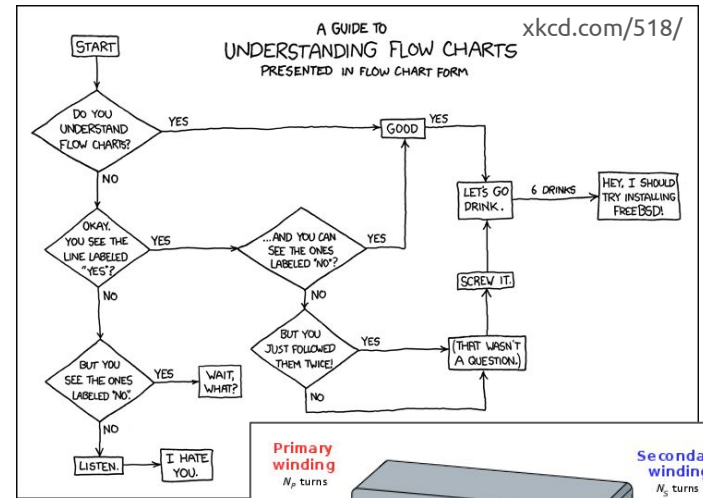
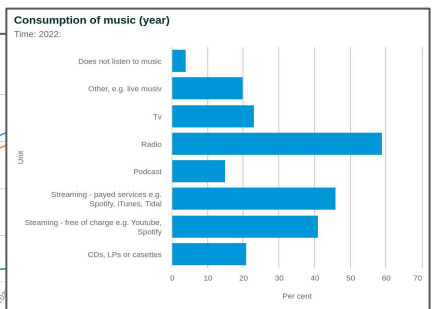
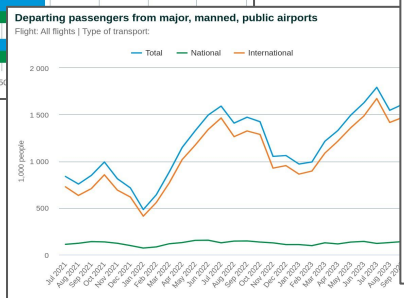
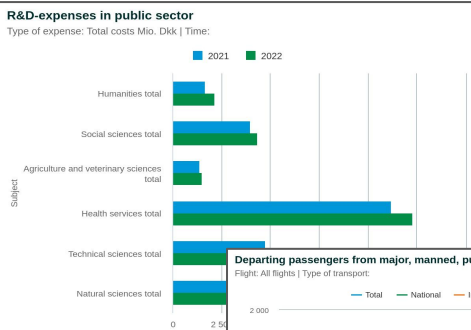
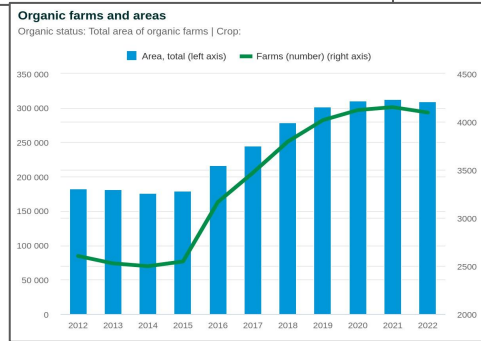
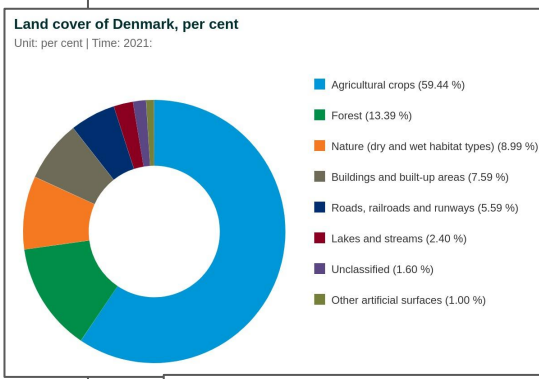
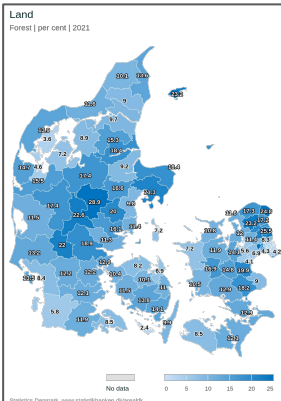


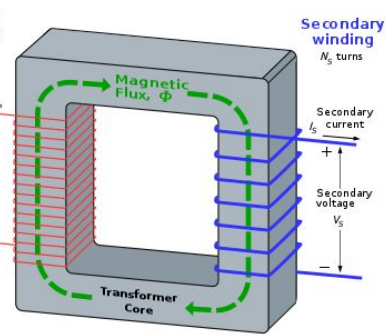
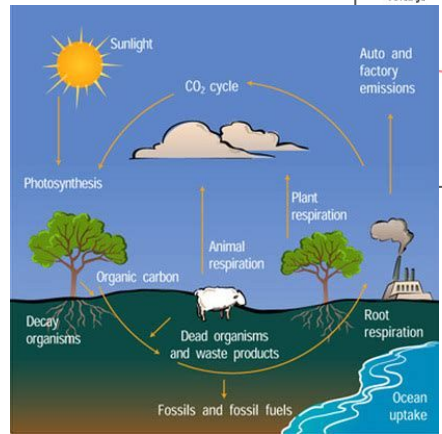
Visualizing Data

