

# Bringing Visual Inference to the Classroom

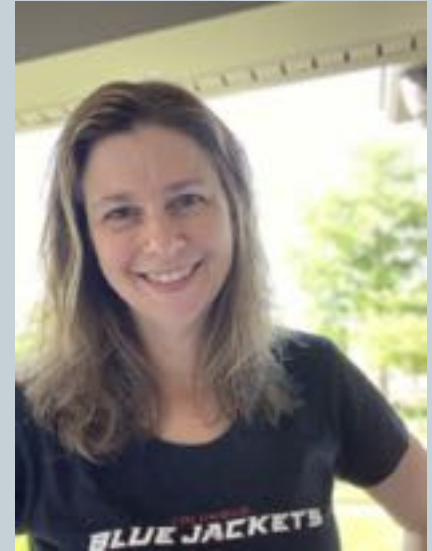


Adam Loy

# CAUSE/JSDSE webinar series



**Welcome from our  
host and moderator**



Leigh Johnson

# What's new in the journal?



## Data Science

Article

### Data Detectives: A Data Science Program for Middle Grade Learners >

Jacoya Thompson & Golnaz Aristoopour Irgens

Accepted author version posted online: 14 Feb 2022

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Views

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Crossref citations

0  
Abstracts

## Research Articles

Article

### Statistical Skills Gaps of Professors of Education at U.S. Universities and HBCUs >

Kimberlee C. Everson

Accepted author version posted online: 14 Feb 2022

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Abstracts

## Rapid Communication

Brief Report

### Alternate Forms of the One-Way ANOVA F and Kruskal-Wallis Test Statistics >

Rogel W. Johnson

Accepted author version posted online: 14 Feb 2022

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Views

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Crossref citations

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Abstracts

# March CAUSE/JSDSE Webinar



Details coming soon!

Signup at <https://www.causeweb.org/cause/webinars>

Webinars are recorded and posted (with slides) at that same site

# Consortium for the Advancement of Undergraduate Statistics Education



<https://www.causeweb.org/cause>

## **Save the date!**

May 23rd - 26th, 2022, with  
pre-conference workshops May 19th - 20th  
<https://causeweb.org/cause/ecots/ecots22/about>

Regional Conferences proposals are due today!!

Virtual posters & beyond and birds-of-a-feather  
proposals are due on March 13th, 2022



# Consortium for the Advancement of Undergraduate Statistics Education



<https://www.causeweb.org/cause>

**USPROC** | The Undergraduate Statistics Project Competition

More info at <https://www.causeweb.org/usproc>

# Adam Loy



Adam Loy is an assistant professor of statistics at Carleton College in Northfield, Minnesota. He received his PhD in statistics from Iowa State University. Adam's research interests include incorporating realistic computation and visualization into the classroom, exploring the potential of visual inference, developing better visualizations to explore complex models, and developing useful and useable R packages.



# Bringing Visual Inference to the Classroom



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# The move to a simulation-based curriculum



Since 2007, we've seen a shift to simulation-based inference in the intro course

Validation studies (Tintle et al. 2014; Maurer & Lock 2014; Hildreth et al. 2018)

Implementation in other courses

- Statistical inference (Cobb 2011; Chihara & Hesterberg 2011)
- Throughout curricula (Tintle et al. 2015)

