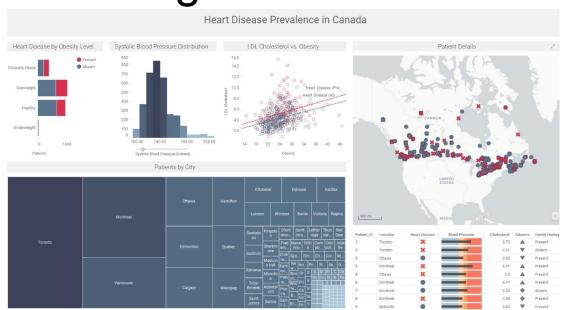
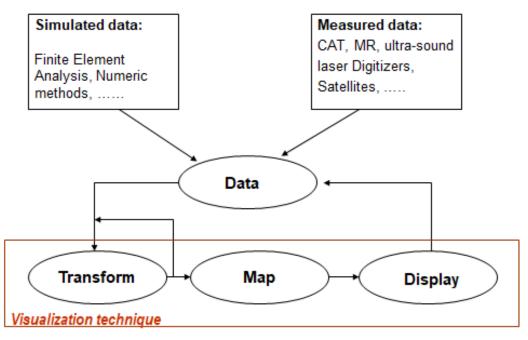


Creating a Visualization



heart-disease-prevalence

Scientific Visualization reference model



(adapted from Schroeder et al., 2006)



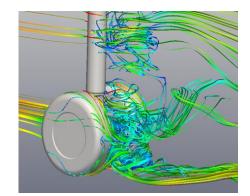
The visualization creator is involved in all phases

The user should get insights from the visualization

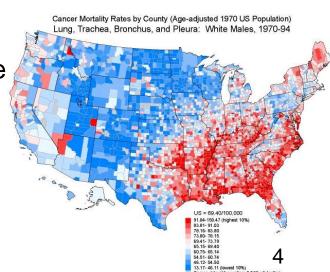
Visual mapping

7 2.5 2 1.5 1 0.5 8

- It is necessary to decide:
 - which visual structures use to **represent** the data
 - their location in the display
- Some types of data can be easily mapped to a spatial location



- Examples:
- . data with a topological or geographical structure
- Abstract data don't have an easy correspondence with the dimensions of the physical space around us



Three **structures** must be defined in the **visual mapping/encoding**:

- spatial substrate
- graphical elements
- graphical properties
- Spatial substrate dimensions in physical space where the visual representation is created (can be defined in terms of axes and type of data)
- **Graphical elements** anything visible appearing in the space points, lines, surfaces, volumes
- **Graphical properties** properties of the graphical elements to which the human retina is very sensitive **retinal variables**:

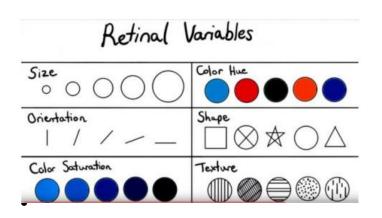
size, orientation, color, texture, and shape

- **Spatial substrate** axes (x, y, ...) type of data (quantitative, ordinal, categorical)

volumes

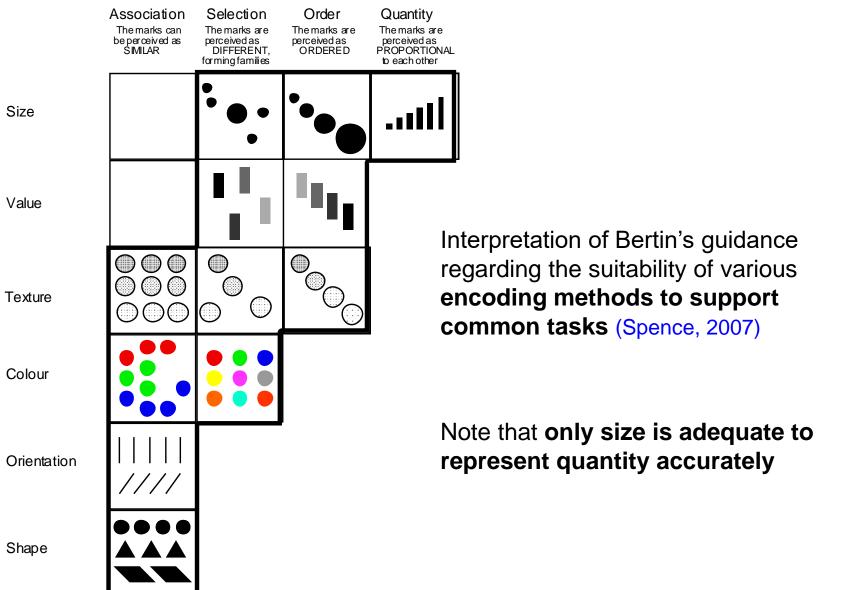
- Graphical elements points lines surfaces

- **Graphical properties** retinal variables:

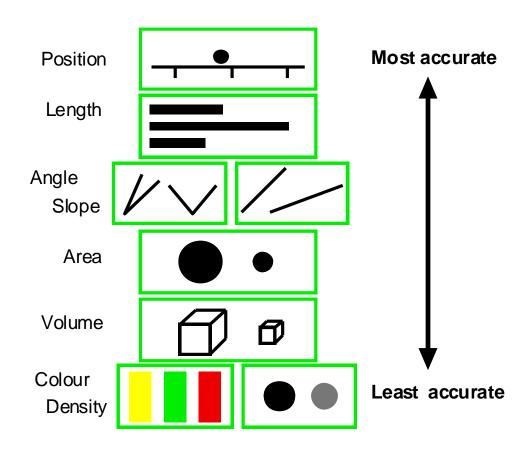


size,
orientation
color (depends on physiology and culture)
texture
shape

How to select visual encodings?



How to select visual encodings to accurately represent quantity?