Stella Frank stfr@diku.dk

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



Not Data Visualizations



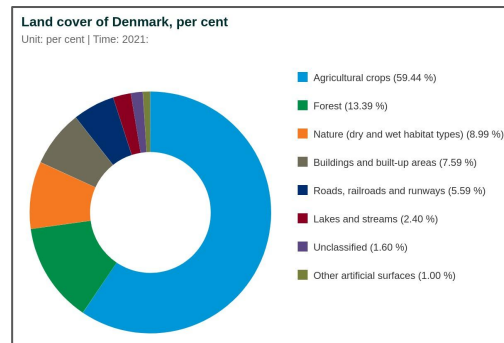
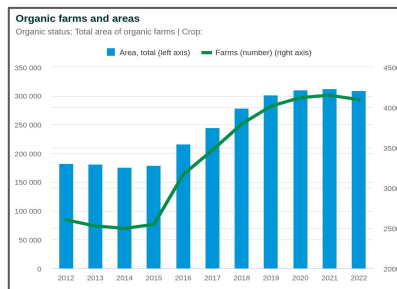
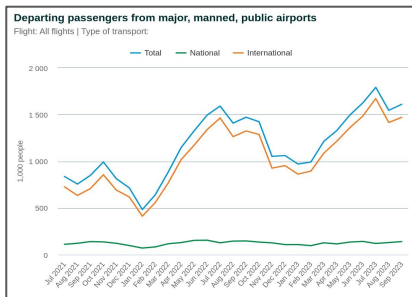
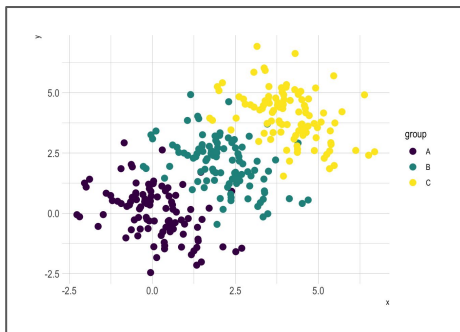
<https://en.wikipedia.org/wiki/Transformer>

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

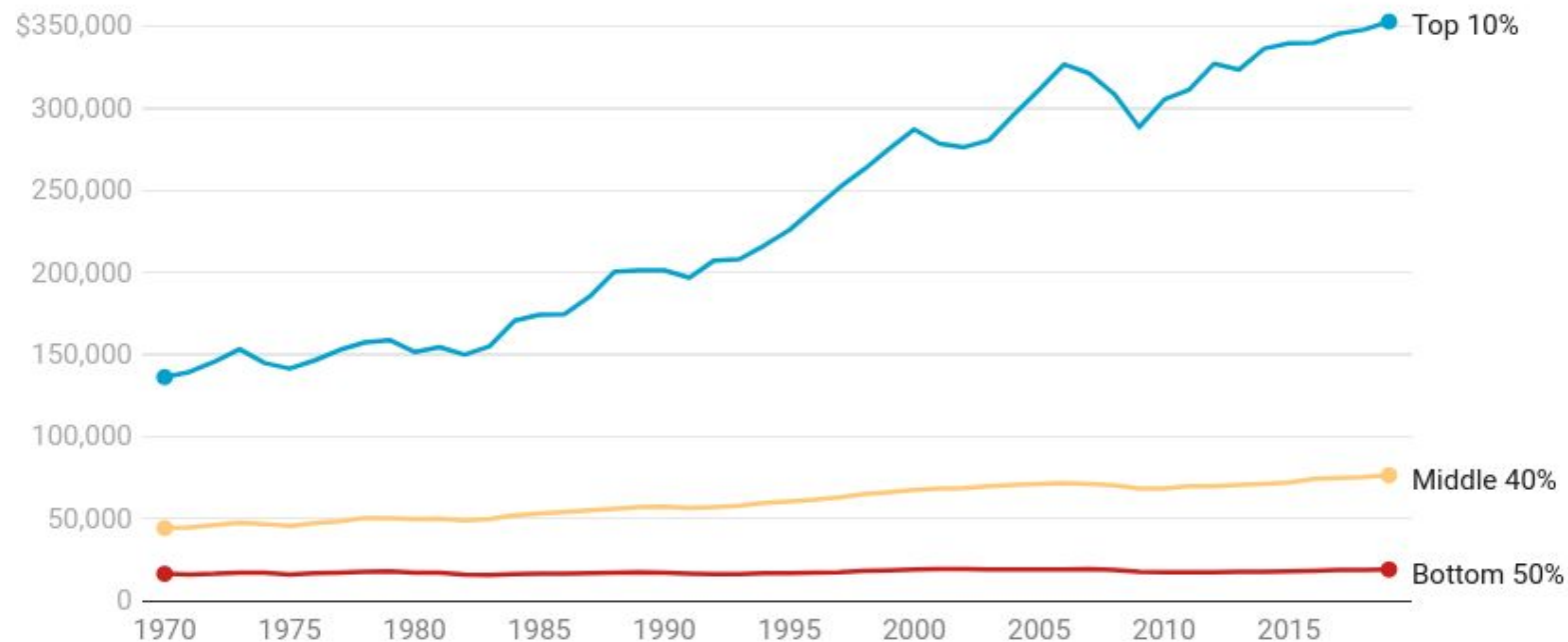
Table 0.1: Average US Adult Income, 1970-2019

| 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](https://handsondataviz.org/believe.html), accessed 2020

