# Mixing Active Learning and Lecturing: Using Interactive Visualization as a Teaching Tool JSM 2018

Jessica Minnier, PhD & Ted Laderas, PhD Oregon Health & Science University

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Slides available at http://bit.ly/jsm-minnier

## Setting

## **OHSU Data Science Institute**

- 2 Day workshop
- 3 Hours for "Introduction to Statistics and Data Exploration"
- Aim of DSI: "bring together researchers, librarians, and information specialists for formal training on key topics in data science"

## Audience

- Librarians, information scientists, researchers
- Very little mathematical/programming background
- Heterogeneous background in science and research

## Goals

## Statistical Concepts

- Start with the didactics
- Use interactive visualizations to illustrate statistical concepts

## **Data Exploration**

- Empower students to explore data (no fear!)
- Encourage understanding of relationships of data

## Interactivity

- Interactive plots for exploration of multi-variable relationships
- Include some coding exercises (as bonus material)

## Methods

## Approach

- Implement as a Tutorial, but used with didactic teaching
- : uses Shiny to build interactive R Markdown style workbooks
- Can be deployed as a website, or on student's computer (requires R/Rstudio)

## **Practicalities**

- Categorical data session and continuous data session
- Hosted on github as a package on Github (https://github.com/laderast/dsiexplore)
- Hosted workbooks on shinyapps.io for real time interactivity

## Interactivity

- Didactic lessons embedded in workbooks with interactive components
- Interactive sliders, dropdown options allow interaction with data filtering and analysis
- Interactive code teaches effect of changing code components on visualizations/analyses

#### Categorical Data

Ted Laderas and Jessica Minnier 06 November, 2017

Learning	<b>Objectives</b>	for this	Session
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What is Exploratory Data Analysis?

What are important associations?

The Data (Whickham)

Let's look at our Outcome

Two Variables Walk Into A Bar

The 2x2 table

The effect of Age on the Data

Not as easy as we thought!

**Chi-Squared Test** 

Your Turn

Your Turn: Calculate your p-value

Missing Values

Be proud of yourself!

References

Start Over

#### Learning Objectives for this Session

Please fill out the pre-assessment before you do anything! https://goo.gl/forms/jRa7TRFWXp5rlF6H3

Make sure to buddy up, and use your post-its!

At the end of this session you should be able to

- Understand why we need to do Exploratory Data Analysis (EDA)
- Understand statistical association between categorical variables
- · Visual ways to identify association
- · Basic understanding of Chi-squared statistics
- · Understand how to assess missing data

**Next Topic** 

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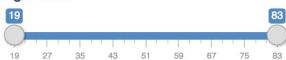
Start Over

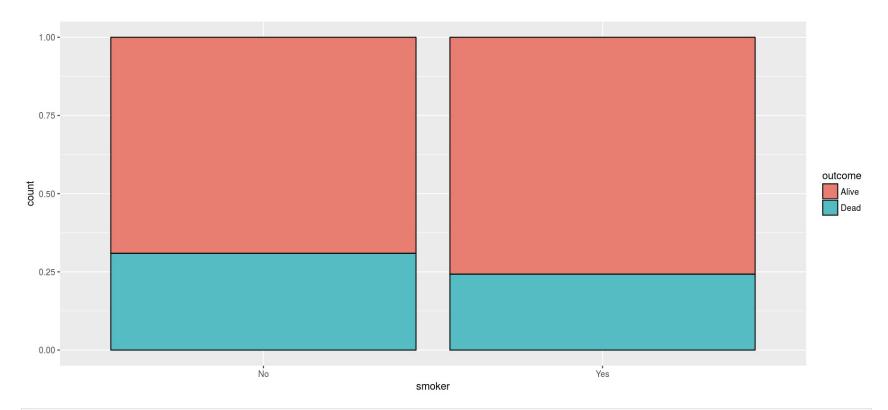
#### Not as easy as we thought!

So as you get older, you're more likely to die. This may be messing up our overall results!

Let's ask the question again, with a younger group: are smokers under 60 more likely to die than non smokers?

#### **Age Cutoff**





For patients who are under 60, is smoking associated with death?

Yes, the proportion of smokers who die is greater than the proportion of non-smokers who die for those patients younger than 60 years.

No, the proportion of smokers who die is smaller than the proportion of non-smokers who die for those patients younger than 60 years.

**Submit Answer** 

0

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#### P-values

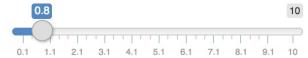
There is a straightforward interpretation to the *p-value*, and it has to do with how unique or rare our case is compared to our distribution of randomly generated cases.

