

INFORMATION VISUALIZATION

Introduction

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INSTRUCTORS

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OFFICE — Digiteo Moulon Building

OFFICE HOURS — By appointment



YOU!

QUICK INTROS

Any particular interests?

COURSE INFO

Visualization

Exam

November - January

TBD

Class website:

See Slack channel – everyone will get an invite

READINGS

mostly for additional interest

will announce readings on a per-lecture basis

ELECTRONICS POLICY

Laptops and devices okay
(in fact you'll need them)
...but use them for work!



BEHAVIOR & SOCIETY

Students are Better Off without a Laptop in the Classroom

What do you think they'll actually use it for?

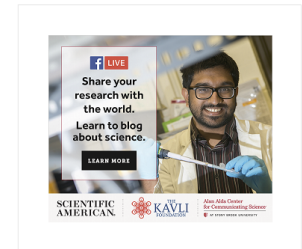
By Cindi May on July 11, 2017



Credit: Getty Images

As recent high school graduates prepare for their migration to college in the fall, one item is sure to top most students' shopping wish lists: a laptop computer. Laptops are ubiquitous on university campuses, and are viewed by most students as absolute must-have items, right alongside laundry detergent, towels, and coffee pots.

Without question, personal laptops can enhance the college experience by facilitating engagement with online course material, providing access to sources for research, maximizing internship searches, and even improving communication with friends and parents. Many students also opt to bring their laptops to class so that they can take notes, view online lecture slides, and search the web for course-related material. This practice, it



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LATEST NEWS



Puerto Rico Looks to Alphabet's X Project Loon Balloons to Restore Cell Service



The Ethical Minefields of Technology



Astronomers Are Finally Mapping the "Dark Side" of the Milky Way

GRADING SCHEME

- Assignments: 66%
 - check the website for due dates of assignments and how to submit them
 - Slack participation
- Exam: 33%

web <https://www.aviz.fr/Teaching2018/InteractiveInfoVis>

(stuff to download for each class)

don't forget to contact Petra for **Slack**  invites!

PROJECT

real data providers bring their data
they want/need to explore

you work on a real problem with real data
see the reality of data visualization challenges
make an impact



EXERCISES

3-4 exercises

1 (individual) with curated data

several (group) building towards your project



DATAFAIR



(before next class)

early next week we'll be given a link
to brief data descriptions (data-briefs)
consider a few you find interesting)

next class (29/11) we meet at building **660** (room TBD)

elevator speech by data providers

“speed data-dating“

form groups and talk more with providers

AFTER TODAY YOU WILL...

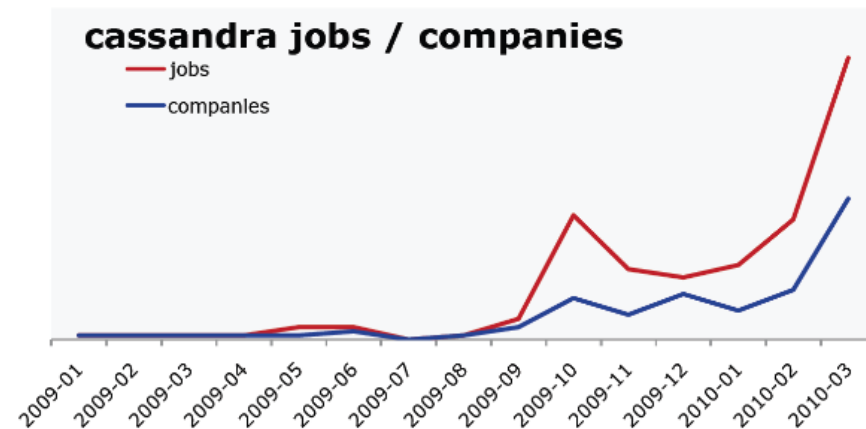
have gained an overview of the research area

learned basic principles of data representation and interaction

Why

INFORMATION VISUALIZATION

Hiring trends for data science



It's not easy to get a handle on jobs in data science. However, data from [O'Reilly Research](#) shows a steady year-over-year increase in Hadoop and Cassandra job listings, which are good proxies for the "data science" market as a whole. This graph shows the increase in Cassandra jobs, and the companies listing Cassandra positions, over time.

"The ability to take data -- to be able to understand it, to process it, to extract value from it, to visualize it, to communicate it - that's going to be a hugely important skill in the next decades."

Hal Varian, chief economist at Google

QUESTION

how can we effectively access data?

- understand its structure?
- make comparisons?
- make decisions?
- gain new knowledge?
- convince others?
- ...

MANY POSSIBLE WAYS TO ADDRESS...



Information Visualization

EXAMPLE

I		II		III		IV	
x	y	x	y	x	y	x	y
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89

Raw Data from Anscombe's Quartet

[Source: Anscombe's quartet, Wikipedia]

STATISTICAL ANALYSIS

For all four columns, the statistics are identical

I		II		III		IV	
x	y	x	y	x	y	x	y
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89

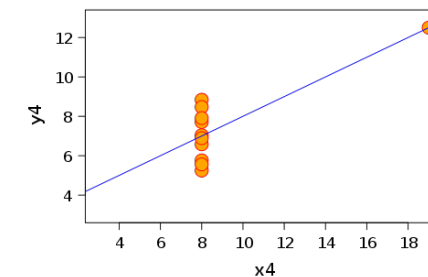
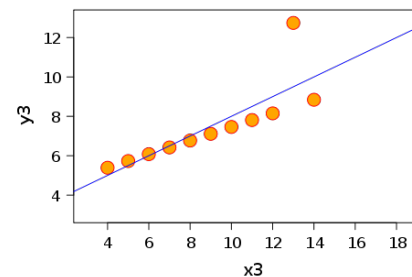
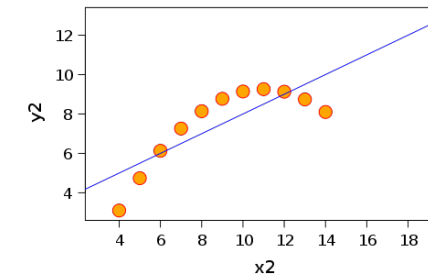
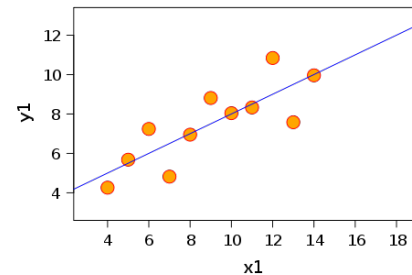
Mean of x	9.0
Variance of x	11.0
Mean of y	7.5
Variance of y	4.12
Correlation between x and y	0.816
Linear regression line	$y = 3 + 0.5x$

[Source: Anscombe's quartet, Wikipedia]

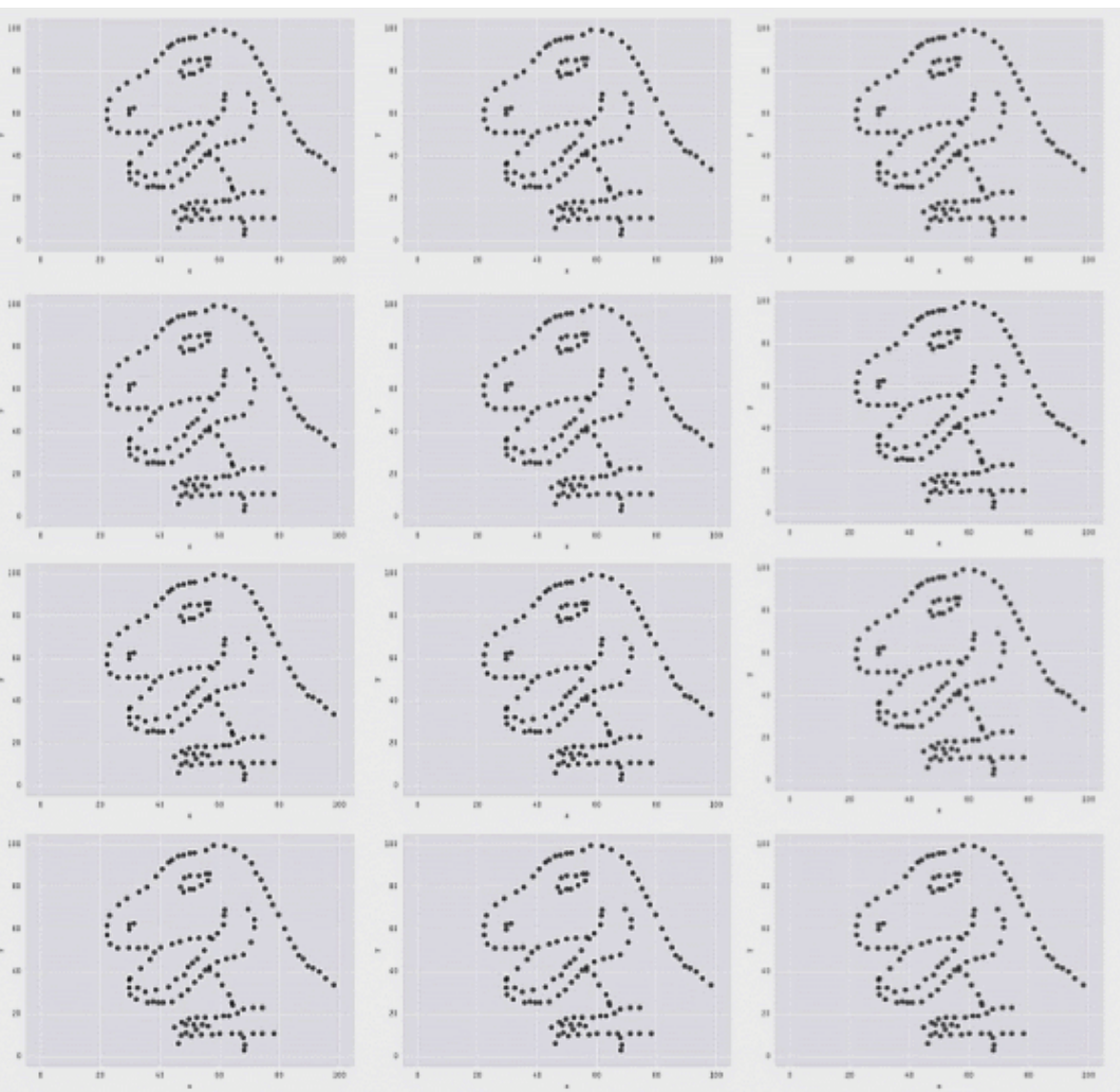
VISUAL REPRESENTATION OF THE DATA

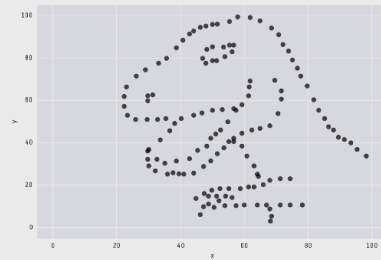
Visual representation reveals a different story

I		II		III		IV	
x	y	x	y	x	y	x	y
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
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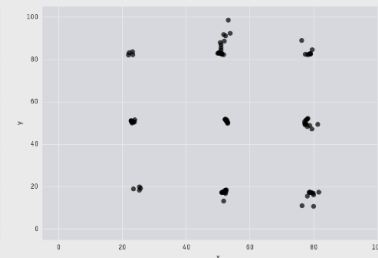
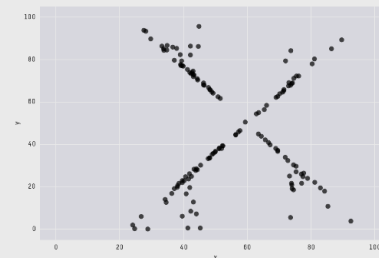
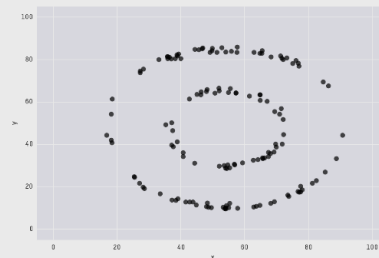
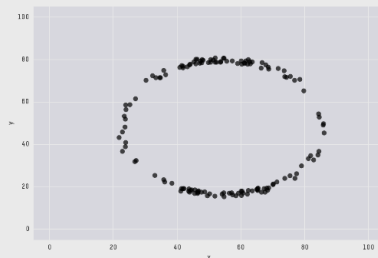
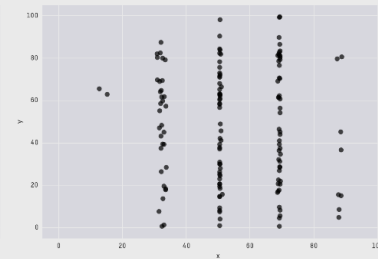
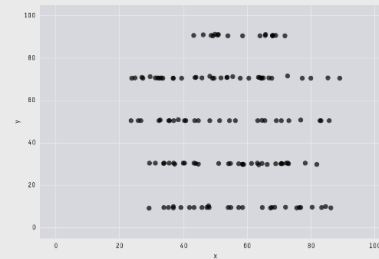
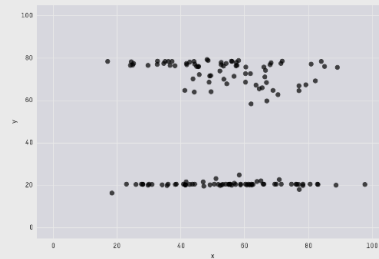
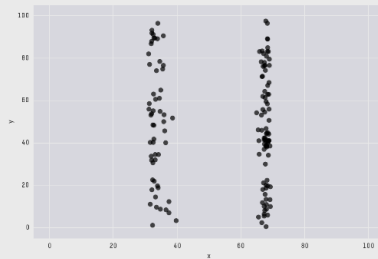
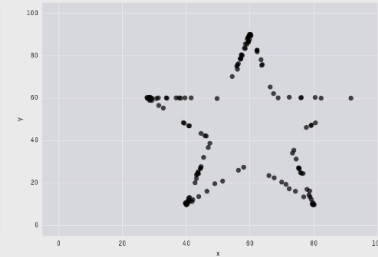
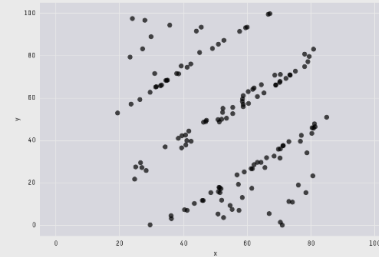
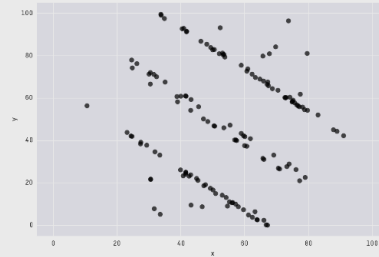
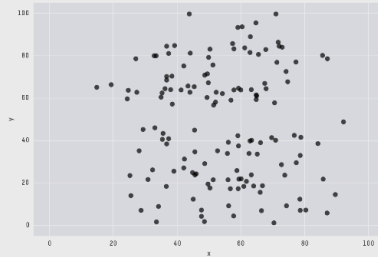


[Source: Anscombe's quartet, Wikipedia]





X Mean: 54.26
 Y Mean: 47.83
 X SD : 16.76
 Y SD : 26.93
 Corr. : -0.06

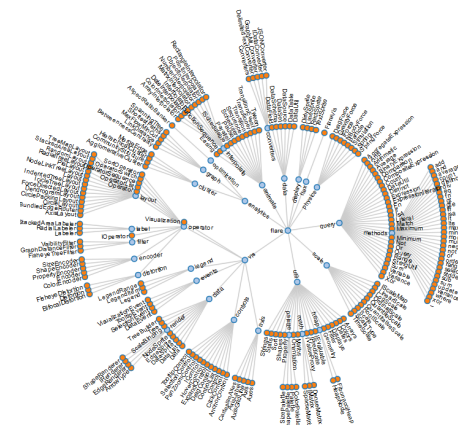


Why visual data representations?