**All have similar approach to visualization of the inferential process**

# Do Distracting Colors Influence the Time to Complete a Game?



20 students randomly assigned to the standard game

20 students a game with a color distracter

Subjects played the game in the same area with similar background noise

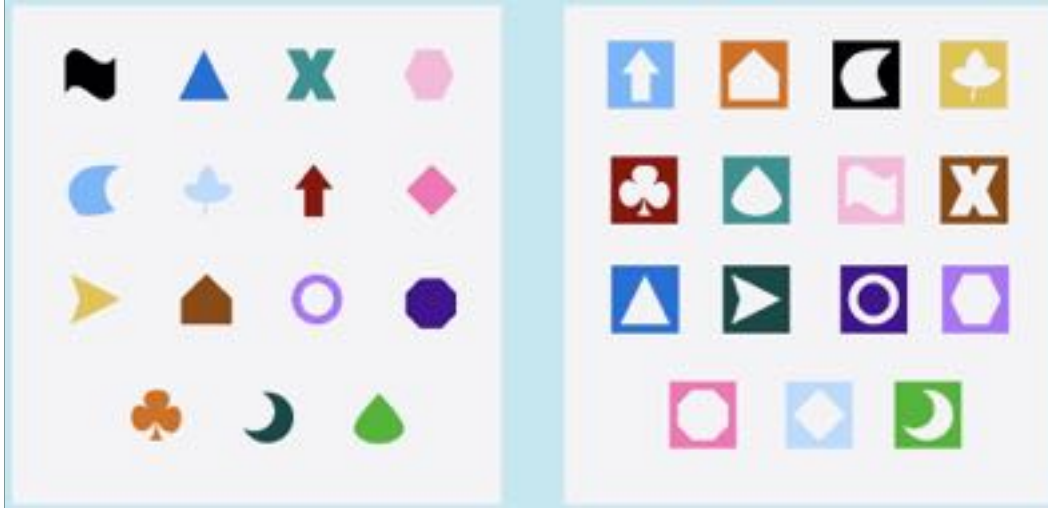
Collected the the time, in seconds, required to complete the game

Example taken from Kuiper and Sklar (2013)