The relative difficulty of **assessing quantitative value** as a function of encoding mechanism, as established by Cleveland and McGill (Spence, 2007)

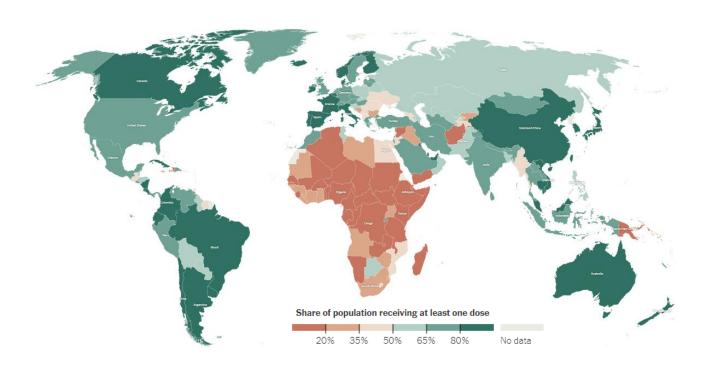
In a nut shell: Do you have a lot of data?

Visualization may be the solution (or part of it)

- Creating a Visualization has several phases
- Visual mapping is core
- There are several possible visual encodings/ visualization techniques
- But,
 How to select techniques? → next topic

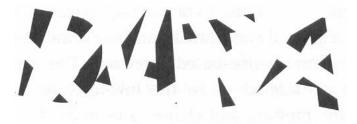


Mapping - Visually encoding value



Remember:

 The Human Visual system is the product of millions of years of evolution



 Although very flexible, it is tuned to data represented in specific ways



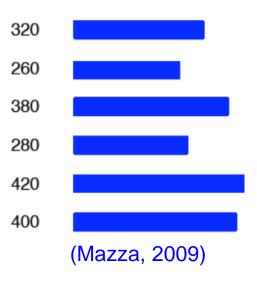
 If we understand how its mechanisms work we will be able to produce better results

Pre-attentive attributes can help observers to see before though

Example: Count the number of 7s

https://www.youtube.com/watch?time _continue=121&v=AiD6etOB6qI Some visual attributes as size, proximity are also quickly processed by visual perception, before the cognitive processes come into play

Example: mapping numerical values to the length of bars:



Procedure to follow to create visual representations of data

- 1. Define the problem and the users' questions
- 2. Examine the **nature of the data** to represent and **pre-process** the data
- 3. Determine the **number of attributes**/variables/dimensions
- 4. Choose the **visual structures to map**

test several ideas ...

Important aspects to consider:

nature of the problem explore confirm

nature of the data to represent ordinal categorical

number of attributes

univariate
bivariate
trivariate
multivariate

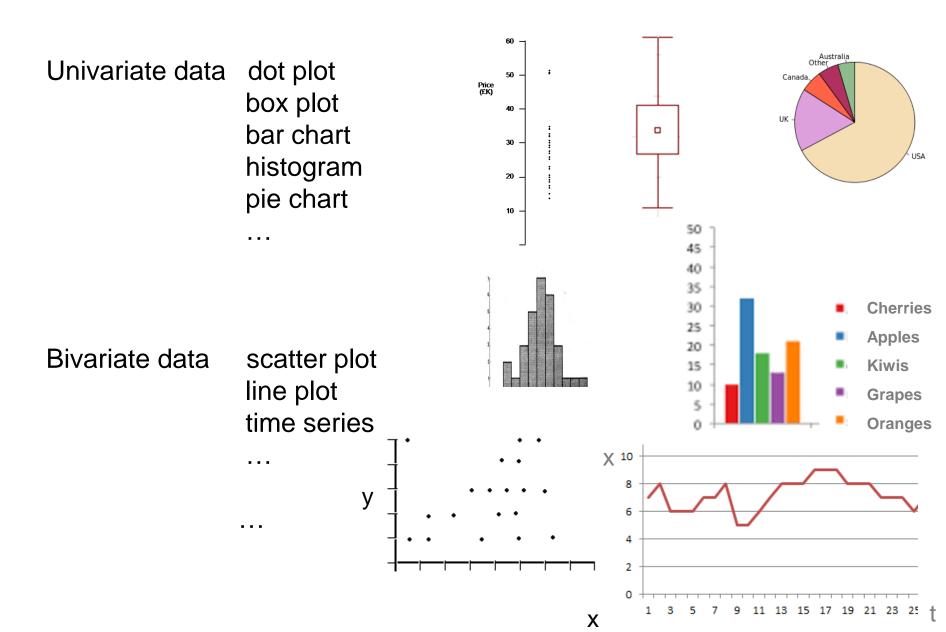
Next: visualization techniques organized according the n. of attributes

tables
networks
spatial or geographical fields
geometry

of tabular data

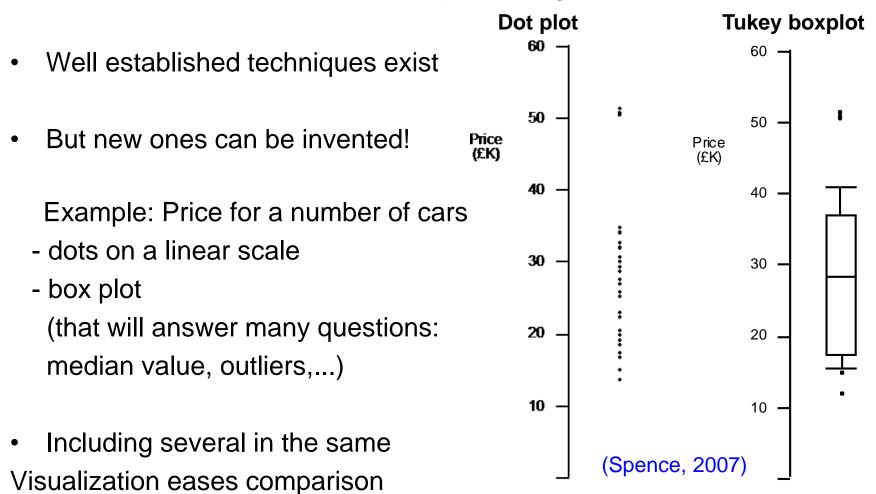


Common Visualization Techniques to visually represent univariate, bivariate data



Representing univariate data

A common situation consists in representing a set of values



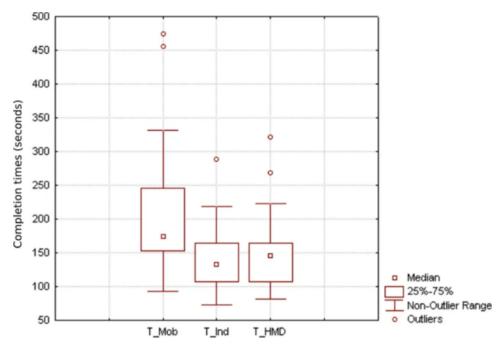
https://www.data-to-viz.com/caveat/boxplot.html