Vision is our most dominant sense

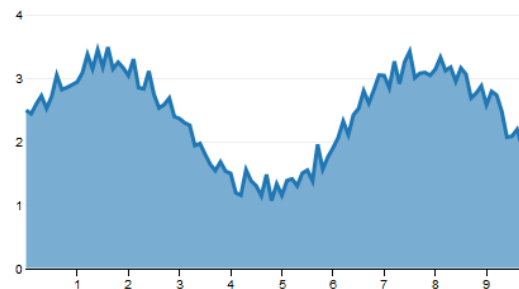
We are very good at recognizing visual patterns

We need to see and understand in order to explain,
reason, and make decisions

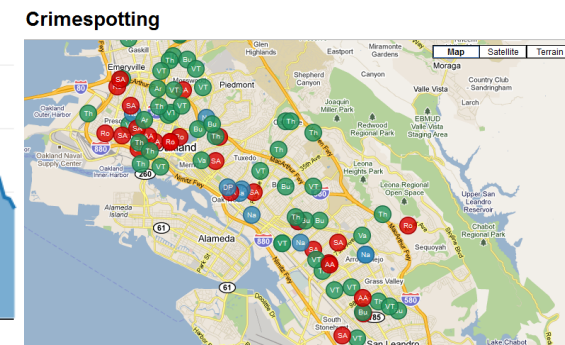


graphs / hierarchies

common examples:



charts



maps

all examples from: <http://vis.stanford.edu/protovis/>

Other benefits of visualization

expand human working memory

- offload cognitive resources to the visual system,

reduce search

- by representing a large amount of data in a small space,

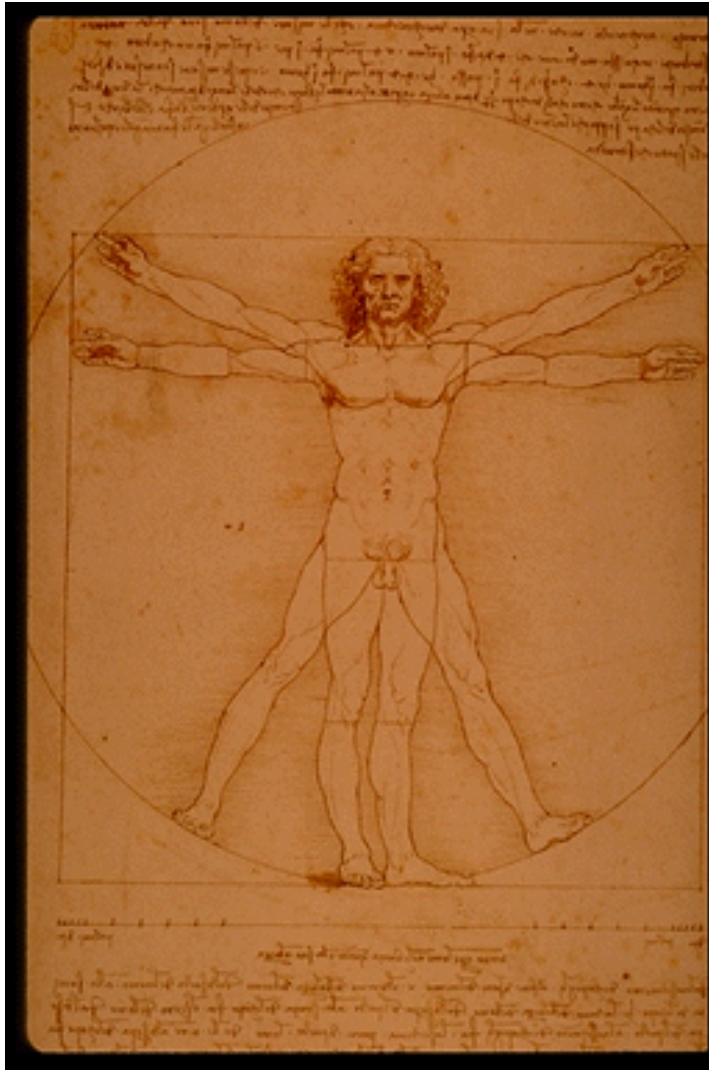
enhance the recognition of patterns

- by making them visually explicit

aid monitoring of a large number of potential events

provides a manipulable medium

- & allows exploration of a space of parameter values.



L'occhio,
che si dice finestra dell'anima,
è la principale via donde il comune senso
può piú copiosamente e magnificamente
considerare
le infinite opere di natura.

Leonardo da Vinci
(1452 - 1519)

The eye...
the window of the soul,
is the principal means
by which the central sense
can most completely and
abundantly appreciate
the infinite works of nature.

百聞不如一見

"One hundred rumors are not comparable to one look."

An Old Chinese Inscription

Information visualization

Create visual representation

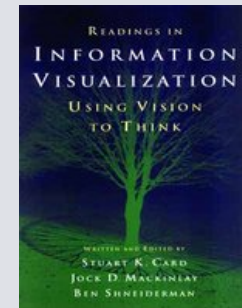
Concentrates on abstract data

Includes interaction

Official Definition:

The use of computer-supported, interactive, visual representations of abstract data to amplify cognition.

[Card et al., 1999]



Functions of Visualizations

Recording information

Tables, blueprints, satellite images

Processing information

needs feedback and interaction

Presenting information

share, collaborate, revise

for oneself, for one's peers and to teach

Seeing the unseen

Visualization of abstract data has been practiced for hundreds of years...

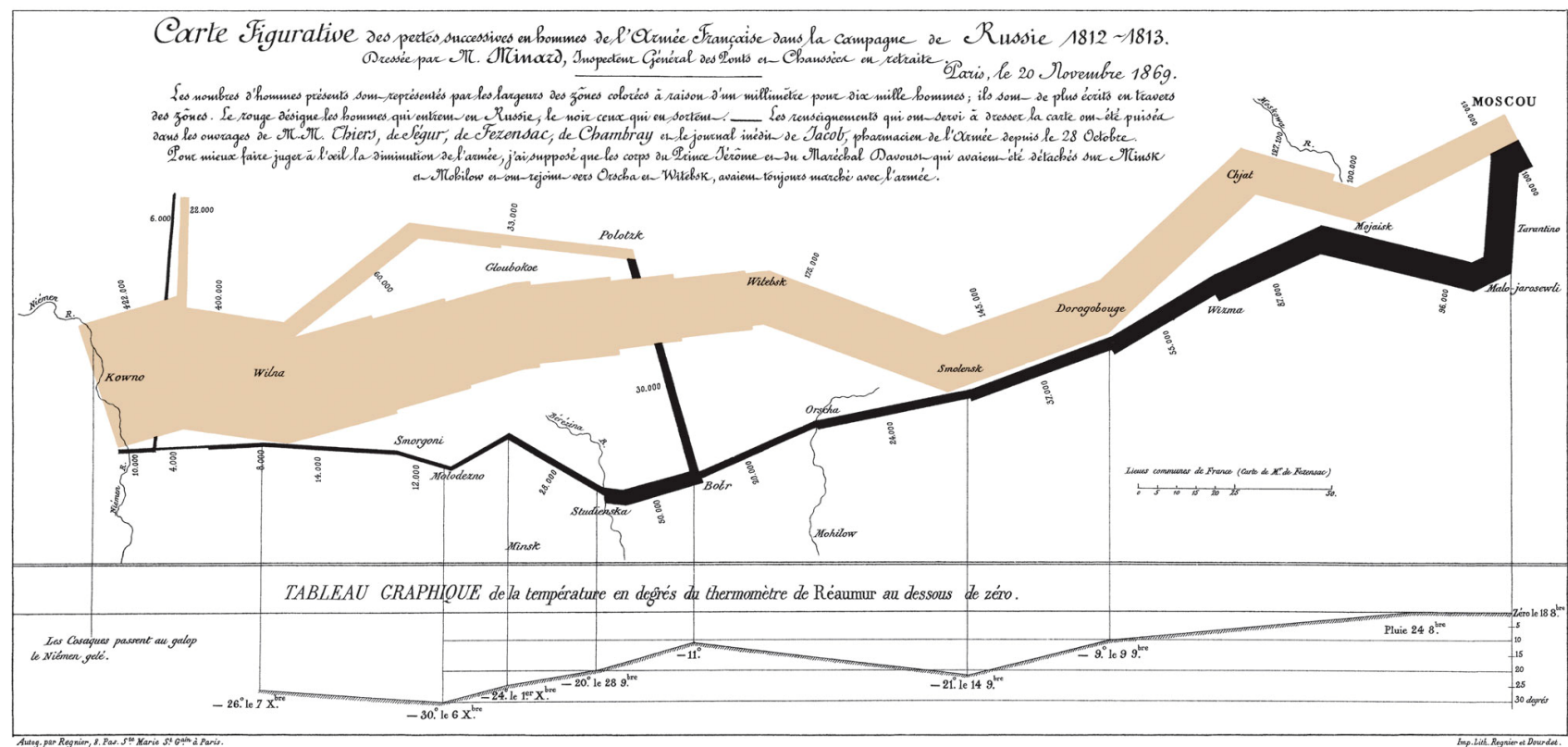
HISTORICAL EXAMPLES

Napoleon's March on Moscow

Charles Minard, 1869

Named the best statistical graphic ever drawn (by Edward Tufte)

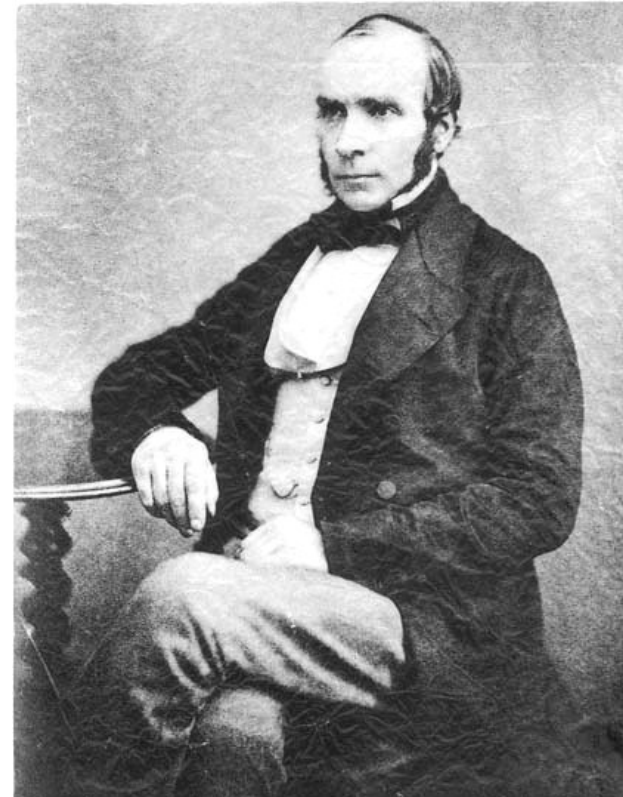
- Includes: spatial layout linked with stats on: army size, temperature, time
- Tells a story in one overview



More info: The Visual Display of Quantitative Information (Tufte)

The Broadway Street Pump

- In 1854 cholera broke out in London
 - 127 people near Broad Street died within 3 days
 - 616 people died within 30 days
- “Miasma in the atmosphere”
- Dr. John Snow was the first to link contaminated water to the outbreak of cholera
- How did he do it?
 - he talked to local residents
 - identified a water pump as a likely source
 - used maps to illustrate his theory
 - convinced authorities to disable the pump

