MassMutual DSDP 2017:

INTRODUCTION TO DATA VISUALIZATION

June 8, 2017

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Statistical & Data Sciences
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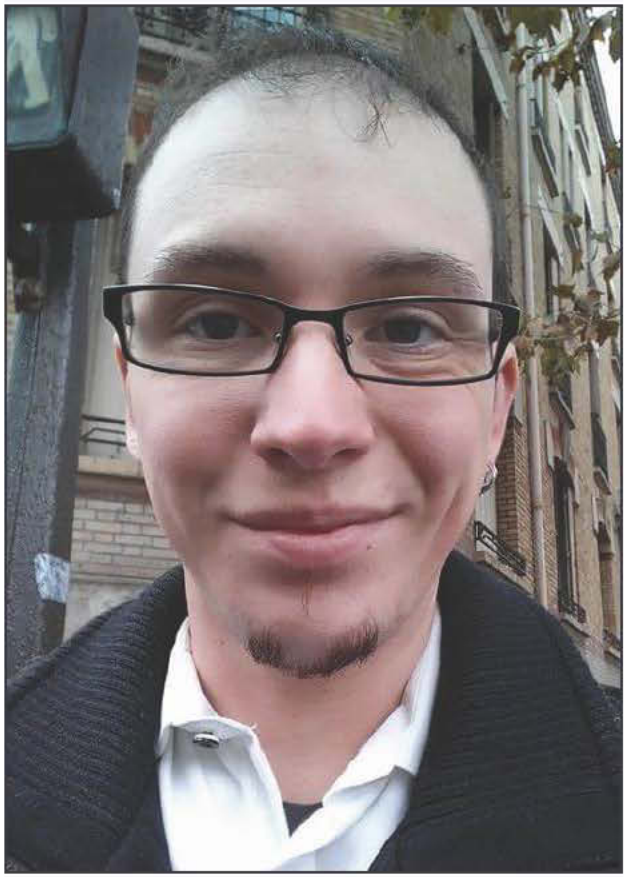
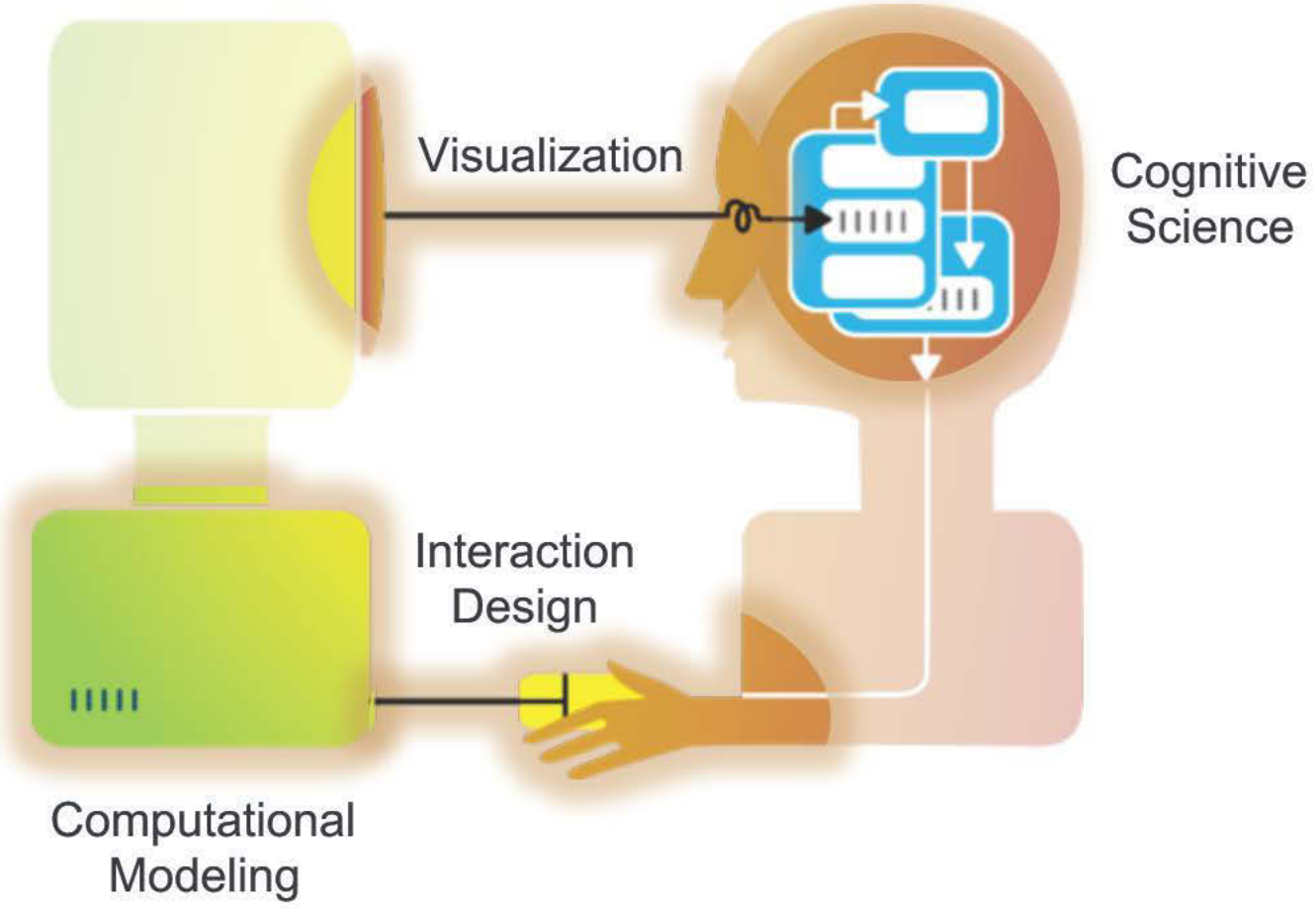
People



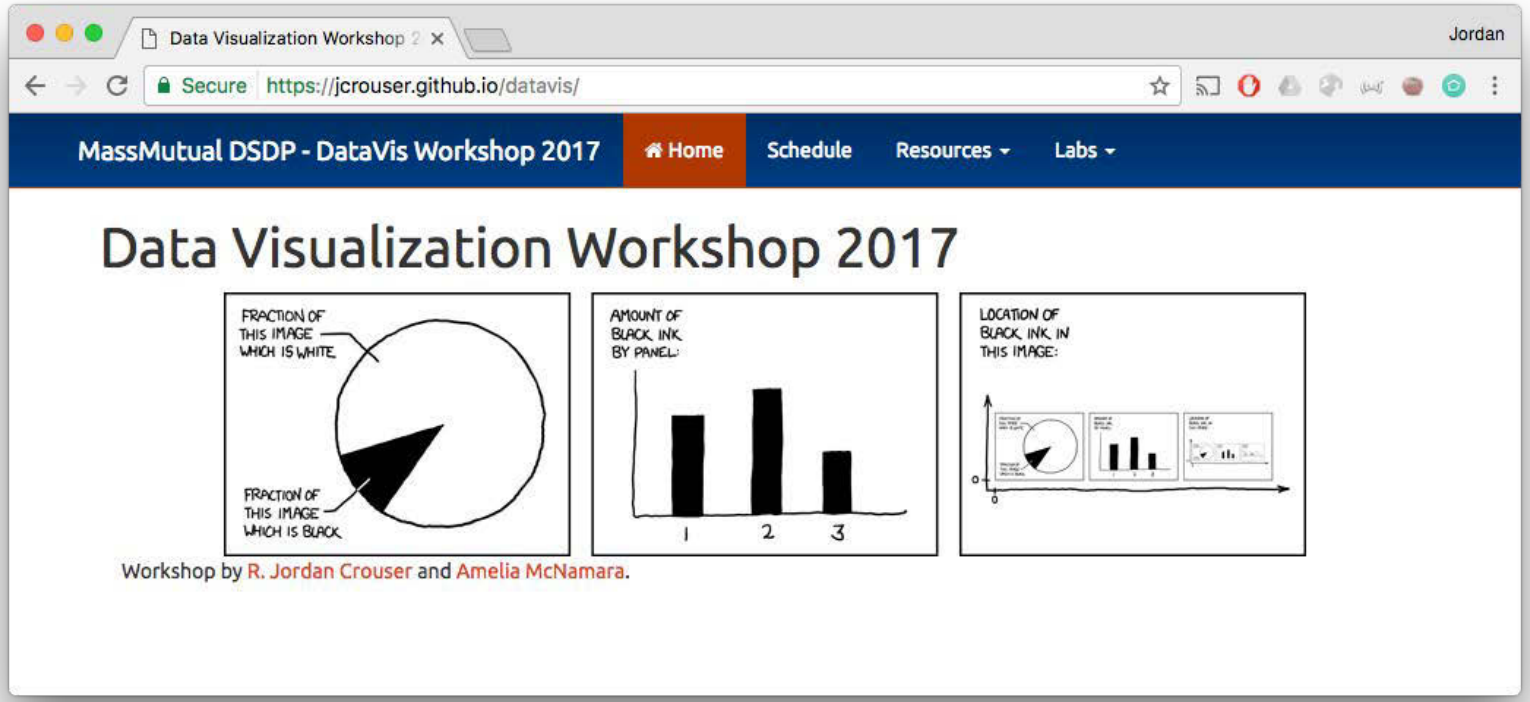
Jordan
(computer scientist)

Amelia
(statistician)

Our research (broadly)

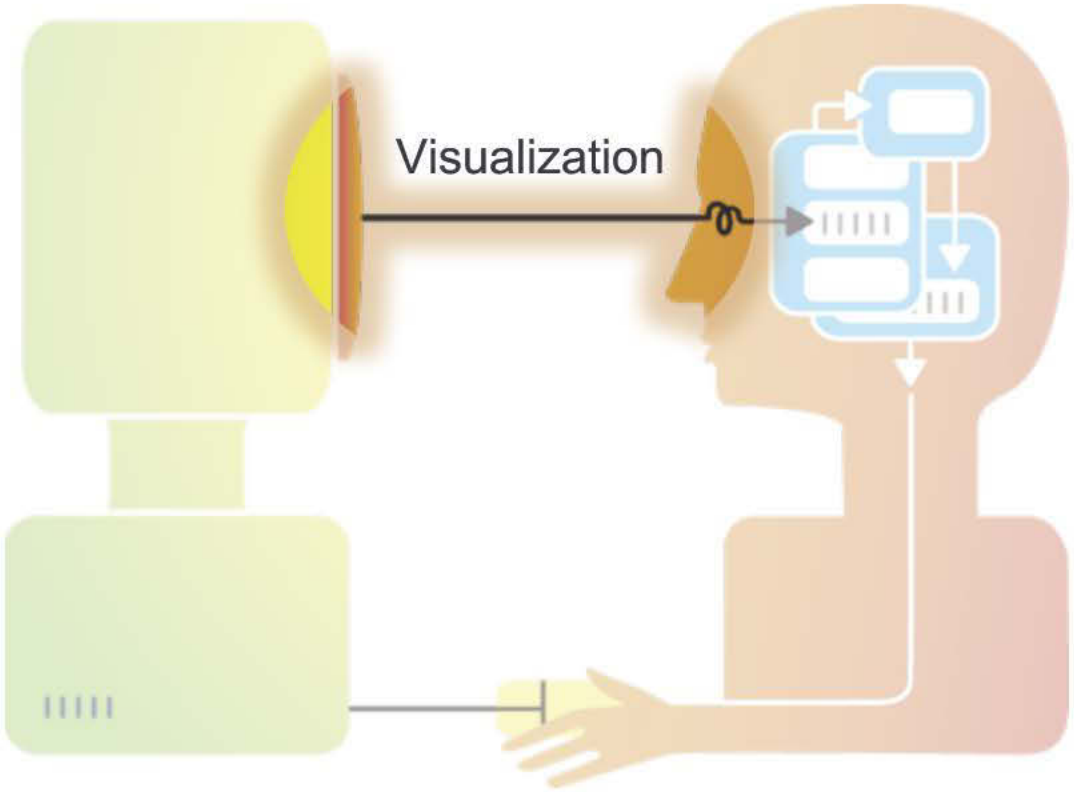


Housekeeping

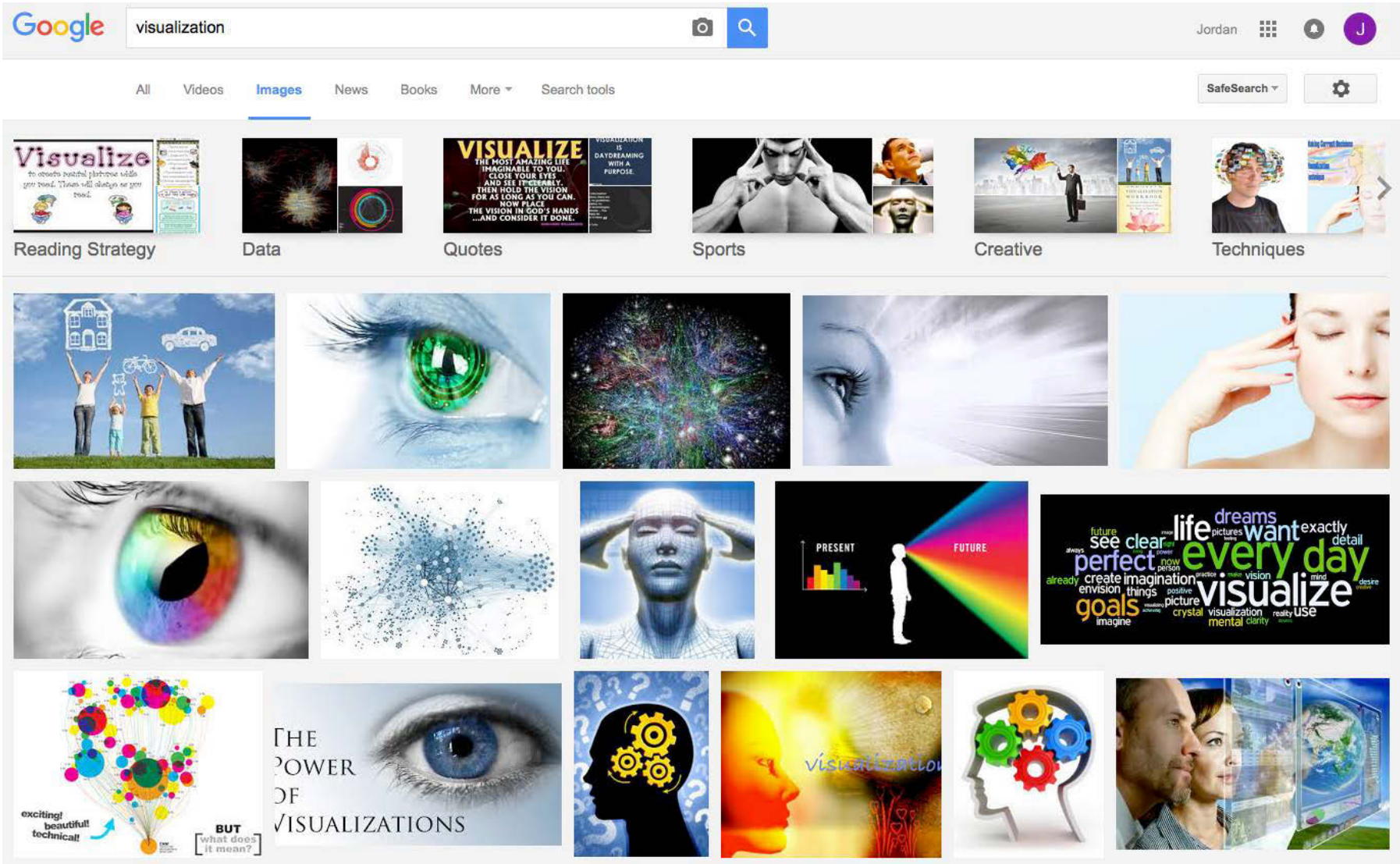


jcrouser.github.io/datavis

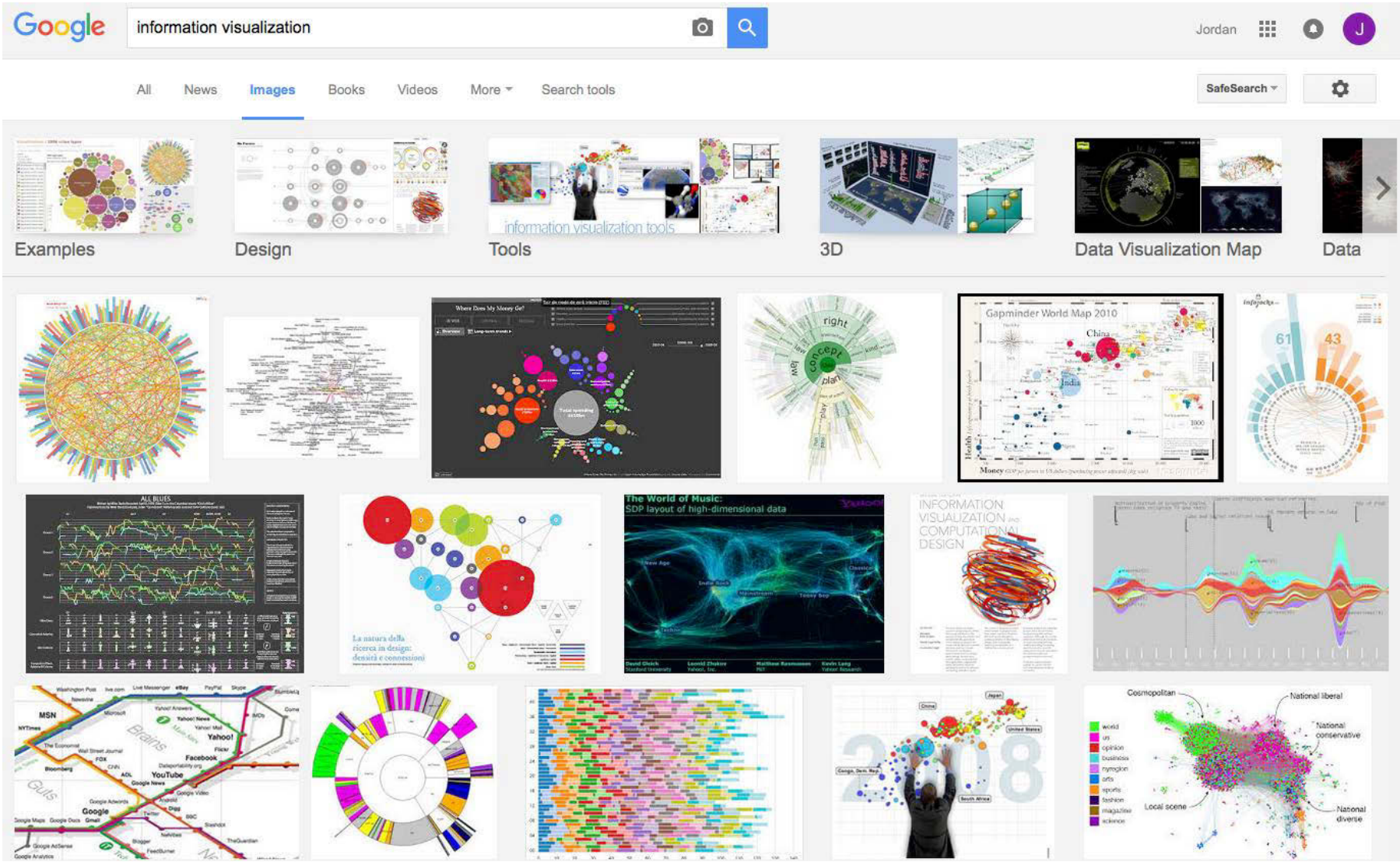
About this course



What is visualization?



What is visualization?