Example:

Including several box plots in the same Visualization allows comparing:

Comparing 3 augmented reality visualization methods for assembly procedures (times)

https://link.springer.com/article/ 10.1007/s10055-021-00557-8



Task performance from the three methods based on time of completion

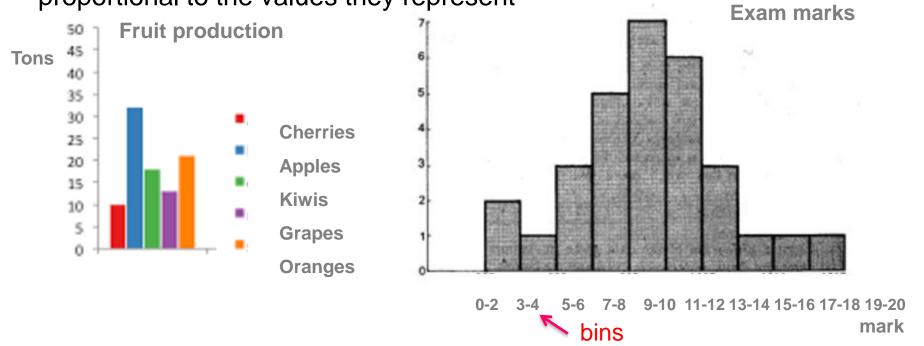
Two common techniques not to be confounded!

Histogram represents a distribution of numerical data

Bar chart represents the number of occurrences of a categorical/

ordinal data

Both represent data by rectangular bars (vertical or horizontal) with length proportional to the values they represent



Another simple (and too common) technique

Pie Chart

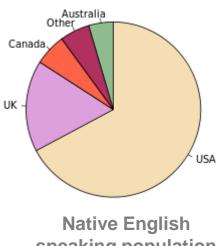
Represents numerical proportion, parts of an whole

The arc length of each slice (its central angle and area), is proportional to the quantity it represents

Are much controversial: many experts recommend avoiding them http://www.perceptualedge.com/articles/08-21-07.pdf

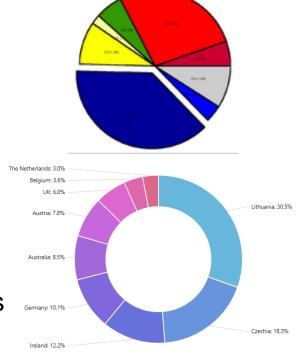


It is difficult to compare different sections of a pie chart, or to compare data across different pie charts



speaking population

Variations of pie charts:



- Simple criteria to determine whether a pie chart is acceptable
- Consider it only if:
- The parts make up a meaningful whole
- + The parts are mutually exclusive
- + There are <6 parts and slices have not very different sizes

If the main purpose is to compare between the parts, use a different chart!

https://eagereyes.org/techniques/pie-charts

Representing bivariate data

The scatterplot is the conventional representation

Each observation is represented by a point on a two dimensional space.

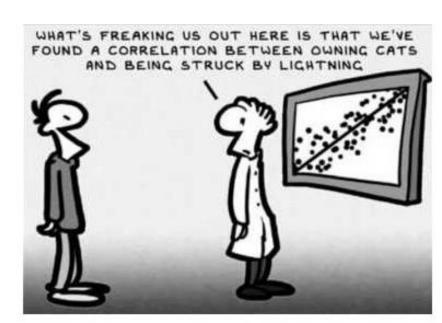
The axes are associated with these two attributes

This representation affords awareness of:

- general trends
- local trade-offs
- outliers



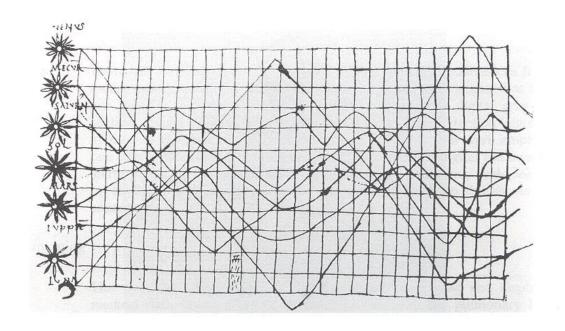
Correlation is not causation



Representing bivariate data

The line chart

One of the oldest known and ubiquitous Visualizations



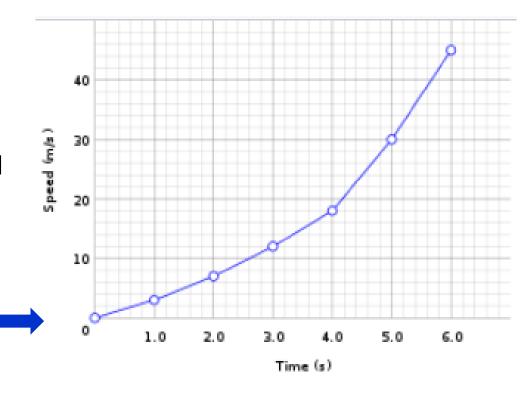
Inclination of orbits along the time - Xth century (Tufte, 1983)

 A line chart or line plot or line graph or curve chart displays information as a series of data points called 'markers' connected by straight line segments

 Basic type of chart common in many fields

 Often used to visualize a trend in data over intervals of time

 Can we use a solid line connecting the samples, in this example?

