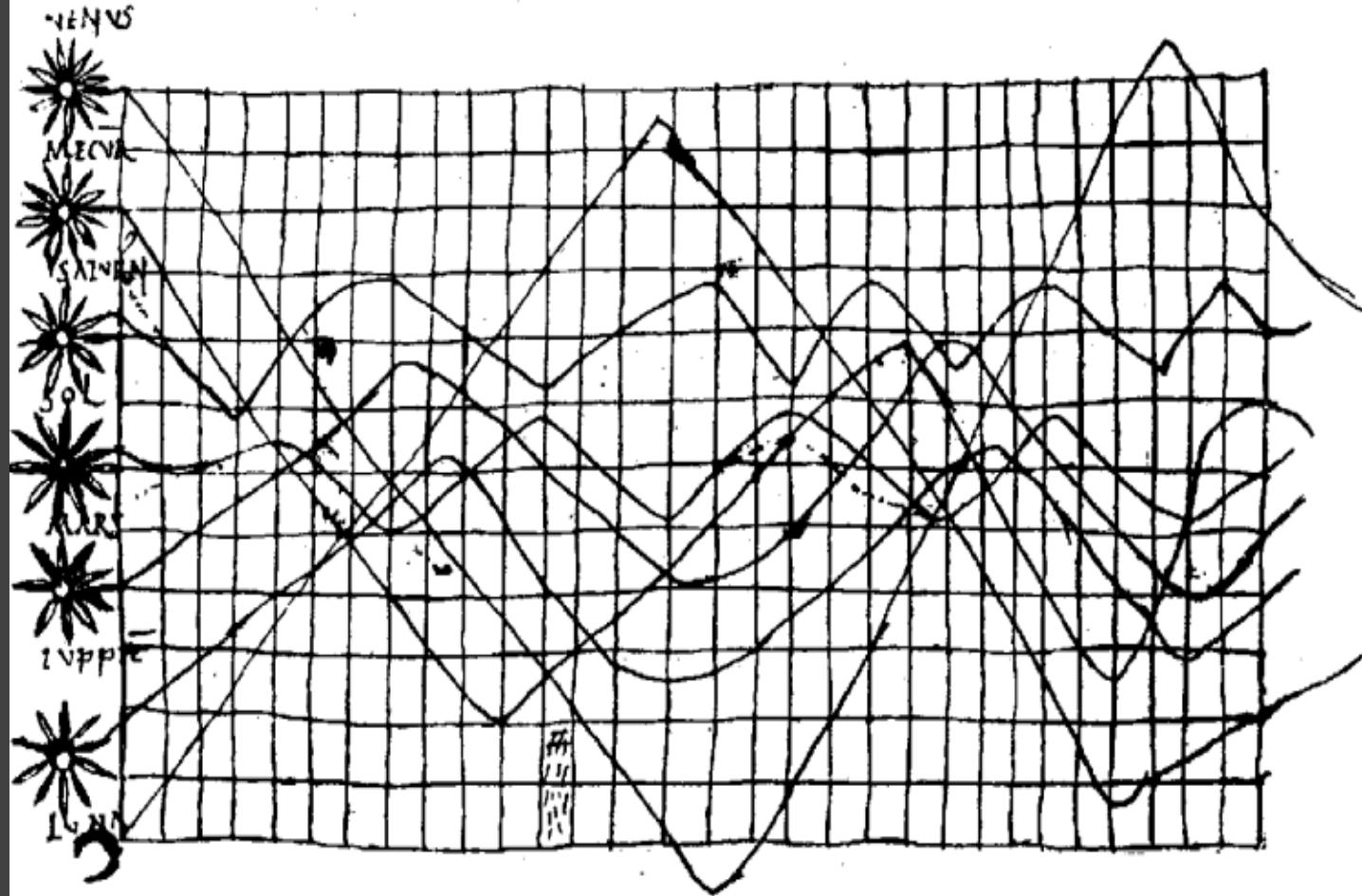


# CSE 442 - Data Visualization

# Introduction



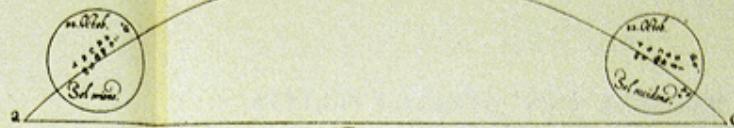
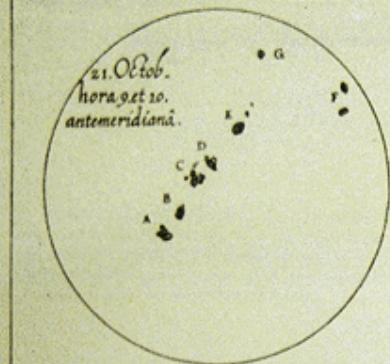
Jeffrey Heer University of Washington



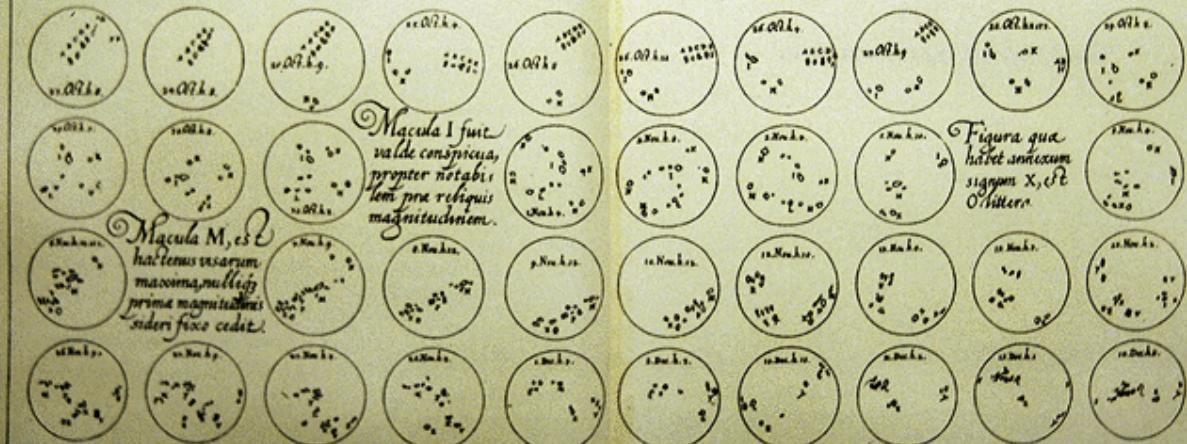
~950 AD Position of Sun, Moon and Planets

MACVLAE IN SOLE APPARENTES, OBSERVATAE  
anno 1611. ad latitudinem grad. 48. min. 40.

b



a c, horizon. a b c, arcus solis diurnus. Solariens ex parte a, maculas exhibet quas vides, occidens vero c, easdem ratione primi motus, non nihil mutat. Et hanc matutinam vespertinamque mutationem, omnes macula quotidie subeunt. Quod semel exhibuit et monuisse, sufficiat.



Alia. Non ag. modis

Compendio

Sunspots over time, Scheiner 1626

TOLEDO.

GRADOS DE LA LONGITUD.



G. Iansonius.

G. Mercator.

I. Schonera.

P. Lansbergius.

T. Brahe.

ROMA

I. Regiomontanus.

Oronius.

C. Clavius.

C. Ptolomeus.

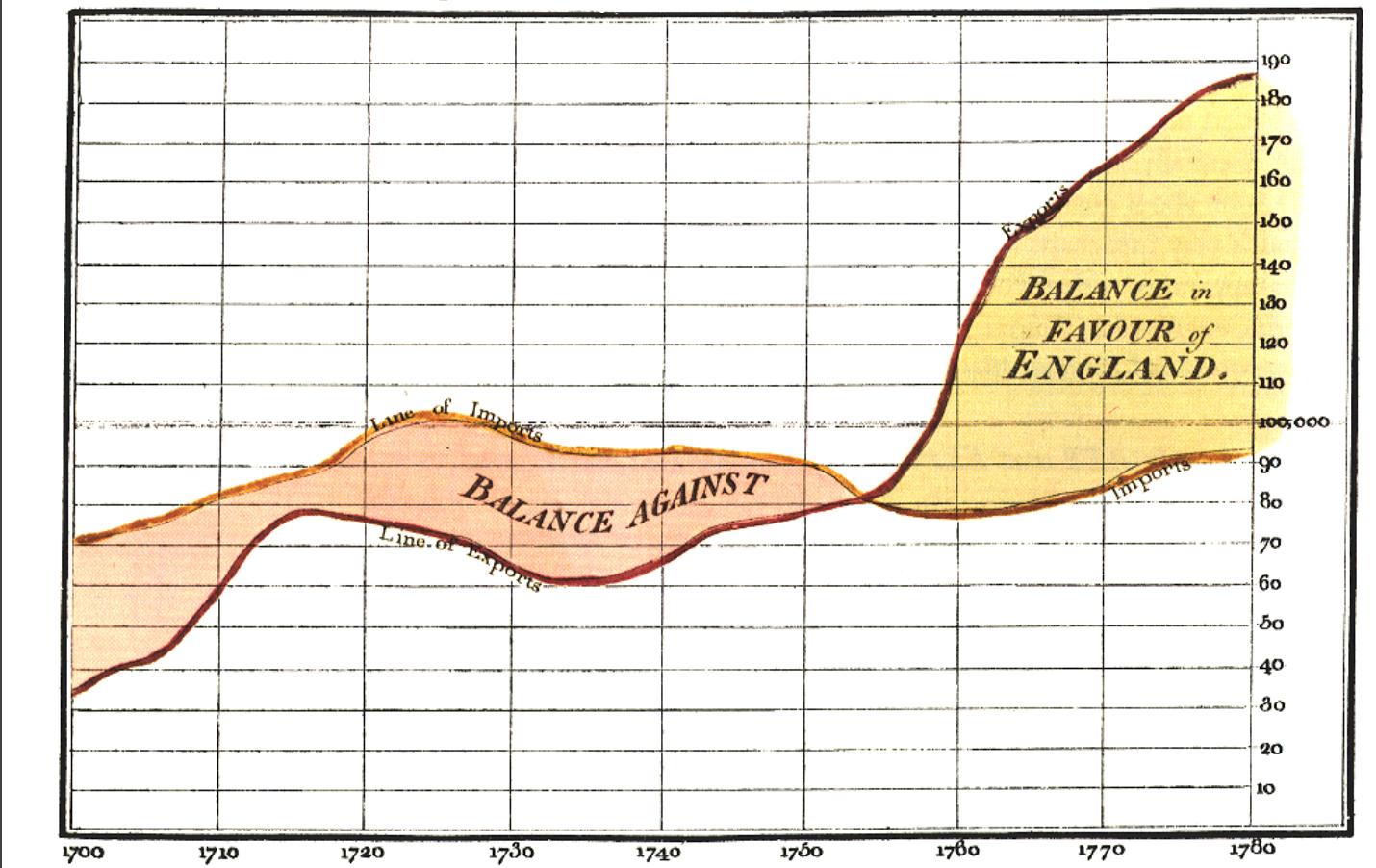
A. Argelius.

A. Maginus.

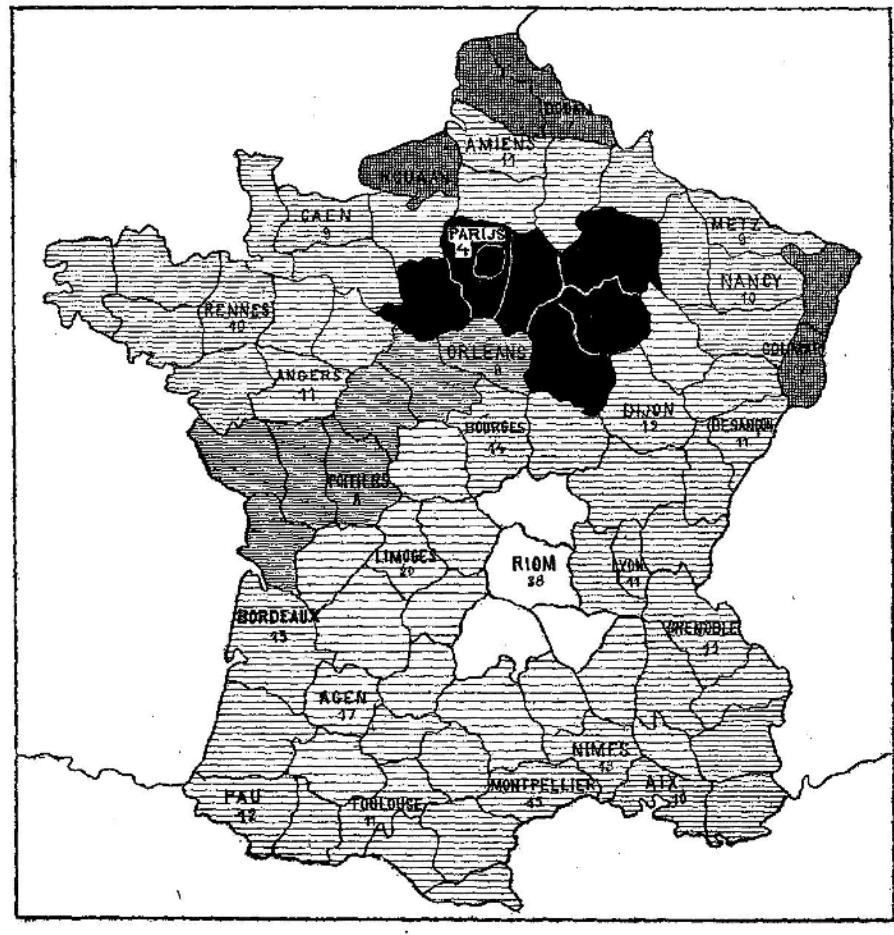
D. Origanus.

Longitudinal distance between Toledo and Rome, van Langren 1644

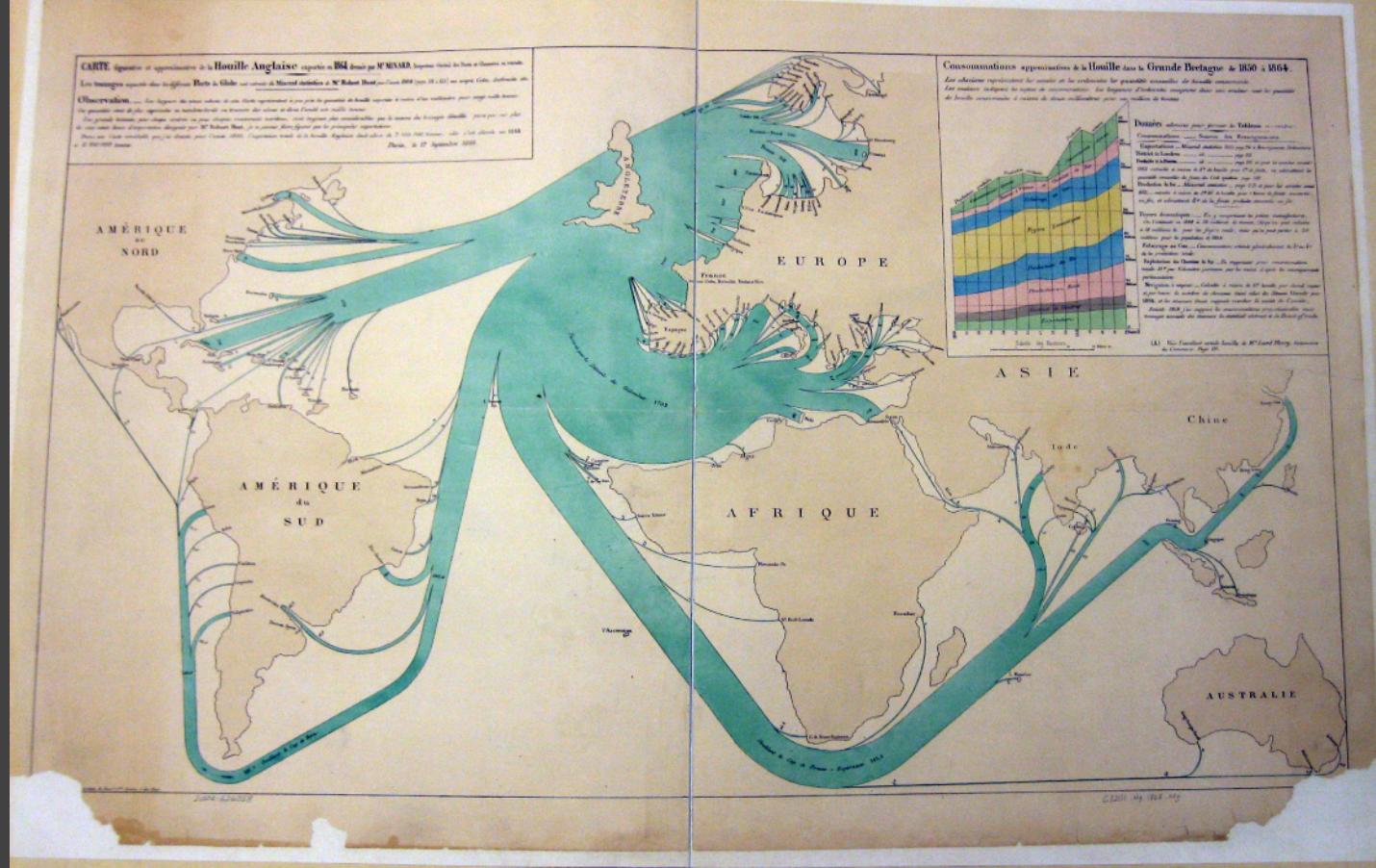
Exports and Imports to and from DENMARK & NORWAY from 1700 to 1780.



The Commercial and Political Atlas, William Playfair 1786



1826(?) Illiteracy in France, Pierre Charles Dupin

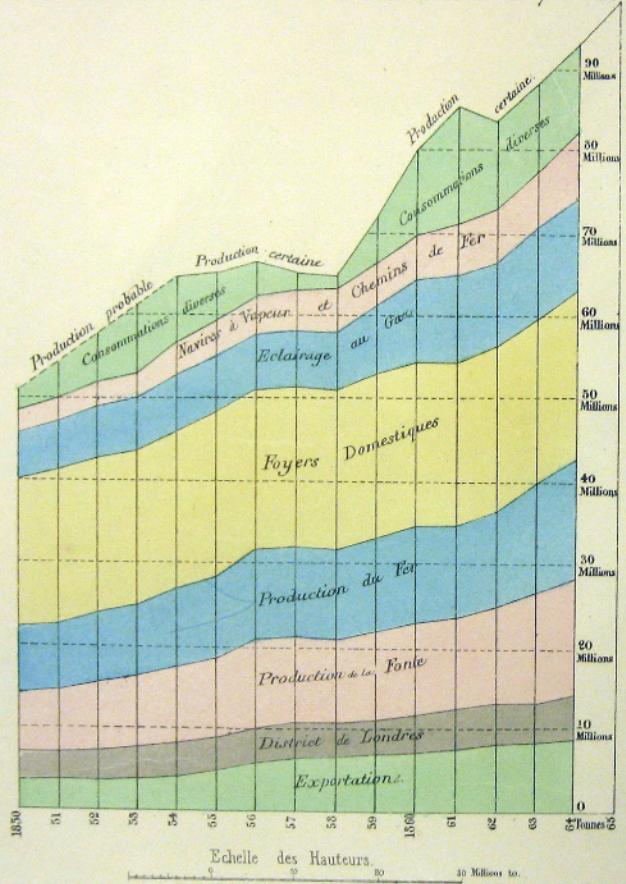


## 1864 British Coal Exports, Charles Minard

# Consommations approximatives de la Houille dans la Grande Bretagne de 1850 à 1864.

Les abscisses représentent les années et les ordonnées les quantités annuelles de houille consommée.

Les couleurs indiquent les espèces de consommations. Les longueurs d'ordonnées comprises dans une couleur sont les quantités de houille consommées à raison de deux millimètres pour un million de tonnes.



(A) Voir l'excellent article houille de M<sup>r</sup> Lamé Fleury, Dictionnaire du Commerce Page III.

Données admises pour former le Tableau ci-contre.

Consommations. — Sources des Renseignements.

Exportations. — *Mineral statistics 1865 page 214 et Renseignements Parlementaires.*  
District de Londres. — *id. page 213*

Produits de la Fonte. — *id. page 215 et pour les années avant 1855 calculé à raison de 3<sup>10</sup>/<sub>35</sub> de houille pour 1<sup>re</sup> de fonte, en admettant les quantités annuelles de fonte du Coal question page 192.*

Production du fer. — *Mineral statistics — page 215 et pour les années avant 1855 — calculé à raison de 3<sup>10</sup>/<sub>35</sub> de houille pour 1 tonne de fonte convertie en fer; et admettant 10<sup>40</sup>/<sub>41</sub> de la fonte produite convertie en fer*

Foyers domestiques: — En y comprenant les petites manufactures. On l'estimait en 1848 à 19 millions de tonnes, (A) qu'on peut réduire à 18 millions to. pour les foyers seuls, mais qu'on peut porter à 20 millions pour la population de 1864.

Eclairage au Gaz. — Consommation estimée généralement du 3<sup>10</sup>/<sub>35</sub> au 3<sup>10</sup>/<sub>36</sub> de la production totale.

Exploitation des Chemins de Fer. — En supposant pour consommation totale 10<sup>40</sup>/<sub>41</sub> par Kilomètre parcouru par les trains d'après les renseignements parlementaires.

Navigation à vapeur. — Calculée à raison de 5<sup>10</sup>/<sub>35</sub> houille par cheval vapeur et par heure, le nombre de chevaux étant celui du Steam Vessels pour 1864, et les steamers étant supposés marcher la moitié de l'année;

Avant 1864 j'ai supposé les consommations proportionnelles aux tonnages annuels des steamers du statistical abstract et du Board of trade.

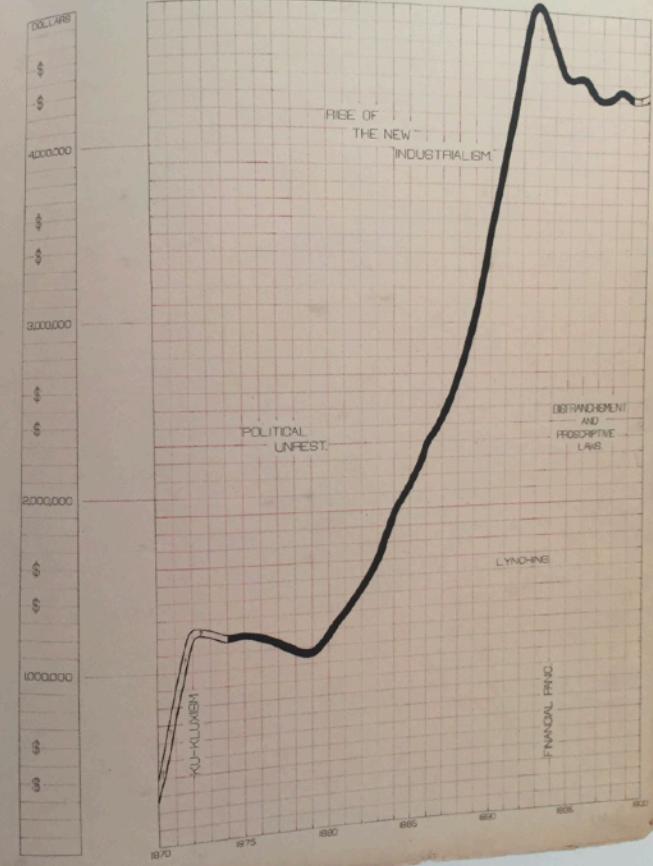
## Negro business men in the United States.

### Nègres Americains dans les affaires.

Done by Atlanta University.



VALUATION OF TOWN AND CITY PROPERTY OWNED  
BY GEORGIA NEGROES.



1900 Visualizing Black America , W. E. B. DuBois et al.

## Entering the 1900s...

Rise of **formal statistical methods** in the physical and social sciences

**Little innovation** in graphical methods

A period of **application and popularization**

Graphical methods enter textbooks, curricula, and **mainstream use**



**Four major influences** act on data analysis today:

1. The formal theories of statistics.
2. Accelerating developments in computers and display devices.
3. The challenge, in many fields, of more and larger bodies of data.
4. The emphasis on quantification in a wider variety of disciplines.



While some of the influences of statistical theory on data analysis have been helpful, others have not.

Data Analysis & Statistics, Tukey & Wilk 1966



**Exposure**, the effective laying open of the data to **display the unanticipated**, is to us a major portion of data analysis...

It is not clear how the **informality** and **flexibility** appropriate to the **exploratory character** of exposure can be fitted into any of the structures of formal statistics so far proposed.



Accordingly, both approaches and techniques need to be structured so as to **facilitate human involvement and intervention**.

Some implications for effective analysis are: (1) it is essential to have convenience of **interaction of people and intermediate results** and (2) at all stages of data analysis, the outputs need to be **matched to the capabilities of the people who use it and want it**.

