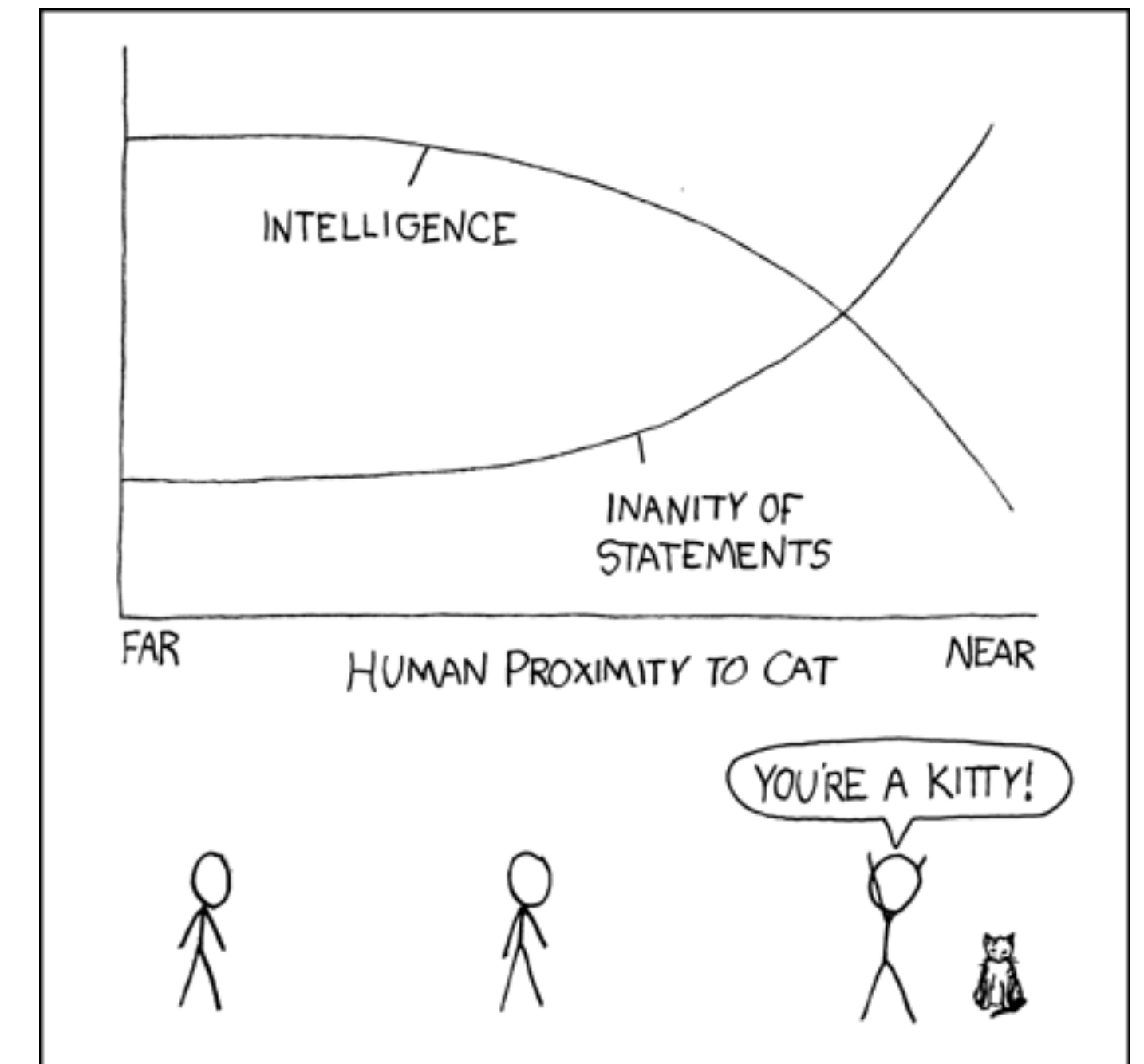


CS-5630 / CS-6630

Visualization for Data Science

Alexander Lex
alex@sci.utah.edu



visualization

pictures

The purpose of computing is insight, not numbers.

- Richard Wesley Hamming

- Card, Mackinlay, Shneiderman

Banana ***M. acuminata***

Date ***P. dactylifera***

Cress ***Arabidopsis thaliana***

Rice ***Oryza sativa***

Sorghum ***Sorghum bicolor***

Brome ***Brachypodium distachyon***



vi · su · al · i · za · tion

1. Formation of mental visual images
2. The act or process of interpreting in visual terms or of putting into visible form

Visualization Definition

Visualization is the process that **transforms** (abstract) **data** into **interactive graphical representations** for the purpose of **exploration, confirmation, or presentation.**

Good Data Visualization

- ... makes data **accessible**
- ... combines strengths of **humans and computers**
- ... enables **insight**
- ... **communicates**

Visualization

“Visualization is really about external cognition, that is, how resources outside the mind can be used to boost the cognitive capabilities of the mind.”



Stuart Card

Why Visualize?

To inform humans: Communication

How is ahead in the election polls?

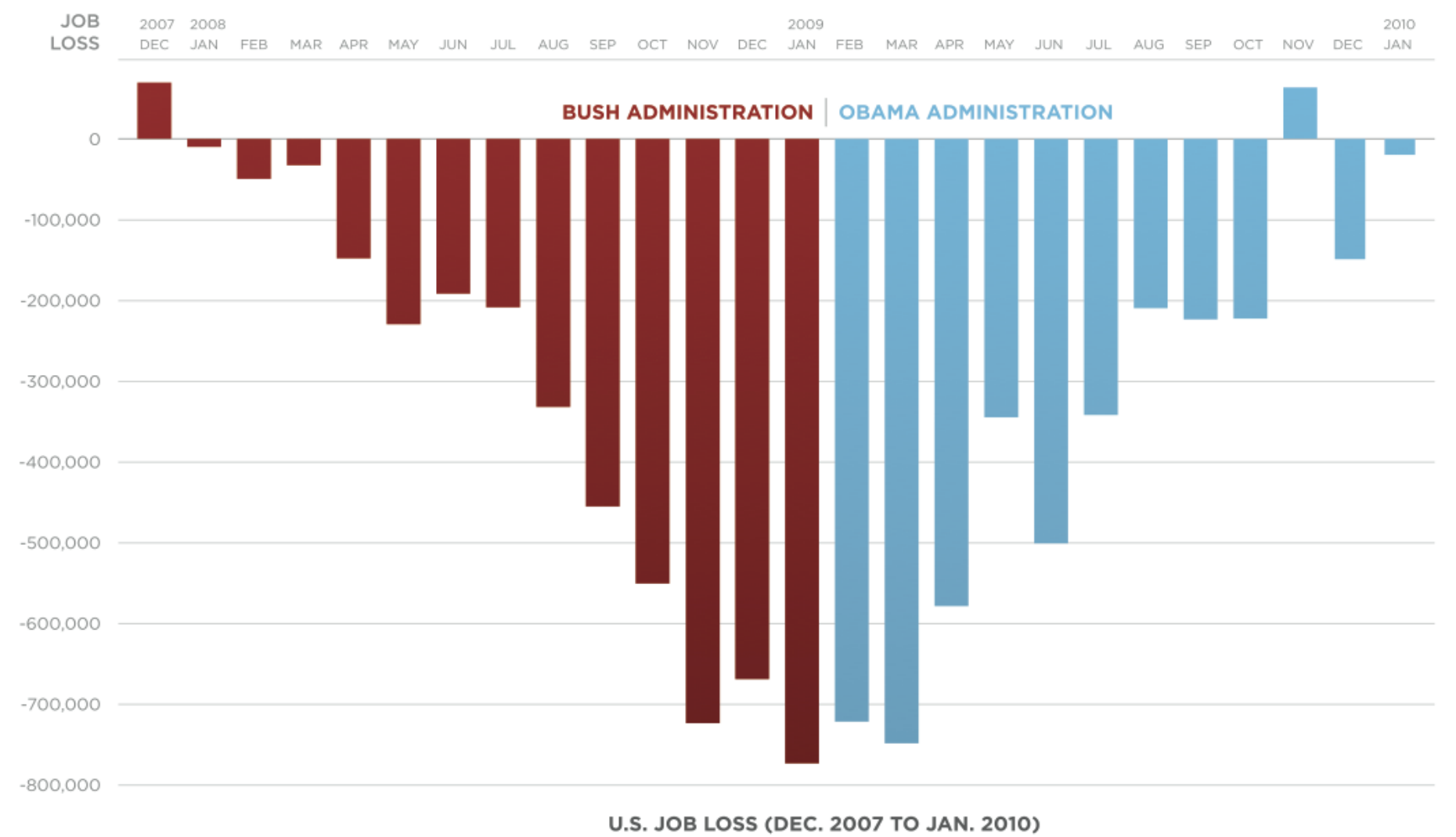
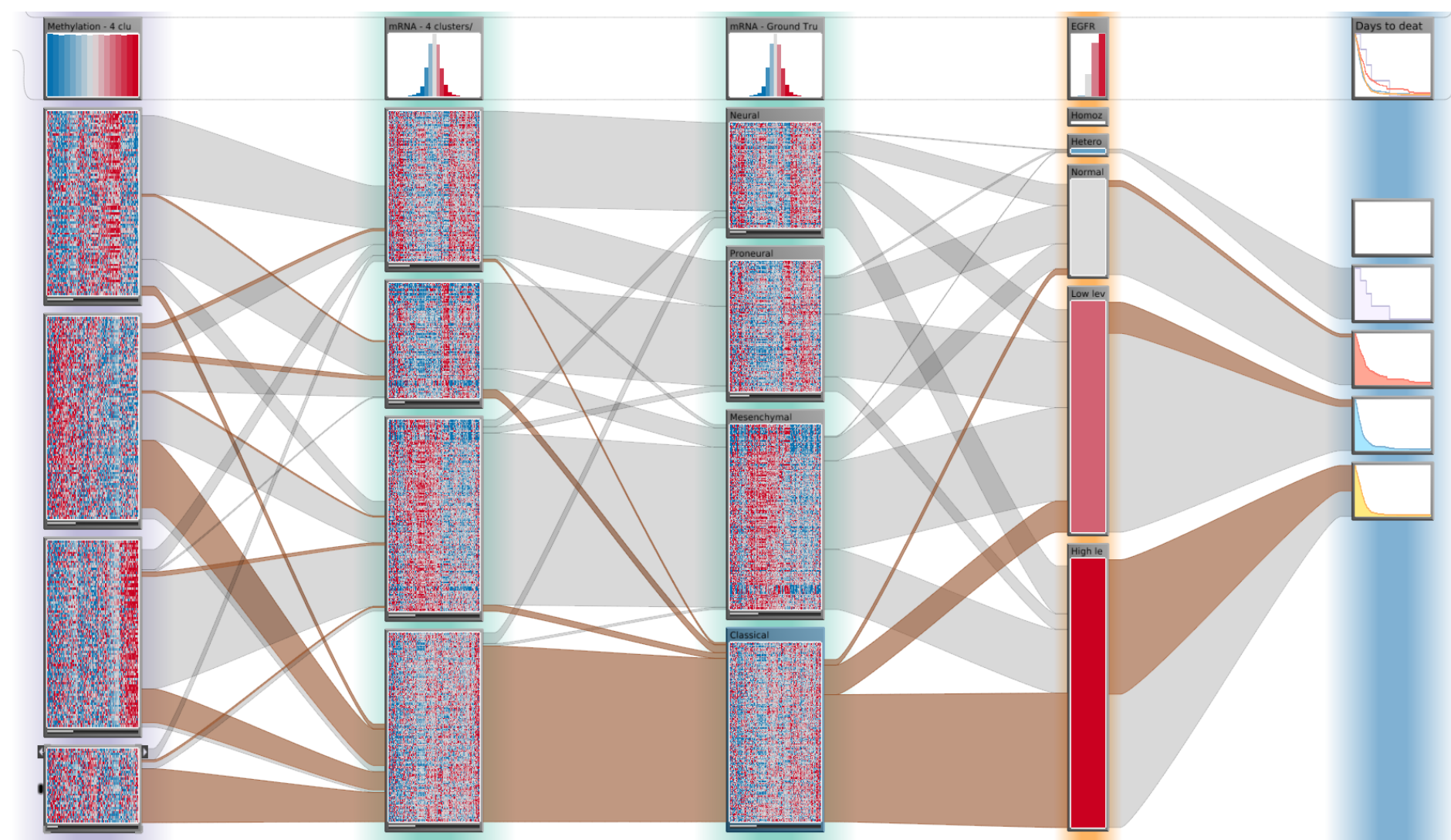
When questions are not well
defined: Exploration

What is the structure of a terrorist network?

Which drug can help patient X?

Purpose of Visualization

[Obama Administration]



Open Exploration

Confirmation

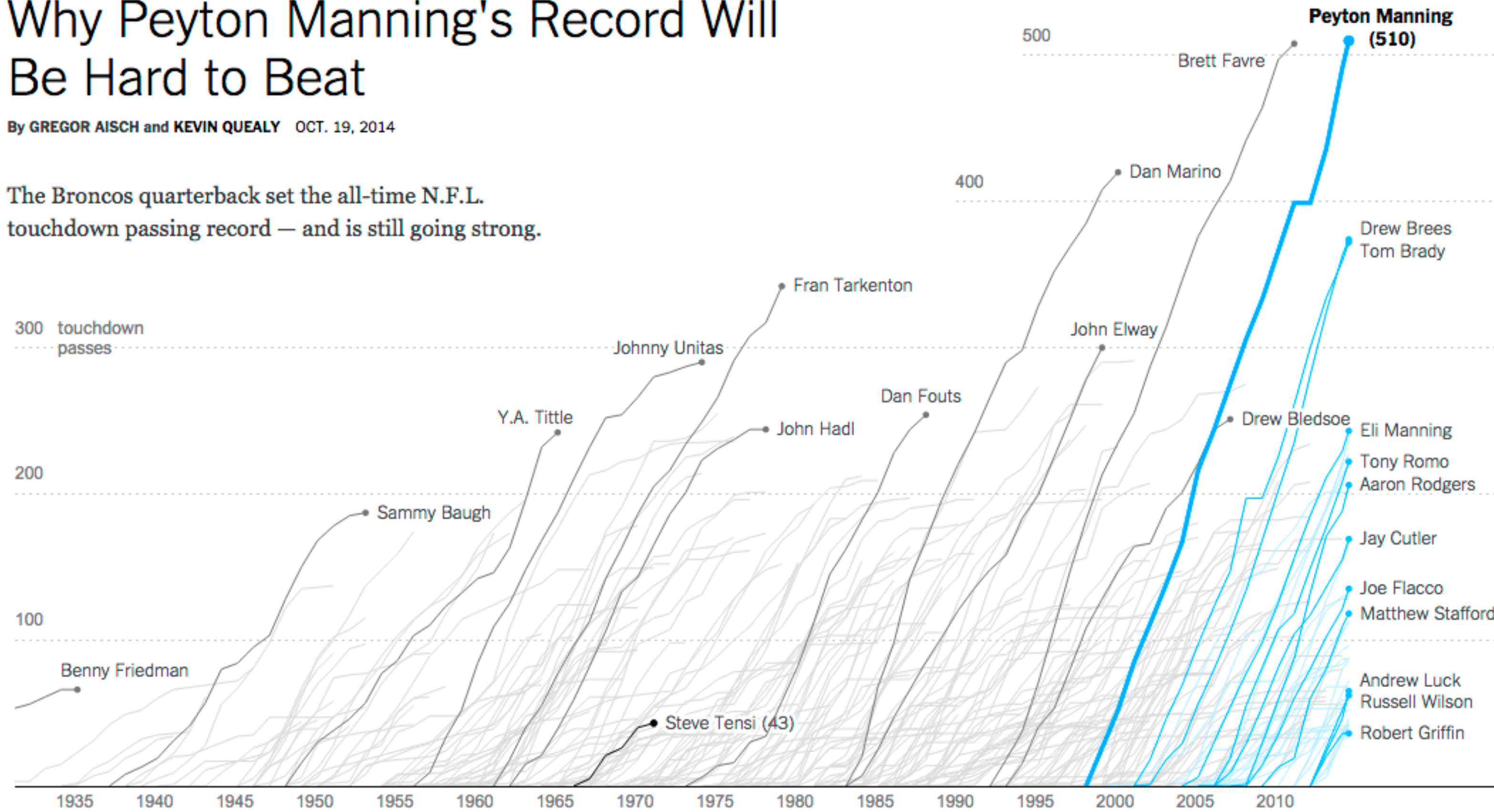
Communication

Example Communication

Why Peyton Manning's Record Will Be Hard to Beat

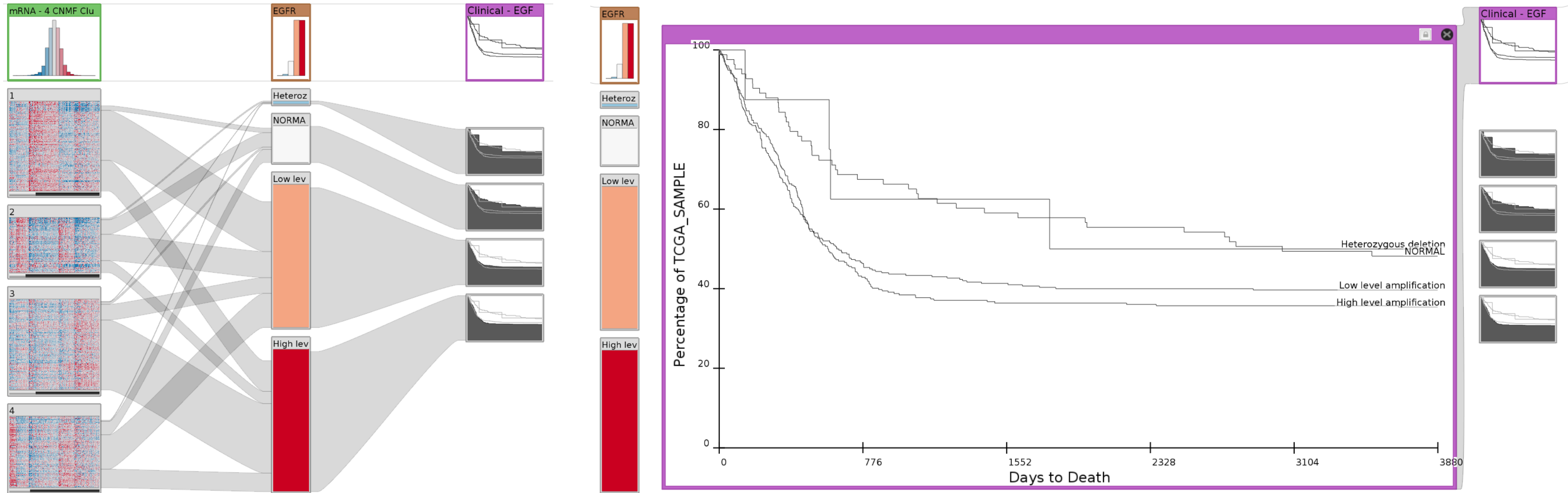
By GREGOR AISCH and KEVIN QUEALY OCT. 19, 2014

The Broncos quarterback set the all-time N.F.L. touchdown passing record — and is still going strong.