So the *p-value* is interpreted as the probability that we will see a random case with the same exact statistic or higher.

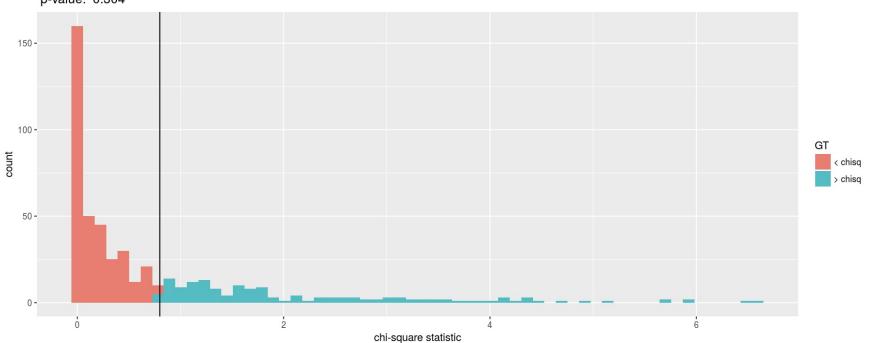
For example, if we had 10,000 random cases, and our p-value was 0.2, that means that of our 10000 cases, we would expect to see 10000 \* 0.2 = 2000 random cases with our statistic or greater.

Try adjusting the value of the chi-square statistic and see how many random cases are expected to have that statistic or higher.

#### slide to adjust statistical cutoff



Chi-square statistic: 0.8 % above chi-square: 30.4 p-value: 0.304



#### Continuous Data

Ted Laderas and Jessica Minnier 11/06/2017

Learning Objectives for this Session

EDA with continuous variables

How do we assess associations

between two continuous variables?

What is a factor that may be associated

with BMI?

Missingness and suspicious data elements

T Test

Explore other factors

Advanced Topic: Linear Models -

multiple predictors/associations

Your Turn

Resources and extra practice

Post-session survey

#### Correlation

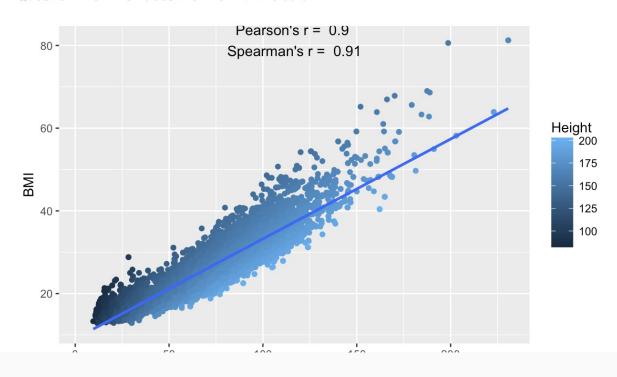
A simple statistical quantification of the association of two continuous variables is the **Pearson's Correlation Coefficient** (often labeled r).

Correlation = a quantity measuring the extent of interdependence of variable quantities

Pearson's correlation coefficient: a measure of the linear correlation between two variables

- · Note that this is quantifying a linear relationship.
- Value between -1 and +1, with 0 denoting no linear correlation
- · We can visually represent the linear relationship with a line through the scatter plot.
- If the relationship is relatively curved or exponential Pearson's correlation will not capture this relationship.
- An alternative might be the Spearman's correlation which essentially is the Pearson's correlation of the ranks. This evaluates monotone
  relationships.

Question: How well does the line "fit" the data?





## Continuous Data Ted Laderas and Jessica Minnier 11/06/2017 Learning Objectives for this Session

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#### ✓ Correlation explorer

Now you can try to get a feel for what correlation (linear and non-linear) looks like. Try a few pairs:

(For fun sometime, play the "guess the correlation" game at guessthecorrelation.com)