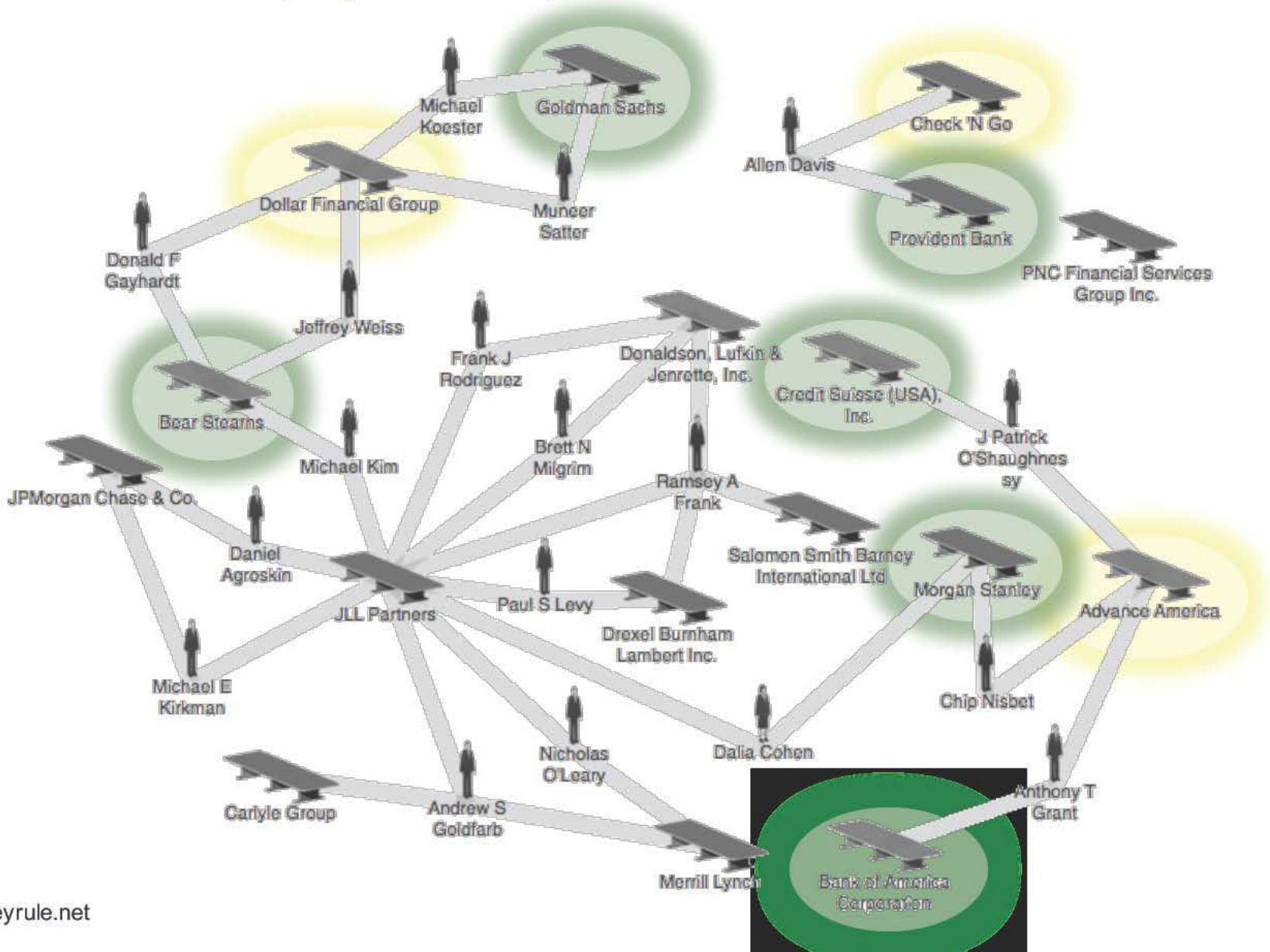
Perhaps a more helpful question:

What are some ways
a “visualization” can be **useful**?

Does it help you spot trends?

More info here: http://en.wikipedia.org/wiki/1854_Broad_Street_cholera_outbreak

Does it help you explore?



Credit: theyrule.net

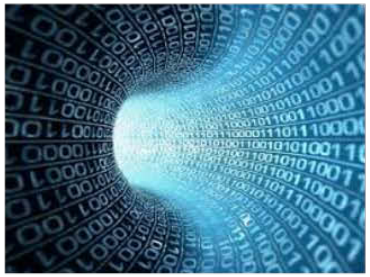
Does it tell a story?

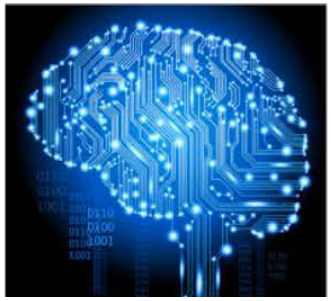
Hans Rosling's 200 Countries, 200 Years, 4 Minutes – The Joy of Stats – BBC Four
https://www.youtube.com/watch?v=jbkSRLYSojo

Visualization (def.)

Visual **representations** of data that reinforce human **cognition**

Wait… what is “data”?





Data: a definition

Data is a set of *variables* that capture various aspects of the world:

Tuition rates, enrollment numbers, public vs. private, etc.

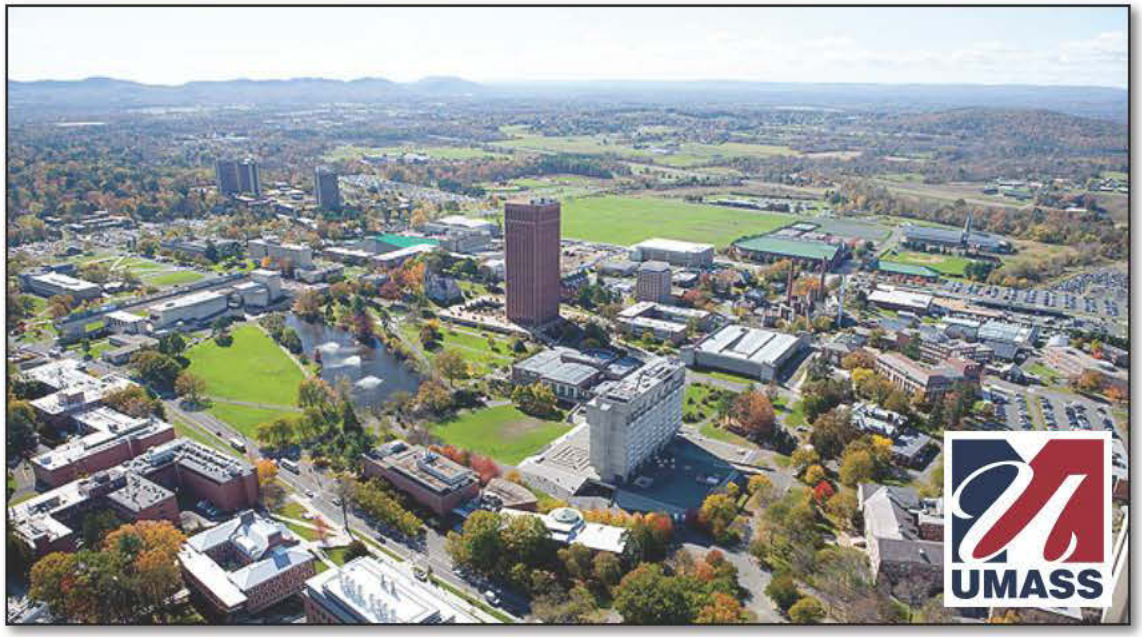
Data: a definition

A dataset also contains a set of *observations* (also called *records)* over these variables. For example:

tuition = $46,288, *enrollment* = 2,563, *private, etc.*

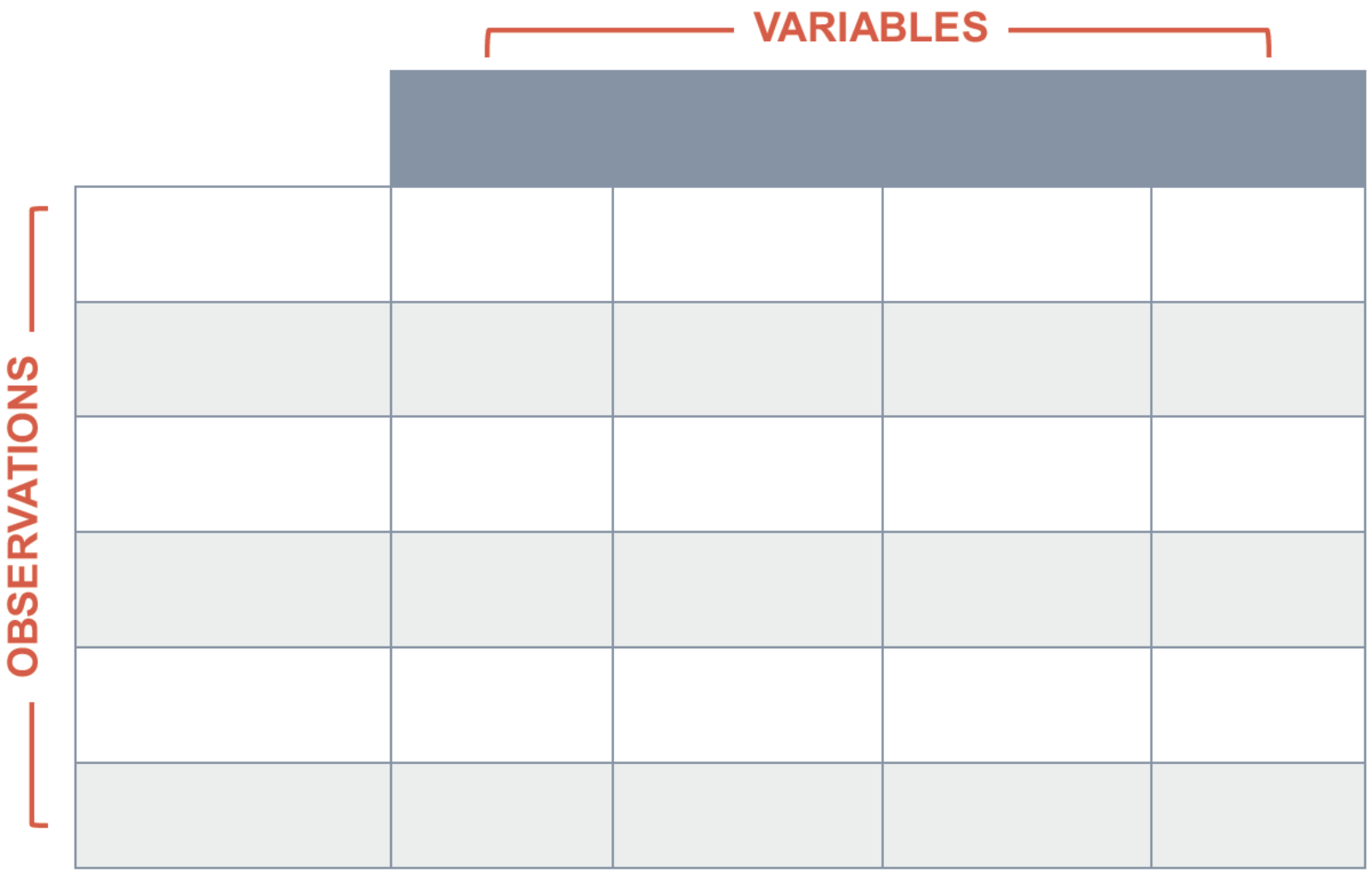
Data: a definition

A dataset also contains a set of *observations* (also called *records)* over these variables. For example:



tuition = $16,115, *enrollment* = 28,635, *public, etc.*

One way to think about this:



Another way to think about this

```
class school_obs:
      def __init__(tuition, enrollment,
                   pub_or_priv):
            self.tuition = tuition
            self.enrollment = enrollment
            self.pub_or_priv = pub_or_priv
```

VARIABLES

OBSERVATIONS

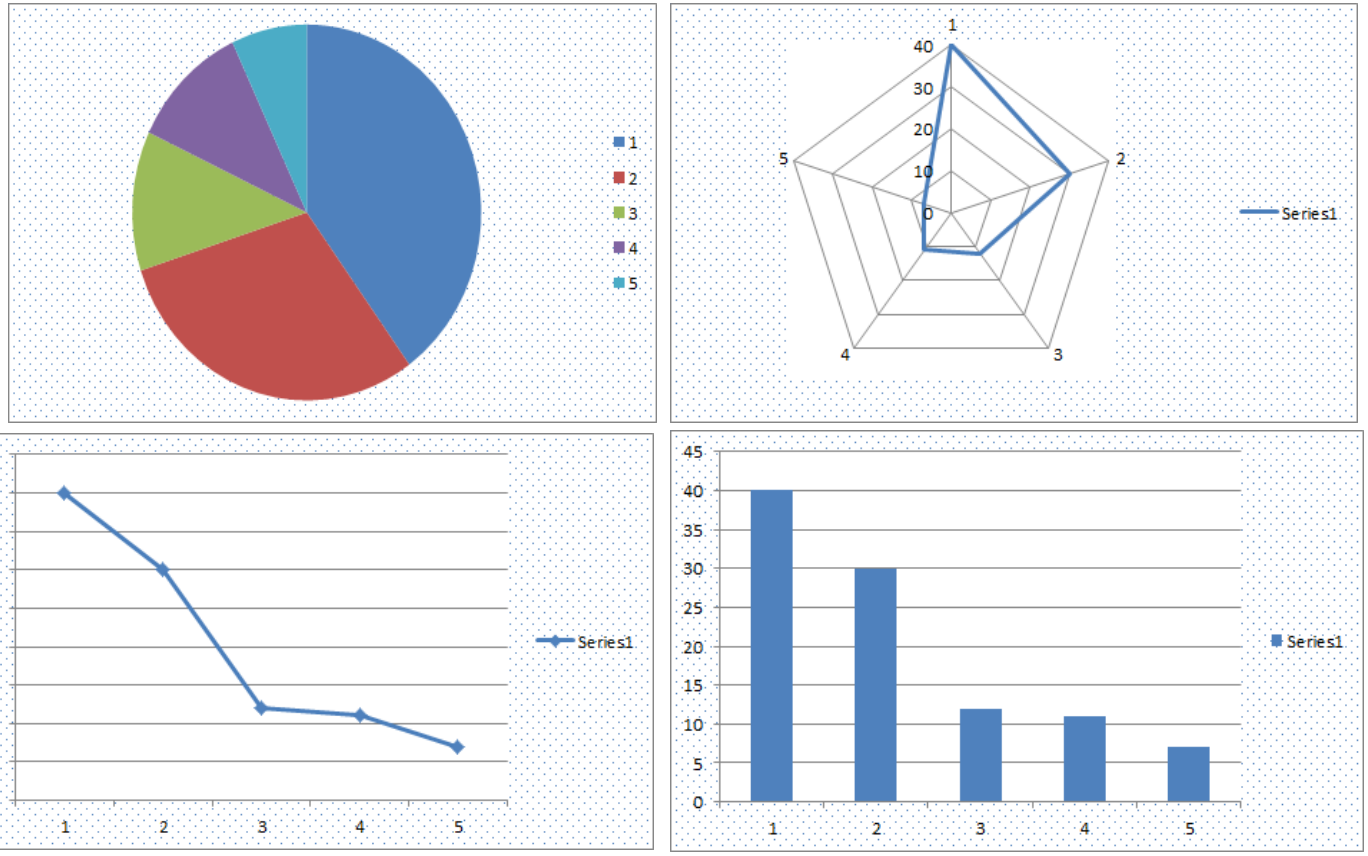
Why is this important?

- Data have dimensions
- Visualizations have dimensions, too
- To build visualizations, we need to **map** data dimensions to visual dimensions