[New York Times]

Example Exploration: Cancer Subtypes



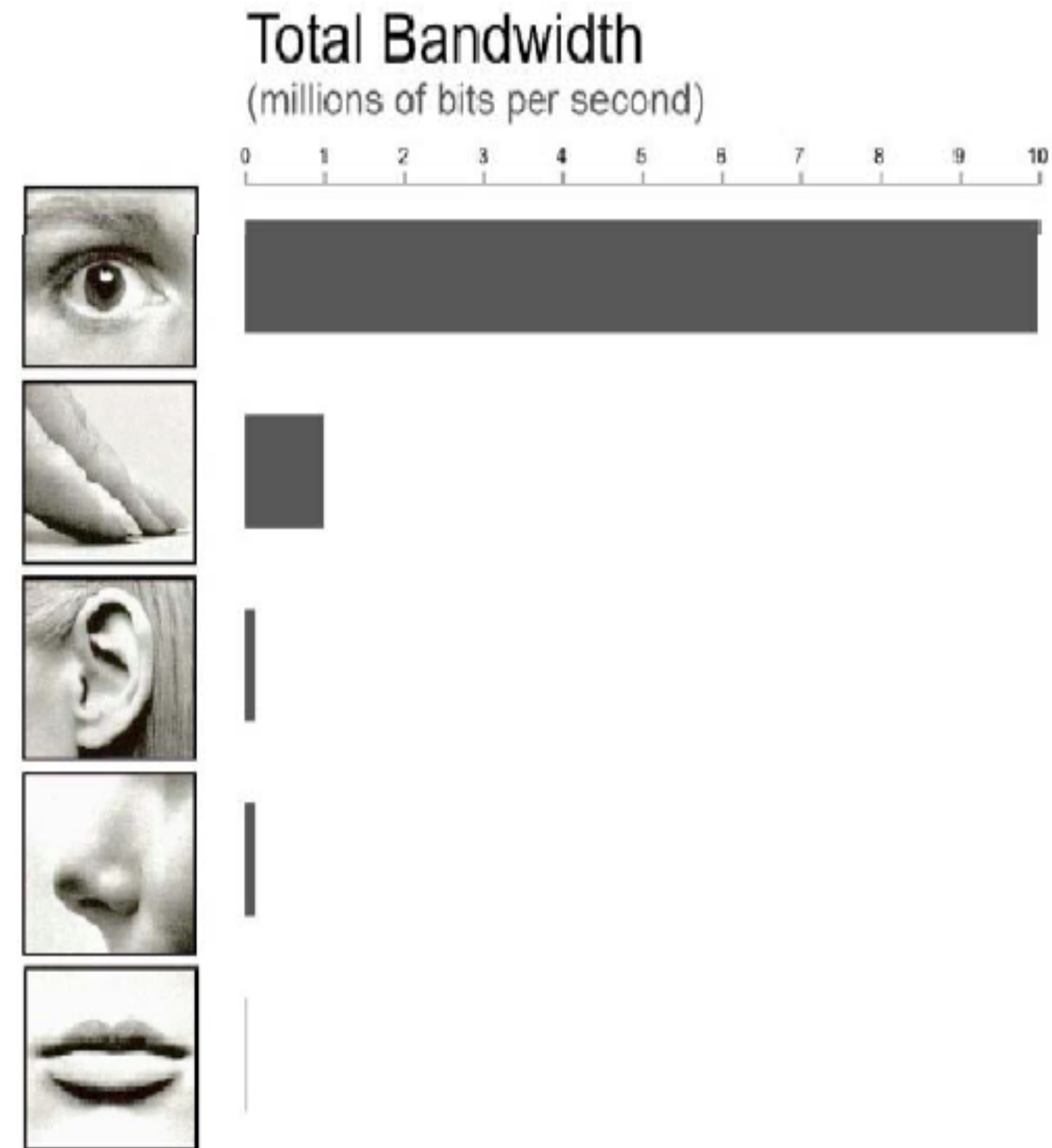
Why Graphics?

Figures are **richer**; provide more information with less clutter and in less space.

Figures provide the *gestalt* effect: they give an overview; **make structure more visible**.

Figures are **more accessible**, easier to understand, **faster to grasp**, more comprehensible, **more memorable**, more fun, and less formal.

list adapted from: [Stasko et al. 1998]



for city's main public hospital was a wreck, and the city's public-housing projects were shuttered.

Campanella then switched to an identically constructed map, only this time based on 2010 census data, and in bits and pieces on the screen there was a simple and arresting picture of what Katrina meant. In the neighborhoods that were once a dense black, many of the little squares had thinned and turned gray. The sharp lines that once separated the teapot from Central City were now blurry: the white areas of the city were pushing north, into the vacuum left by the exodus. The Bywater was graying, as it gentrified still further. "Before Katrina, an American Community Survey estimate of New Orleans Parish population was four hundred and fifty-five thousand, and about sixty-eight per cent black," Campanella said. "Now the latest estimate is three hundred and eighty-four thousand, and it's about

Textual description of a map of the effects of hurricane Katrina on New Orleans.
New Yorker, posted by Alberto Cairo

Katrina's Diaspora

Katrina's Diaspora

The victims of Hurricane Katrina have filed for assistance from FEMA from every state. The map shows the distribution and number of the 1.36 million individual assistance applications as of Sept. 23.

Counties from which families filed applications

Circles are sized according to the number of applications from a ZIP code

10,000
5,000
1,000
100
10

Miles
0 100 200 300 400

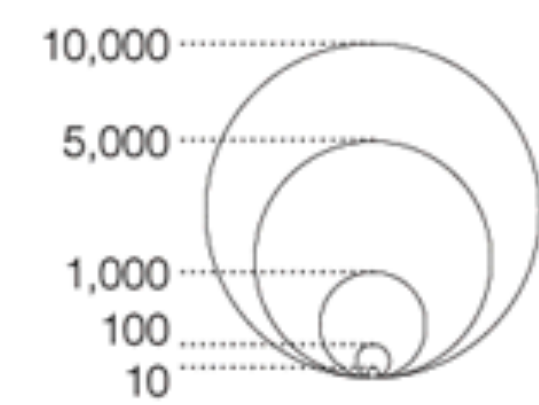
Selected metropolitan areas and application counts:

- San Francisco: 1,954
- Los Angeles: 4,435
- San Diego: 1,203
- Las Vegas: 1,210
- Phoenix: 1,400
- Tucson: 364
- El Paso: 568
- Honolulu: 105
- Anchorage: 67
- Albuquerque: 405
- Salt Lake City: 448
- Boise: 88
- Portland: 520
- Seattle: 920
- San Antonio: 6,035
- Corpus Christi: 1,169
- Houston: 84,749
- Dallas-Ft. Worth: 37,113
- San Antonio: 6,035
- Chicago: 4,773
- Minneapolis-St. Paul: 966
- St. Paul: 966
- Chicago: 4,773
- Detroit: 1,651
- New York: 4,186
- Philadelphia: 1,562
- Washington: 4,852
- Atlanta: 29,252
- Jacksonville: 2,797
- Orlando: 2,693
- Ft. Walton Beach: 3,343
- Tampa: 2,907
- Miami-Ft. Lauderdale: 4,188
- New Orleans: 183,617

Number of applications from selected metropolitan areas

Counties from which families filed applications

Circles are sized according to the number of applications from a ZIP code



PUERTO RICO

When not to visualize?

When to automate?

Well defined question on well-defined dataset

Which gene is most frequently mutated in this set of patients?

What is the current unemployment rate?

No human intervention possible/necessary

Decisions needed in minimal time

High frequency stock market trading: which stock to buy/sell?

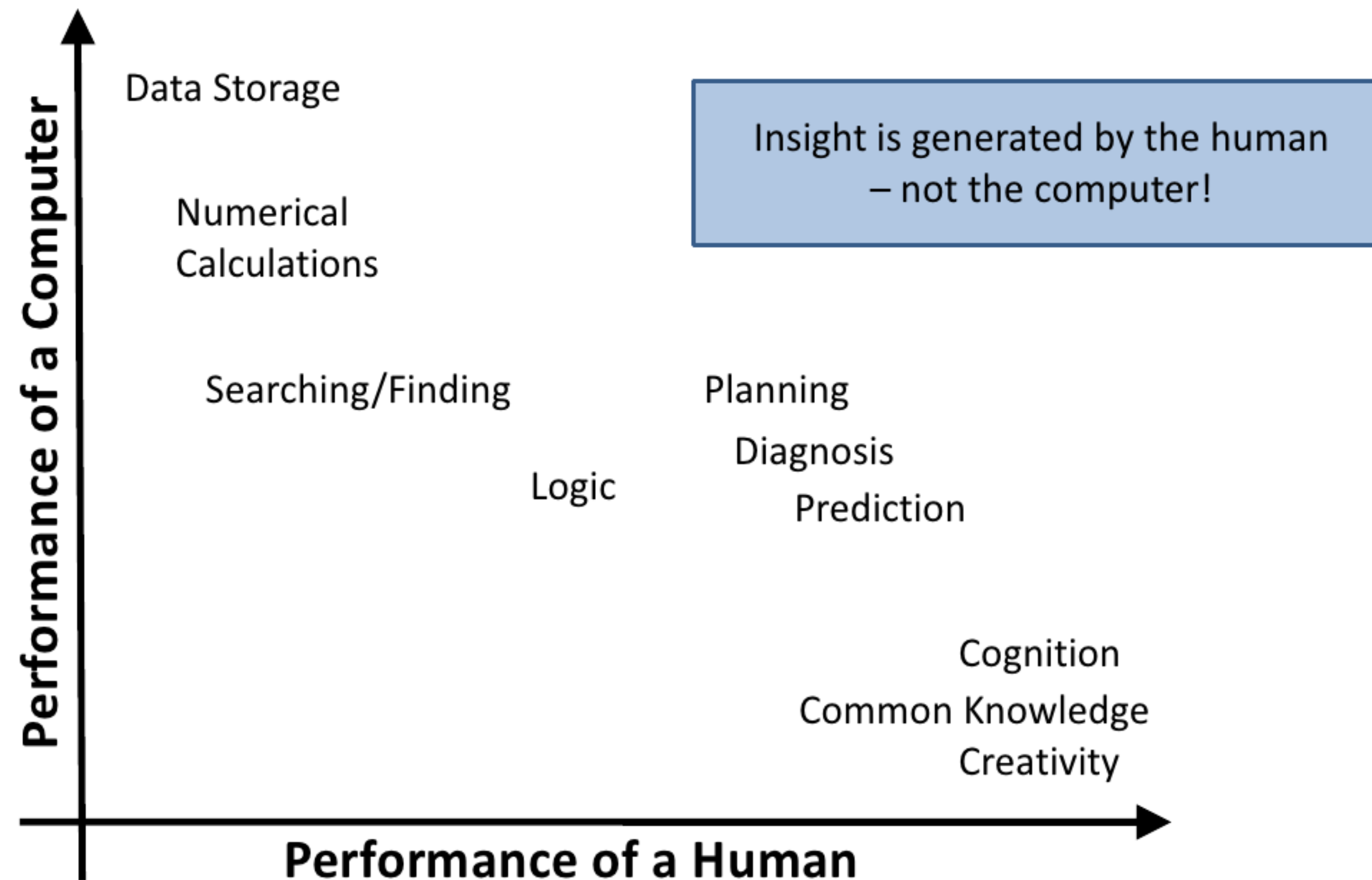
Manufacturing: is bottle broken?

Impractical for human to be involved

Automatic data products



The Ability Matrix



Why Use Computers?

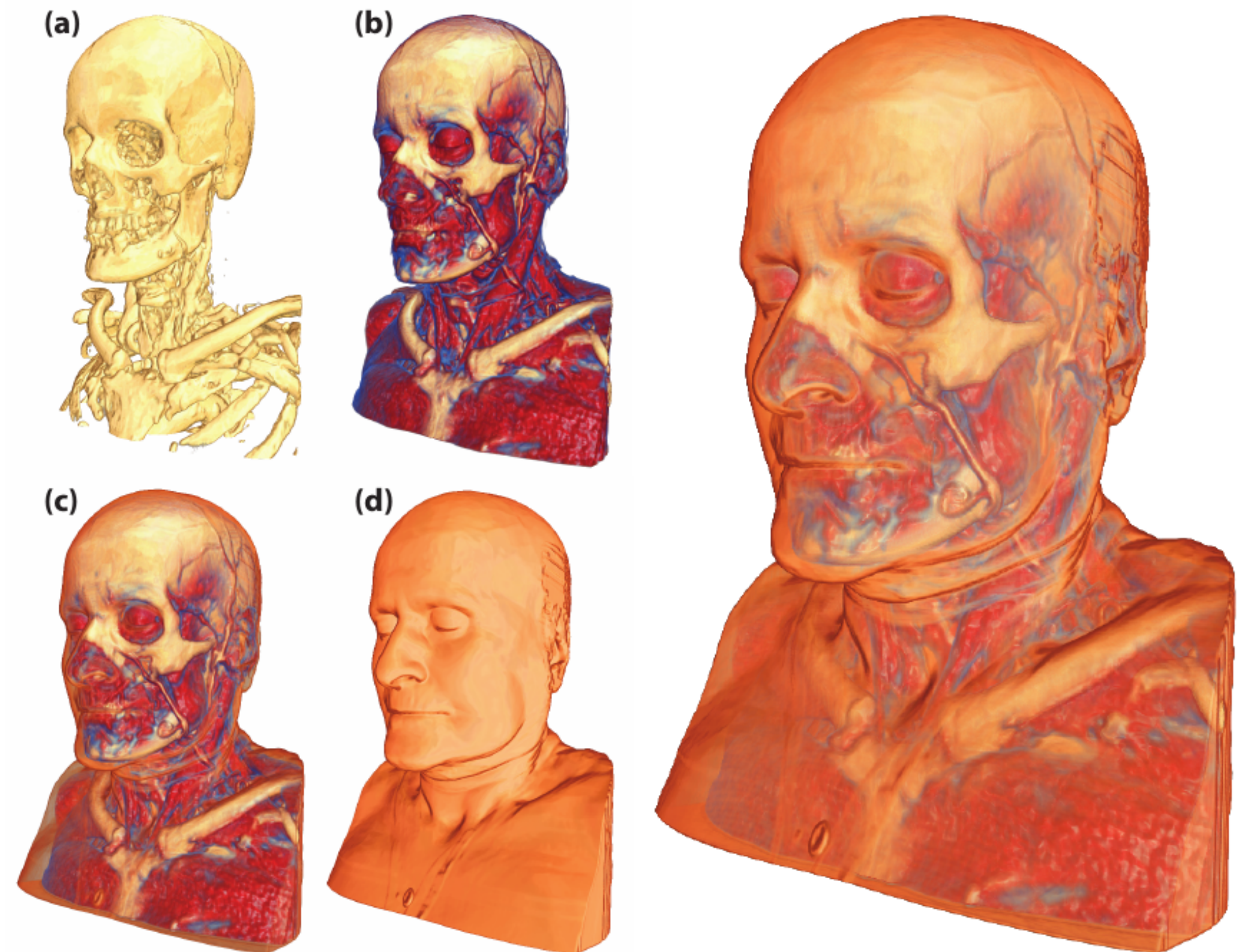
Scale

Drawing by hand (or Illustrator)

infeasible

inflexible (updates!)

How to draw an MRI scan?



[Bruckner 2007]

Why Use Computers?

