

# Visualizing Data

# Recap

- We have spent **a lot** of time working with `ggplot2`
- Everything covered this week can be applied with `ggplot2`
- Goal of this week is to get you to stop and think about your plots *before* you make them. What are the kinds of things you should be considering?

# Stuff you already know

Data visualization is both an art *and* a science

Art:

- Aesthetically pleasing
- The story you're trying to tell

Science:

- Graphical representation of specific types of data
- The story you're trying to tell

# Warning

**Making figures for academic purposes  $\neq$  making figures for pure data viz purposes**

If you go into data science (and aren't constrained by academia), you'll want to check out R shiny (for interactive plots), web design, typography, etc.

# Today

- What kinds of things should you be thinking about when it comes to data viz *for academic papers?*
- What **not** to do (I'm gonna rant a bit)
- Hopefully helpful resources (no memorizing!)

# What should we be thinking about?

1. Telling your story
2. Contrast
3. Accessibility

# Telling your story

- Raw data are not intuitive. For the most part, you can't look at a spreadsheet of numbers and decipher any patterns. Especially with really big spreadsheets!
- We need a way to graphically show the data so that our human eyes can try to make sense of the data.
- It is so easy to **LIE** with data! Balancing act of conveying your message and not lying.
- As our datasets become more complex and high dimensional, data visualization can become more challenging.
- The goal is NOT to show every bit of data you have collected. It's to show the relationship you care about in an honest manner.

# What makes a good figure?

- Clear, descriptive title
- Axes are clearly labeled with variables and units of measurement
  - **Label. Your. Damn. Axes.**
- Scale is:
  - consistent across axis
  - easily interpreted
  - chosen so that data are evenly distributed (remember restricted range from correlation?)
- Data points are represented clearly, with a good key/legend if needed
- Graph is the *appropriate* type for your data (nominal, ordinal, interval, ratio etc.)

# Appropriateness

There are no 100% right answers, but there are wrong ones...

- Pie charts are never the answer; 3D pie charts are the worst of the worse
  - Instead, try stacked bar plot or even better, stacked dot plot
- 3D bar plots are never the answer
- Typically, you should show either the raw data or at least a distribution
  - Shelly is anti bar plot, generally
  - Shelly is anti box plots on their own (nice when combined)
  - Shelly is skeptical of lines -- need to be careful

# Why Bar Plots Can Suck

Stephanie Spielman, PhD @stephspiel · Nov 27, 2020

This visualization perfectly displays why barplots with error bars are often egregiously misleading. #dataviz

Nature Research @nresearchnews · Nov 27, 2020

Many children are willing to make personal sacrifices to punish wrongdoers and even more so if they believe punishment will teach the transgressor a lesson, according to a study in @NatureHumBehav. go.nature.com/3nUWtma