Set A		Set B		Set C		Set D	
X	Y	X	Y	X	Y	X	Y
10	8.04	10	9.14	10	7.46	8	6.58
8	6.95	8	8.14	8	6.77	8	5.76
13	7.58	13	8.74	13	12.74	8	7.71
9	8.81	9	8.77	9	7.11	8	8.84
11	8.33	11	9.26	11	7.81	8	8.47
14	9.96	14	8.1	14	8.84	8	7.04
6	7.24	6	6.13	6	6.08	8	5.25
4	4.26	4	3.1	4	5.39	19	12.5
12	10.84	12	9.11	12	8.15	8	5.56
7	4.82	7	7.26	7	6.42	8	7.91
5	5.68	5	4.74	5	5.73	8	6.89

### Summary Statistics

$$u_X = 9.0 \quad \sigma_X = 3.317$$

$$u_Y = 7.5 \quad \sigma_Y = 2.03$$

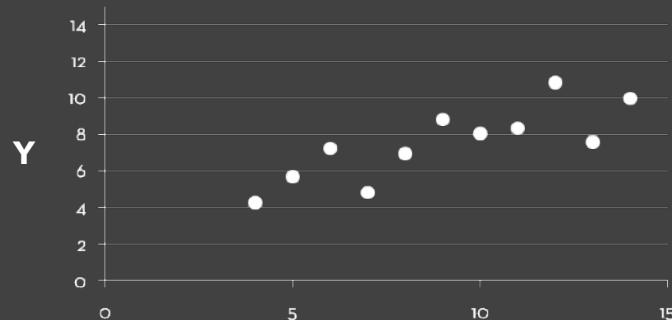
### Linear Regression

$$Y = 3 + 0.5 X$$

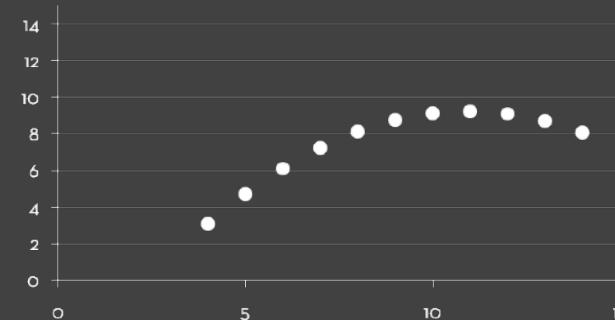
$$R^2 = 0.67$$

[Anscombe 1973]

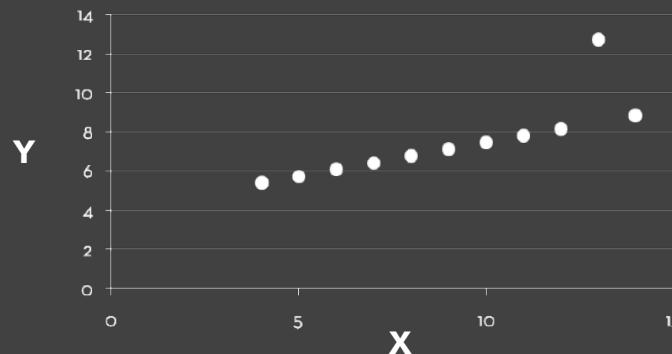
### Set A



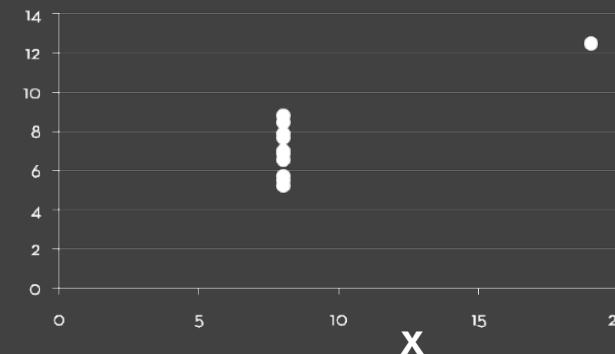
### Set B



### Set C



### Set D

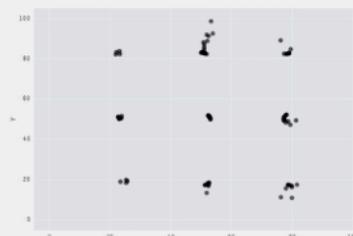
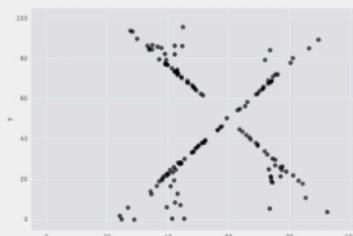
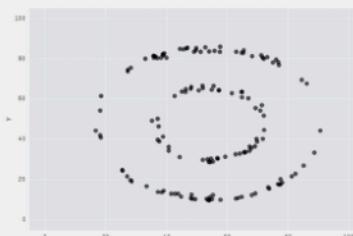
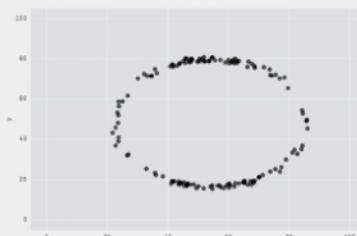
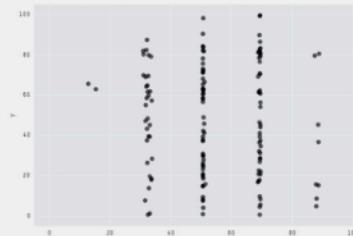
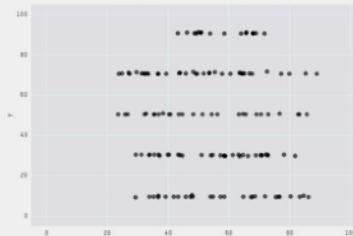
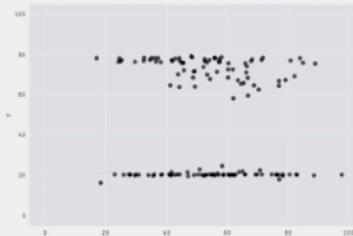
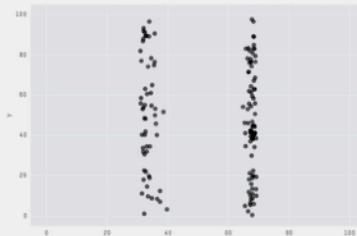
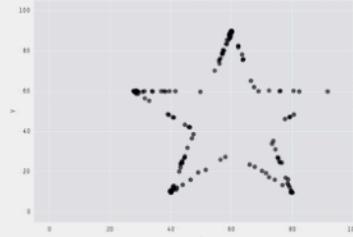
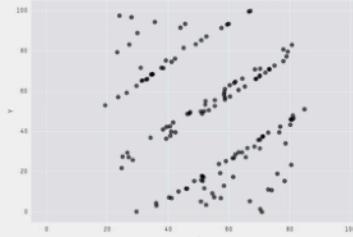
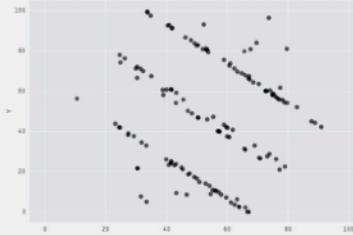
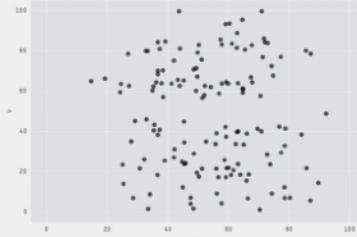


[Anscombe 1973]



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

[Matejka 2017]



# What is Visualization?

"Transformation of the symbolic into the geometric"

[McCormick et al. 1987]

"... finding the artificial memory that best supports our natural means of perception." [Bertin 1967]

"The use of computer-generated, interactive, visual representations of data to amplify cognition."

[Card, Mackinlay, & Shneiderman 1999]

# Goals of Visualization Research

## 1 **Understand** how visualizations convey information

What do people perceive / comprehend?

How do visualizations inform mental models?

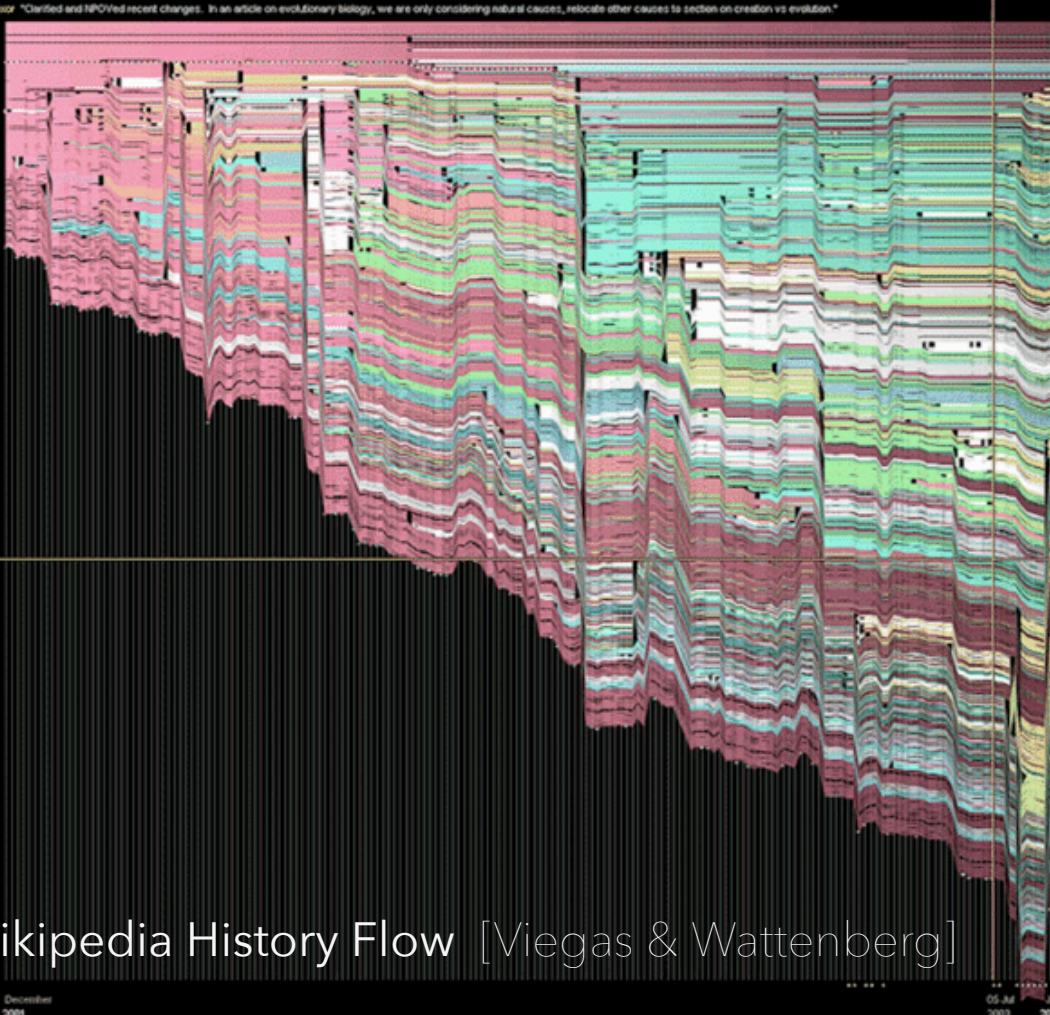
## 2 **Develop principles and techniques** for creating effective visualizations and supporting analysis

Leverage perception & augment cognition

Improve ties between visualization & mental model

Example:  
Wikipedia Edits

authors	posts
Desirell	3
Hennies	3
Henzel	4
Ed Poor	4
Tiarr	1
Suzanne_Brasser	1
Paul Drye	1
AxelBolt	1
Conversion script	1
Rigembie	1
Shubenstein	12
Bryan Denkison	1
Maverick49	5
Vicci.Rosenzweig	1
Josh Grasse	1
Robert Merkel	1
Pierre-Albert	1
Fredbeuler	1
Tjitz	1
Gog	4
Emmett	3
Q	1
Jehnouse	5
Canenbeert	1
Graff	3
Dermvera	1
AdwinRitchie	17
Ques zzz brown	2
Heron	1
Ryguessu	1
The Anone	1
Alan Peakall	1
Sneelichu	1
Eloquence	7
Malusz	1
Cyde	1
RX	3
Fred Beuler	5
JNSchmidt	1
Michael	1
Rewry	1
Zoe	1
Leonor	13
Someone else	1
Tenace	1
Zundark	1
Jedibreak	1
Mirage	1
1 Data Aligned	1



Therefore, over time, the types of organisms that have traits better **adapted** to their environment will tend to become the dominant ones in an environment, while organisms poorly adapted to their environment will become extinct. Natural selection also provides for a mechanism by which life can sustain itself over time. Since, in the long run, environments always change, if successive generations of organisms make adaptations which allow them to survive and reproduce, species would simply die out as their biological niches die out. Therefore, life is allowed to persist over great spans of time, in the form of evolving species. The central role of natural selection in evolutionary theory has created a strong connection between that field and the study of **biology**.

### Genetic drift.

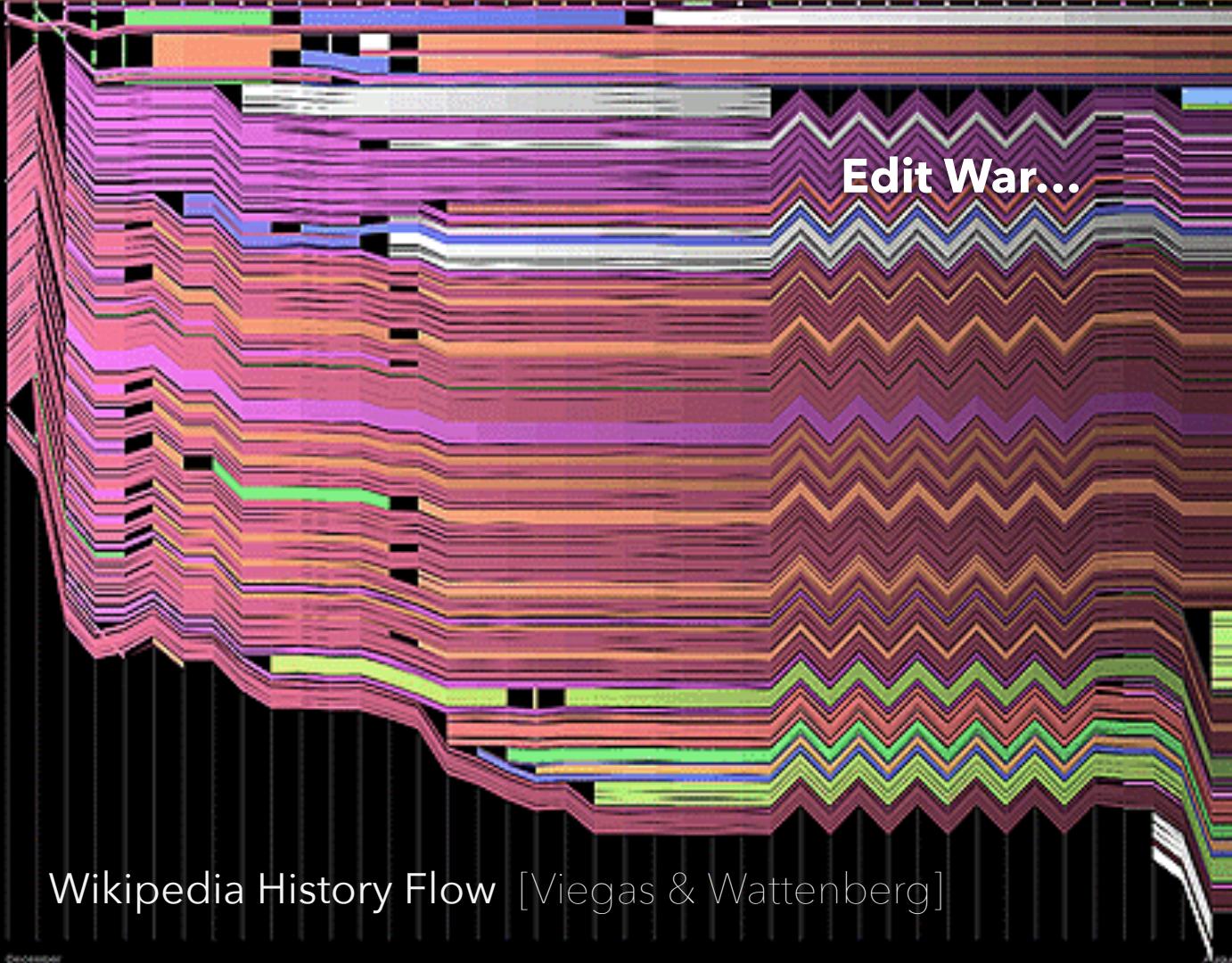
Genetic drift describes changes in gene frequency that cannot be ascribed to selective pressures, but are due instead to events that are unrelated to selection. This is especially important in small mating populations, where simply cannot have enough offspring to maintain the same gene distribution as the parental generation. Such fluctuations in gene frequency between successive generations may result in some genes disappearing from the population. Two separate populations that begin with the same gene frequency might, therefore, "drift" by random fluctuation into two divergent populations with different gene sets (i.e. genes that are present in one but have been lost in the other). Rare stochastic events (volcanoes, solar, meteor impact, etc.) might contribute to genetic drift by altering the gene frequency outside of "normal" selective pressures.

### Development of evolutionary theories

As science has uncovered more and more information about the biology of life, such as genetics and molecular biology, theories of evolution have changed. The general trend has been not to overturn well-supported theories, but to supplement them with more detailed and therefore more complex ones.

While transmutation was accepted by a sizeable number of scientists before 1859, it was the publication of Charles Darwin's *The Origin of Species* which provided the first cogent mechanism by which evolutionary change could persist: his mechanism of natural selection. The *Galápagos* timeline outlines the major steps of evolution on Earth as expounded by the theory's proponents.

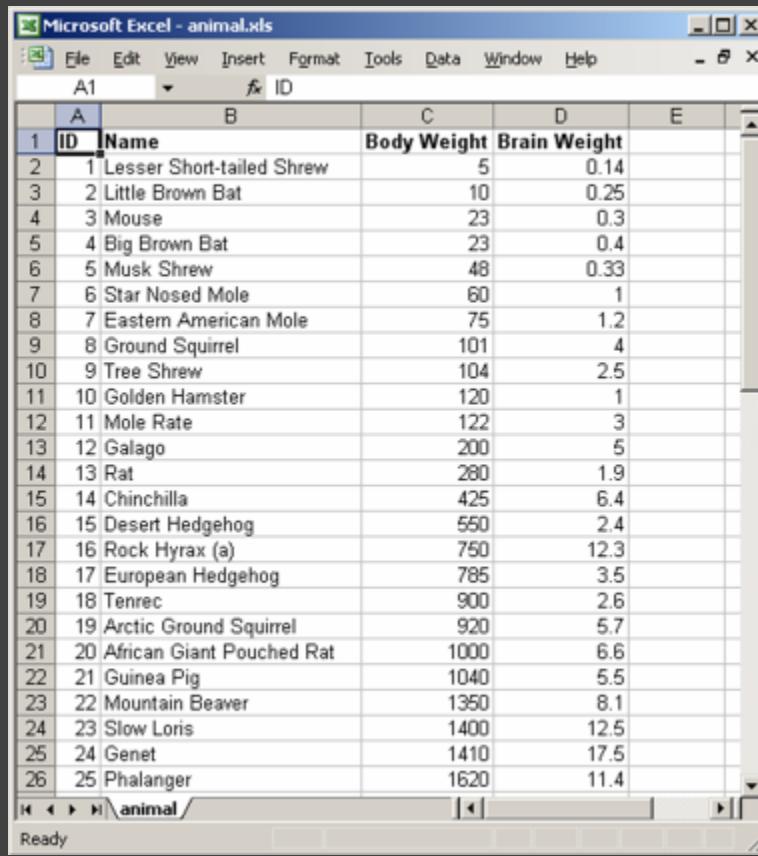
Following the dawn of molecular biology, it became clear that a major mechanism for variation within a population is the mutagenesis of DNA. An essential component to evolutionary theory is that during the cell cycle, DNA is copied fairly, but not entirely, faithfully. When these rare copying errors occur, they are said to introduce genetic mutations of three general consequences relative to the current environment: good, bad, or neutral. By definition, individuals with "good" mutations will have an a stronger propensity to propagate, individuals with "bad" mutations will have a lower chance of successful reproduction, and those carrying "neutral" mutations will have neither an advantage nor a disadvantage. These definitions assume that the environment remains stable. Considered at the level of a single gene, these variations just described represent different **genetic alleles**. Following environmental change, alleles may retain their classification of good, bad, or neutral, or may shift into one of the other categories. Those genes that are found to be "good" in one environment may become "bad" as other bear favorable mutations.



Wikipedia History Flow [Viegas & Wattenberg]

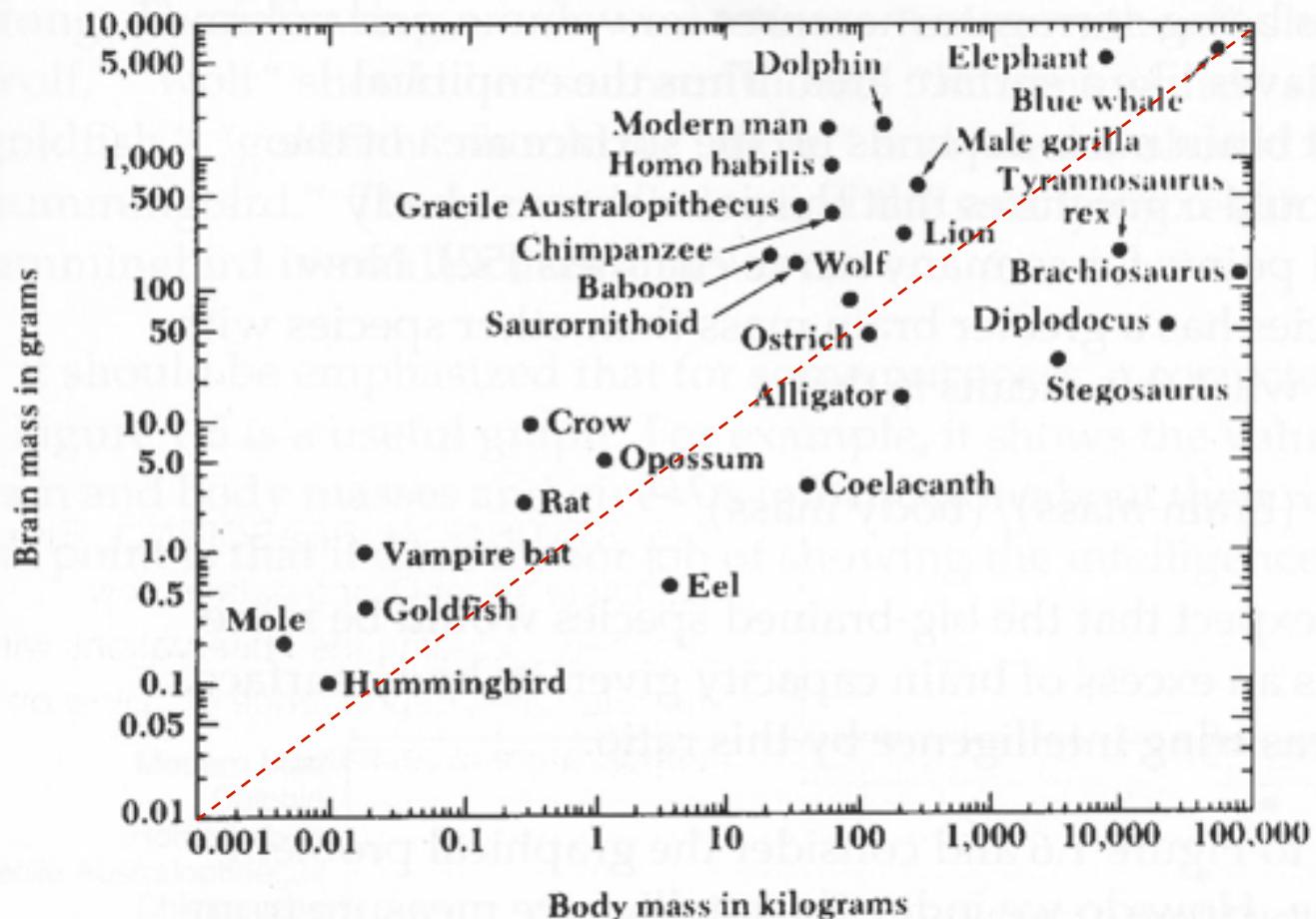
Example:  
Animal Brains

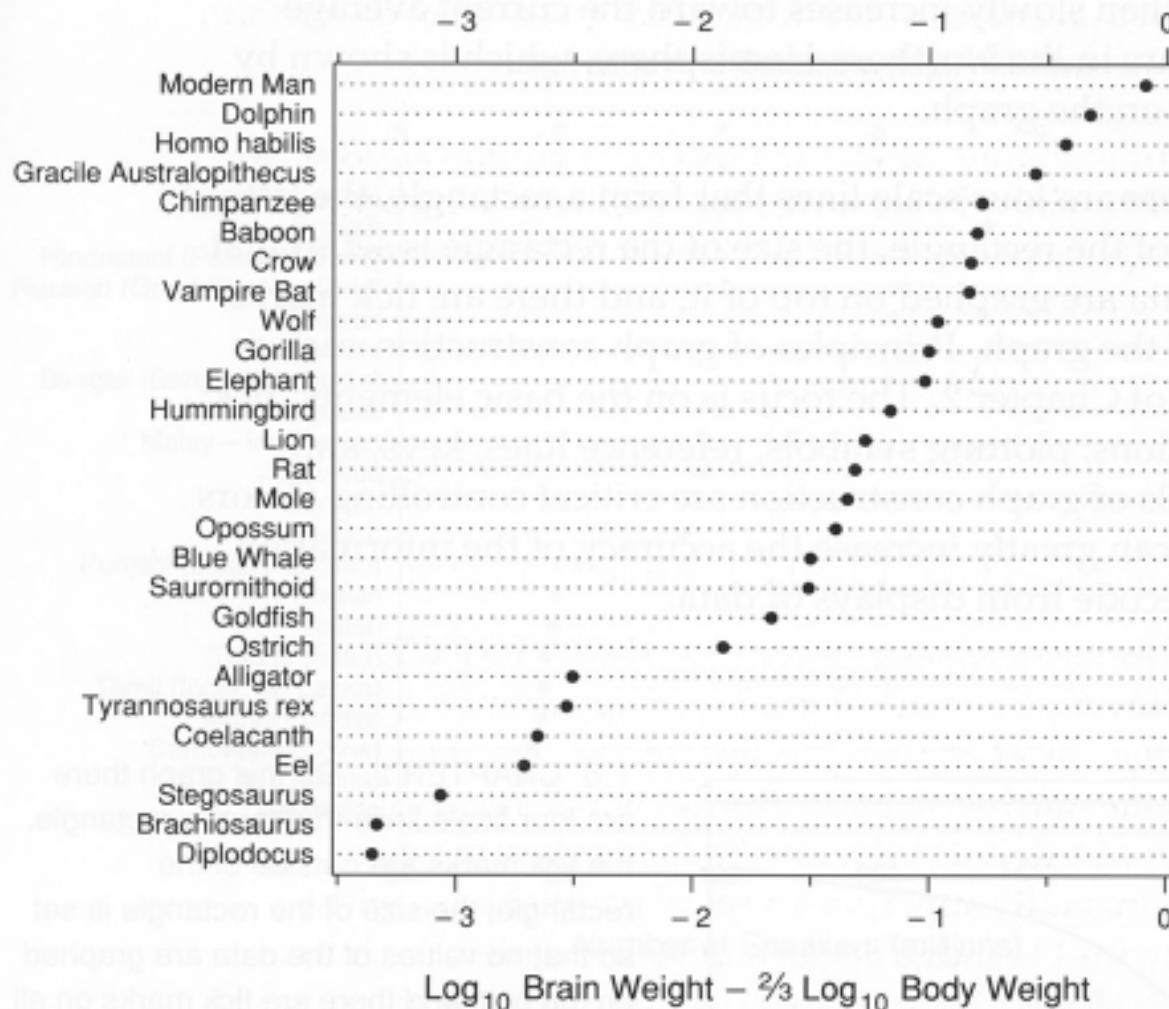
# Which animals are the “smartest”?