<https://stat2games.sites.grinnell.edu>



Standard game



Color distracter

# Initial group discussion

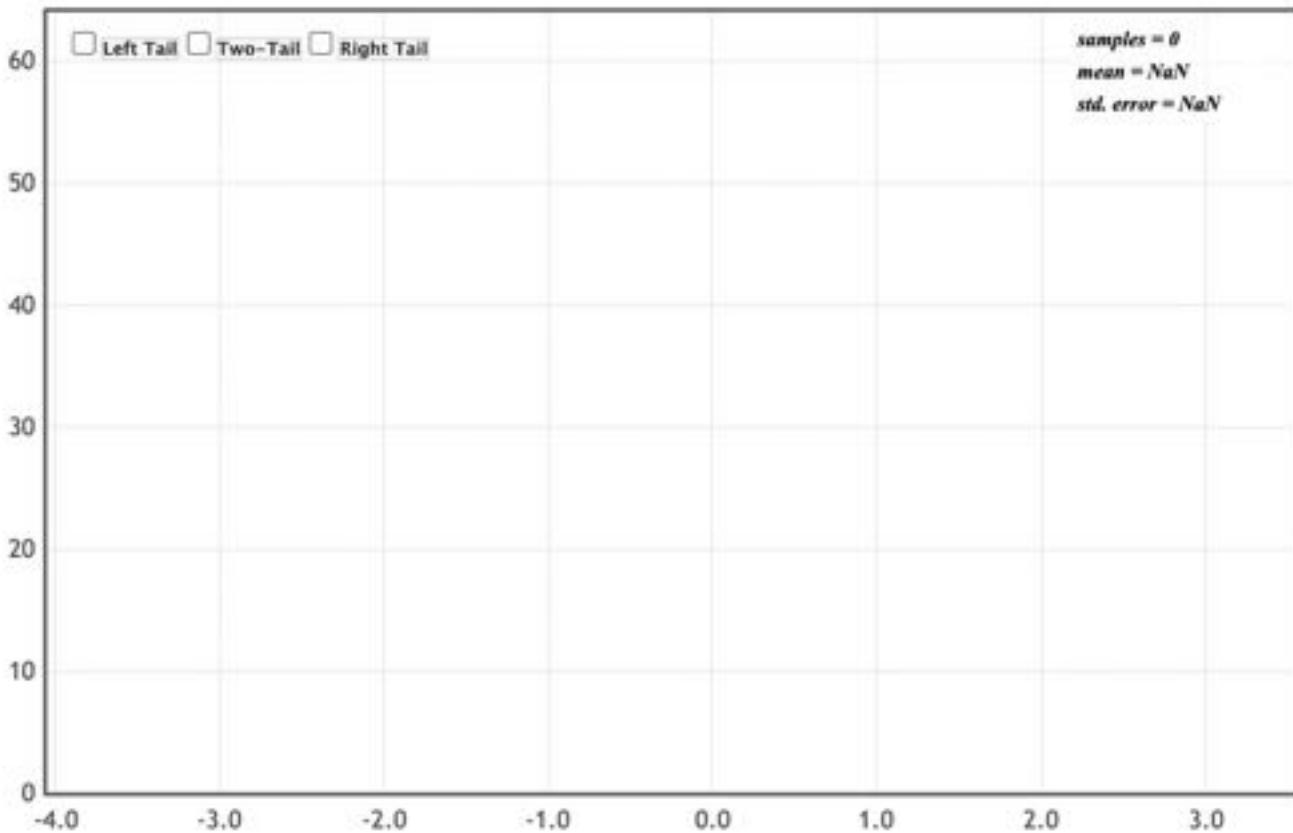


What competing claims are being investigated in this study?

What do the sample data have to say?  $\bar{x}_1 - \bar{x}_2 = 2.55$

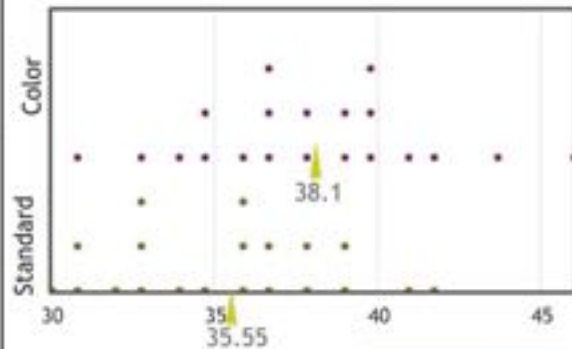
# The gap between apps and understanding

Randomization Dotplot of  $\bar{x}_1 - \bar{x}_2$ , Null hypothesis:  $\mu_1 = \mu_2$



