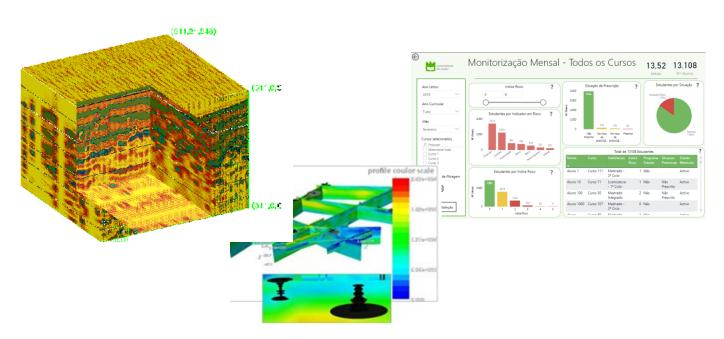
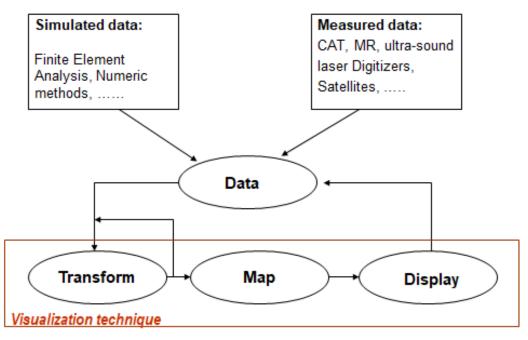




## Creating a Visualization



### Scientific Visualization reference model



(adapted from Schroeder et al., 2006)

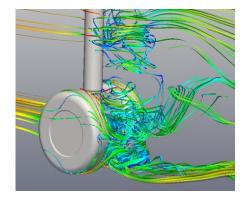


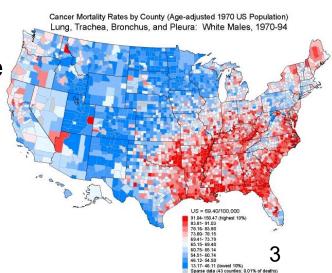
The visualization creator is involved in all phases



## Visual mapping

- 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**:

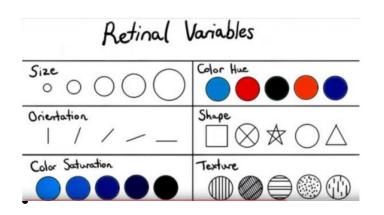
- spatial substrate
- graphical elements
- graphical properties
- Spatial substrate dimensions in physical space where the visual representation is created (can be defined in terms of axes and type of data)
- **Graphical elements** anything visible appearing in the space points, lines, surfaces, volumes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

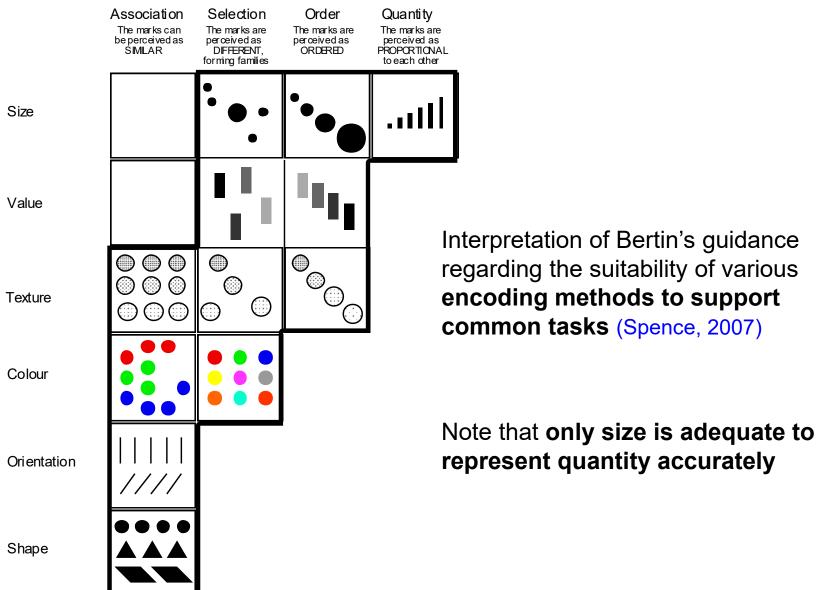
- Graphical elements points lines surfaces volumes