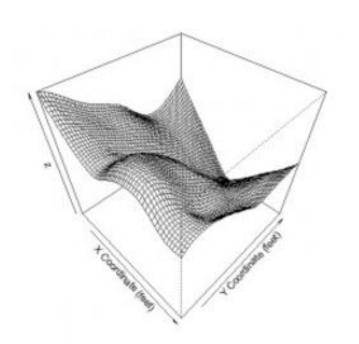


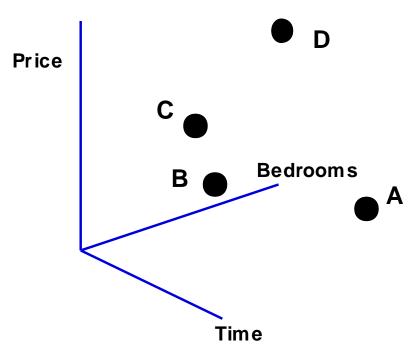
24

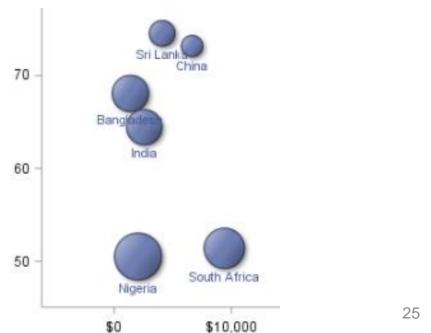
Common Visualization Techniques to represent trivariate data

Trivariate data surface plot contour plot 3D representation bubble plot/chart

. . .







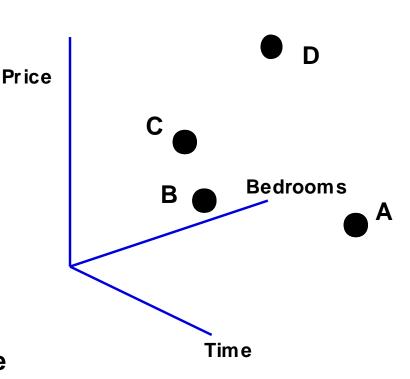
Representing Trivariate data

 Since we live in a 3D world, representing trivariate data as points in a 3D space and displaying a 2D view is natural

 However, these representations of abstract data can be ambiguous

 This can be solved by interaction, allowing the user to reorient the representation

"for 3D to be useful, you' ve got to be able to move it" (Spence, 2007)



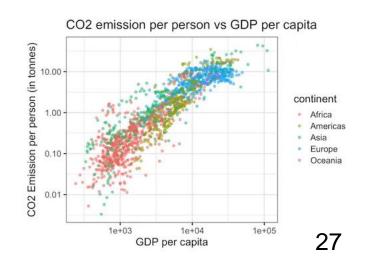
Other Simple (and common) representations of trivariate data

In a bubble chart data are represented as a disk that expresses two
of the values through the disk's xy location and the third through its

size (radius or area?)

 Mapping the variable to circle size must be done carefully. The interpretation may be ambiguous Representing one more dimension through color in a scatter plot

https://visage.co/data-visualization-101-bubble-charts/



Important aspects to consider:

nature of the problem explore confirm

nature of the data to represent ordinal categorical

number of attributes bive

univariate
bivariate
trivariate
multivariate

Next: visualization techniques organized according the n. of attributes

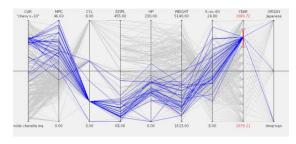
tables
networks
spatial or geographical fields
geometry

of tabular data

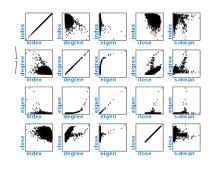


Techniques for Multivariate (or Hypervariate) data

Coordinate plots ——parallel coordinate plots



star (radar/spider) plots





Scatterplot Matrix



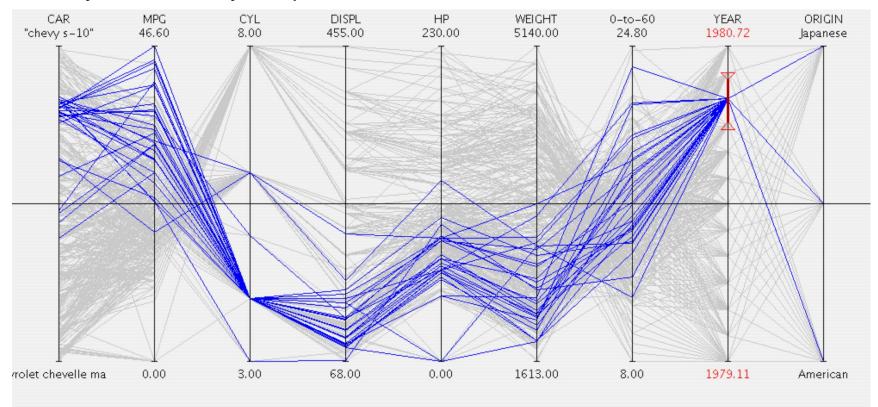
Icons/glyphs





 Parallel coordinates plots are one of the most popular techniques for hypervariate data

They have a very simple basis

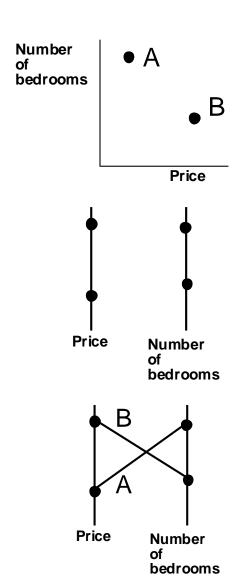


Consider a simple case of bivariate data (data concerning houses to sell)

1- A scatterplot represents the price and number of bedrooms associated with two houses

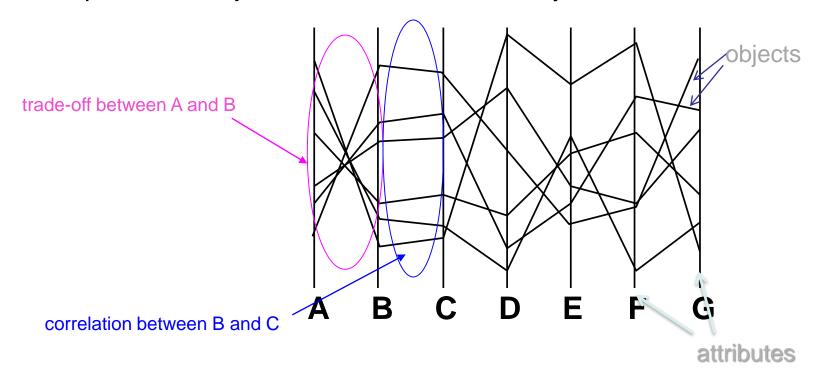
2- the axes are detached and made parallel; each house is represented by a point on each axis