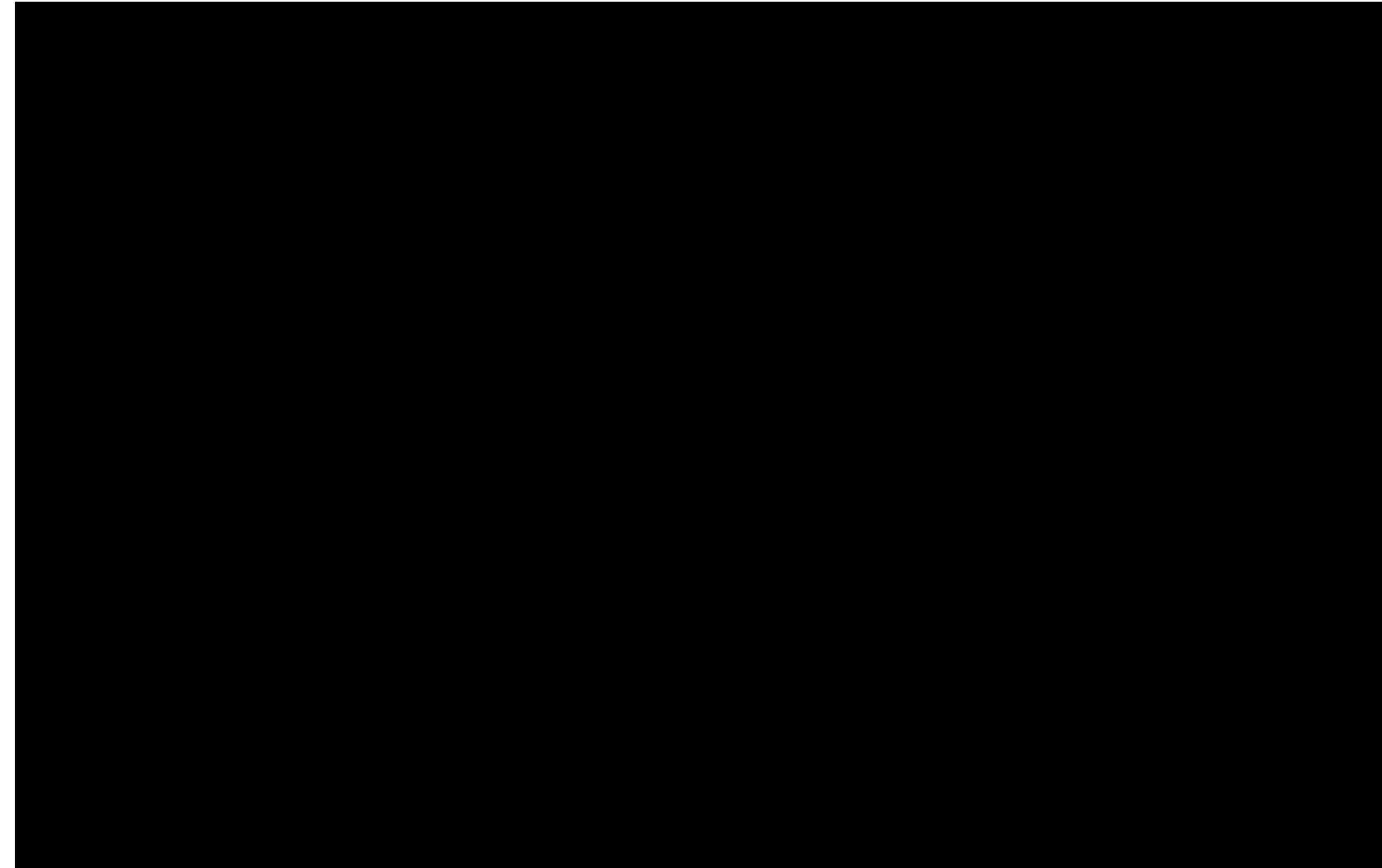
Interaction

Interaction allows to “drill down” into data

Integration

Integration with algorithms

Make visualization part of a data analysis pipeline



[Sunburst by John Stasko, Implementation in Caleydo by Christian Partl]

Why User Computers?

Efficiency

Re-use charts / methods for
different datasets

Quality

Precise data driven rendering

Storytelling

Use time

Tell Stories

[New York Times]



Why not just use Statistics?

I		II		III		IV	
x	y	x	y	x	y	x	y
10	8.0	10	9.1	10	7.4	8	6.5
8	6.9	8	8.1	8	6.7	8	5.7
13	7.5	13	8.7	13	12.	8	7.7
9	8.8	9	8.7	9	7.1	8	8.8
11	8.3	11	9.2	11	7.8	8	8.4
14	9.9	14	8.1	14	8.8	8	7.0
6	7.2	6	6.1	6	6.0	8	5.2
4	4.2	4	3.1	4	5.3	19	12.
12	10.	12	9.1	12	8.1	8	5.5
7	4.8	7	7.2	7	6.4	8	7.9
5	5.						6.8

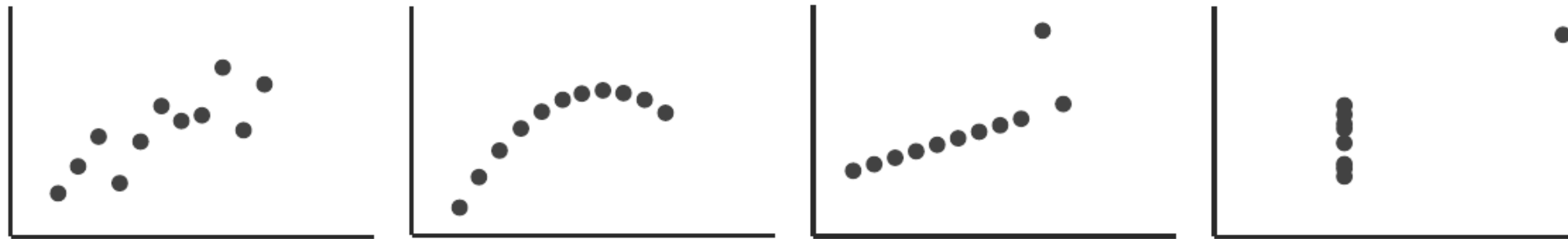
Mean x: 9 y: 7.50

Variance x: 11 y: 4.122

Correlation x – y: 0.816

Linear regression: $y = 3.00 + 0.500x$

Anscombe's Quartett

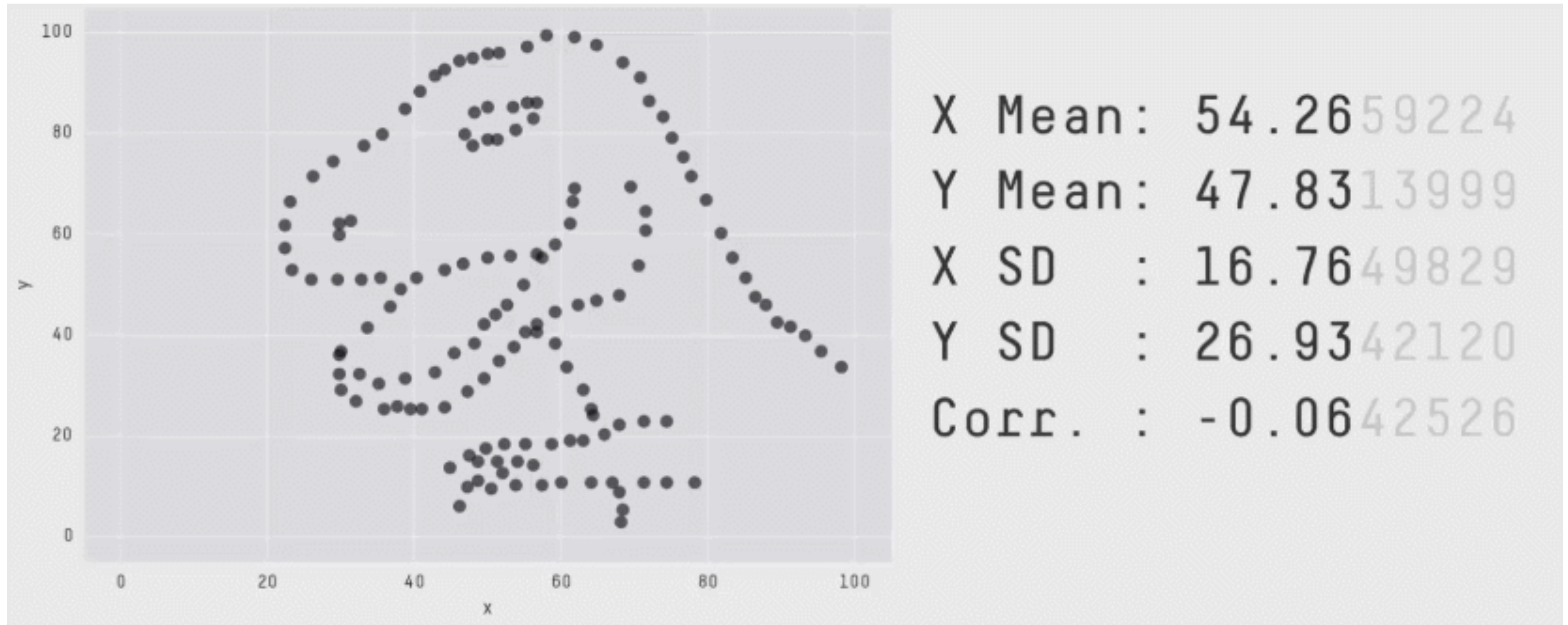


Mean x: 9 y: 7.50

Variance x: 11 y: 4.122

Correlation x – y: 0.816

Linear regression: $y = 3.00 + 0.500x$



Same Stats, Different Graphs: Generating Datasets with Varied Appearance and Identical Statistics through Simulated Annealing, CHI 2017, Justin Matejka, George Fitzmaurice

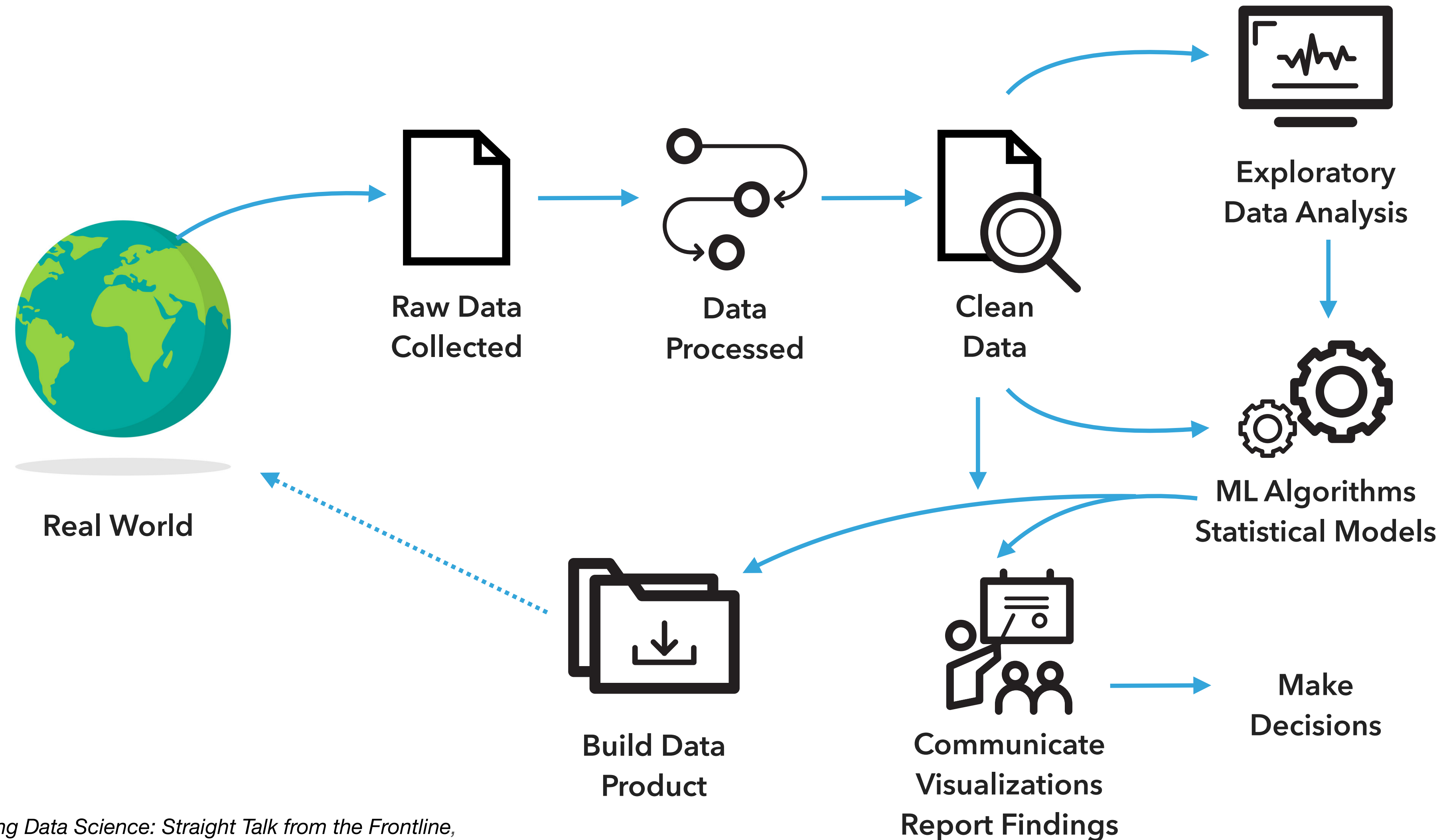
Visualization =

Human Data Interaction

Data

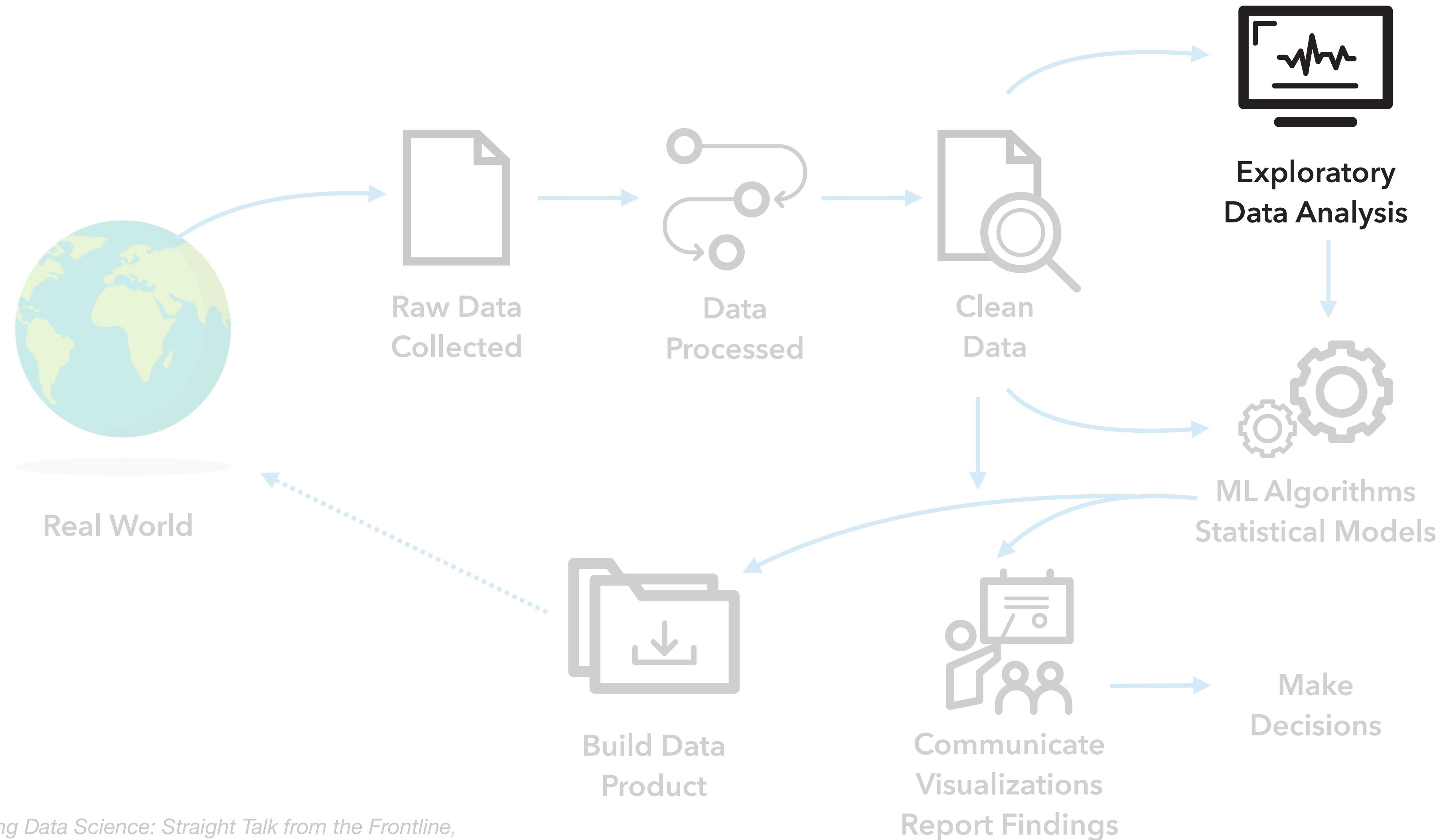
Human-Data Interaction

What's the Role of **Visualization** in **Data Science**?