- **Graphical properties** retinal variables:

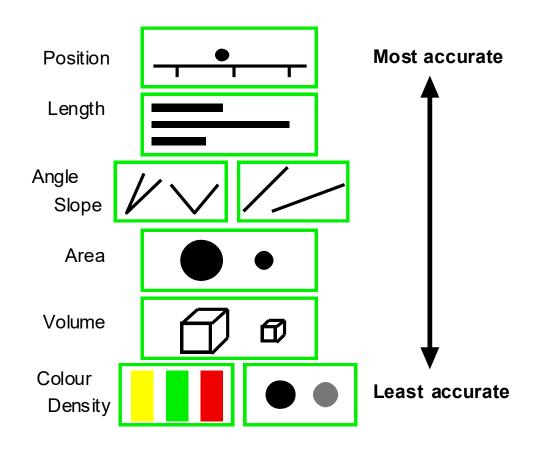


size, orientation color (depends on physiology and culture) texture shape

#### 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:

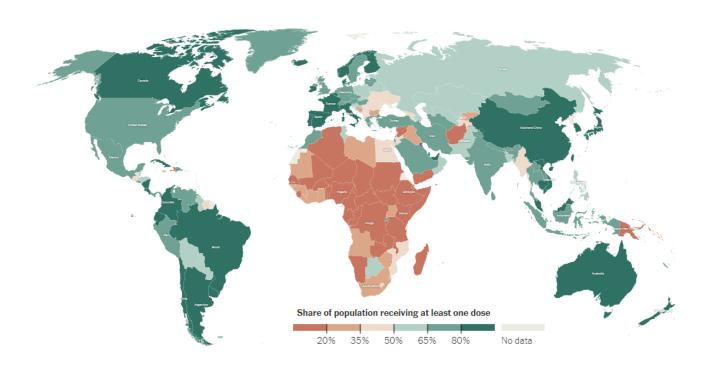
Do you have a lot of data?

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

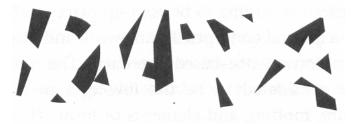


## Mapping - Visually encoding value

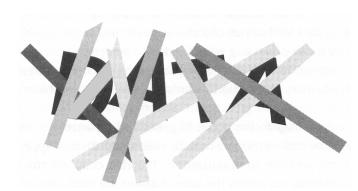


#### Remember:

 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



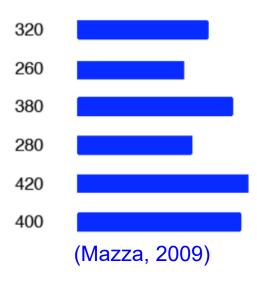
 If we understand how its mechanisms work we will be able to produce better results

## 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  • Other 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 ...

nature of the problem communicate 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