nature  
human behaviour

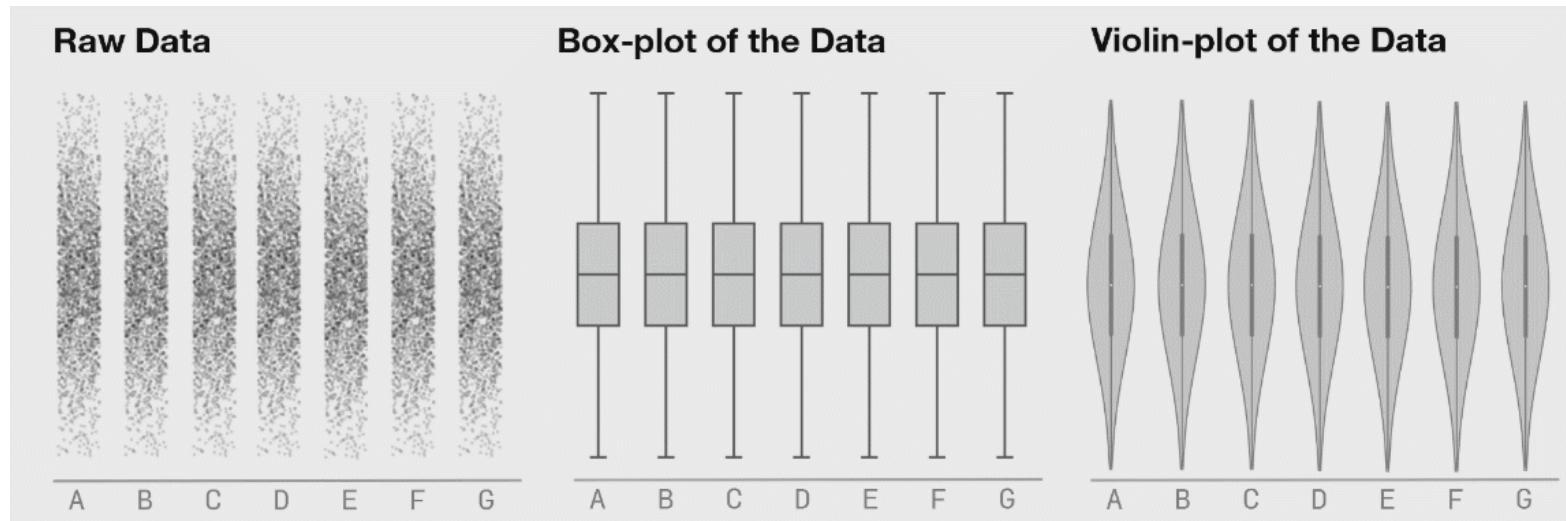
Condition	Strength of Consequentialist motive	Strength of Retributive motive
Baseline control	~10%	~10%
Non-communicative	~40%	~40%
Communicative	~75%	~75%

Stephanie Spielman, PhD @stephspiel · Nov 27, 2020

I must also add: Big props to authors for overlaying the jitter plot here. The combined visualization does make the actual underlying data distribution extremely clear (while also revealing the flaws of barplots as standalone viz).

- Different alternatives exist, with dot plots being the best option probably
- Bars are more appropriate when you have proportions or counts; but even still -- dotplots
- Tracey Weissgerber has an excellent thread with many resources on why barplots suck and how to maximize the utility of dotplots:  
<https://tinyurl.com/3e6222a4>

# Why box plots on their own can suck



<https://www.autodesk.com/research/publications/same-stats-different-graphs>

# Why adding lines can be misleading



# Lines Cont...

 **Бай Михал** @BajMihal · Jul 5, 2020

Replying to [@AmihaiGlazer](#)

And that explains everything:

**Academic Positions**

University of California, Irvine, Professor of Economics 1988-2020; Associate Professor of Economics, 1984-88; Assistant Professor of Economics, 1979-84  
Chair, Department of Economics, University of California, Irvine, 1995-1998  
Director, Focused Research Program in Public Choice, University of California, Irvine, 1989-1992  
Visiting Professor of Economics, Graduate School of Industrial Administration, Carnegie Mellon University, 1991-1992  
Visiting Senior Research Associate, Graduate School of Business, Stanford University, Fall 1985

**Education**

Ph.D. Yale University (1978), Economics  
M.A. Yale University (1975), Economics  
B.A. Cornell University (1974, Phi Beta Kappa), Economics

**Grants**

Gift from Charles Koch Foundation to support Program in Corporate Welfare Studies, March 2017, \$251,000  
Gift from Troesh Family Foundation to support Program in Corporate Welfare Studies, December 2017, \$150,000  
Gift from Charles Koch Foundation to support Program in Corporate Welfare Studies, February 2018, \$445,000  
Gift from Troesh Family Foundation to support Program in Corporate Welfare Studies, October 2018, \$240,000  
Gift from Troesh Family Foundation to support Program in Corporate Welfare Studies, January 2019, \$130,000

 3  14  14 

 **Jack Pondit vsp** @Jack\_Pondit · Jul 2, 2020

Replying to [@AmihaiGlazer](#)

This is a horrible misuse and misinterpretation of statistics.