#### X-axis

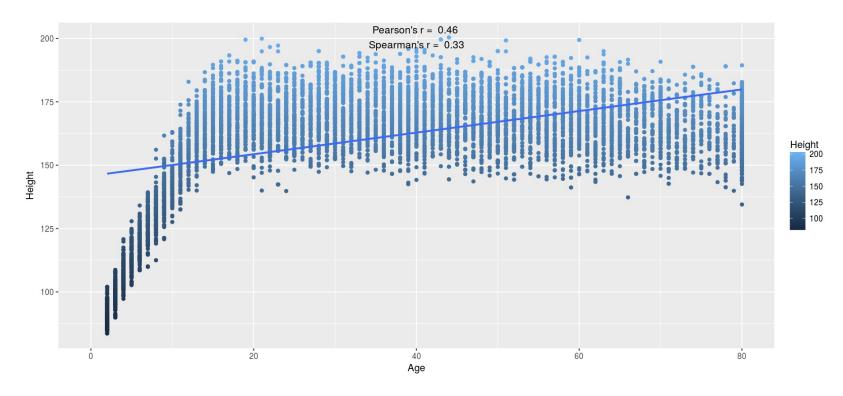


#### Y-axis



#### Color





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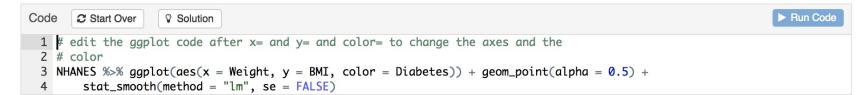
Resources and extra practice

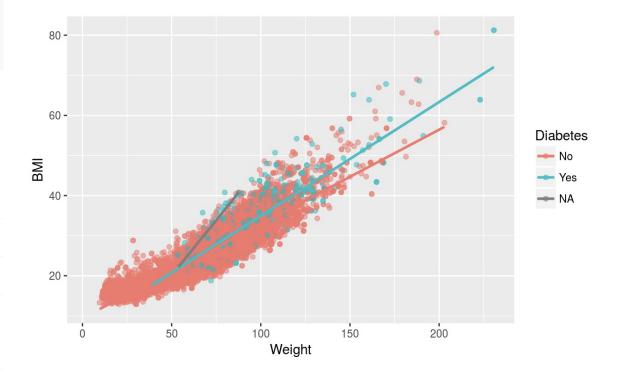
Post-session survey

Start Over

#### Practice Coding

If you want to practice coding a scatter plot, try editing ggplot2 code below to show a scatter plot of Age vs Height, colored by Gender:





Previous Topic

Next Topic

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However! The T-test is pretty robust to slight violations of the normality assumption, especially since we have a large sample size

• statistics side note: thanks to the Central Limit Theorem, our test is still *valid* as in we preserve our type I error; for a nice explanation of this see this Stats Geek blog post and Lumley T, et al 2002)

So, let's run a t-test (yay!) to assess the difference in means of BMI comparing diabetics and non-diabetics:

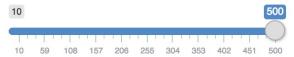
Difference in Means	Means No	Means Yes	T Statisitic P Value
-6.4	26.16	32.56	-20.83 3.9e-78

Note the p-value is extremely small. This is because we have a very large sample size and the difference in means is pretty large.

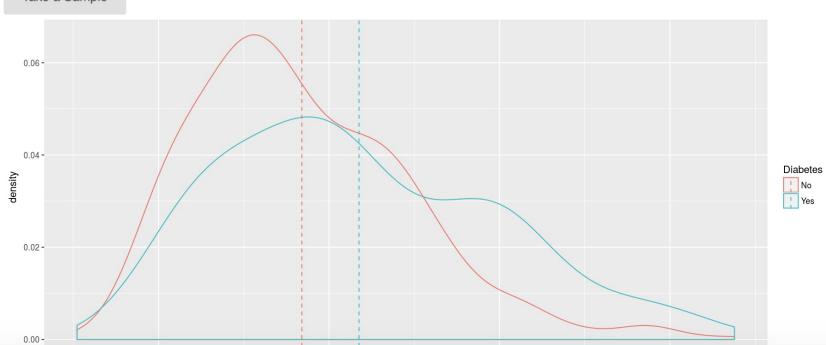
#### Smaller sample size

What happens if we have a much smaller sample size? We can examine the effect of sample size by randomly sampling a subset of the data. Look at our test statistic and p-value, as well as the difference in means.