dataset types

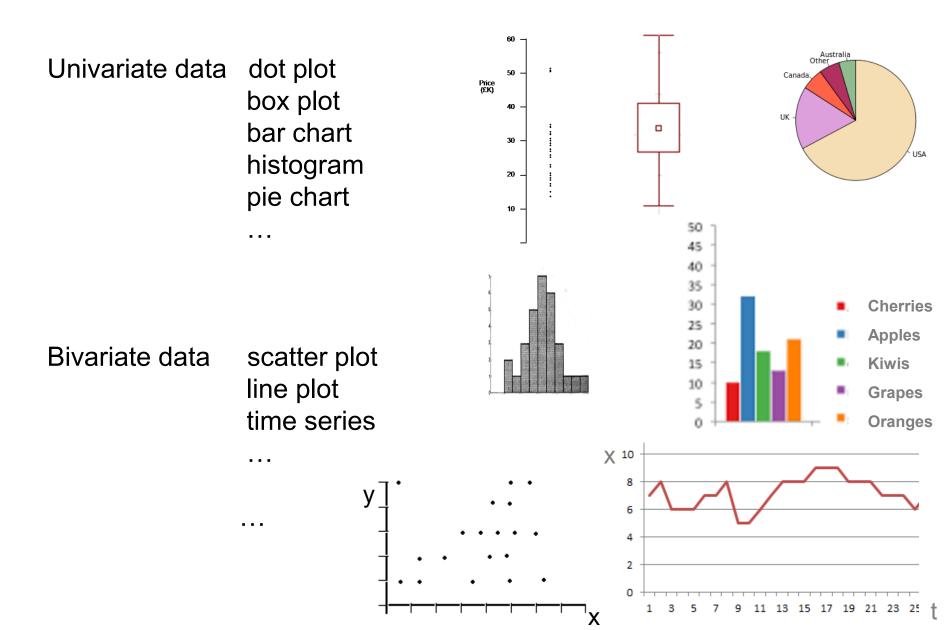
tables
networks
spatial or geographical
fields
geometry

• • •

#### of tabular data



## Common Visualization Techniques to visually represent univariate, bivariate data



## Representing univariate data

A more common situation consists in representing a set of values



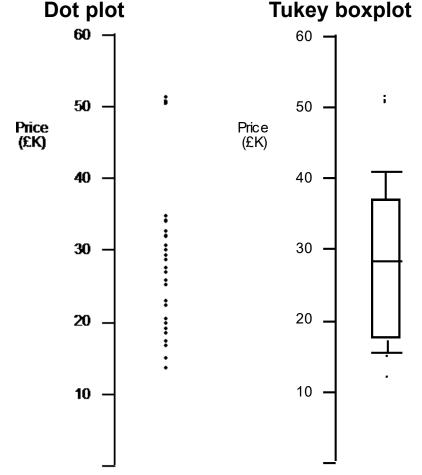
But new ones can be invented!

#### Example:

Price for a number of cars:

- dots on a linear scale
- box plot(that will answer many questions: median value, outliers,...)

(Spence, 2007)



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

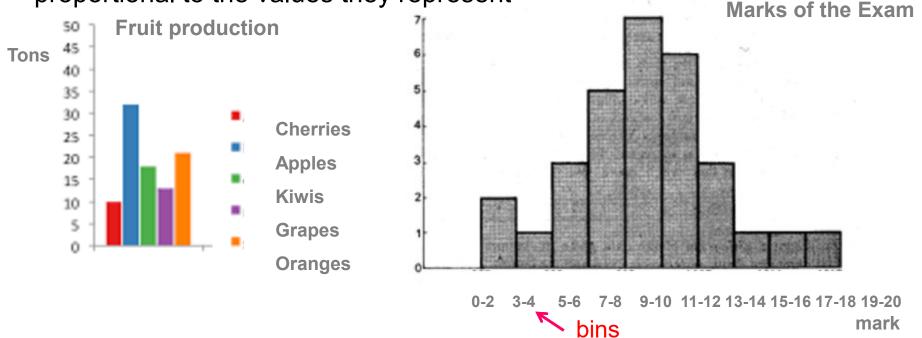
## Two common techniques not to be confounded!

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

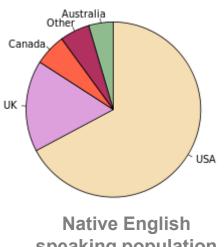
Represents numerical proportion, parts of an whole

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

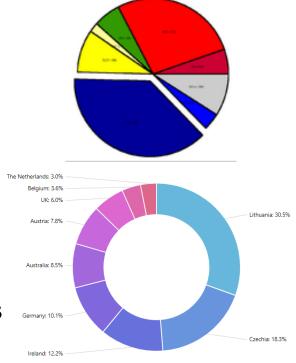


It is difficult to compare different sections of a pie chart, or to compare data across different pie charts



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</li>

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.