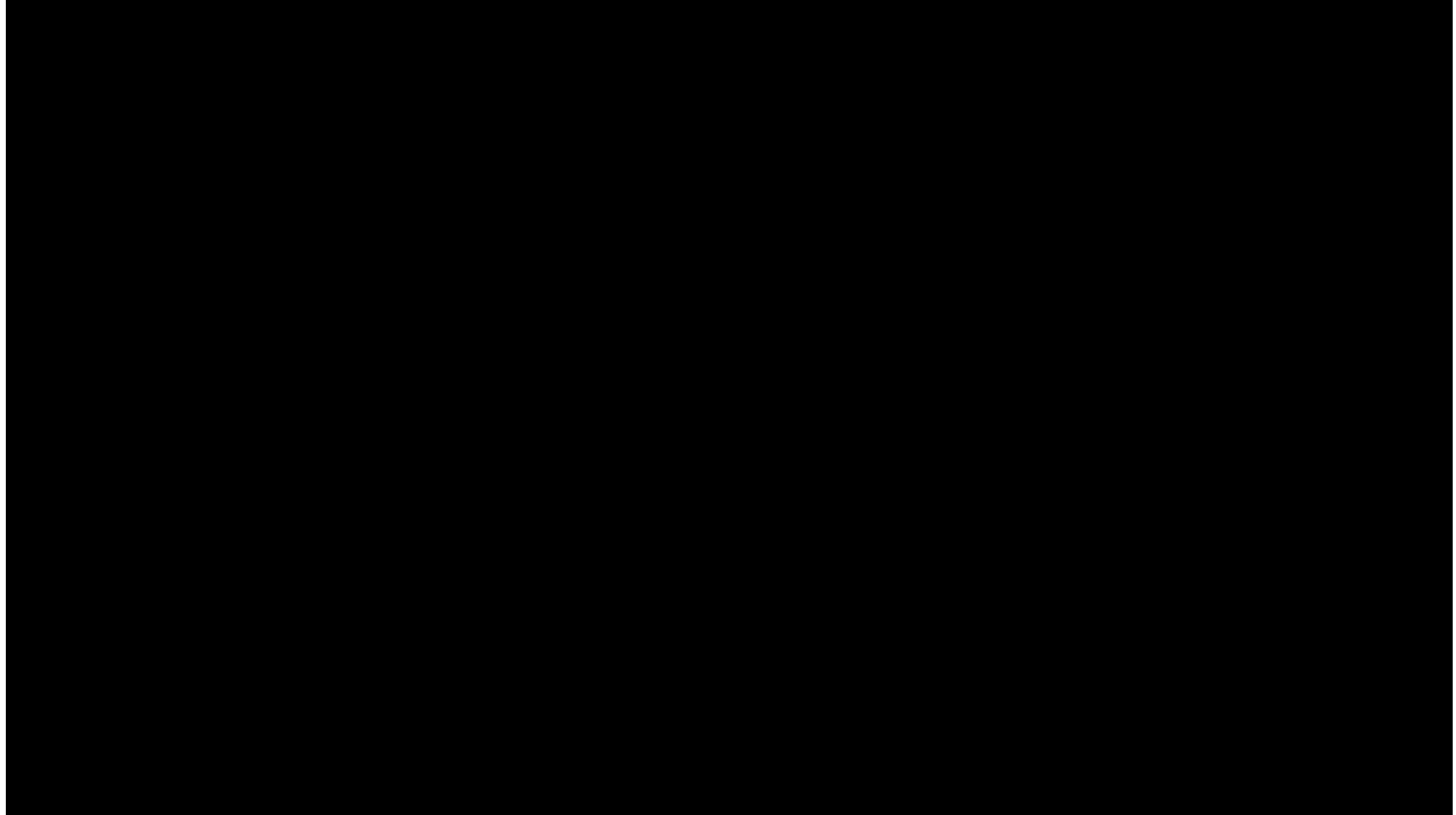


More info here: http://en.wikipedia.org/wiki/1854_Broad_Street_cholera_outbreak



... AND MORE RECENTLY

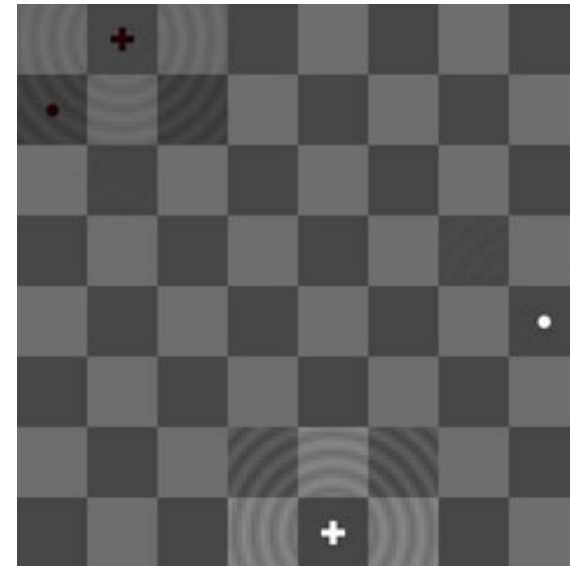
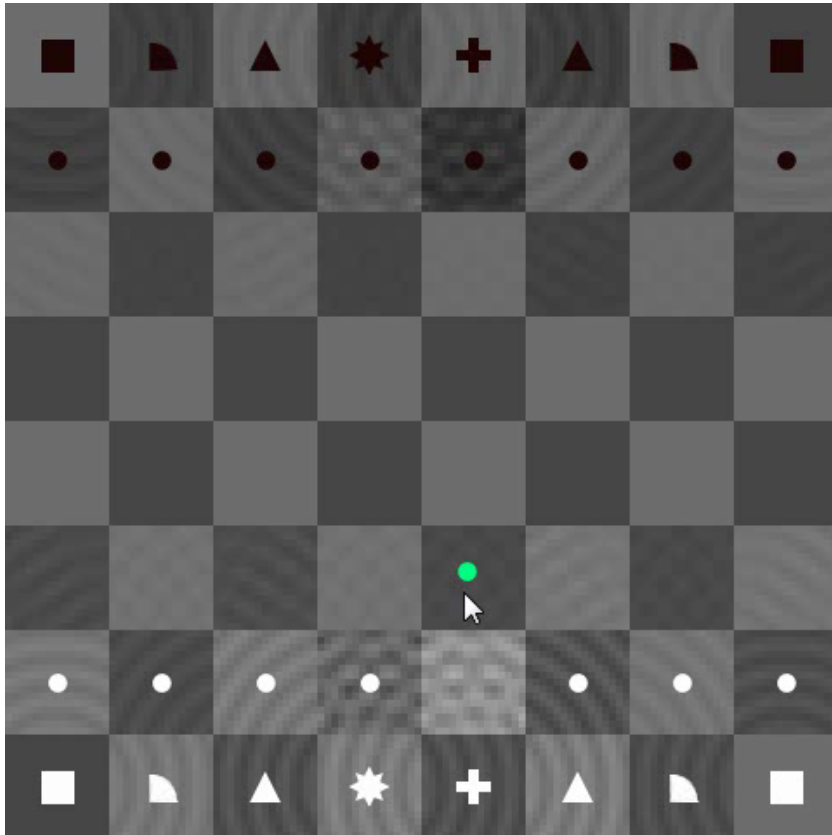
TrashTrack



Winner of the NSF International Science & Engineering Visualization Challenge!

<http://senseable.mit.edu/trashtrack/>

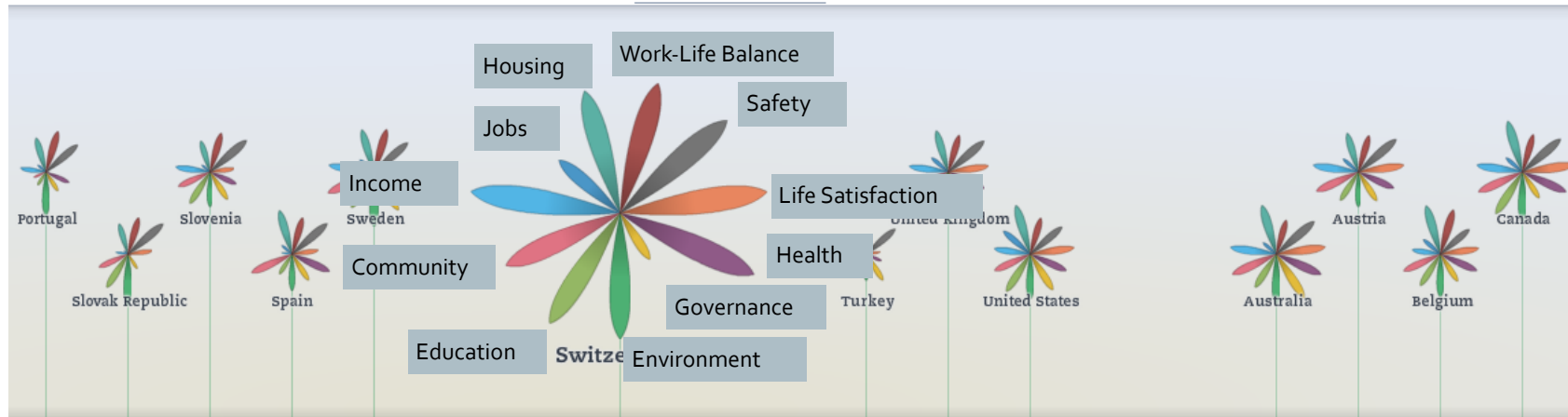
Artificial Intelligence



<http://www.turbulence.org/spotlight/thinking/chess.html>

Open Data

- Movement making government data freely available
- Encourage participation by everyone

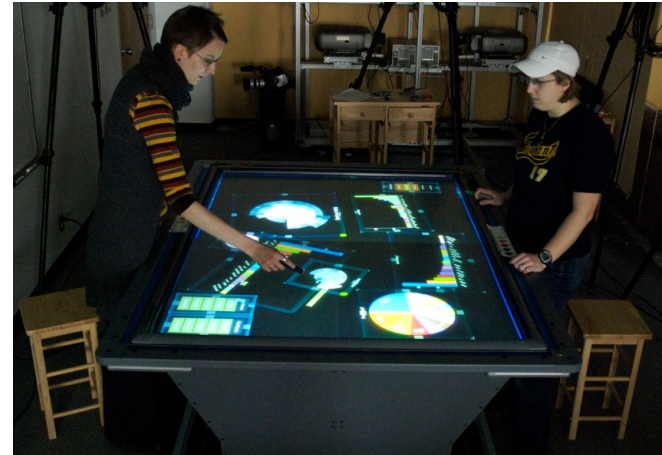


OECD Better Life Index: <http://www.oecdbetterlifeindex.org/>

Specific Visualization Environments



Molecular visualisation in the Reality Cube
University of Groningen, NL



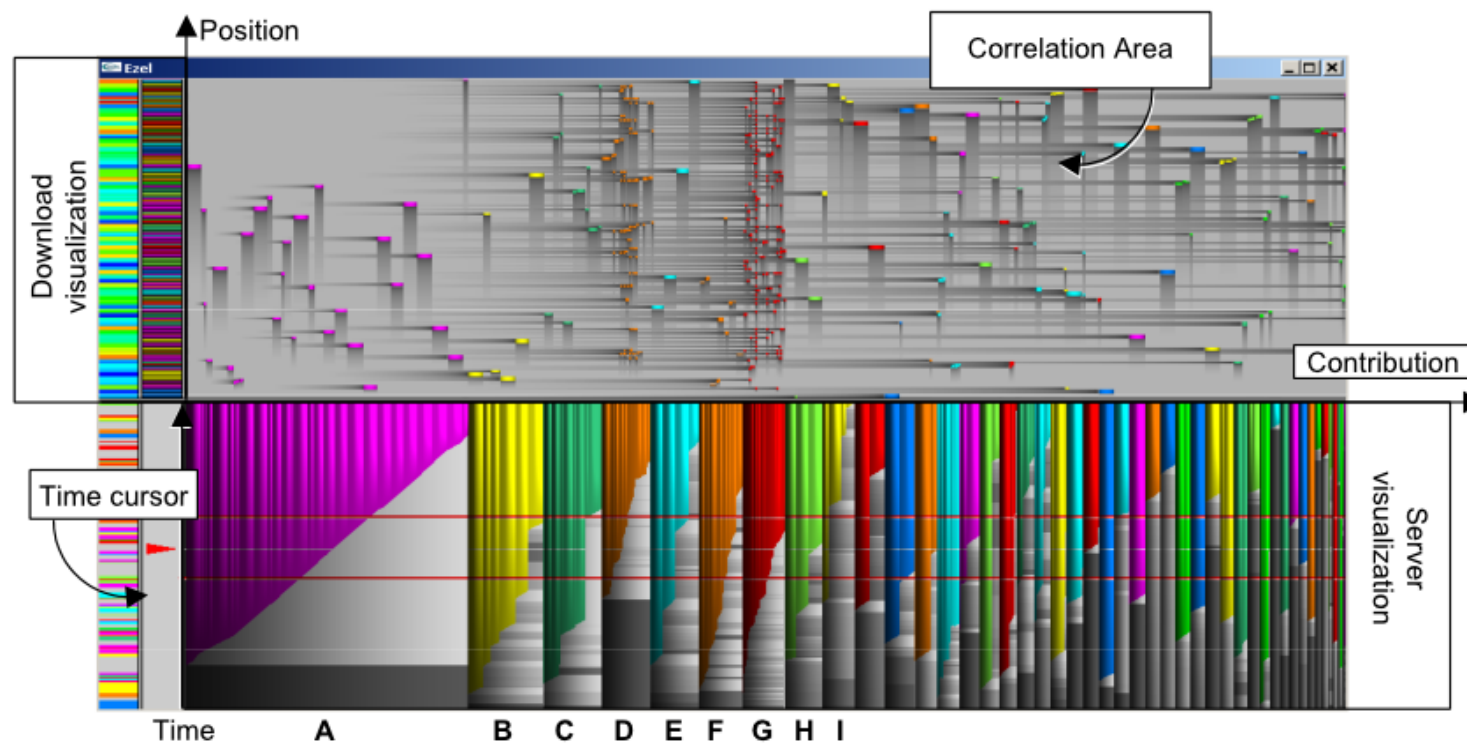
Tabletops for Visualization
University of Calgary



WILD Wall, INRIA

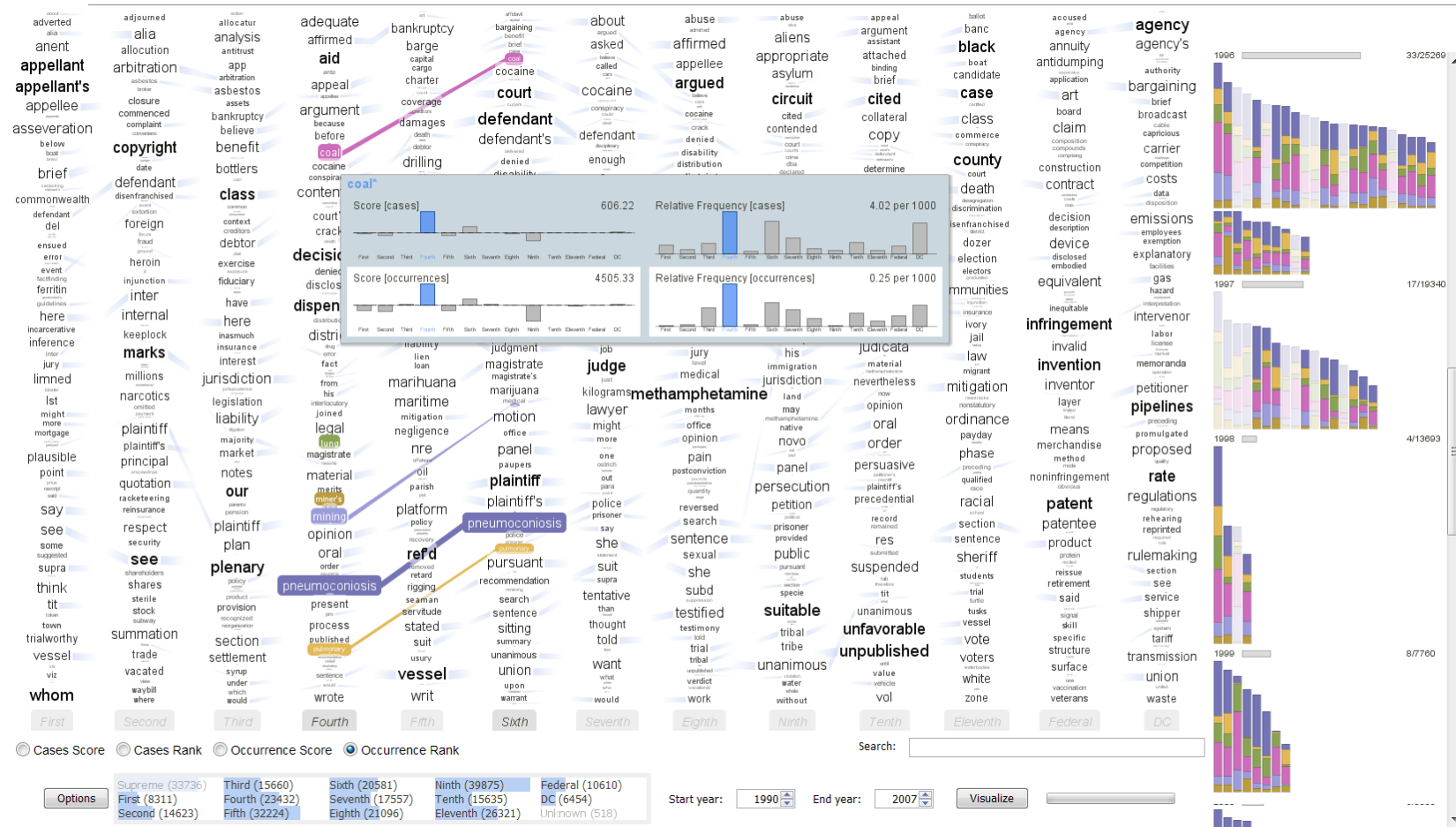
Software Visualization

EZEL: a Visual Tool for Performance Assessment of Peer-to-Peer File-Sharing Networks (Voinea et al., InfoVis, 2004)