# Things to avoid

- Chartjunk
- Misleading text/axes
- Inaccurate plotting
- So many COVID-19 data visualizations...

# Chartjunk



# Misleading Text/Axes

## Top 5 Counties with the Greatest Number of Confirmed COVID-19 Cases

The chart below represents the most impacted counties over the past 15 days and the number of cases over time. The table below also represents the number of deaths and hospitalizations in each of those impacted counties.

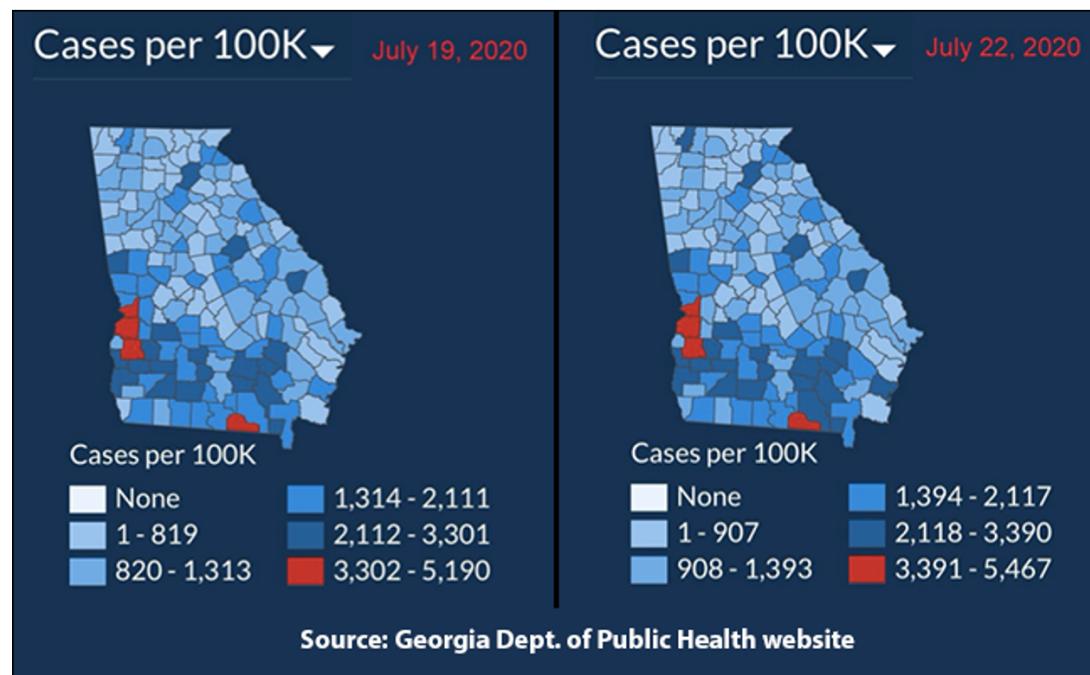


davidmcarlson  
@davidmcarlson

On the official Georgia COVID stats web page is this graph. Looks good, getting better, right? Look closer at the dates on the X-axis. They have arranged the dates out of order to create a declining appearance.  
@GaDPH @georgiagov

# Inaccurate Plotting

This strange plot was put out by Georgia's Health Department. It's trying to show that basically there haven't been any real changes in COVID-19 statewide. But look at the values in the legend...They've changed them to basically keep the same graph. WUT?!



# Misleading Shapes



**Scrimphony Warchestra** · 8h

Replies to [@reina\\_sabah](#)

DATA VISUALIZATION IS MY  
PASSION



1



7



494



**Sabah Ibrahim** @reina\_sa... · 8h

STOP YELLING YOU WILL ALERT  
THE LATVIANS



12



7



940



# Accessibility is IMPORTANT

- Colorblindness sucks. ~1 in 12 men are colorblind (much lower in women)
- People have poor vision (glasses, anyone?)
- Journals scale your figure sizes down so that it fits within the article (like within a column of text)

What can we do?

- Colors
- Contrast
- Big text size (bigger/bolder you can get away with less color contrast)

# Color Palettes

Also super helpful! `RColorBrewer` and `ggsci` are great. But there are millions of others.