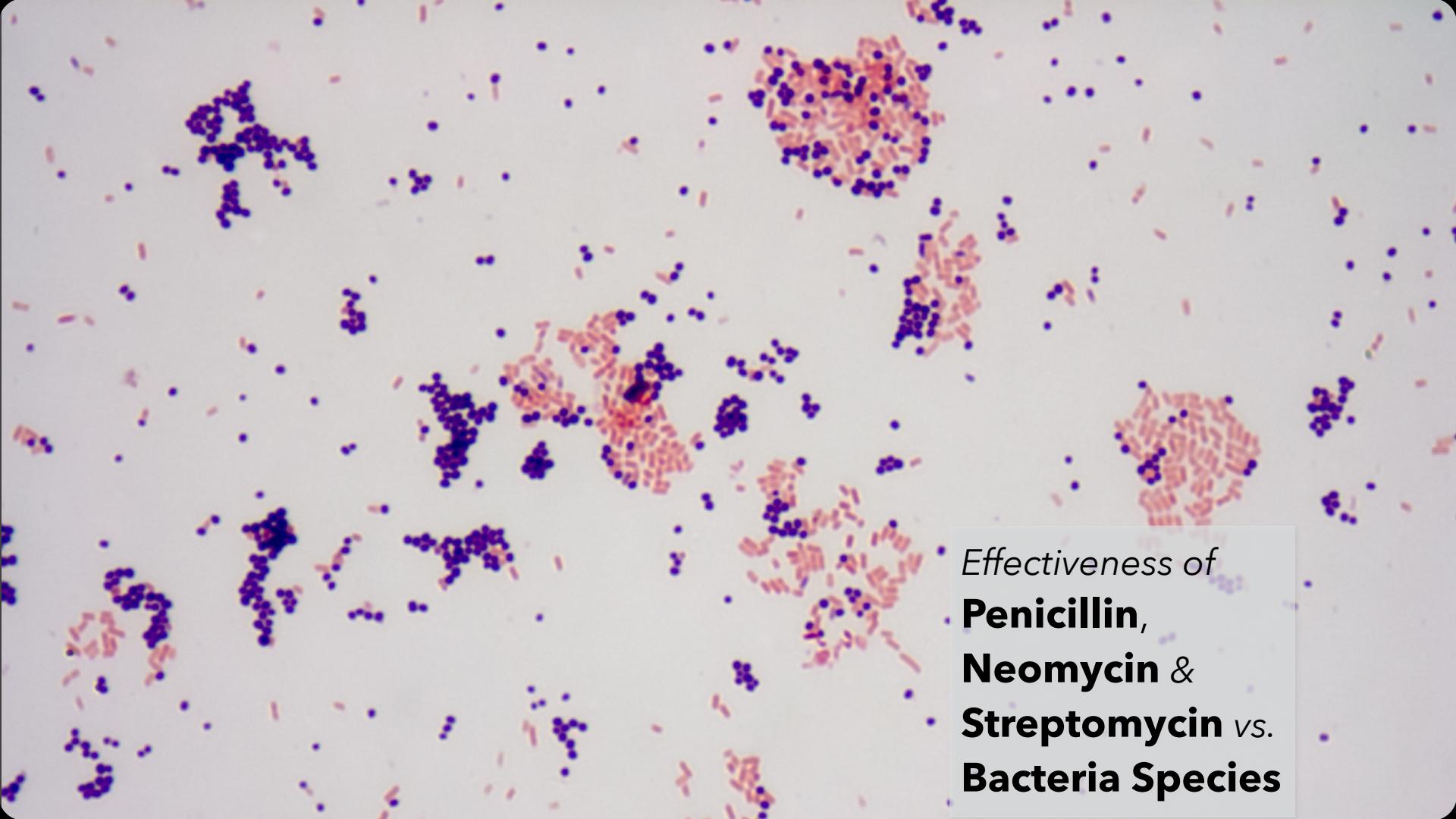
A screenshot of a Microsoft Excel spreadsheet titled "animal.xls". The spreadsheet contains data for 26 different animals, listed in rows 2 through 27. The columns are labeled A, B, C, D, and E. Column A is labeled "ID" and contains numerical values from 1 to 26. Column B is labeled "Name" and lists the names of the animals. Column C is labeled "Body Weight" and column D is labeled "Brain Weight". The data shows that the animal with the highest brain weight relative to body weight is the Tenrec, with a brain weight of 3.5 and a body weight of 900.

A	B	C	D	E
1	ID	Name	Body Weight	Brain Weight
2	1	Lesser Short-tailed Shrew	5	0.14
3	2	Little Brown Bat	10	0.25
4	3	Mouse	23	0.3
5	4	Big Brown Bat	23	0.4
6	5	Musk Shrew	48	0.33
7	6	Star Nosed Mole	60	1
8	7	Eastern American Mole	75	1.2
9	8	Ground Squirrel	101	4
10	9	Tree Shrew	104	2.5
11	10	Golden Hamster	120	1
12	11	Mole Rat	122	3
13	12	Galago	200	5
14	13	Rat	280	1.9
15	14	Chinchilla	425	6.4
16	15	Desert Hedgehog	550	2.4
17	16	Rock Hyrax (a)	750	12.3
18	17	European Hedgehog	785	3.5
19	18	Tenrec	900	2.6
20	19	Arctic Ground Squirrel	920	5.7
21	20	African Giant Pouched Rat	1000	6.6
22	21	Guinea Pig	1040	5.5
23	22	Mountain Beaver	1350	8.1
24	23	Slow Loris	1400	12.5
25	24	Genet	1410	17.5
26	25	Phalanger	1620	11.4





# Example: Antibiotic Effectiveness

A microscopic image showing numerous bacterial colonies. The colonies are irregular, cluster-like structures composed of many small, dark purple, rod-shaped bacteria. They are scattered across a light-colored, textured background, which appears to be a petri dish or slide. The lighting is bright, highlighting the contrast between the colonies and the background.

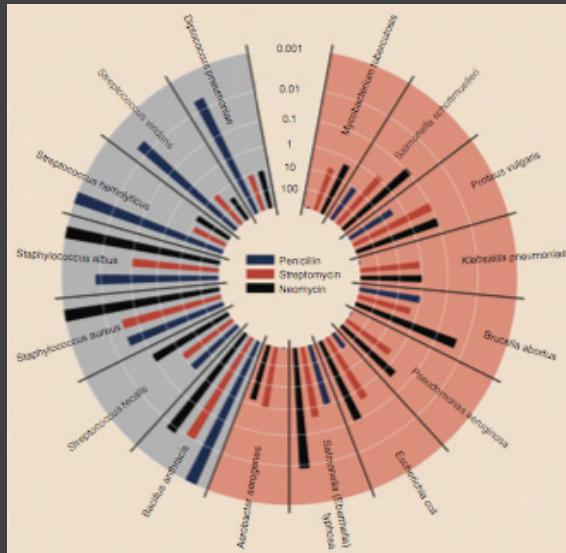
# *Effectiveness of Penicillin, Neomycin & Streptomycin vs. Bacteria Species*

# What questions might we ask?

Table 1: Burtin's data.

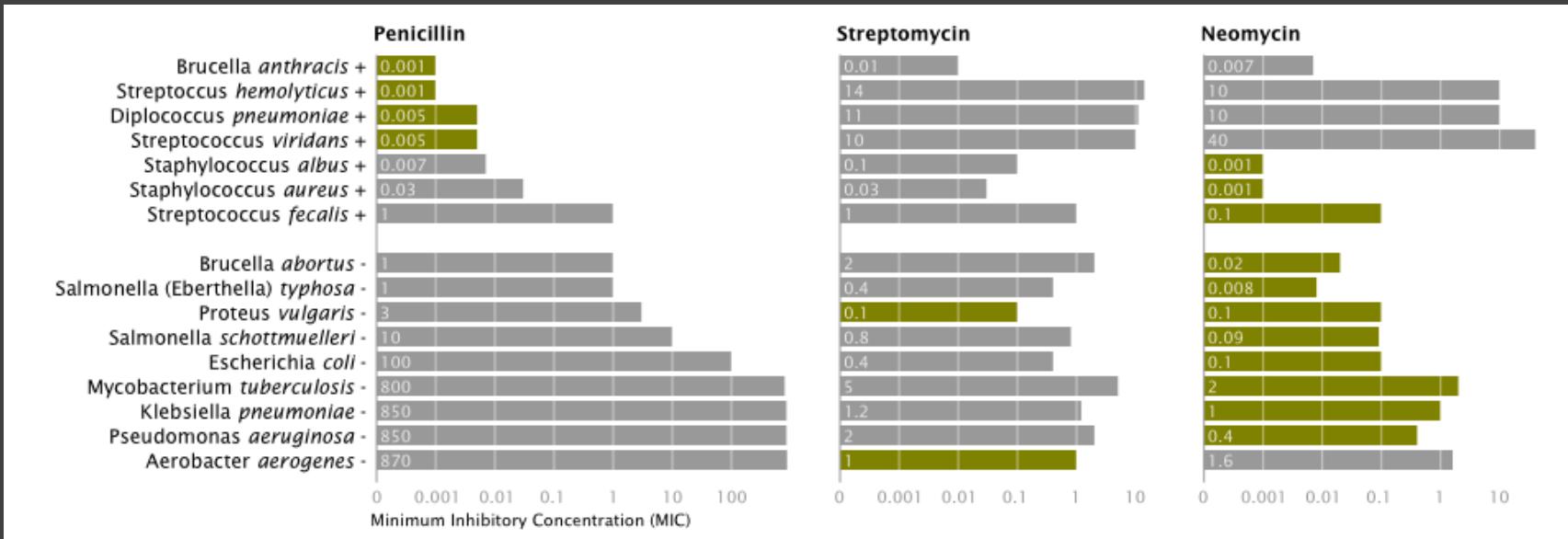
Bacteria	Antibiotic			
	Penicillin	Streptomycin	Neomycin	Gram Staining
<i>Aerobacter aerogenes</i>	870	1	1.6	negative
<i>Brucella abortus</i>	1	2	0.02	negative
<i>Brucella anthracis</i>	0.001	0.01	0.007	positive
<i>Diplococcus pneumoniae</i>	0.005	11	10	positive
<i>Escherichia coli</i>	100	0.4	0.1	negative
<i>Klebsiella pneumoniae</i>	850	1.2	1	negative
<i>Mycobacterium tuberculosis</i>	800	5	2	negative
<i>Proteus vulgaris</i>	3	0.1	0.1	negative
<i>Pseudomonas aeruginosa</i>	850	2	0.4	negative
<i>Salmonella (Eberthella) typhosa</i>	1	0.4	0.008	negative
<i>Salmonella schottmuelleri</i>	10	0.8	0.09	negative
<i>Staphylococcus albus</i>	0.007	0.1	0.001	positive
<i>Staphylococcus aureus</i>	0.03	0.03	0.001	positive
<i>Streptococcus fecalis</i>	1	1	0.1	positive
<i>Streptococcus hemolyticus</i>	0.001	14	10	positive
<i>Streptococcus viridans</i>	0.005	10	40	positive

# Which antibiotic is most effective?



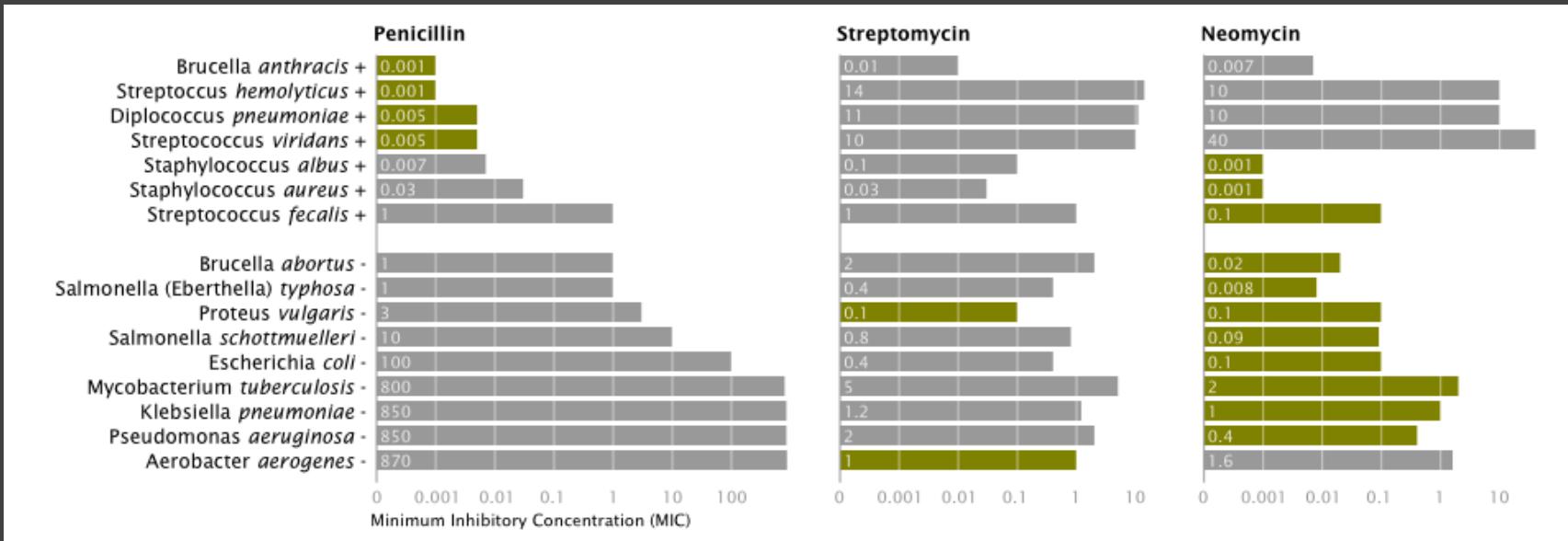
Bacteria	Penicillin	Antibiotic Streptomycin	Neomycin	Gram stain
<i>Aerobacter aerogenes</i>	870	1	1.6	-
<i>Brucella abortus</i>	1	2	0.02	-
<i>Bacillus anthracis</i>	0.001	0.01	0.007	+
<i>Diplococcus pneumoniae</i>	0.005	11	10	+
<i>Escherichia coli</i>	100	0.4	0.1	-
<i>Klebsiella pneumoniae</i>	850	1.2	1	-
<i>Mycobacterium tuberculosis</i>	800	5	2	-
<i>Proteus vulgaris</i>	3	0.1	0.1	-
<i>Pseudomonas aeruginosa</i>	850	2	0.4	-
<i>Salmonella (Eberthella) typhosa</i>	1	0.4	0.008	-
<i>Salmonella schottmuelleri</i>	10	0.8	0.09	-
<i>Staphylococcus albus</i>	0.007	0.1	0.001	+
<i>Staphylococcus aureus</i>	0.03	0.03	0.001	+
<i>Streptococcus faecalis</i>	1	1	0.1	+
<i>Streptococcus hemolyticus</i>	0.001	14	10	+
<i>Streptococcus viridans</i>	0.005	10	40	+

# Which antibiotic is most effective?



Mike Bostock

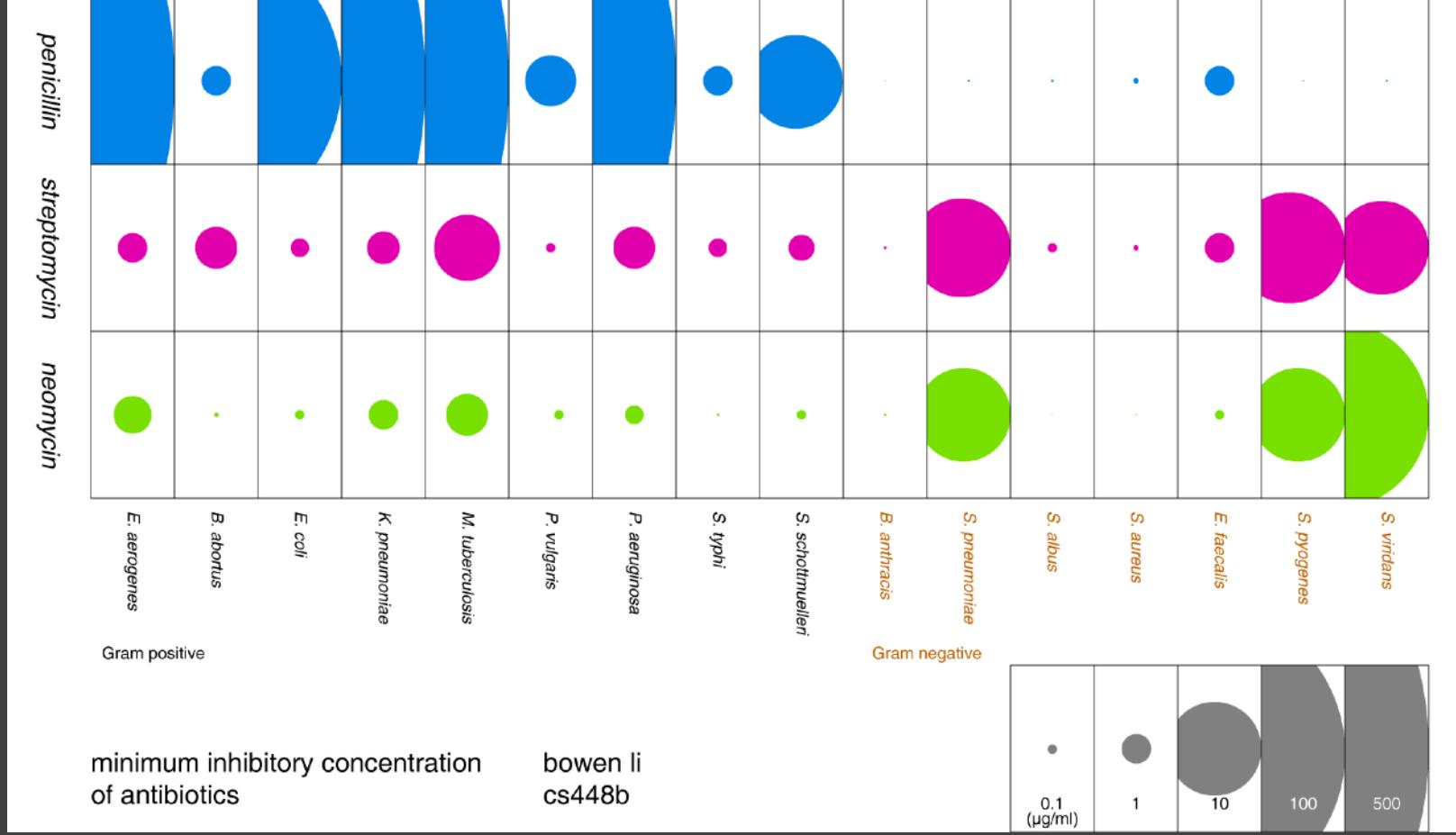
# Which antibiotic is most effective?



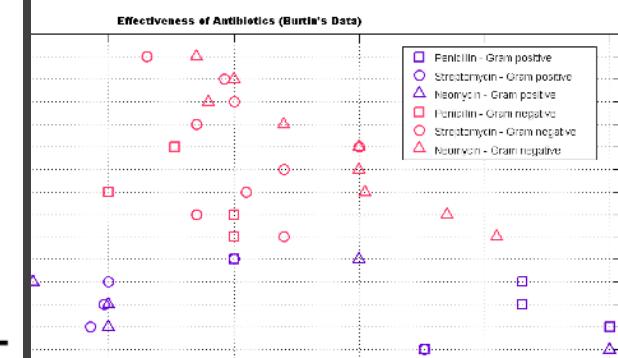
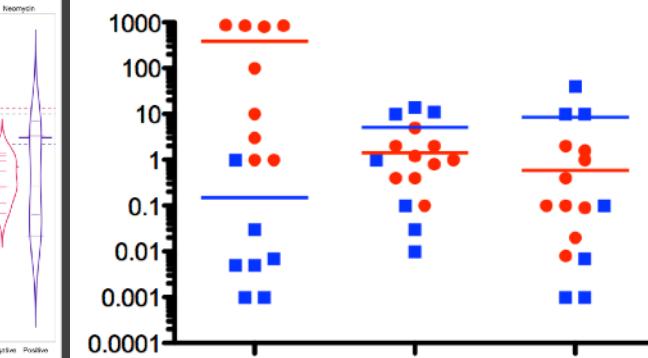
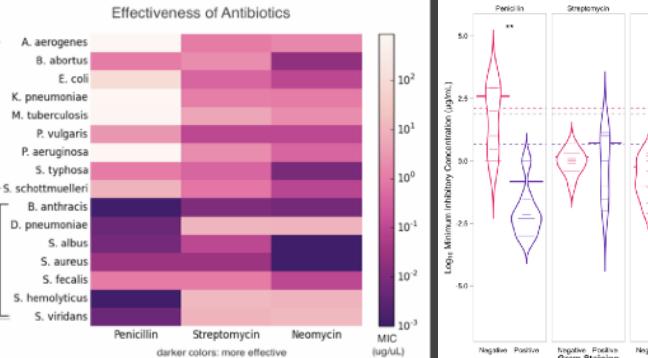
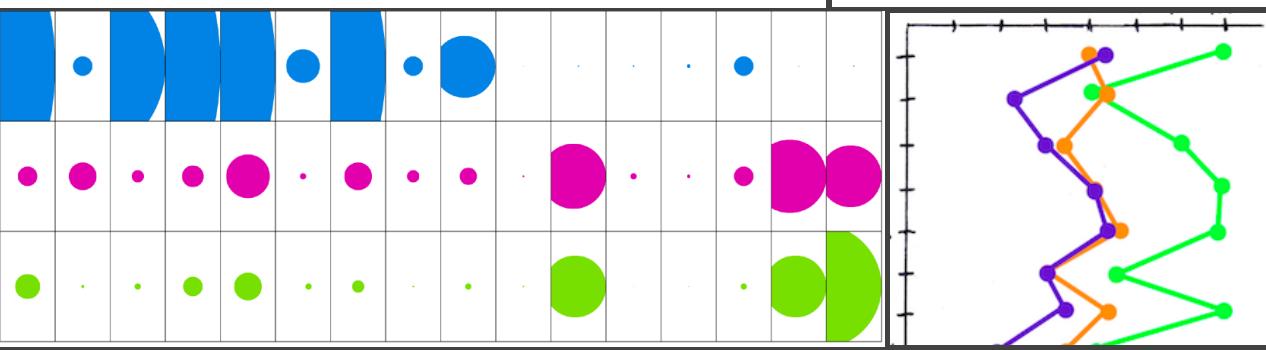
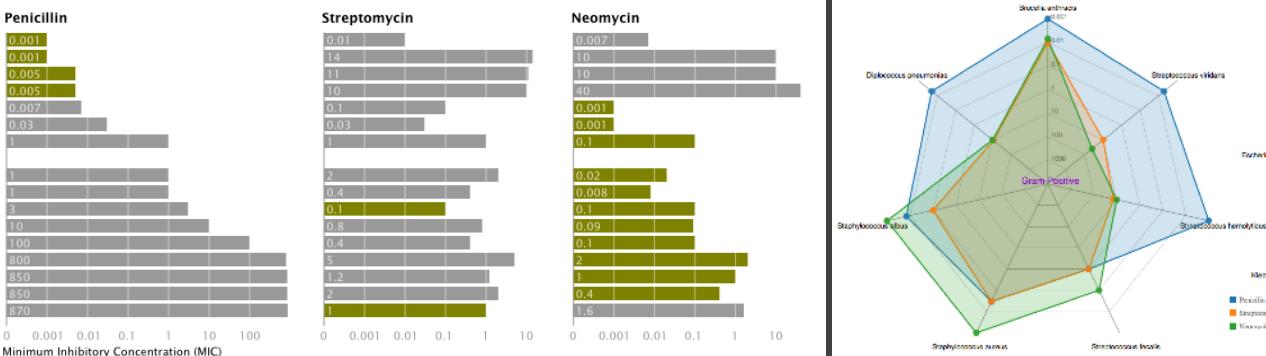
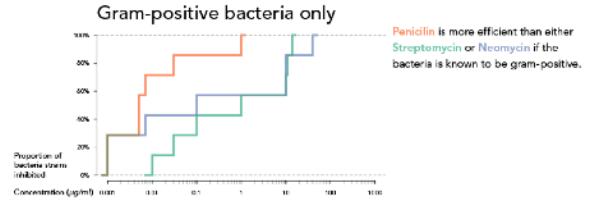
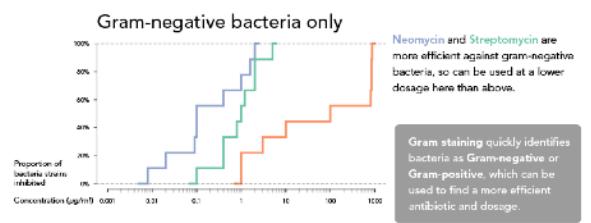
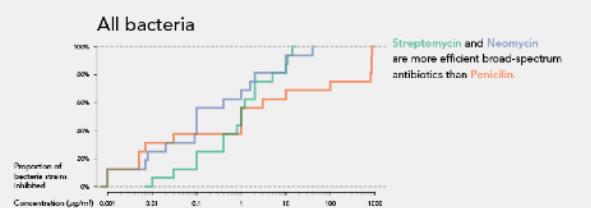
X-axis: Antibiotic |  $\log(\text{MIC})$