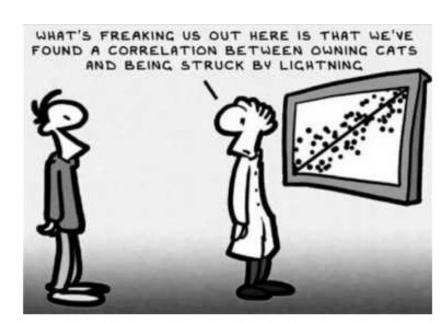
The axes are associated with these two attributes

This representation affords awareness of:

- general trends
- local trade-offs
- outliers



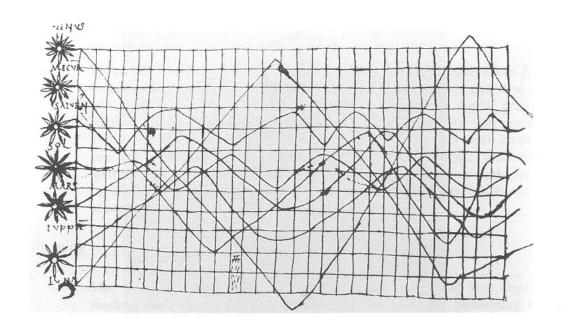
**Correlation is not causation** 



## Representing bivariate data

The line chart

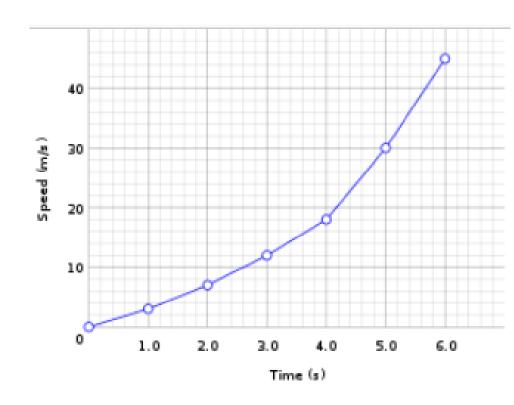
One of the oldest known and ubiquitous Visualizations



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

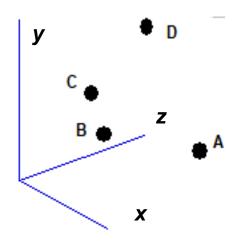
 A line chart or line plot or line graph or curve chart displays information as a series of data points called 'markers' connected by straight line segments

- Basic type of chart common in many fields
- Often used to visualize a trend in data over intervals of time

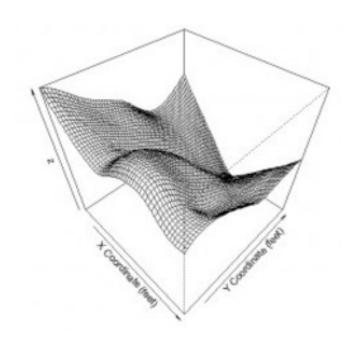


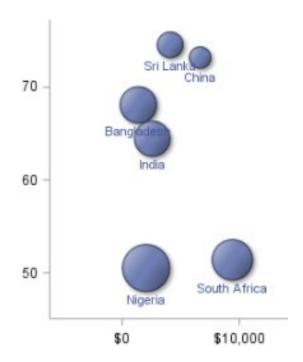
## Common Visualization Techniques to represent trivariate data

