Visualizing Data

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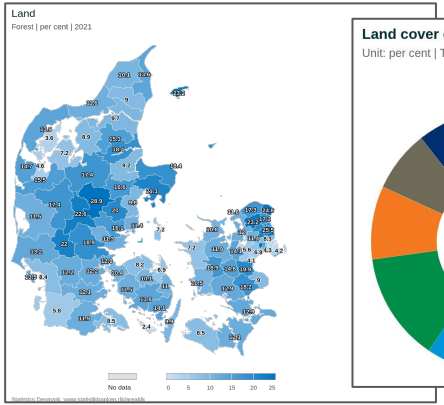
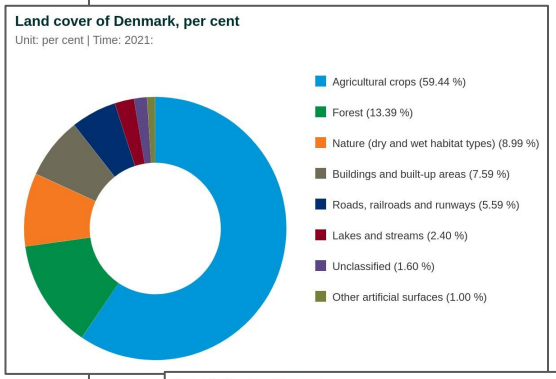
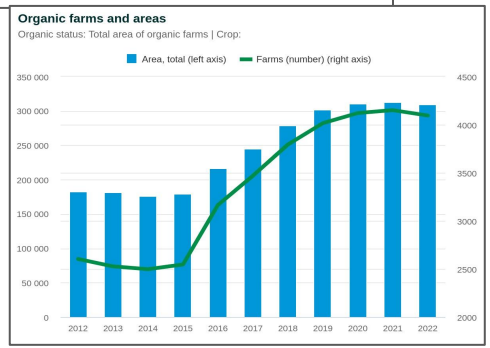
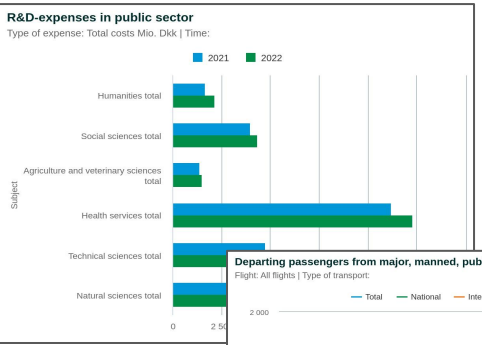
Please have your chart ready, or find one at www.dst.dk