Y-axis: Gram-Staining | Species

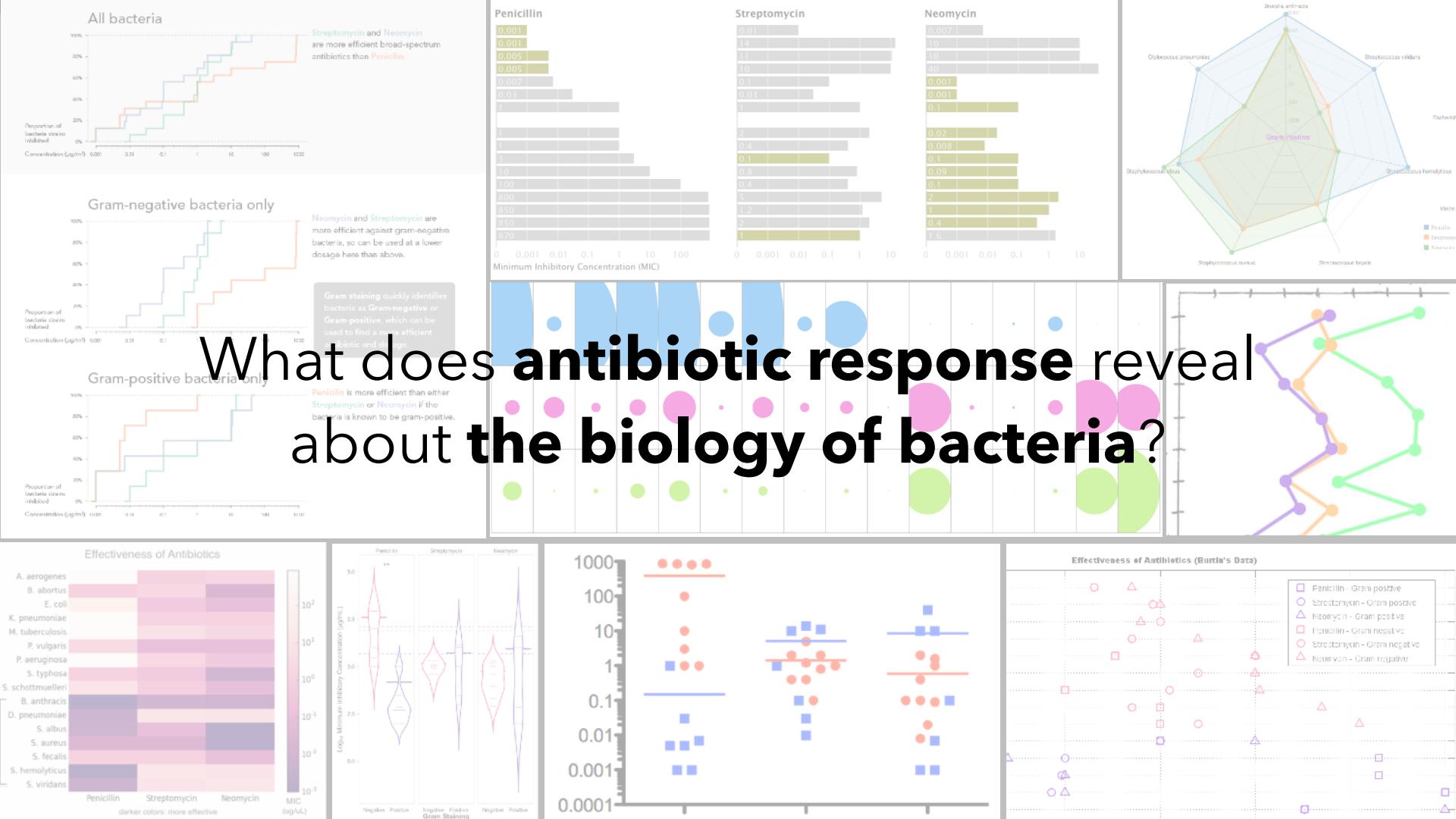
Color: Most-Effective?



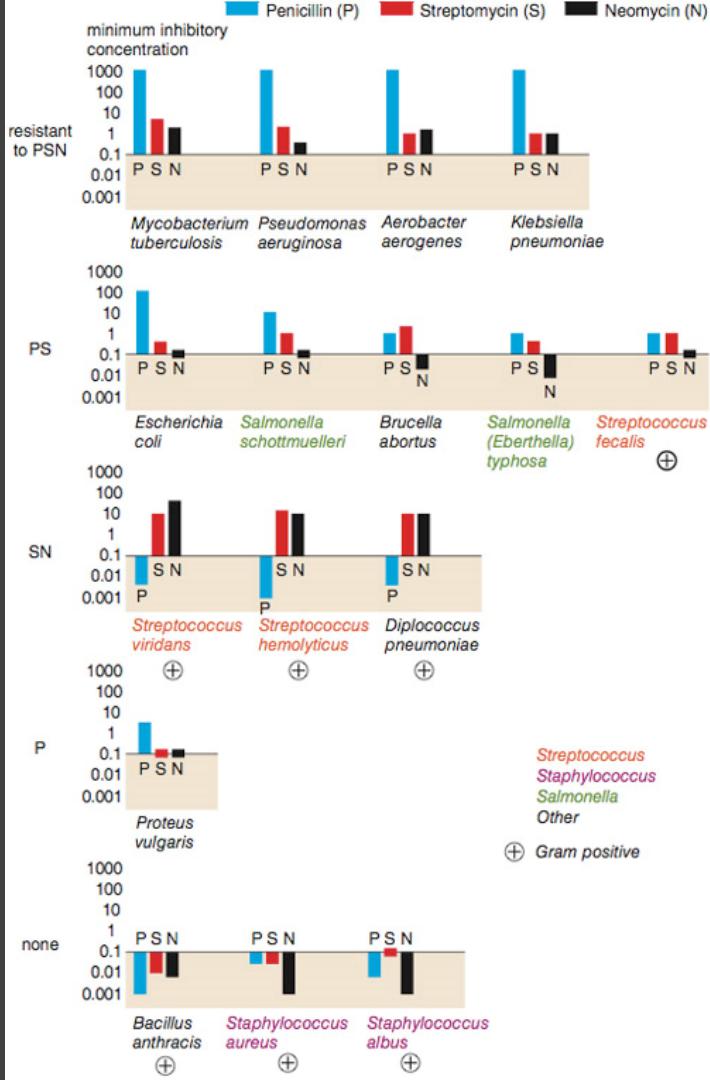
Bowen Li







Do the bacteria  
group by antibiotic  
resistance?

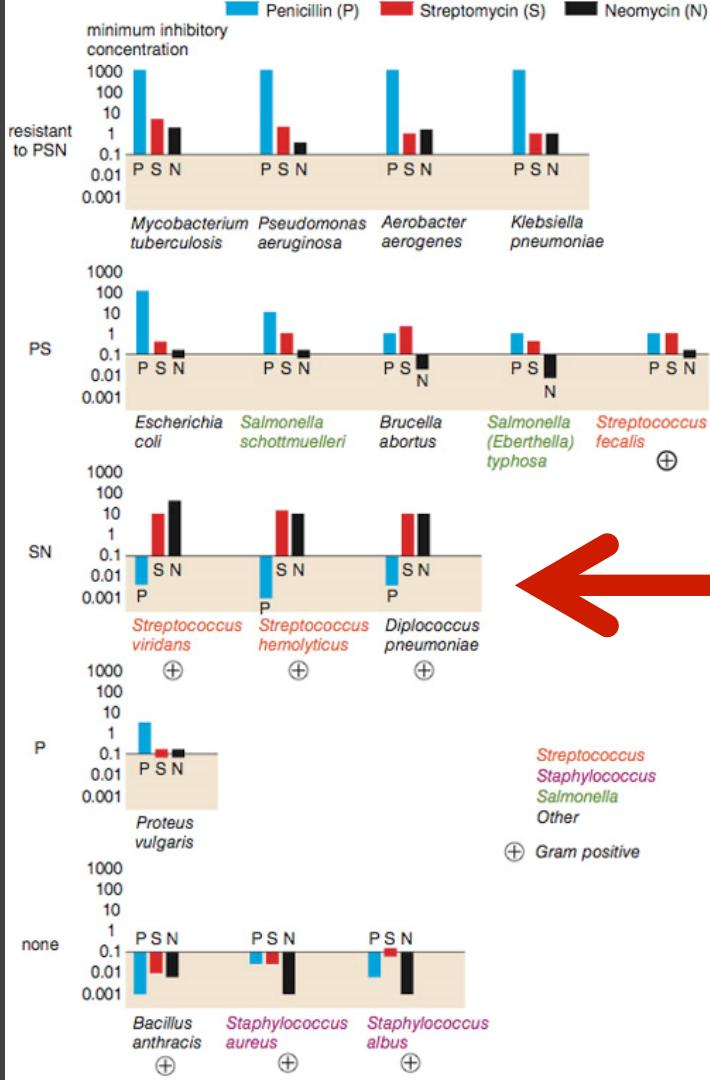


# Do the bacteria group by antibiotic resistance?

*Streptococcus*  
*Staphylococcus*  
*Salmonella*  
*Other*

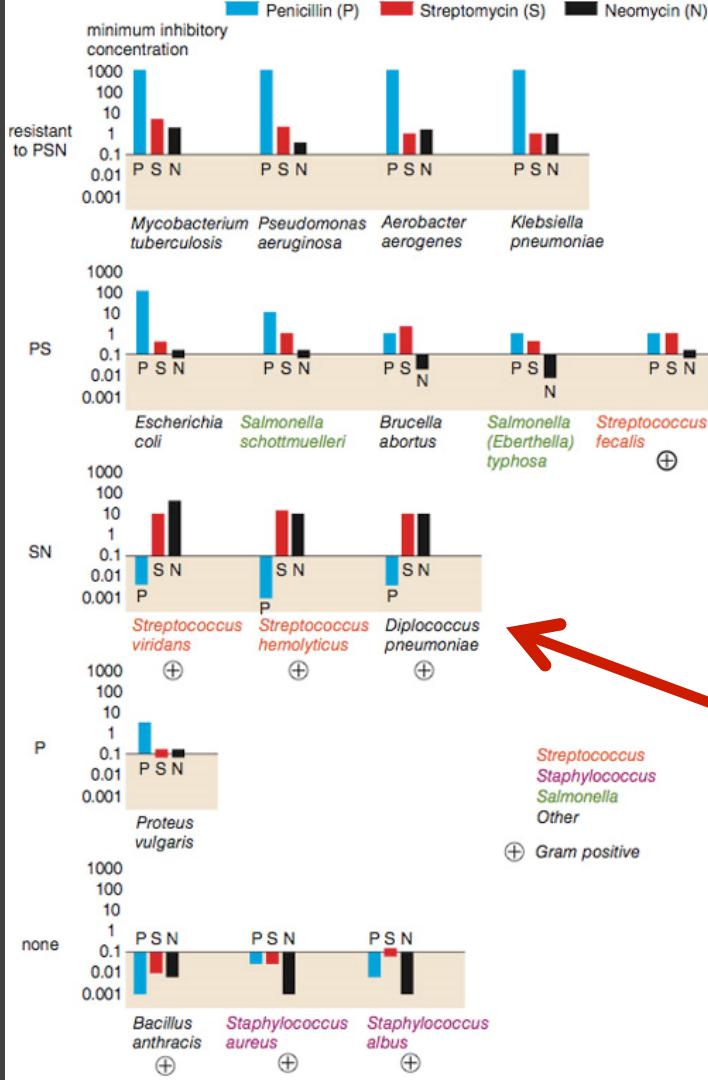
⊕ Gram positive

Wainer & Lysen  
*American Scientist*, 2009



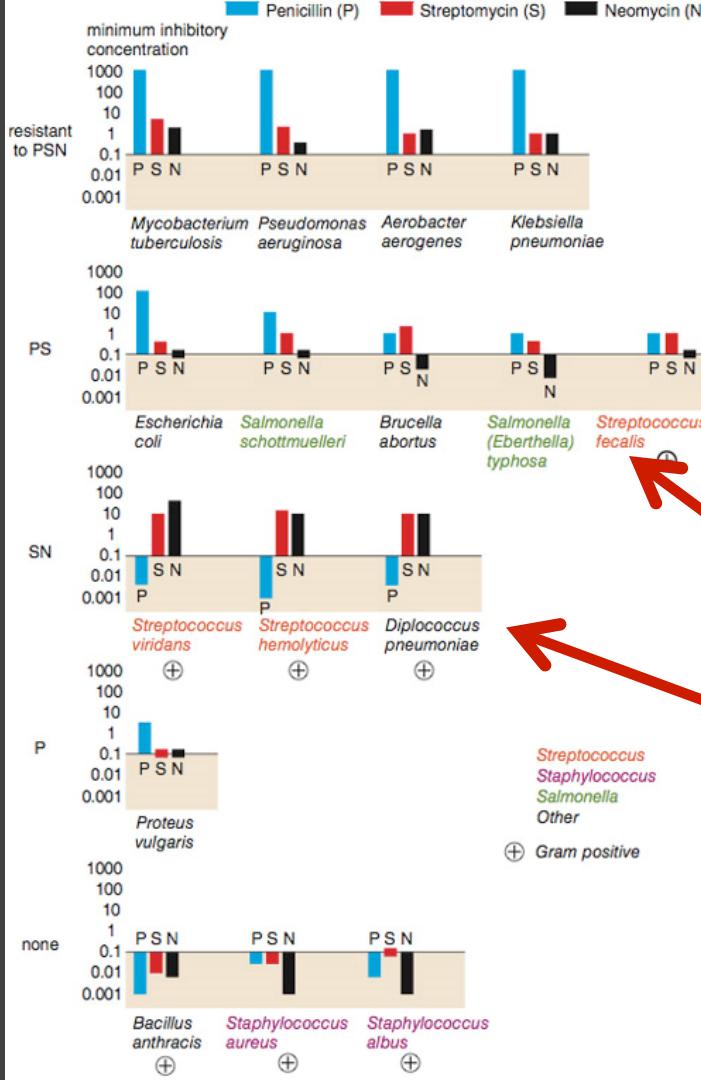
# Do the bacteria group by antibiotic resistance?

Wainer & Lysen  
American Scientist, 2009



# Do the bacteria group by antibiotic resistance?

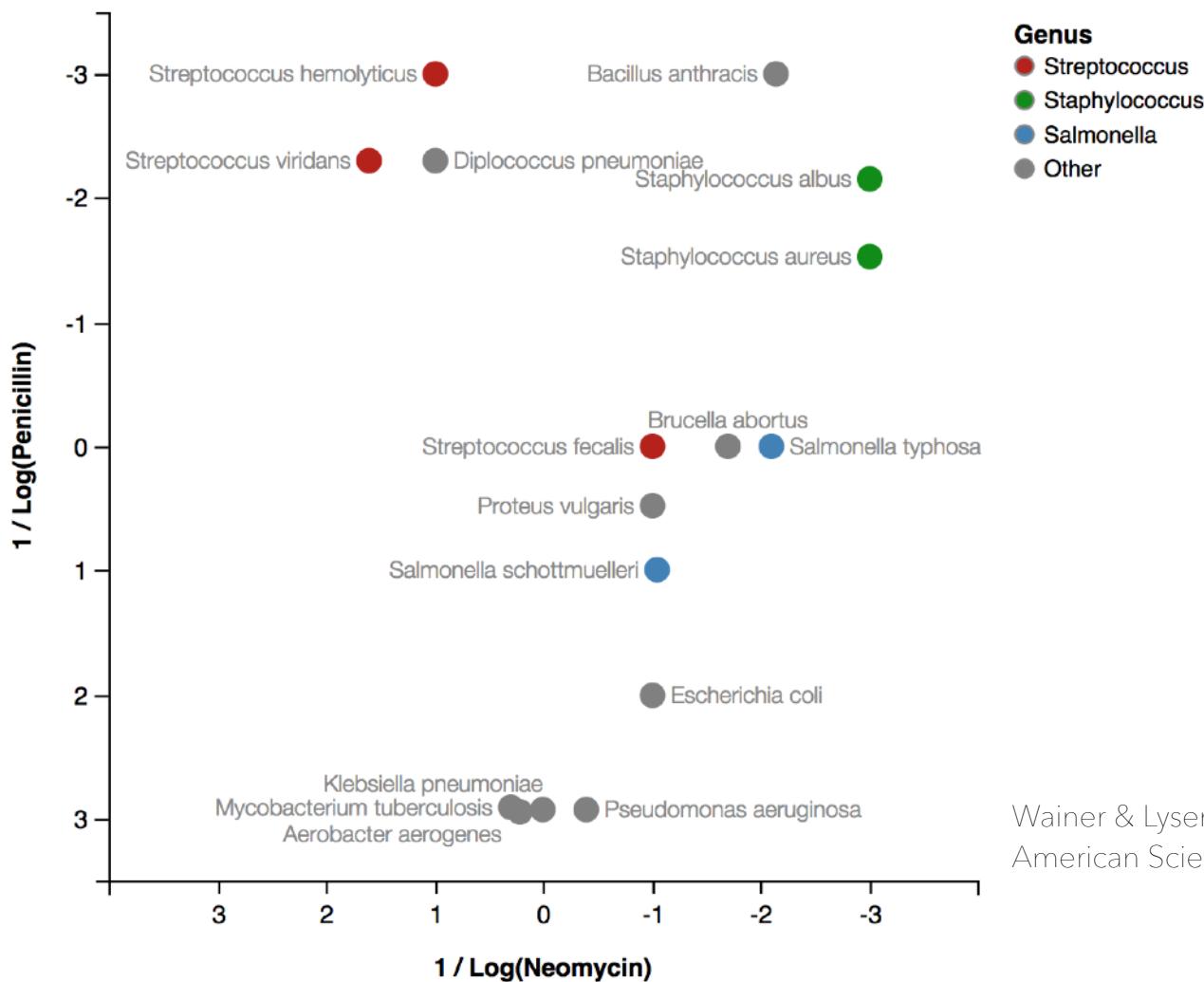
Really a streptococcus!  
(realized ~20 yrs later)



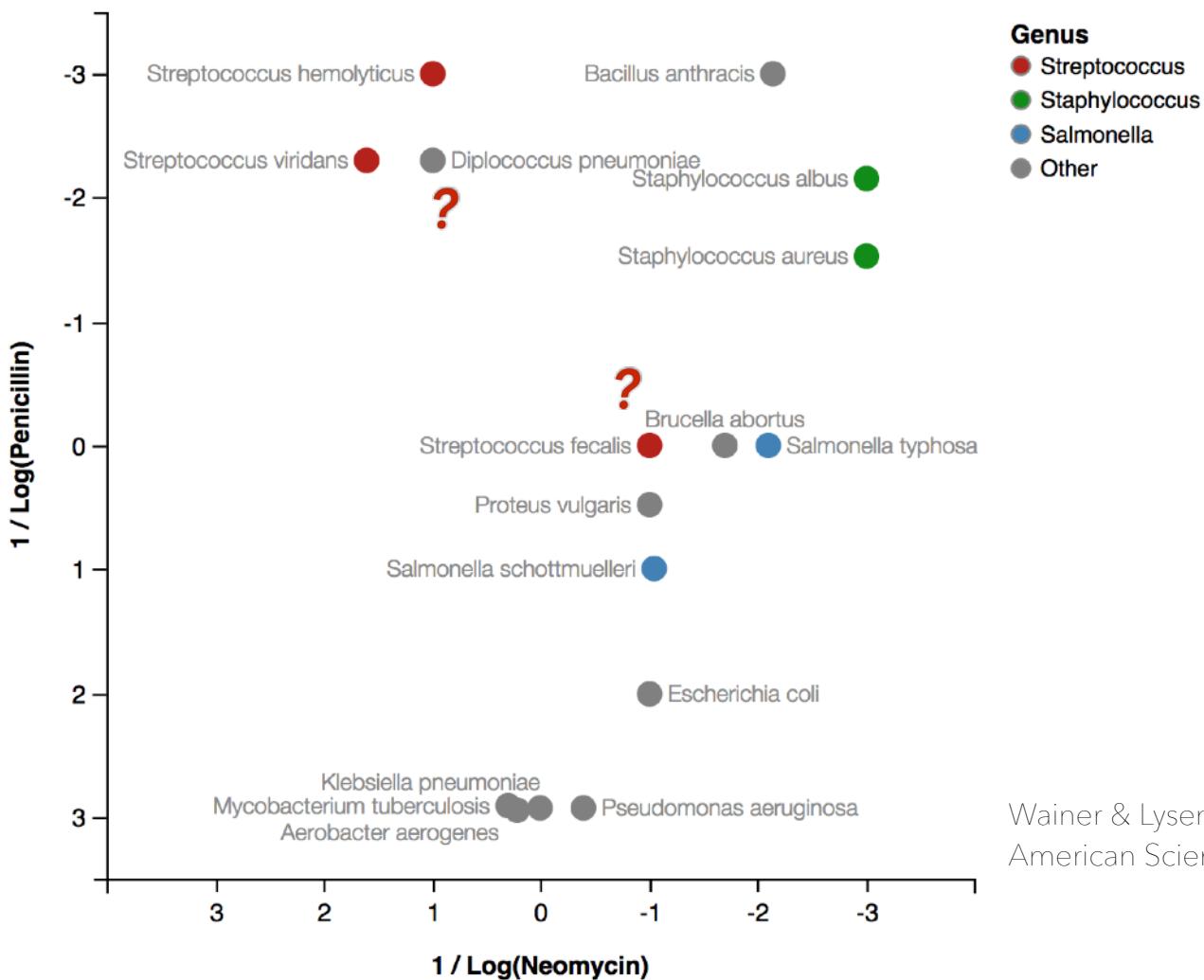
# Do the bacteria group by antibiotic resistance?

Not a streptococcus!  
(realized ~30 yrs later)

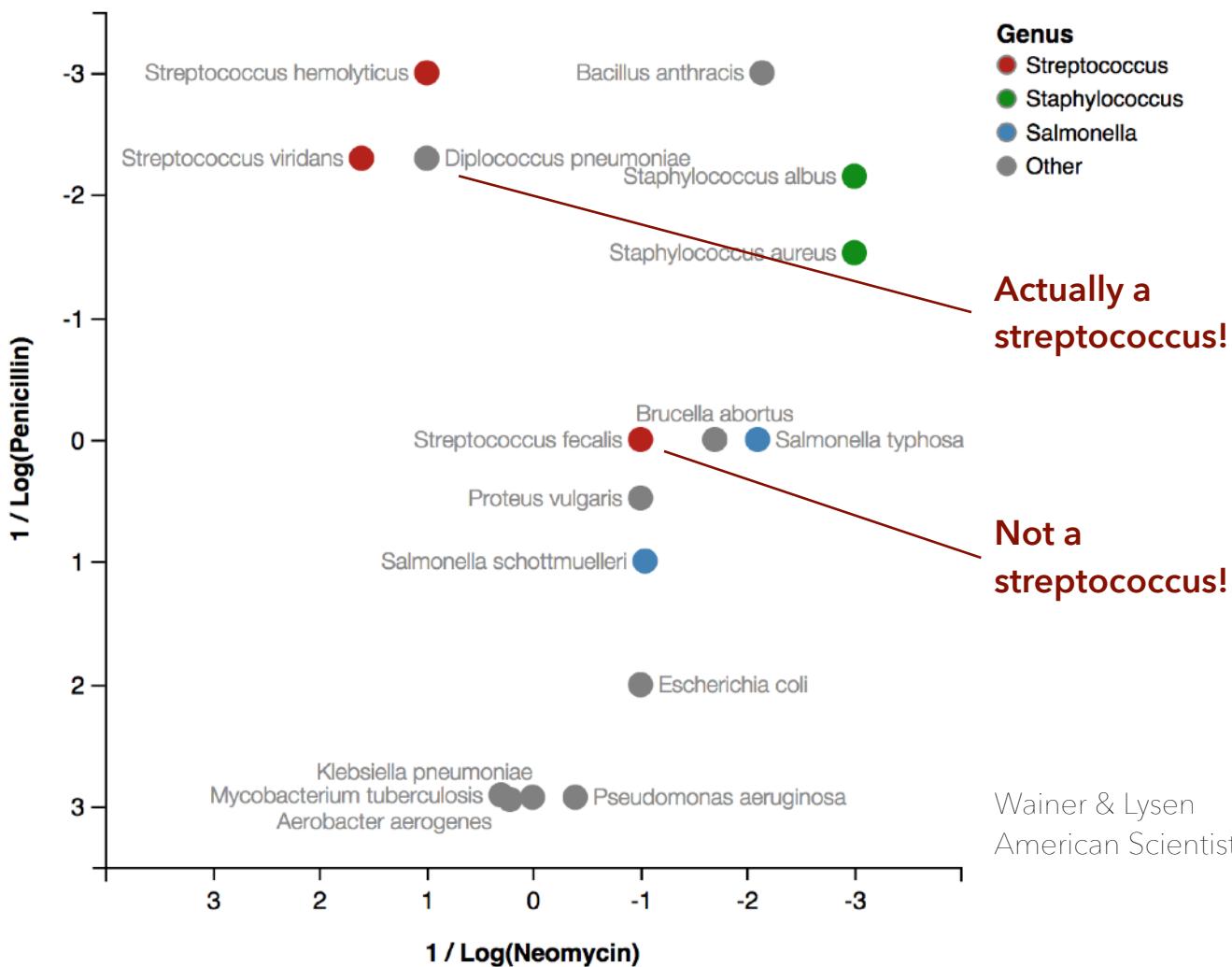
Really a streptococcus!  
(realized ~20 yrs later)



Wainer & Lysen  
American Scientist, 2009



Wainer & Lysen  
American Scientist, 2009



# Lesson: Iterative Exploration

## **Exploratory Process**

- 1 Construct graphics to address questions
- 2 Inspect “answer” and assess new questions
- 3 Repeat...

