IR 101 DATA VISUALIZATION CAIR 2020

Scott Heil UC Riverside scott.heil@ucr.edu



OVERVIEW OF THIS SESSION

Platform agnostic

• Focus on concepts, not technical how-to

 Practical examples, including some that work well and some that need help • Examine how data types relate to visual conventions

• Top 10 tips for upping your visualization game

A few suggested resources

DO THESE WORK?







This uses a lot of space to convey very little information



4-Year Graduation Rate by Ethnicity





WERE THOSE HARDER THAN THEY NEEDED TO BE?

A FEW SAMPLES THAT WORK WELL

Category	Year 2006	#	At least somewhat dissatisfied 21%	Percentage of respondents by response category				At least somewhat satisfied	Select Year(s) Multiple values
Afr Amr		1,328		12%	33%	33%	12%	79%	Characteristics Race/Ethnicity
	2018	2,190	25%	<mark>6%</mark> 15%	33%	31%	11%	75%	Race/ Lennercy
Amr Ind	2006	341	18%	11%	25%	36%	22%	82%	
	2018	369	21%	12%	32%	36%	10%	79%	Very Dissatisfied
Asian	2006	22,590	22%	<mark>5%</mark> 15%	36%	33%	9%	78%	Dissatisfied
	2018	20,993	24%	<mark>6%</mark> 15%	38%	31%	7%	76%	Somewhat Dissatisfie
His/Lat	2006	7,103	15%	10%	27%	41%	17%	85%	Satisfied
	2018	14,558	21%	<mark>5%</mark> 13%	33%	35%	11%6	79%	Very Satisfied
Interntnl	2006	1,199	20%	12%	31%	37%	13%	80%	
	2018	5,021	19%	13%	36%	35%	10%	81%	
White	2006	19,700	13%	8%	26%	43%	18%	87%	
	2018	13,916	18%	10%	28%	39%	15%	82%	Π

Responses by Race/Ethnicity and Year at UC

Source: University of California, Office of the President, 2018

Scale centered at the midpoint **Colors are meaningful and** match visual expectations Responses by Race/Ethnicity and Year at UC **Provides multiple** Select Year(s) At least At least Percentage of respondents by response Multiple values levels of summary Category Year somewhat some what # category dissatisfied satisfied and detail along with Characteristics Afr Amr 1,328 21% 12% 33% 79% 2000 Race/Ethnicity change over time 2018 2,190 25% 15% 33% 75% 82% Amr Ind 2006 341 18% 11% 25% Very Dissatisfied 2018 369 21% 12% 32% 79% Dissatisfied Asian 2006 22,590 22% 5% 15% 36% 78% Somewhat Dissatisfie 2018 20,993 24% 15% 38% 76% Somewhat Satisfied His/Lat 15% 85% 2006 7,103 10% 27% Satisfied 2018 14,558 21% 5% 13% 33% 79% Very Satisfied 31% 80% 2006 20% 12% Interntnl 1,199 13% 36% 81% 2018 5,021 19% White 2006 19,700 13% 8% 26% 87% 2018 13,916 18% 28% 82% 10%

Source: University of California, Office of the President, 2018

Not worth losing sleep over

Sleep length per night by age and employment status





The Economist



Source: The Economist, September 2020

Reduces complex pattern to illustrate the message



Source: The Economist, September 2020



REVENUE AND MARGIN GROWTH SCENARIOS Assuming 9% membership growth.

Source: Scott Berinato, Harvard Business Review, August 2016

Isolates each scenario with the others in the background



Source: Scott Berinato, Harvard Business Review, August 2016



Source: California State University, Chancellor's Office, 2020



Source: California State University, Chancellor's Office, 2020

WHAT MAKES A VISUAL SUCCESSFUL?

Sound quantitative reasoning

Clear message or focus

Intuitive and easy to read

Aesthetically pleasing

DATA TYPES AND VISUAL REPRESENTATIONS

Nominal comparisons

Categorical

No natural order

• Typical examples: demographic groups, student types, academic units

Nominal examples



Ordinal comparisons

• Clear sequence of levels

 Typical examples: Likert scales, age brackets, score ranges, letter grades

Ordinal examples



Proportions of the whole

 Measures that are expressed as percentages

• Typical examples: Demographic distributions, graduation rates, retention rates, admission rates

Proportion examples



Time series

• Multiple measurements over some time interval

• Typical examples: Annual graduation rates, enrollment by term, demographic change over time

Time series examples



Correlations

• One or more quantitative measures

• Typical examples: Subgroup comparisons, associations across multiple measures

Correlation examples

Multiple Line

Scatterplot/Bubble





Combination



Heatmap



TIPS FOR BETTER DATA GRAPHICS



Use ranked lists



*Avoid what Edward Tufte called the "tyranny of the alphabet"

(But this may not be appropriate for ordinal categories)



Remove clutter

Just the essentials

More formatting, but not necessarily more meaning





Small graphs can be effective

Reasonable size for four data points



Bigger does not always improve readability





Ensure the visual matches the message

Here the visual conclusion is lots of change

But in this format the message is stability over time



(Data values are identical for the two graphs)



Make the legend part of the graph

Here the meaning is instantly clear and the reader does not have to hunt for anything in a separate legend

Group A Group B



Zero is not needed in trend line graphs

Nothing is lost here if the yaxis does not begin at zero But for bars and columns zero is necessary







Try small multiples

In this format the reader can quickly scan across multiple series and spot similarities and differences

Group A

Group B

Group C



Use descriptive titles and annotations

Incoming GPA vs. Retention Rate

Cohort grade point average compared with one-year fall-to-fall retention





Be deliberate in the use of color

Gradients of the same color often convey the meaning well



Higher contrast colors can be reserved for highlights and special information The rainbow effect is often a distraction and a lost opportunity to add meaning



TIP 10

Not every data point needs to have a label



Labeling key points with easily readable text is almost always better than many points with tiny labels

WRAP-UP

Closing tips

• Let the meaning of the data help drive your visualizations

• Simple is often better than complicated

• Favor readability and intuitive understanding over complexity

RESOURCES

Cheat sheets

• Core Principles of Data Visualization at policyviz.com

- Visual Vocabulary at <u>ft.com</u>
- Data Visualization Cheatsheet at <u>kaggle.com</u>

Books

- *Good Charts,* Scott Berinato,
 2016
- *Effective Data Visualization*, Stephanie Evergreen, 2019

THANK YOU