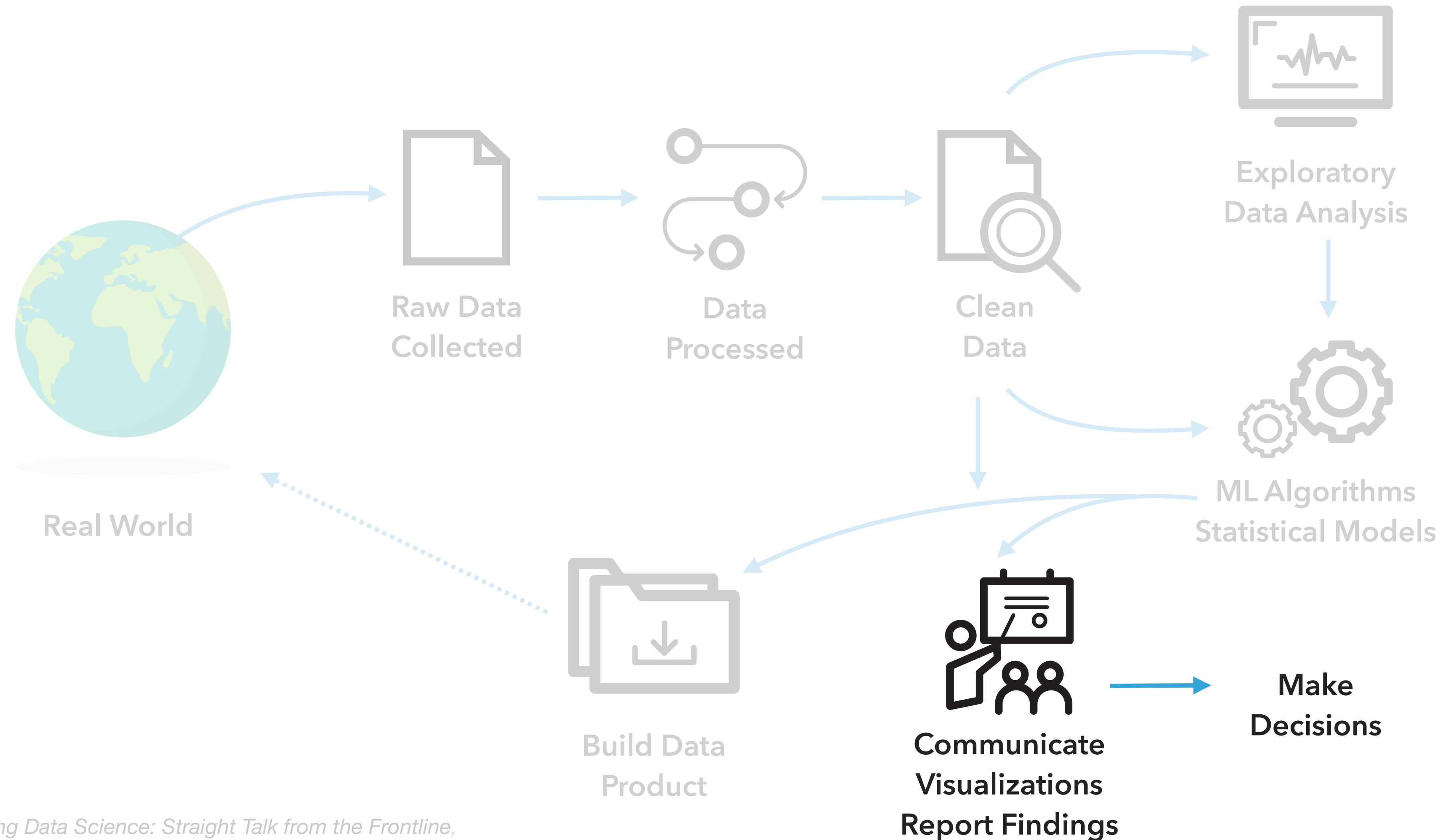


*Adapted from Doing Data Science: Straight Talk from the Frontline,
[O'Neil and Schutt, 2013]*

What's the Role of Visualization in Data Science?

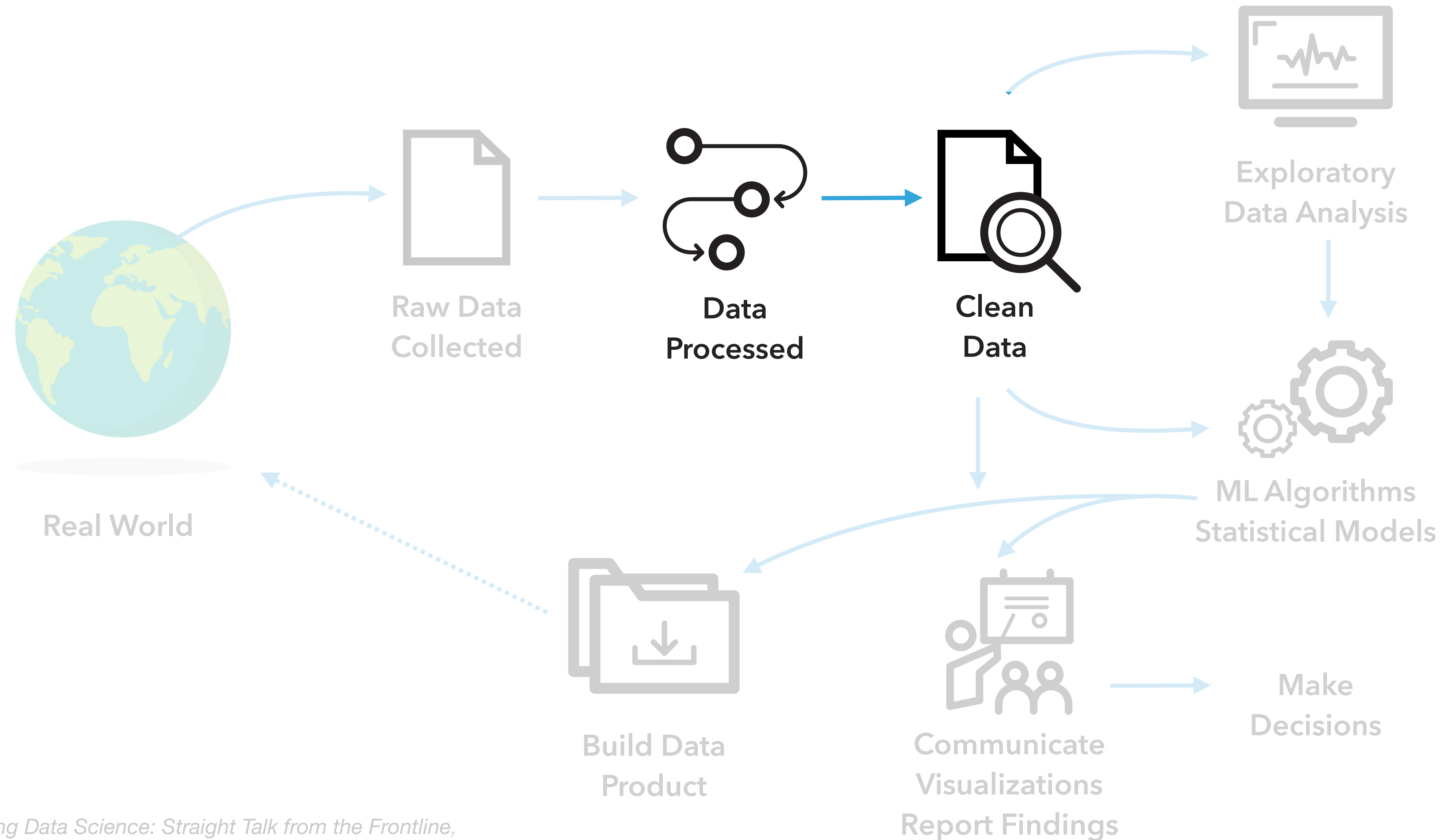


What's the Role of **Visualization** in **Data Science**?



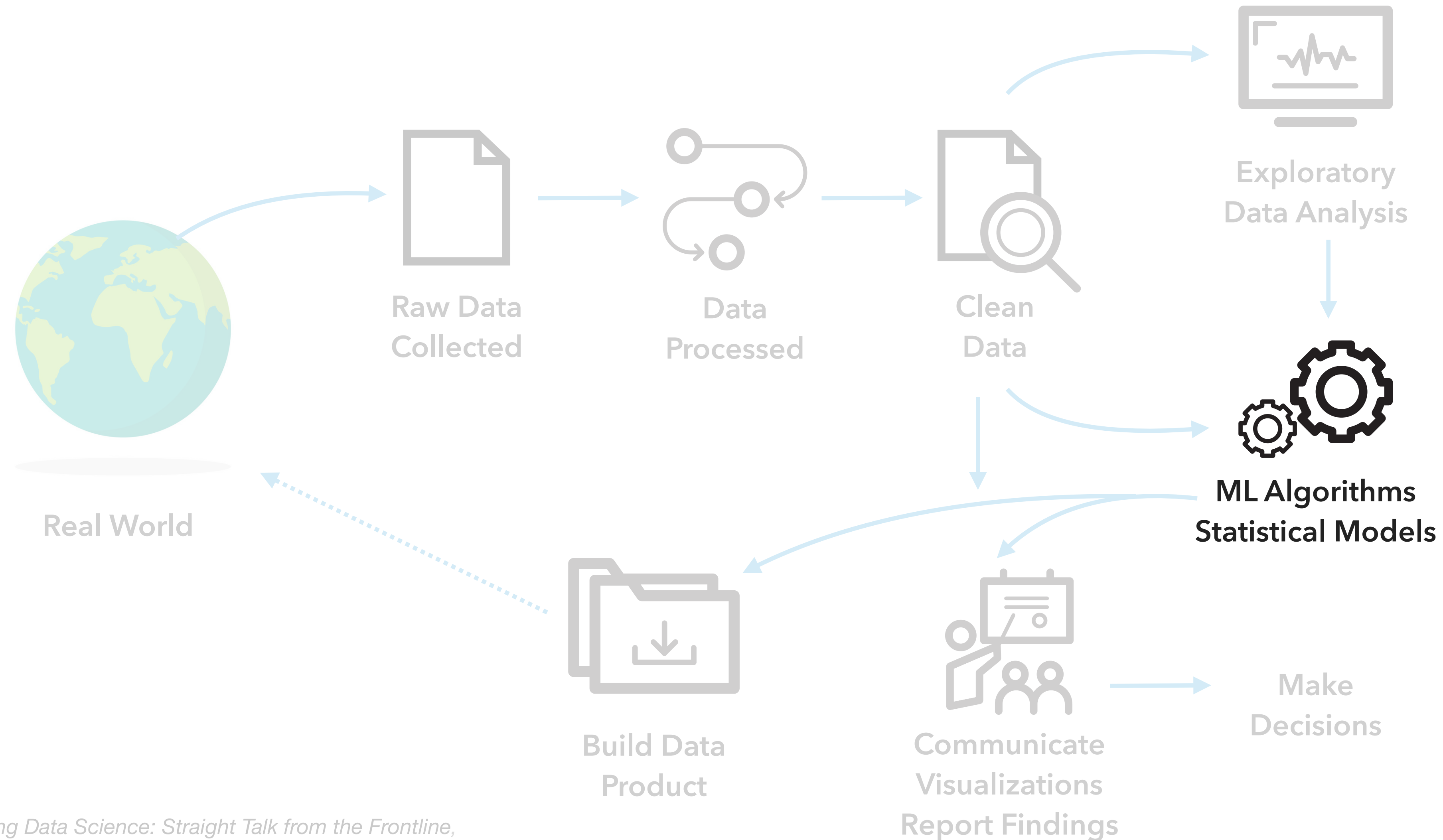
*Adapted from Doing Data Science: Straight Talk from the Frontline,
[O'Neil and Schutt, 2013]*

What's the Role of **Visualization** in **Data Science**?



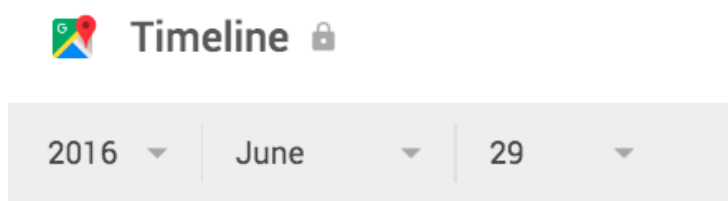
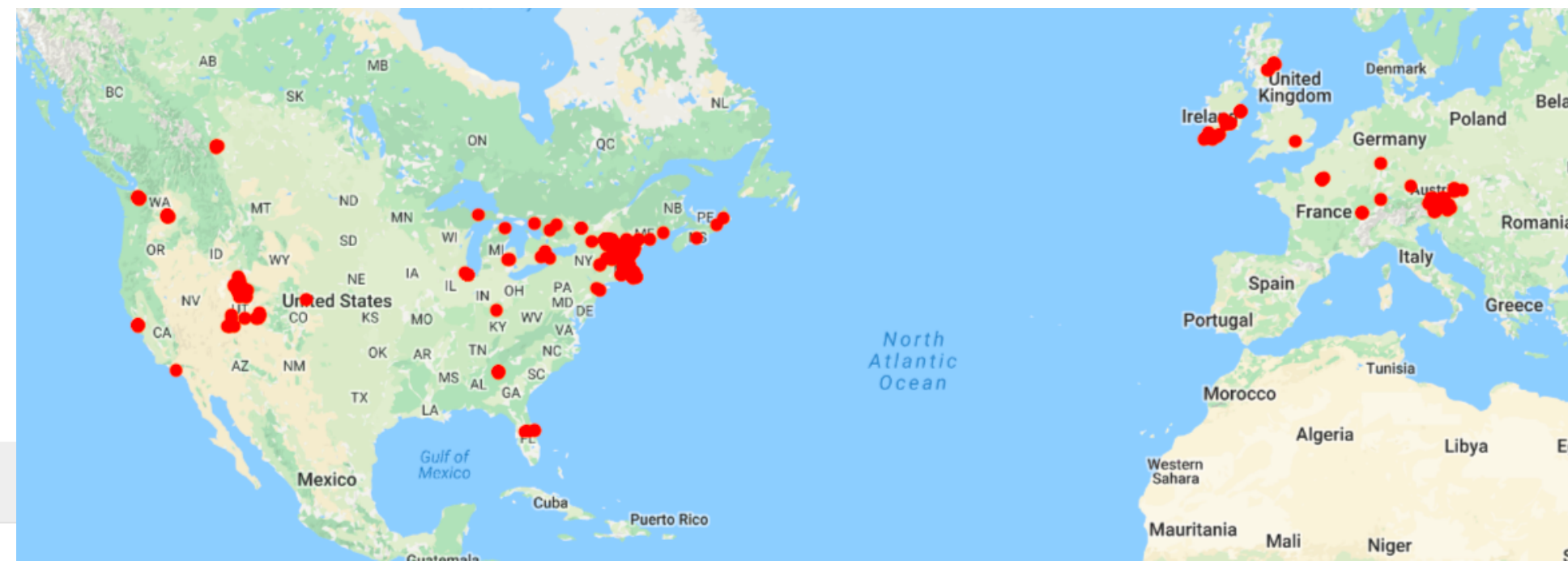
*Adapted from Doing Data Science: Straight Talk from the Frontline,
[O'Neil and Schutt, 2013]*

What's the Role of **Visualization** in **Data Science**?