Text Visualization

Parallel Tag Clouds to Explore Faceted Text Corpora (Collins et al., VAST 2009)



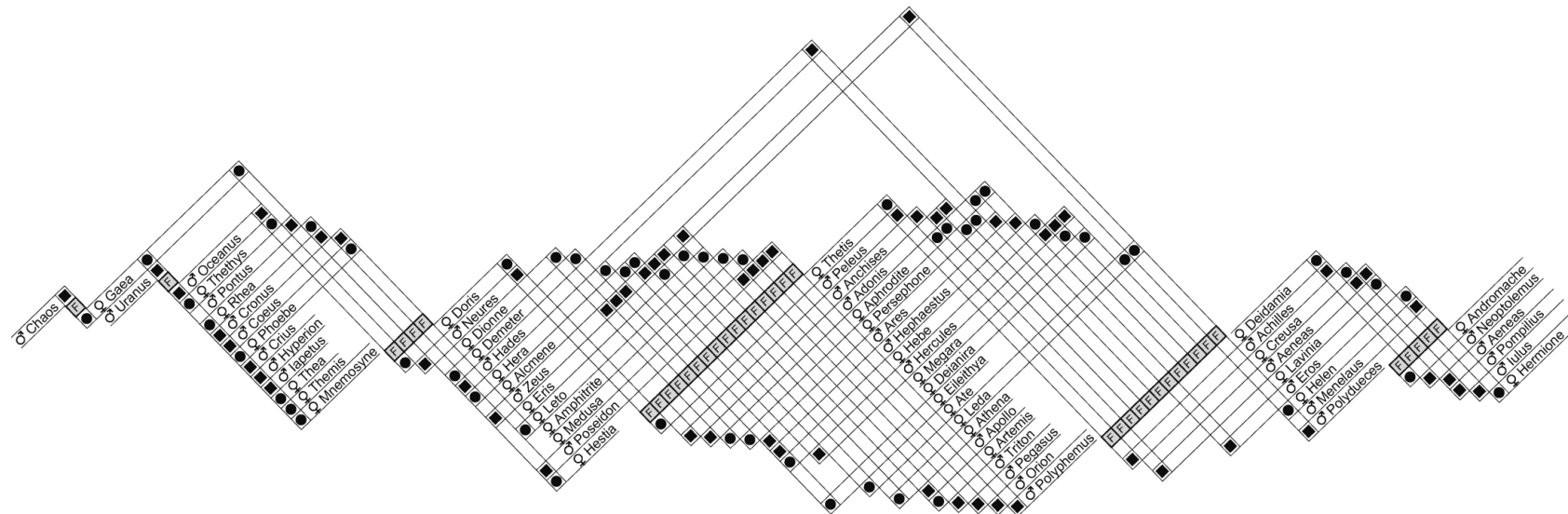
Graphs



http://www.facebook.com/note.php?note_id=469716398919

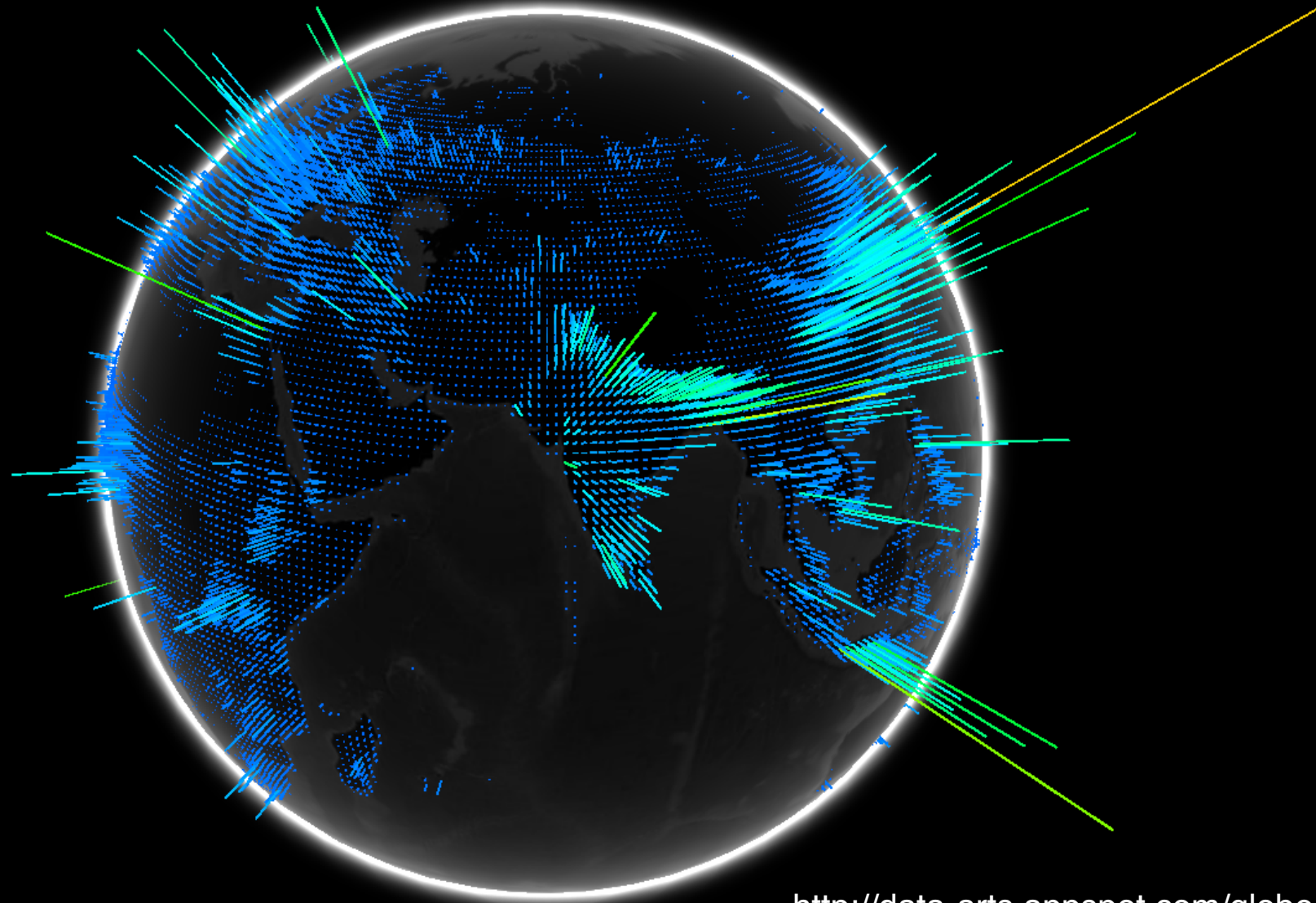
Visualizing Friendships by [Paul Butler](#) on Tuesday, December 14, 2010

Family Trees

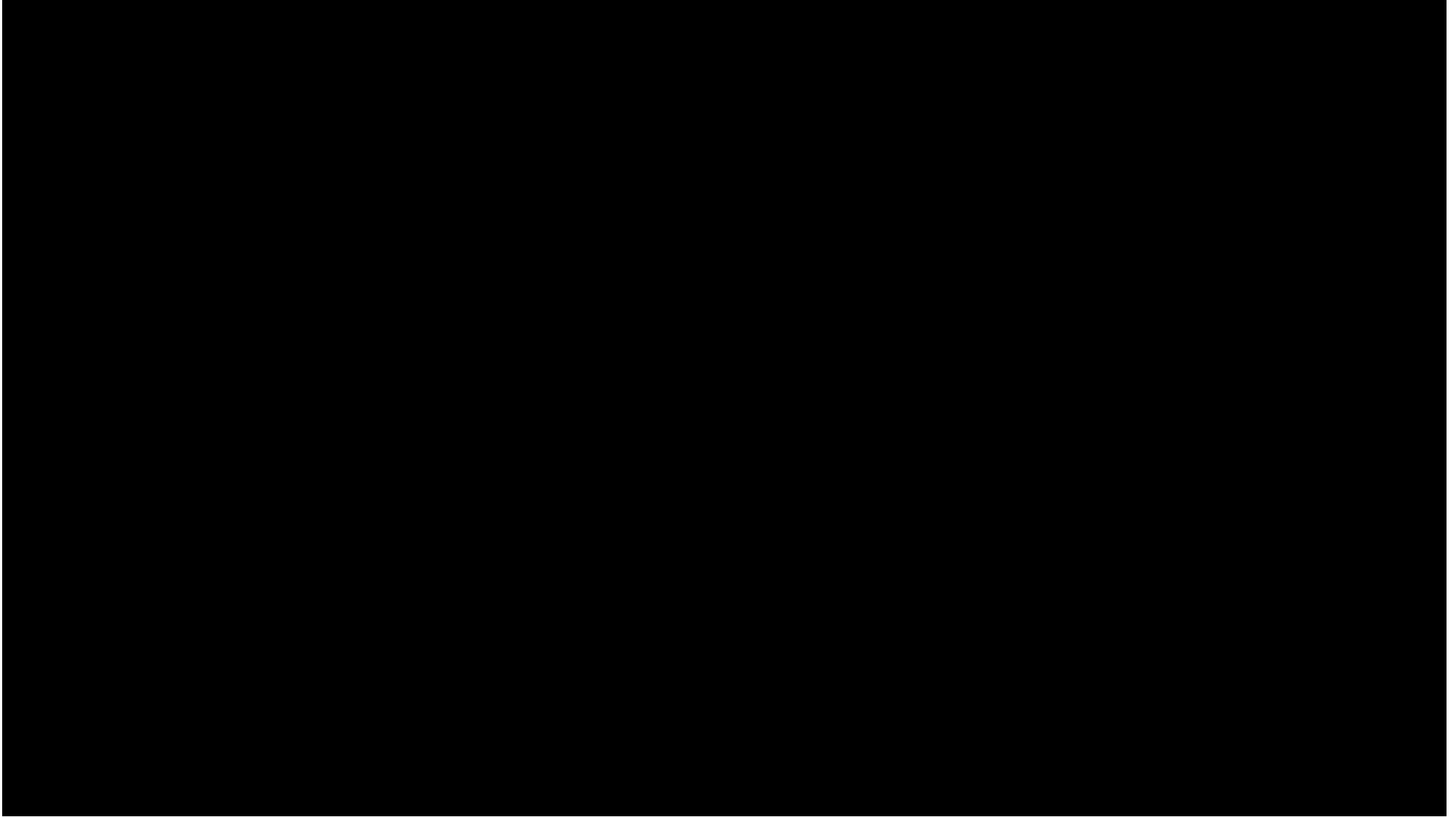


<http://www.aviz.fr/geneaquilts/>

Geographic Visualization



<http://data-arts.appspot.com/globe>



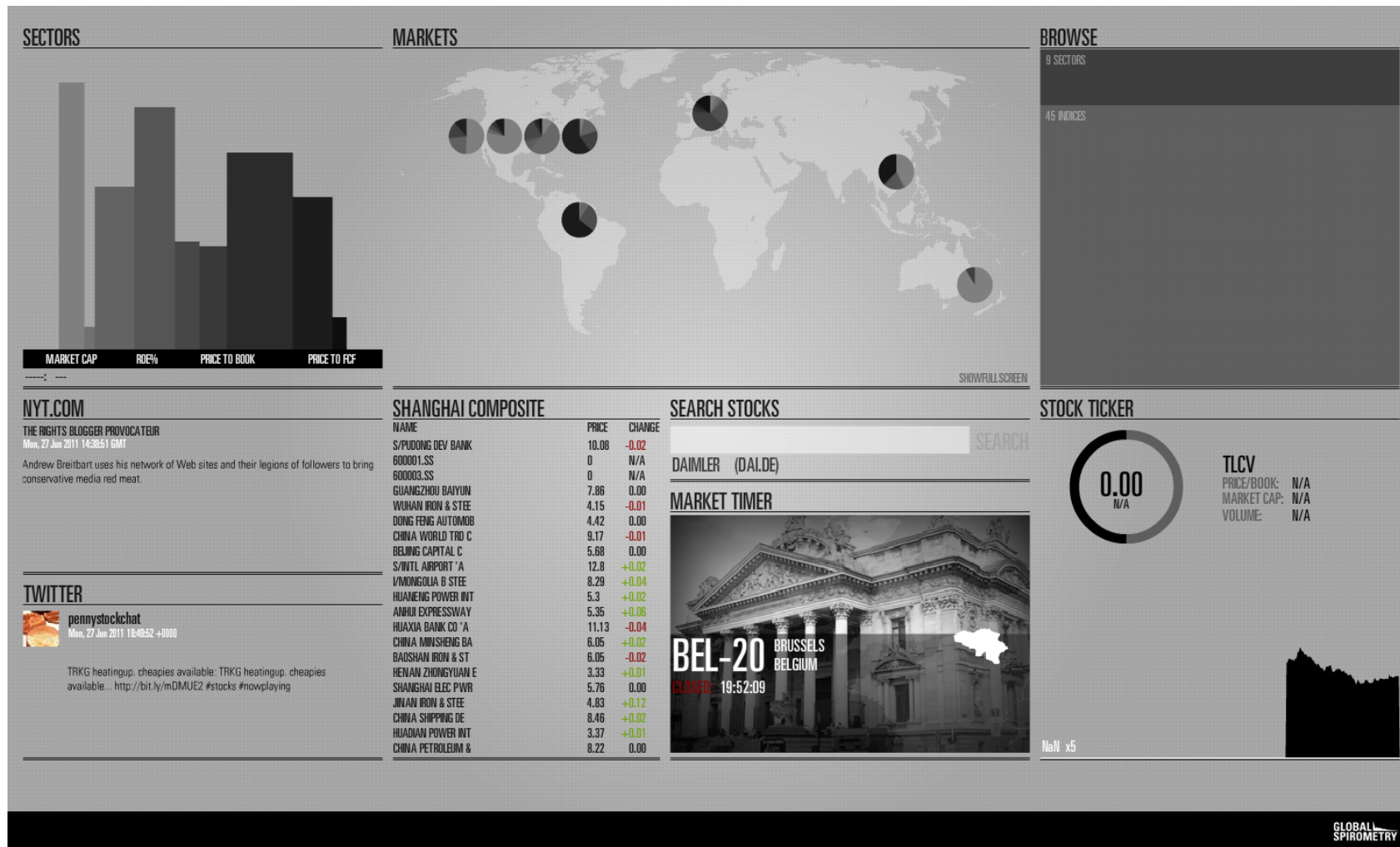
http://players.brightcove.net/679256133001/NkgrDczuol_default/index.html?videoId=5373956201001

Weather



<http://weatherspark.com/>

Data Dashboards



<http://globalspirometry.com>

Resources for more examples

Visualization conferences

Blogs

<http://infosthetics.com/>

<http://felinlovewithdata.com/>

<http://eagereyes.org/>

<http://flowingdata.com/>

<http://www.informationisbeautiful.net/>

Books

Textbooks

Readings in Information Visualization: Using Vision to Think (a bit old now but good intro)

Information Visualization (Robert Spence – a light intro, I recommend as a start)

Information Visualization Perception for Design (Colin Ware, focused on perception and cognition)

Interactive Data Visualization: Foundations, Techniques, and Applications (Ward et al.)