Ex: the color palette for all of the slides on this website? The Aussie color palette from <https://flatuiccolors.com/palette/au>

**All you need are hex codes** (6 digits, alphanumeric). This is true for all color palettes (including monochromatic).

Different types of palettes (unordered, sequential, divergent etc.). Look [at this blog post](#) to learn more about these

# Contrast

When you have something side-by-side, you can have different colors. OR you can have the same color but a different shade/tone/tint.



# Grayscale

The most obvious of this is grayscale (technically, it's not a hue, and you're dealing with saturation, but that's completely unimportant for this intro):



# Providing Contrast is Important

It lets the reader *easily* extract meaningful information.

Colors, shapes, size (as in bubble plots), sometimes transparency etc.

Need help picking different shades/tones/tints of the same color? A ton of websites can help! Ex: <https://www.colorhexa.com/>

# Text Size

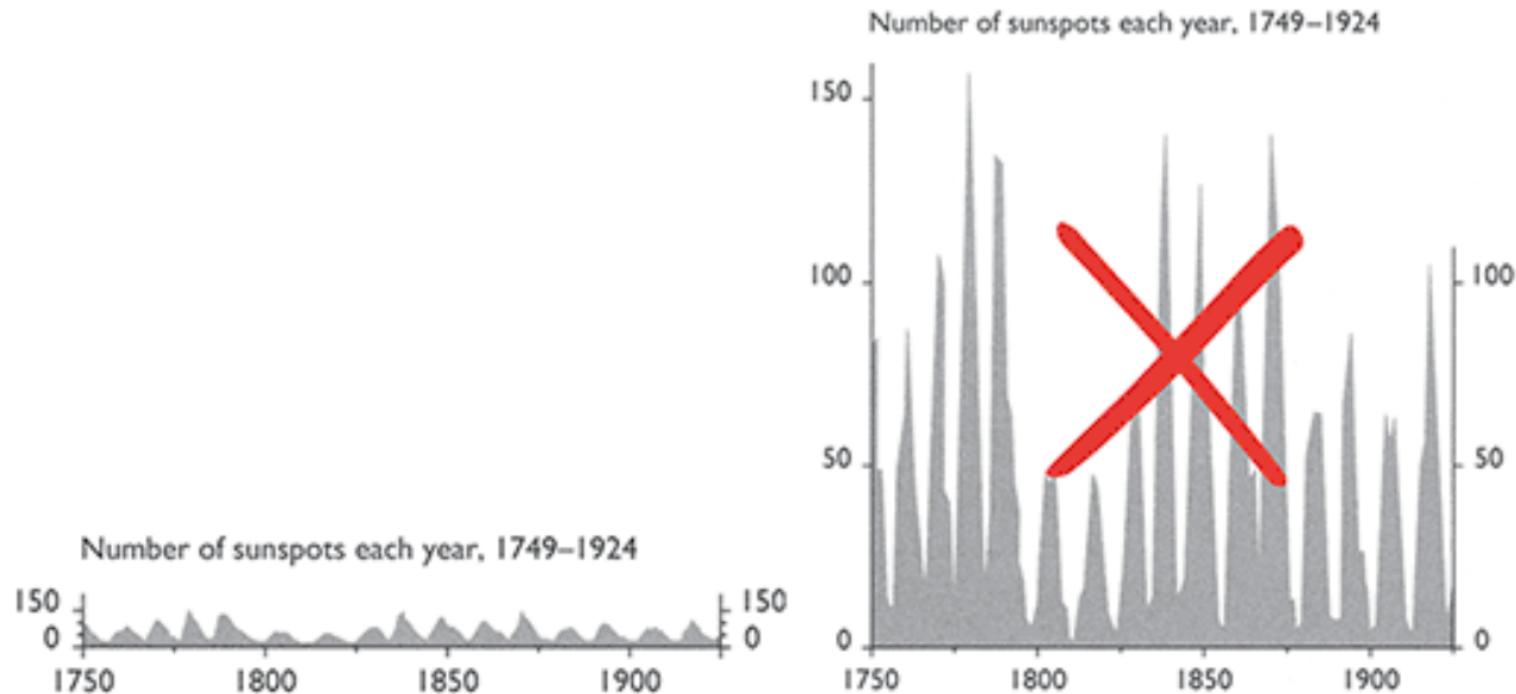
**Don't make your font size super uber small!**