Trivariate data surface plot contour plot 3D representation bubble plot



- - -





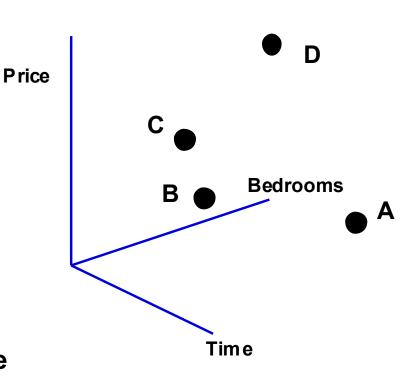
## Representing Trivariate data

 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)

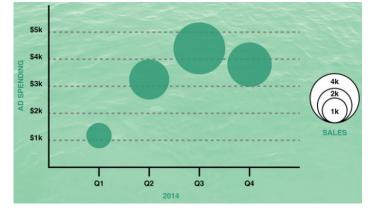


## Other Simple (and common) representations of trivariate data

• 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

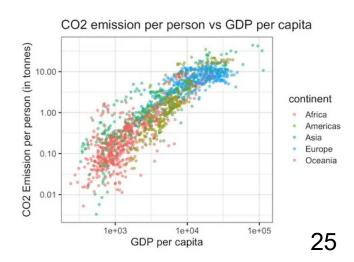
size (radius or area?)

 Mapping the variable to size must be done carefully. The interpretation of size may be ambiguous



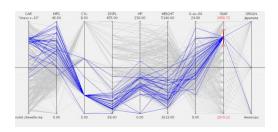
 Representing one more dimension through color

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

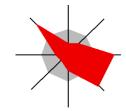


## Techniques for Multivariate (or Hypervariate) data

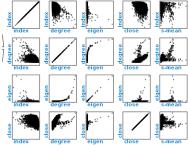
Coordinate plots ——parallel coordinate plots



star (radar/spider) plots



Scatterplot Matrix



Maps



Icons/glyphs





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

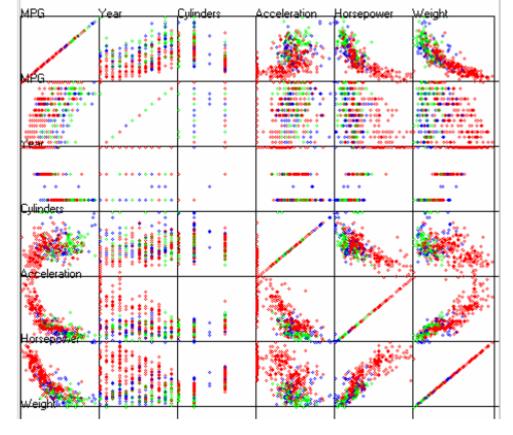
However, as the number of attributes increase, the number of different pairs

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:

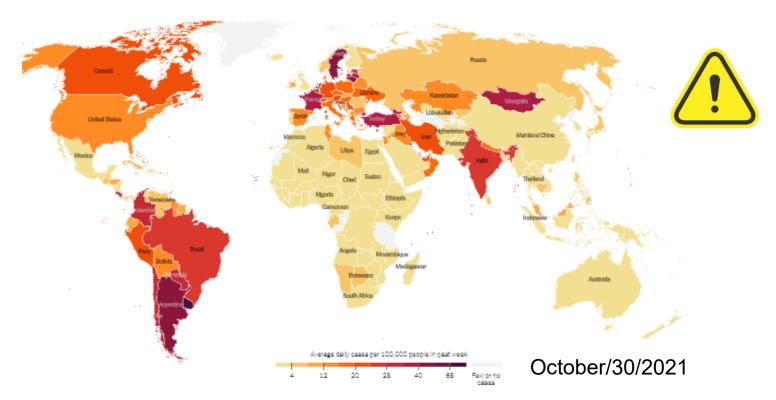
Dimensionality reduction!