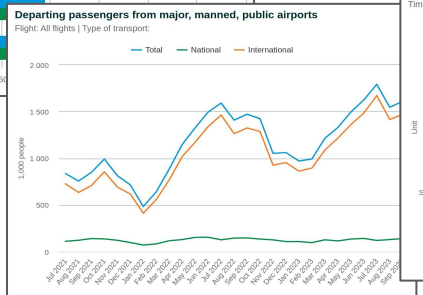
Data Visualizations

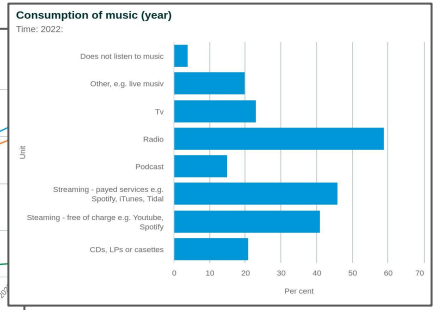






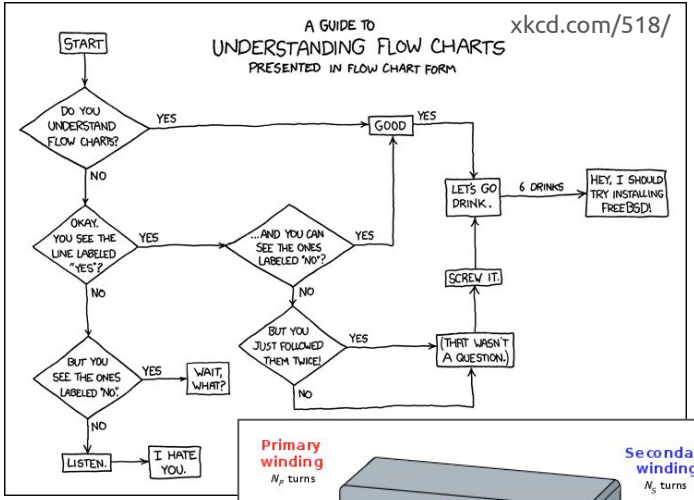




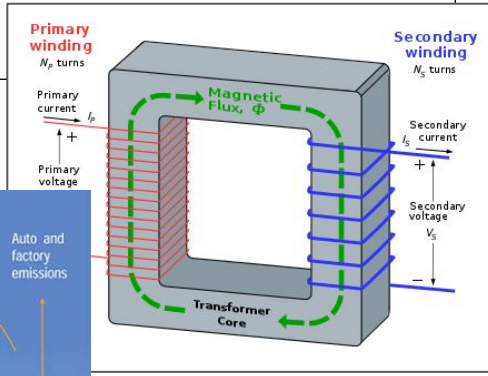


all charts from dst.dk

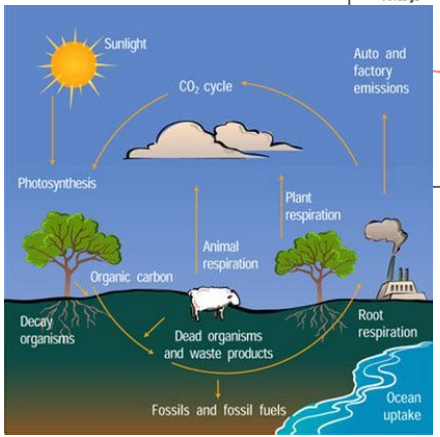
Not Data Visualizations



xkcd.com/518/



https://en.wikipedia.org/wiki/Transformer

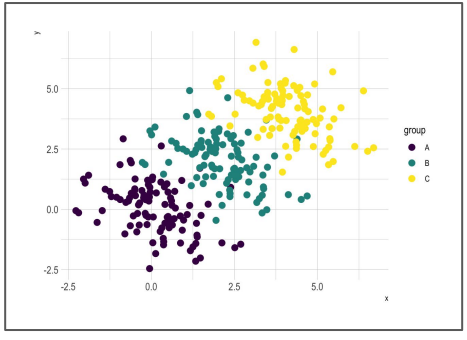


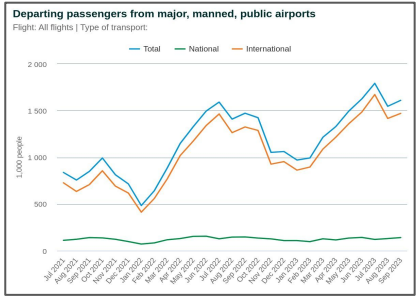
en.wikipedia.org/wiki/Ecosystem_respiration

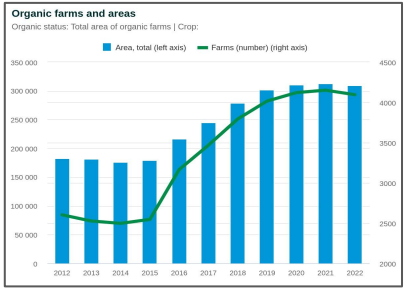
Data visualizations today

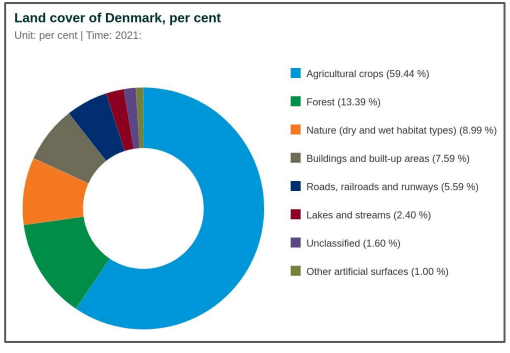
- Frameworks for thinking about different kinds of charts* & when to use them
- Some best practices for visualization design based on human perception
- (Examples of using pandas and seaborn in python.)

*Nomenclature is subtle and inconsistent: chart, graph, plot are ~interchangeable.









more plot

more chart

Visualization is for communication about data

Visualizations go from computer-readable data to human-usable information.

Humans have cognitive limitations:

- Terrible at comparing more than a handful of numbers at once
- Much better at understanding distributions & comparative values graphically
- Still: Bad at comparing more than ~4 variables

Visualizing data can help a lot, but we still need to respect our human limitations.

Example: Numbers vs Graph

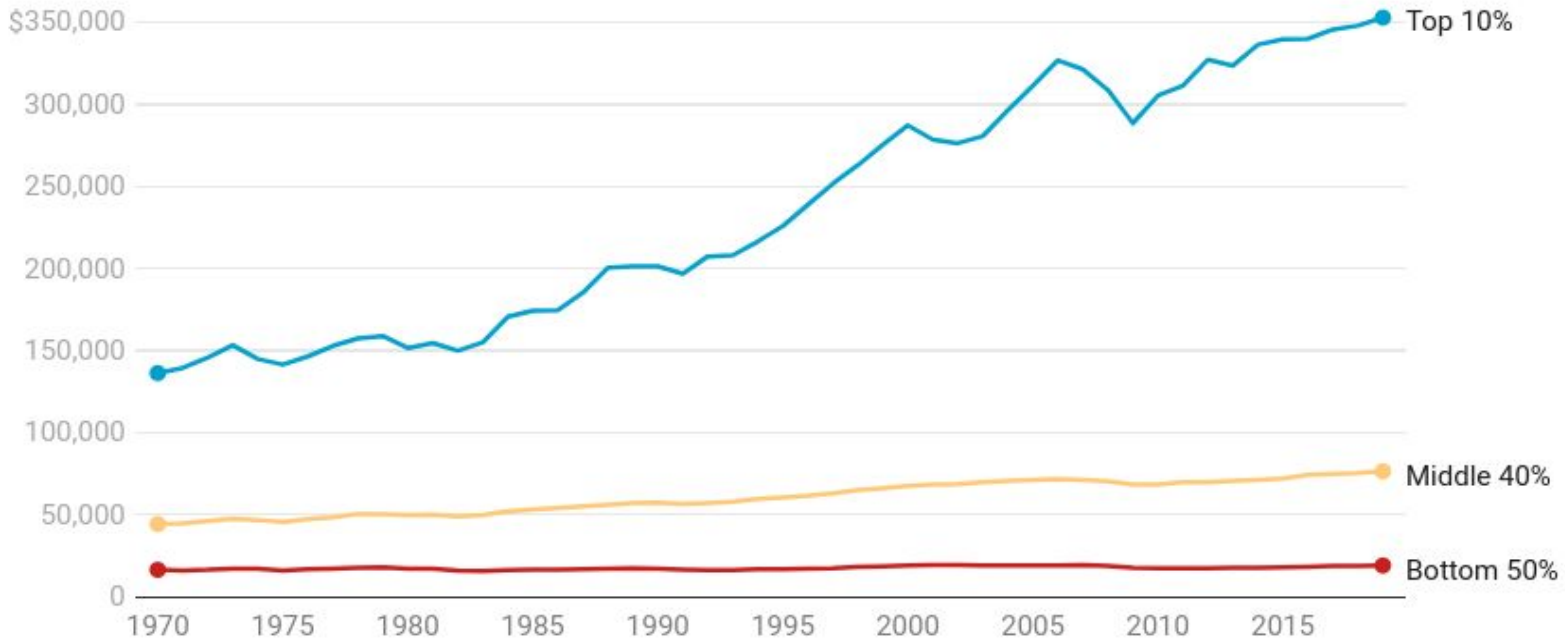
Table 0.1: Average US Adult Income, 1970-2019

US Income Tier	1970	2019
Top 10 Percent	$136,308	$352,815
Middle 40 Percent	$44,353	$76,462
Bottom 50 Percent	$16,515	$19,177