Some of us are getting old and our eyesight is fading (*I'm not bitter...yes I am...*)

Also, academic journals scale down your figures. Better to make the text size larger so that when it gets scaled down, it's still readable.

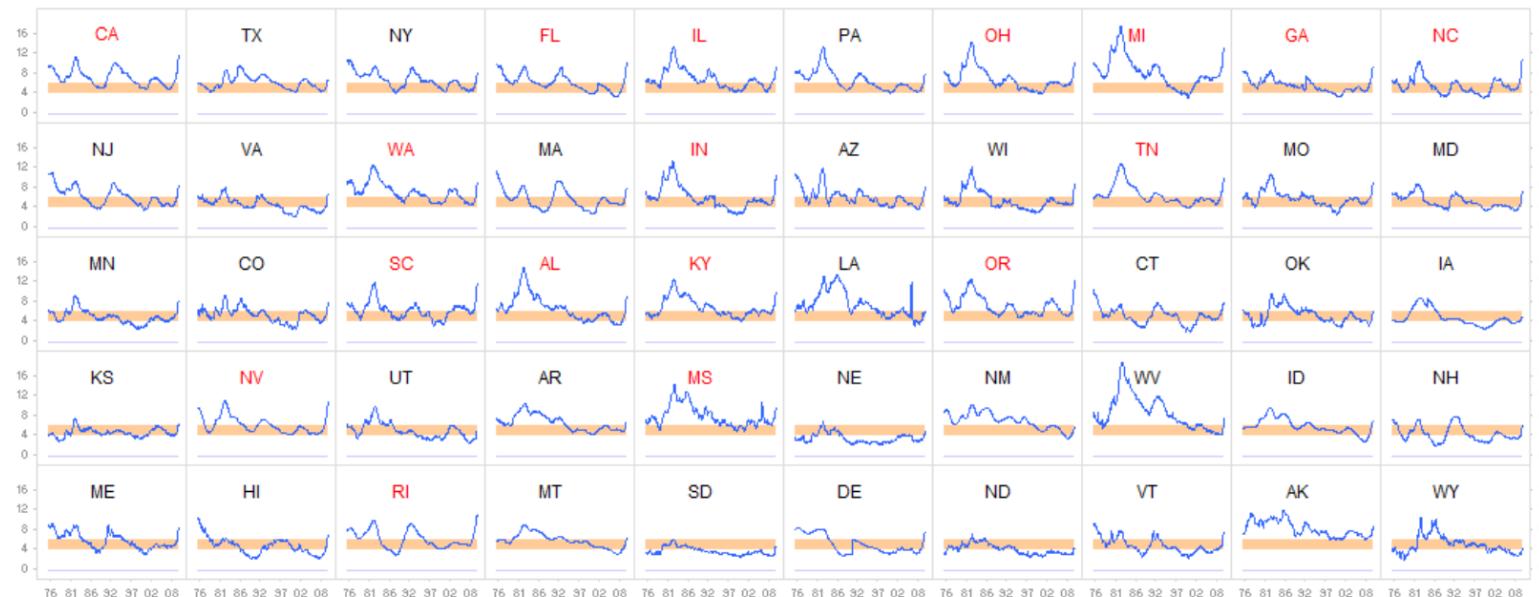
# Aspect Ratios

You want to display your figures faithfully, but you don't want to take up extra space you don't need. Think about the aspect ratio of your plots!