Scatterplot matrix for 6 attributes of a car dataset

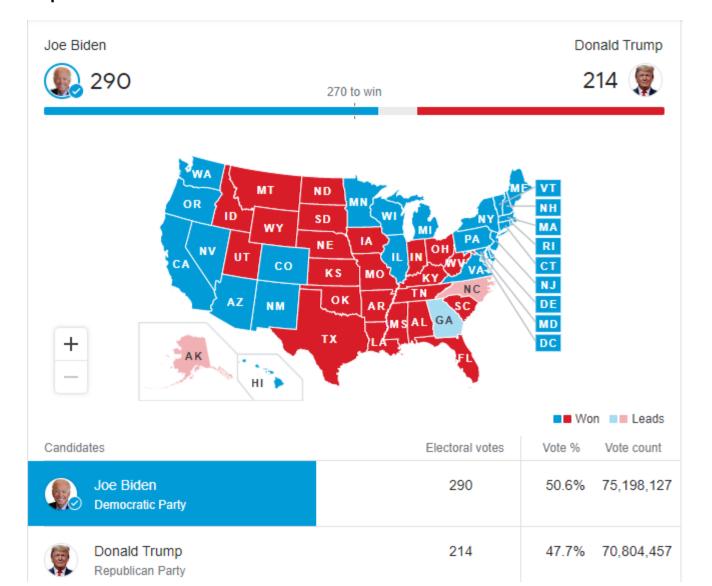
**Choropleth maps -** A standard approach to communicating aggregated data by geographical areas using color encoding of the geographic area

They require some care: what are the possible issues?



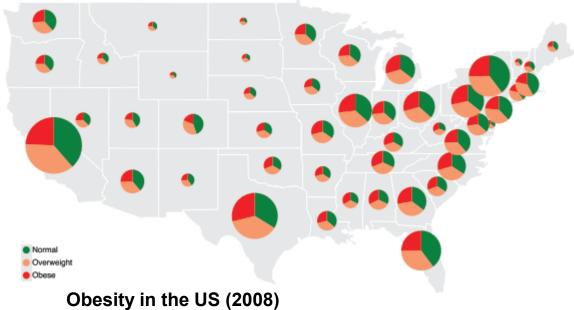
https://www.nytimes.com/interactive/2020/world/coronavirus-maps.html

# Visualizations of the US 2020 Election (choropleth + bar) the bar helps better understand the ratio of votes



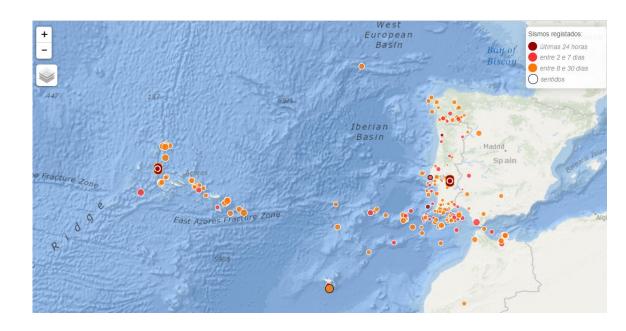
## Simple representations of attributes on a map

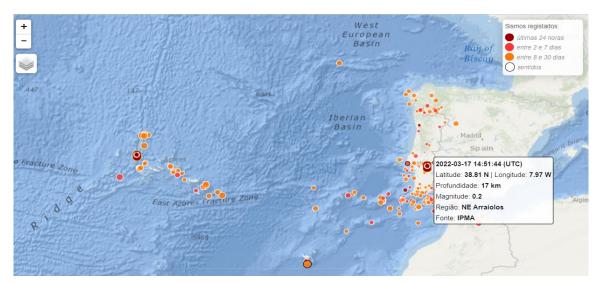
- Graduated Symbol Maps are an alternative to the choropleth map;
- Symbols are placed over an underlying map; may show more dimensions
- Avoid confounding geographic area with data values



(Heer et al., 2010)

Seismic activity:
Is something missing in this visualization?