Note: Shown in constant 2019 US dollars. National income for individuals aged 20 and over, prior to taxes and transfers, but includes pension contributions and distributions. Source: World Inequality Database, accessed 2020

https://handsondataviz.org/believe.html

Average US Adult Income, by Percentile, 1970-2019



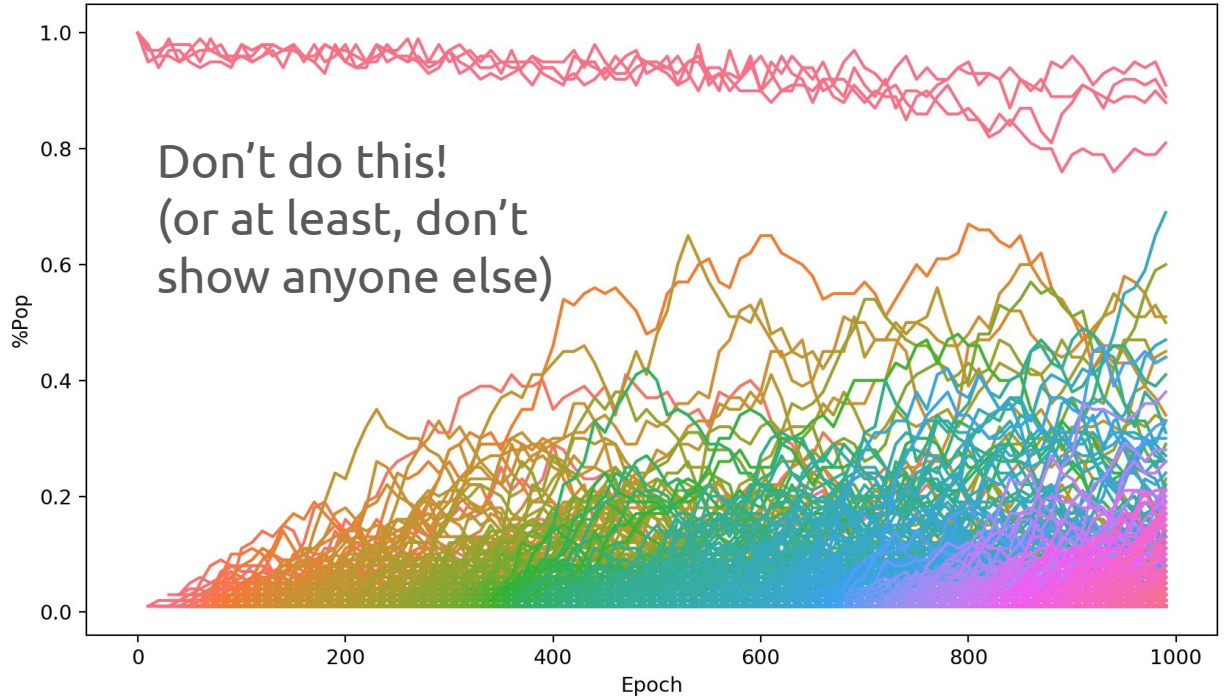
Note: Shown in constant 2019 US dollars. National income for individuals aged 20 and over, prior to taxes and transfers, but includes pension contributions and distributions.

Chart: by HandsOnDataViz • Source: World Inequality Database 2020 • Get the data • Created with Datawrapper

https://handsondataviz.org/believe.html

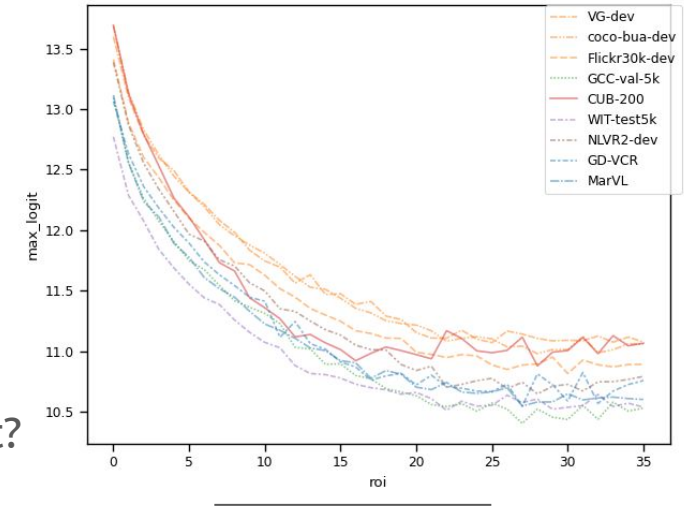
Good communication requires selection & summarization

If you plot everything, you show nothing.

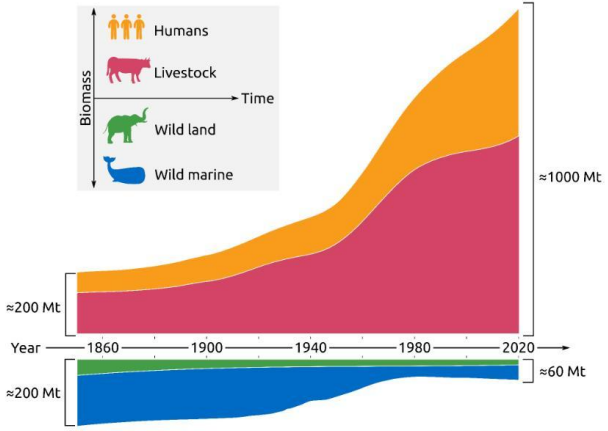


Two types of visualizations

1. **Explorative visualizations**
   Goal: *understand* what is happening in the data.
   - Visualizations as thinking tool; audience = you.
   - Each visualization represents a hypothesis,
   an expectation about what you will see. What is it?



2. **Persuasive visualizations**
   Goal: *convince* your audience that your conclusions are correct.
   - Match between chart/plot & text (data & story)
   - Important to understand & respect your audience's knowledge, attention, defaults.
   Test & revise!



Greenspoon et al, in prep.

Data

What kind of data are represented in your visualizations?

Pairwise exercise, in a minute - but first, let's think about what data looks like.