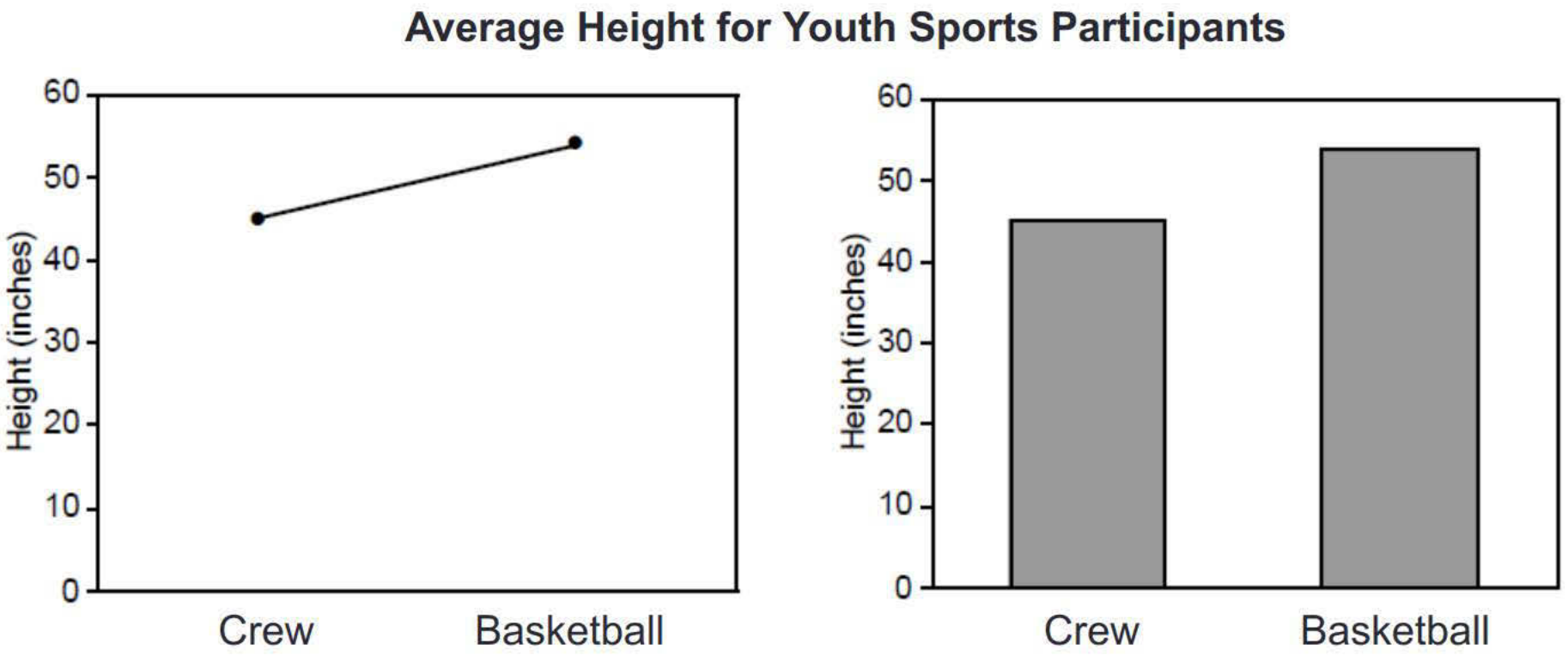
Key question for this course

Which **data dimension** should be mapped

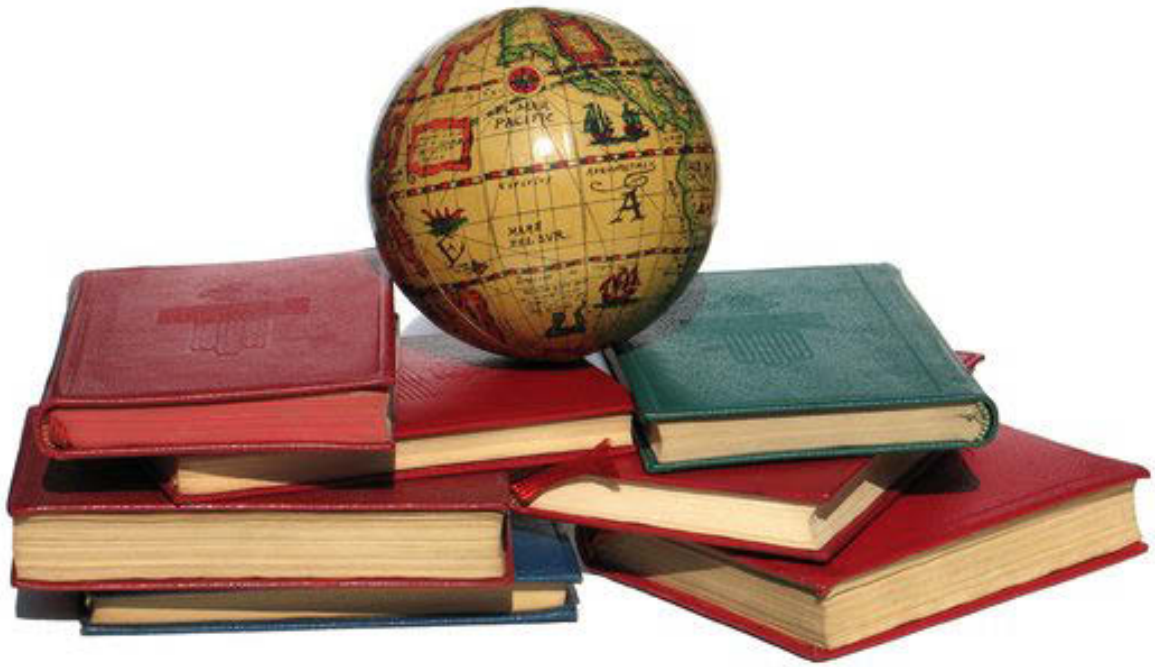
to which **visual dimension?**



Answer: it depends



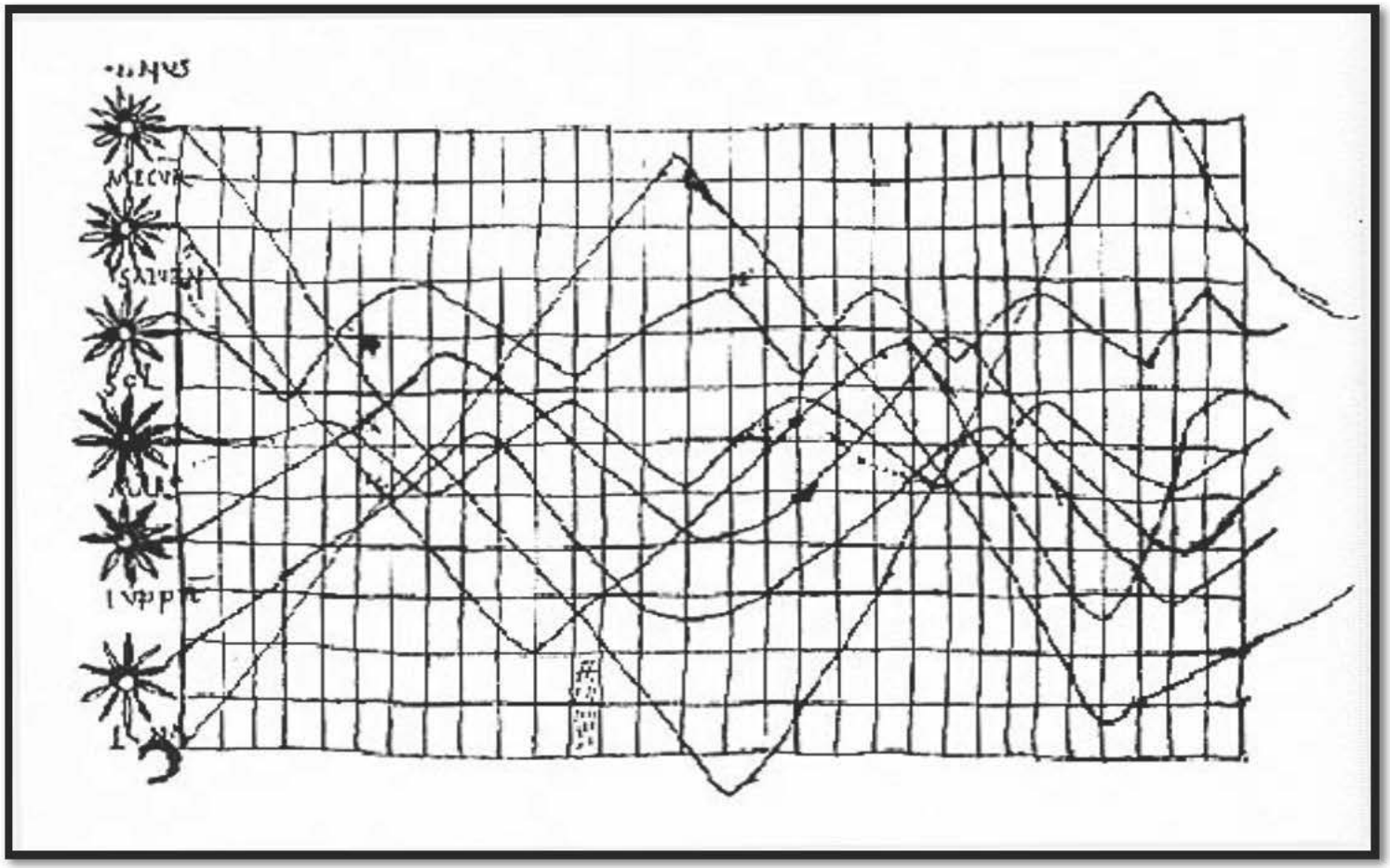
A quick history lesson…



(Incomplete) History of Visualization: 15,000BC

15,000 BC. Laxcaux, France

(Incomplete) History of Visualization: 900s

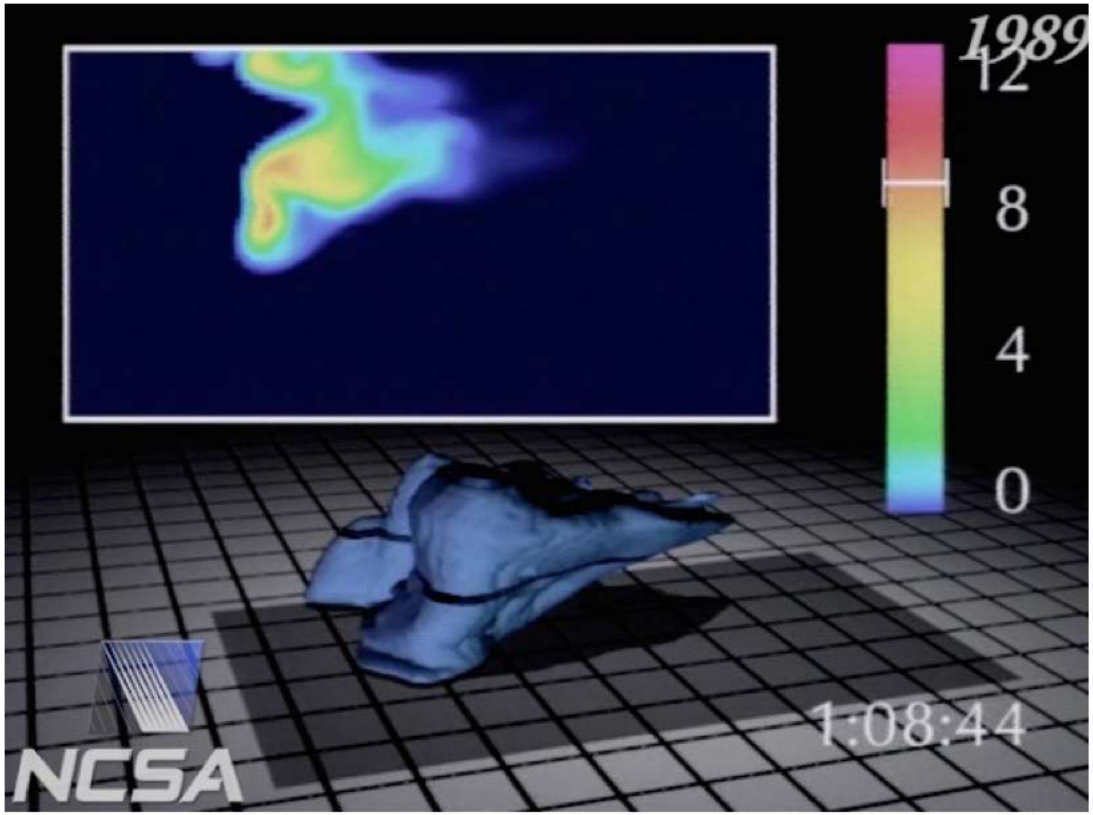


"De cursu per zodiacum", illustrator unknown

(Incomplete) History of Visualization: 1970s

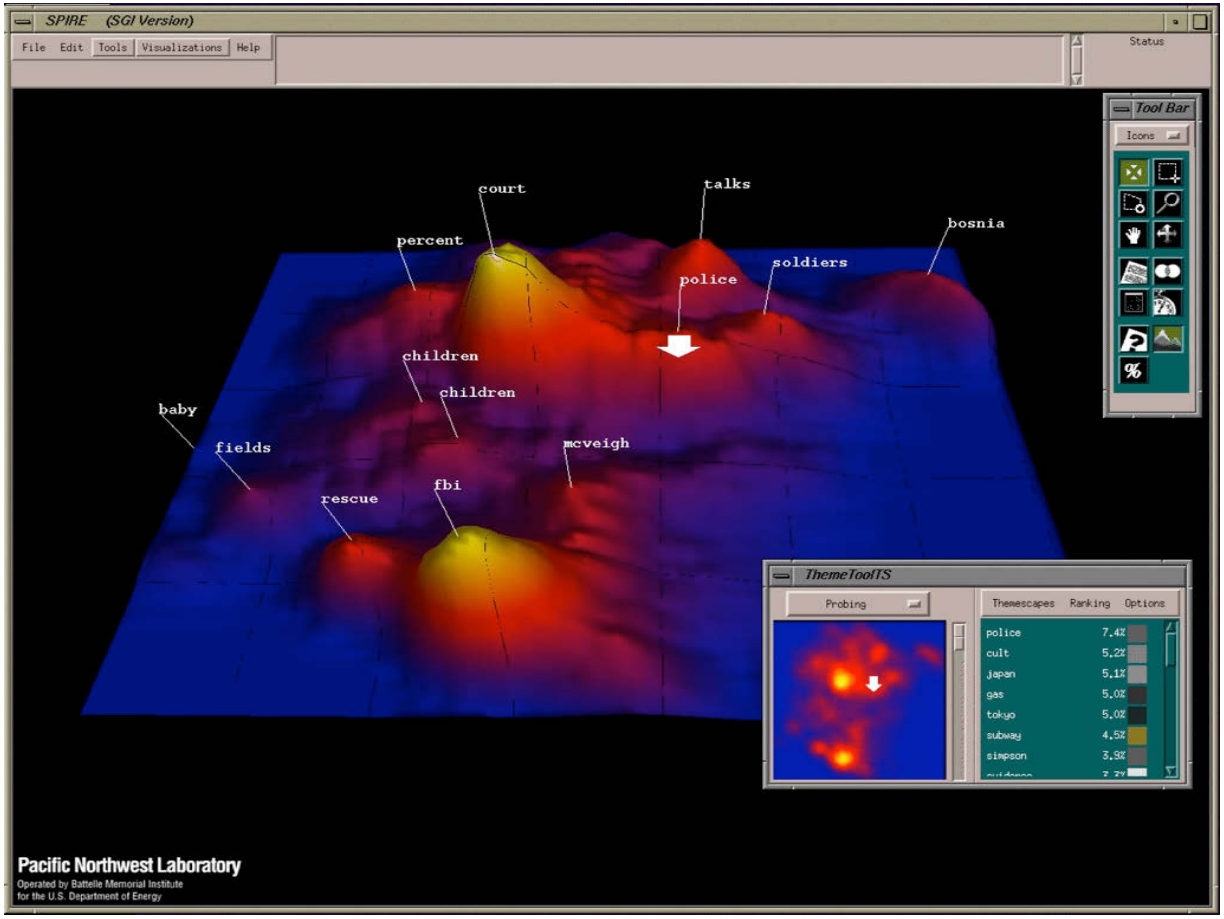
- CAD/CAM, building cars, planes, chips
- Starting to think about: 3D, animation, edu, medicine

(Incomplete) History of Visualization: 1980s



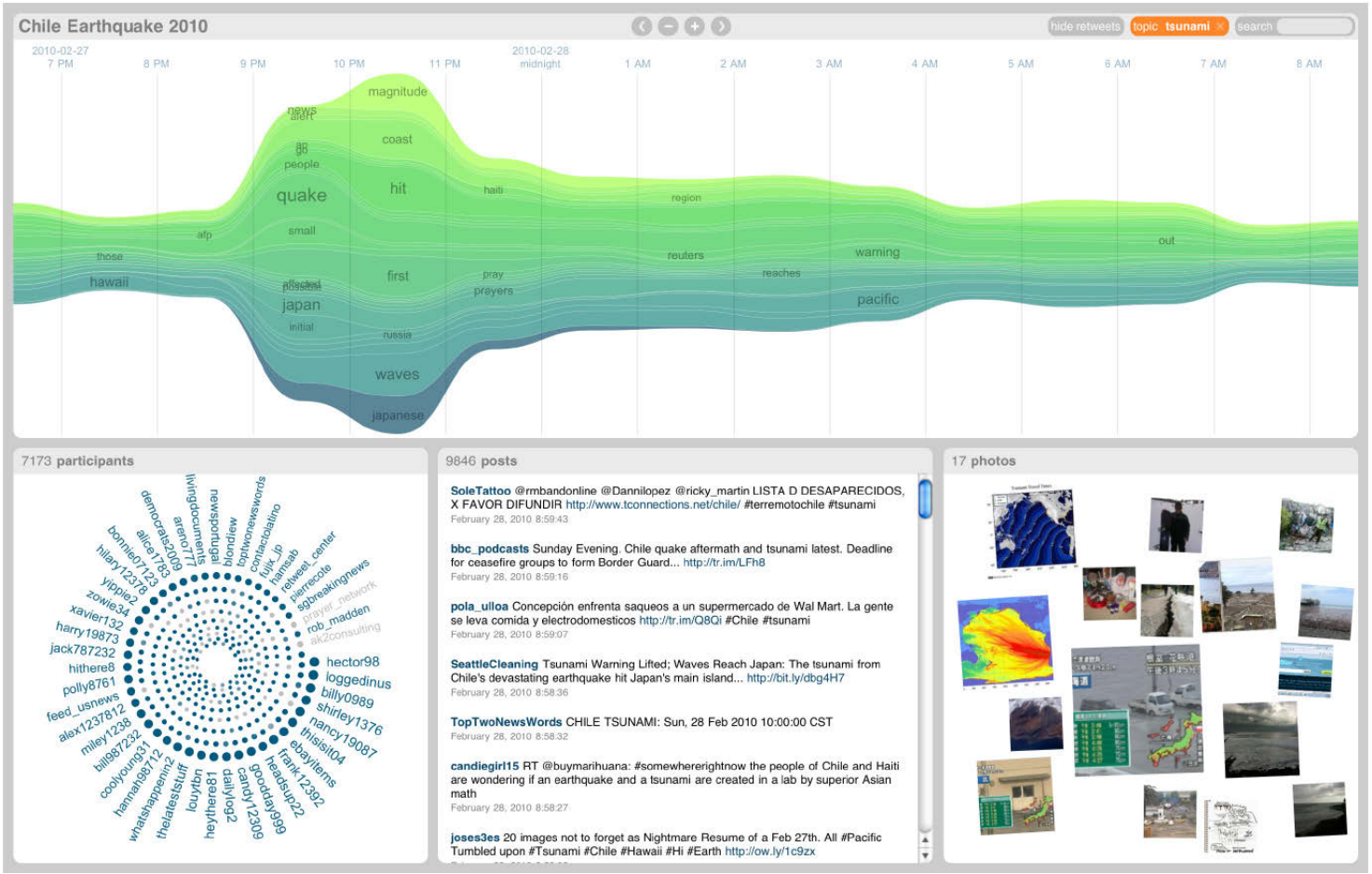
- Scientific visualization, physical phenomena
- Starting to think about: photorealism, entertainment

(Incomplete) History of Visualization: 1990s



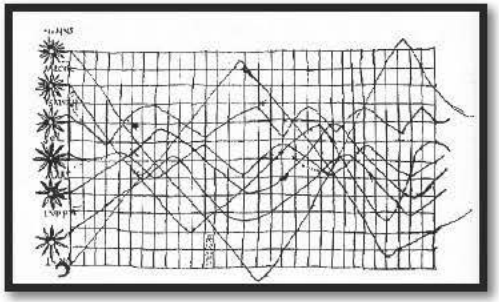
- Information visualization, storytelling
- Starting to think about: online spaces, interaction

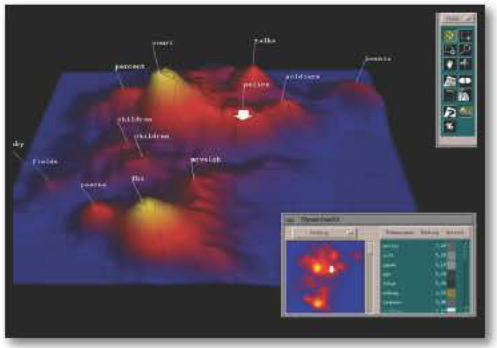
(Incomplete) History of Visualization: 2000s

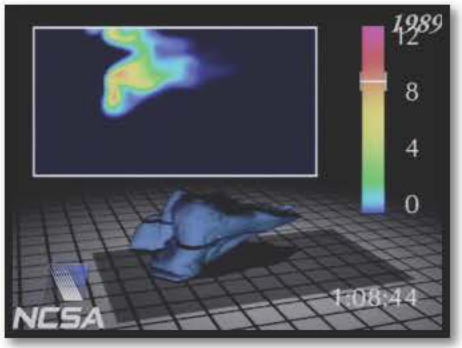


- Coordination across multiple views, interaction
- Starting to think about: sensemaking, provenance

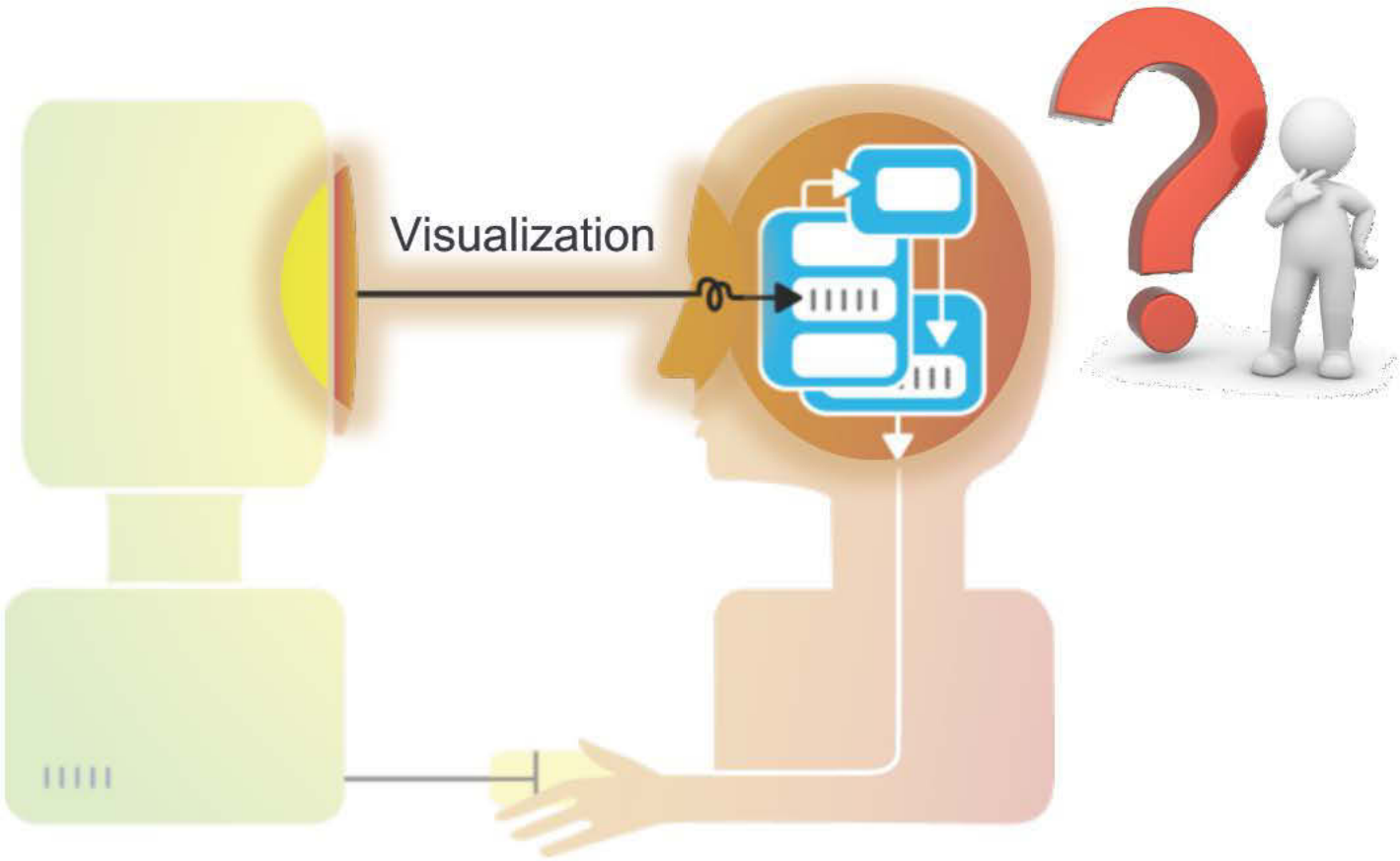
Discussion: what are they all trying to do?