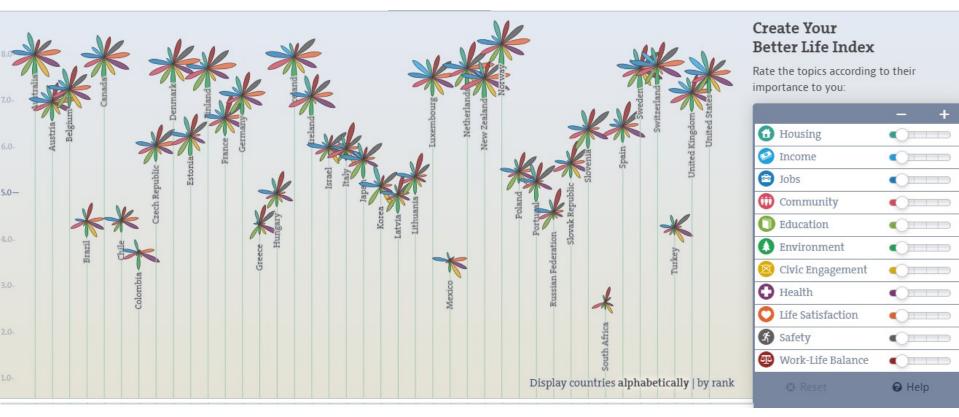




https://www.ipma.pt/pt/geofisica/sismicidade/index.jsp

## Glyph chart example

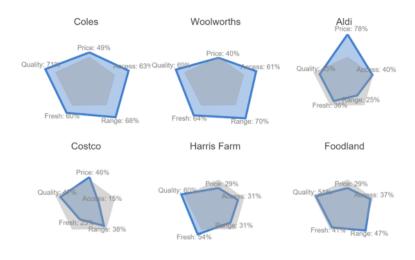
The physical properties of the shape represent different categorical variables sized according to the associated quantitative value and distinguished through color

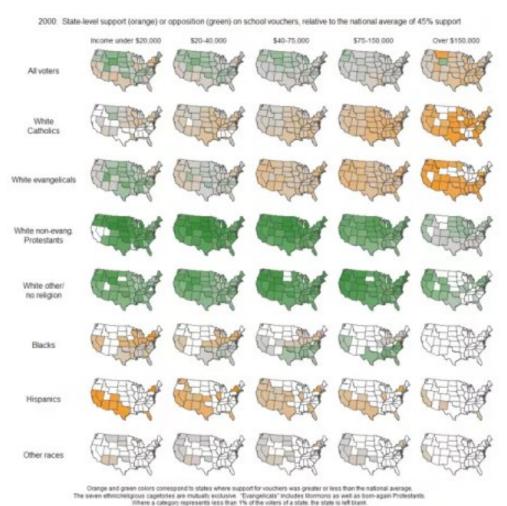


## Small multiples:

arrangement approach that facilitates efficient and effective comparisons

(Kirk, 2012)







nature of the data to represent ordinal categorical

number of attributes

bivariate

trivariate

multivariate

multivariate

univariate

bivariate

trivariate

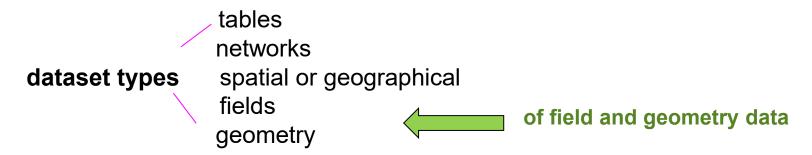
multivariate

number of attributes

Next: visualization techniques

organized according the n. of

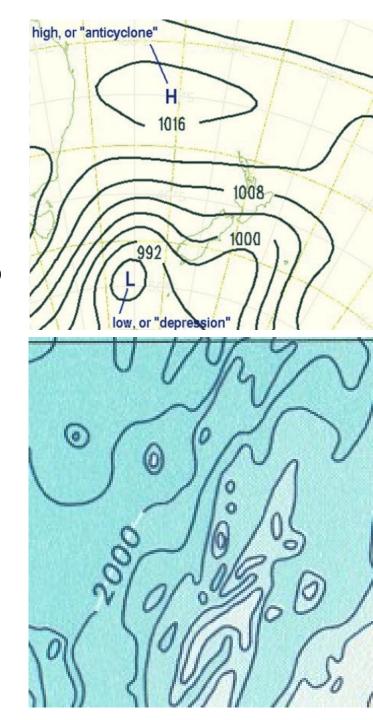
attributes



. . .