#### **Total Sample Size**



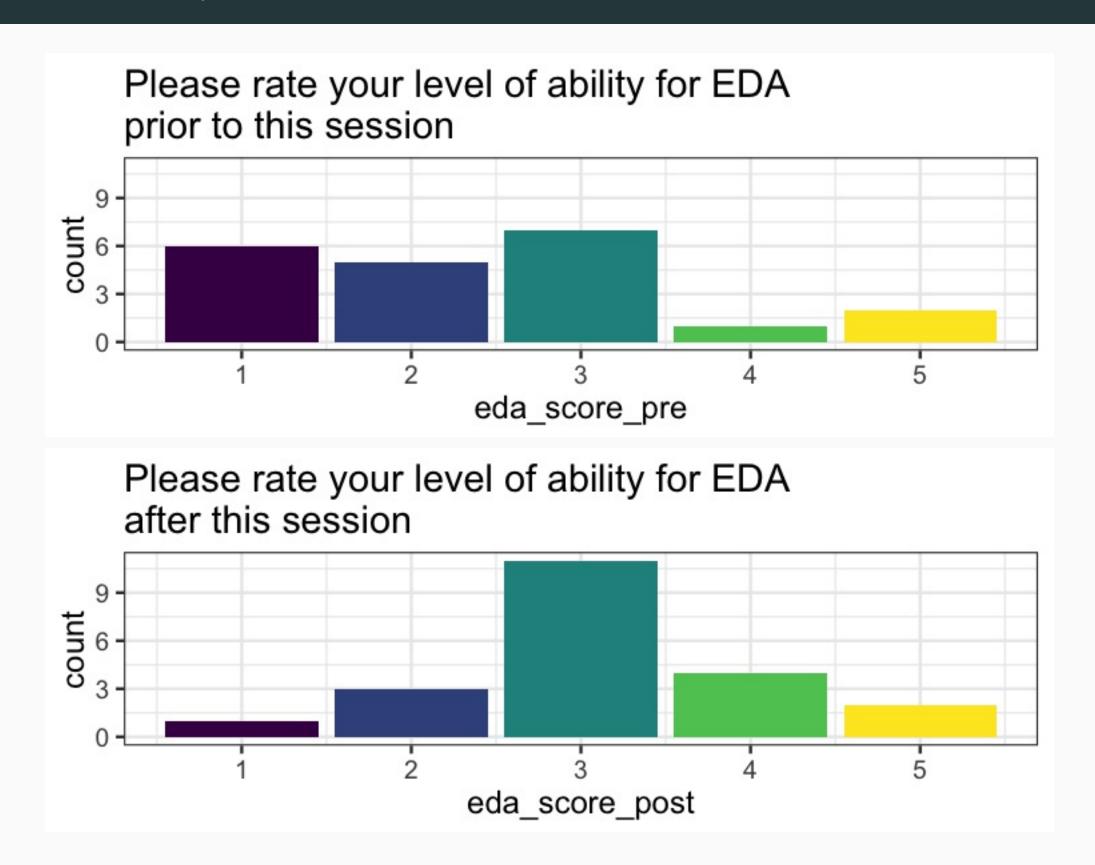
#### Take a Sample



## Conclusions & Results

- LearnR package + Shiny in R → interactive workbooks
- Students were empowered to learn
- Students liked the visualizations
  - "Very well done and methodical treatment the sliders were great!"
- Students felt engaged with the subject
  - "Explanation of key statistical concepts was effective and really made me want to learn more."
- Pre/Post-workshop survey: 95% of learners felt they gained practical knowledge (n=22)

## Survey Results



## **Impact**

### Pros:

- Accessible to beginners
- Mathematical concepts are more memorable
- Sparks discussions
- Empowers and engages students in scientific discovery/analysis

## Cons:

- Advanced students may require more challenging activities
- Visualizations must be tested for effectiveness
- Requires programming skills to implement

## Future Work and Adaptations

- Expand materials with more advanced statistical concepts
- Longer workshops \(\longrightarrow\) more interactive material, more topics
- Determine which interactive explorations are most effective

## Introduction to Visualization/Data Literacy

- Extension of this work: https://tladeras.shinyapps.io/dataLiteracy/
- HMSP410, Health Informatics for OHSU-PSU School of Public Health (co-taught by Ted Laderas and Bill Hersh)

## **Further Information**

- eCOTS e-poster: https://www.causeweb.org/cause/ecots/ecots18/posters/3-03
- Categorical Data: https://tladeras.shinyapps.io/categoricalData/
- Continuous Data: https://minnier.shinyapps.io/ODSI\_continuousData/
- LearnR package: https://rstudio.github.io/learnr/
- DSIexplore LearnR package: https://github.com/laderast/DSIExplore

## Thank you!

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Ted Laderas, PhD ♥ laderas, ♠ laderast ♠ https://laderast.github.io/
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Contact me: minnier-[at]-ohsu.edu, datapointier, iminnier

Slides available at http://bit.ly/jsm-minnier

Code for slides available at https://github.com/jminnier/talks\_etc

Slides created via the R package xaringan by Yihui Xie with the metropolis theme