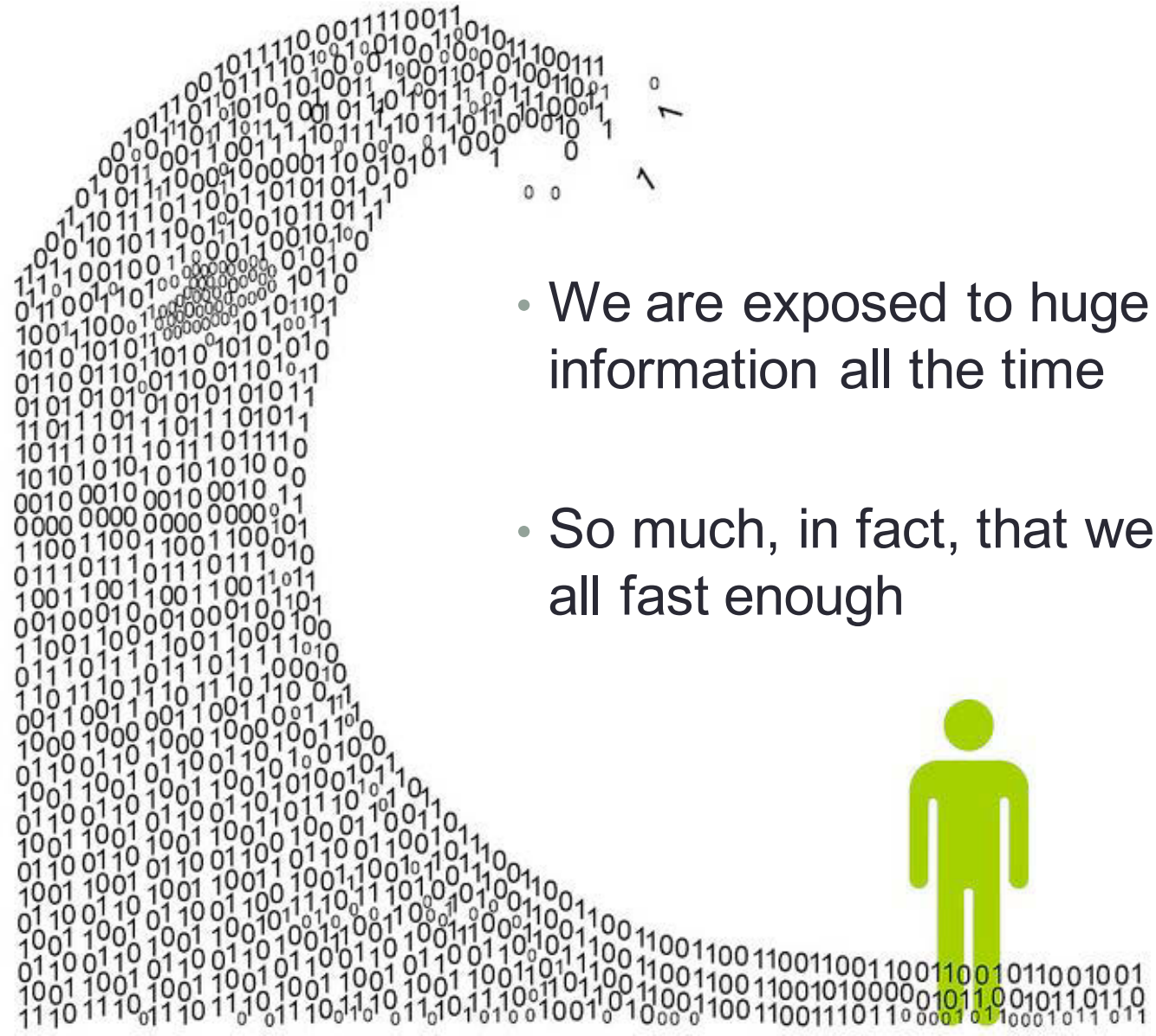




Visualization helps shape *mental models*



Information overload

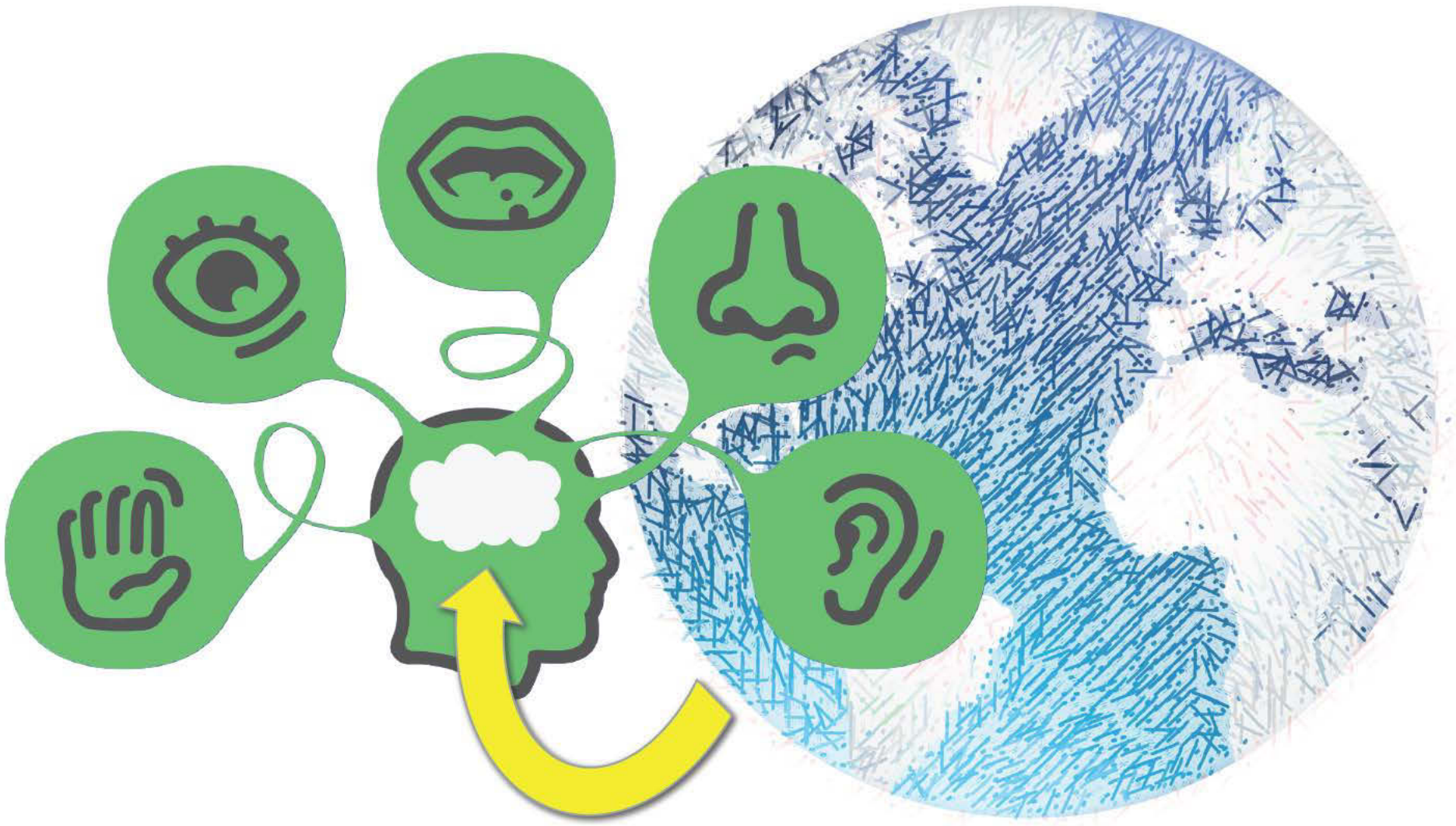


- We are exposed to huge amounts of information all the time
- So much, in fact, that we can't process it all fast enough

Mental models

To cope, we construct **mental models:** abstracted, simplified versions of the world that are more manageable

Mental Models: a Sketch

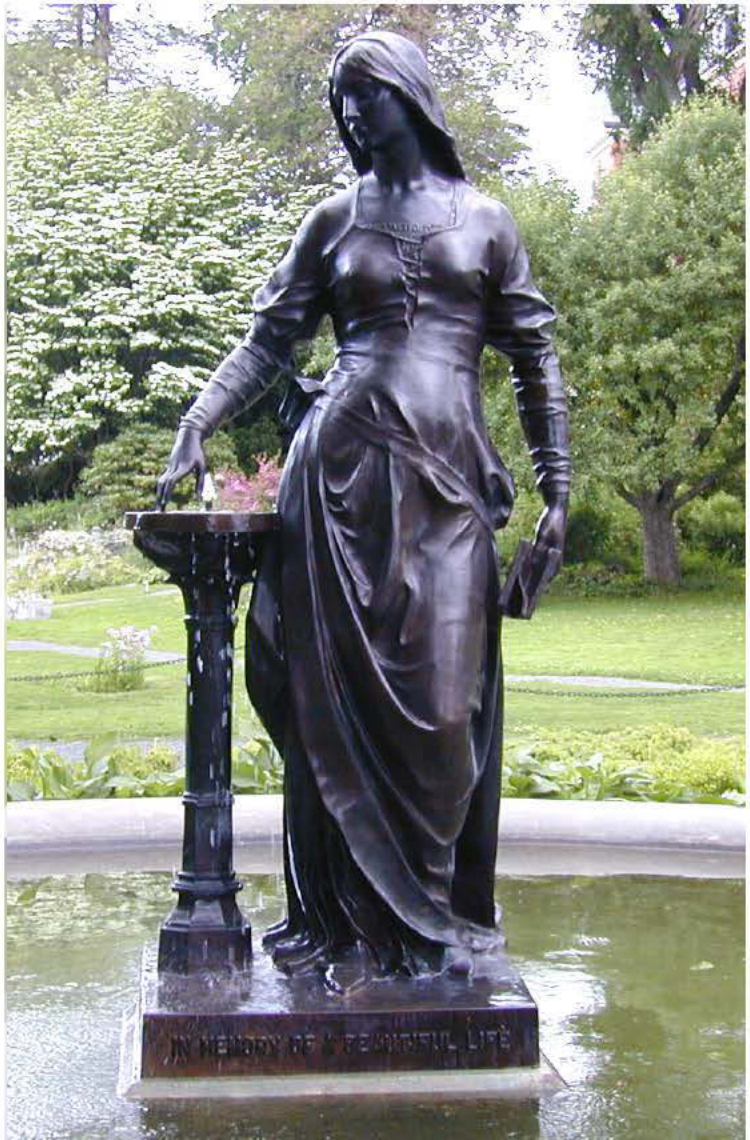


1. We tend to see what we expect to see

2. Mental models form quickly, & update slowly

3. New information gets incorporated into the existing model

4. Initial exposure interferes with accurate perception



Blur size

128px
64px
32px
16px
8px
None

The good, the bad, and the ugly…

The good:

- Well-tuned mental models let us process information quickly
- Frees up more processing power to synthesize information

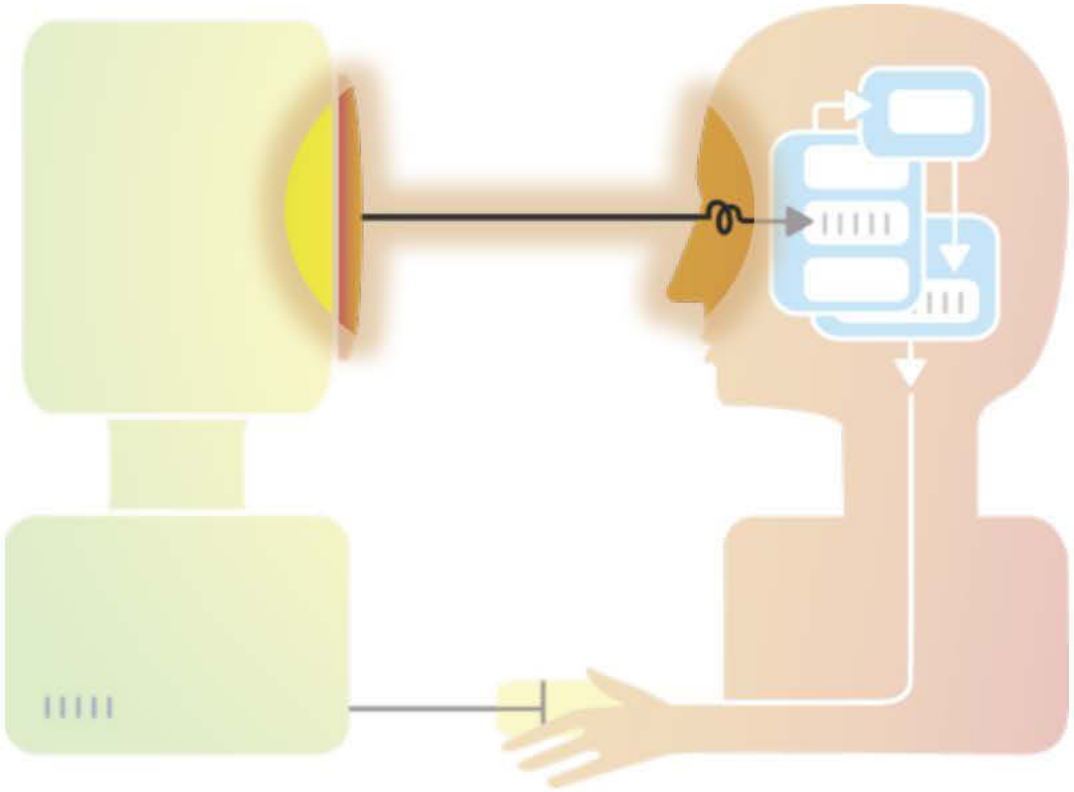
The bad:

- People (esp. experts) tend not to notice information that contradicts their mental model
- A “fresh pair of eyes” can be beneficial

The ugly:

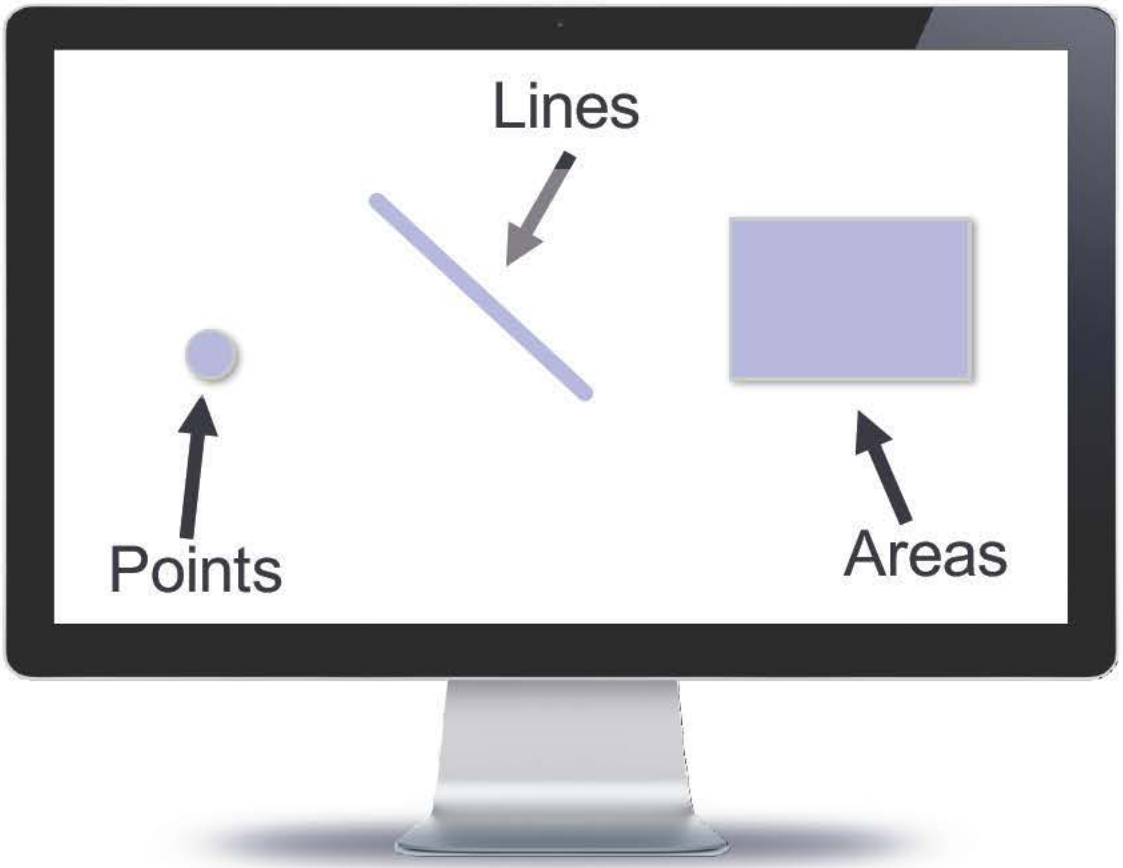
- Mental models are unavoidable: everyone has them, and they’re all different
- **Key:** be aware of how mental models form, how they shape perception, and how to support (or challenge) them

So what do we have to work with?



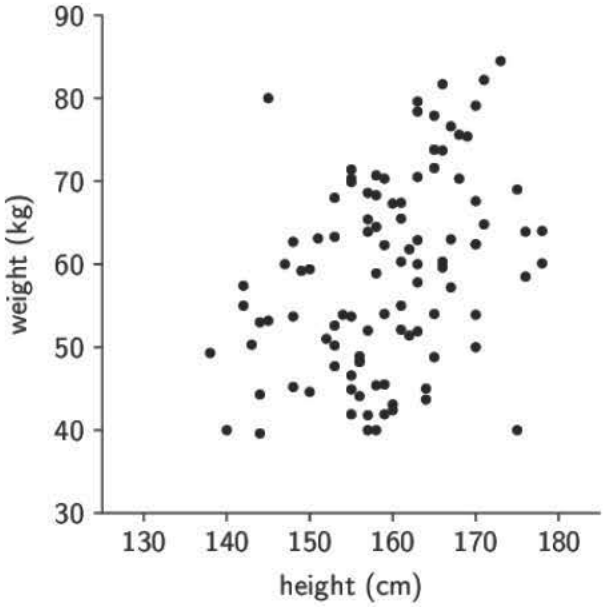
Graphical primitives

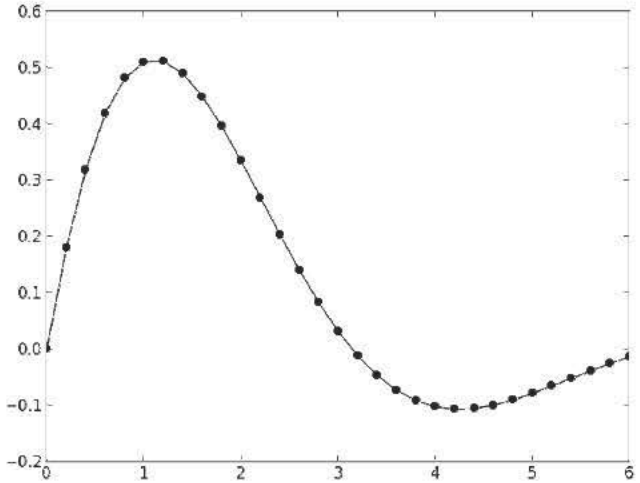
The images we draw are composed of marks: like ink



Visual dimension: position

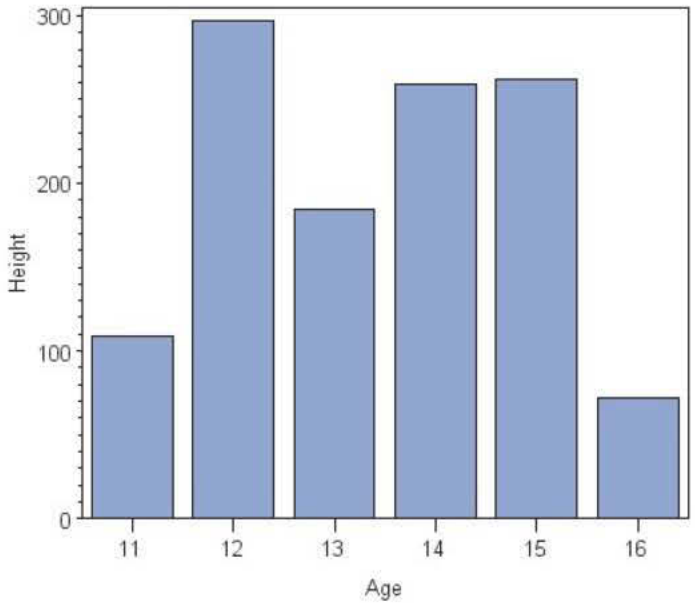
- Encode information using **where** the mark is drawn
- Some examples:

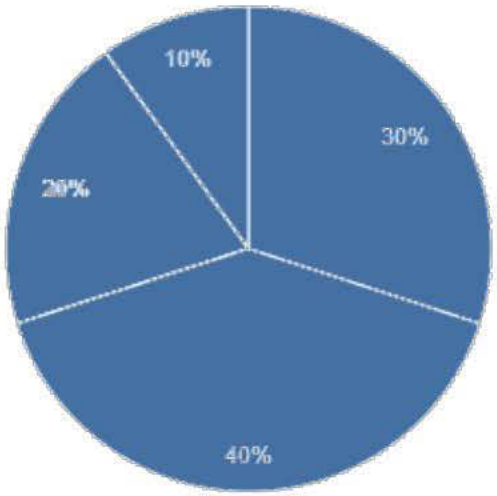


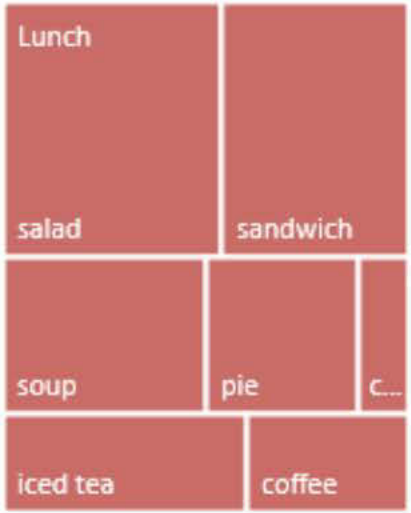


Visual dimension: size

- Encode information using **how big** the mark is drawn
- Examples:

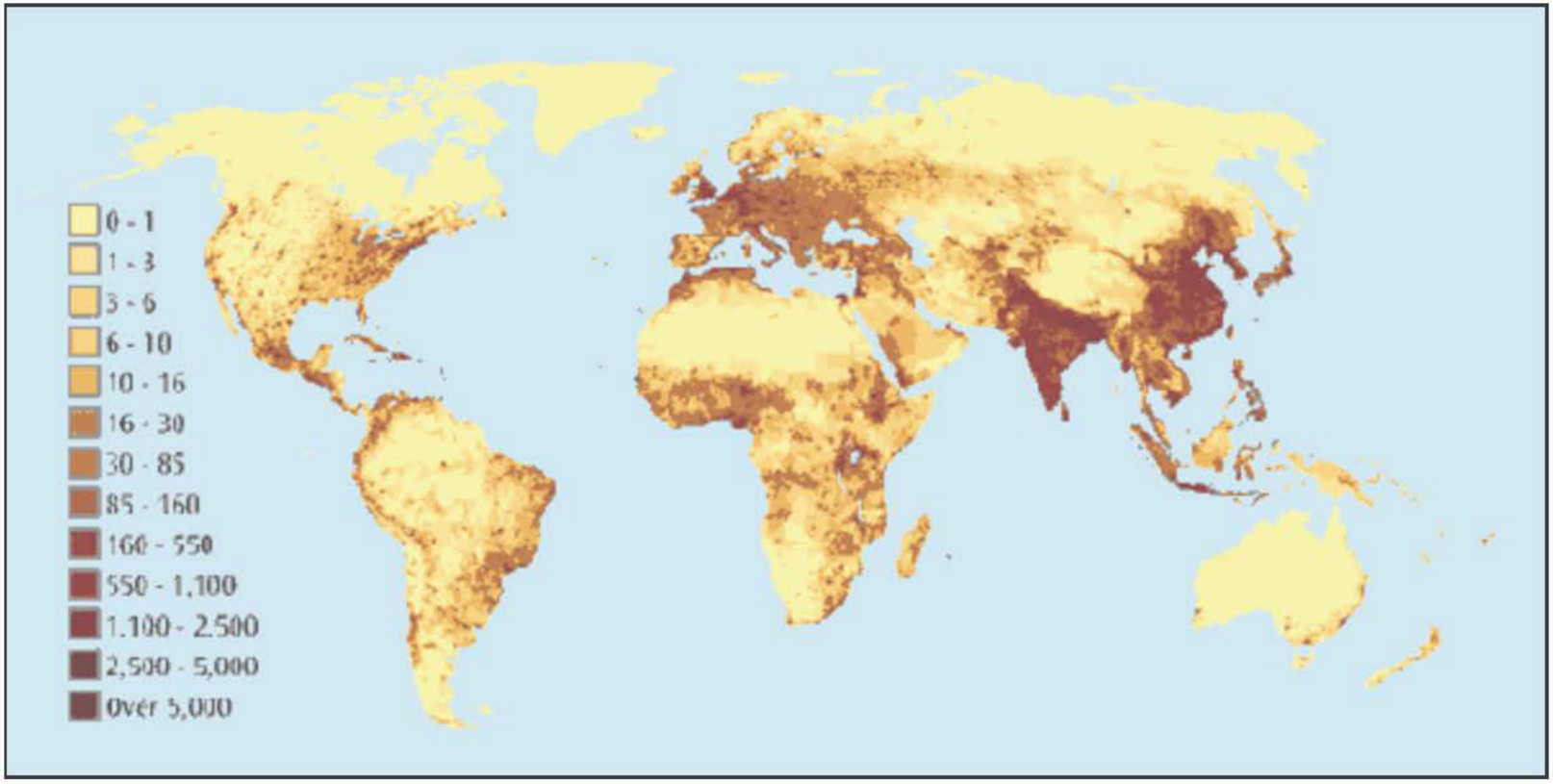






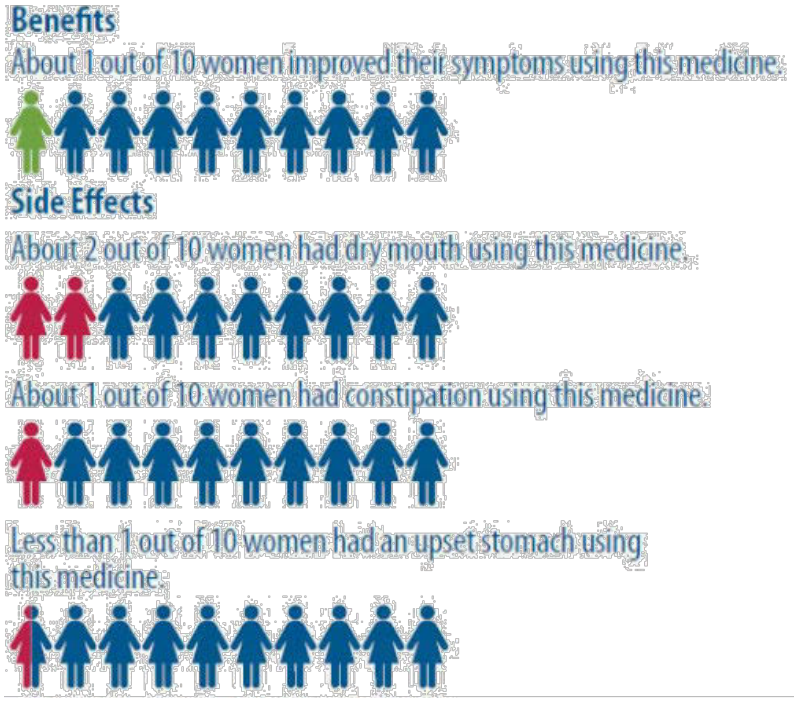
Visual dimension: value

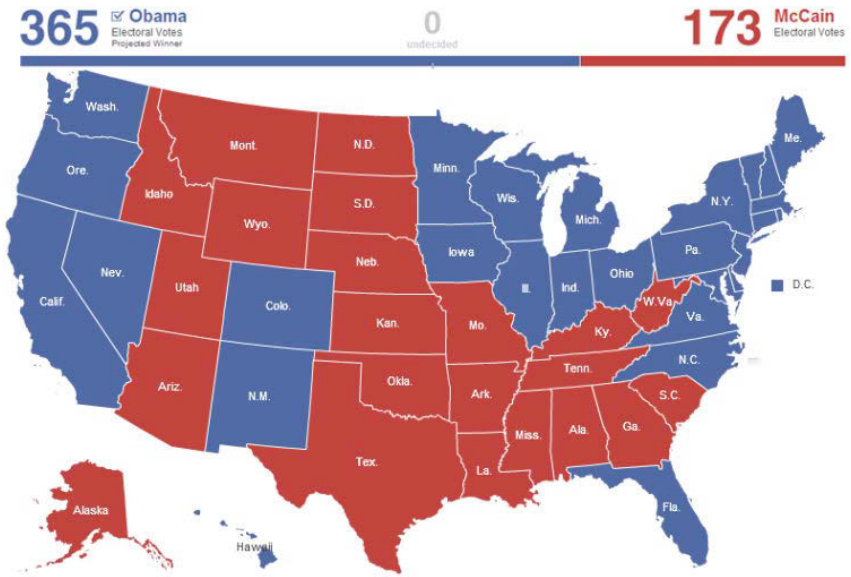
- Encode information using **how dark** the mark is drawn
- Example:



Visual dimension: color

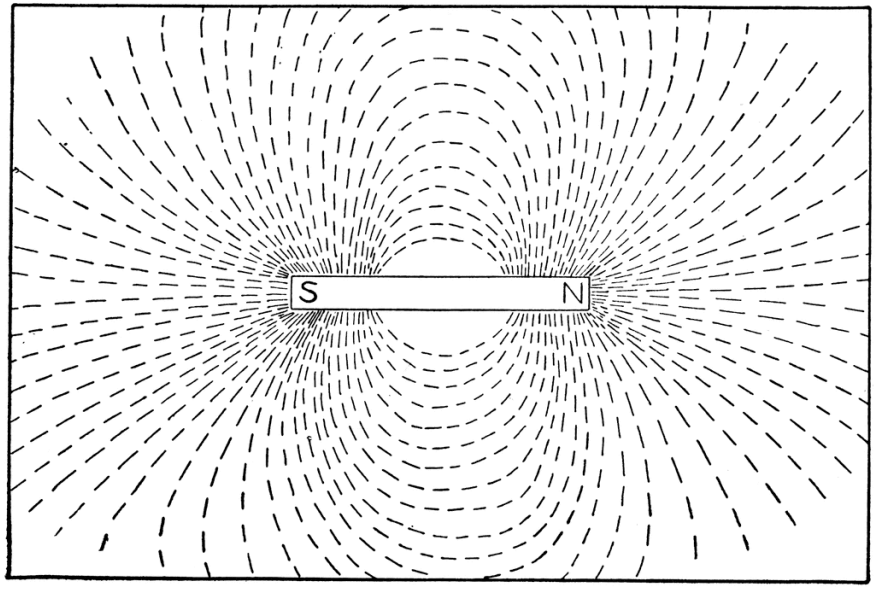
- Encode information using the **hue** of the mark
- Examples:





Visual dimension: orientation

- Encode information using how the mark is **rotated**
- Examples:



Visual dimension: shape

- Encode information using how the mark is **shaped**
- Examples:

Discussion

What makes a **good** encoding?

Principle 1: expressiveness

- Encodes **all** the facts
- Example:

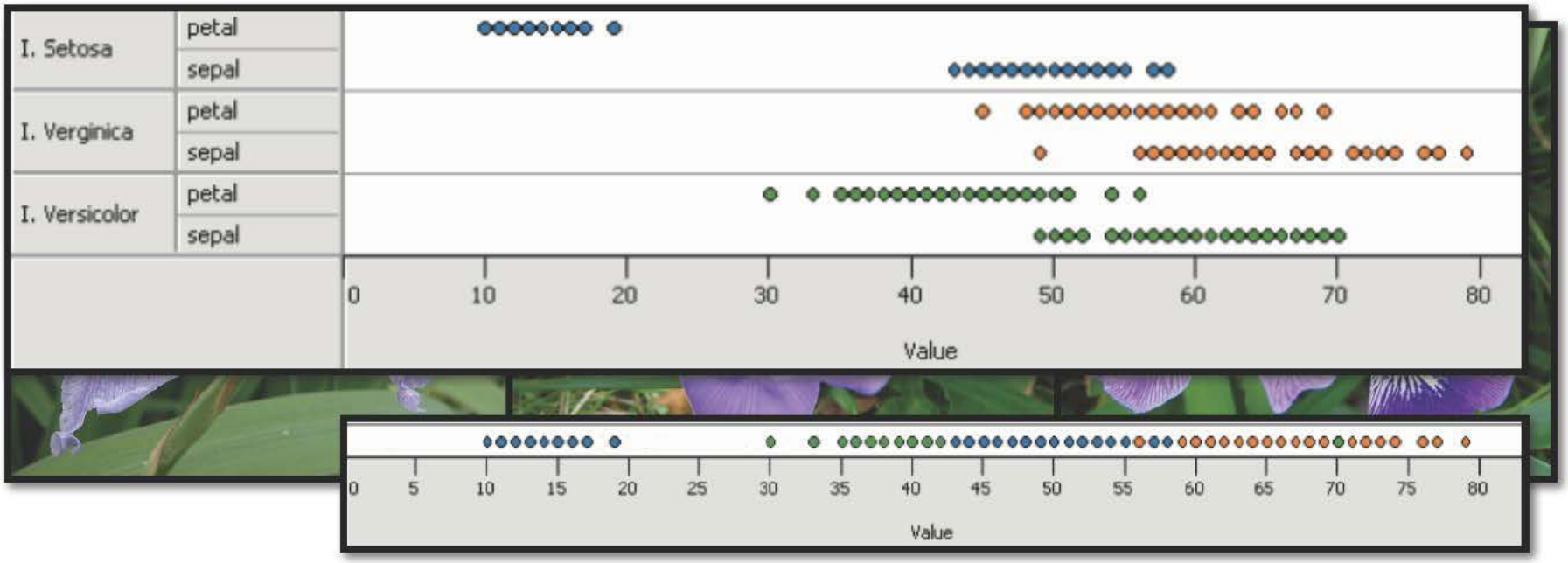
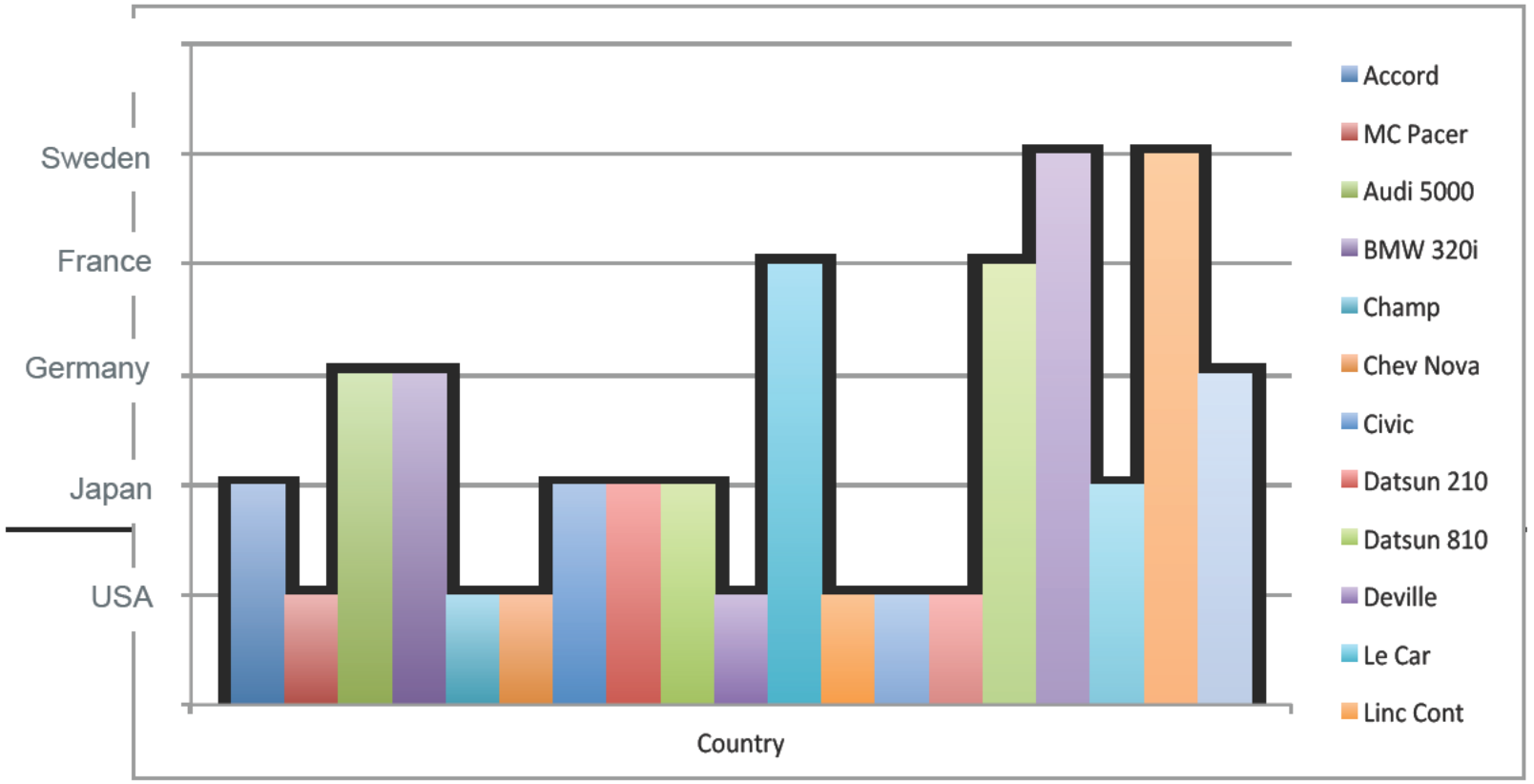


Fig. Courtesy of M Krzywinski

Principle 1: expressiveness

- Encodes **only** the facts
- Example:

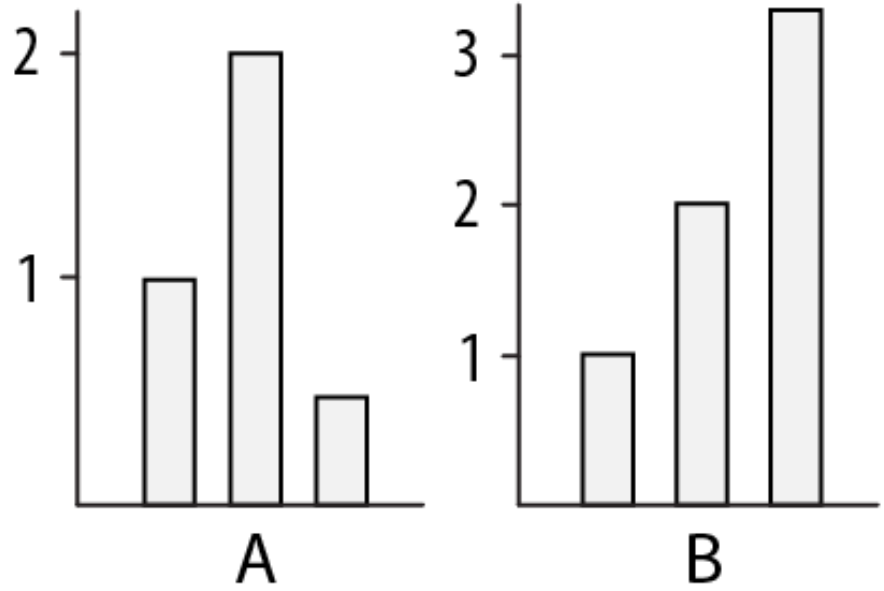


Adapted from Mackinlay J (1986) Automating the design of graphical presentations of relational information.