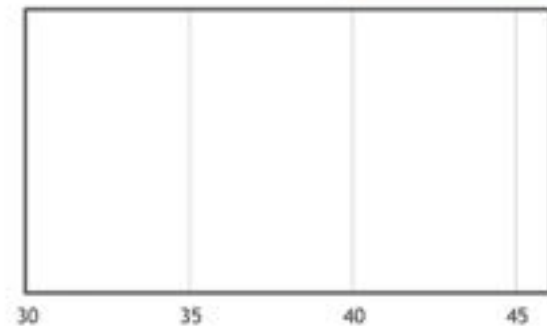
Original Sample

$\bar{x}_1 - \bar{x}_2 = -2.55$ ,  $n_1 = 20$ ,  $n_2 = 20$

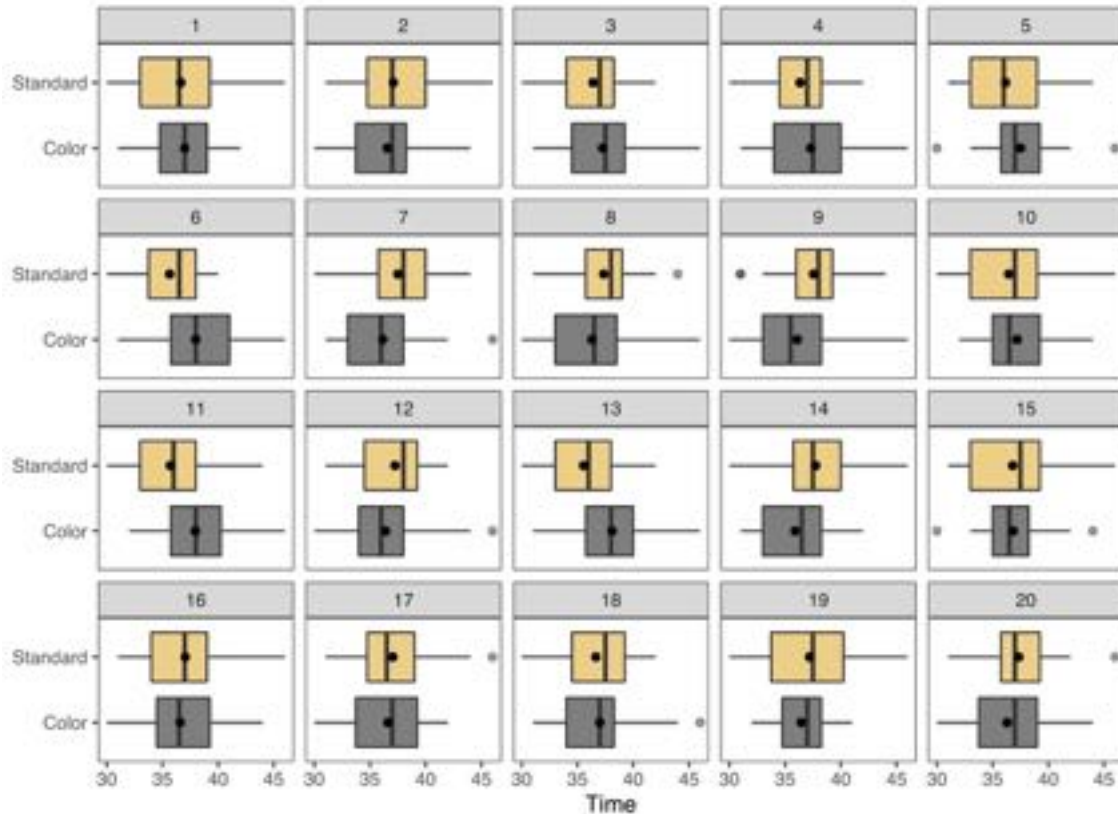


Randomization Sample

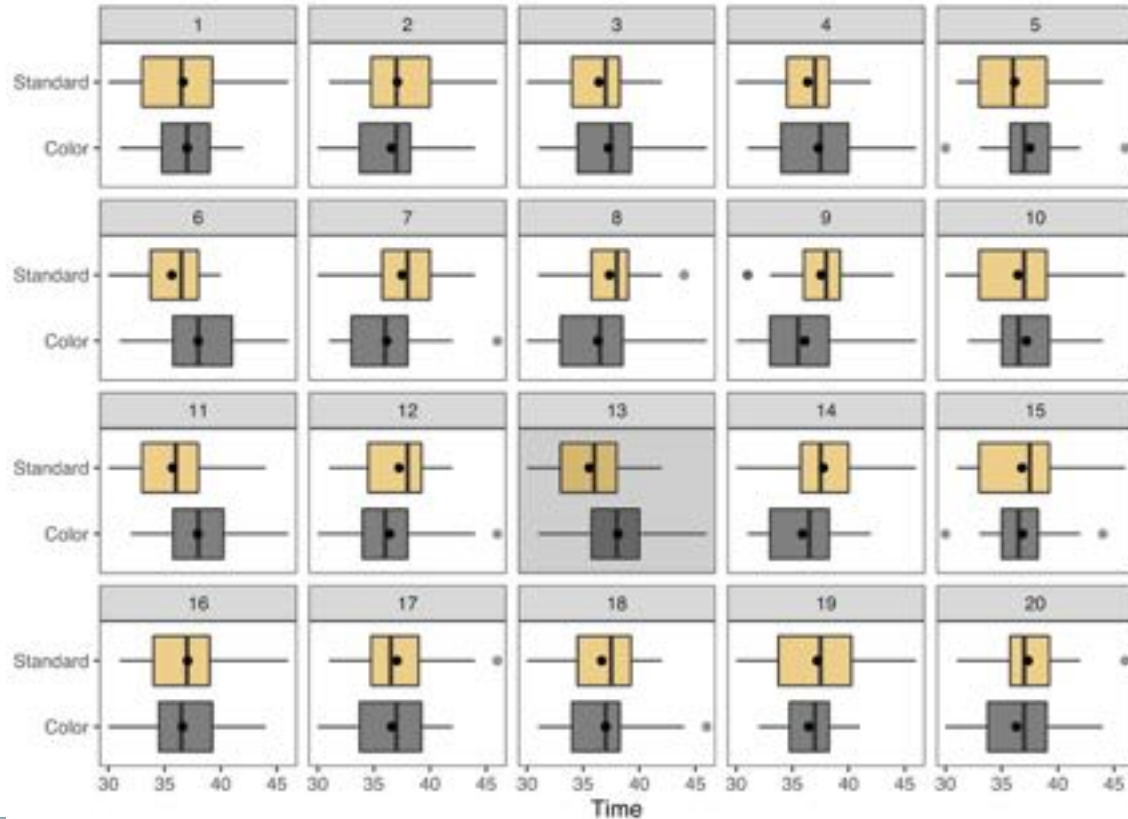
[Show Data Table](#)