3- To avoid ambiguity the pair of points representing a house are joined and labeled



 For objects characterized by many attributes the parallel coordinate plots offer many advantages

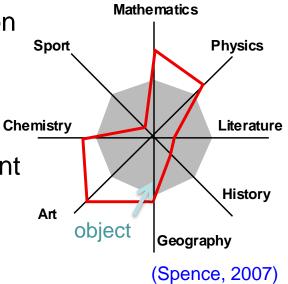
A example for six objects, each characterized by seven attributes:



The trade-off between A and B, and the correlation between B and C, are immediately apparent. The trade-off between B and E, and the correlation between C and G, are not.

Star plots have many features in common with parallel coordinate plots

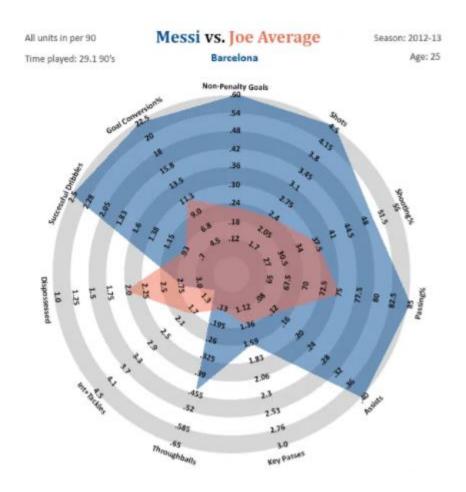
An attribute value is represented by a point on a coordinate axis

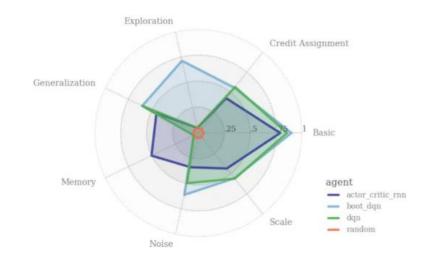


Attribute axes radiate from a common origin

For a given object, points are joined by straight lines

Other useful information such as average values or thresholds can be encoded

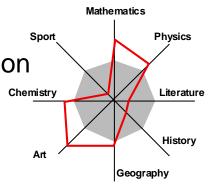




https://syncedreview.com/2019/08/16/deepmind-bsuite-evaluates-reinforcement-learning-agents/

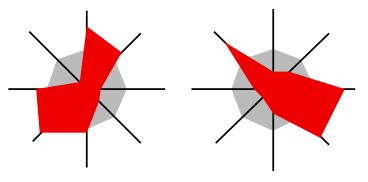
Properties of star plots:

 Their shape can provide a reasonably rapid appreciation of the attributes of the objects



They offer object visibility and are suitable to compare objects

(by visibility it is meant the ability to gain insight pre-attentively; without a great cognitive effort)



Bob's performance

Tony's performance

(Spence, 2007)

The scatterplot matrix (SPLOM) is applicable to higher n. of variables

However, as the number of attributes increases, the number of different pairs of

attributes increases rapidly:

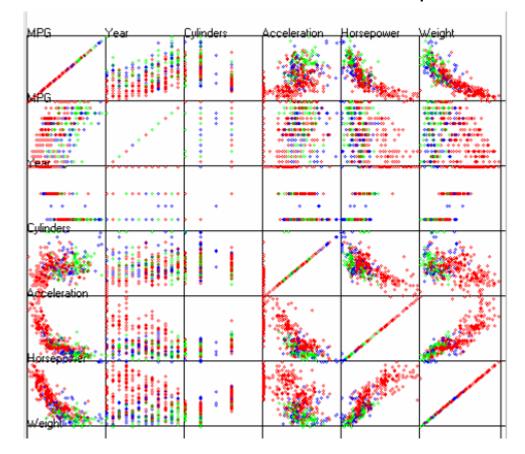
2 variables-> 1 scatterplot

3 variables -> 3 scatterplots

4 variables -> 6 scatterplots

We may try to reduce the number of dimensions keeping the more relevant:

Dimensionality reduction!

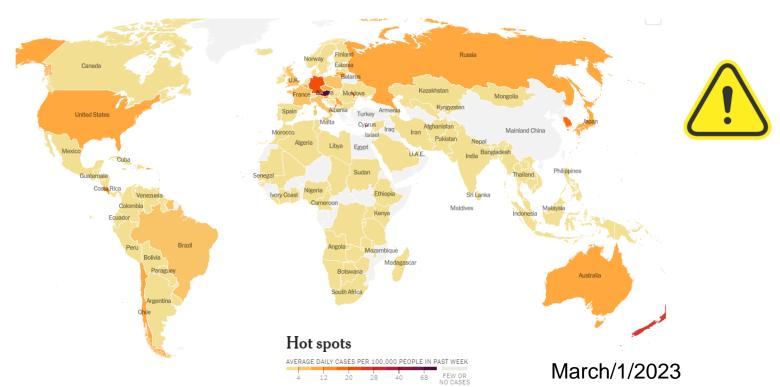


Scatterplot matrix for 6 attributes of a car dataset

Simple representations of attributes on a map

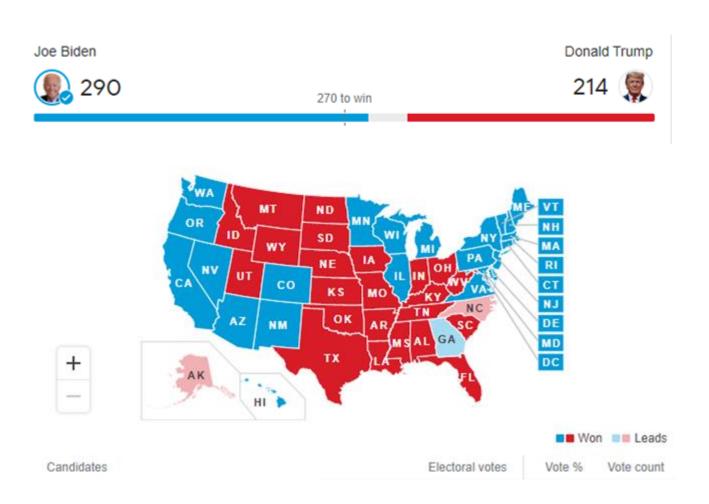
Choropleth maps - A standard approach to communicating aggregated data by geographical areas using color encoding of the geographic area

They require some care: what are the possible issues?