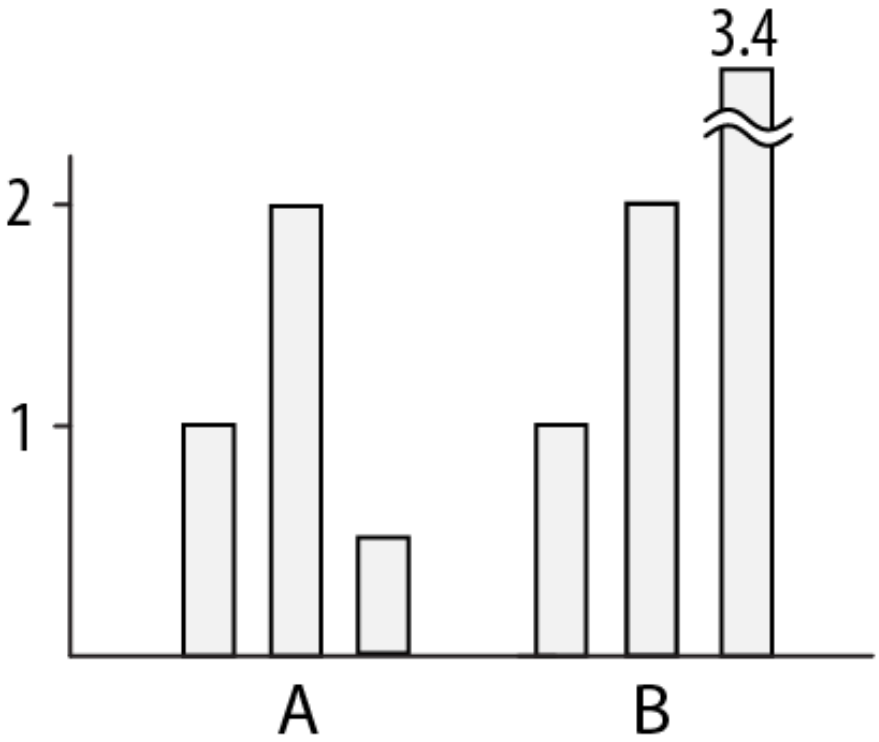
Principle 2: consistency

- Use **consistent axes** when comparing charts

misleading



improved

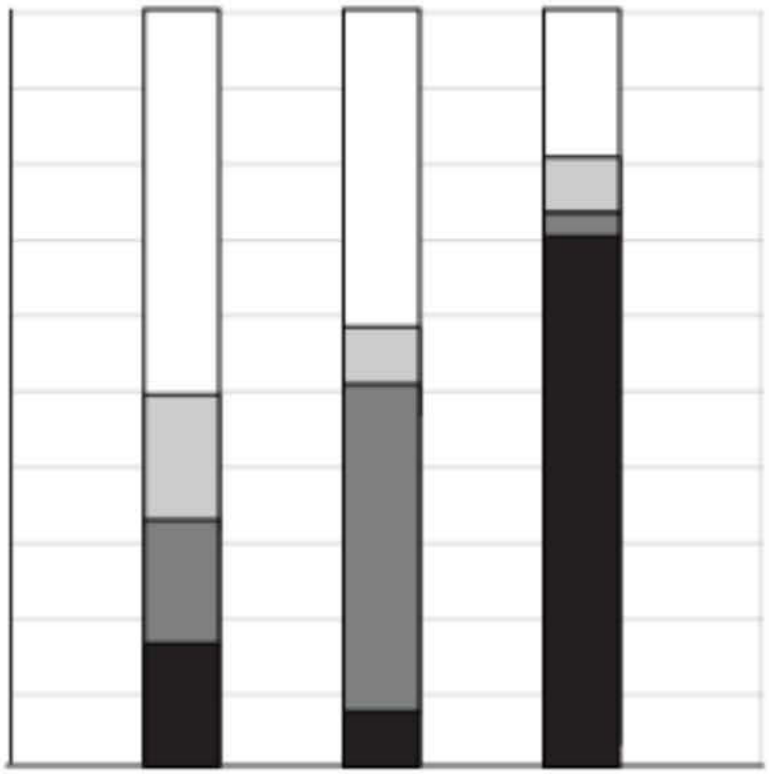


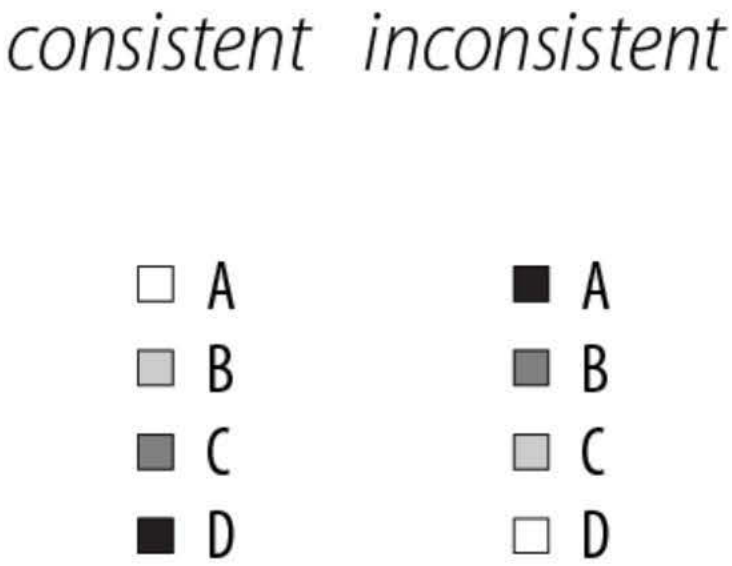
Raina SZ, et al. (2005) Evolution of base-substitution gradients in primate mitochondrial genomes. Genome Res 15: 665-673.

M. Krzwinski, behind every great visualization is a design principle, 2012

Principle 2: consistency

- A note on **legends**: order items according to appearance

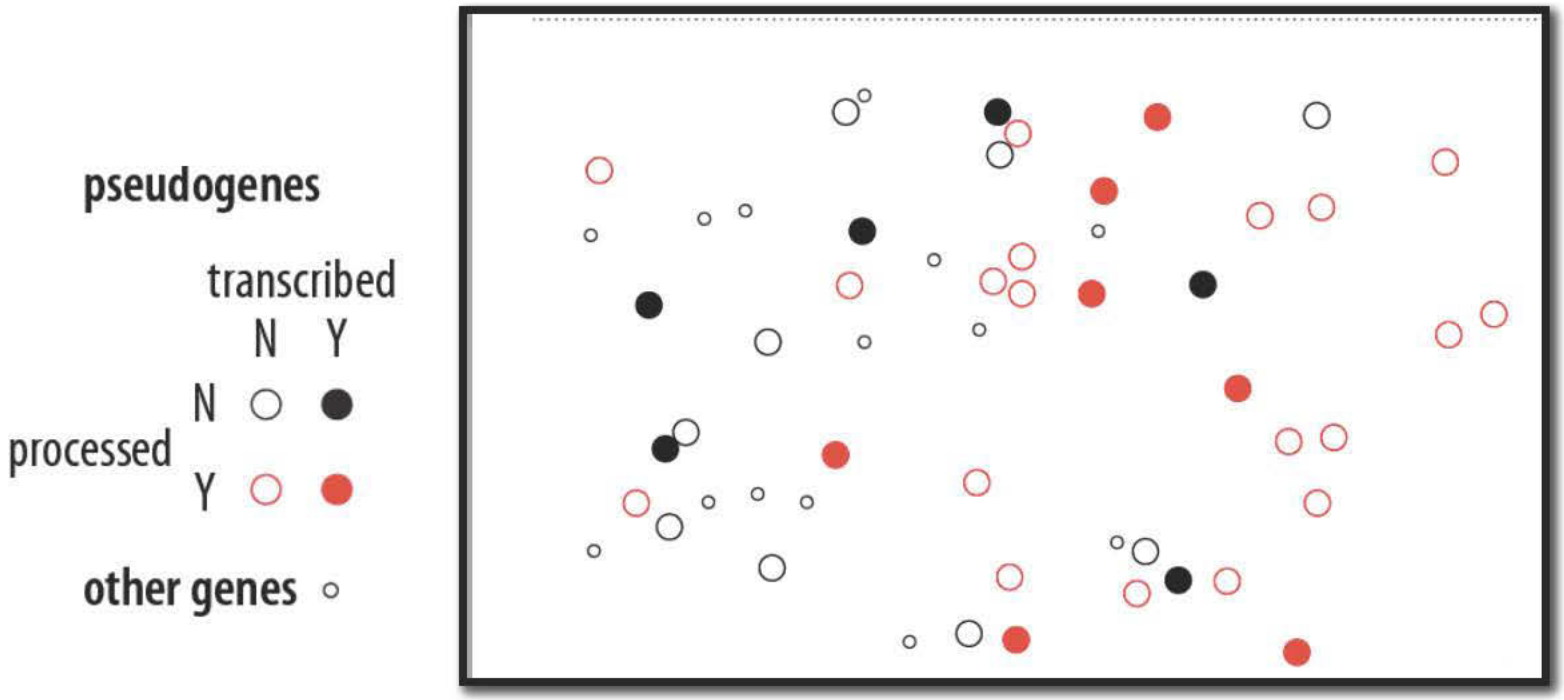




M. Krzwinski, behind every great visualization is a design principle, 2012

Principle 2: consistency

- Visual variation should **reflect and enhance** the underlying variation in the data
- Avoid **visually similar** encodings for independent variables
- Example:



Principle 2: consistency

- Uniform size and alignment reduces visual complexity and aids interpretation
- Example:

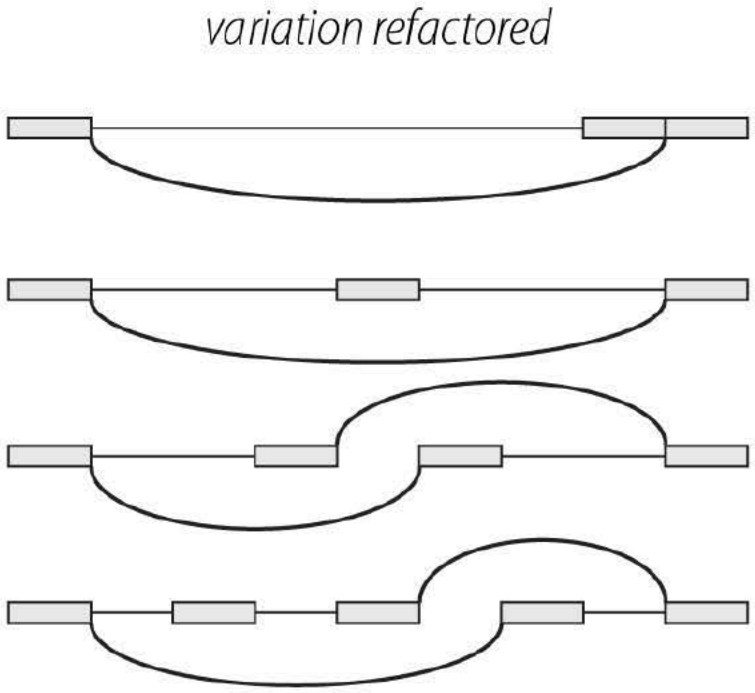
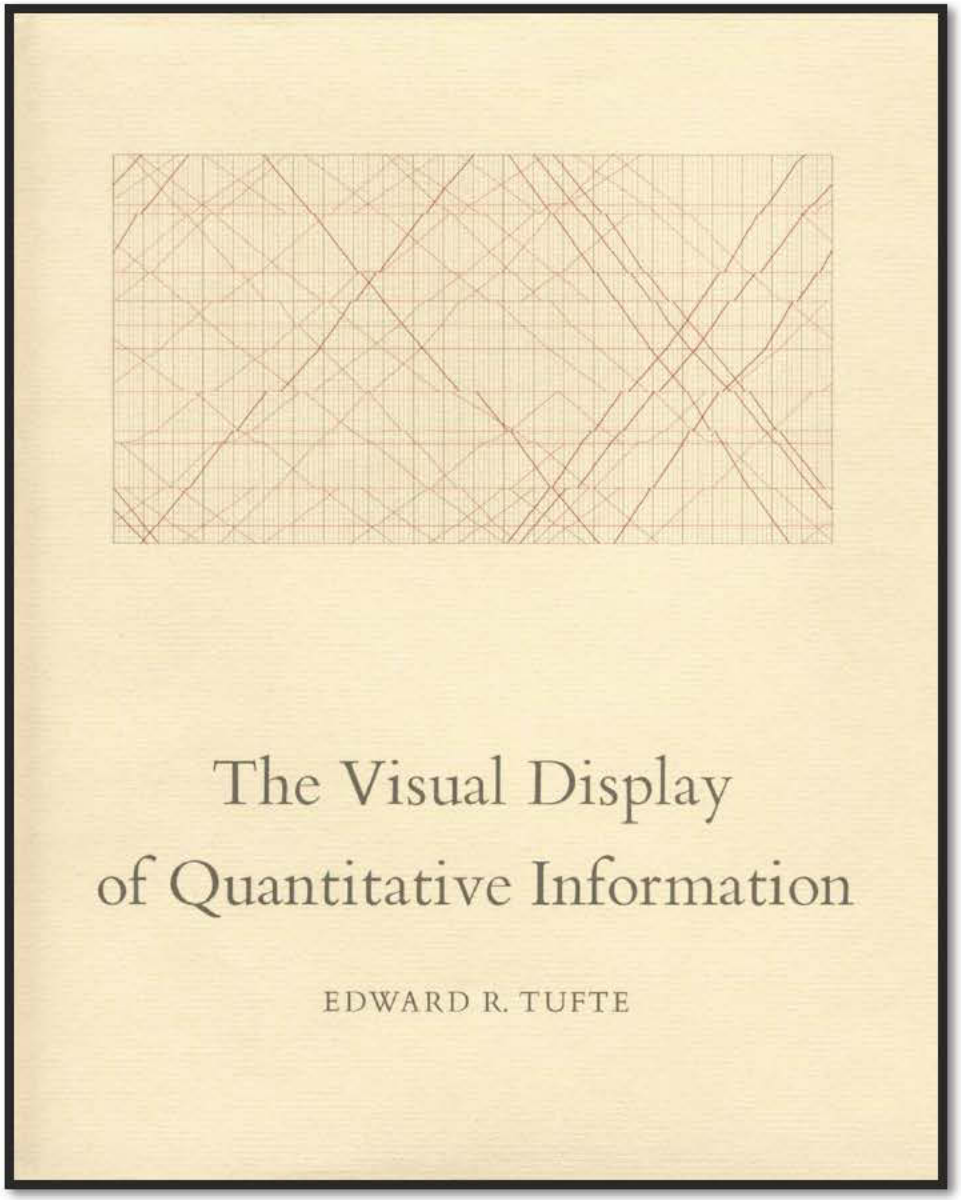


Fig. 1: Sharov AA et al. (2005) Genome-wide assembly and analysis of alternative transcripts in mouse. Genome Res 15: 748-754.
Fig. 2: M. Krzwinski, behind every great visualization is a design principle, 2012

Tufte, 1983

"Above all else, show the data."



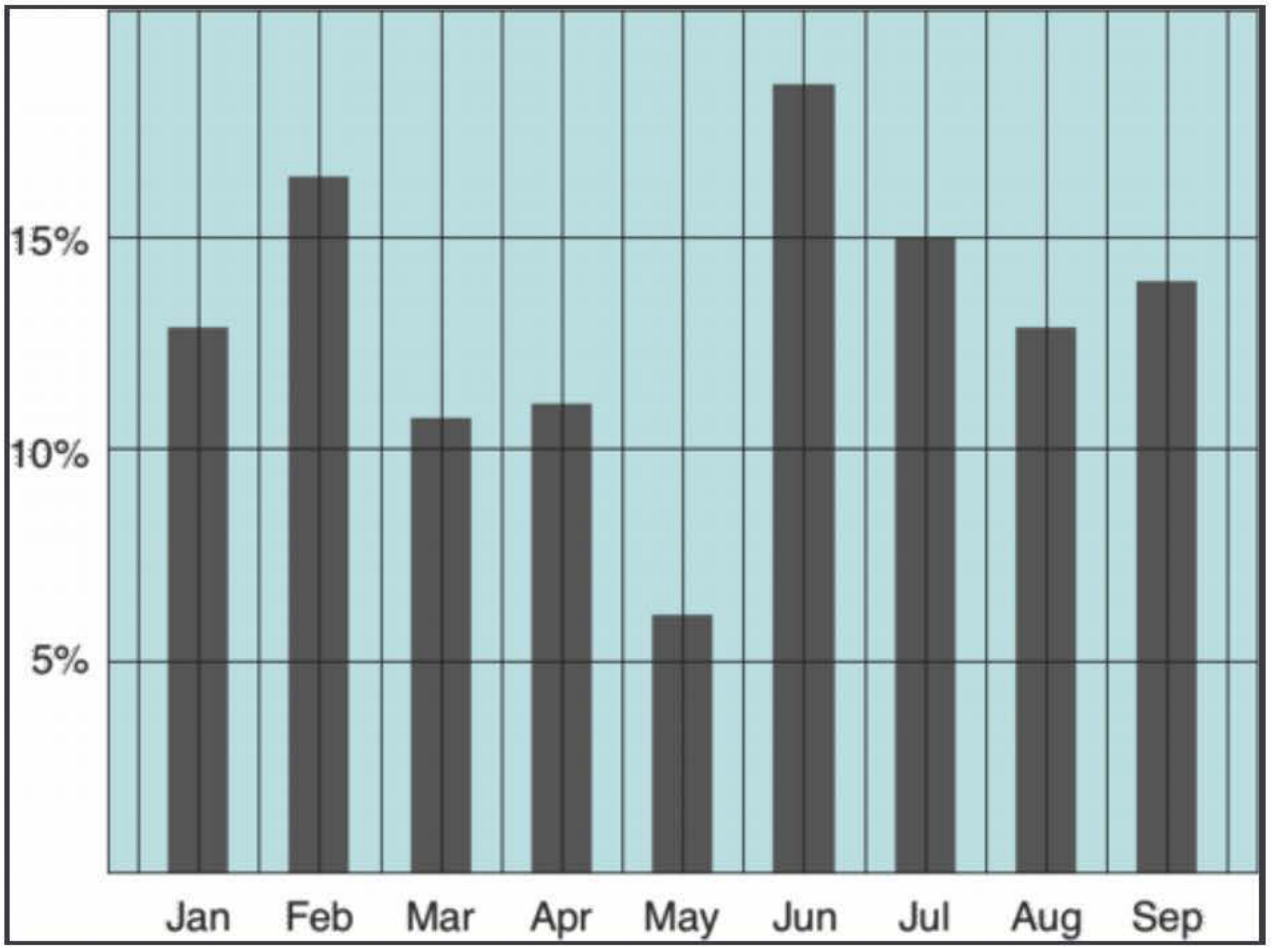
Tufte, 1983

$$\text{Data-ink ratio} = \frac{\text{Data-ink}}{\text{Total ink used to print the graphic}}$$

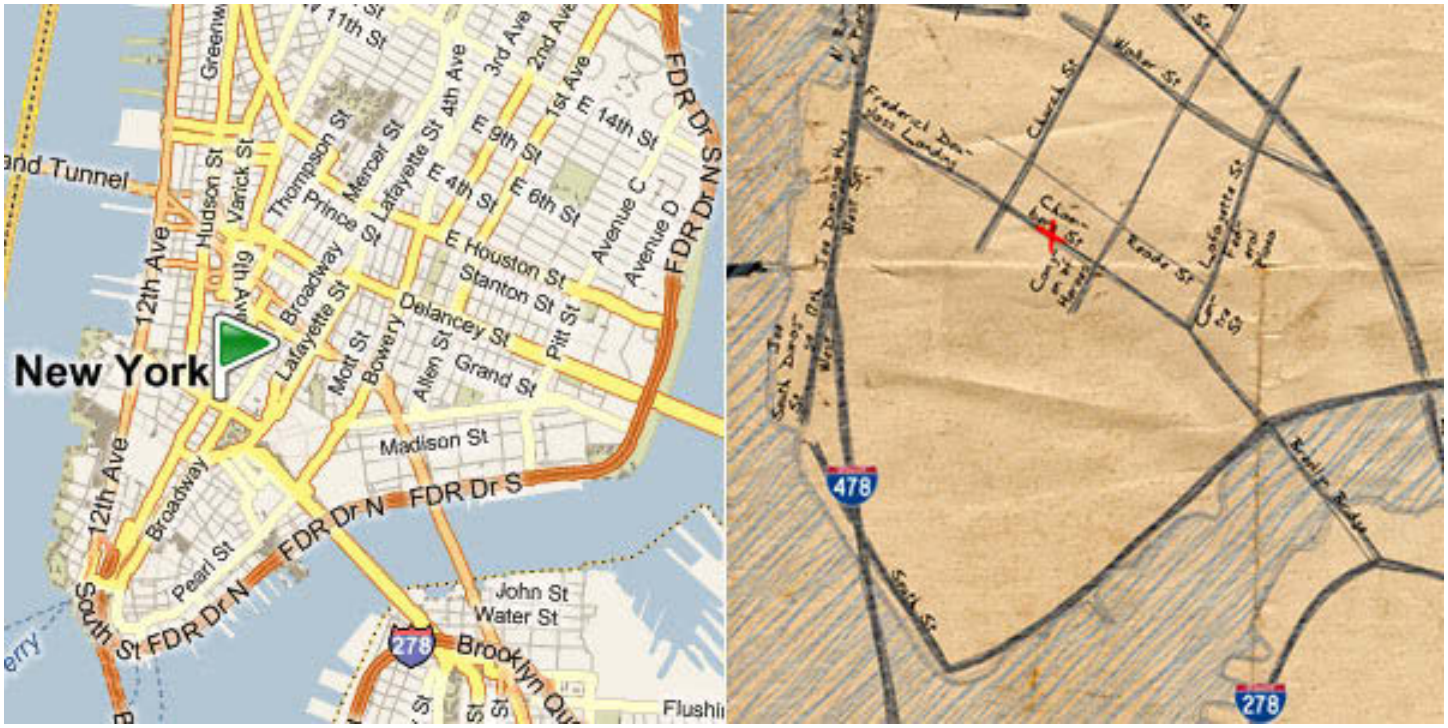
= proportion of a graphic's ink devoted to the non-redundant display of data-information