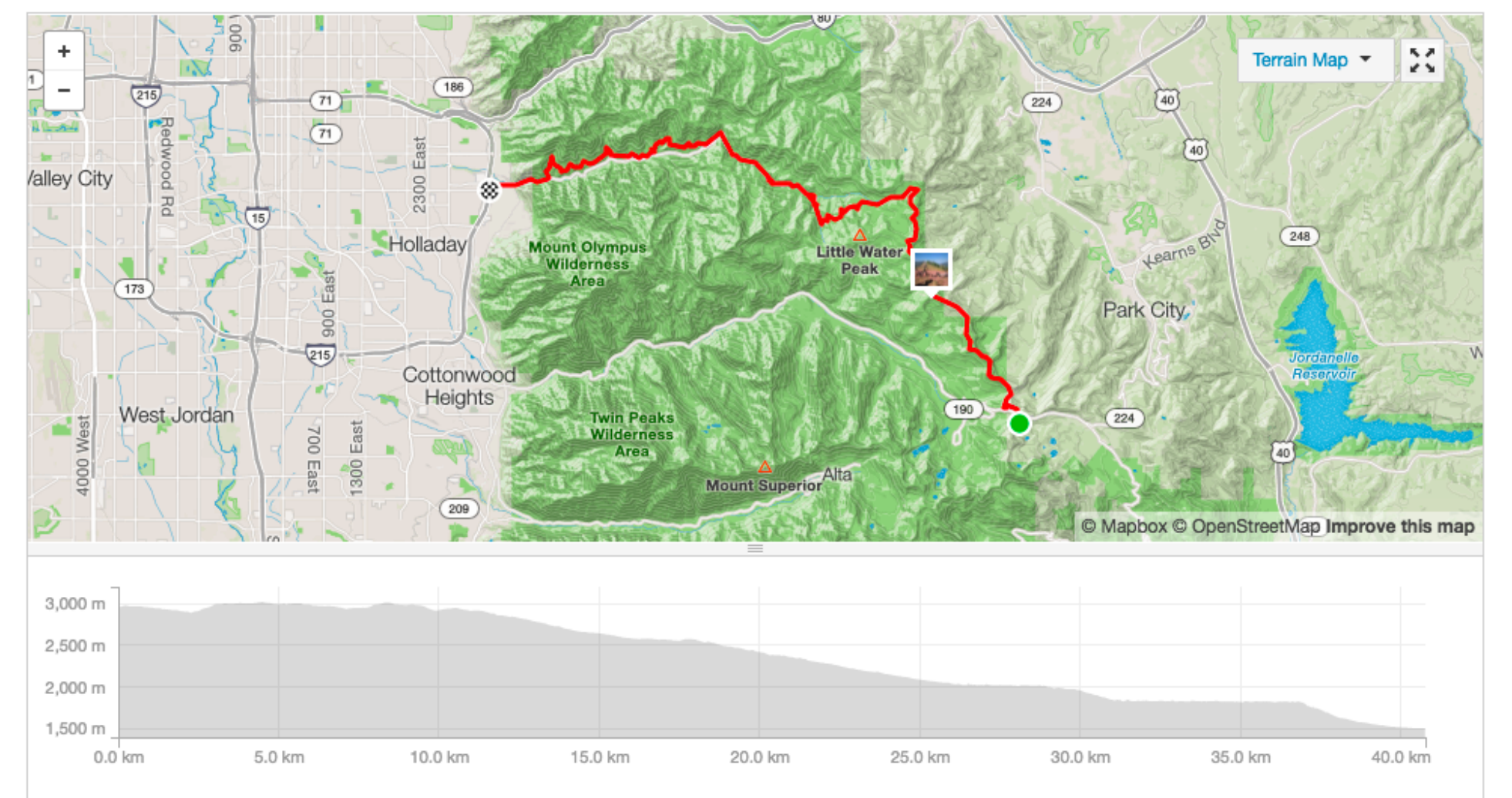
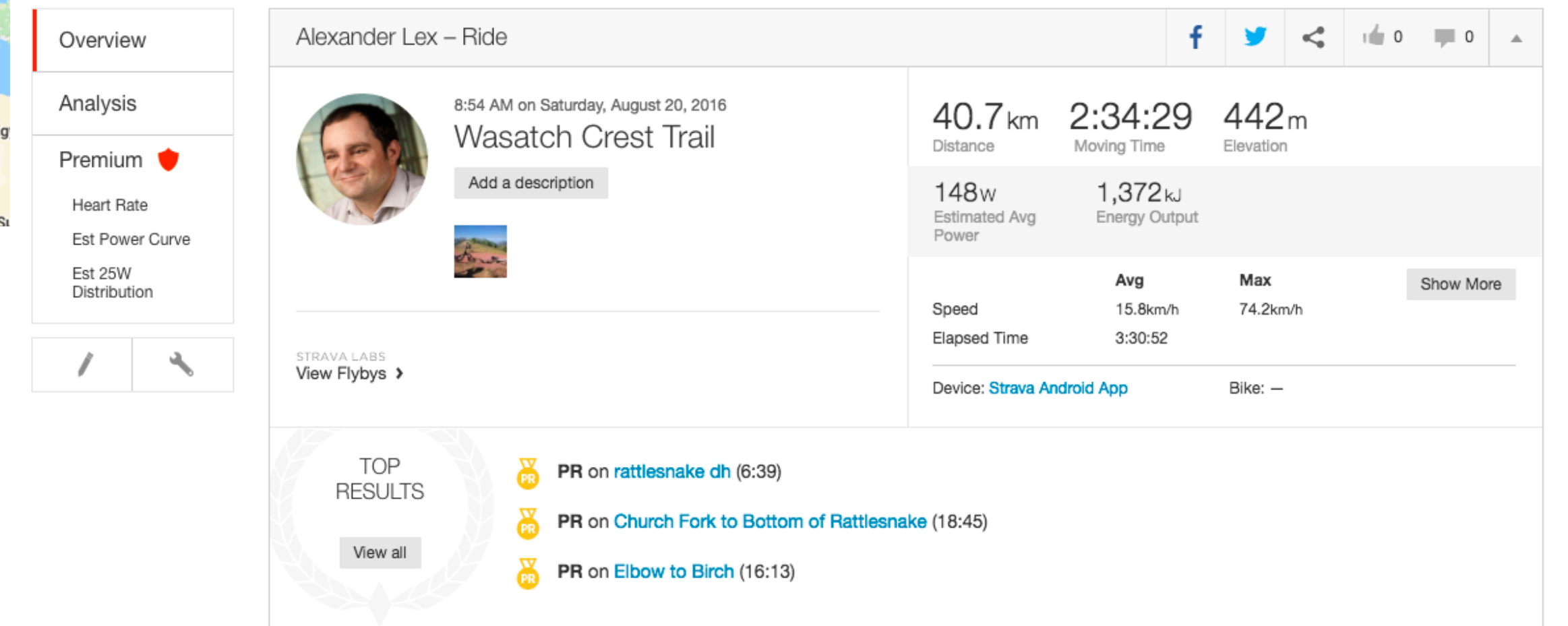
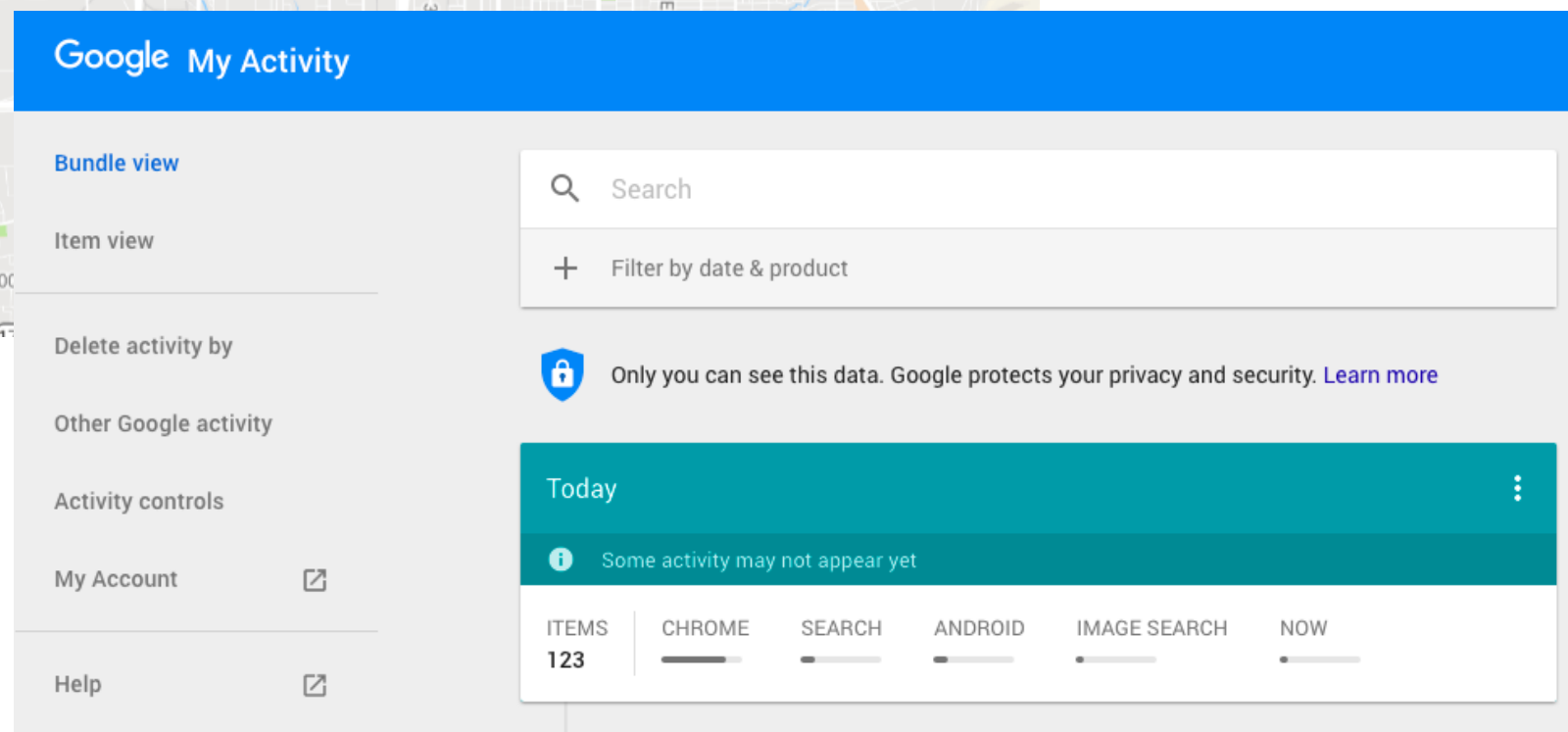
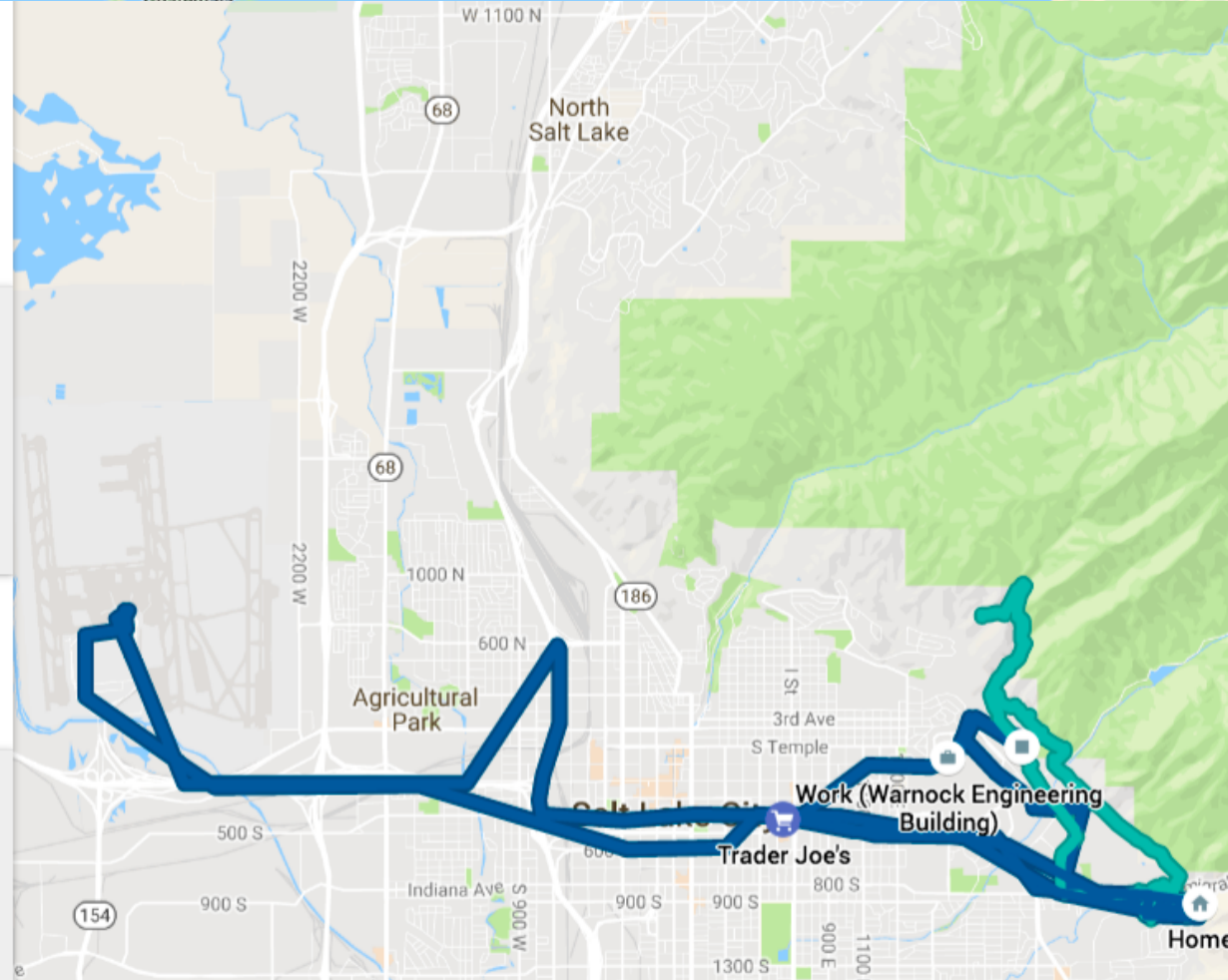
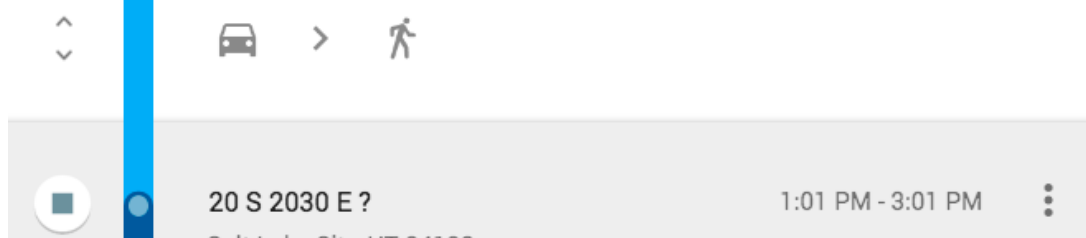
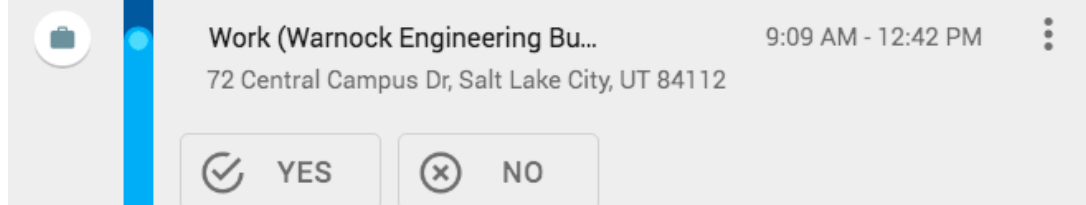
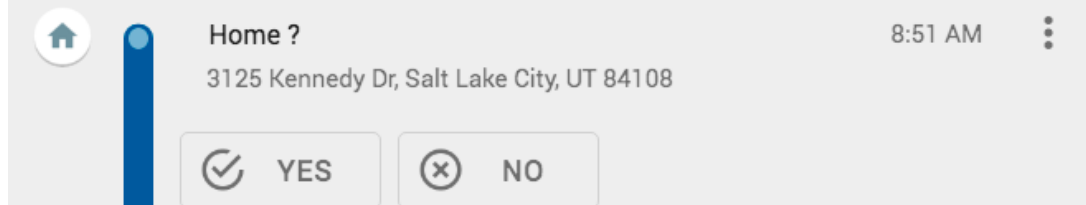


*Adapted from Doing Data Science: Straight Talk from the Frontline,
[O'Neil and Schutt, 2013]*

Example: Personal Data



Wednesday, June 29, 2016



Big Data in Science and Engineering

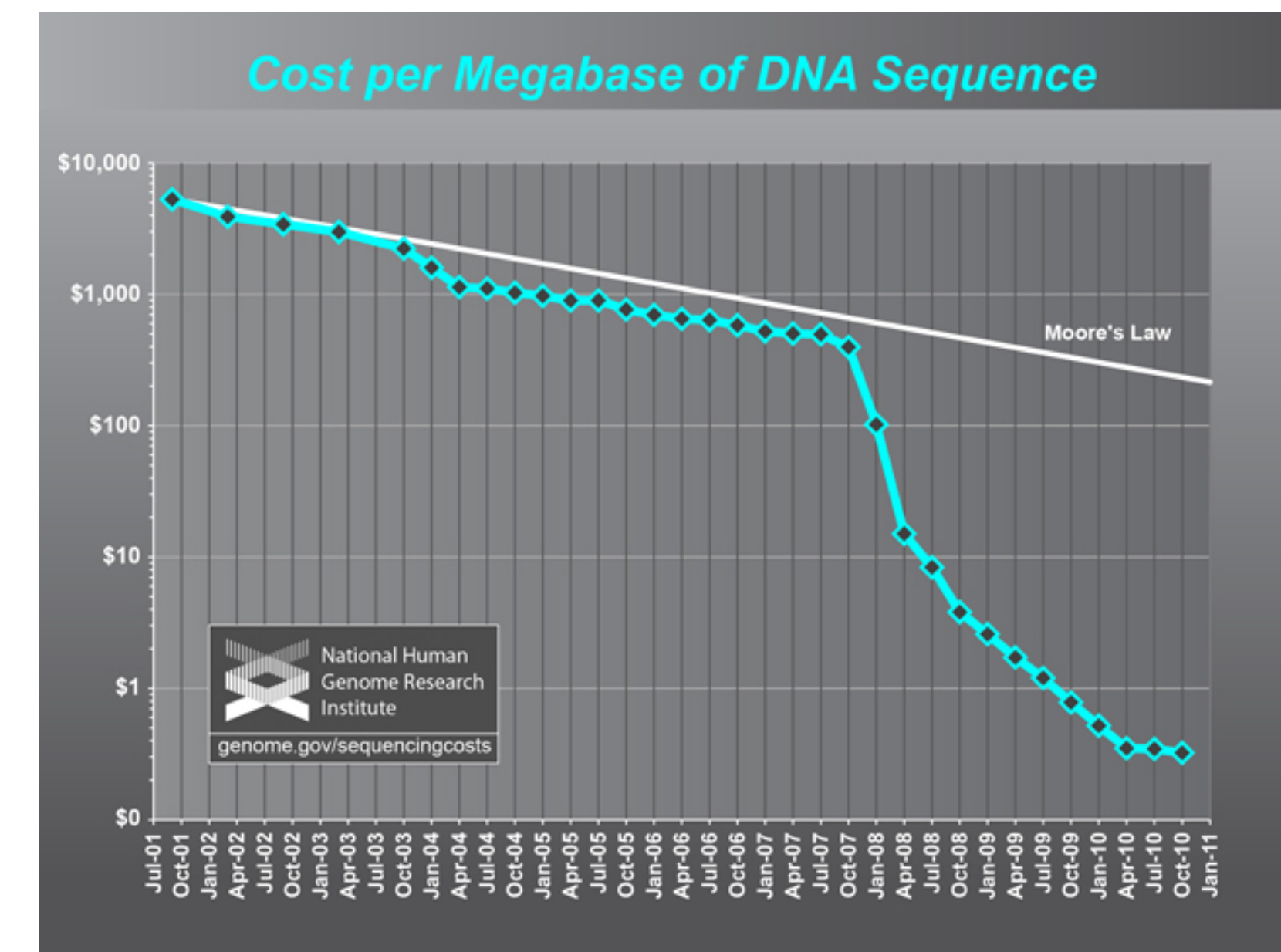
“Big Data” has transformed science and engineering.