Visualizations of the US 2020 Election

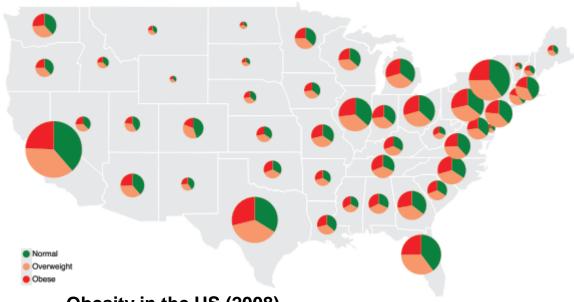
(choropleth + bar) the bar helps better understand the ratio of votes (and the number provides the maximum accuracy)



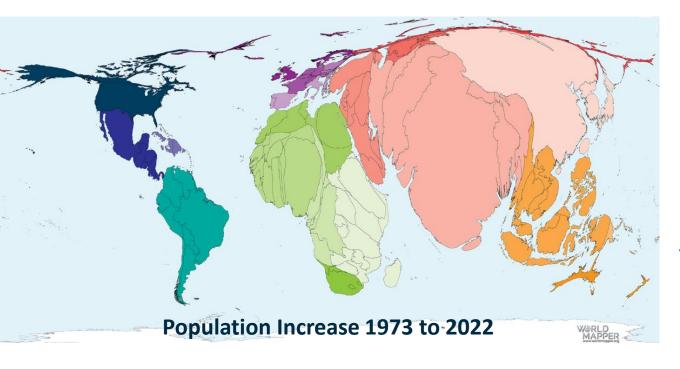
Simple representations of attributes on a map

- Graduated Symbol Maps are an alternative to the choropleth map;
- Symbols are placed over an underlying map; may show more attributes
- Avoid confounding geographic area with data values

What is missing in this visualization?



Obesity in the US (2008) (Heer et al., 2010)



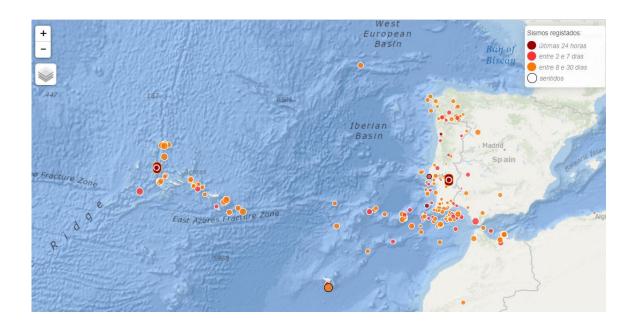
Cartograms: Using distortion to represent a variable

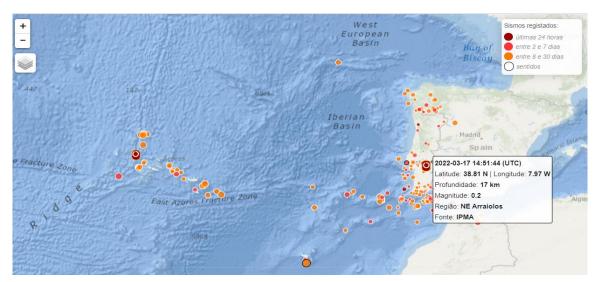
https://worldmapper.org/





Seismic activity:
Is something missing in this visualization?

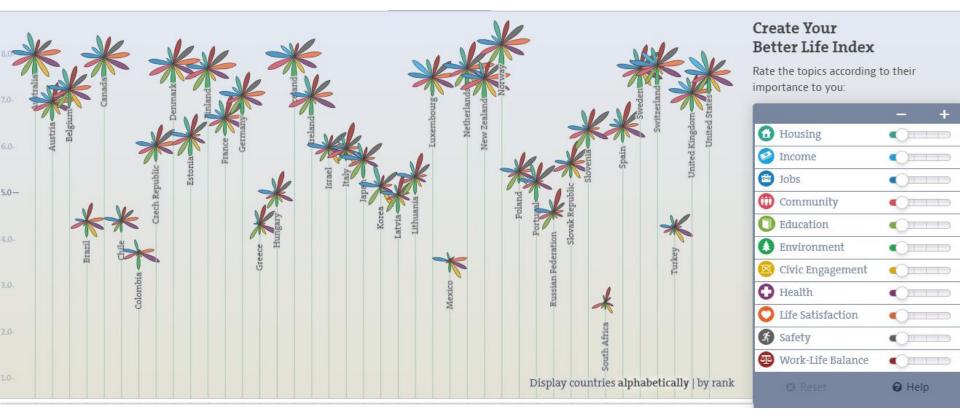




https://www.ipma.pt/pt/geofisica/sismicidade/index.jsp

Glyph chart example

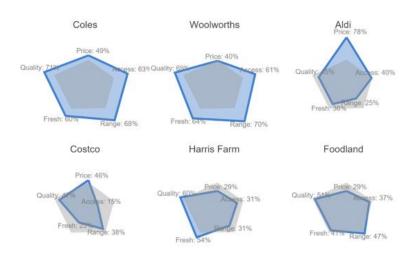
The physical properties of the shape represent different categorical variables sized according to the associated quantitative value and distinguished through color