Visualization Analysis and Design (Tamara Munzner, most recent book)

Examples

Beautiful Data (McCandless)

Dear Data (Lupi, Posavec)

Now You See it (Few)

Tufte Books: Visual Display of Quantitative Information (and others)

... (many more, ask me for details)

It is difficult to create

CREATE VISUALIZATIONS

GOOD



What is a representation?

A representation is

a formal system or mapping by which the information can be specified (D. Marr)
a sign system in that it stands for something other than its self.

for example: the number thirty-four

34
decimal

100010
binary

XXXIV
roman

Presentation

different representations reveal different aspects of the information

- decimal: counting & information about powers of 10,
- binary: counting & information about powers of 2,
- roman: impress your friends (outperformed by positional system)

presentation

how the representation is placed or organized on the screen

34, **34**, 34

Principles of Graphical Excellence

Well-designed presentation of interesting data

– a matter of *substance, statistics, design*

Complex ideas communicated with clarity, precision, efficiency

Gives the viewer the greatest number of ideas in the shortest time
with the least ink in the smallest space

Involves almost always multiple variables

Tell the truth about the data

Or a bit more simply...

Solving a problem simply means representing it
so as to make the solution transparent ... (Simon, 1981)

Good representations:

- allow people to find relevant information

 - information may be present but hard to find

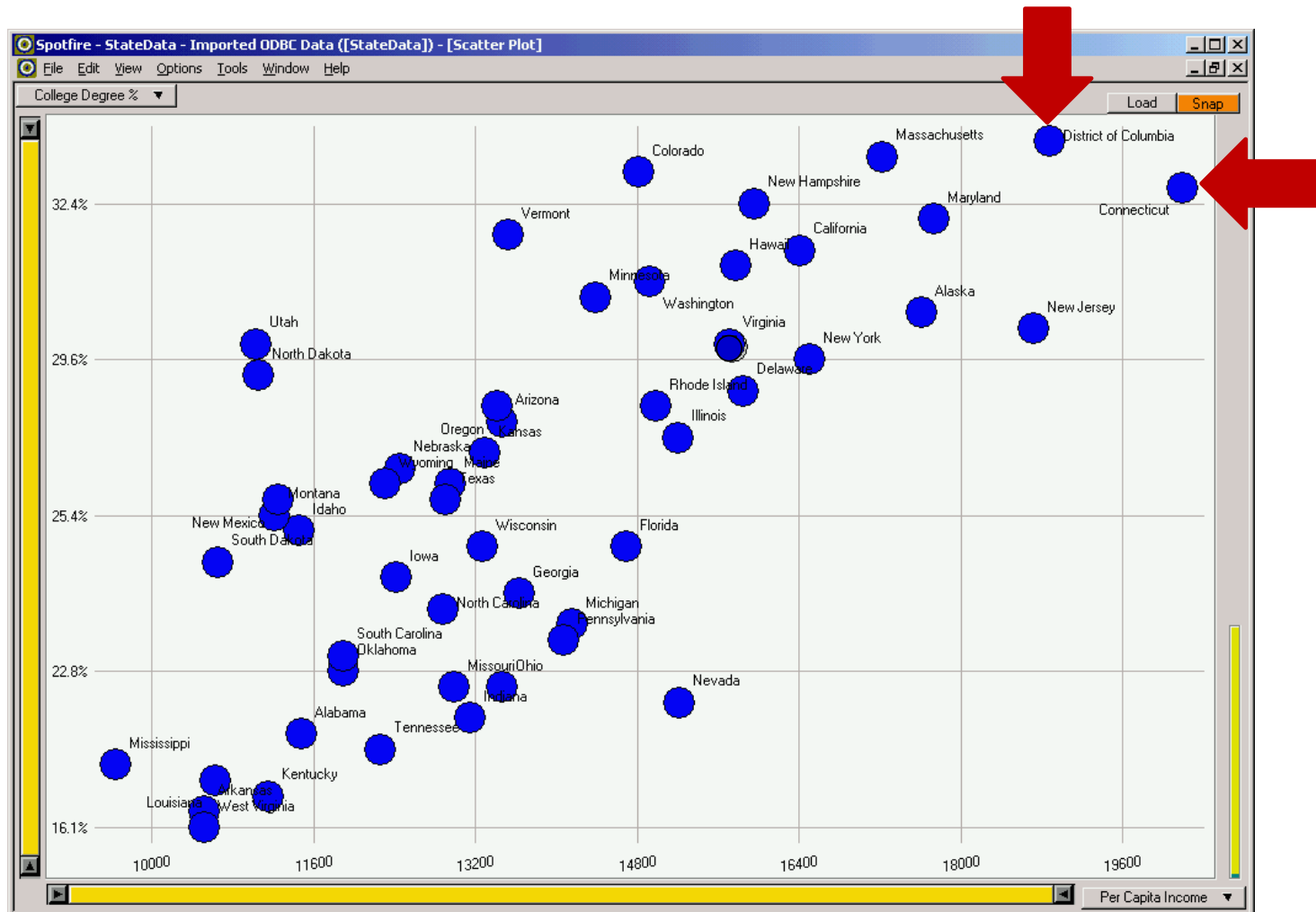
- allow people to compute desired conclusions

 - computations may be difficult or “for free” depending on representations

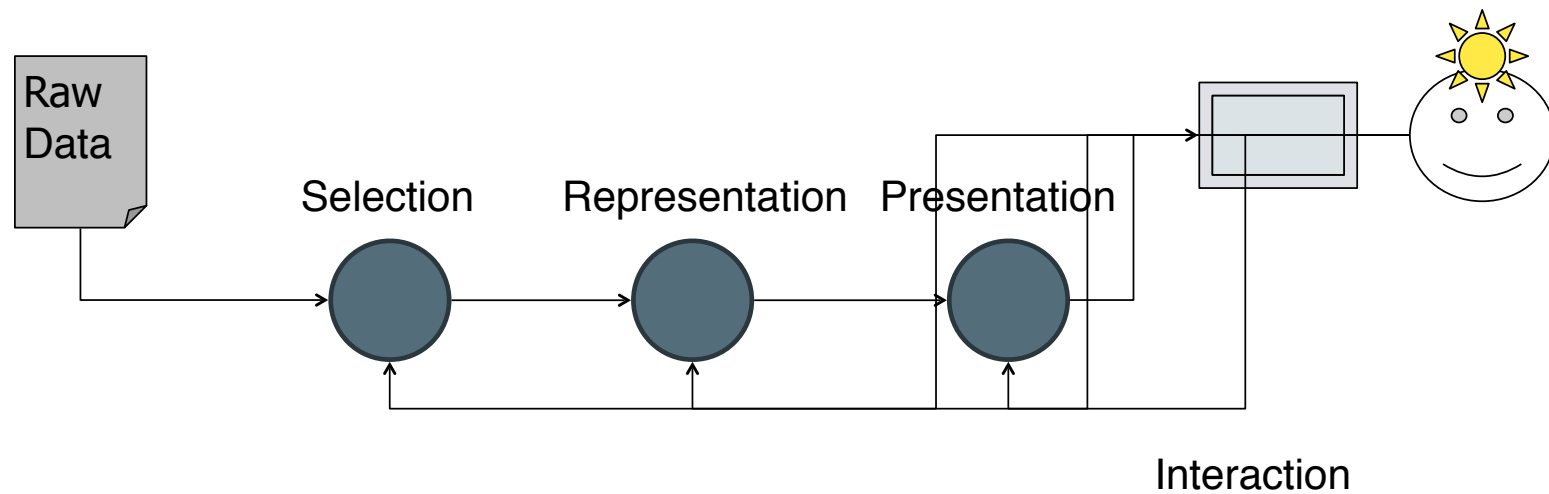
Good representation?

Table - StateData ()			Load	Snap
State	College Degree %	Per Capita Income		
Alabama	20.6%	11486		
Alaska	30.3%	17610		
Arizona	27.1%	13461		
Arkansas	17.0%	10520		
California	31.3%	16409		
Colorado	33.9%	14821		
Connecticut	33.8%	20189		
Delaware	27.9%	15854		
District of Columbia	36.4%	18881		
Florida	24.9%	14698		
Georgia	24.3%	13631		
Hawaii	31.2%	15770		
Idaho	25.2%	11457		
Illinois	26.8%	15201		
Indiana	20.9%	13149		
Iowa	24.5%	12422		
Kansas	26.5%	13300		
Kentucky	17.7%	11153		
Louisiana	19.4%	10635		
Maine	25.7%	12957		
Maryland	31.7%	17730		
Massachusetts	34.5%	17224		
Michigan	24.1%	14154		
Minnesota	30.4%	14389		
Mississippi	19.9%	9648		
Missouri	22.3%	12989		
Montana	25.4%	11213		
Nebraska	26.0%	12452		
Nevada	21.5%	15214		
New Hampshire	32.4%	15959		
New Jersey	30.1%	18714		
New Mexico	25.5%	11246		
New York	29.6%	16501		
North Carolina	24.2%	12885		
North Dakota	28.1%	11051		
Ohio	22.3%	13461		
Oklahoma	22.8%	11893		
Oregon	27.5%	13418		
Pennsylvania	23.2%	14068		
Rhode Island	27.5%	14981		
South Carolina	23.0%	11897		
South Dakota	24.6%	10661		
Tennessee	20.1%	12255		
Texas	25.5%	12904		
Utah	30.0%	11029		
Vermont	31.5%	13527		
Virginia	30.0%	15713		
Washington	30.9%	14923		
West Virginia	16.1%	10520		
Wisconsin	24.9%	13276		
Wyoming	25.7%	12311		

Good representation!



How do we arrive at a visualization?

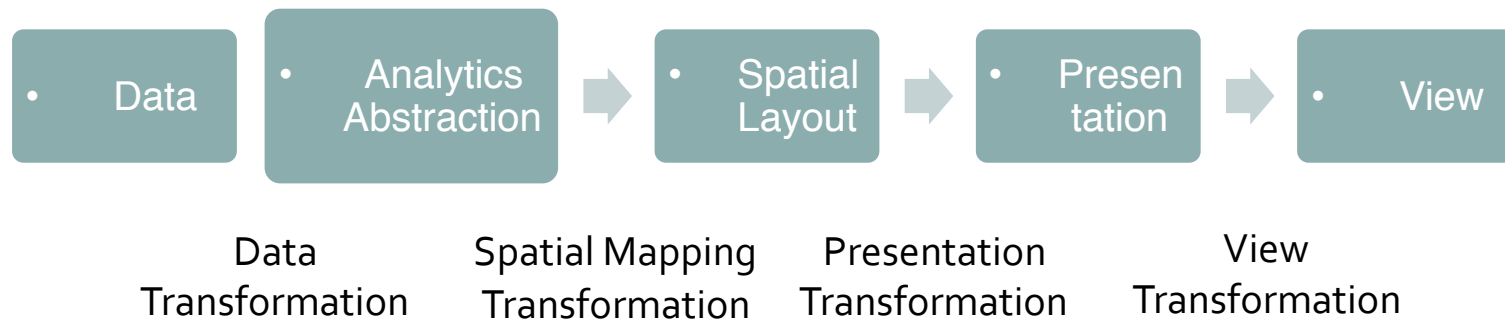


The Visualization Pipeline

From [Spence, 2000]