Structured vs Unstructured data:

Unstructured: piles of text, images, sounds, DNA

Structured: formatted, classified, 'structured' data

most common: tabular data

also: network/graph data, GIS data, etc.

Tabular data - *tidy* data formatting is recommended

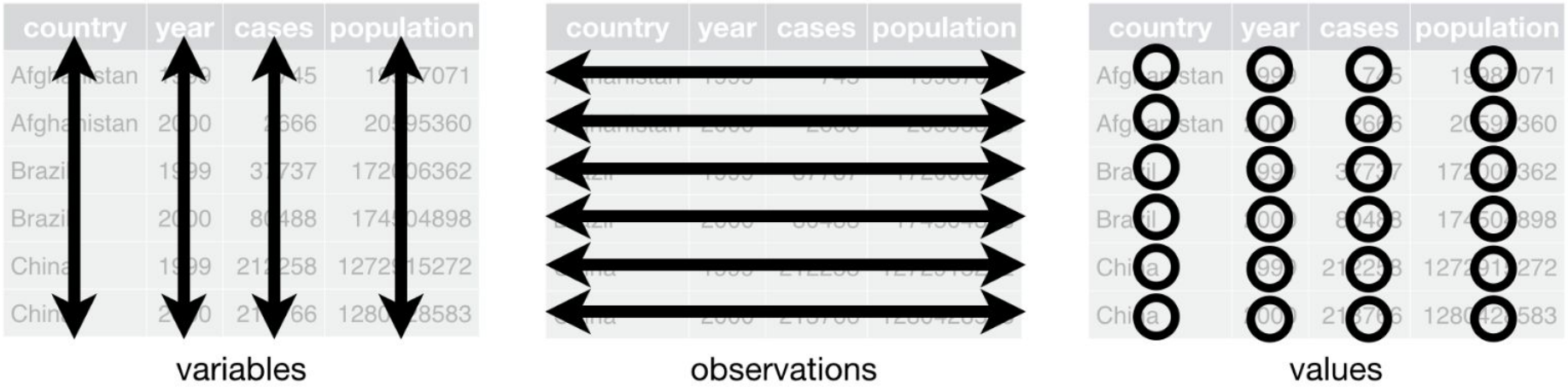


Figure 12.1: Following three rules makes a dataset tidy: variables are in columns, observations are in rows, and values are in cells.

Wickham & Grolemund, R for Data Science

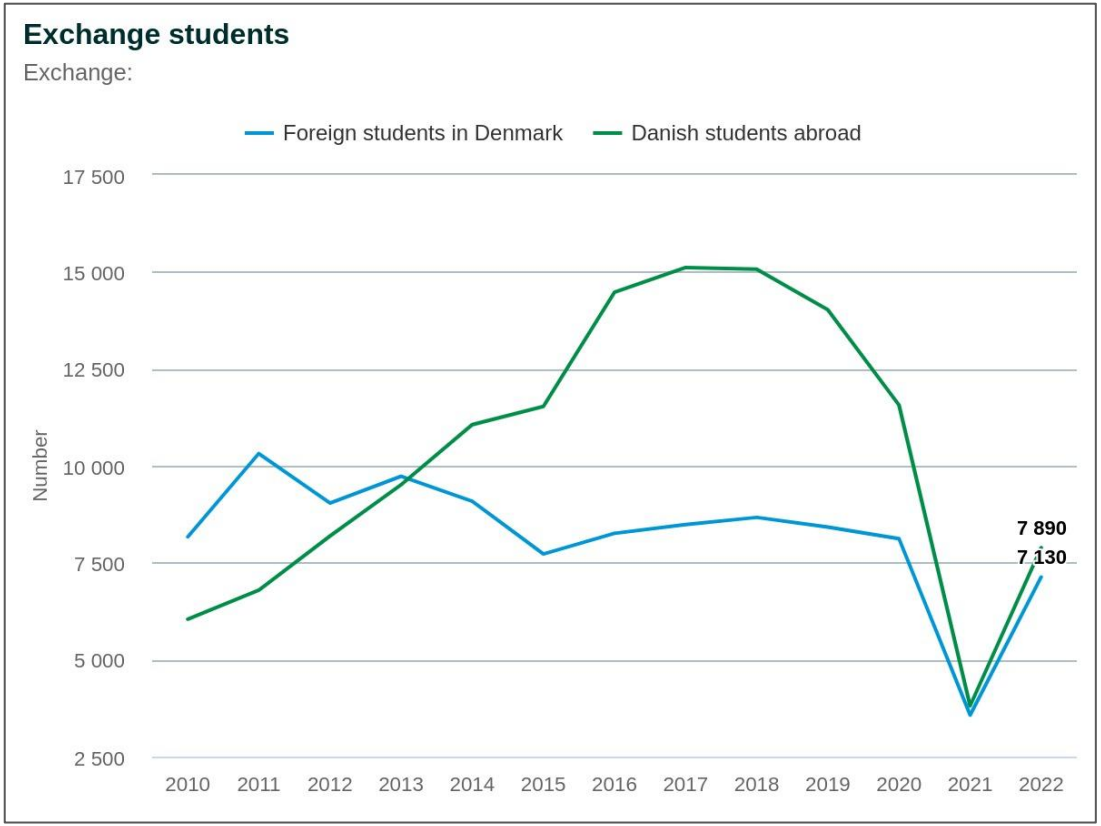
Pairwise, 4 min: What is the data behind your charts?

Look at your charts to determine:

- What are the *observations*?
- What are the *variables?*
- What are the *values?*
- What *units* do the values have?

Does the plot show individual data points, or does it show summarising statistics (e.g. averages)?

Data types: Are the variables' values categorical or quantitative? Discrete or continuous? Maybe even ordinal?