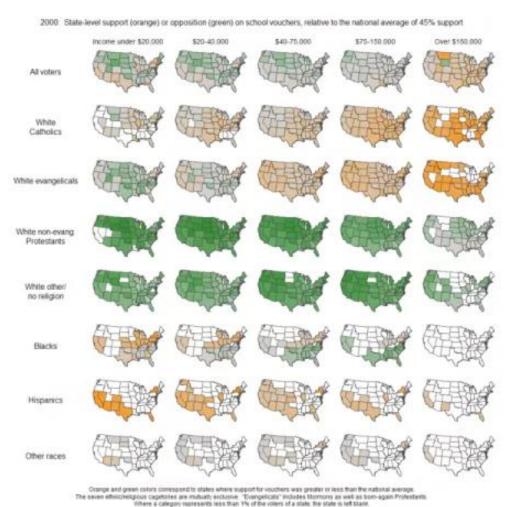


Small multiples:

arrangement approach that facilitates efficient and effective comparisons

(Kirk, 2012)





Dashboards

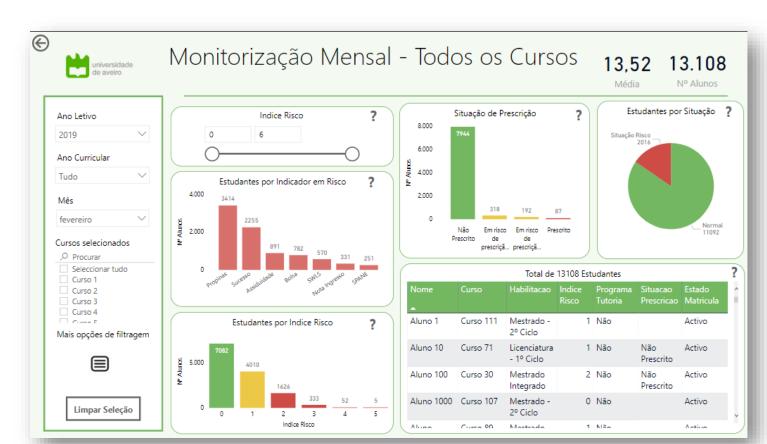
Visual display summarizing a dataset providing information at-a-glance (e.g. KPIs)

"A dashboard is a visual display of the most important information needed to achieve one or more objectives; consolidated and arranged on a single screen so the information can be monitored at a glance. " (Few, 2004)

https://www.nngroup.com/articles/dashboards-preattentive/

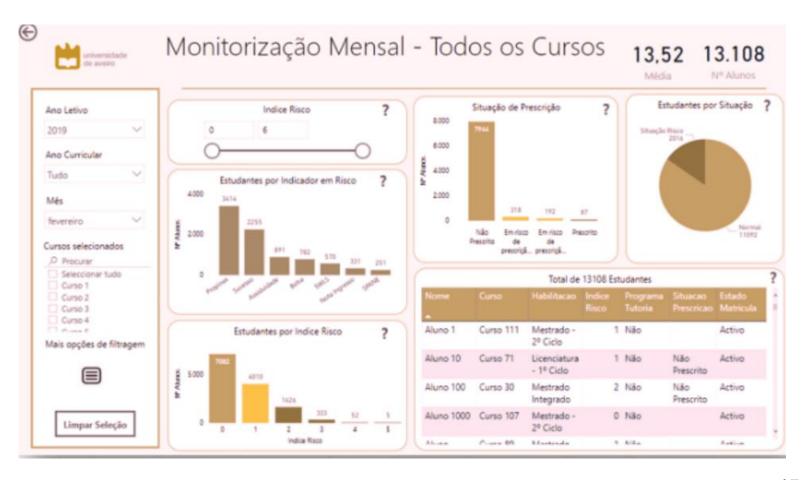
Prototype: "Portal dos indicadores, UA"

Any issues about the color?



Testing readability to color-blind users:

As seen by someone with deuteranopia (red-green blindness) https://www.color-blindness.com/coblis-color-blindness-simulator/)





nature of the data to represent ordinal categorical

number of attributes

bivariate
trivariate
trivariate
multivariate

univariate
visualization techniques
organized according the n. of
attributes

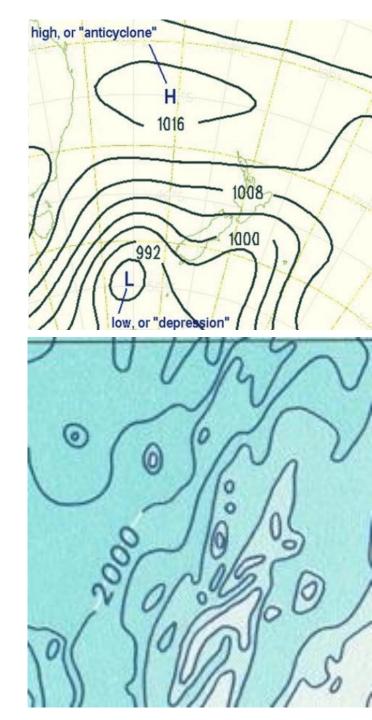
tables
networks
spatial or geographical
fields
geometry

Next: of field data

. . .

Representations of a scalar in a 2D field

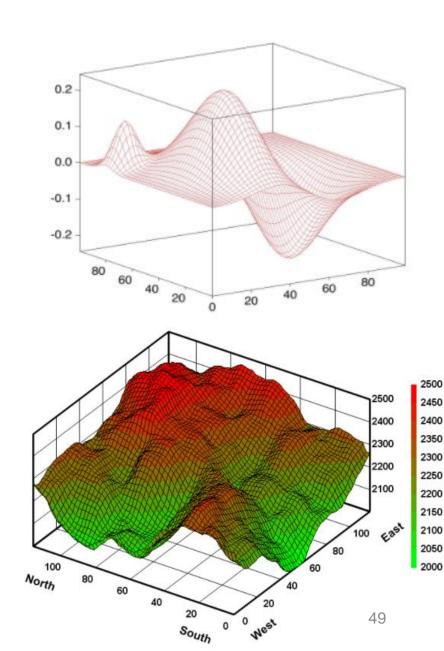
- Contour plots
- contour line (also isoline, isopleth, or equipotential curve) of a function of two variables is a curve along which the function has a constant value, so that the curve joins points of equal value.
- Typical in meteorological charts (isobars and isothermal curves)
- and maps (to represent altitude or depth)