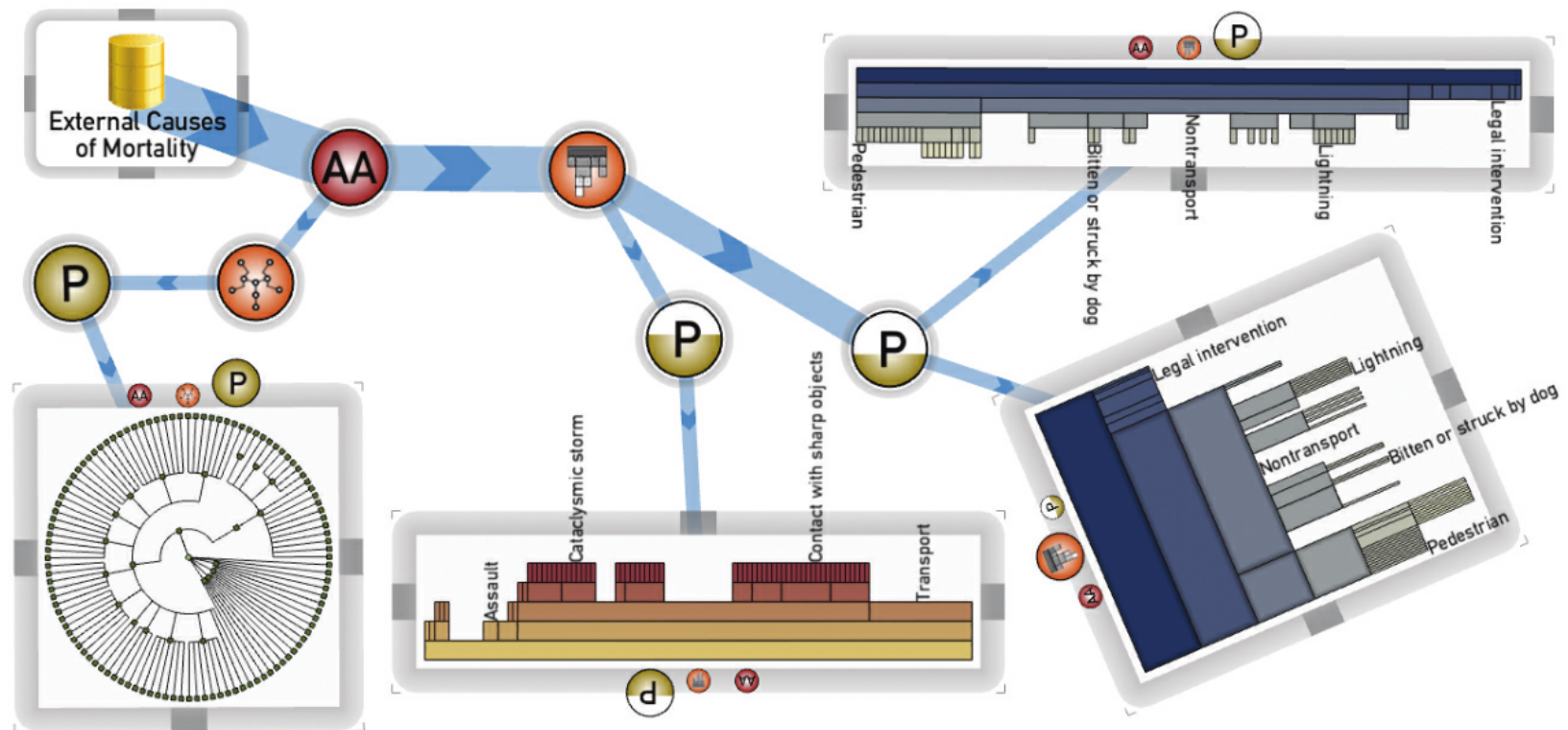
Visualization Reference Model

Also a visualization pipeline a bit expanded



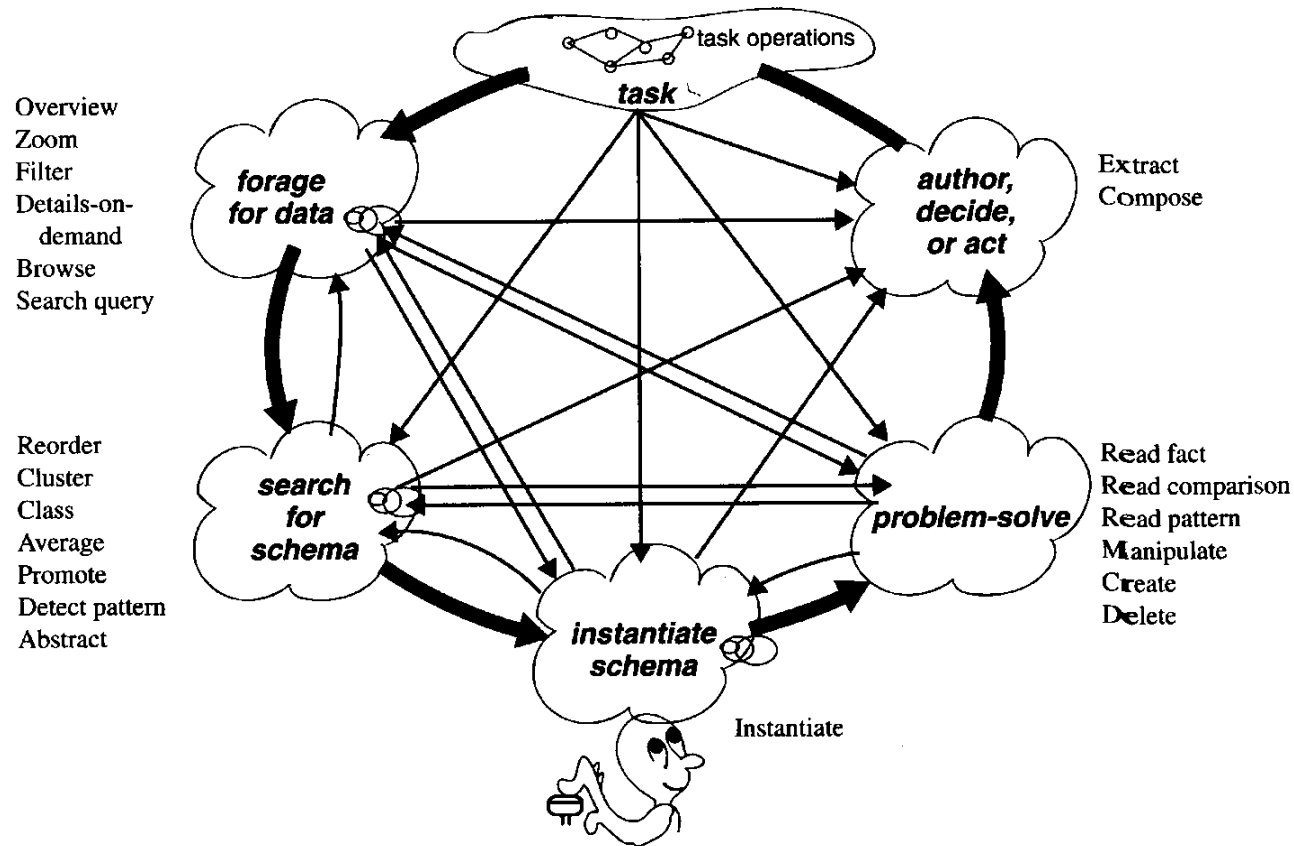
From [Card et al., Readings in Information Visualization]

Visualization pipeline in an image



[Tobiasz et al., 2009]

Knowledge Crystallization Cycle



Working with visualizations in NOT a linear process

[Card et al., 1999]

Pitfalls

- Selecting the wrong data
- Selecting the wrong data structure
- Filtering out important data
- Failed understanding of the types of things that need to be shown
- Choosing the wrong representation
- Choosing the wrong presentation format
- Inappropriate interactions provided to explore the data

Recap

- So far you
 - learned what information visualization is
 - learned about the advantages of visualization
 - saw a number of examples (historical and new)
- Next
 - you will get to know your data
 - you will learn about the basic components of visualization

Data

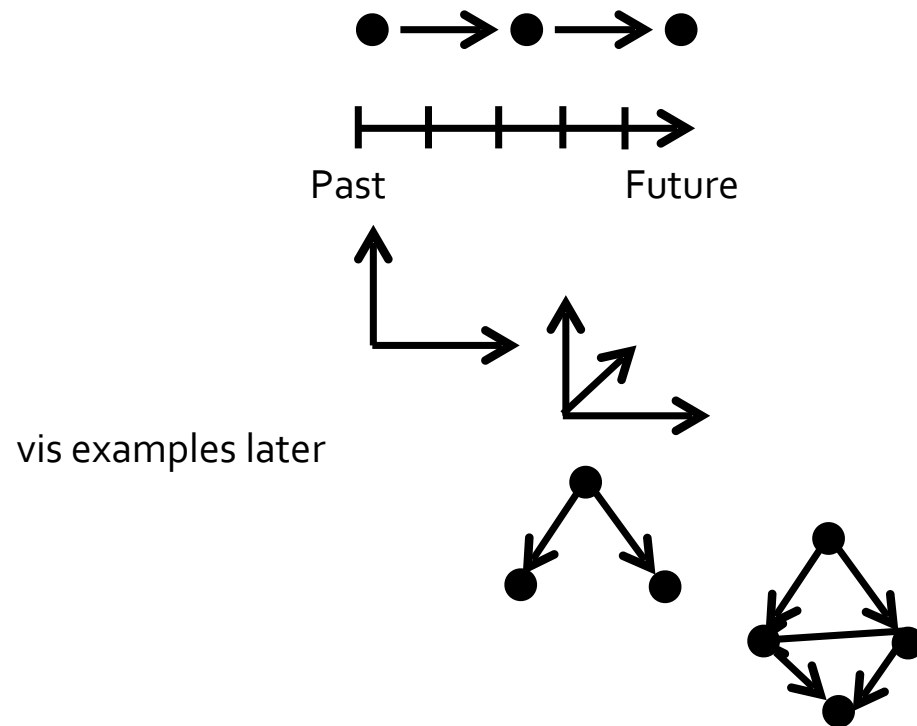
- Data is the foundation of any visualization
- The visualization designer needs to understand
 - the data properties
 - know what meta-data is available
 - know what people want from the data

Nominal, Ordinal and Quantitative

- Nominal / Categorical (labels)
 - Fruits: apples, oranges
- Ordered
 - Quality of meat: grade A, AA, AAA
 - Can be counted and ordered, but not measured
- Quantitative
 - Intervals or Ratios
 - Can do arithmetic on it

Data-Type Taxonomy

- 1D (linear)
- Temporal
- 2D (maps)
- 3D
- nD (relational)
- Trees (hierarchies)
- Networks (graphs)



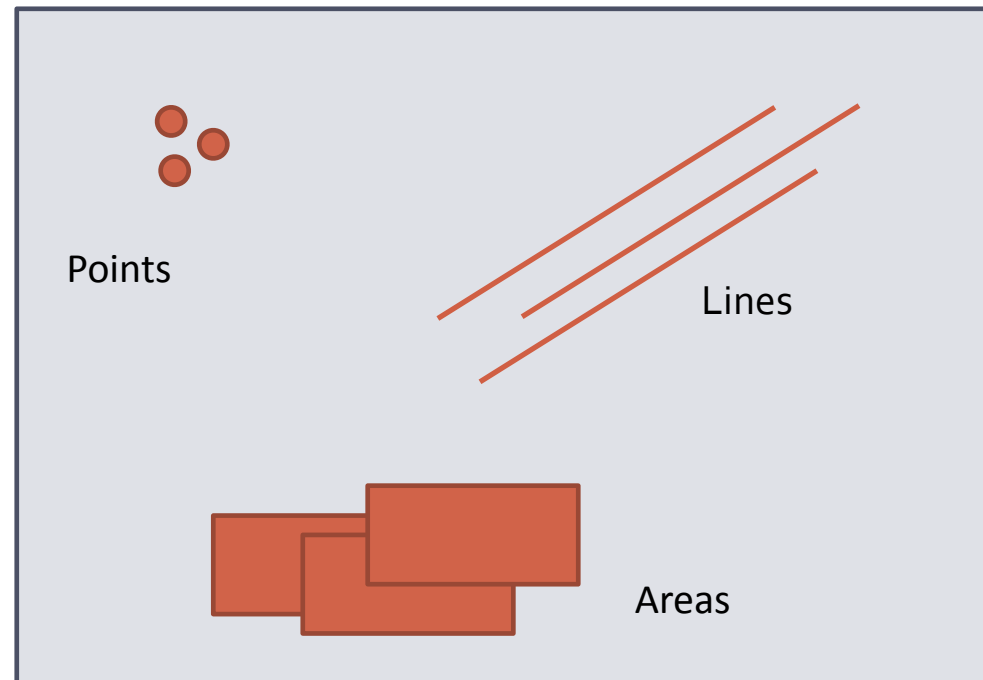
Shneiderman: The Eyes Have It

Why is this important?

- Nominal, ordinal, and quantitative data are best expressed in different ways visually
- Data types often have inherent tasks
 - temporal data (comparison of events)
 - trees (understand parent-child relationships)
 - ...
- But:
 - any data type (1D, 2D,...) can be expressed in a multitude of ways!

Visualization's Main Building Blocks

Marks which represent:

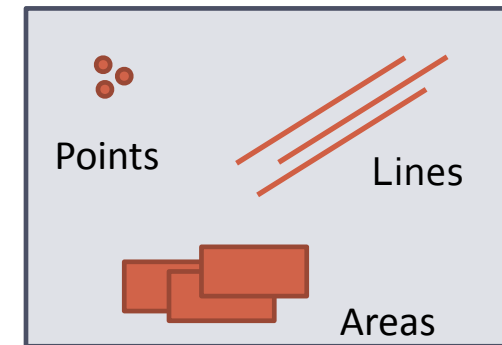


From Semiology of Graphics (Bertin)

The following slides on the topic adapted from Sheelagh Carpendale

Points

“A point represents a location on the plane that has **no theoretical length or area**. This signification is independent of the size and character of the mark which renders it visible.”

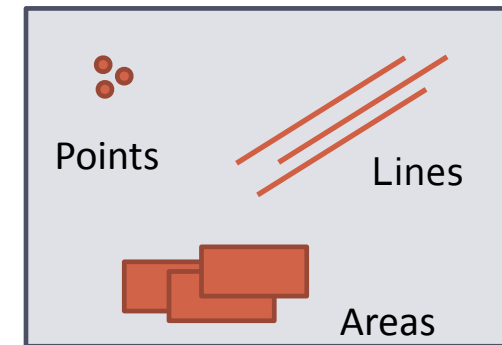


a location

marks that indicate points can vary in all visual variables

Lines

“A line signifies a phenomenon on the plane which has **measurable length but no area**. This signification is independent of the width and characteristics of the mark which renders it visible.”

