606	674	

Victim's race	Defendant's race	DP	No DP	Total	% DP
Caucasian	Caucasian	53	414	467	11.3%
Caucasian	African American	11	37	48	22.9%
African American	Caucasian	0	16	16	0%
African American	African American	4	139	143	
Total		68	606	674	

Victim's race	Defendant's race	DP	No DP	Total	% DP
Caucasian	Caucasian	53	414	467	11.3%
Caucasian	African American	11	37	48	22.9%
African American	Caucasian	0	16	16	0%
African American	African American	4	139	143	2.8%
Total		68	606	674	

Victim's race	Defendant's race	DP	No DP	Total	% DP
Caucasian	Caucasian	53	414	467	11.3%
Caucasian	African American	11	37	48	22.9%
African American	Caucasian	0	16	16	0%
African American	African American	4	139	143	2.8%
Total		68	606	674	

Who is more likely to get the death penalty?

▶ People of one race are more likely to murder others of the same race, murdering a Caucasian is more likely to result in the death penalty, and there are more Caucasian defendants than African American defendants in the sample.

- ▶ People of one race are more likely to murder others of the same race, murdering a Caucasian is more likely to result in the death penalty, and there are more Caucasian defendants than African American defendants in the sample.
- Controlling for the victim's race reveals more insights into the data, and changes the direction of the relationship between race and death penalty.

- ▶ People of one race are more likely to murder others of the same race, murdering a Caucasian is more likely to result in the death penalty, and there are more Caucasian defendants than African American defendants in the sample.
- Controlling for the victim's race reveals more insights into the data, and changes the direction of the relationship between race and death penalty.
- ➤ This phenomenon is called *Simpson's Paradox*: An association, or a comparison, that holds when we compare two groups can disappear or even be reversed when the original groups are broken down into smaller groups according to some other feature (a confounding/lurking variable).

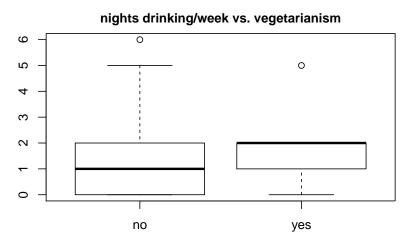
Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- 7. Use side-by-side box plots to visualize relationships between numerical and categorical variables

3. Summary

How do drinking habits of vegetarian vs. non-vegetarian students compare?



Housekeeping

2. Main ideas

- 1. Always start your exploration with a visualization
- When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
 - 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
 - 6. Be aware of Simpson's paradox
- Use side-by-side box plots to visualize relationships between numerical and categorical variables

3. Summary

- 1. Always start your exploration with a visualization
- 2. When describing numerical distributions discuss shape, center, spread, and unusual observations
- Robust statistics are not easily affected by outliers and extreme skew
- 4. Use box plots to display quartiles, median, and outliers
- Use mosaic plots for visualizing relationship between two categorical variables
- 6. Be aware of Simpson's paradox
- 7. Use side-by-side box plots to visualize relationships between numerical and categorical variables