Choose which plot is most different from the others and justify your choice



Choose which plot is most different from the others and justify your choice



# What did we just do?



We compared the **data plot** with **decoy (null) plots** of samples where, by construction, there is no association

This forces us to make decisions by comparing what we observe to what we would expect under the null

All of this is done using "Sesame Street logic"



# How do I use it in class?



Brief overview of logic of hypothesis testing → group discussion

Quick discussion of how decoy plots are created

→ individual evaluation of lineup

→ group discussion

Reveal observed panel

→ group discussion of implications of identification/no identification

Debrief

## Simulation-based inference

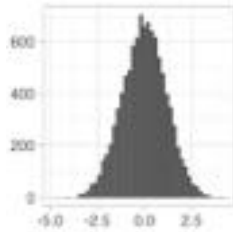
Hypotheses

$H_0$ : equal means

Test statistic

$$T(x) = \bar{x}_1 - \bar{x}_2$$

Reference distribution

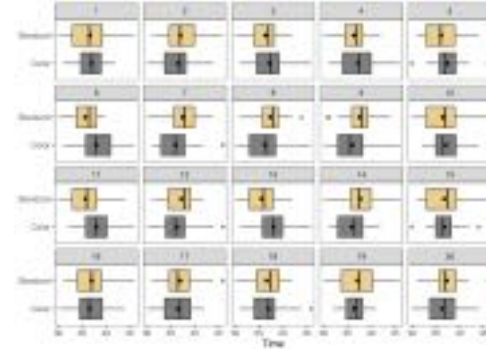
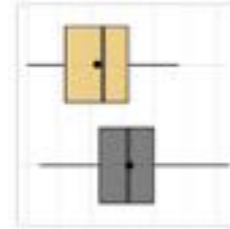


Evidence against null if...

Far out in the tail(s)