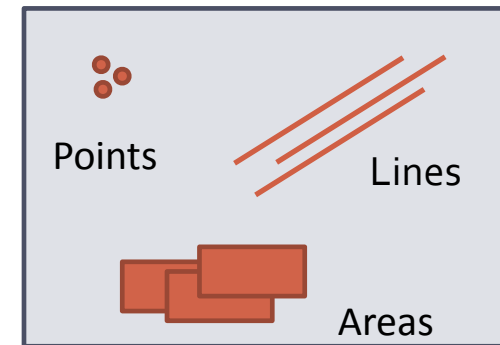


a boundary, a route, a connection

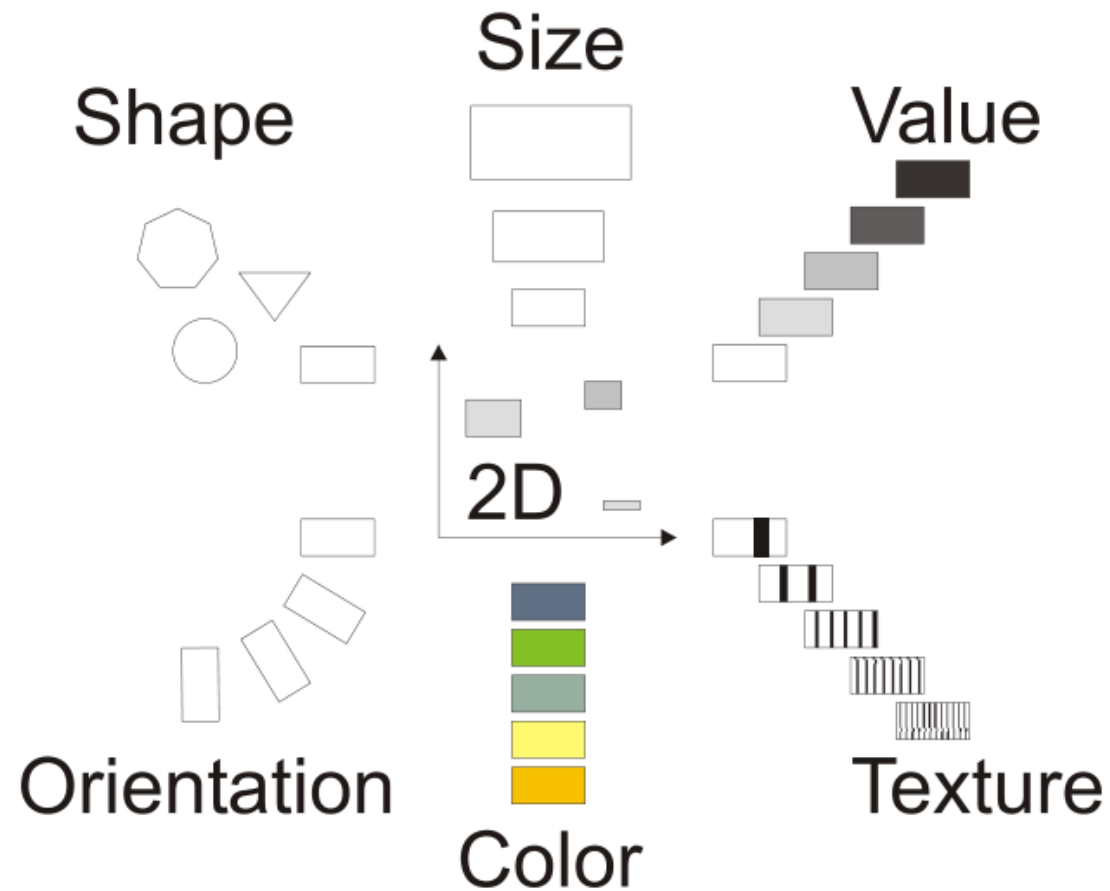
Areas

“An area signifies something on the plane that **has measurable size**.
This signification applies to the entire area covered by the visible mark.”

an area can change in position but
not in size, shape or orientation
without making the area itself have a
different meaning



Visual Variables Applicable to Marks



From Semiology of Graphics (Bertin)

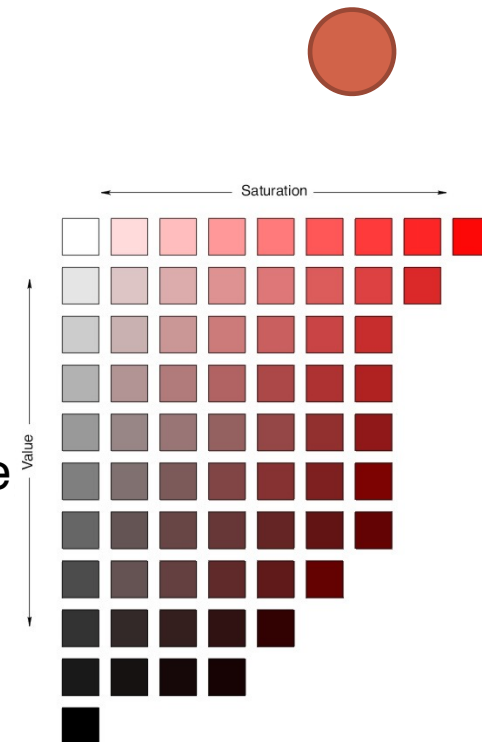
Additional Variables for Computers

- **motion**

- direction, acceleration, speed, frequency, onset, ‘personality’

- **saturation**

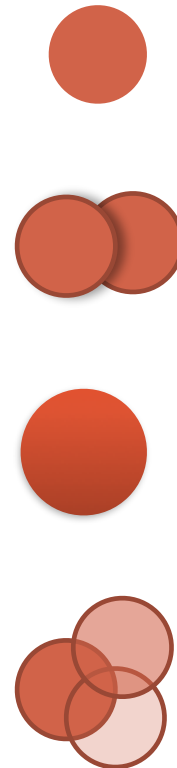
- colour as Bertin uses largely refers to hue
saturation != value



Extending those from Semiology of Graphics (Bertin)

Additional Variables for Computers

- **flicker**
 - frequency, rhythm, appearance
- **depth? ‘quasi’ 3D**
 - depth, occlusion, aerial perspective, binocular disparity
- **Illumination**
- **transparency**



From Semiology of Graphics (Bertin)

Characteristics of Visual Variables

Selective:

Can this variable allow us to spontaneously differentiate/isolate items from groups?

Associative:

Can this variable allow us to spontaneously group items in a group?

Ordered:

Can this variable allow us to spontaneously perceive an order?

Quantitative:

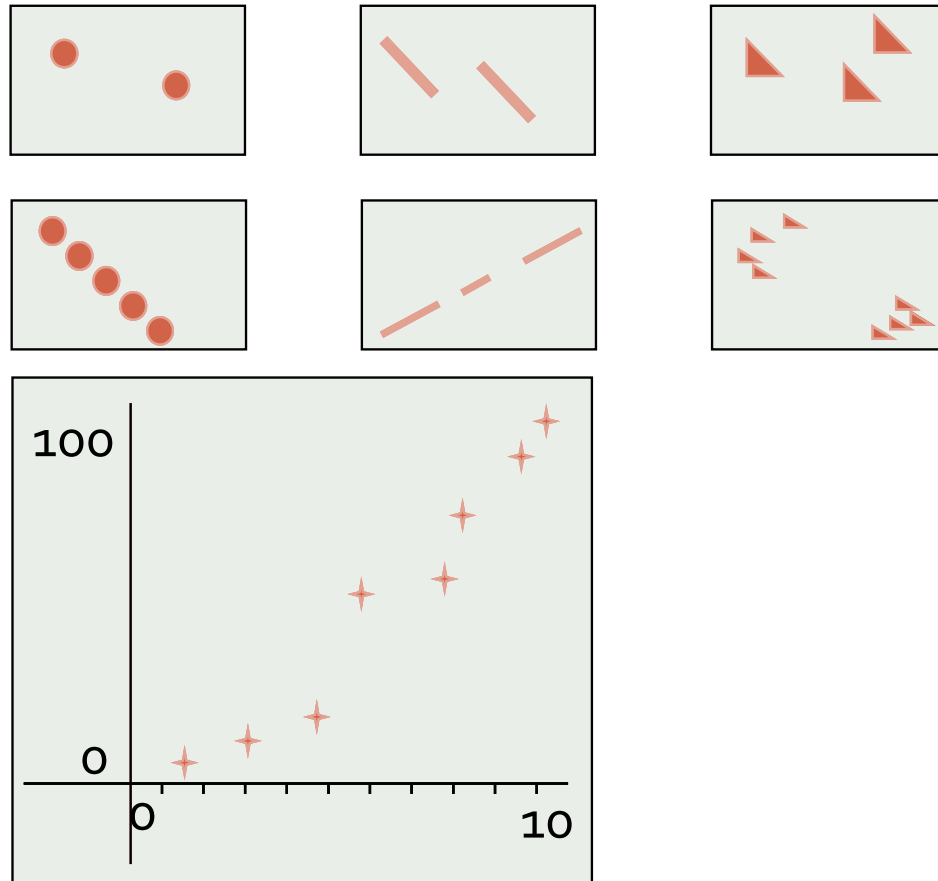
Can the difference between two marks in this variable be interpreted numerically ?

Length (resolution):

Across how many changes in this variable are distinctions possible?

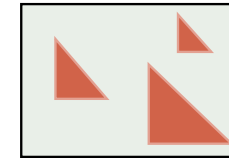
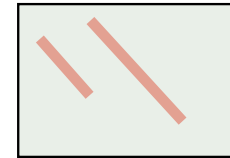
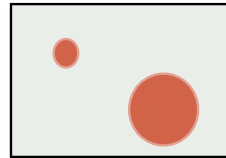
Visual Variable: Position

- ✓ • selective
- ✓ • associative
- ✓ • quantitative
- ✓ • order
- ✓ • length
(resolution)

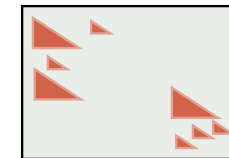
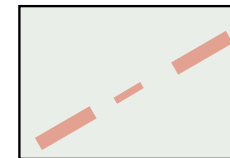
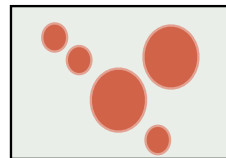


Visual Variable: Size

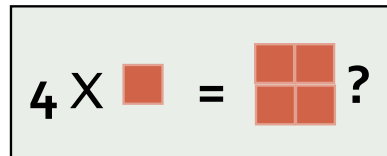
✓ • selective



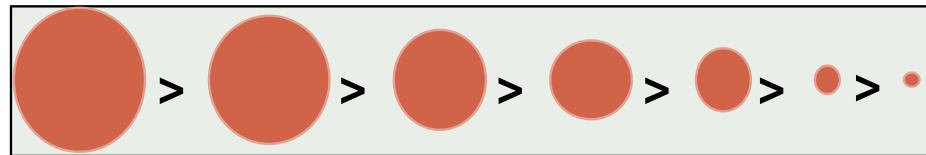
✓ • associative



✗ • quantitative



✓ • order



✓ • length
(resolution)

Size



points



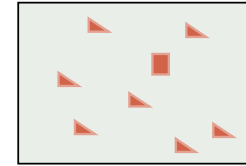
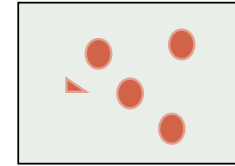
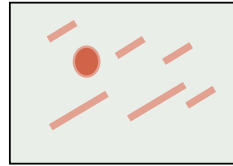
lines



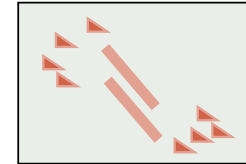
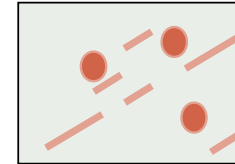
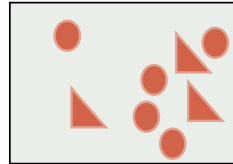
areas

Visual Variable: Shape

 • selective




 • associative

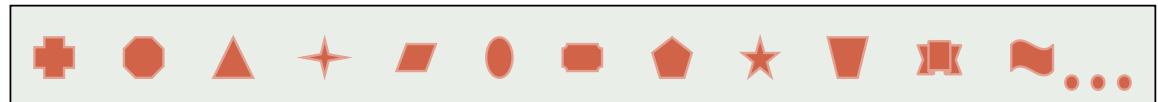


 • ordered



 • quantitative

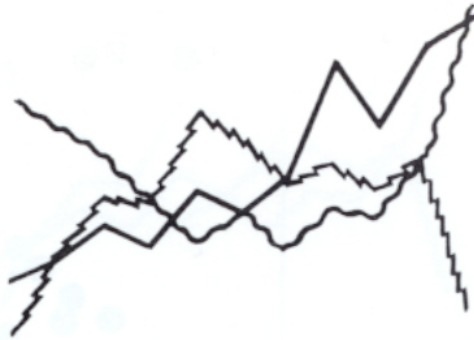
 • length
(resolution)



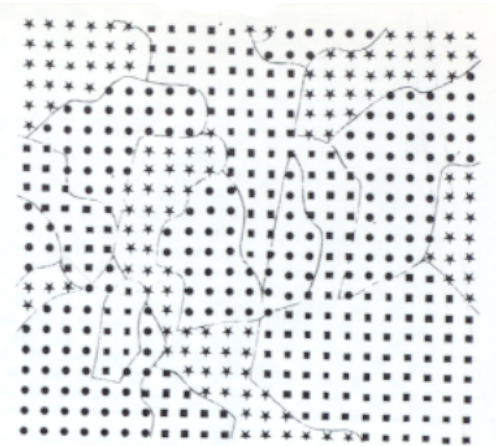
Shape



points



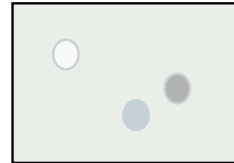
lines



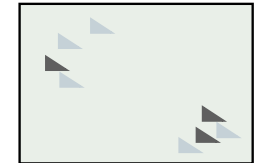
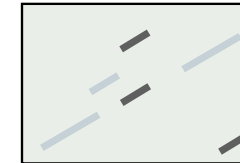
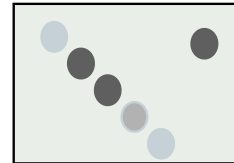
areas

Visual Variable: Value

✓ • selective

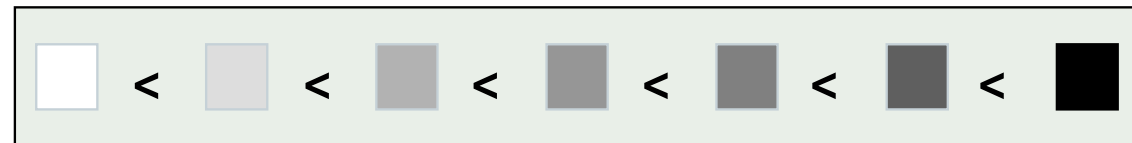


✓ • associative



≠ • quantitative

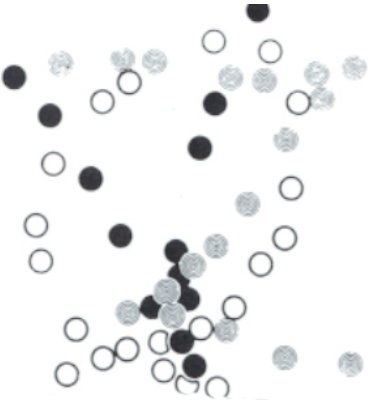
✓ • order



✓ • length (resolution)

- theoretically infinite but practically limited
- association and selection ~ < 7 and distinction ~ 10