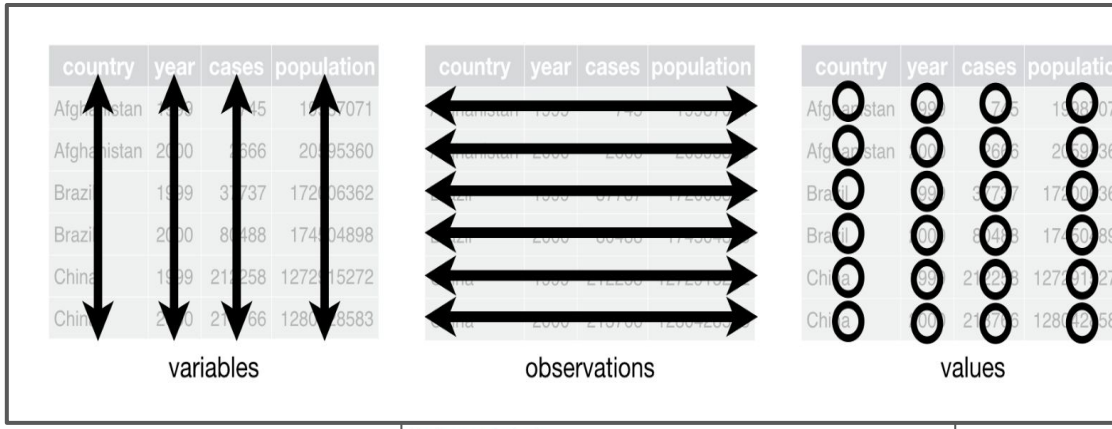


https://www.dst.dk/en/Statistik/emner/uddannelse-og-forskning/fuldtidsuddannelser/udvekslingsstuderende

Pairwise, 5 min: What is the data behind your charts?

Look at your charts to determine:

- What are the *observations*?
- What are the *variables?*
- What are the *values?*
- What *units* do the values have?



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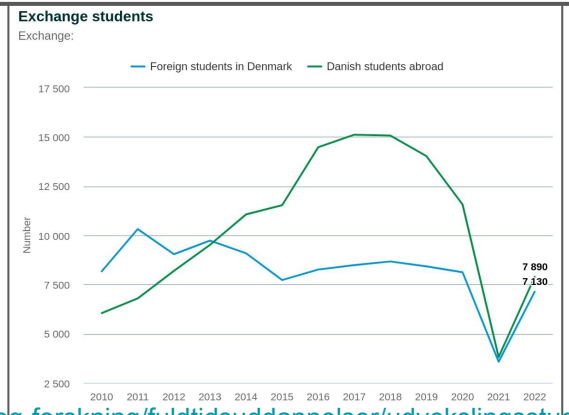

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What did you find?

Have data ⇛ need chart: best practices

Lots of different kinds of charts - good for different kinds of messages.

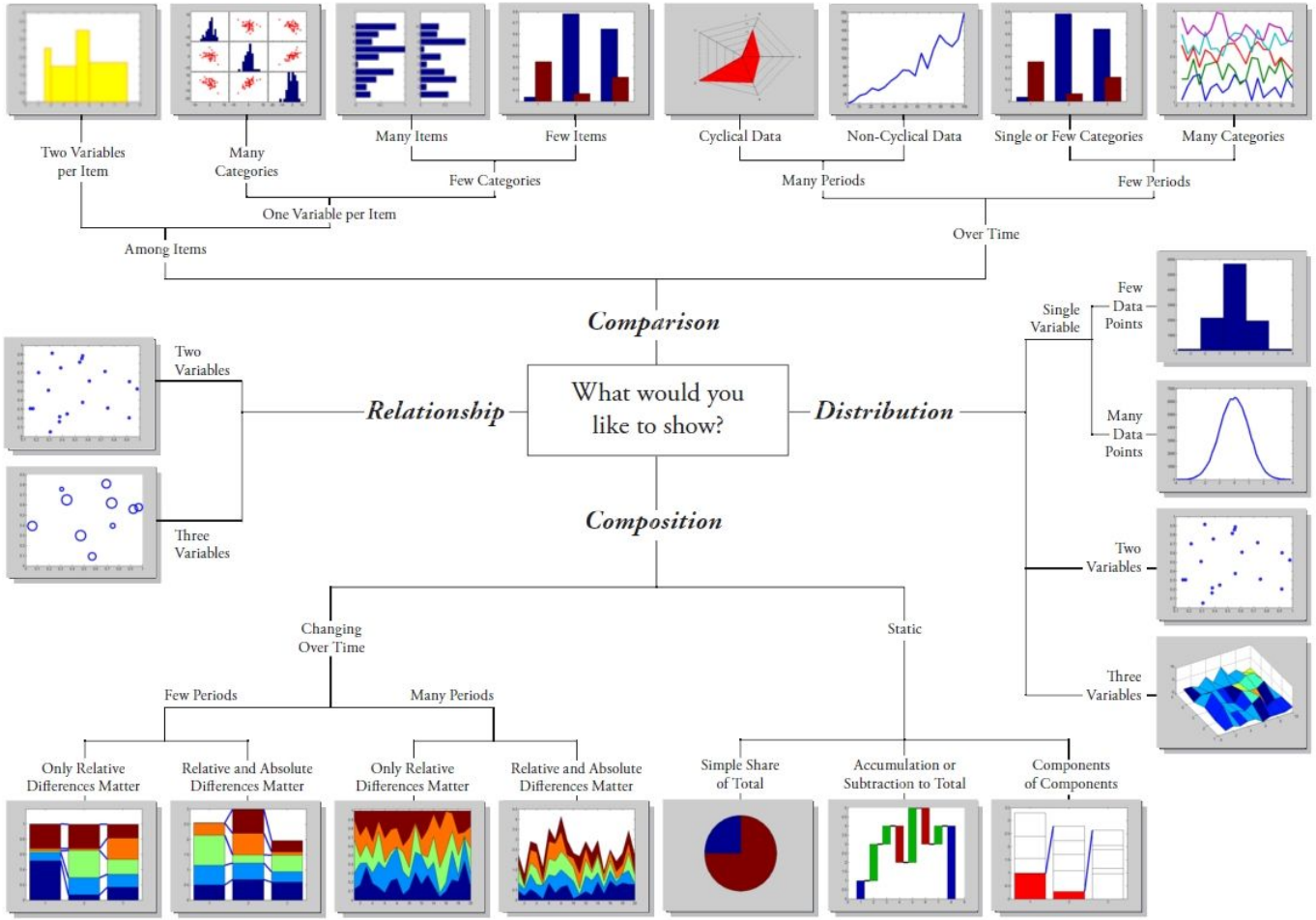
Deciding which type of chart to construct depends on data & overall story.