# Small Multiples

Monthly Unemployment Rates by State, Jan 1976 - Apr 2009



Source: Bureau of Labor Statistics

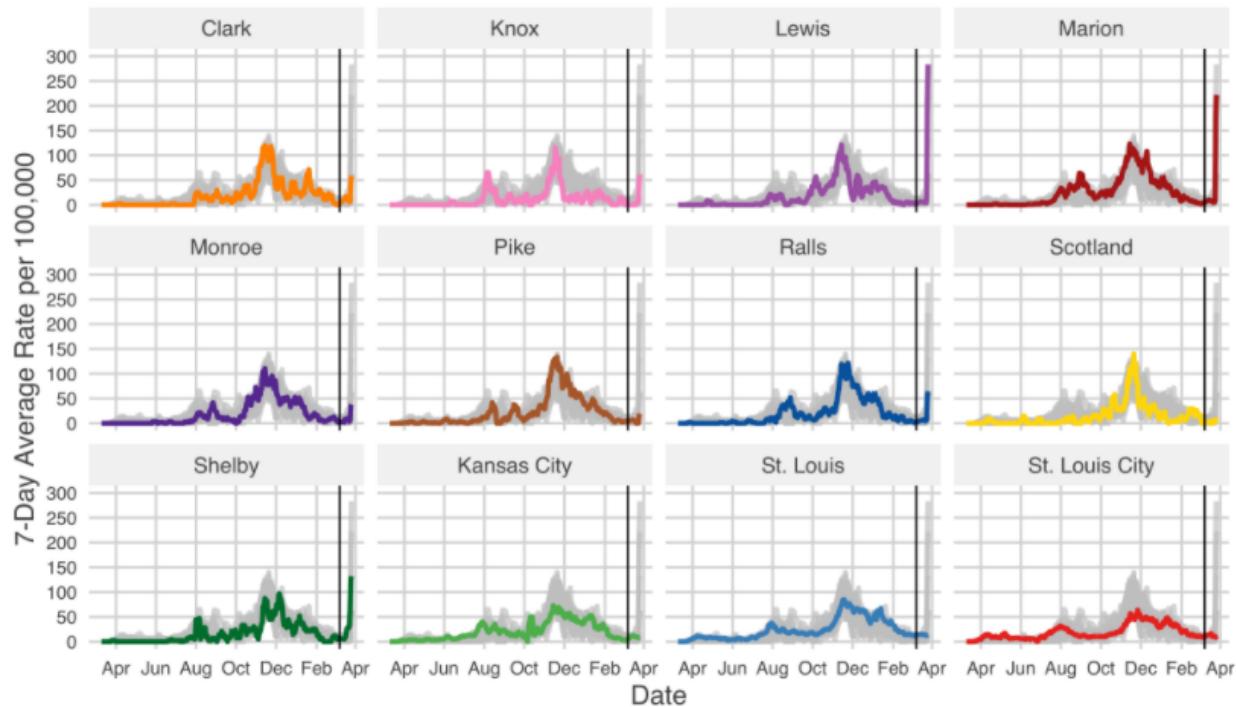
Notes: The orange band denotes a "normal" unemployment rate (4%-6%);  
State code in red: unemployment rate in April 2009 is higher than the US average

<https://www.juiceanalytics.com/writing/better-know-visualization-small-multiples>