Value



points



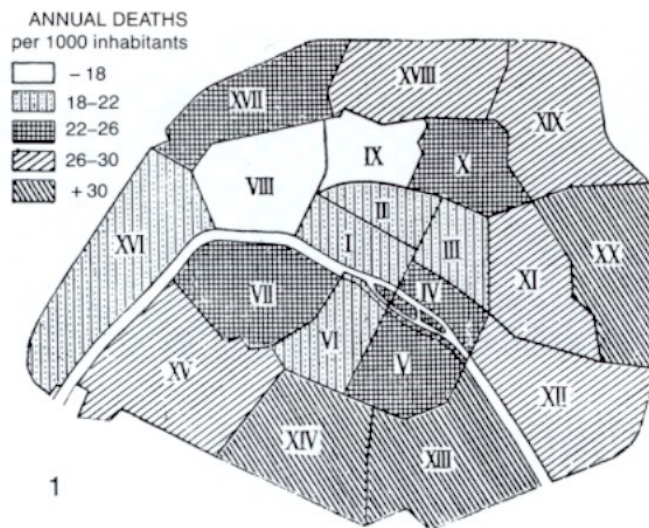
lines



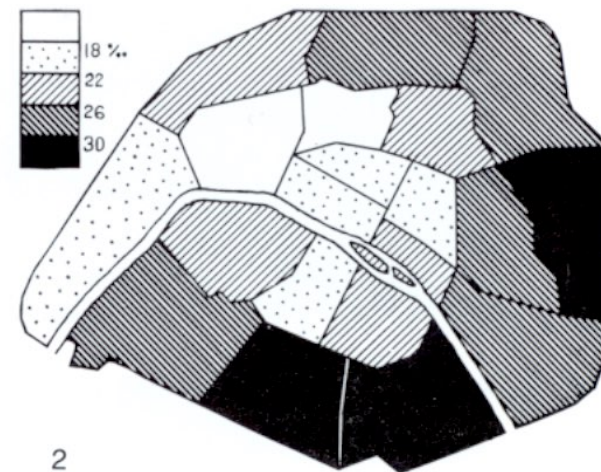
areas

Value

ordered, cannot be reordered



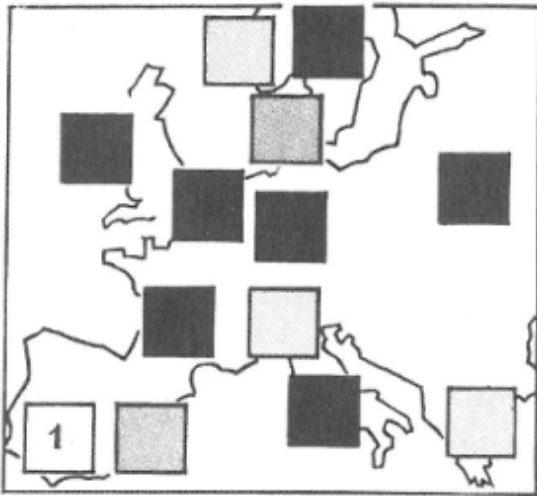
Values not ordered correctly according to scale
Information has to be read point by point



Values ordered correctly
Image much more useful

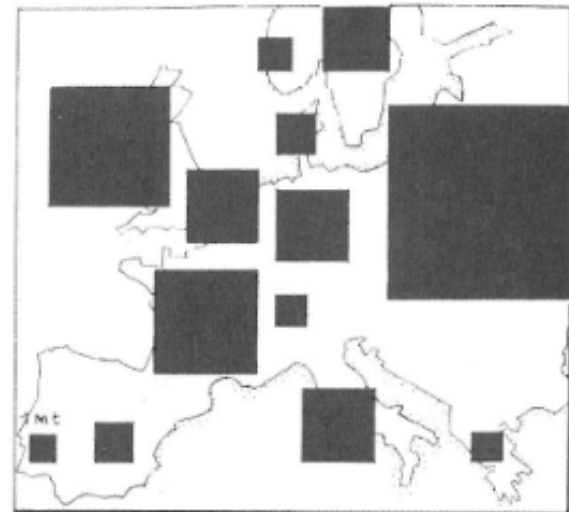
Value

is not quantitative



if Portugal is 1, what is France?

you need a legend!

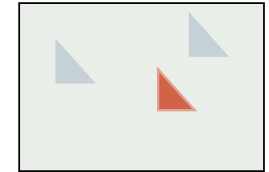
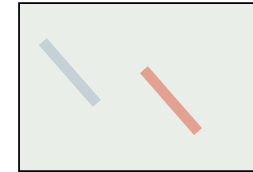
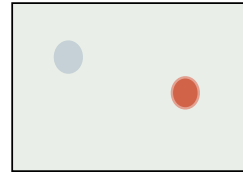


if Portugal is 1, what is France?

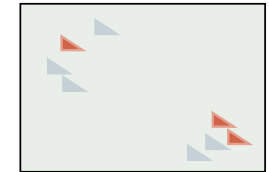
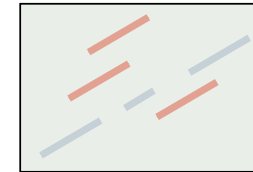
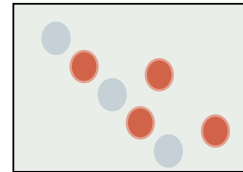
still hard, but doable

Visual Variable: Colour

✓ • selective



✓ • associative



≠ • quantitative

≠ • order



✓ • length (resolution)

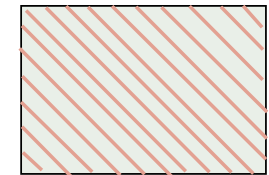
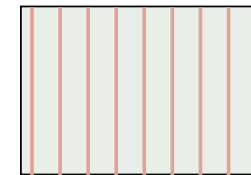
- theoretically infinite but practically limited
- association and selection ~ < 7 and distinction ~ 10

Visual Variable: Orientation

✓ • selective

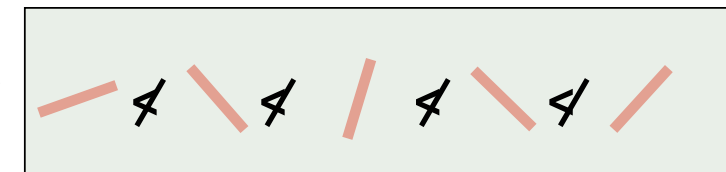
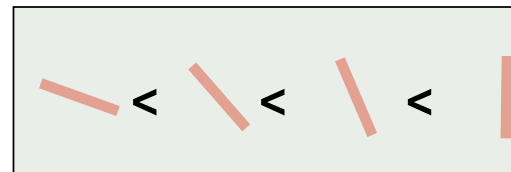


✓ • associative



✗ • quantitative

✗ • order



✓ • length (resolution)
• ~5 in 2D; ? in 3D

Orientation



points



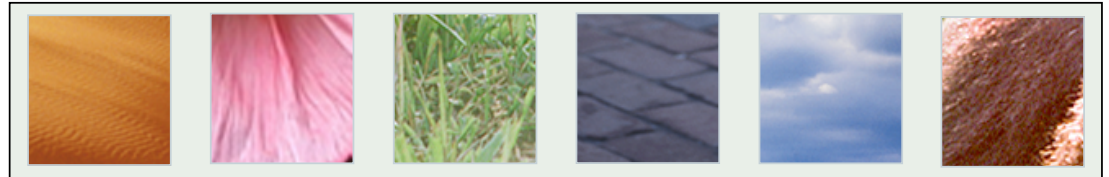
lines



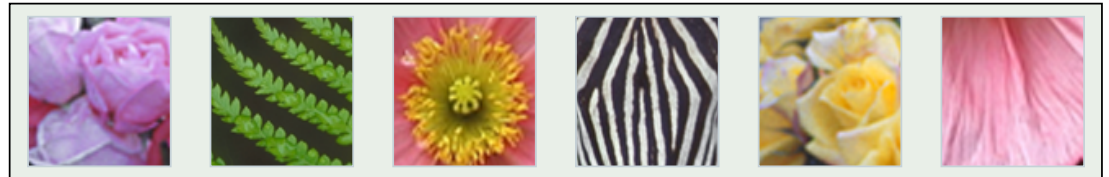
areas

Visual Variable: Texture

✓ selective

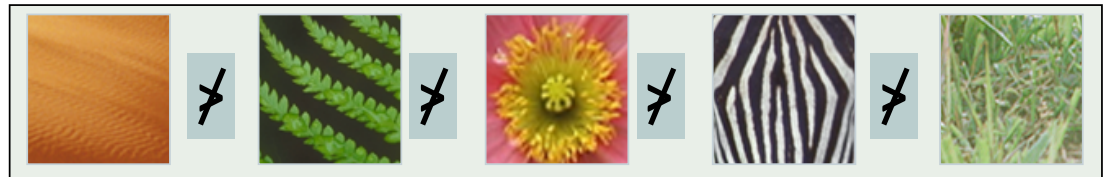


✓ associative



≠ quantitative

≠ order



✓ length
(resolution)
• theoretically infinite

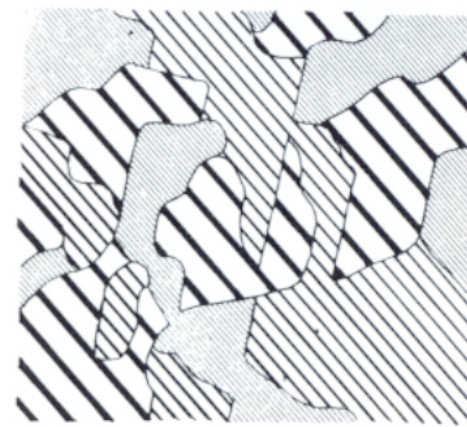
Texture



points



lines

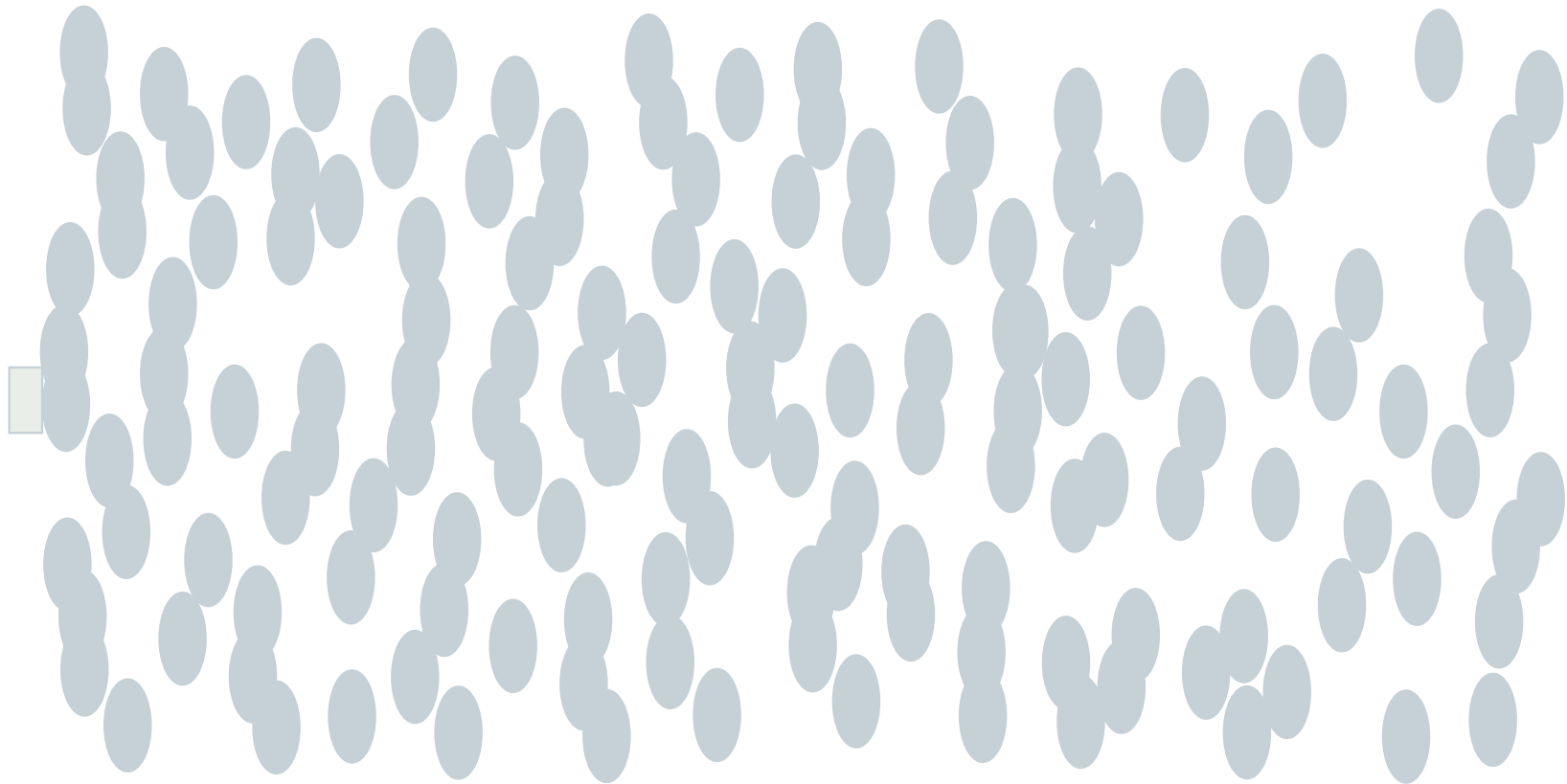


areas

Visual Variable: Motion

- ✓ **selective**
motion is one of our most powerful attention grabbers
- ✓ **associative**
moving in unison groups objects effectively
- ≠ **quantitative**
subjective perception
- ≠ **order**
- ? **length (resolution)**
distinguishable types of motion?

Motion



Visual Variables

Visual Variable	Selective	Associative	Quantitative	Order	Length
Position	Yes	Yes	Yes	Yes	Dependant on resolution
Size	Yes	Yes	Approximate	Yes	Association: 5; Distinction: 20
Shape	With Effort	With Effort	No	No	Infinite
Value	Yes	Yes	No	Yes	Association: 7; Distinction: 10
Hue	Yes	Yes	No	No	Association: 7; Distinction: 10
Orientation	Yes	Yes	No	No	4
Grain	Yes	Yes	No	No	5
Texture	Yes	Yes	No	No	Infinite
Motion	Yes	Yes	No	Yes	Unknown

Summary

More Accurate



Less Accurate

Jacques Bertin refined by Cleveland&McGill then by Card&Mackinlay



<https://www.youtube.com/watch?v=AdSZJzb-aX8>

The Art of Data Visualization | Off Book | PBS Digital Studios

Summary

- Now you know the main building blocks are **marks**
- Marks are modified by **visual variables**
- Visual variables have **specific characteristics**
- These characteristics influence how the data will be perceived