Representations of a 3D scalar function

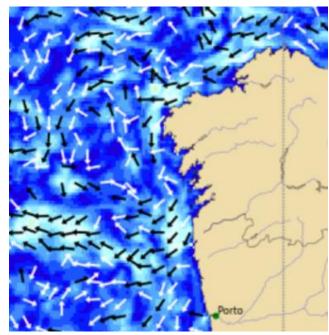
- Surface plots
- May be combined with color

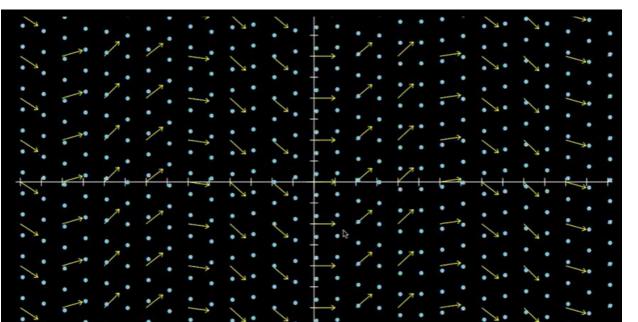
(preferably in a redundant way and carefully selecting the scale)



Representing vectors in a 2D field







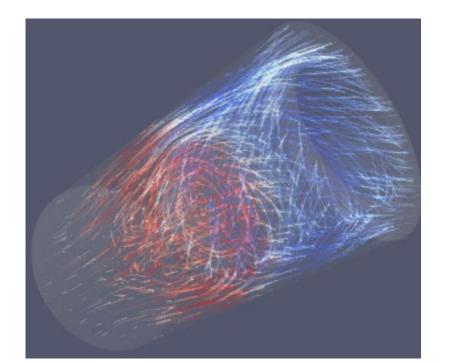
https://www.ipma.pt/pt/maritima/currents/

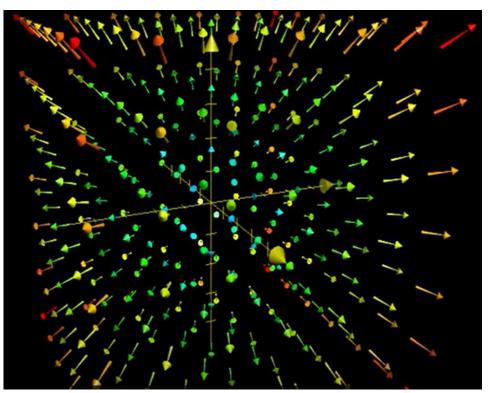
https://www.khanacademy.org/math/multivariable-calculus/thinking-about-multivariable-function/visualizing-vector-valued-functions/v/fluid-flow-and-vector-fields

Representations of vectors in a 3D field

Vector data visualization: in 3D space a vector variable is visualized

Stream Lines are another interesting representation





3d vector field example (video) | Khan Academy

New Animated Stream Lines Representation for ParaView 5.3 (kitware.com)

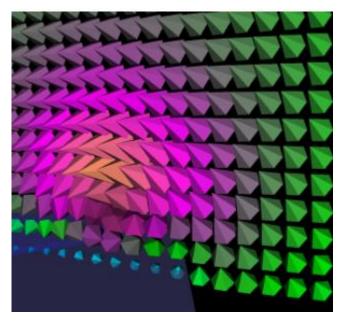
Glyphs

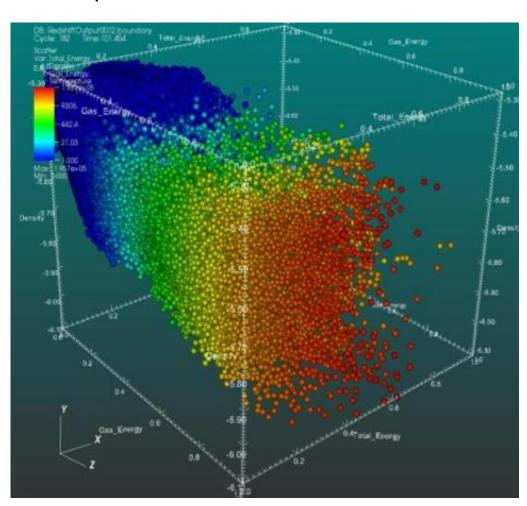
Four-dimensional data visualization: in 3D space a fourth scalar variable

is visualized using colored glyphs

Glyphs for Visualizing a 3D Vector Field







 These are only some of the visualization techniques to represent a value

There are a lot more ...

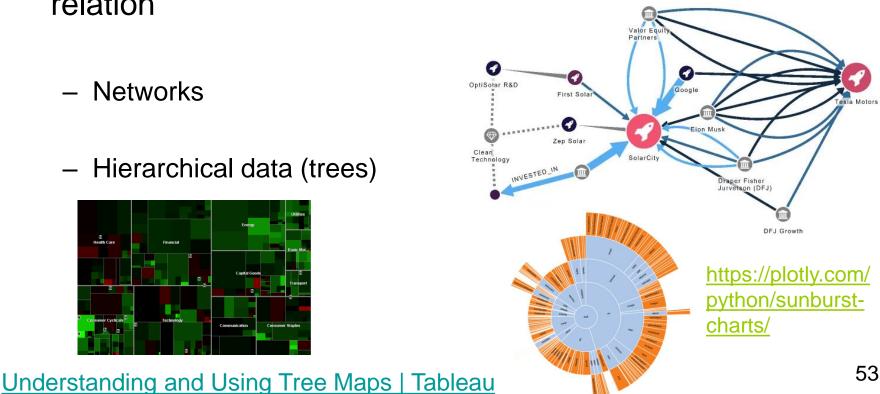
And we may want to visually represent beyond value:

relation

Networks

Hierarchical data (trees)





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 Communications of the ACM, vol. 8, n.1, 2010
 https://doi.org/10.1145/1743546.1743567