<https://handsondataviz.org/believe.html>

Average US Adult Income, by Percentile, 1970-2019

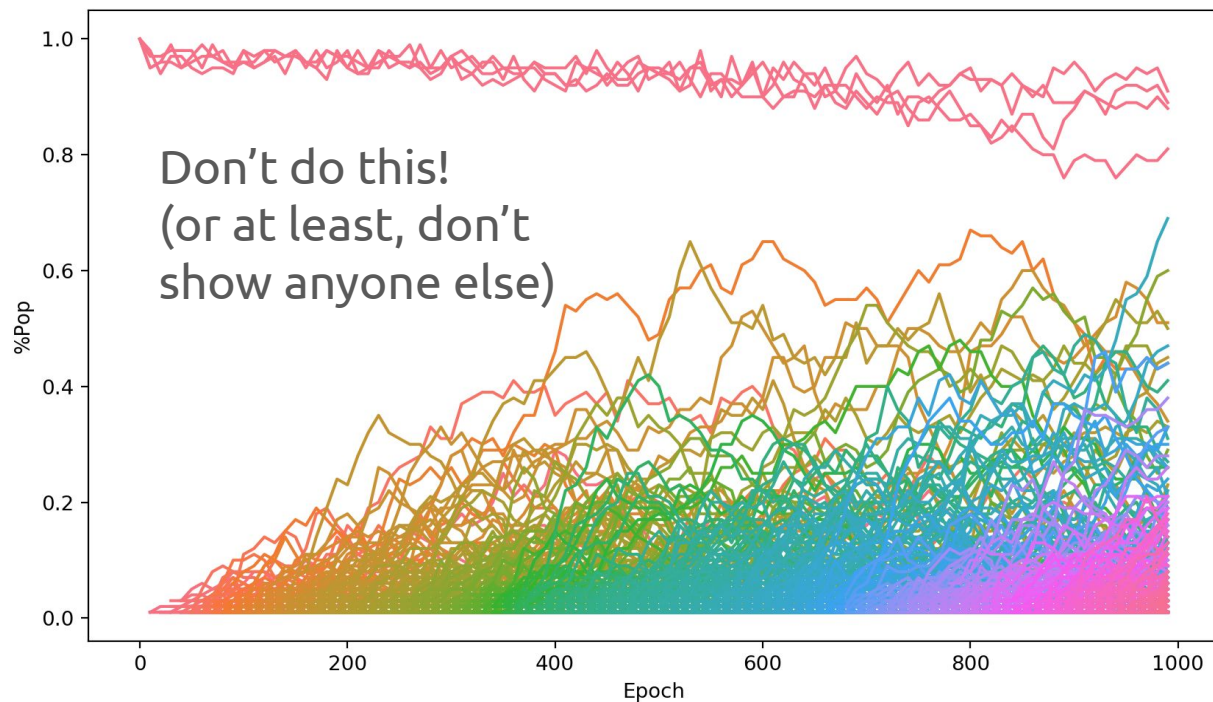


Note: Shown in constant 2019 US dollars. National income for individuals aged 20 and over, prior to taxes and transfers, but includes pension contributions and distributions.

Chart: by HandsOnDataViz • Source: [World Inequality Database 2020](#) • [Get the data](#) • Created with [Dataviz](#)

Good communication requires selection & summarization

If you plot everything, you show nothing.

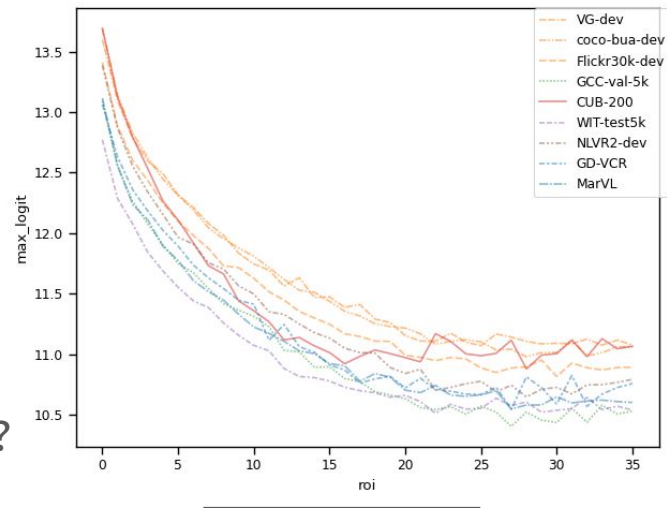


Two types of visualizations

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?

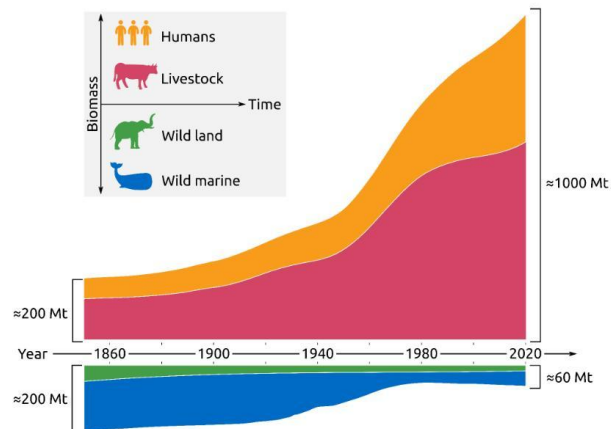


2. Persuasive visualizations

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!