## Lineup

$H_a$ : mean larger for color distractor



Identifiable

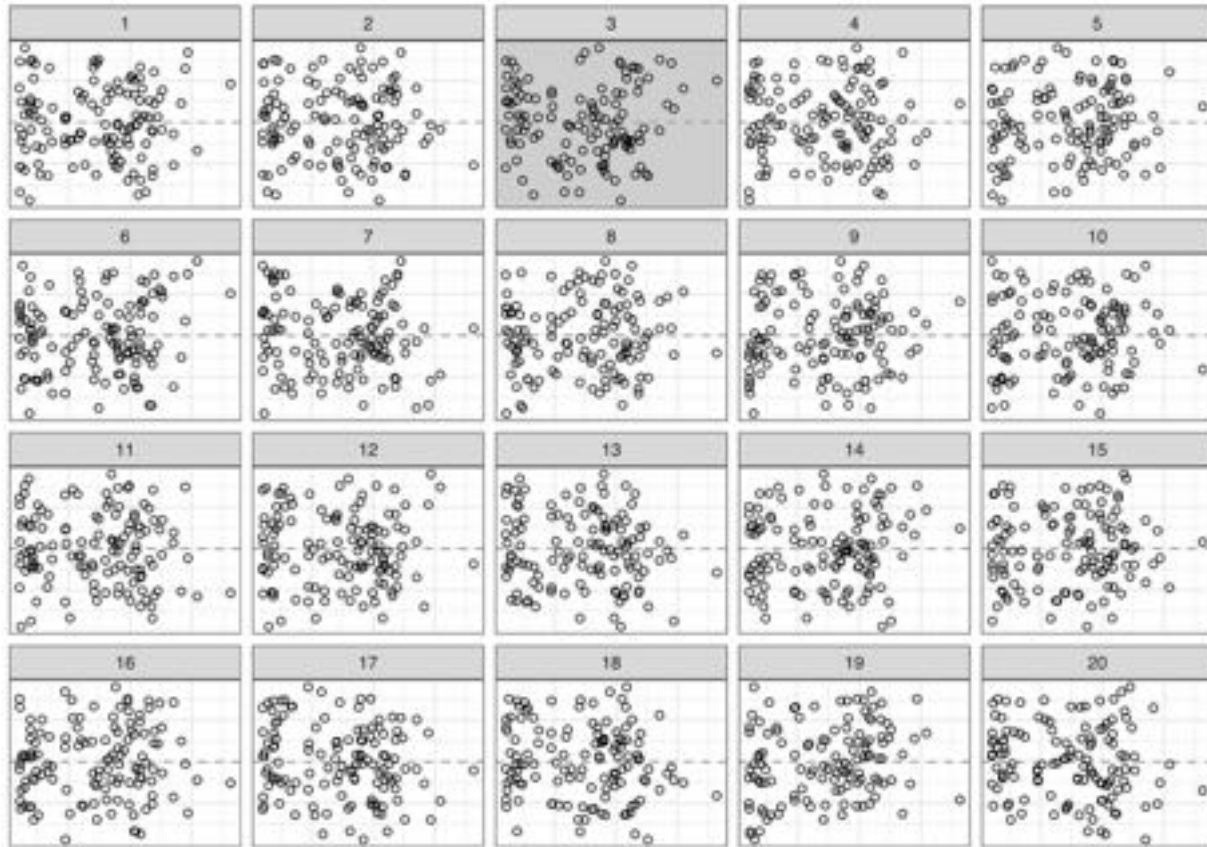
# Where else is the lineup protocol useful?