**Transform data** appropriately (e.g., invert, log)

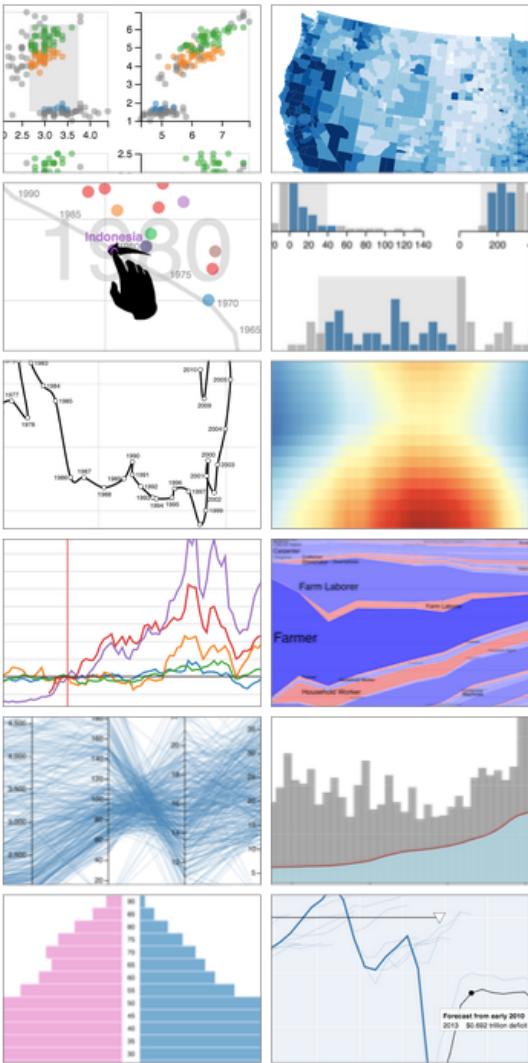
**Formulate clear analysis questions & goals**

**Don’t trust your data!**

# Exploration Tasks

# Data Exploration Tasks

**Profile:** learn the shape and structure of the data, assess data quality, check modeling assumptions  
**GOAL:** Is the data actionable? What can we ask?



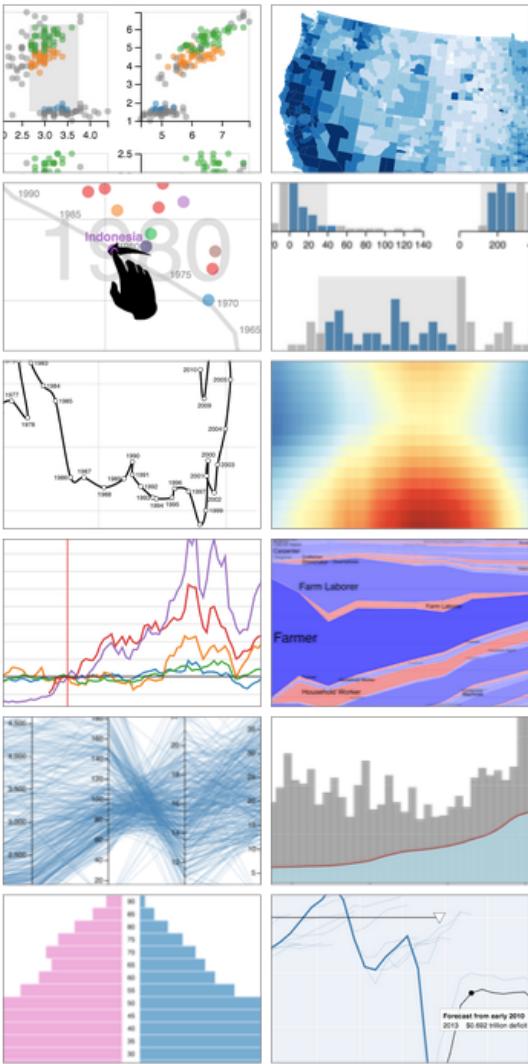
# Data Exploration Tasks

**Profile:** learn the shape and structure of the data, assess data quality, check modeling assumptions

## **GOAL:** Is the data actionable? What can we ask?

**Search:** identify specific data points or relations of interest to form an evidentiary chain

**GOAL:** Fact-finding, isolate important points/connections



# Data Exploration Tasks

**Profile:** learn the shape and structure of the data, assess data quality, check modeling assumptions

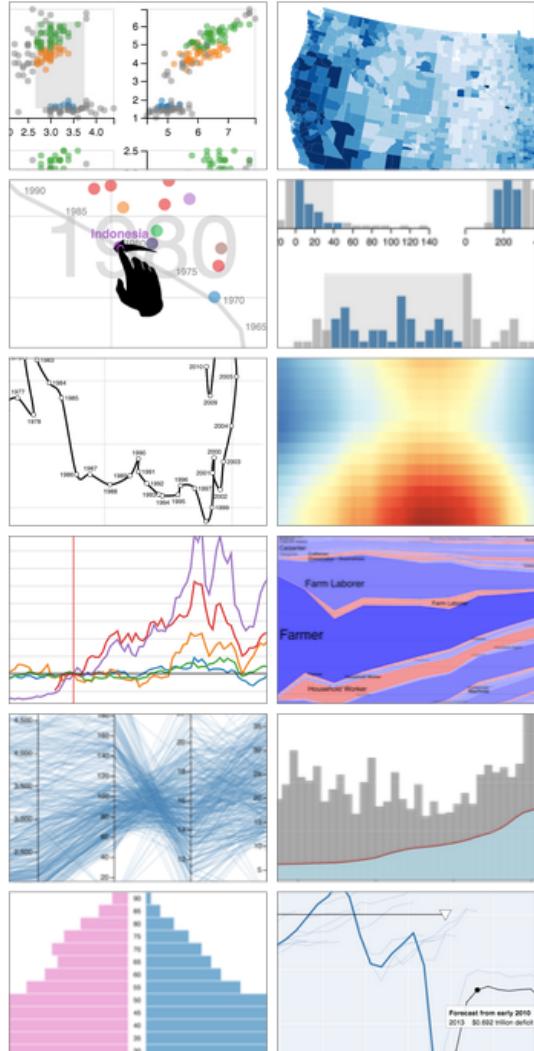
**GOAL:** Is the data actionable? What can we ask?

**Search:** identify specific data points or relations of interest to form an evidentiary chain

**GOAL:** Fact-finding, isolate important points/connections

**Infer:** generalize from observed patterns, ascribe observations to specific factors or causes

**GOAL:** Inform modeling and decision making

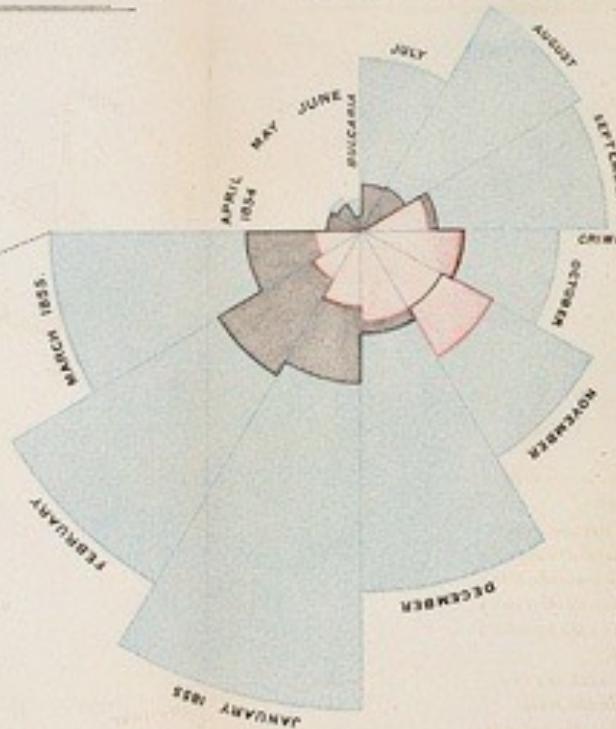
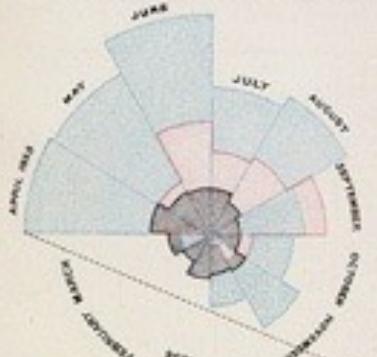


# Communication Tasks

2.  
APRIL 1855 TO MARCH 1856.

1.  
APRIL 1854 TO MARCH 1855.

DIAGRAM OF THE CAUSES OF MORTALITY  
IN THE ARMY IN THE EAST.

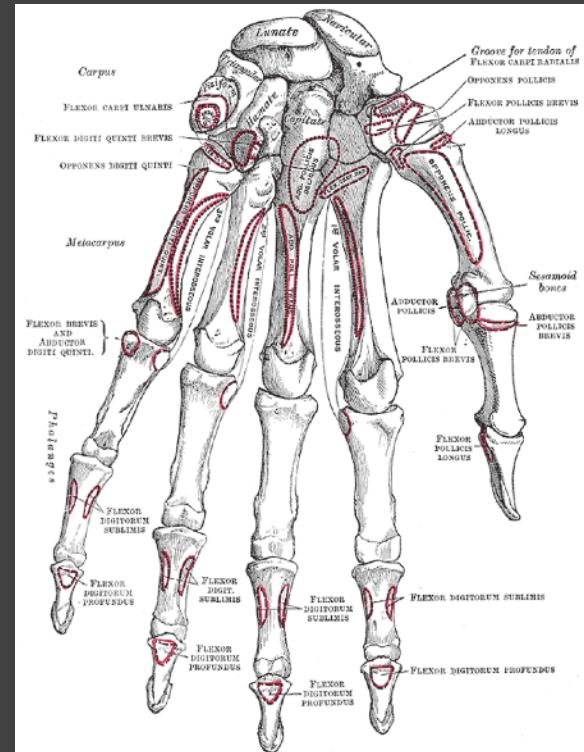
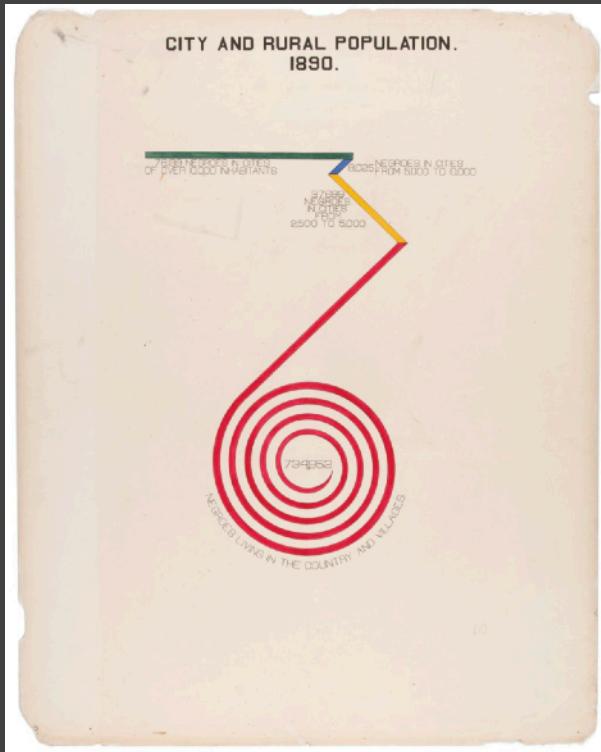


"to affect thro' the Eyes  
what we fail to convey to  
the public through their  
word-proof ears"

Harriet Martineau, 1858

1856 "Coxcomb" of Crimean War Deaths, Florence Nightingale

# Communicate, Inform, Inspire

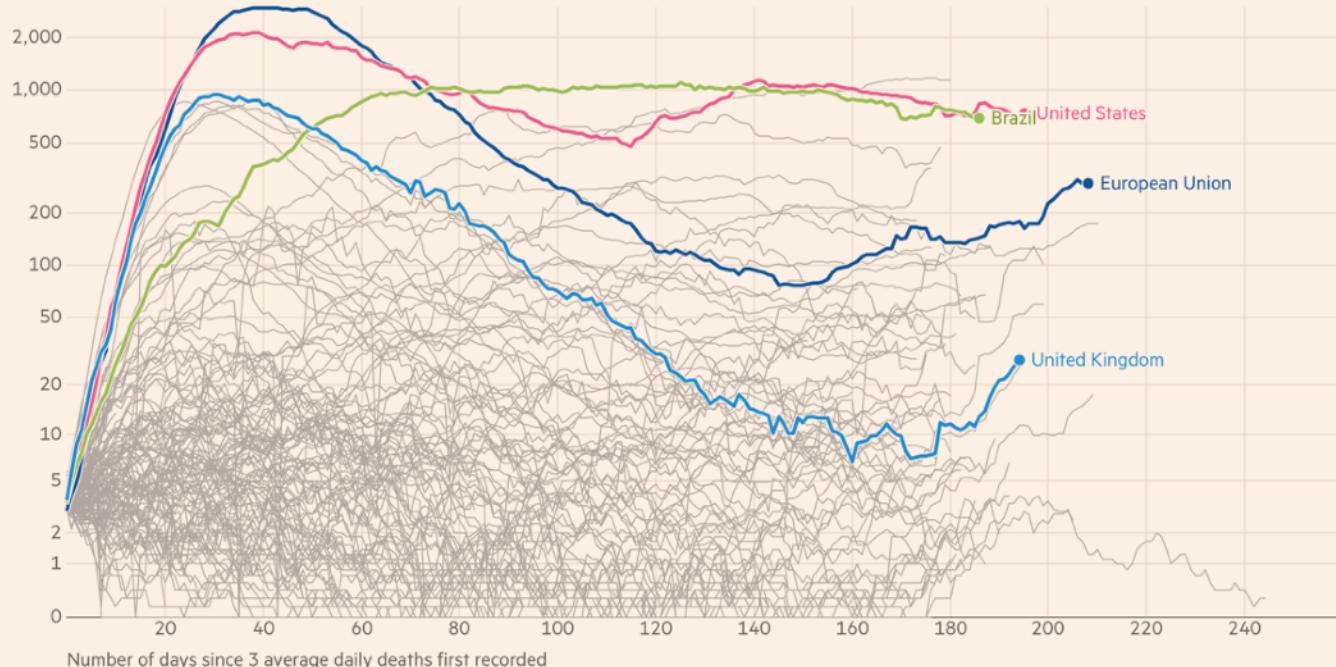


Visualizing Black America, Du Bois et al. 1900

Bones in hand, Gray's Anatomy 1918 ed.

## New deaths attributed to Covid-19 in European Union, United States, Brazil and United Kingdom

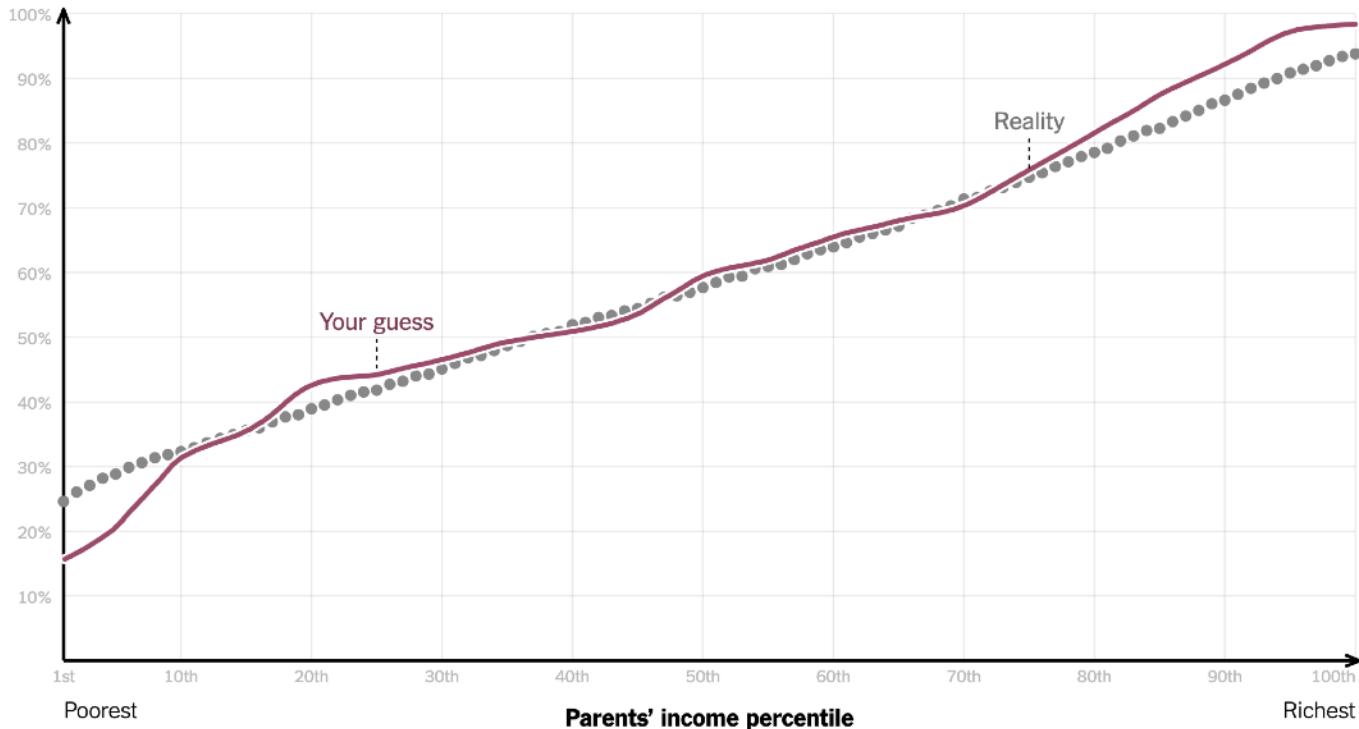
Seven-day rolling average of new deaths, by number of days since 3 average daily deaths first recorded



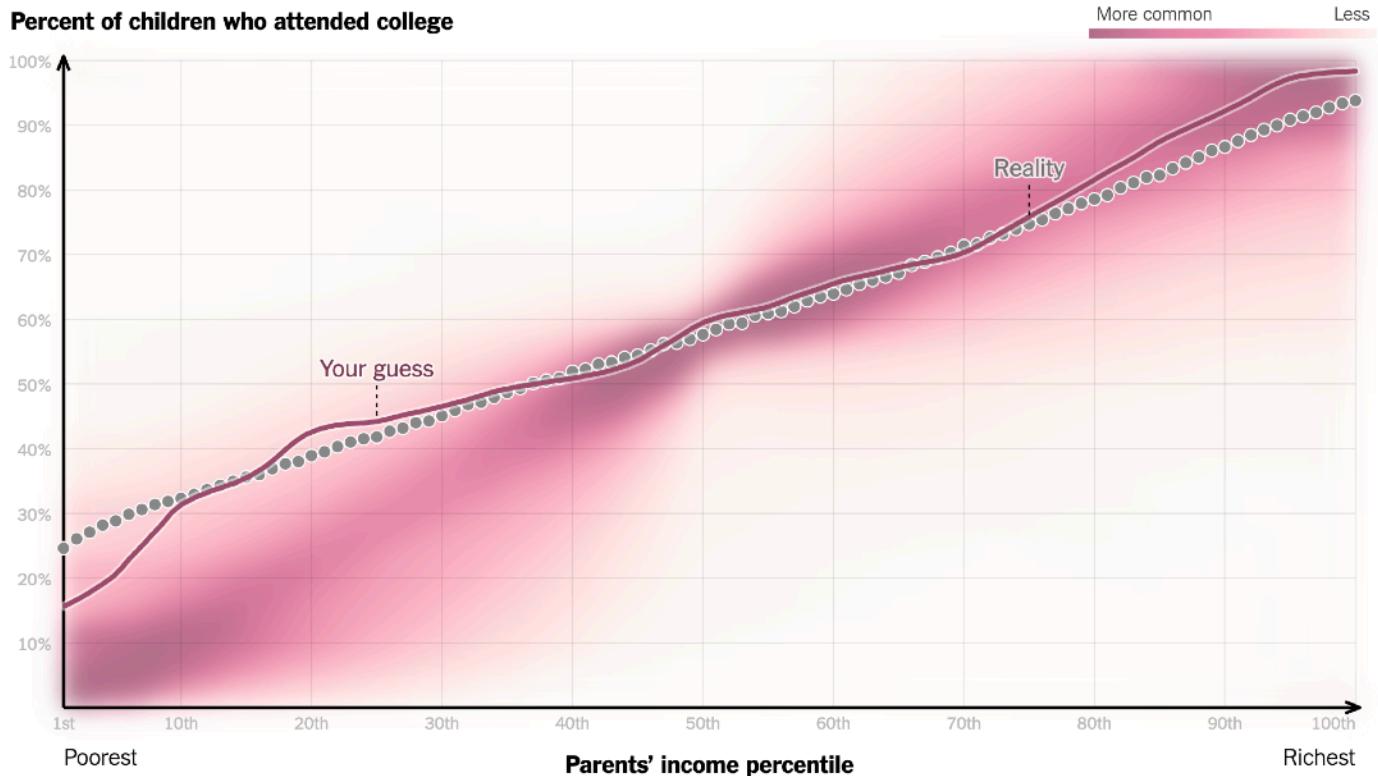
Source: Financial Times analysis of data from the European Centre for Disease Prevention and Control, the Covid Tracking Project, the UK Dept of Health & Social Care and the Spanish Ministry of Health.  
Data updated September 25 2020 12.46pm BST. Interactive version: [ft.com/covid19](http://ft.com/covid19)

FINANCIAL TIMES

### Percent of children who attended college



You Draw It: How Family Income Predicts Children's College Chances  
[New York Times, May 28, 2015]



You Draw It: How Family Income Predicts Children's College Chances  
[New York Times, May 28, 2015]

# Course Overview

# Course Overview

**W1:** Introduction & Visualization Tools Part 1

**W2:** Visual Encoding & Deceptive Visualization

**W3:** Data Transformation & Dimensionality Reduction

**W4:** Interaction & Mapping

**W5:** Visualization Tools Part 2 & D3.js Tutorial

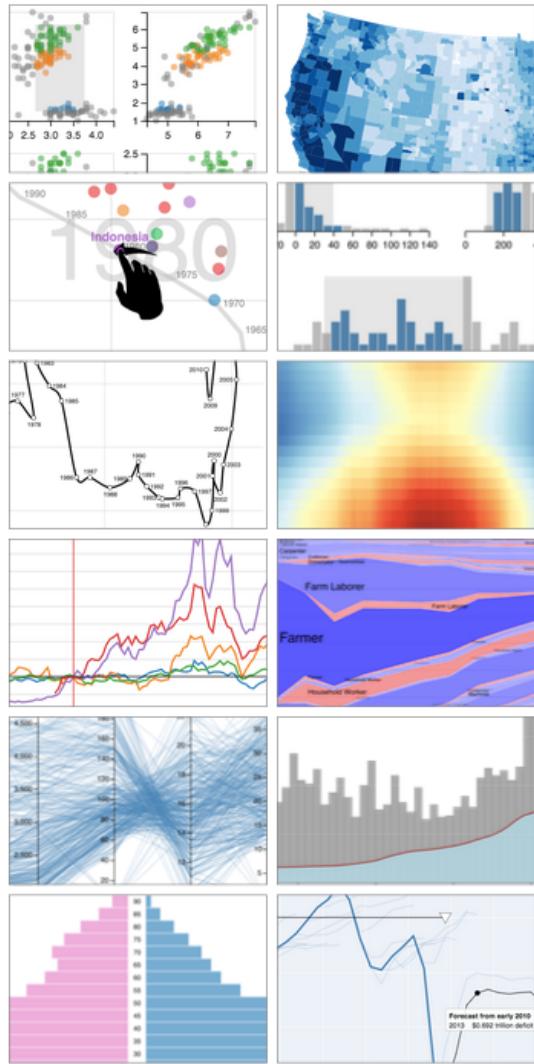
**W6:** Animation & Color

**W7:** Perception & Final Project Kick-Off

**W8:** Networks & Uncertainty

**W9:** Scalable Visualization & Final Project Peer Review

**W10:** Evaluation & Final Project Showcase



# Visual Encoding

LES VARIABLES DE L'IMAGE							12	14
XY 2 DIMENSIONS DU PLAN	POINTS	LIGNES	ZONES				OQ	≠
Z	TAILLE						OQ	≠
	VALEUR						O	≠
LES VARIABLES DE SÉPARATION DES IMAGES							13	
GRAIN							O	≠
COULEUR							≡	≠
ORIENTATION							≡	≠