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

| country | year | cases | population |
|-------------|------|--------|------------|
| Afghanistan | 1999 | 745 | 19987071 |
| Afghanistan | 2000 | 2666 | 20095360 |
| Brazil | 1999 | 31737 | 172006362 |
| Brazil | 2000 | 80488 | 174004898 |
| China | 1999 | 212258 | 1272915272 |
| China | 2000 | 216706 | 1280425583 |

variables

| country | year | cases | population |
|-------------|------|--------|------------|
| Afghanistan | 1999 | 745 | 19987071 |
| Afghanistan | 2000 | 2666 | 20095360 |
| Brazil | 1999 | 31737 | 172006362 |
| Brazil | 2000 | 80488 | 174004898 |
| China | 1999 | 212258 | 1272915272 |
| China | 2000 | 216706 | 1280425583 |

observations

| country | year | cases | population |
|-------------|------|--------|------------|
| Afghanistan | 99 | 745 | 19987071 |
| Afghanistan | 00 | 2666 | 20095360 |
| Brazil | 99 | 31737 | 172006362 |
| Brazil | 00 | 80488 | 174004898 |
| China | 99 | 212258 | 1272915272 |
| China | 00 | 216706 | 1280425583 |

values

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

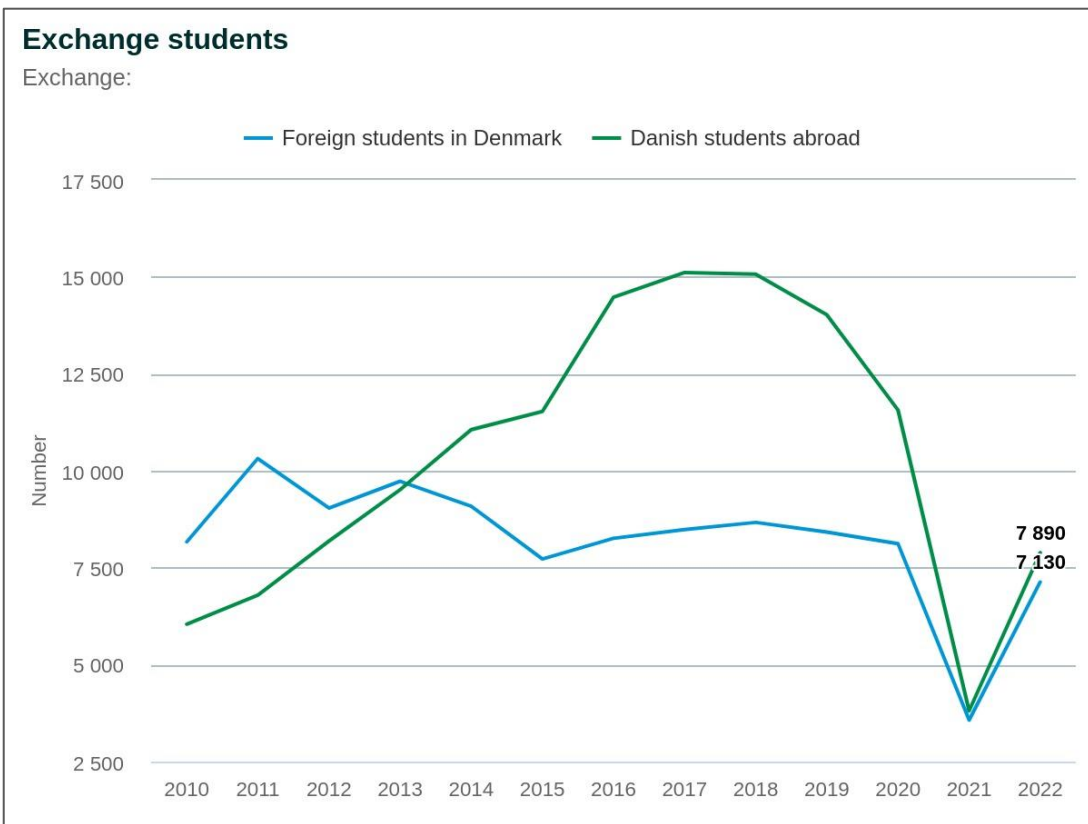
Pairwise, 4 min: What is the data behind your charts?

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?