## Apophenia

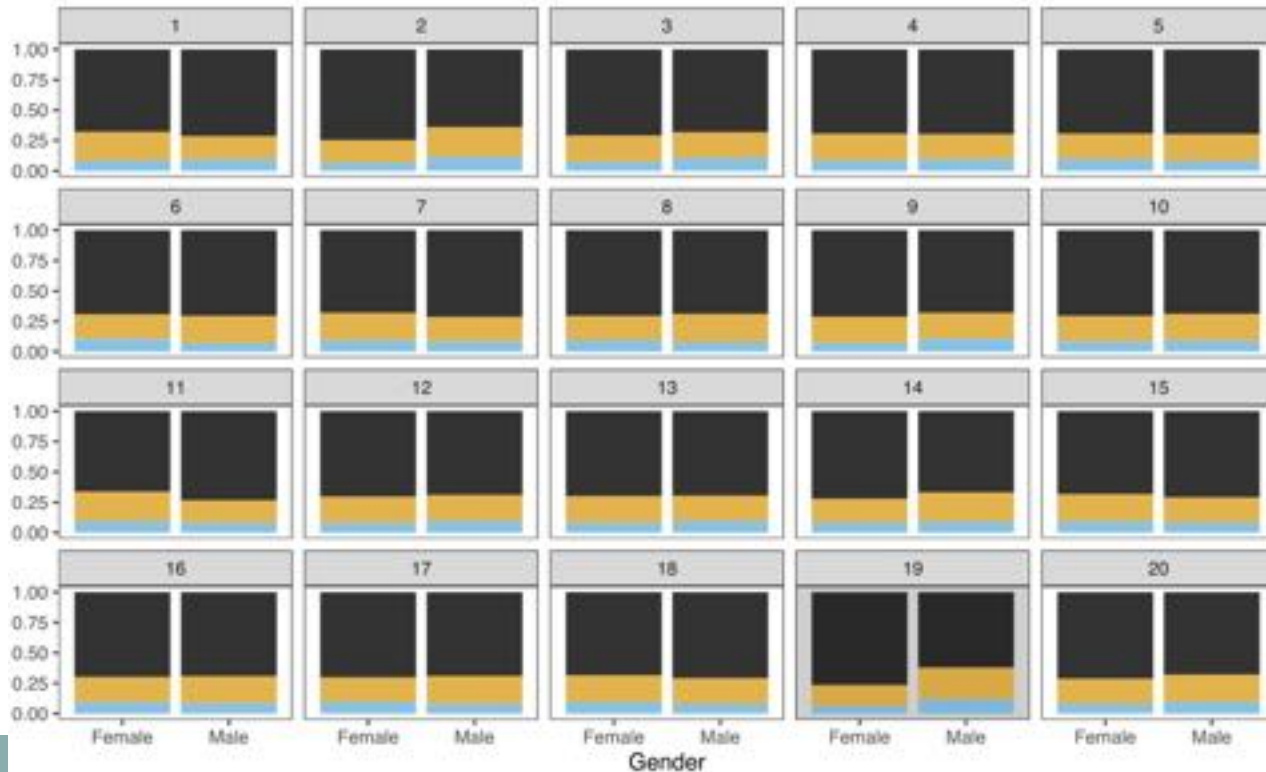
“the tendency to perceive a connection or meaningful pattern between unrelated or random things (such as objects or ideas)”