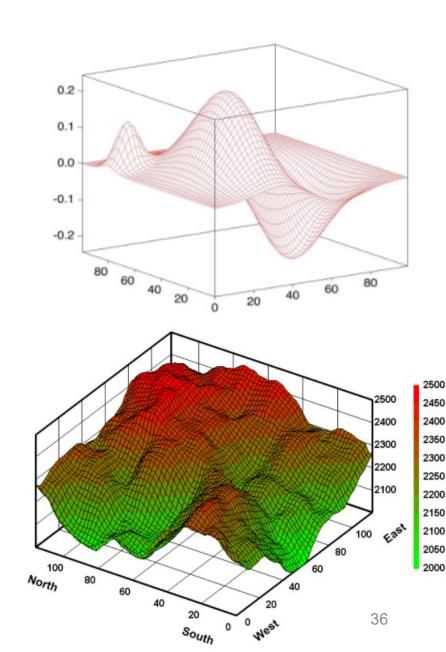
## Representations of a scalar in a 2D field

- 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)



- Surface plots
- May be combined with color

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



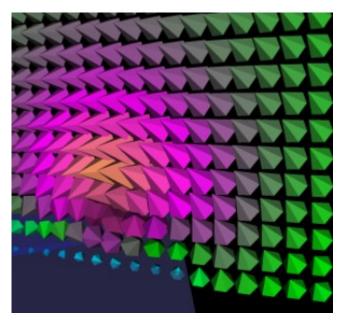
### Representations of scalars and vectors in a 3D field

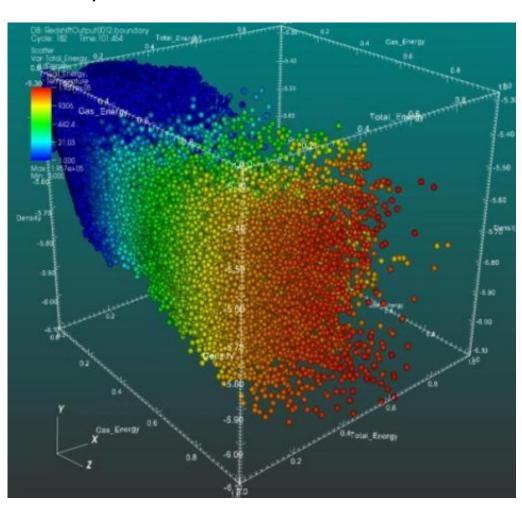
Four-dimensional data visualization: in 3D space a fourth scalar variable

is visualized using colored glyphs

Glyphs for Visualizing a 3D Vector Field







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

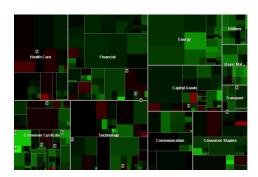
There are a lot more ...

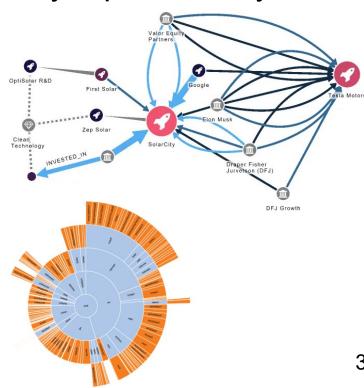
And we may visually want to visually represent beyond

value: relation

Networks

Hierarchical data





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- Heer, J., Bostock, M., & Ogievetsky, V. A tour through the visualization zoo.
   Communications of the ACM, vol. 8, n.1, 2010
   <a href="https://doi.org/10.1145/1743546.1743567">https://doi.org/10.1145/1743546.1743567</a>