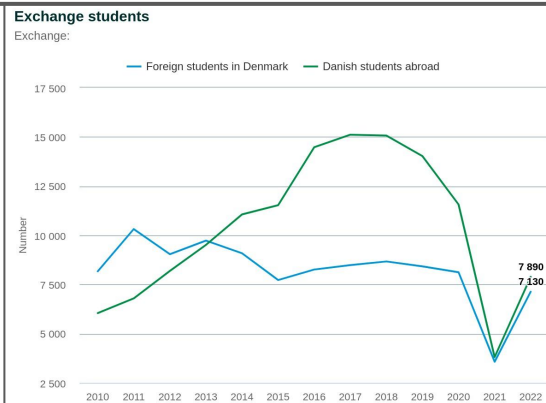
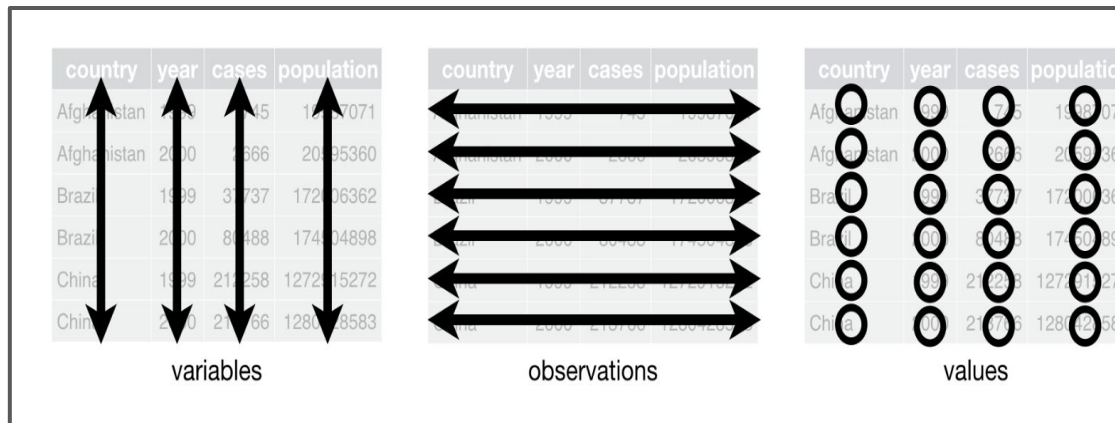
Pairwise, 5 min: What is the data behind your charts?

Look at your charts to determine:

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

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?



What did you find?

Have data ⇒ need chart: best practices

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

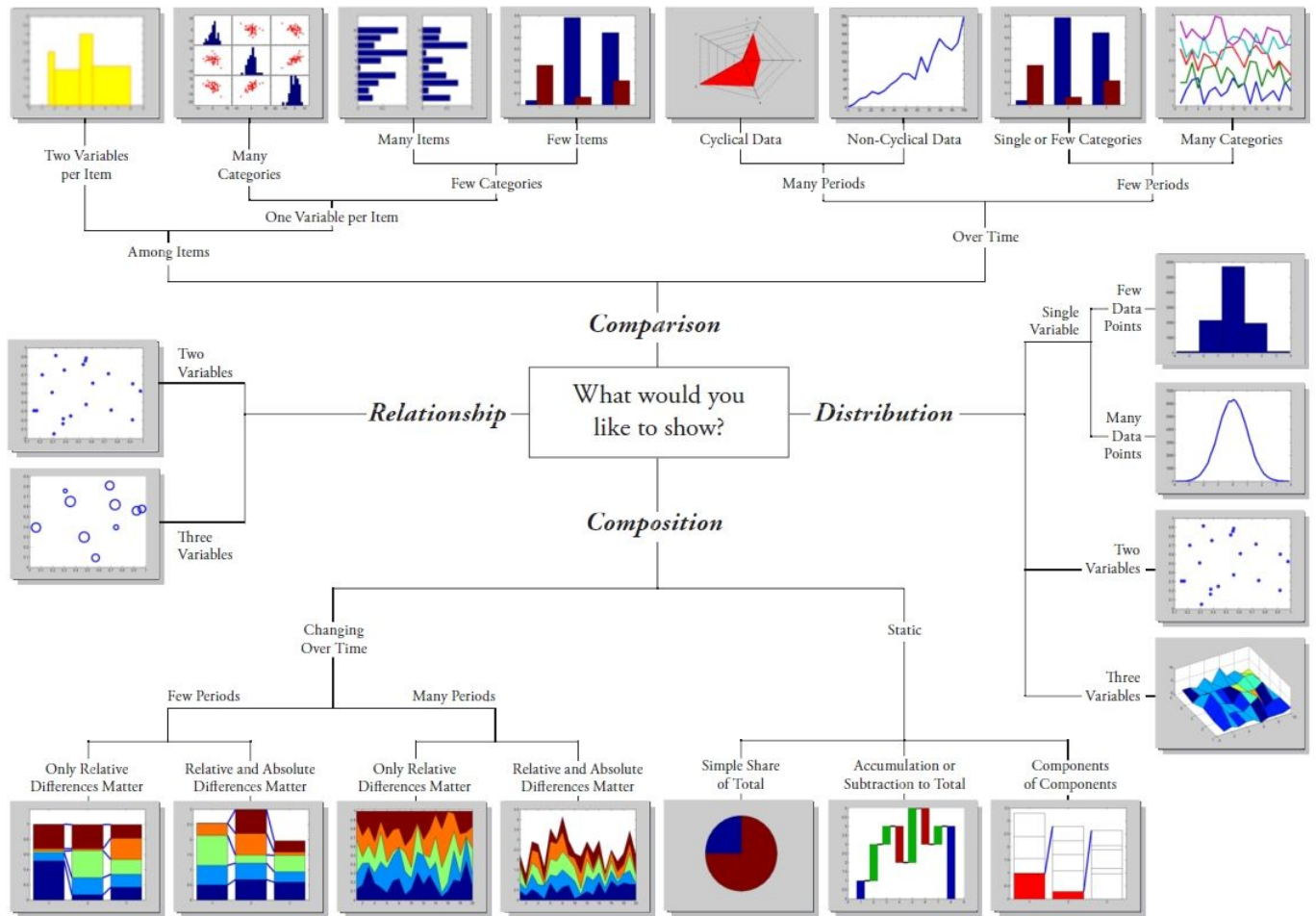
Deciding which type of chart to construct depends on data & overall story.