↪ What message is your chart conveying? What is the context?

Two tools to help structure this decision space:

Abela's Chart Chooser & FT Visual Vocabulary

Abela's Chart Chooser

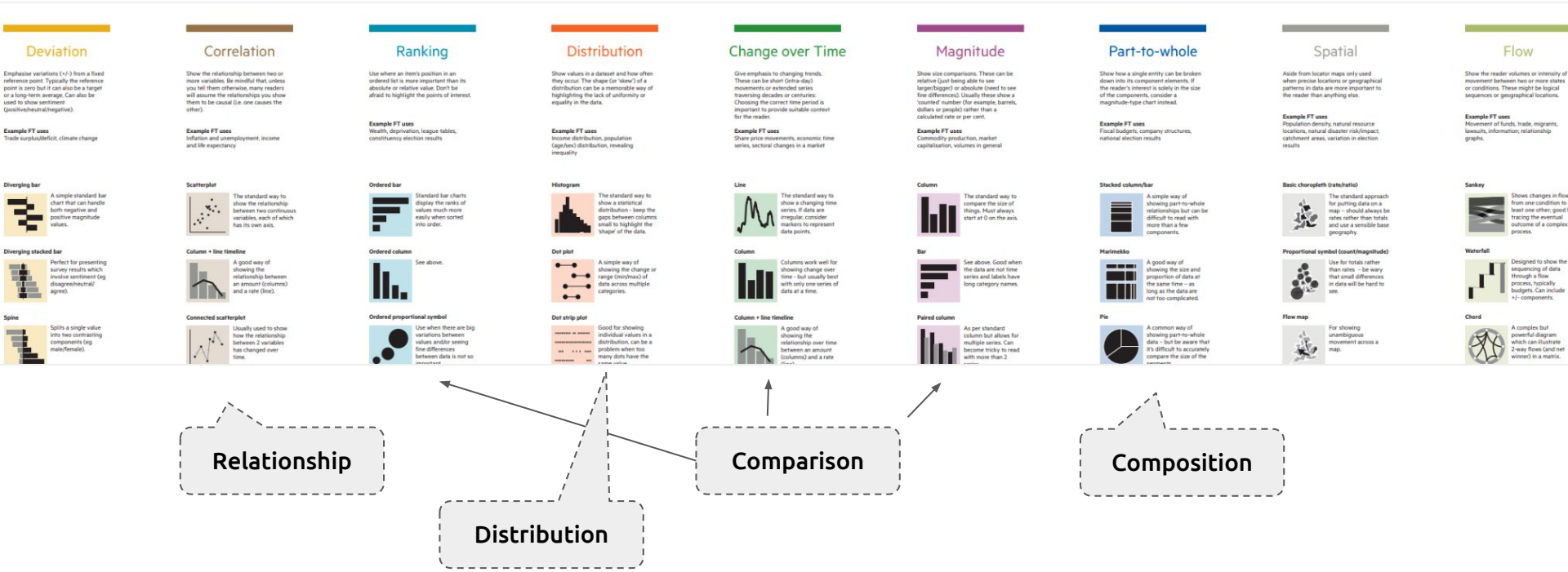


Modified with permision -Doug Hull
blogs.mathworks.com/videos
hull@mathworks.com 2009

FT's Visual Vocabulary

ft-interactive.github.io/visual-vocabulary/

Remix: public.tableau.com/views/VisualVocabulary/VisualVocabulary



Which category fits your chart?

https://github.com/Financial-Times/chart-doctor/blob/66caa2f126d223b3c68c7360888a035123c4ede6/visual-vocabulary/Visual-vocabulary-en.pdf

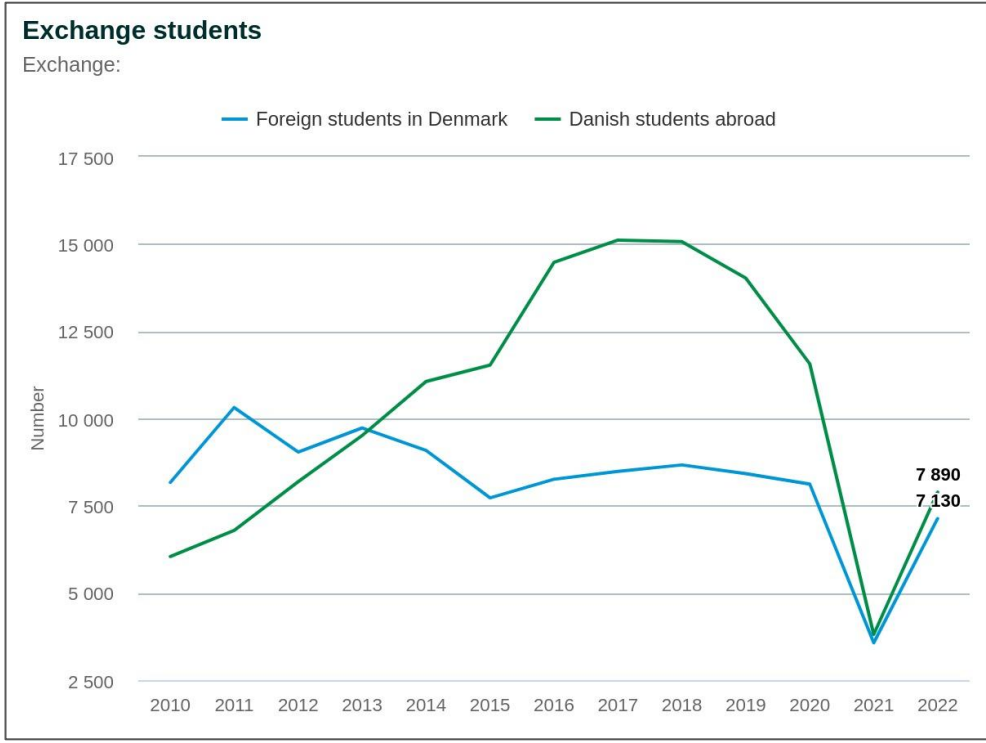
Pairwise, 4 min: What is the story motivating your chart?