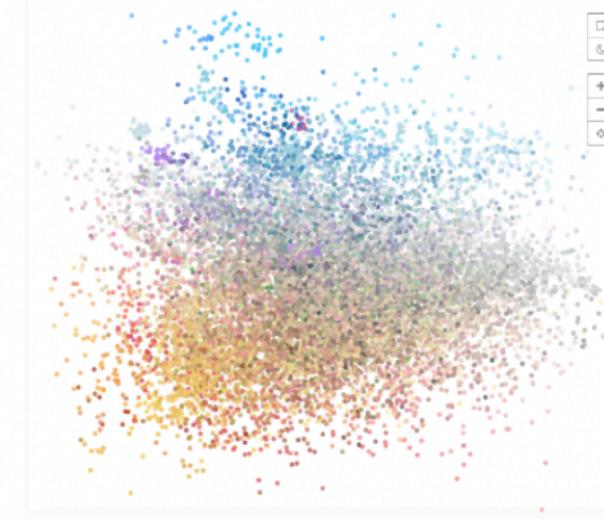
# Data Transformation



t-SNE

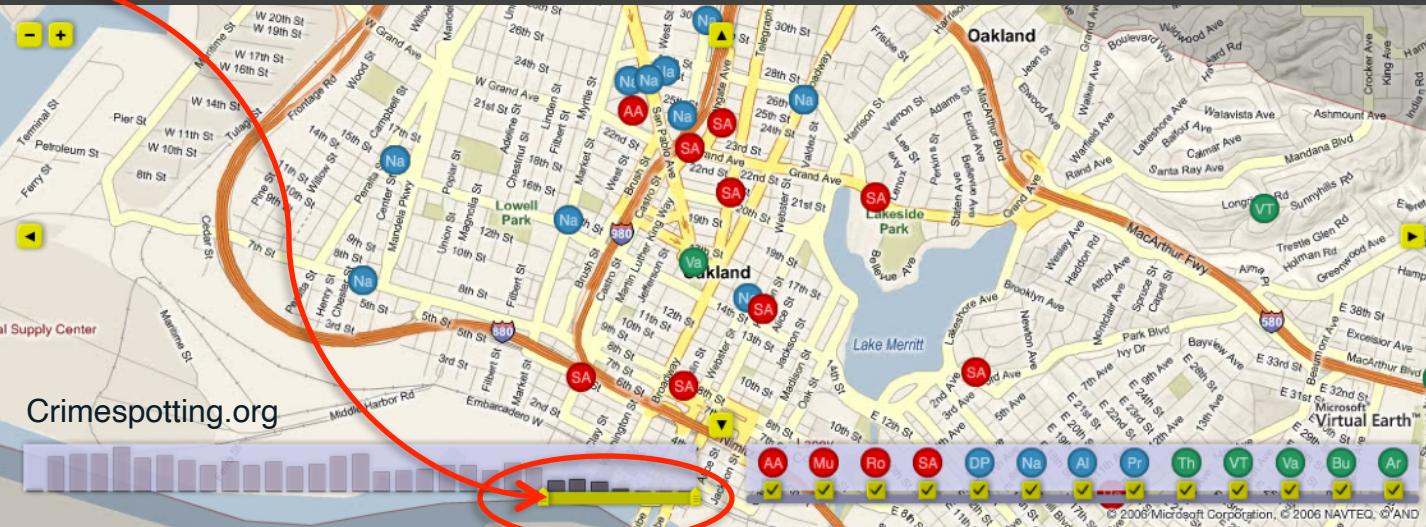
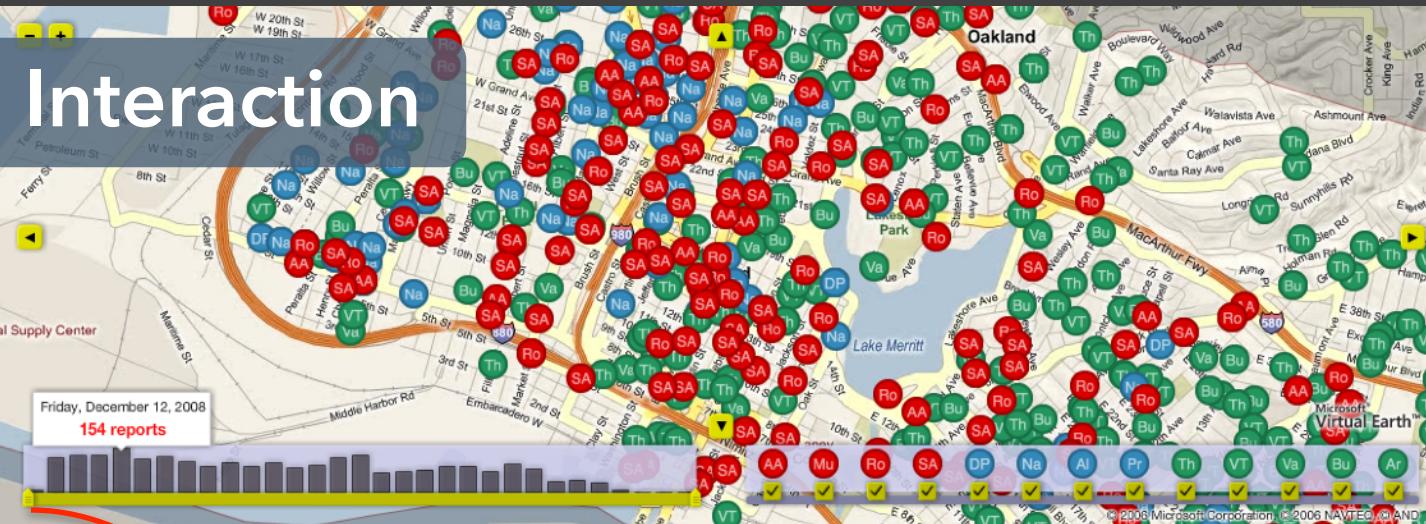


UMAP

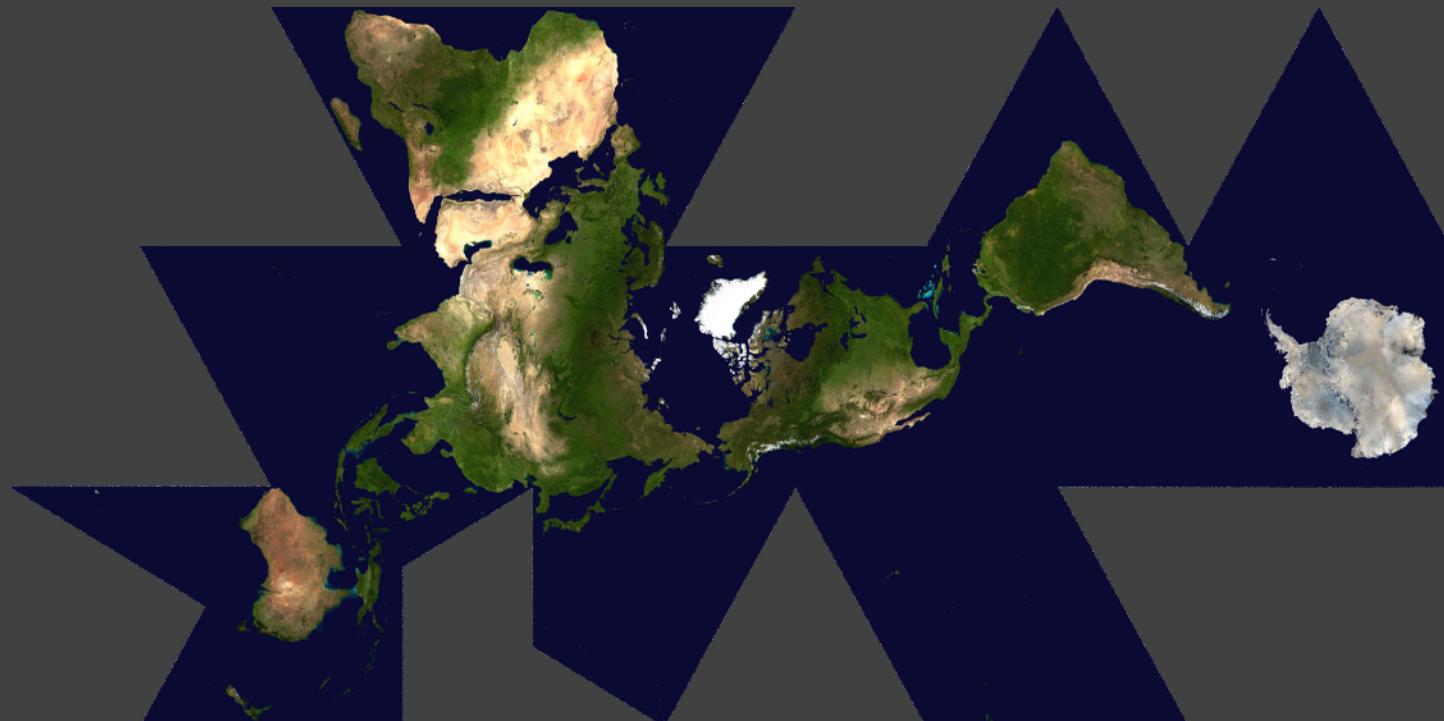


PCA

# Interaction

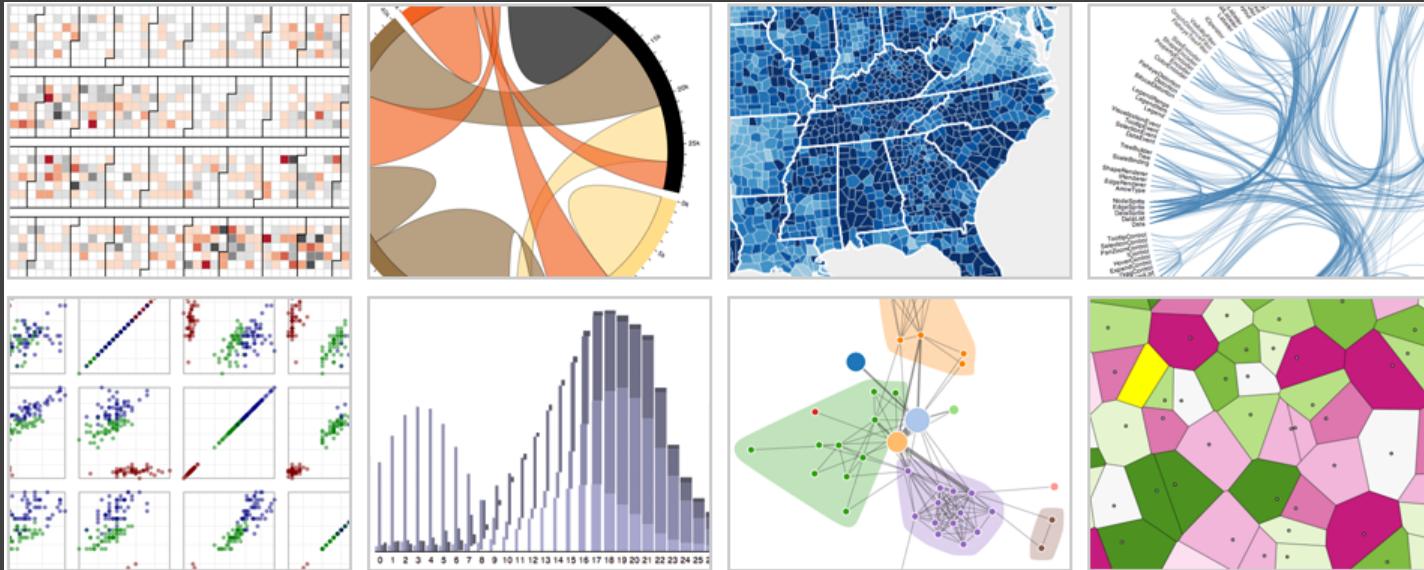


# Mapping & Cartography



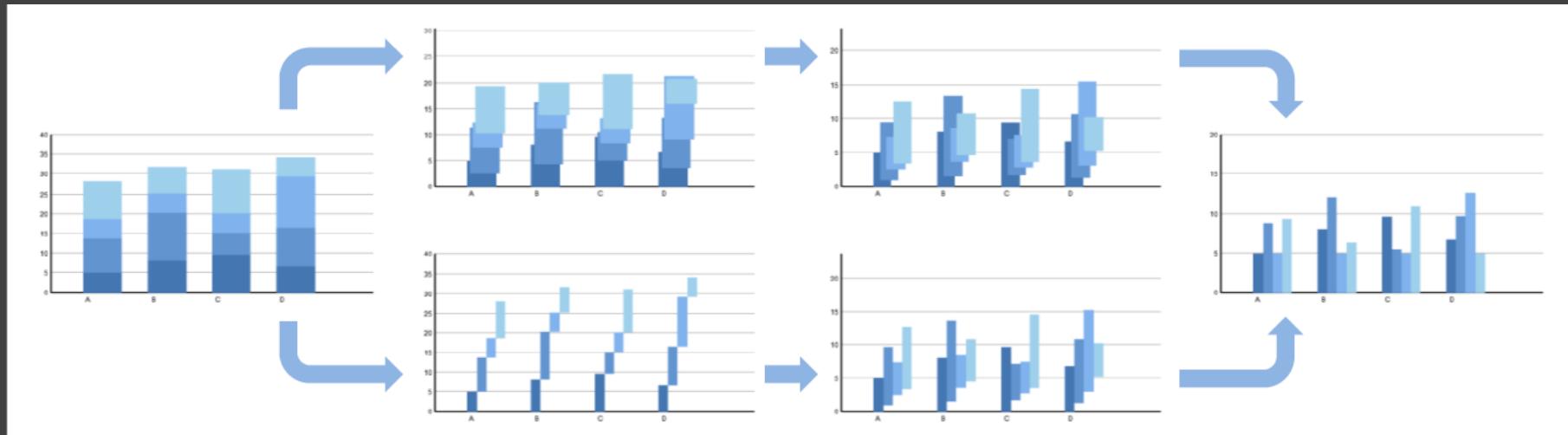
Dymaxion Maps [Fuller 46]

# Visualization Tools



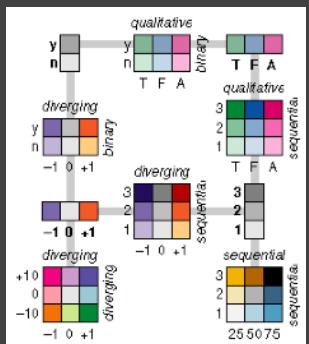
D3: Data-Driven Documents

# Animation

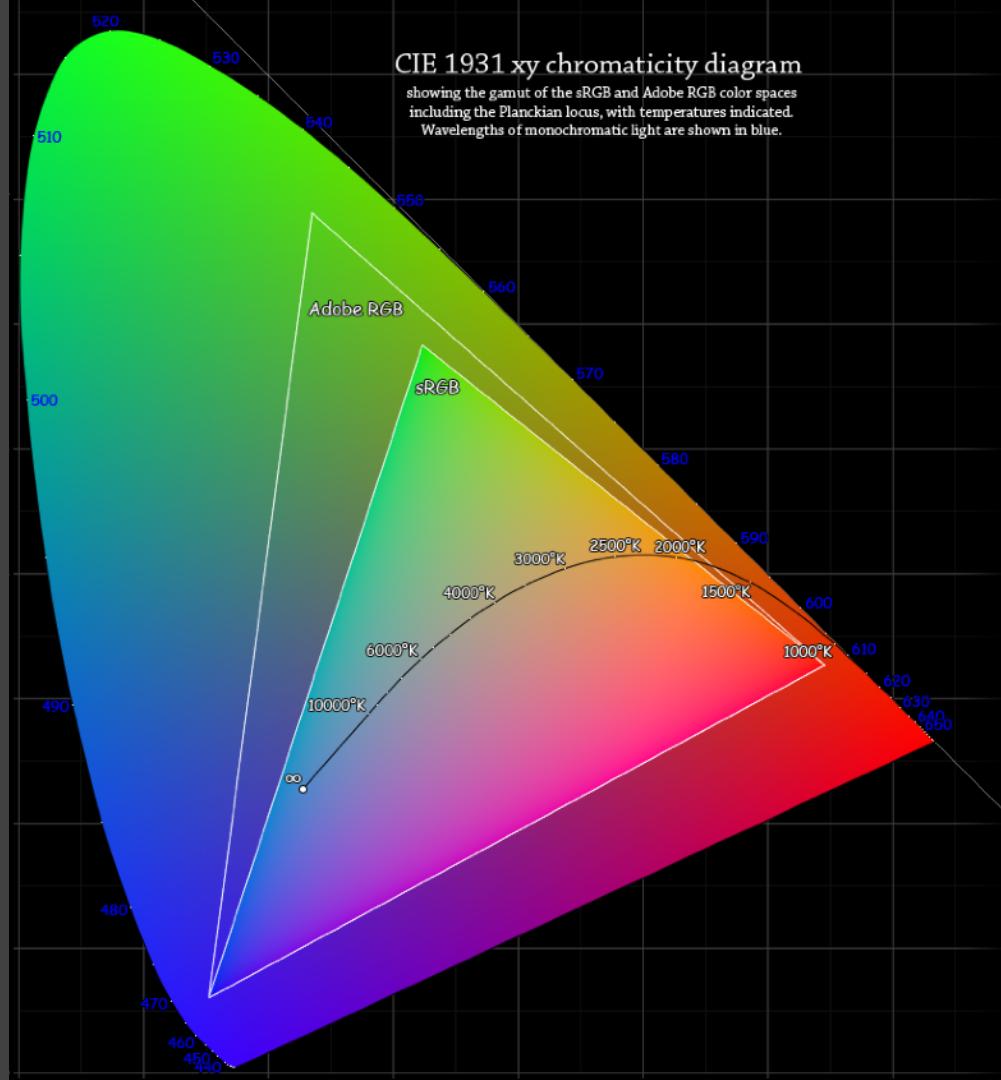


Animated transitions in statistical data graphics [Heer & Robertson 07]

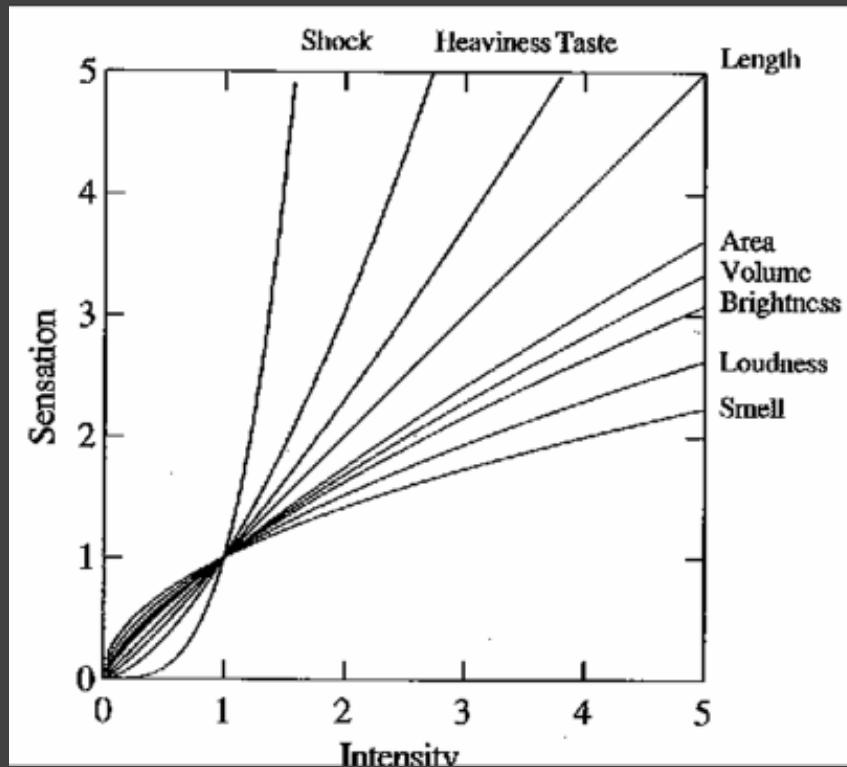
# Color



## Color Brewer

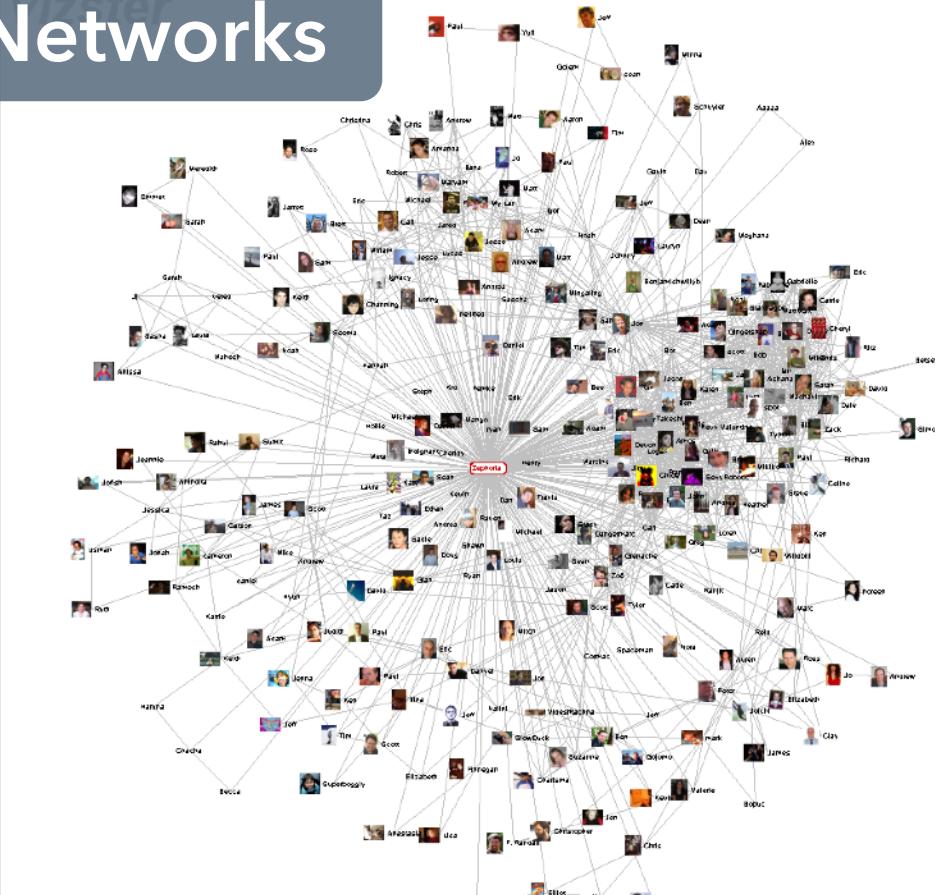


# Graphical Perception



The psychophysics of sensory function [Stevens 61]

# Networks



community &gt;&gt;

Enable

search &gt;&gt;

## Zephoria

**User ID** 21721  
**Friends** 266  
**Age** ??  
**Gender** Female  
**Status** Single  
**Location** San Francisco, CA  
**Hometown** Lancaster, PA  
**Occupation** researcher: social networks, identity, context  
**Interests** apophenia, observing people, culture, questioning power, reading, buddhism, ipsoity, computer-mediated communication, social networks, technology, anthropology, stomping, psytrance/goaltrance [Infected Mushroom, Son Kite, IbogaDigital Structures], Ani Difranco, downtempo, Thievery Corporation, Beth Orton, Morpheme, Ween, White Stripes  
**Music** Authors: Irving Goffman, Stanley Milgram, Jeanette Winterson, Eric Schlosser, Leslie Feinberg, Dorothy Allison, Italo Calvino, Hermann Hesse  
**Books** ??  
**TV Shows** Koyaanisqatsi, Amelie, Waking Life, Tank Girl, The Matrix, Clockwork Orange, American Beauty, Fight Club, Boys Don't Cry  
**Movies** ??  
**Member Since** 2003-10-21  
**Last Login** 2003-10-21  
**Last Updated** 2003-10-21  
**About** [Some know me as danah...]  

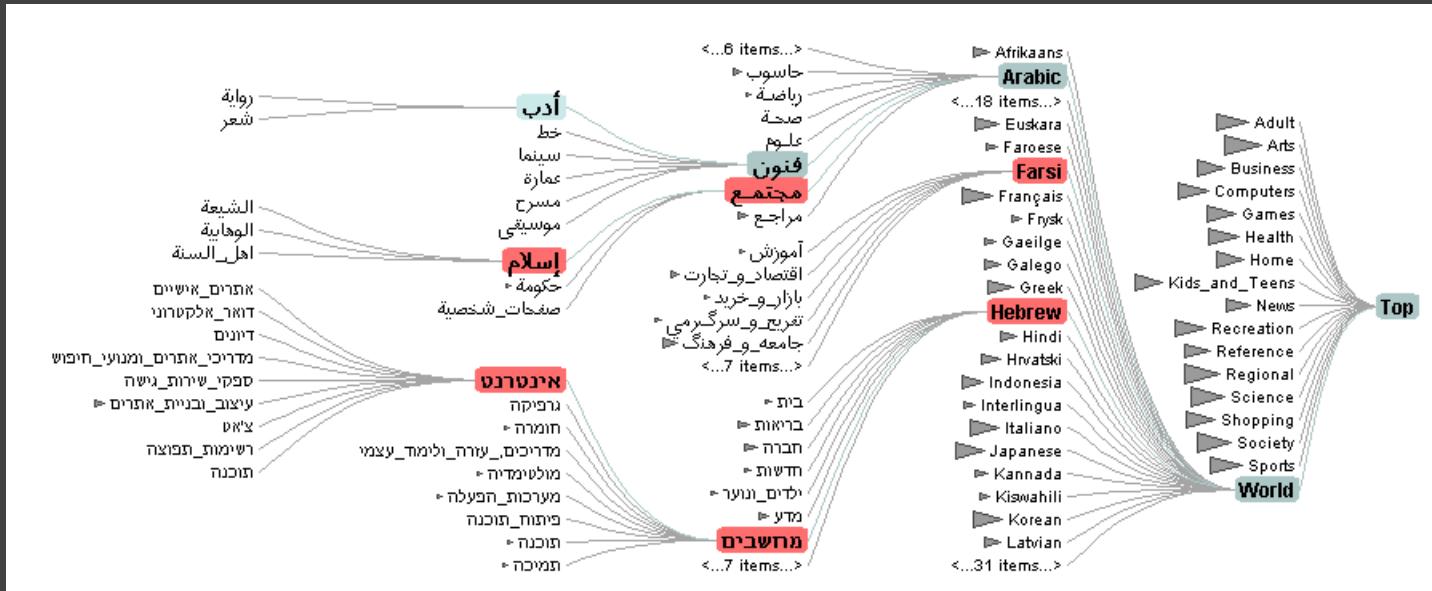
I'm a geek, an activist and an academic, fascinated by people and society. I see life as a very large playground and enjoy exploring its intricacies. I revel in life's chaos, while simultaneously providing my own insane element.