↪ 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



FT's Visual Vocabulary

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

| Deviation | Correlation | Ranking | Distribution | Change over Time | Magnitude | Part-to-whole | Spatial | Flow |
|--|---|---|---|--|---|--|--|---|
| Emphasize variations (+/-) from a fixed reference point. Typically the reference point is zero but it can also be a target or a long-term average. Can also be used to show sentiment (positive/negative/neutral). | Show the relationship between two or more variables. Be mindful that unless you tell them otherwise, many readers will assume the relationships you show them to be causal (i.e. one causes the other). | Use where an item's position in an ordered list is more important than its absolute or relative value. Don't be afraid to highlight the points of interest. | Show values in a dataset and how often they occur. The shape (or 'skew') of a distribution can be a memorable way of highlighting the lack of uniformity or equality in the data. | Give emphasis to changing trends. These can be short (one-day) movements or extended series traversing decades or centuries. Choosing the correct time period is important to provide suitable context for the reader. | Show size comparisons. These can be relative (just being able to see larger/smaller) or absolute (need to see fine differences). Usually these show a 'counted number' (for example, barrels, dollars or people) rather than a calculated rate or per cent. | Show how a single entity can be broken down into its component elements, if the reader's interest is solely in the size of the components, consider a magnitude-type chart instead. | Aids the reader values or intensity of movement between two or more states or conditions. These might be logical sequences or geographical locations. | Shows the reader values or intensity of movement between two or more states or conditions. These might be logical sequences or geographical locations. |
| Example FT uses Trade surplus/deficit, climate change | Example FT uses Inflation and unemployment, income and life expectancy | Example FT uses Wealth, deprivation, league tables, constituency election results | Example FT uses Income distribution, population (age/sex) distribution, revealing inequality | Example FT uses Share price movements, economic time series, sectoral changes in a market | Example FT uses Commodity production, market capitalisation, volumes in general | Example FT uses Fiscal budgets, company structures, national election results | Example FT uses Population density, natural resource locations, natural disaster risk/impact, catchment areas, variation in election results | Example FT uses Movement of funds, trade, migration, lawsuits, information, relationship graphs. |
| Deriving bar  A simple standard bar chart that can handle both negative and positive magnitude values. | Scatterplot  The standard way to show the relationship between two continuous variables, each of which has its own axis. | Ordered bar  Standard bar charts display the ranks of values much more easily when sorted into order. | Histogram  The standard way to show a statistical distribution - keep the gaps between columns small to highlight the 'shape' of the data. | Line  The standard way to show a changing time series. If data are irregular, consider markers to represent data points. | Column  The standard way to compare the size of things. Must always start at 0 on the axis. | Stacked column/bar  A simple way of showing part-to-whole relationships but can be difficult to read with more than a few components. | Basic choropleth (rate/ratio)  The standard approach for putting data on a map - should always be rates rather than totals and use a sensible base geography. | Sanku  Shows changes in flow from one condition to at least one other, good for tracing the eventual outcome of a complex process. |
| Deriving stacked bar  Perfect for presenting survey results which involve sentiment (eg. disagree/neutral/agree). | Column + line timeline  A good way of showing the relationship between an amount (columns) and a rate (line). | Ordered column  See above. | Dot plot  A simple way of showing the change or range (complexity) of data across multiple categories. | Column  Columns work well for showing change over time - but usually best with only one series of data at a time. | Bar  See above. Good when the data are not time series and labels have long category names. | Marimekko  A good way of showing the size and proportion of data at not too complicated. | Proportional symbol (count/magnitude)  Use for totals rather than rates - be wary that small differences in data will be hard to see. | Waterfall  Designed to show the sequencing of data through a flow process, typically budgets. Can include +/- components. |
| Spin  Splits a single value into two contrasting components (eg. male/female). | Connected scatterplot  Usually used to show how the relationship between 2 variables has changed over time. | Ordered proportional symbol  Use when there are big variations between values and/or seeing fine differences between data is not as important. | Dot strip plot  Good for showing individual values in a distribution, can be a problem when has many data have the same value. | Column + line timeline  A good way of showing the relationship over time between an amount (columns) and a rate (line). | Paired column  As per standard column but allows for multiple series. Can become tricky to read with more than 2. | Pie  A common way of showing part-to-whole data - but be aware that it's difficult to accurately compare the size of the | Flow map  For showing unambiguous movement across a map. | Chord  A complex but powerful diagram which can illustrate 2-way flows (and net flows) in a matrix. |

Relationship

Comparison

Composition

Distribution

Which category fits your chart?