Cheap sensors (e.g. imaging) have changed everything

Examples:

- Large physics experiments and observations
- Cheaper and automated genome sequencing
- Smart buildings / cities (blyncsy)

Controversy: Hypothesis or data driven methods

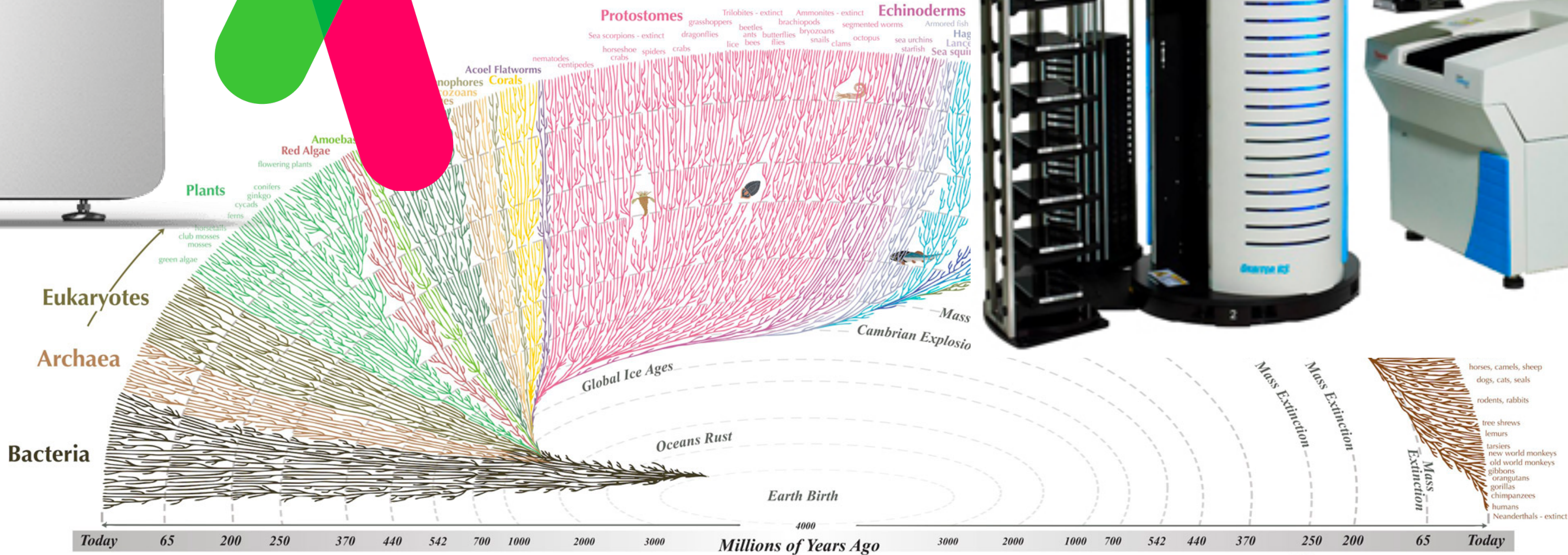


Example: Genomics



23andM

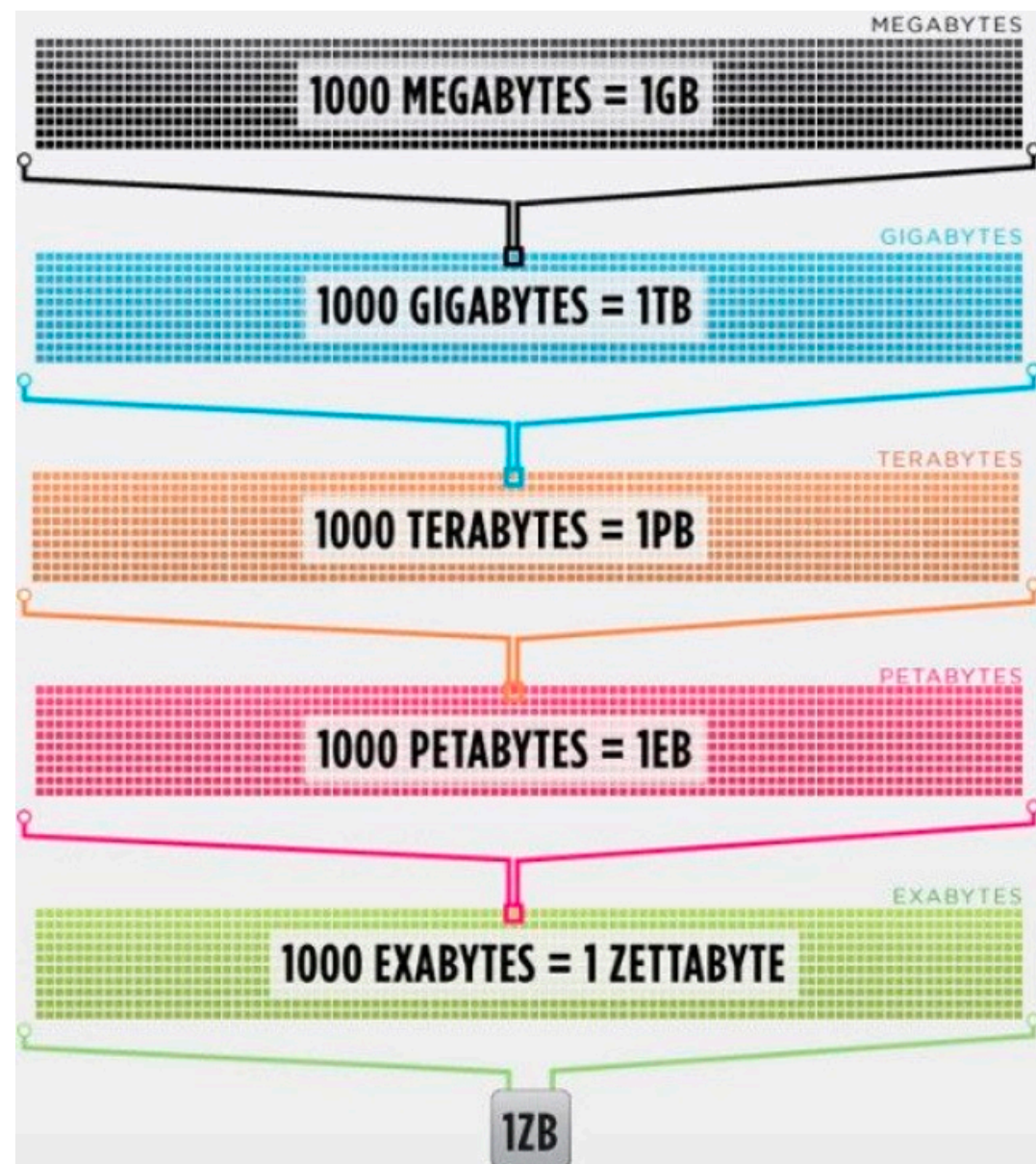
Example TCGA: 1 Petabyte



All the major and many of the minor living branches of life are shown on this diagram, but only a few of those that have gone extinct are shown. Example: Dinosaurs - extinct

NSA Utah Data Center (Bluffdale, Utah)

Storage Capacity?
estimates vary,
NPR/NG estimate
five zettabytes of
data



“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, ... because now we really do have **essentially free and ubiquitous data.**”

Hal Varian, Google’s Chief Economist
The McKinsey Quarterly, Jan 2009

Humans!

Human Data Interaction

Why Humans?

Leveraging human capabilities

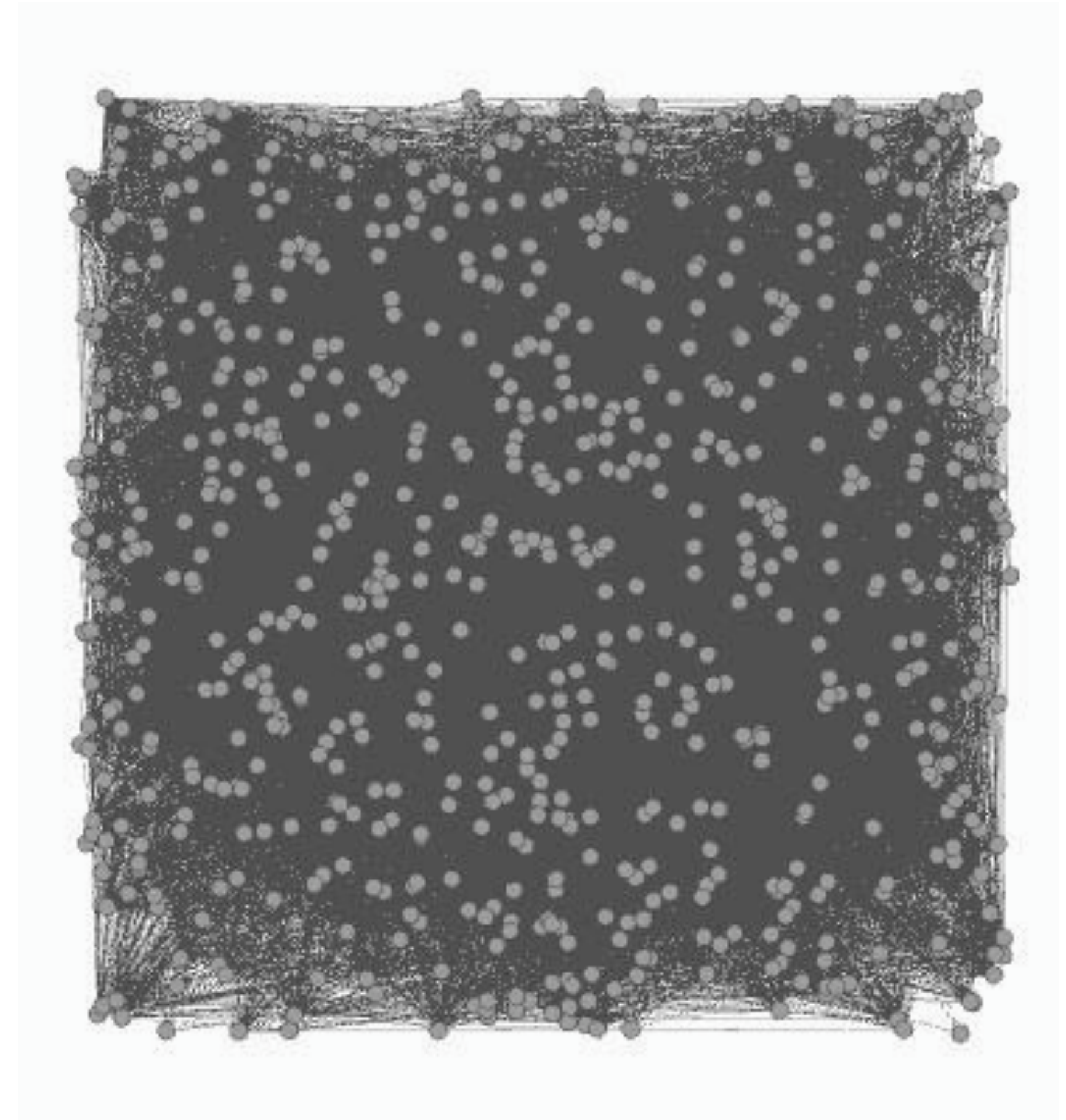
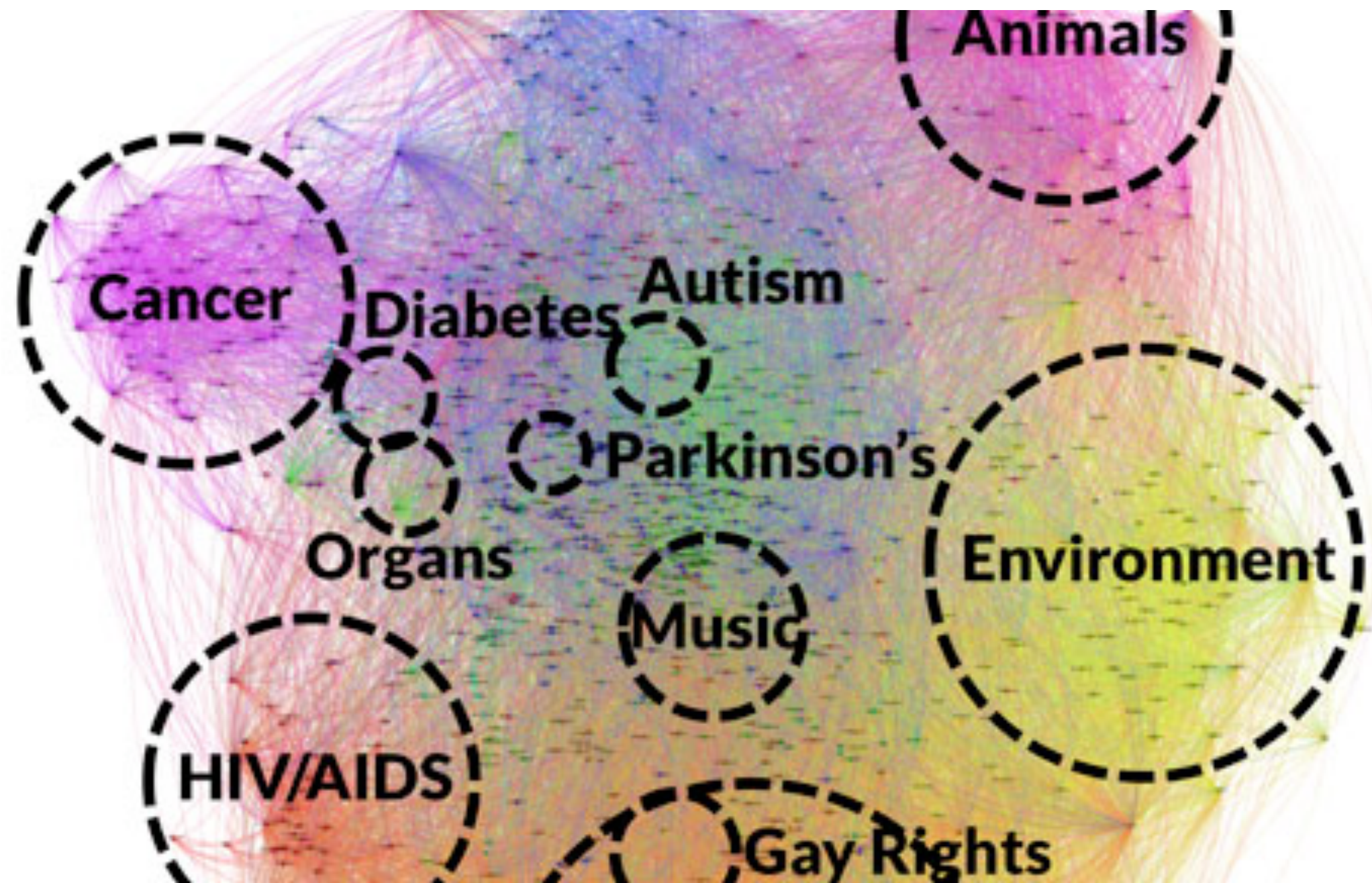
Pattern Discovery: clusters, outliers, trends

Contextual Knowledge: expectations for dataset, explanations for patterns

Action: humans learn and take action

But: we also have to **design for Humans and their limitations**

Not everything that can be drawn can be read!