Person A: Explain the main point the chart is communicating.
Which VV category does it fall into?

Person B: Do you understand? Are you convinced?

Switch!

(This might be harder with dst charts: use your imagination.)



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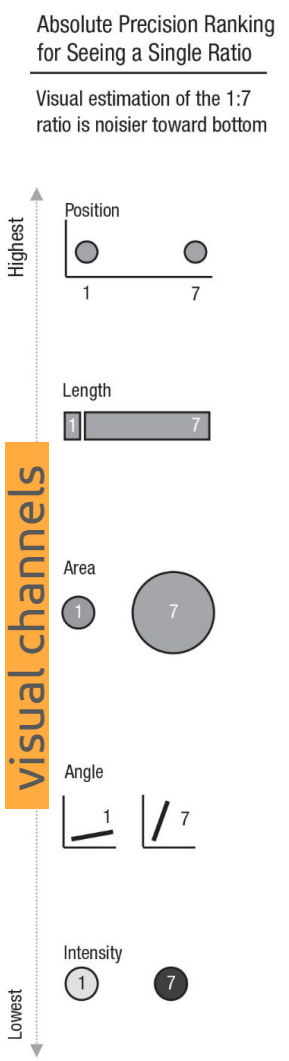
What did you find?

Strengths in Visual Processing



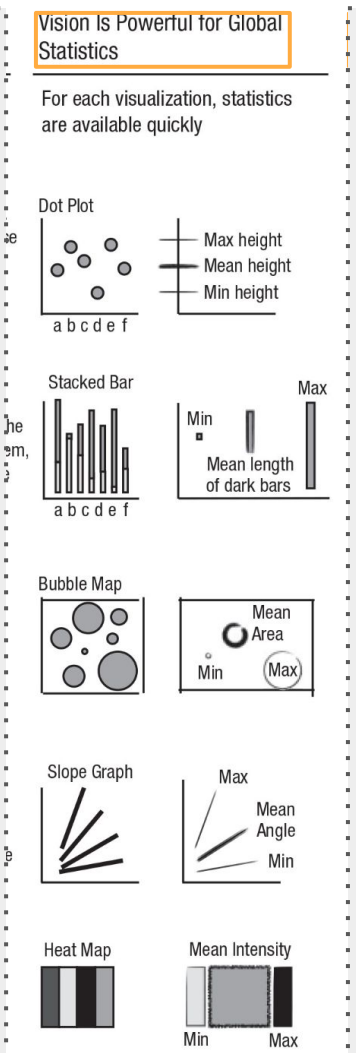
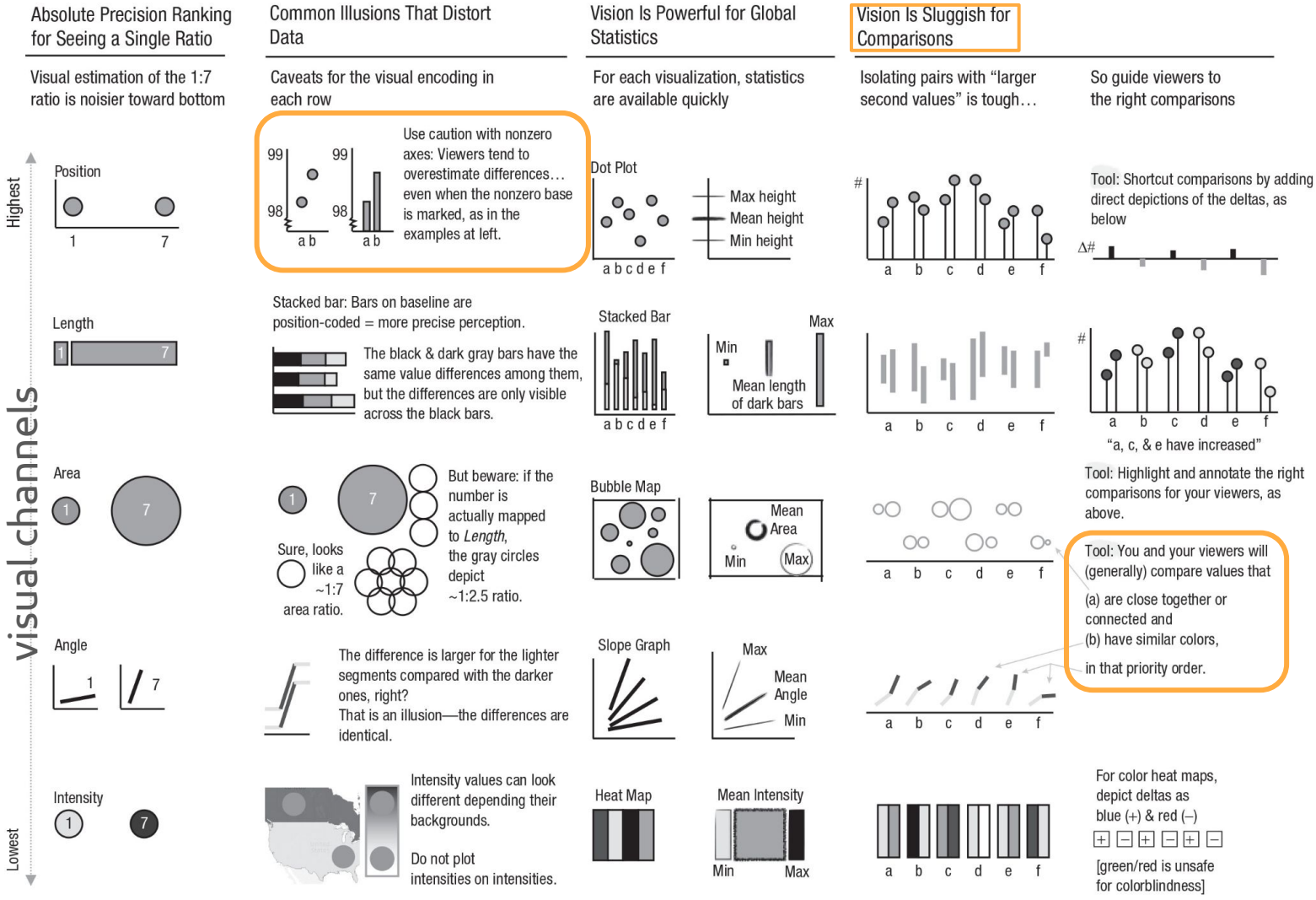


Figure 2 from:
Franconeri, S. L., Padilla, L. M., Shah, P., Zacks, J. M., & Hullman, J. (2021). The Science of Visual Data Communication: What Works. Psychological Science in the Public Interest, 22(3), 110-161. https://doi.org/10.1177/1529100611051956

Strengths and Weaknesses in Visual Processing



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Bars and Lines
Zacks & Tversky 1999

Experiment: ask people to describe the relationship shown on one of four graphs.
Response is ‘discrete comparison’ or ‘trend assessment’?