<https://github.com/FinancialTimes/chart-doctor/blob/66caa2f126d223b3c68c7360888a035123c4ede6/visual-vocabulary/Visual-vocabulary-en.pdf>

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

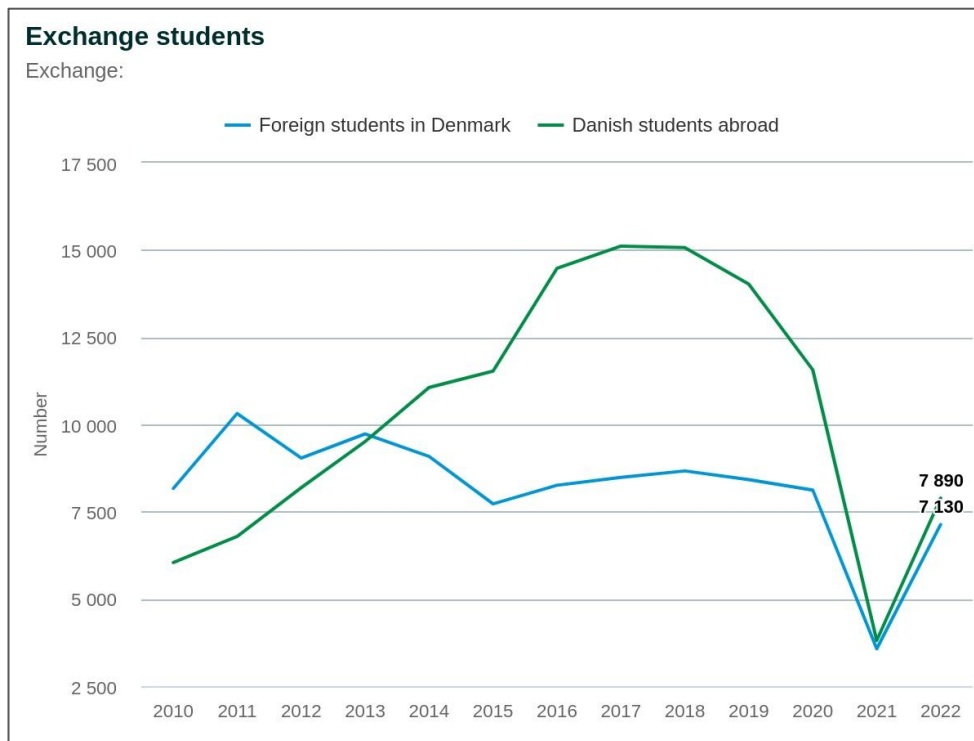
Person A: Explain the main point the chart is communicating.

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.)

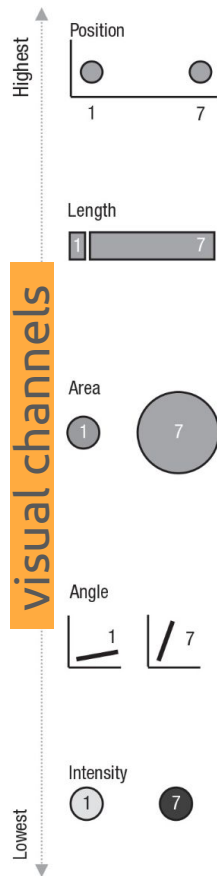


What did you find?

Strengths in Visual Processing

Absolute Precision Ranking for Seeing a Single Ratio

Visual estimation of the 1:7
ratio is noisier toward bottom



Vision Is Powerful for Global Statistics

For each visualization, statistics
are available quickly

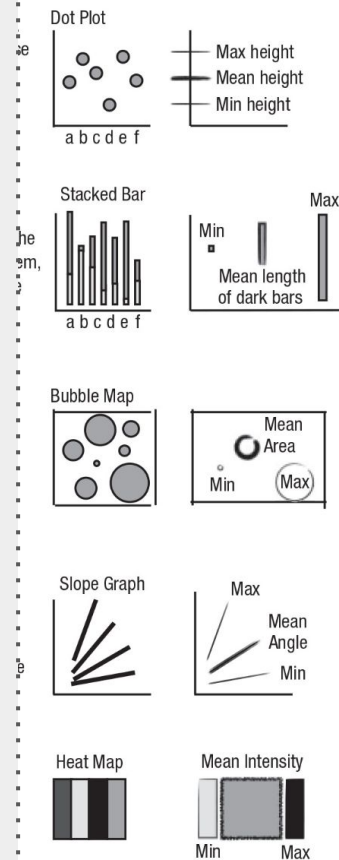
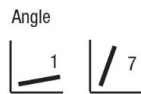
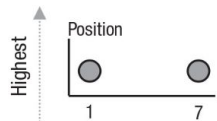


Figure 2 from:
Franconeri, S. L., Padilla, L.
M., Shah, P., Zacks, J. M., &
Hullman, J. (2021). The
Science of Visual Data
Communication: What
Works. *Psychological Science*
in the Public Interest, 22(3),
110-161.
<https://doi.org/10.1177/15291006211051956>

Strengths and Weaknesses in Visual Processing

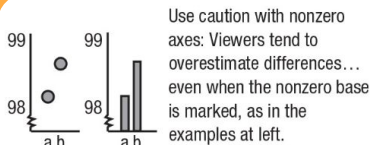
Absolute Precision Ranking for Seeing a Single Ratio

Visual estimation of the 1:7 ratio is noisier toward bottom

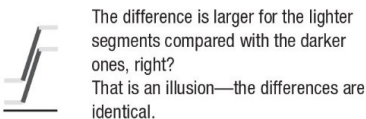
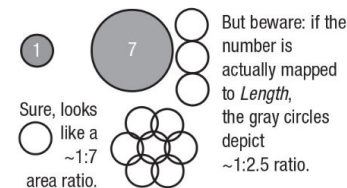
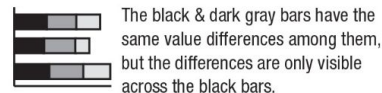


Common Illusions That Distort Data

Caveats for the visual encoding in each row

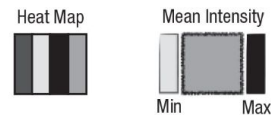
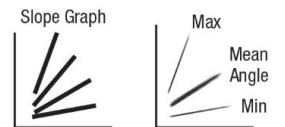
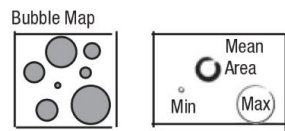
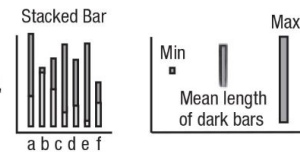
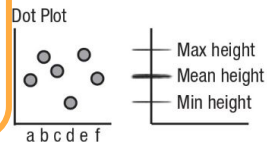


Stacked bar: Bars on baseline are position-coded = more precise perception.