My musings:  
<http://www.zephoria.org/thoughts/>

Want to Meet

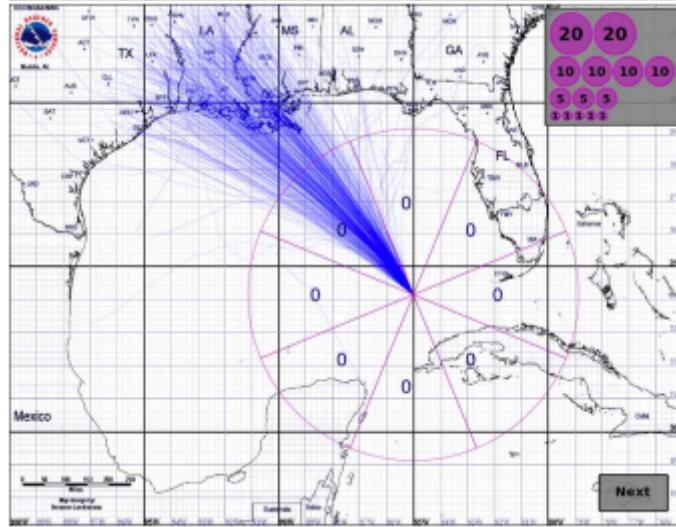
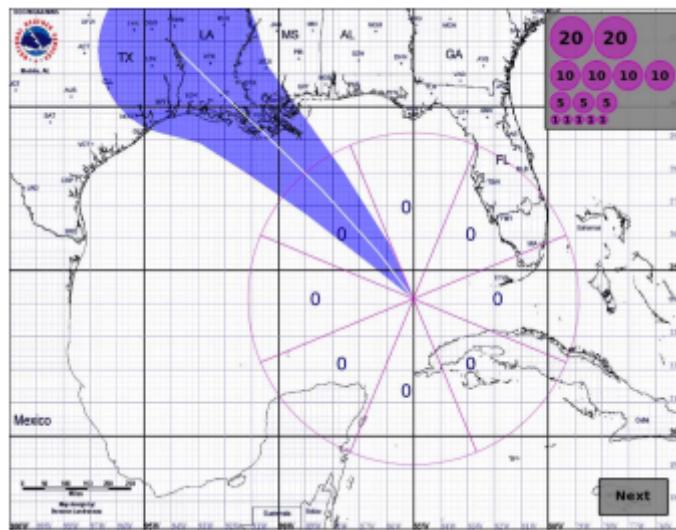
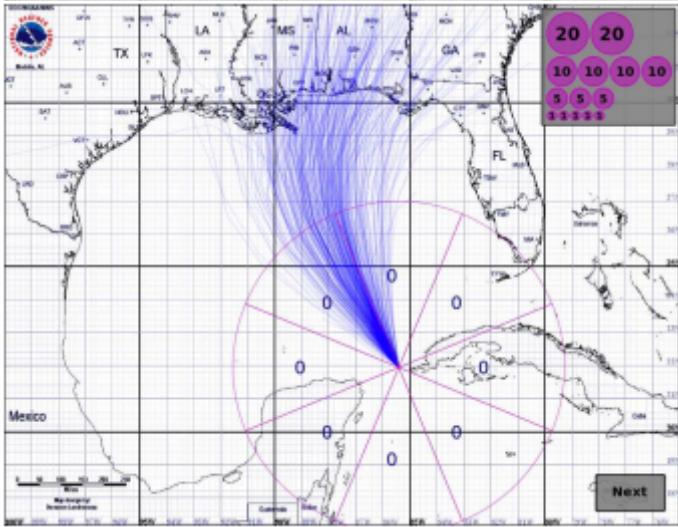
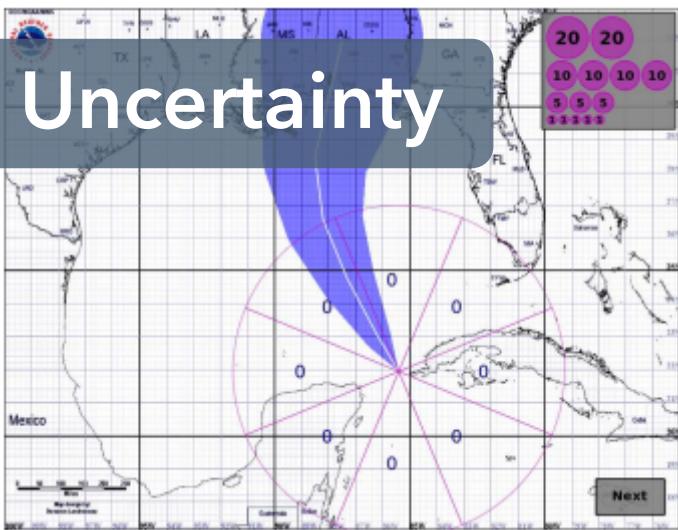
Someone who makes life's complexities seem simply elegant.

# Networks



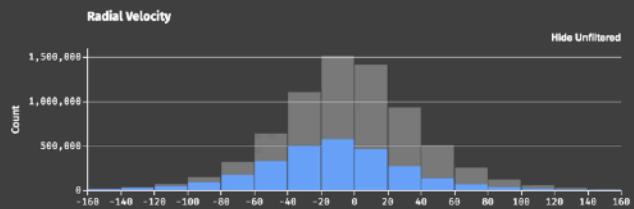
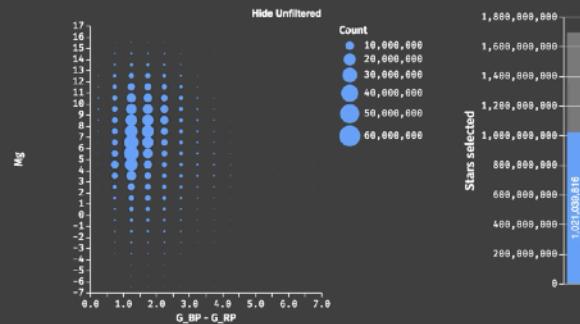
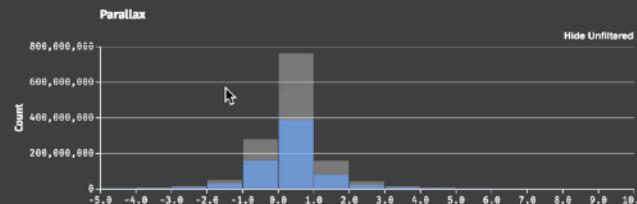
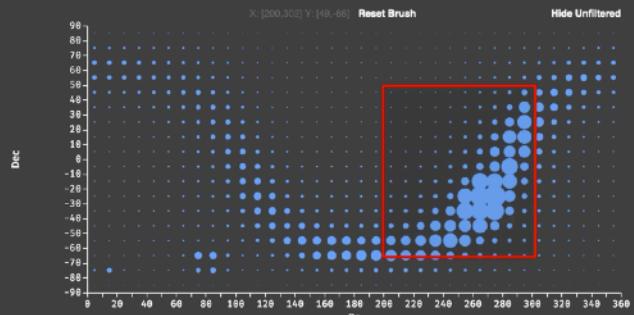
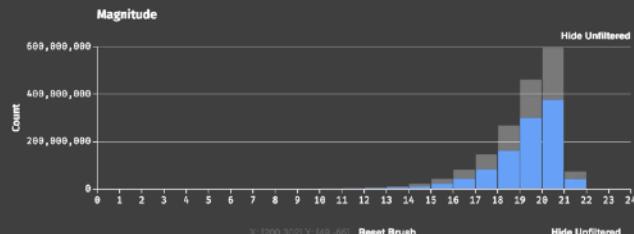
## Degree-Of-Interest Trees [Heer & Card 04]

# Uncertainty



# Scalability

localhost:1234



Interactive querying of 1.7B stars  
(1.2TB) in Falcon [Moritz et al. 2019]

Powered by Falcon 0.13.0

# You should expect to:

- 1 *Evaluate and critique* visualization designs
- 2 *Learn* visualization techniques & theory
- 3 *Implement* interactive data visualizations
- 4 *Develop* a substantial visualization project

# Instructors

cse442@cs

## *Lead Instructor*

**Jeffrey Heer** OH: Tue 10-11am, Gates 302

## *Teaching Assistants*

**Soham Bhosale** OH: Fri 1:30-2:30pm, Gates 121

**Shaan Chopra** OH: Wed 9:30-10:30am, Zoom

**Jason Kim** OH: Online / By Appointment

**Sebastin Santy** OH: Online / By Appointment

**Luke Snyder** OH: Thu 1-2pm, Gates 153

**Jiawen Zhu** OH: Mon 11am-12pm, Allen 624

# Soham Bhosale

- 3rd year undergrad
- **Interests:**
  - Data Science
  - Operating Systems and networks
  - ML
- **Hobbies**
  - Cooking and trying new recipes
  - Tennis
  - Traveling



# **Shaan Chopra** (she/her) | 5th year, CSE PhD

## Research:

- Human-Computer Interaction
- AI & Personal Health Informatics
- Participatory & Inclusive Design of Health Technologies
- Personal Data Use in Clinical Settings
- Health Equity & Community-Based Health

## Personal Interests / Hobbies:

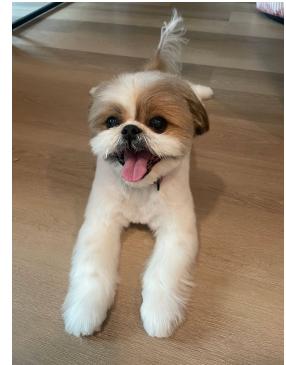
- Outdoor / sports: rowing, hiking, biking, basketball, discus throw, swimming...
- Creative: painting, live sketching, crocheting
- Food-related: drinking mochas, baking & eating desserts



# Jason Kim

4th Year Undergrad

- Former 457 TA (x2) and Now 442 TA
- **Email:** [kimjason@cs.washington.edu](mailto:kimjason@cs.washington.edu)
- **Interests:** Human-Computer Interaction, Autonomous Robotics, Computer Vision
- **Hobbies:** Hiking, fishing, cooking, tennis, video games, ice hockey, and working on my car.



# Sebastin Santy

5th year PhD student

Curious about what intelligence is —

*Human and Machine*

# Luke Snyder

5th year CSE PhD

Office hours: Th 1-2 PM

## Research

- Data visualization & HCI
- Tools for interactive visualization
- Human-AI workflows for scholarly synthesis

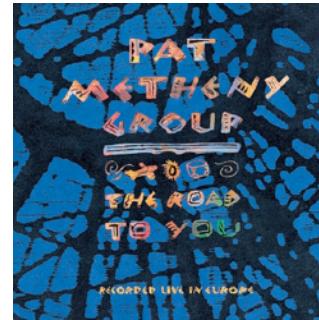


## Technical Expertise

- JavaScript / web programming, D3, Vega-Lite

## For Fun

- Music, running, travel
- Progressive Jazz, Weather Report, Pat Metheny Group





# Jiawen Stefanie Zhu

## About Me

2nd Year CSE PhD Student

*Advisors:* Katharina Reinecke, Tanu Mitra

*Hobbies:* martial arts, piano, watercolour painting

## Research Interests

My broad interest lies in Human-Computer Interaction (HCI), with a focus on augmenting human cognition.

E.g. foster diverse information consumption by facilitating the discovery and sensemaking of cross-lingual information



[stef-zjw.github.io](https://stef-zjw.github.io)



[jiawenz2@uw.edu](mailto:jiawenz2@uw.edu)

# Lectures, Activities & Office Hours

Course sessions will alternate among lecture and in-class exercises (often Thursdays).

All lectures will be in-person and recorded.

Please attend in-person but **NOT** if you feel ill.

Office hours are a mix of in-person and Zoom.

Links for virtual office hours are on Canvas.

Use Ed to post questions and seek help!

# Readings

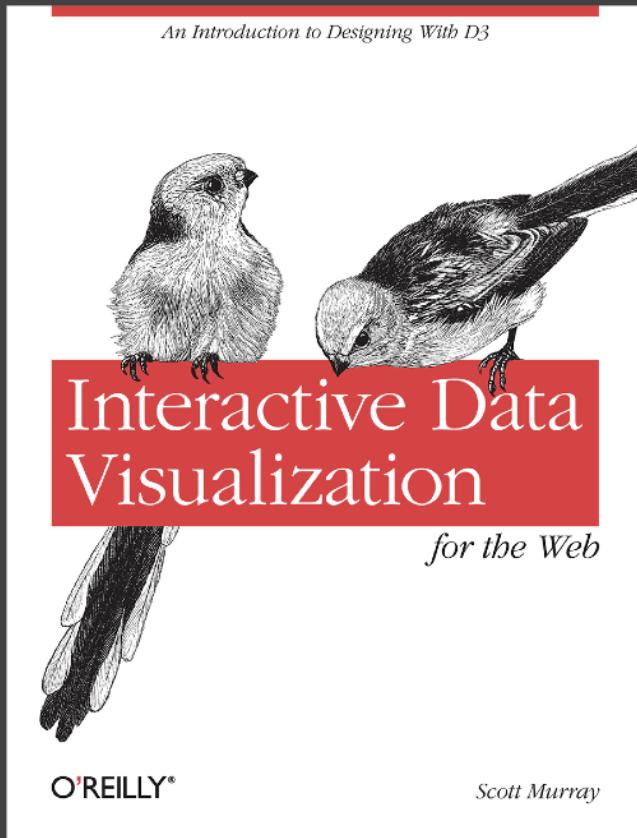
There is no one universal textbook on visualization!

We draw on books, notebooks, and linked articles...

Material in class will loosely follow readings.

Readings should be read by start of class.

# Textbook



## Interactive Data Visualization for the Web, 2nd Edition

*For learning D3!*

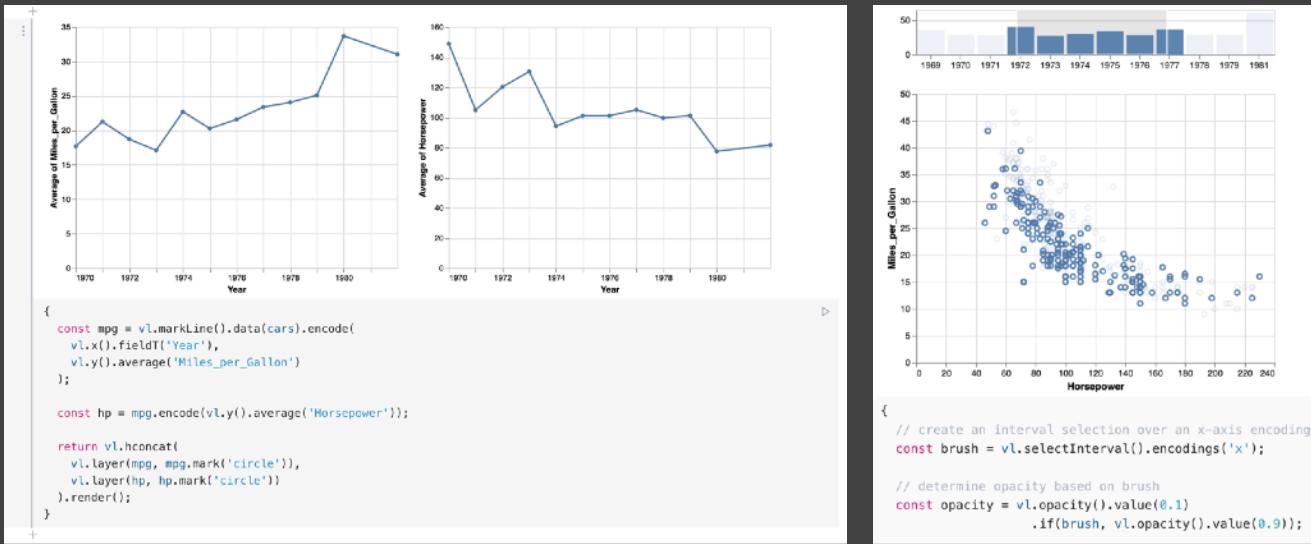
Book available online.

Code / examples on GitHub.

We will be using D3 v7.

<https://d3js.org>

# Interactive Notebooks



Hands-on engagement with course concepts and tools using Observable (JavaScript) notebooks.

# Assignments

**CP** Class Participation (10%)

**A1** Expository Visualization (10%)

**A2** Deceptive Visualization (15%)

Peer Review (5%)

**A3** Interactive Prototype (20%)

Peer Review (5%)

**FP** Final Project (35%)

Proposal

Demonstration Video

Final Prototype

# Grading Philosophy

A *great* submission gets a *great grade* (A- to A, 3.6 - 3.8), but an *exceptional* grade (A+, 3.9 - 4.0) requires *exceptional* effort.

## **Example: Typical A1 grades (out of 10 points).**

Everyone starts with a high score (9/10).

We then *deduct* points for errors and also *add* points for going above and beyond the assignment requirements.

The median score for A1 is typically 8.5 out of 10, which maps to an A-.

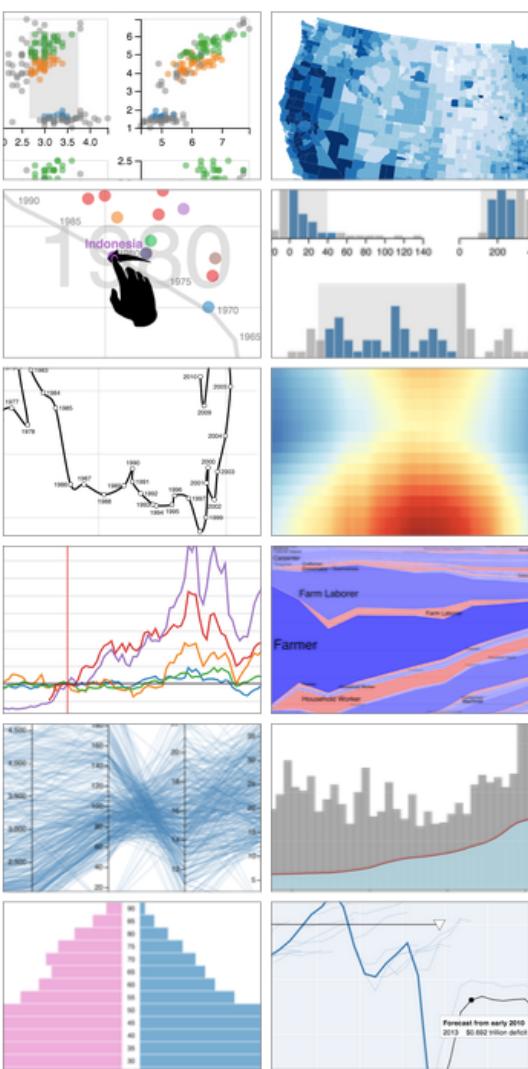
# Course Participation

## Lecture Attendance

Please attend lectures in person! That said, we know that illness, travel, etc. can prevent attendance. If you can't attend class, please review the recordings online.

## Weekly Exercises

We have in-class exercises each week. Complete them even if you can't attend in person. We use "best-effort" grading, so it's OK if you don't complete everything during class time. Focus on assignments, not exercises, between sessions. You also get one exercise "pass".



# Warm-Up Design Activity

# Visual Encoding Exercise

5 17

How many visualizations can you think of for conveying these two numbers? Feel free to invent tasks or contexts. **Sketch as many as you can!**

*Don't stress over quality, go for quantity.*

Time: ~5 minutes

# Visual Encoding Exercise

5 17

Take a photo or screenshot of your visualizations, and post it to the shared thread on Ed.

# Visual Encoding Exercise

5 17

Share your designs with students near you. Introduce yourselves! Then compare your designs. How many ideas are the same? How many are different?

What do you find highly effective? Highly creative?

# Visual Encoding Design

# The Big Picture

task

questions, goals  
assumptions

data

physical data type  
conceptual data type

domain

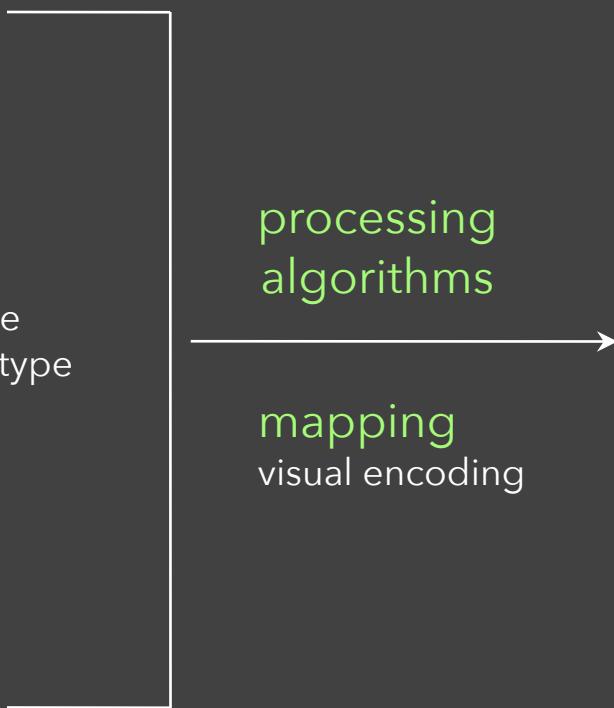
metadata  
semantics  
conventions

processing  
algorithms

mapping  
visual encoding

image

visual channel  
graphical marks



# Data Models

Represent data as a *table* (relation)

Each *row* (tuple) represents a record

Each *column* (field) represents a typed variable

*Physical Type*: integer, float, date, boolean, string...

*Conceptual Type*: temperatures, dollars, products...

For visualization it is helpful to classify fields according to the type of comparisons we wish to make:

*Nominal* (N), *Ordinal* (O), and *Quantitative* (Q) types

# Nominal, Ordinal & Quantitative

# Nominal, Ordinal & Quantitative

N - Nominal (labels or categories)

- Fruits: apples, oranges, ...

# Nominal, Ordinal & Quantitative

N - Nominal (labels or categories)

- Fruits: apples, oranges, ...

O - Ordered

- Quality of meat: Grade A, AA, AAA

# Nominal, Ordinal & Quantitative

N - Nominal (labels or categories)

- Fruits: apples, oranges, ...

O - Ordered

- Quality of meat: Grade A, AA, AAA

Q - Interval (location of zero arbitrary)

- Dates: Jan, 19, 2006; Location: (LAT 33.98, LON -118.45)
- Only differences (i.e., intervals) may be compared

# Nominal, Ordinal & Quantitative

N - Nominal (labels or categories)

- Fruits: apples, oranges, ...