# Small Multiples with COVID

## Pace of New COVID-19 Cases in Select Missouri Counties

Northeastern Missouri Focus  
2020-03-10 through 2021-03-26

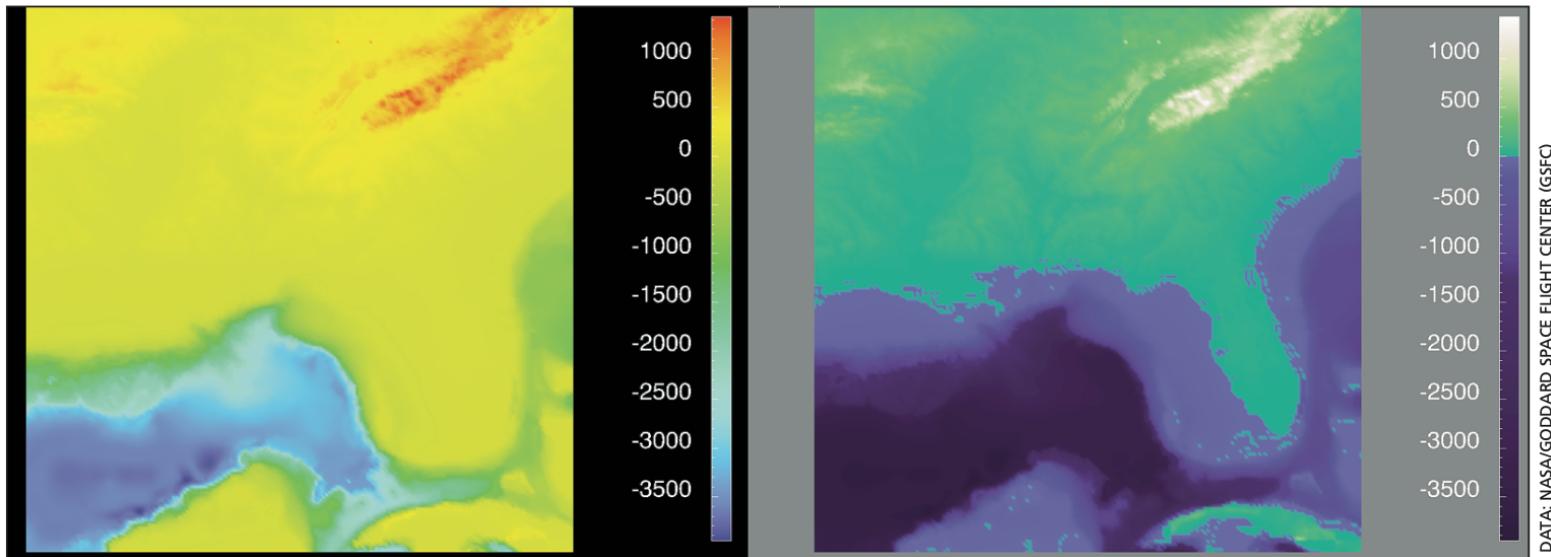


Plot by Christopher Prener, Ph.D.

Data via the New York Times COVID-19 Project and the U.S. Census Bureau

Vertical line represents addition of antigen test data for most Missouri counties on 2021-03-08

# Don't Use Rainbow Colormaps



Rogowitz & Treinish, IEEE Spectrum, 35(12):52-59. 1998

# Self-evaluation

How do you know if you've made a good figure?

- Does it **EASILY** communicate what you want?
- Do readers need to read and re-read your figure legend, or is your message clear?
- Is it accessible to people with poor eyesight or colorblindness?
- Does it *faithfully* reflect your data? Beauty + truth

# Rules

- Don't make a plot when a table will do
- Represent data with appropriate significant figures
- Use appropriate plot types for the data types
- **Label your axes**
- **Label your axes**
- **Label your axes**
- Title your figures
- Whenever possible, show all the data (or at least the distribution)
- Don't rely on a legend or caption/text
- Don't rely on default plotting conventions
- If possible, show outliers rather than removing them
- Sort categorical data accordingly
- Exploit small multiples to great effect (in R, use faceting)
- Strive to maintain the same color conventions/palettes across all figures
- Start with what you want your plot to look like, then work backwards
- **SHOW. YOUR. CODE.**

# Helpful Things in R

- `geom_dotplot`; better than bar plots, typically (see earlier slide)
- Revisit our section on `ggplot2`; I bet you missed a lot...
- Raincloud plots; [blog post here](#)
- Using the same theme modifications for all plots? Make a function to store that theme and call it later (more on functions soon!)
- **Google is your friend; Google is your professor**
- Interested in making generative art in R? Check out the Twitter accounts of [Danielle Navarro](#) and [Ijeamaka Anyene](#); then look at their websites!
- Want examples of good COVID plotting with open R code? Check out [Chris Prener's Twitter account](#). If you'd like this as a weekly newsletter, check out his [River City Data](#) tracking site!