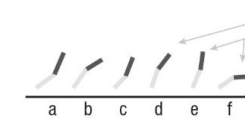
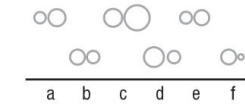
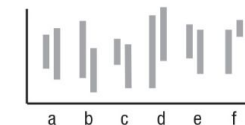
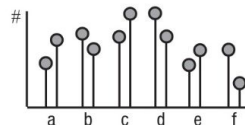
Vision Is Powerful for Global Statistics

For each visualization, statistics are available quickly



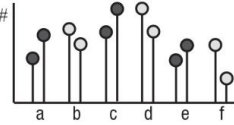
Vision Is Sluggish for Comparisons

Isolating pairs with "larger second values" is tough...



So guide viewers to the right comparisons

Tool: Shortcut comparisons by adding direct depictions of the deltas, as below



Tool: Highlight and annotate the right comparisons for your viewers, as above.

Tool: You and your viewers will (generally) compare values that (a) are close together or connected and (b) have similar colors, in that priority order.

For color heat maps, depict deltas as blue (+) & red (-)
[green/red is unsafe for colorblindness]

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/10.1177/15291006211051956>

Bars and Lines

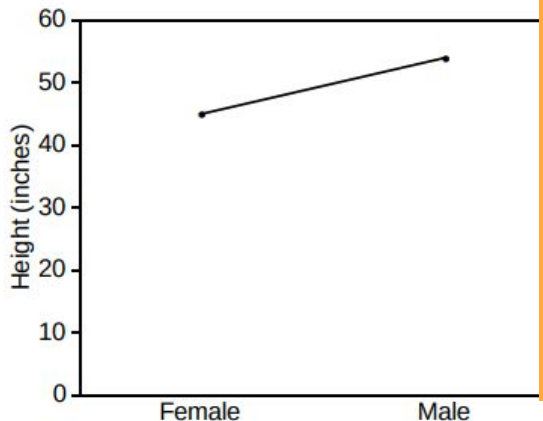
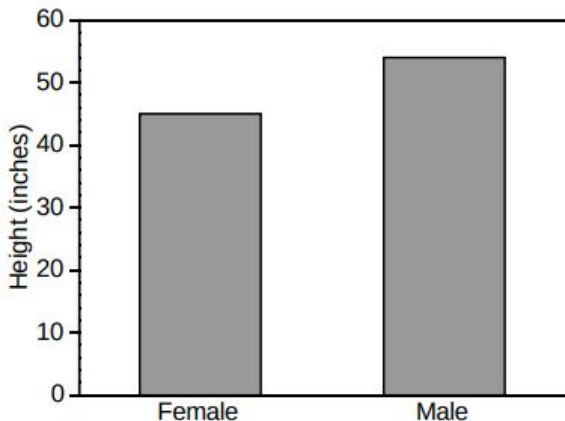
Zacks & Tversky 1999

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

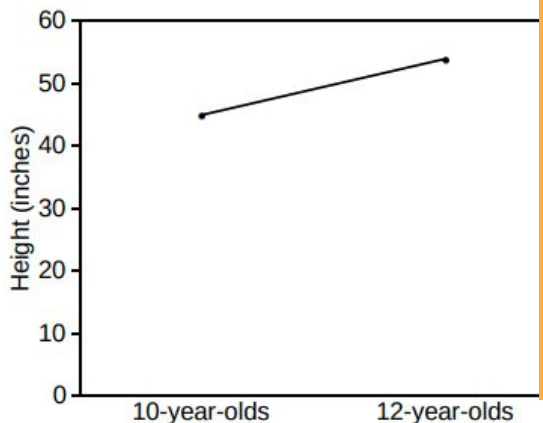
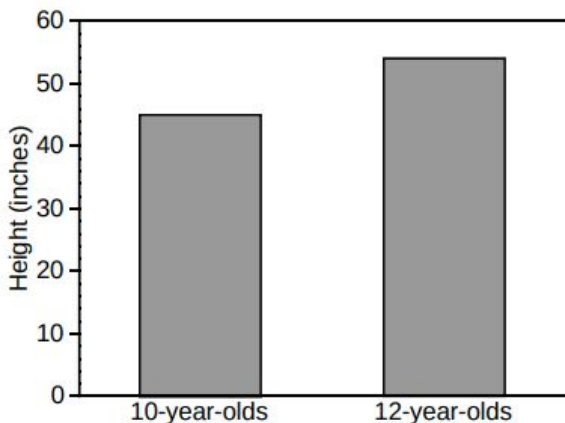
Results: Mostly congruent BUT:
"effect of graph type was about twice that of conceptual domain"

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!



Discrete Groups



Continuous Var.

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

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