Limits of Cognition

Daniel J. Simons and Daniel T. Levin, Failure to detect changes to people during a real world interaction, 1998



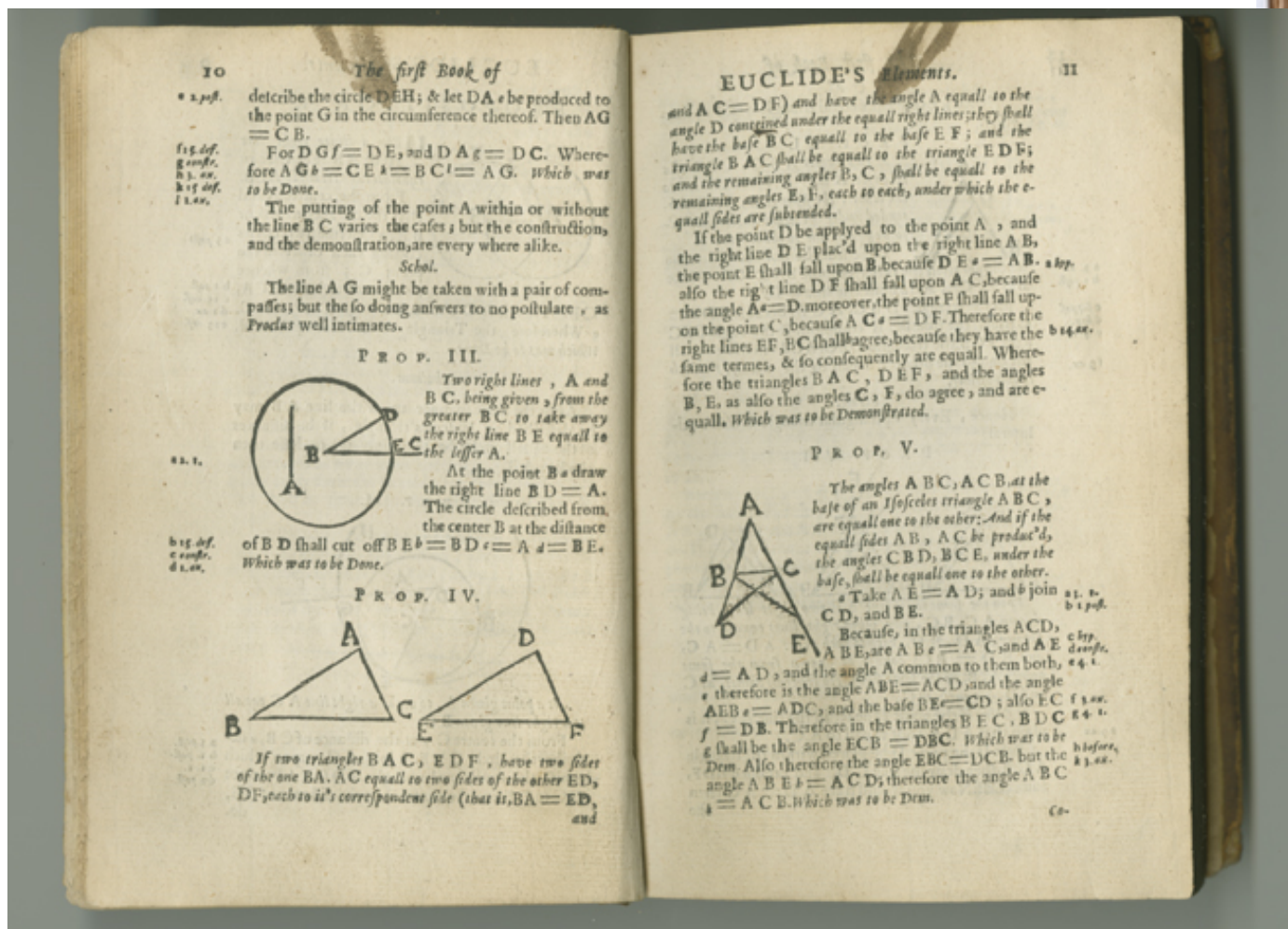
How did we get here?

A bit of history

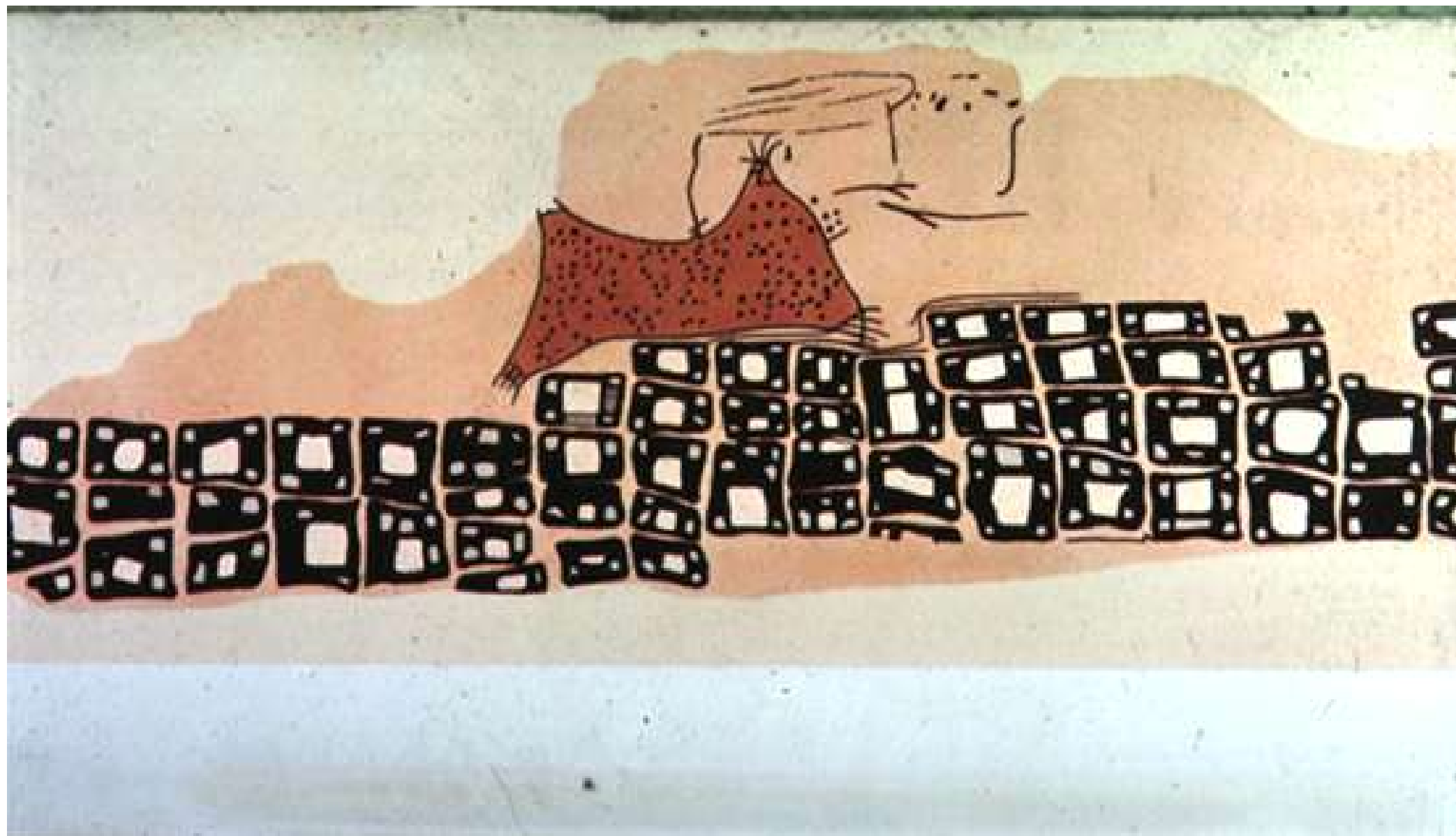
“It is things that make us smart”