O - Ordered

- Quality of meat: Grade A, AA, AAA

Q - Interval (location of zero arbitrary)

- Dates: Jan, 19, 2006; Location: (LAT 33.98, LON -118.45)
- Only differences (i.e., intervals) may be compared

Q - Ratio (zero fixed)

- Physical measurement: Length, Mass, Time duration, ...
- Counts and amounts

# Nominal, Ordinal & Quantitative

N - Nominal (labels or categories)

- Operations:  $=, \neq$

O - Ordered

- Operations:  $=, \neq, <, >$

Q - Interval (location of zero arbitrary)

- Operations:  $=, \neq, <, >, -$
- Can measure distances or spans

Q - Ratio (zero fixed)

- Operations:  $=, \neq, <, >, -, \%$
- Can measure ratios or proportions

# Visual Language is a Sign System



Jacques Bertin

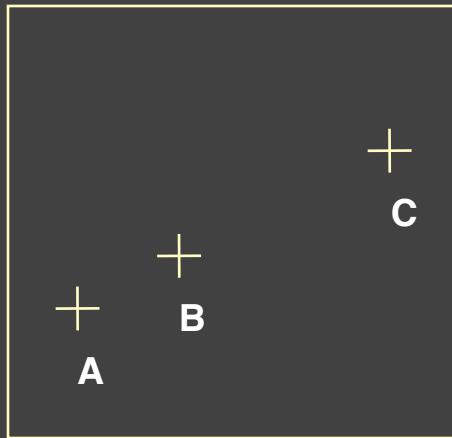
Images perceived as a set of signs

Sender encodes information in signs

Receiver decodes information from signs

Sémiologie Graphique, 1967

# Bertin's Semiology of Graphics



1. A, B, C are distinguishable
2. B is between A and C.
3. BC is twice as long as AB.

∴ Encode quantitative variables

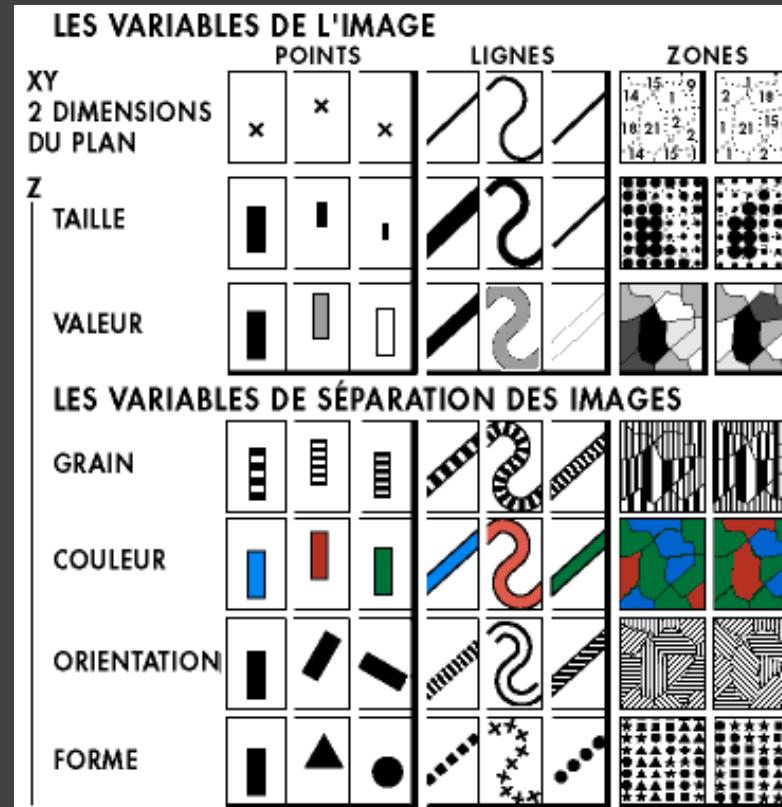
*"Resemblance, order and proportional are the three signfields in graphics."* - Bertin

## LES VARIABLES DE L'IMAGE

	POINTS	LIGNES	ZONES
XY 2 DIMENSIONS DU PLAN			
Z TAILLE			
VALEUR			
LES VARIABLES DE SÉPARATION DES IMAGES			
GRAIN			
COULEUR			
ORIENTATION			
FORME			

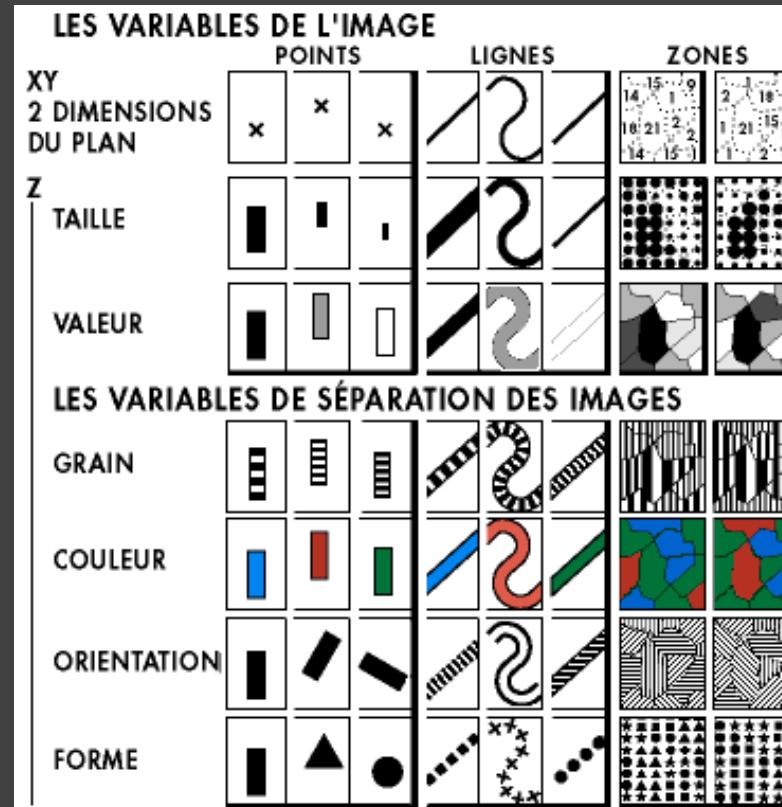
# Visual Encoding Channels

Position (x 2)  
Size  
Value  
Texture  
Color  
Orientation  
Shape

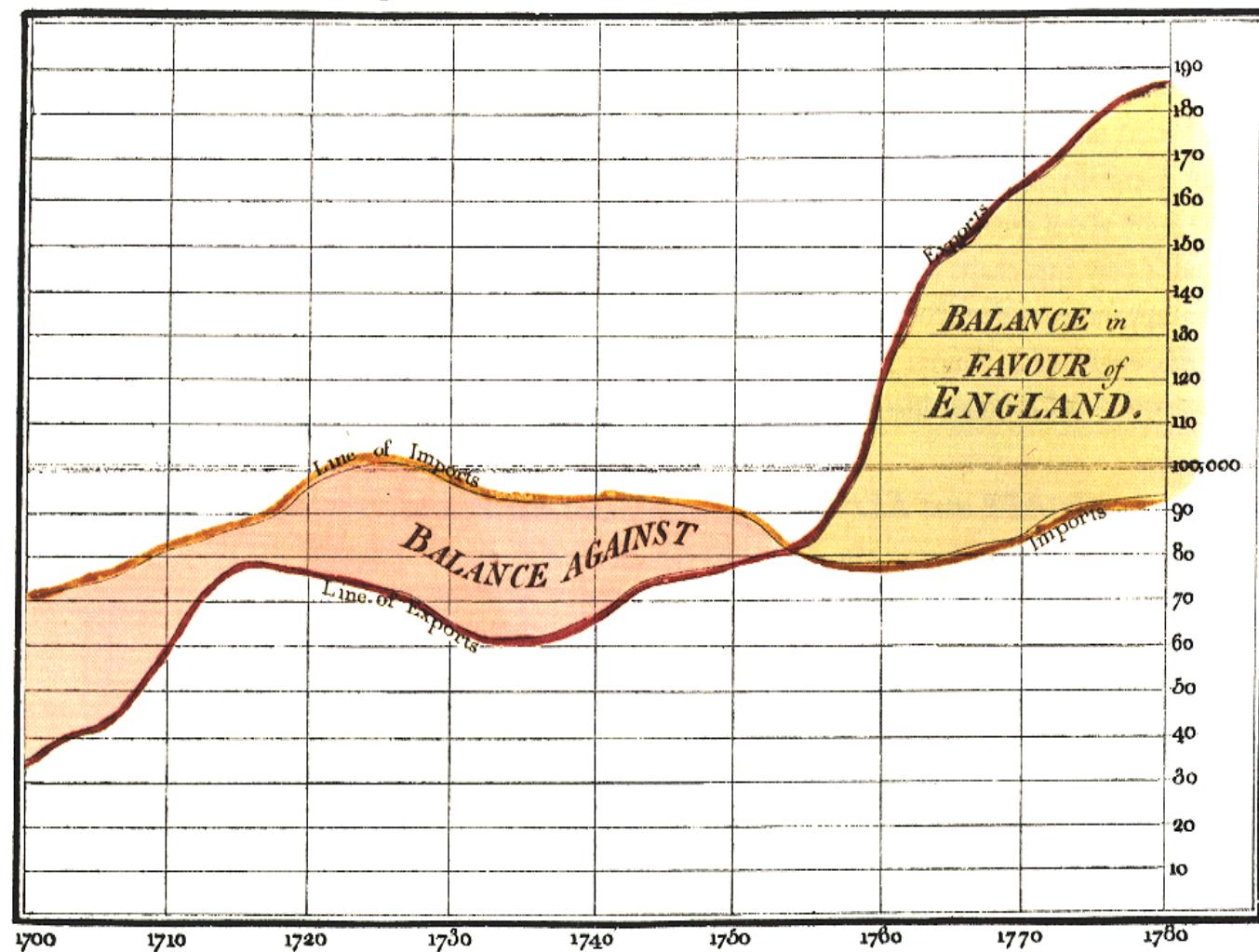


# Visual Encoding Channels

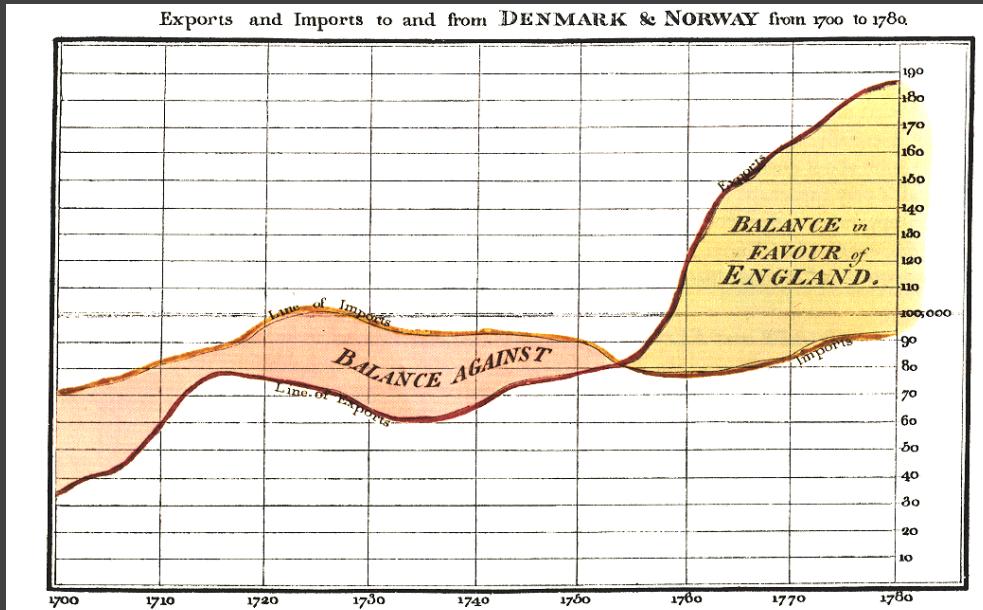
Position  
**Length**  
**Area**  
**Volume**  
Value  
Texture  
Color  
Orientation  
Shape  
**Transparency**  
**Blur / Focus ...**



Exports and Imports to and from DENMARK & NORWAY from 1700 to 1780.



# William Playfair, 1786



X-axis: year (Q)

Y-axis: currency (Q)

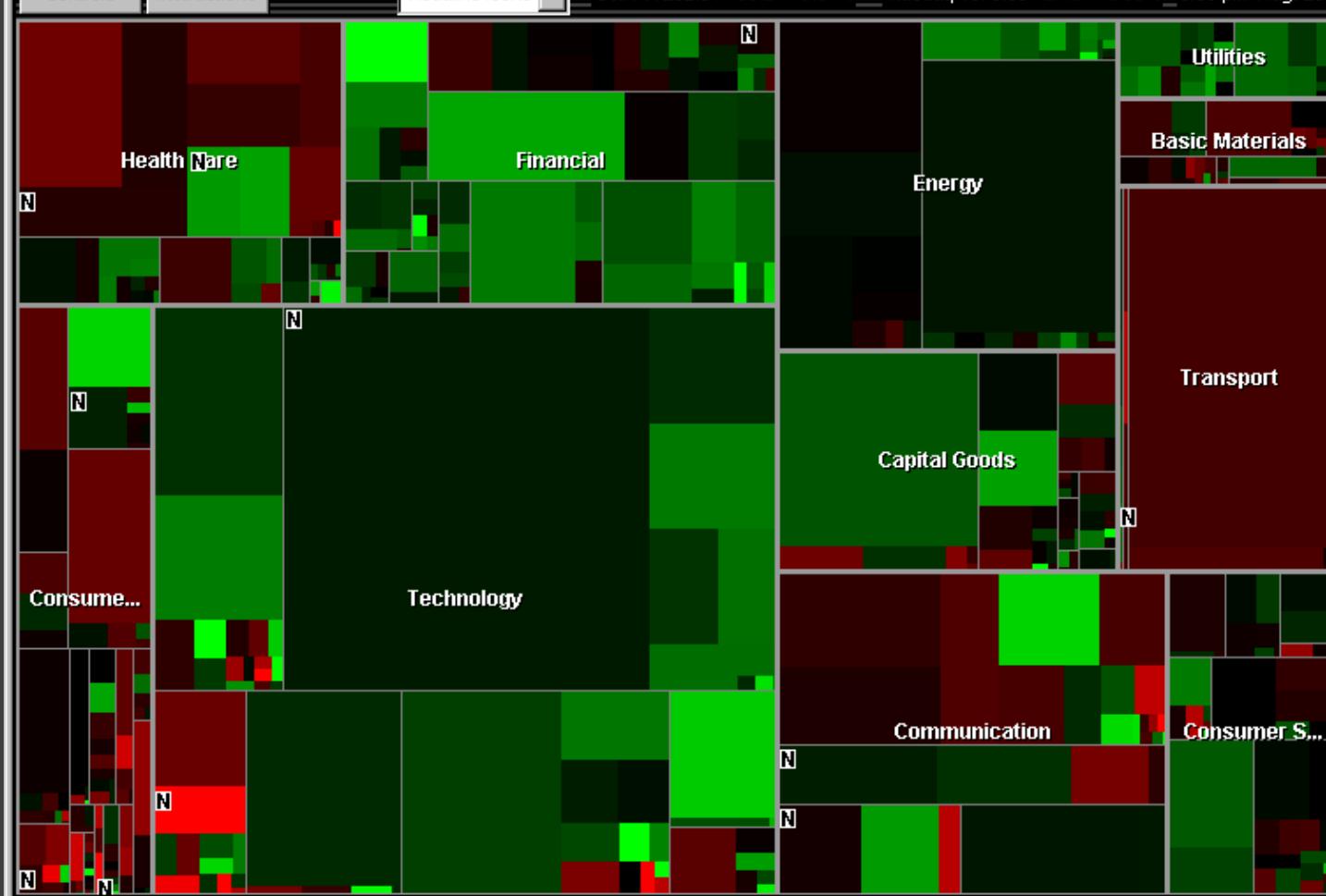
Color: imports/exports (N, O)

Controls

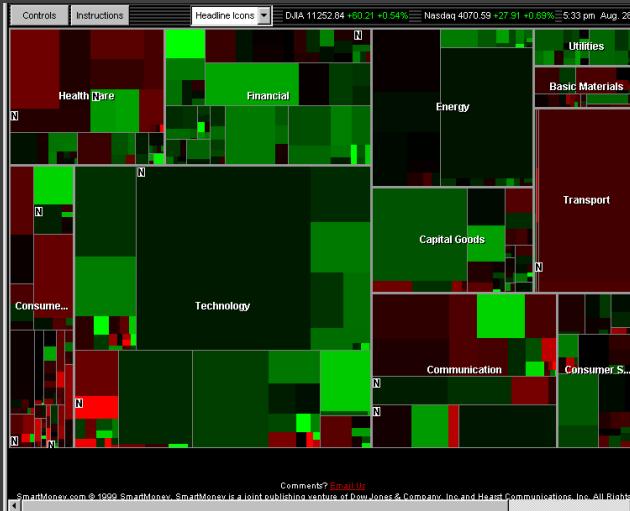
Instructions

Headline Icons

DJIA 11252.84 +60.21 +0.54% Nasdaq 4070.59 +27.91 +0.69% 5:33 pm Aug. 28



# Wattenberg's Map of the Market



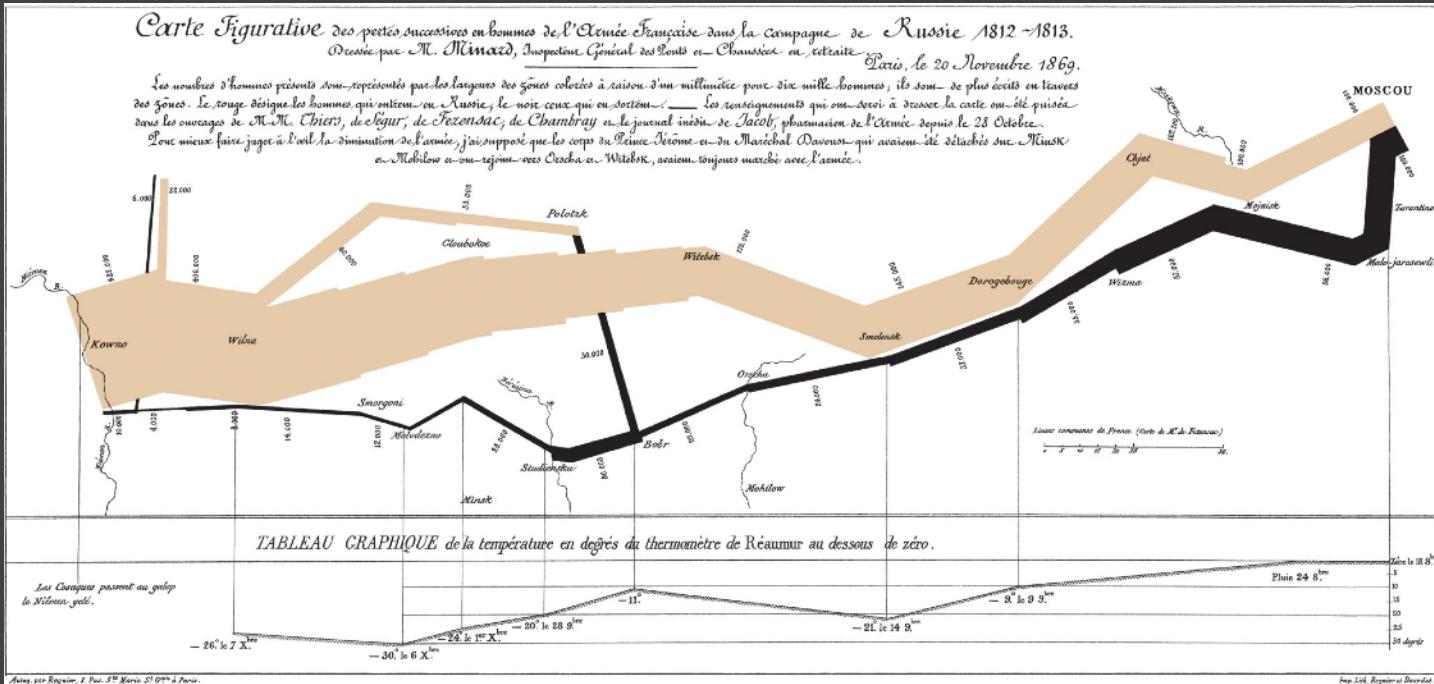
Rectangle Area: market cap (Q)

Rectangle Position: market sector (N), market cap (Q)

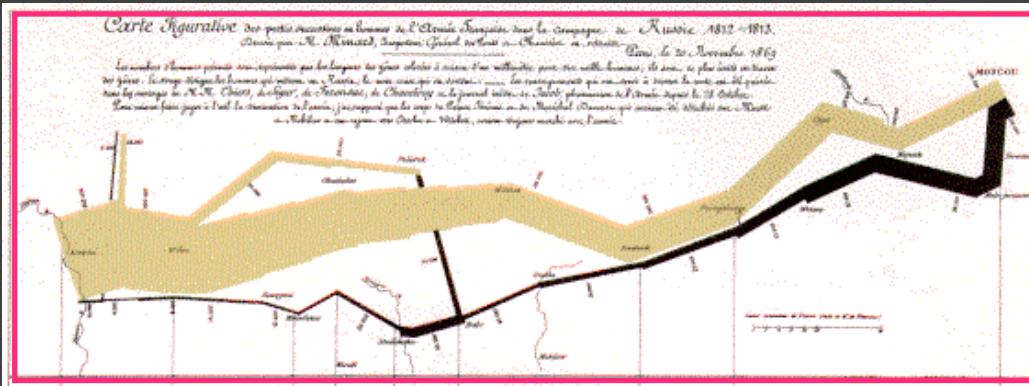
Color Hue: loss vs. gain (N, O)

Color Value: magnitude of loss or gain (Q)

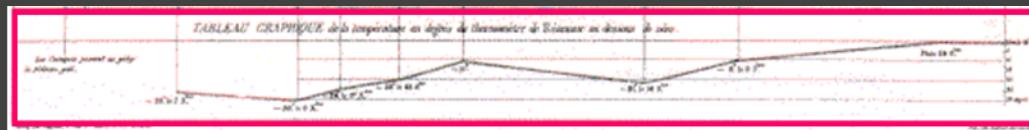
# Minard 1869: Napoleon's March



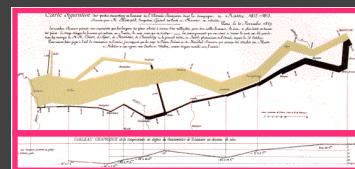
# Single-Axis Composition



+



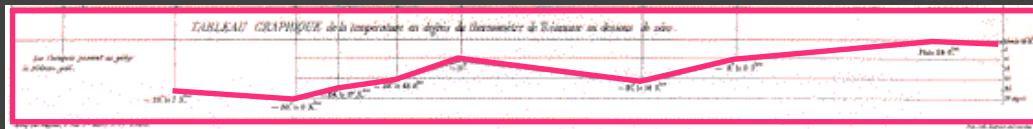
=



# Mark Composition

Y-axis: temperature (Q)

**+** X-axis: longitude (Q) / time (O)



## Temp over space/time (Q x Q)

# Mark Composition

Y-axis: latitude (Q)

+

X-axis: longitude (Q)

+

Width: army size (Q)



Army position ( $Q \times Q$ ) and army size (Q)

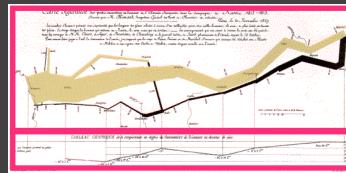
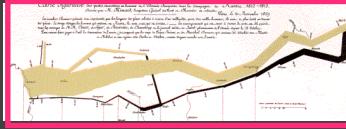
latitude (Q)

longitude (Q)

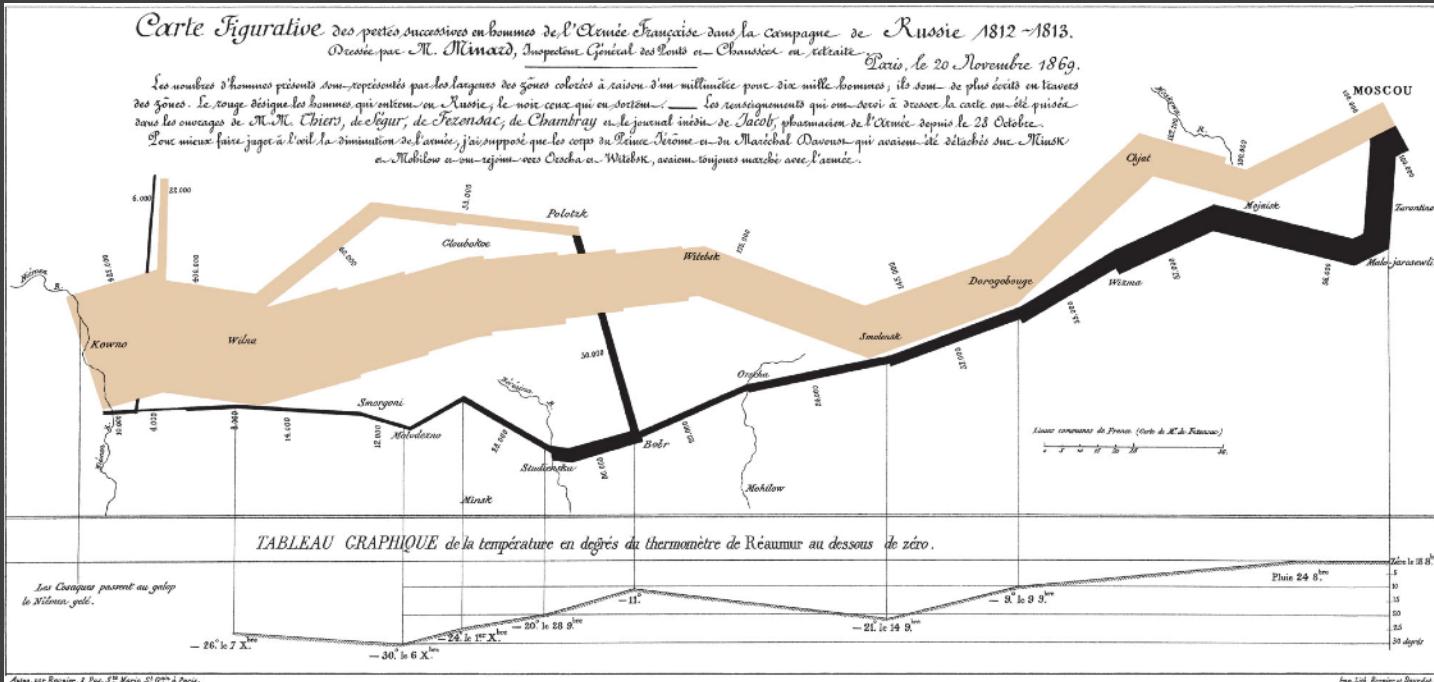
army size (Q)

temperature (Q)

longitude (Q) / time (O)



# Minard 1869: Napoleon's March



Depicts at least 5 quantitative variables. Any others?

# Assignment 1

# Hours of Sunshine

The climate of a place can have a tremendous impact on people's lived experience. You will examine average monthly climate measurements for six major U.S. cities, roughly covering the edges of the continental United States.

Our in-class **exercise** is to get hands-on experience creating and publishing visualizations. You will create and revise a line chart of average monthly sunshine hours for six U.S. cities.

The **assignment** is to then design your own graphic.

# A1: Expository Visualization

Using the given climate data set...

Pick a **guiding question**, use it to title your vis.

Design a **static visualization** for that question.

You are free to **use any visualization tool**.

***This builds on our next in-class exercise, be sure to read the exercise before diving in!!***

**Deliverables** via Gradescope

Image of your visualization (PNG or JPG format)

Short description + design rationale ( $\leq 4$  paragraphs)

Due by **EOD, Tue January 13**.