= 1 - proportion of a graphic that can be erased

Tufte: maximize the data-ink ratio



Familiar example

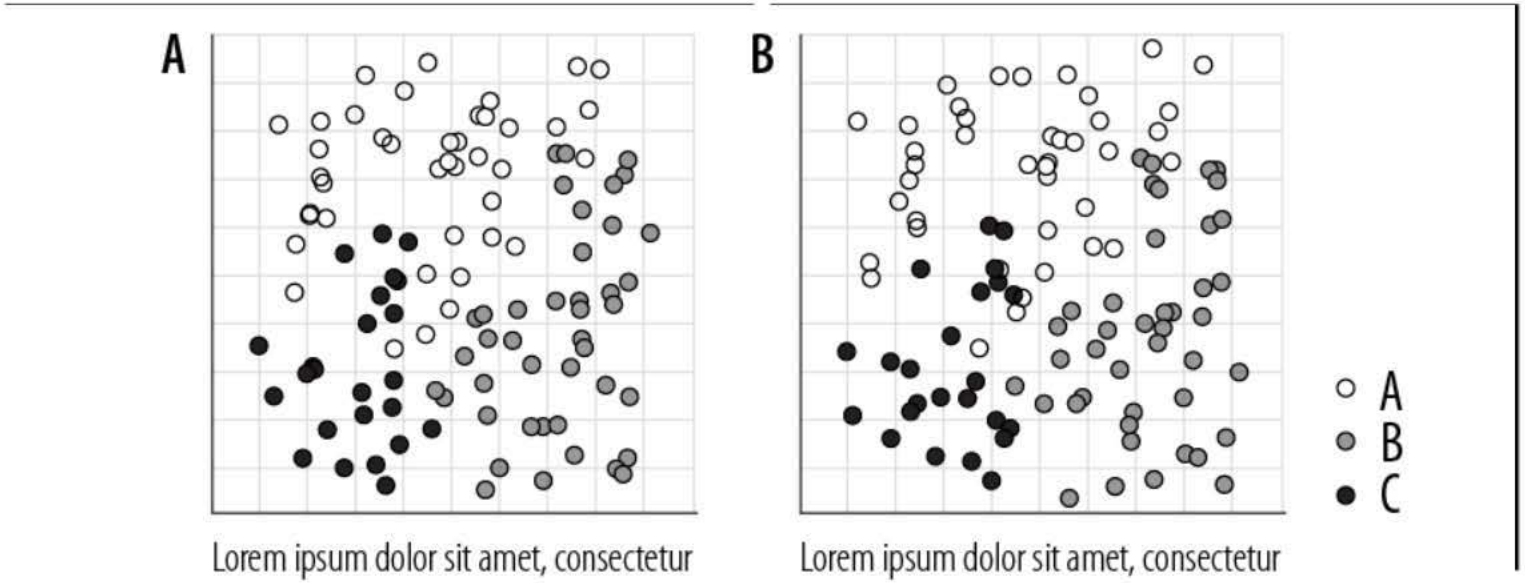


Discussion

- What do you think of the data-ink ratio?
- Consider ways to **maximize** it…

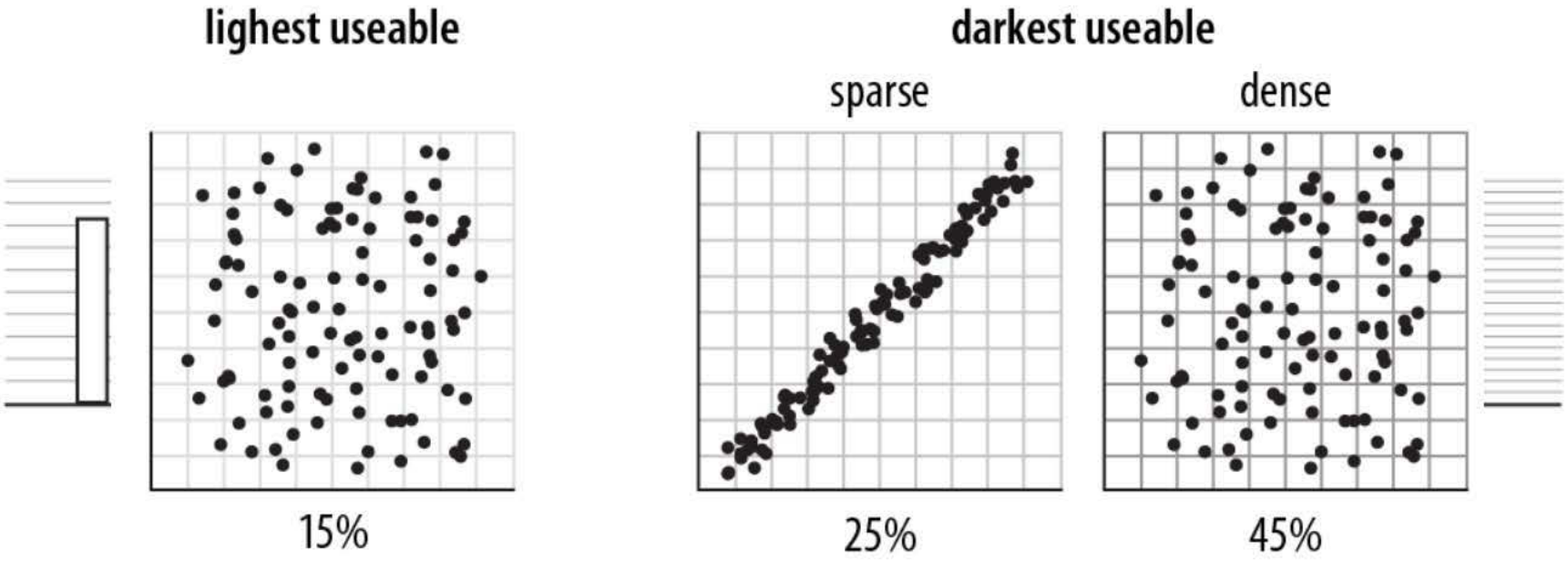
Principle 3: importance ordering

- Avoid unnecessary containment and repetition
- Example



Principle 3: importance ordering

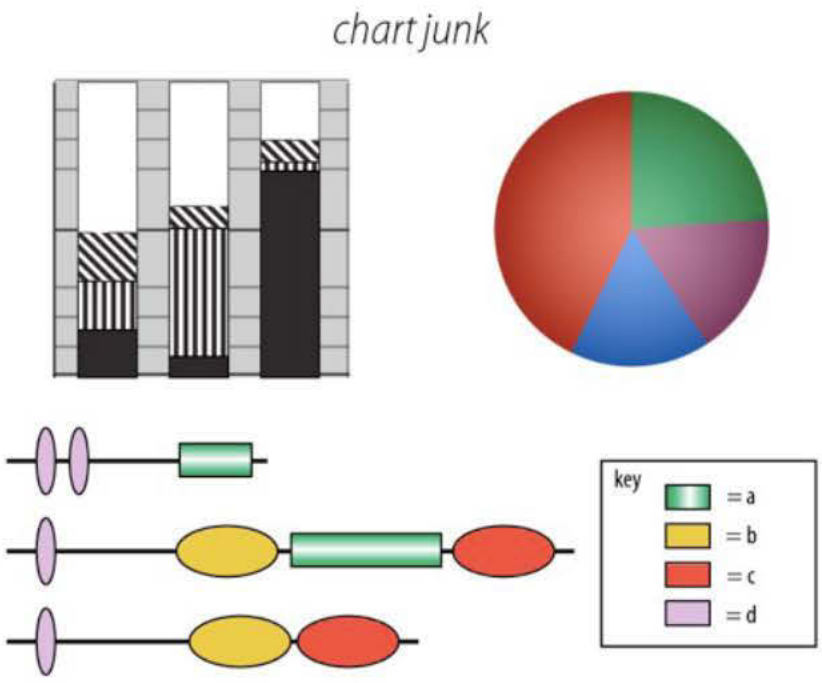
- Navigational aids shouldn't compete with data
- Avoid: heavy **axes**, **error bars** and **glyphs**

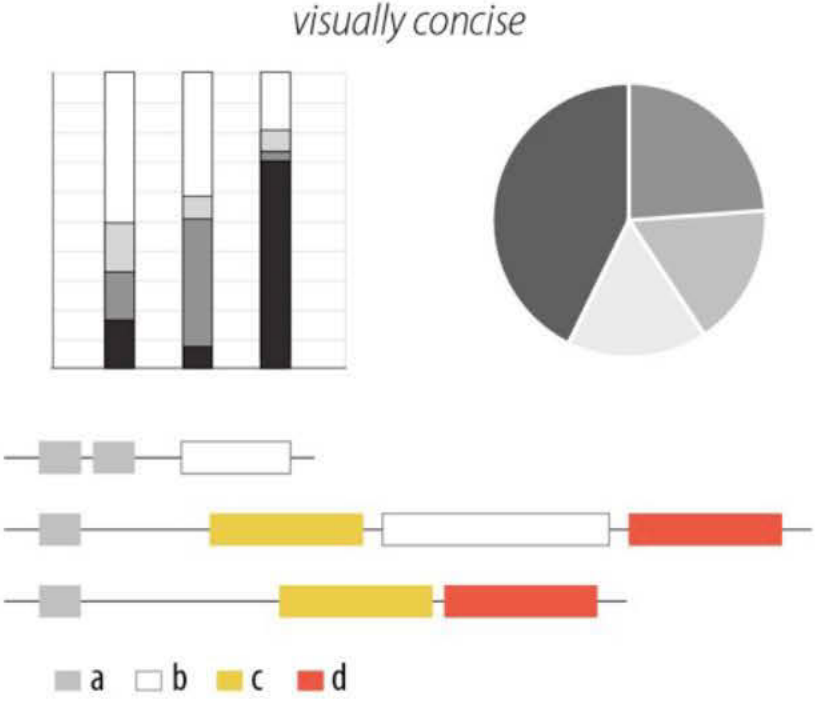


Heer J, Bostock M (2010) Crowdsourcing graphical perception: using mechanical turk to assess visualization design. Proceedings of the 28th international conference on Human factors in computing systems. Atlanta, Georgia, USA: ACM. pp. 203-212.

Principle 3: importance ordering

- Simplify, simplify, simplify…

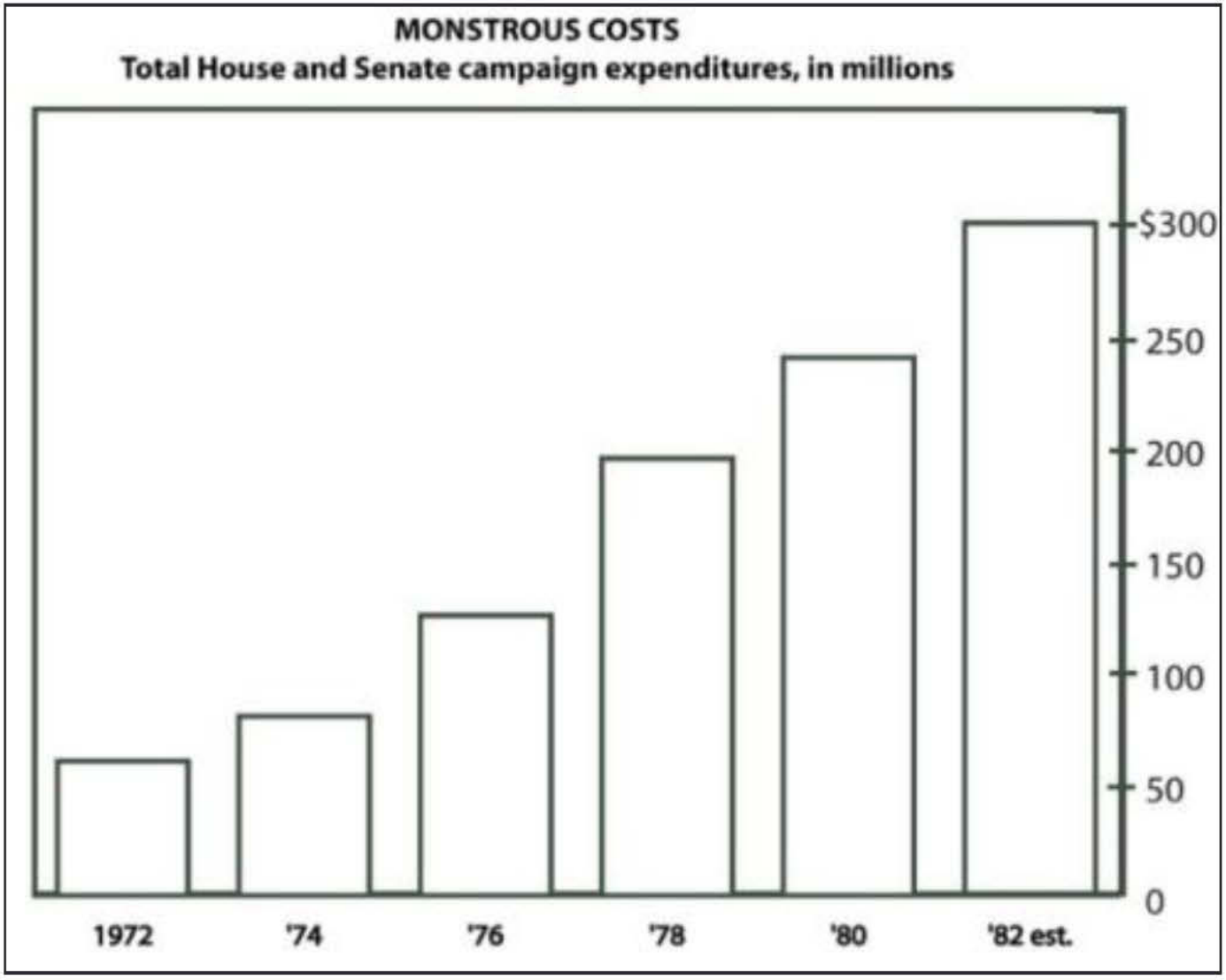




Sharov AA, et al (2006) Genome Res 16: 505-509.
Peterson J, et al. (2009) Genome Res 19: 2317-2323.
Thomson NR, et al. (2005) Genome Res 15: 629-640.
DB, Ko MS (2005) Genome Res 15: 748-754.

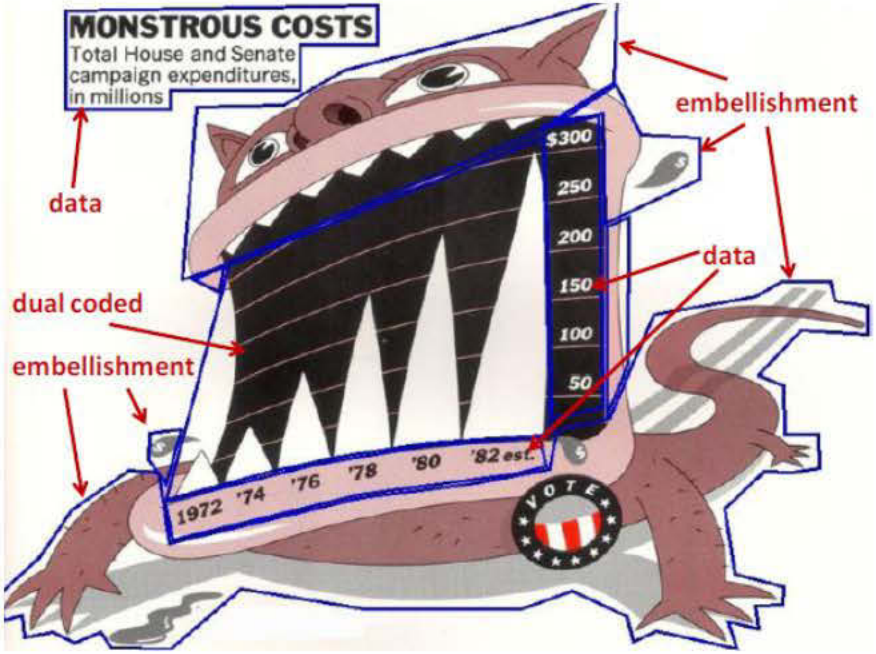
M. Krzwinski, behind every great visualization is a design principle, 2012

A caveat: “chart junk” and recall



Bateman et al. “Useful Junk? The Effects of Visual Embellishment on Comprehension and Memorability of Charts”, CHI 2010

Chart junk and eye gaze



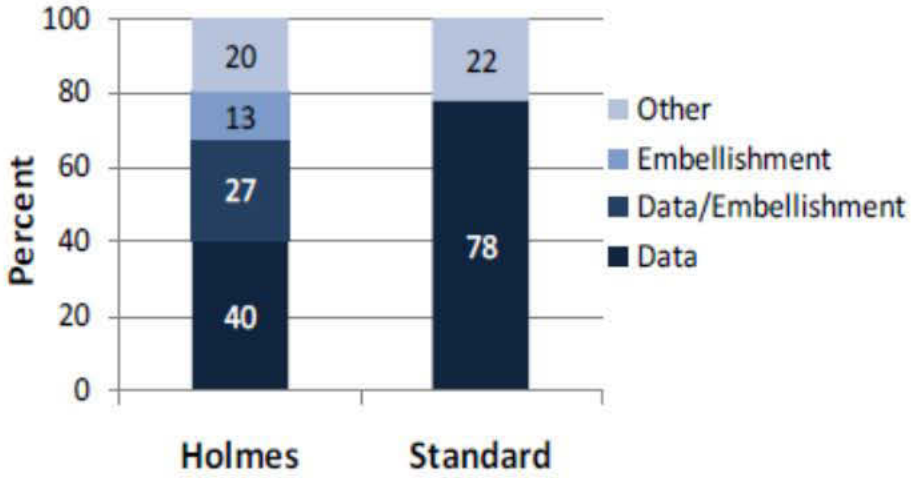


Figure 9. Percentage of on-screen time spent looking at different chart elements for Holmes and Plain charts.

Lab 1: Deconstructing Data Graphics

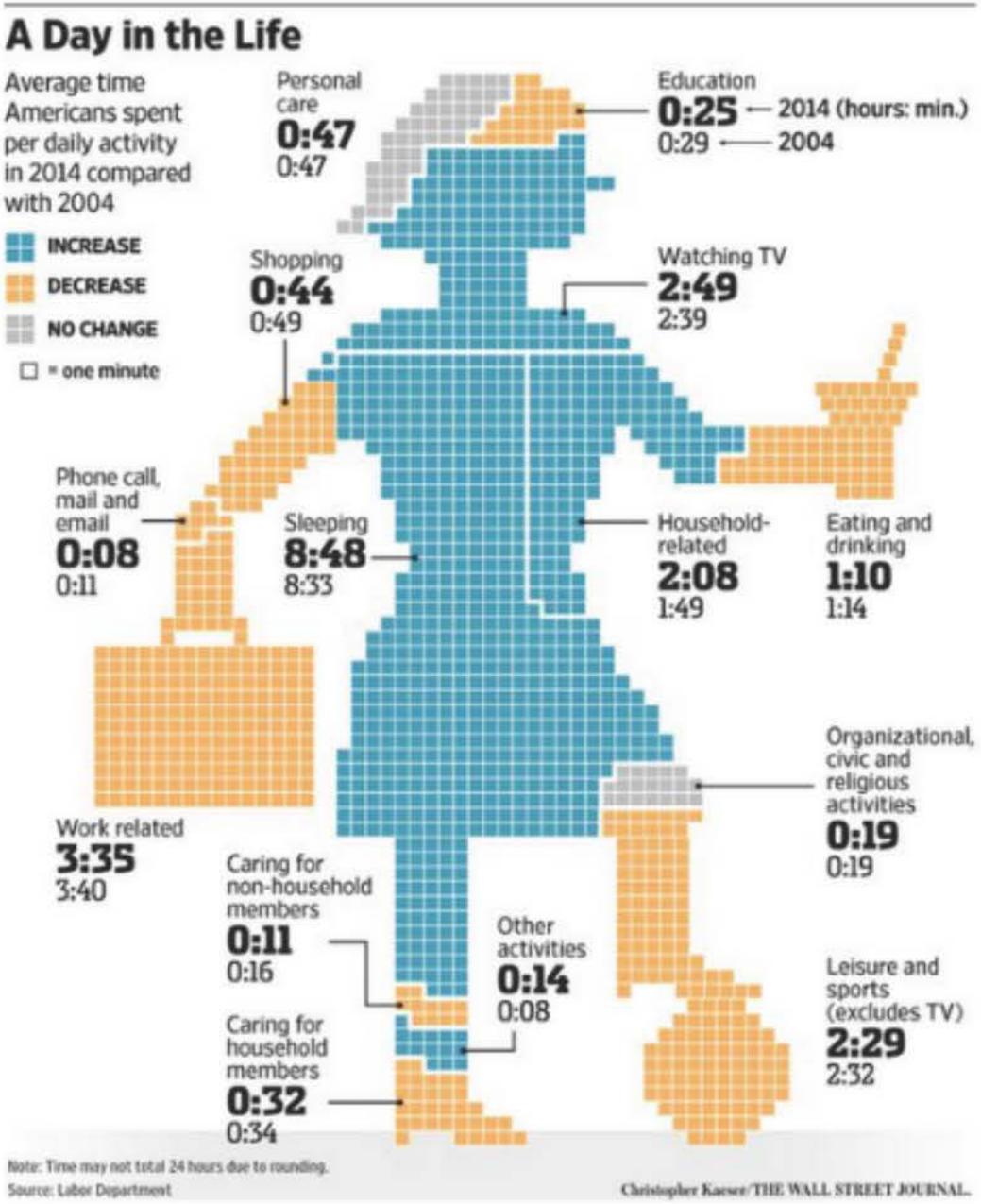
- Break into groups of 2-3 people, and go to:

  jcrouser.github.io/datavis/lab1.html

- During this lab, we'll **critique** some professionally-made visualizations using these principles
- Try to think about the following questions:
  - What is the **first thing you notice** about this visualization?
  - What **point** is this visualization trying to make?
  - Who is the **intended audience**?
  - What is the visualization **doing well**?
  - What **problems** do you see with the visualization design?
  - **Why** do you think the designer made those choices?