Results: Mostly congruent BUT: “effect of graph type was about twice that of conceptual domain”

Leading to statements like: “One tends to get taller as one becomes more [male/Danish/only-child]”

Takeaway: use congruent graphs for your data type!

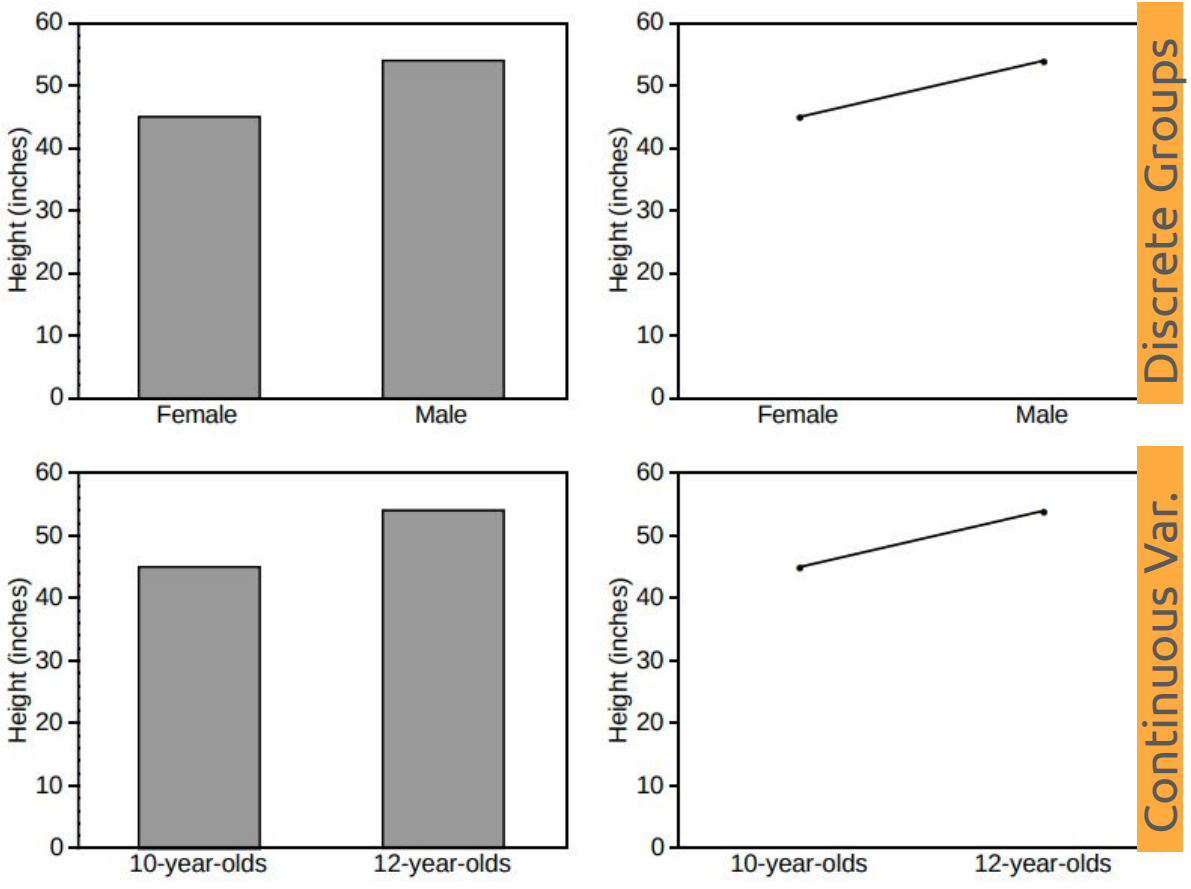

Figure 2. Examples of the bar and line graph stimuli and the continuous and categorical conceptual domains used in Experiment 2.

Discrete areas

Continuous lines

Color Palettes

Qualitative - e.g. lines, bars

Paired

Ordered

Continuous - e.g. heatmaps, maps

Divergent

Don’t use gradient palettes for categories (without good reason)

NOT IDEAL BETTER

Use intuitive colors - match expectations

NOT IDEAL BETTER

https://seaborn.pydata.org/tutorial/color_palettes.html

https://blog.datawrapper.de/colors/

Multiple Figures in one Document

When a document has multiple figures, help the reader: **be consistent**

Things that are the **same, stay the same visually:**

- Same observations keep the same colors
  - e.g. “EU” is always blue;
    - higher values are always more darker
- Same kind of relations use the same kind of chart
  - e.g. Change over time is always a lineplot

Things that are **different look different**

- Don’t use the same kind of chart for two different relations
  - e.g. lineplots for change over time and also change over size

There’s more on Absalon

Python Jupyter notebook using dst data: intro to pandas & seaborn libraries

`Files -> L1 Bibliotek notebook`

Related reading section - see Useful Resources Page on Absalon:

Fundamentals of Data Visualization by Claus O. Wilke

Friends don't let Friends make Bad Graphs by Chenxin Li

& more -

& lots of resources for beginning Python programming.

Next

Thursday lecture: Statistics with Morten Akhøj

Thursday TA sessions: Intro to Python, setup for Assignment 1

Questions?

Admin/logistics/assignment: Daniel Hershkovich - dh@diku.dk

Visualizations/lecture content: Me - stfr@diki.dk