# Does the observed residual plot stand out?

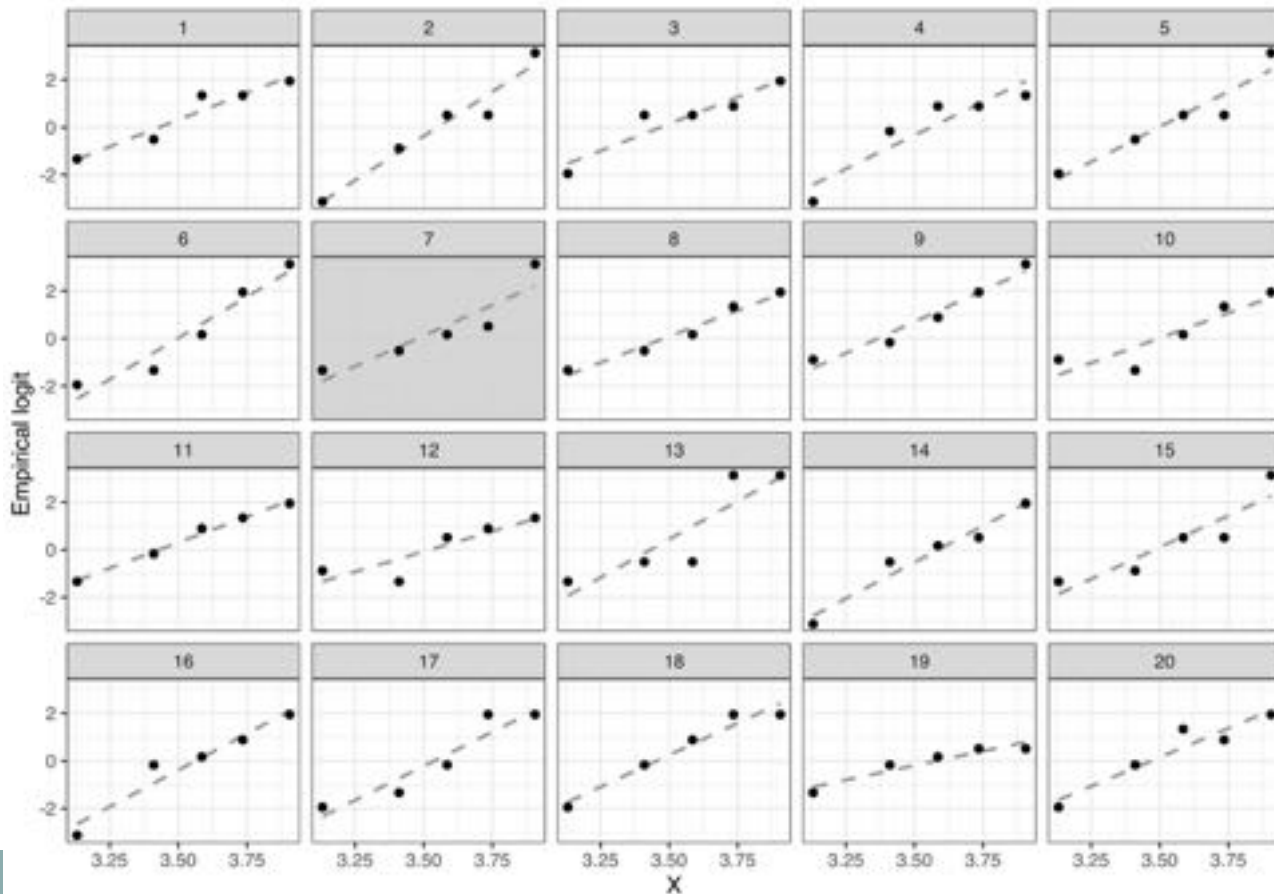


# Is it rude to bring a baby on a plane?

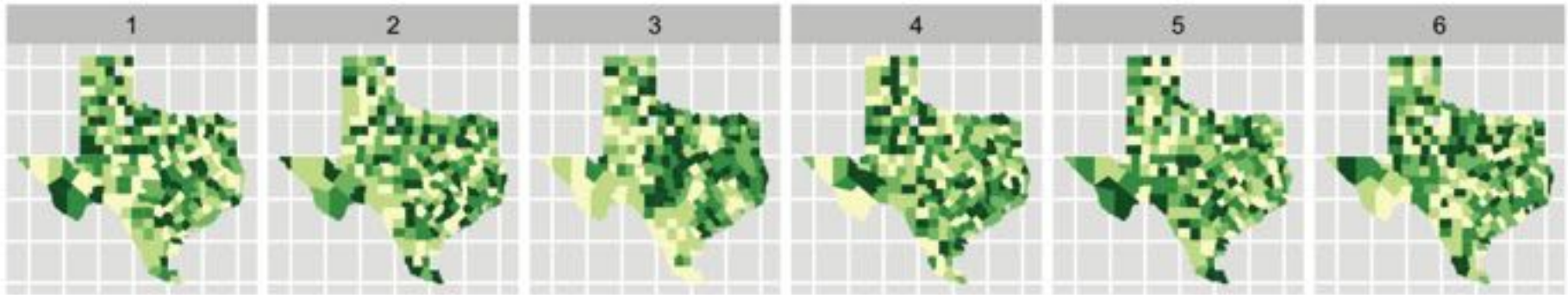
Is it rude to bring a baby on a plane?  
■ No, not at all rude   ■ Yes, somewhat rude   ■ Yes, very rude



# Are the empirical odds linear?



# Is there spatial association in this chloropleth map?



# Conclusions



Lineup introduces students to logic behind testing without need for technical discussions

Lineup provides a framework to help students interpret new statistical graphics

Lineup is a rigorous tool for statistical investigation later in the curriculum



# Classroom support



<https://aloy.github.io/classroom-vizinf/>

Two activities + instructor guide

Tutorial on creating lineups in R

Shiny apps to create lineups for class activities