WSJ News Graphics @WSJGraphics · 3m

Study shows Americans are working more, sleeping more and watching more TV on.wsj.com/1QRHRBf

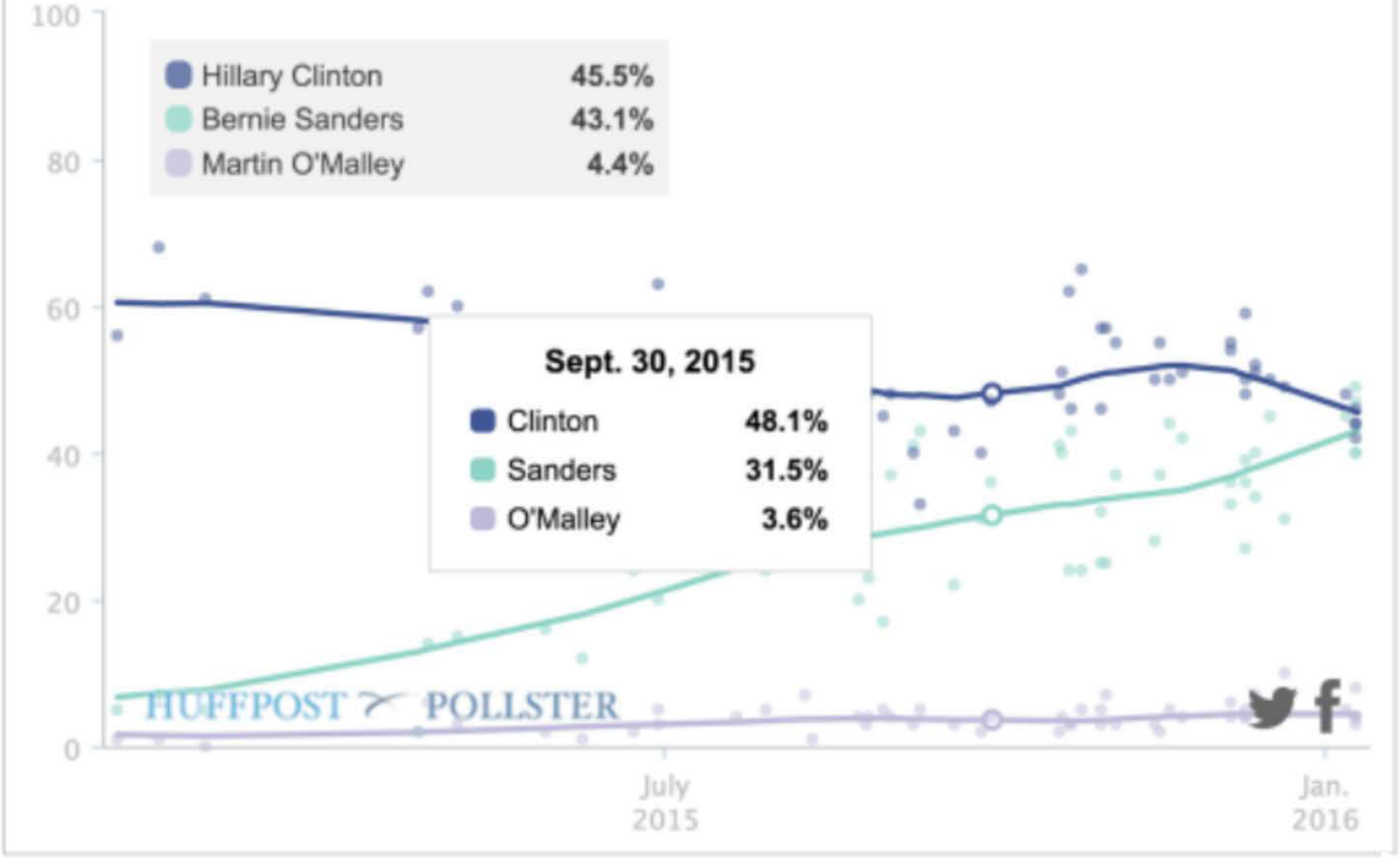


What your

BRAND COLOR

SAYS ABOUT YOUR BUSINESS

2016 Iowa Democratic Presidential Caucus



EVENTS CONTRIBUTING TO DROP OF EURO

Investors think subprime would be the U.S. only crisis

$1.4738

Investors realize EU is also vulnerable to subprime

$1.3919

The decision of European Central Bank to increase the interest rate backfires

$1.2545

Investors are worried about the weakness of the EU economy

$1.20

USD per 1 EUR

Greek debt crisis flares up

$1.2640

Euro plummets as the Greece debt crisis worsens

$1.2149

The Ukraine crisis starts to heat up

$1.36

Fear that election may result in Greece leaving Eurozone

$1.21

ECB's purchase of eurobonds devalues the Euro

$1.06

Traders push Euro down and invest in Pound

$1.0479

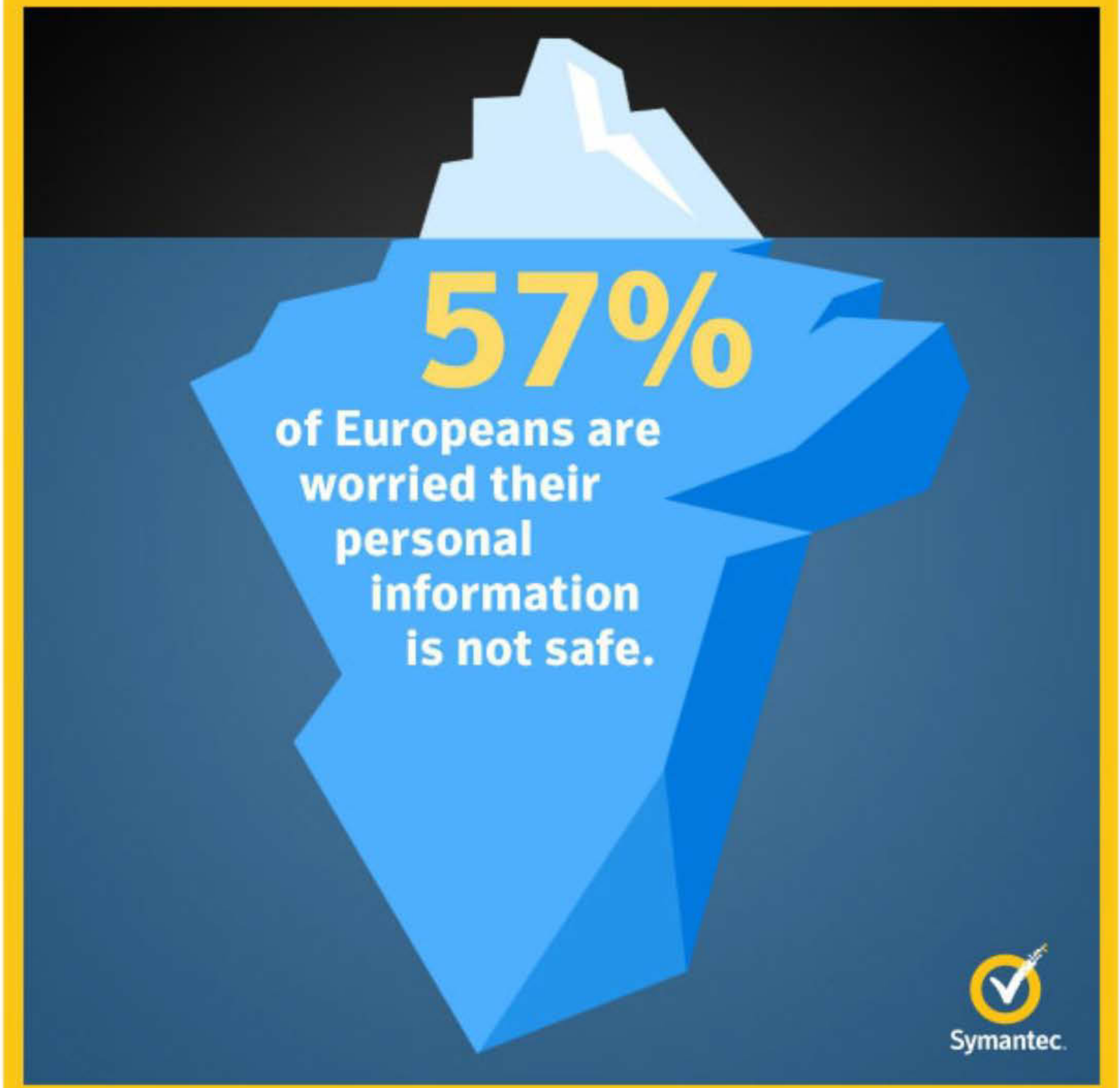
2007 2008 2009 2010 2011 2012 2013 2014 2015

UNDER PRESIDENT OBAMA,

MORE STUDENTS ARE EARNING THEIR HIGH SCHOOL DIPLOMAS THAN EVER BEFORE

#LeadOnEducation

SOURCE: U.S. DEPARTMENT OF EDUCATION, NATIONAL CENTER FOR EDUCATION STATISTICS


57%
of Europeans are worried their personal information is not safe.
Symantec.

Illinois

Variable: Net Job Creation (Per 100)
Employees, Same Sex and Age Group
Year: 2000 Quarter:1
Sex: All and Age Group: Ages 19–21

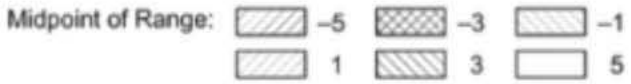


Fig. 5.7 Job creation for young workers, by county, Illinois

Who do Nike sponsor?

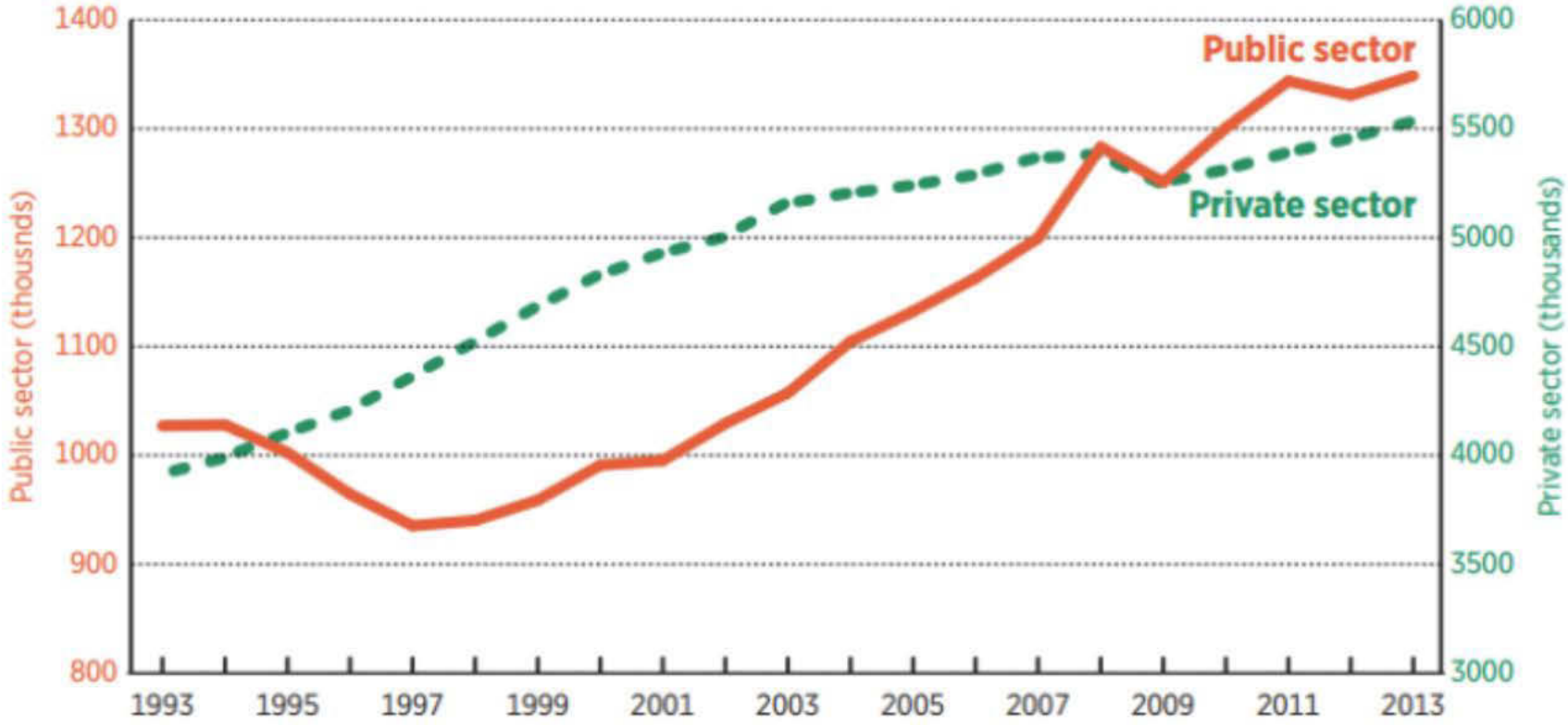
International sports and events sponsor

The American based company is the largest sports supplier in the world, suppling equipment, shoes and apparel.

1,016
athletes sponsored worldwide

$230m
spent on athlete sponsors

Figure 10: Public- and private-sector jobs (000s) in Ontario, 1993–2013



Source: Statistics Canada, CANSIM Table 282-0089: *employment by class of worker and sex, seasonally adjusted and unadjusted; Ontario; Public sector and private sector employees; Both sexes; Seasonally adjusted (x 1,000).*

2011
193,600

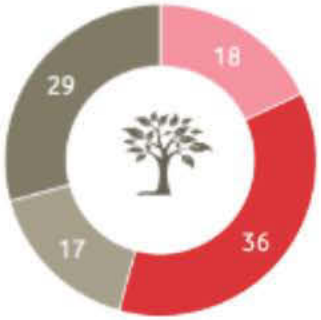
2015
117,161

Despite the hysteria, the number of refugees in the UK has actually fallen by 76,439 since 2011.

Vodafone Institute
for Society
and Communications

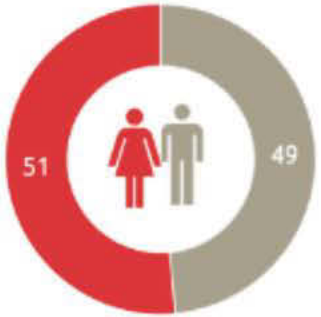
Sample Description

Age

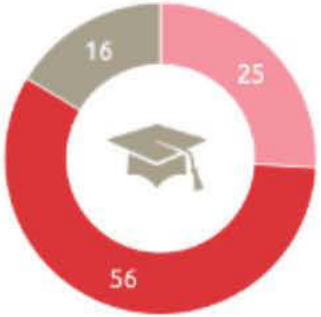


Gender*

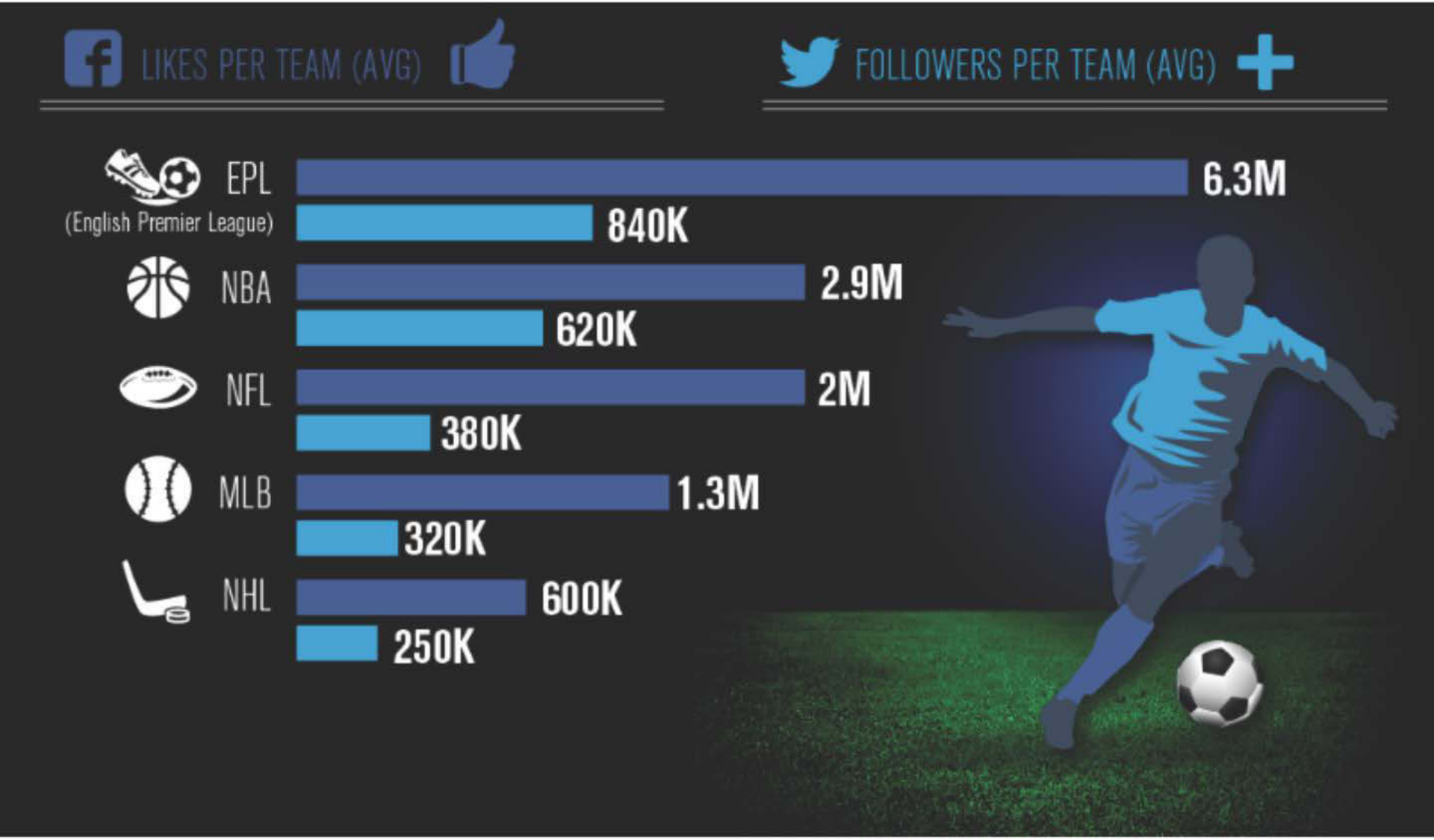
%



Educational level



* Due to rounding, numbers presented in this document may not add up precisely to 100%


LIKES PER TEAM (AVG)
FOLLOWERS PER TEAM (AVG)
EPL
(English Premier League)
6.3M
840K
NBA
2.9M
620K
NFL
2M
380K
MLB
1.3M
320K
NHL
600K
250K

Coming up next

- Grammar of graphics
- Introduction to `ggplot2`
- Lab: Make a Scatterplot