Donald A. Norman

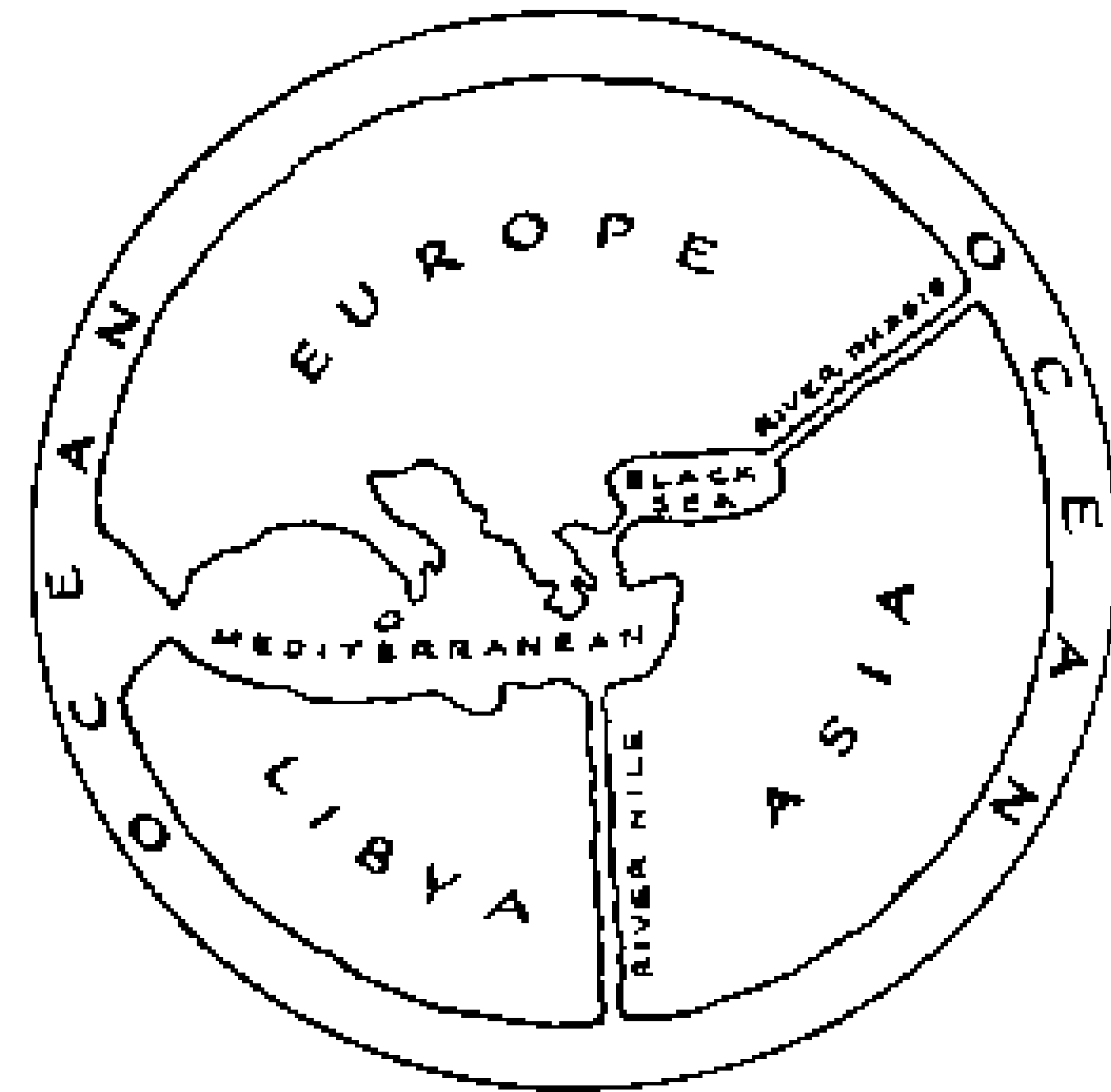




Record



Konya town map, Turkey, c. 6200 BC



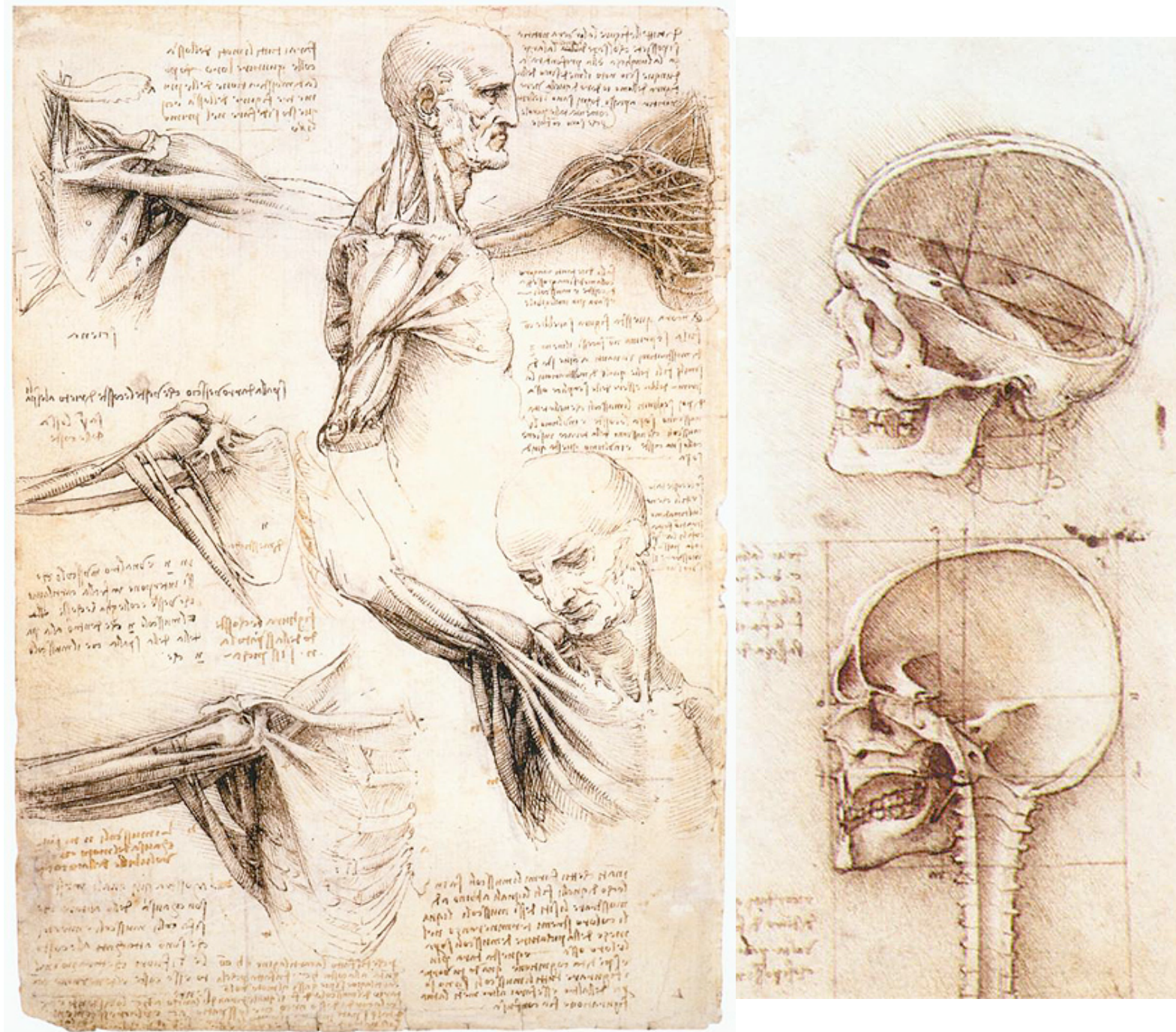
Anaximander's Map of the World

Anaximander of Miletus, c. 550 BC

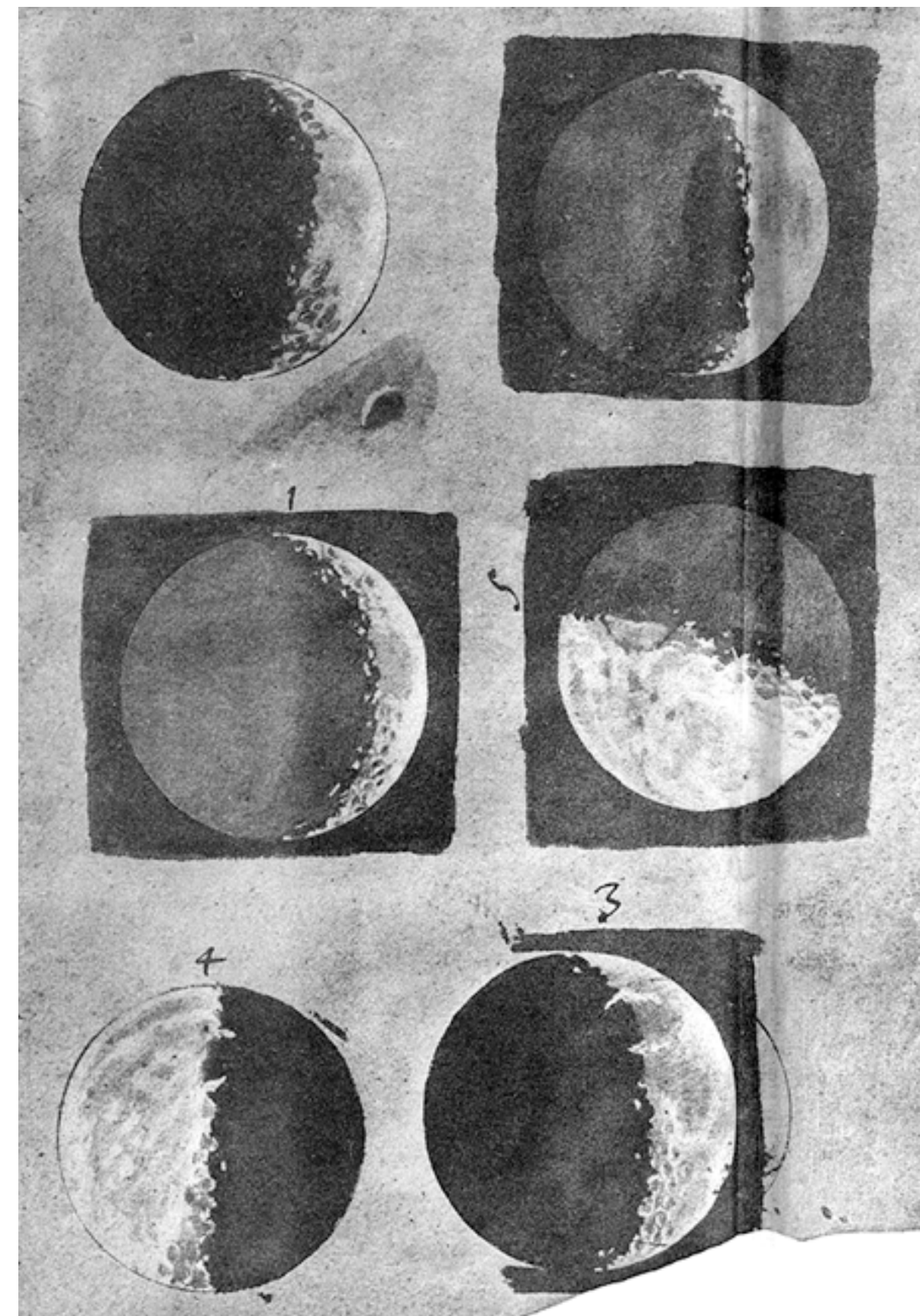
Tabula Peutingeriana – Roman Road Map



Record



Leonardo Da Vinci, ca. 1500



Galileo Galilei, 1616

Donald Norman

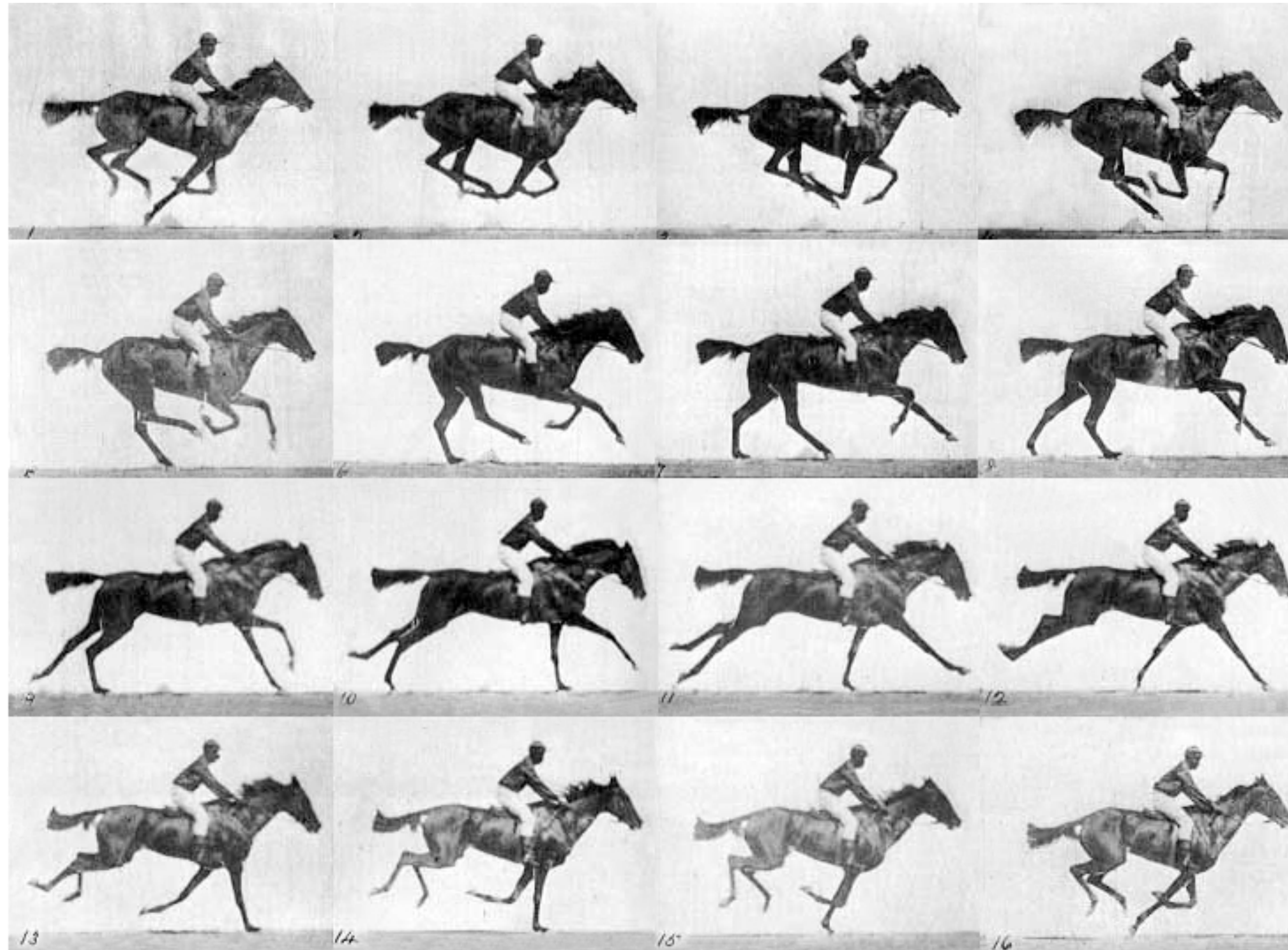


William Curtis (1746-1799)

The History of Visual Communication

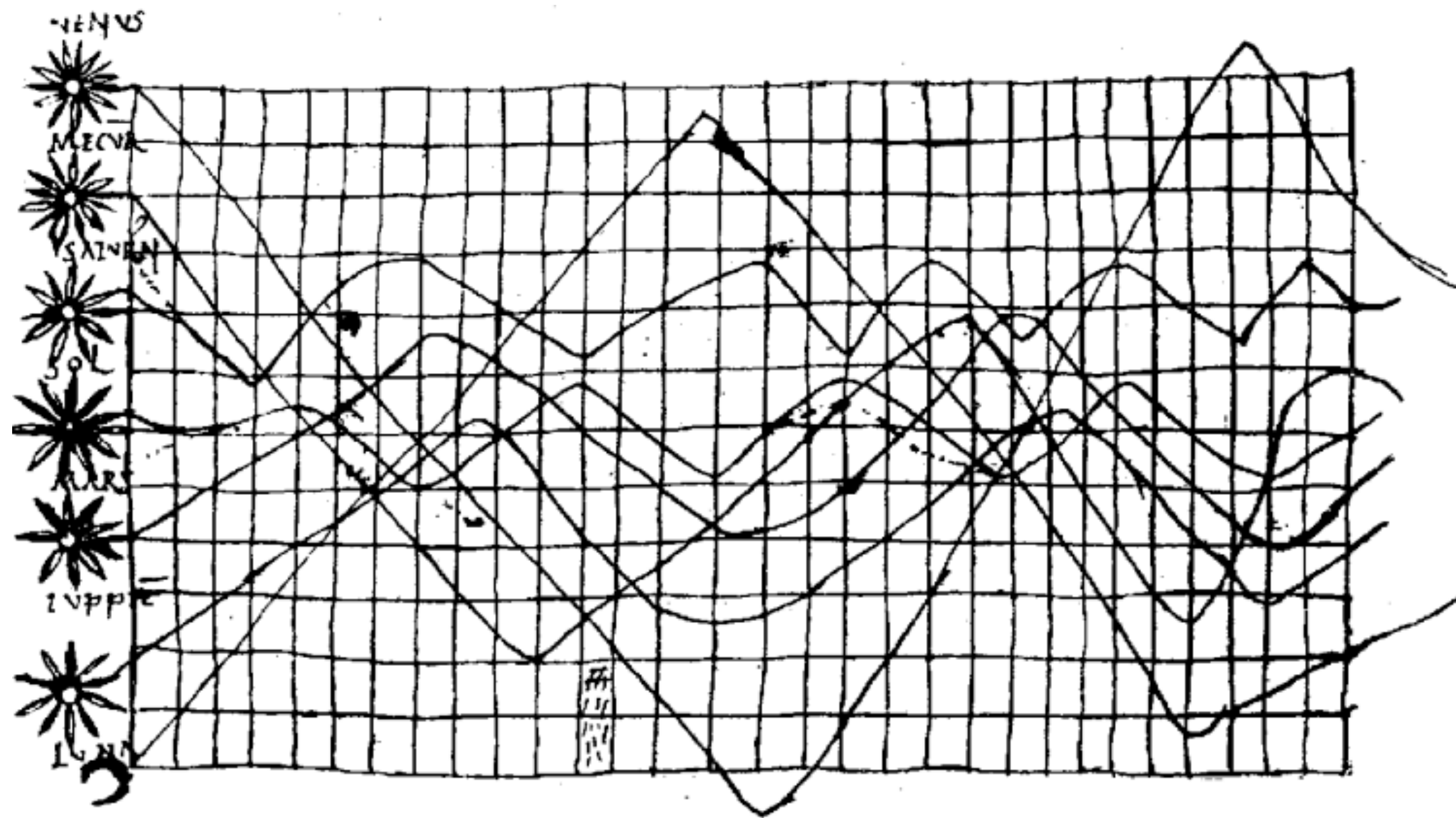
The Galileo Project, Rice University

Record

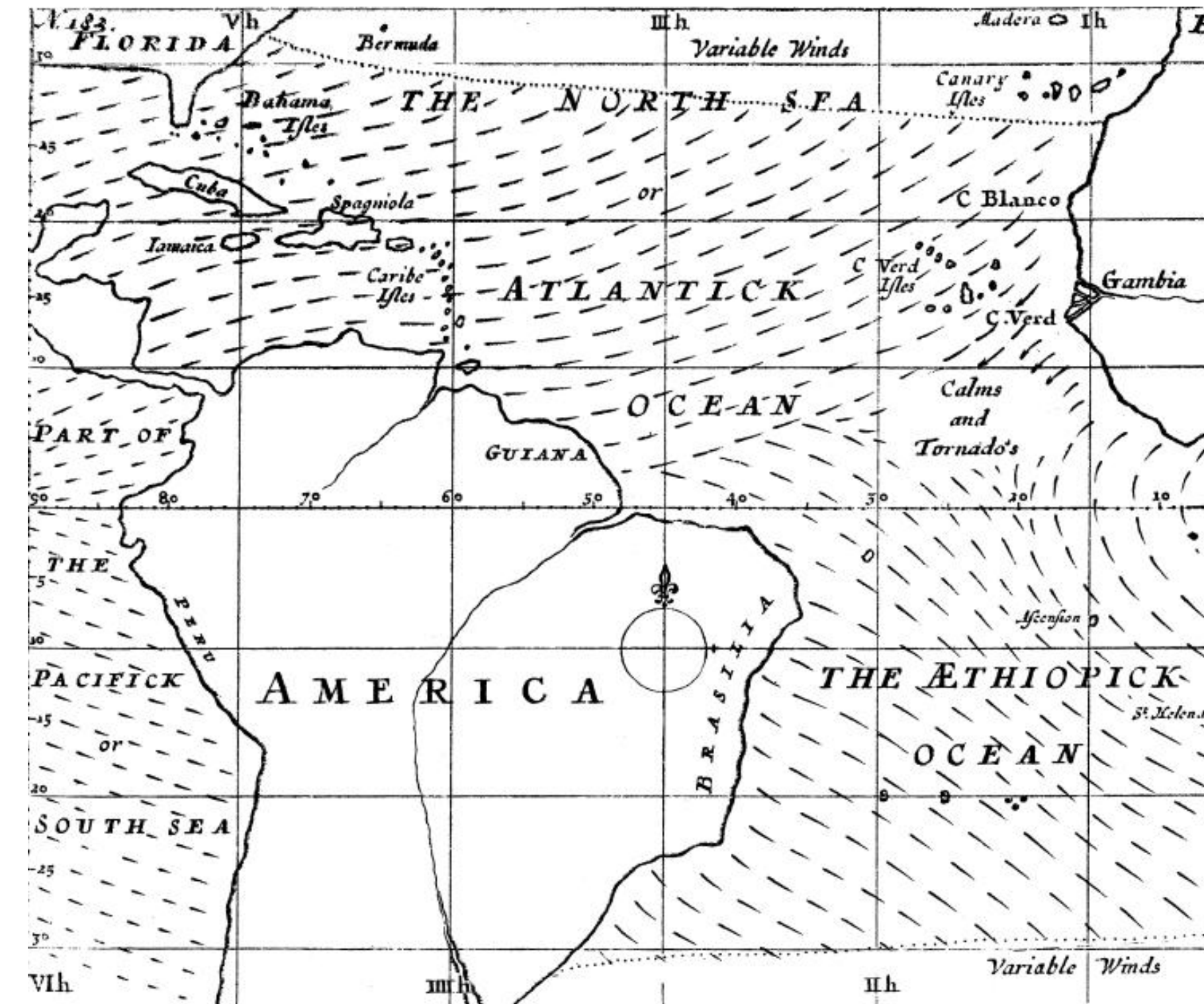


Eadweard J. Muybridge, 1878

Analyze



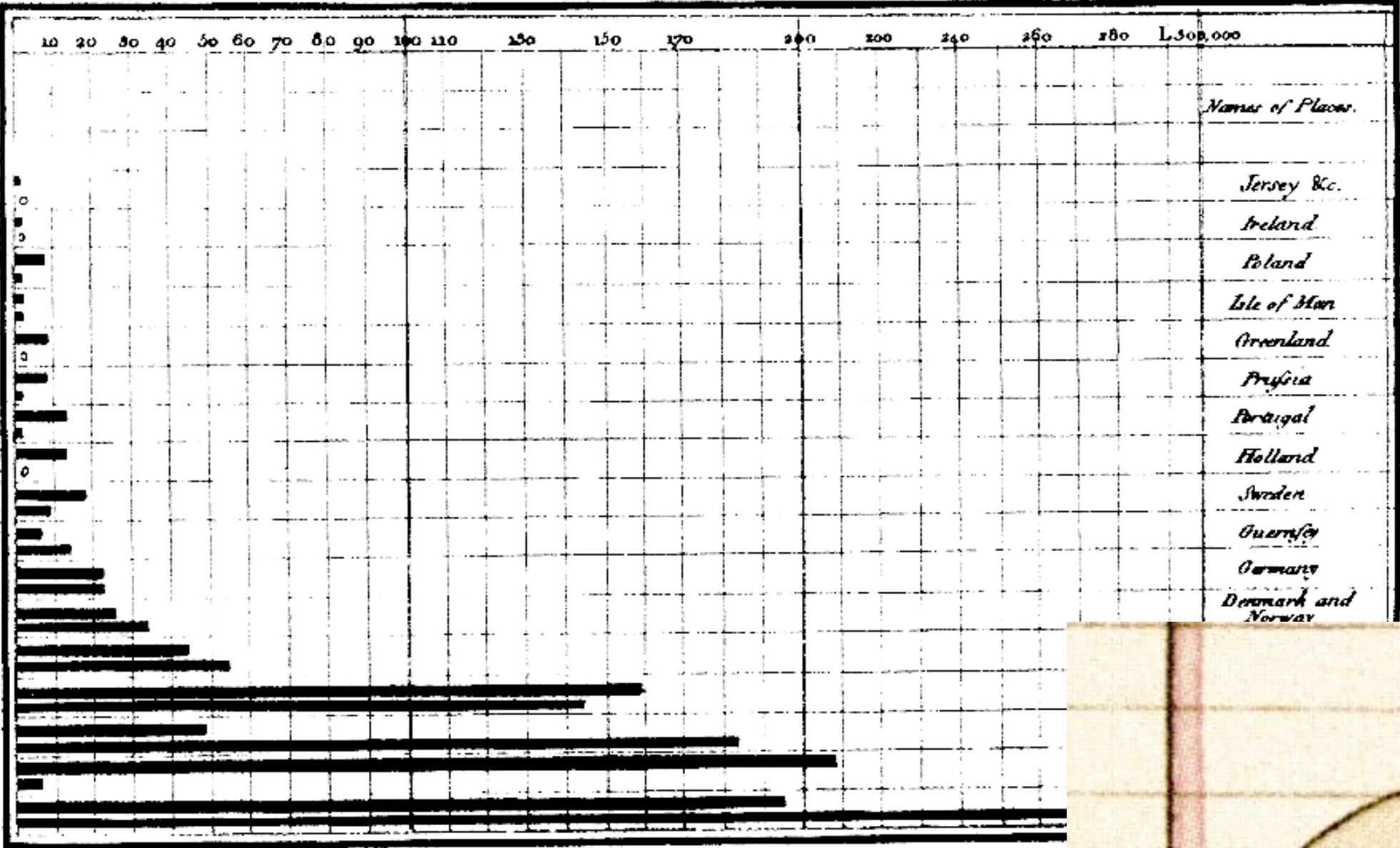
Planetary Movement Diagram, c. 950



Halley's Wind Map, 1686

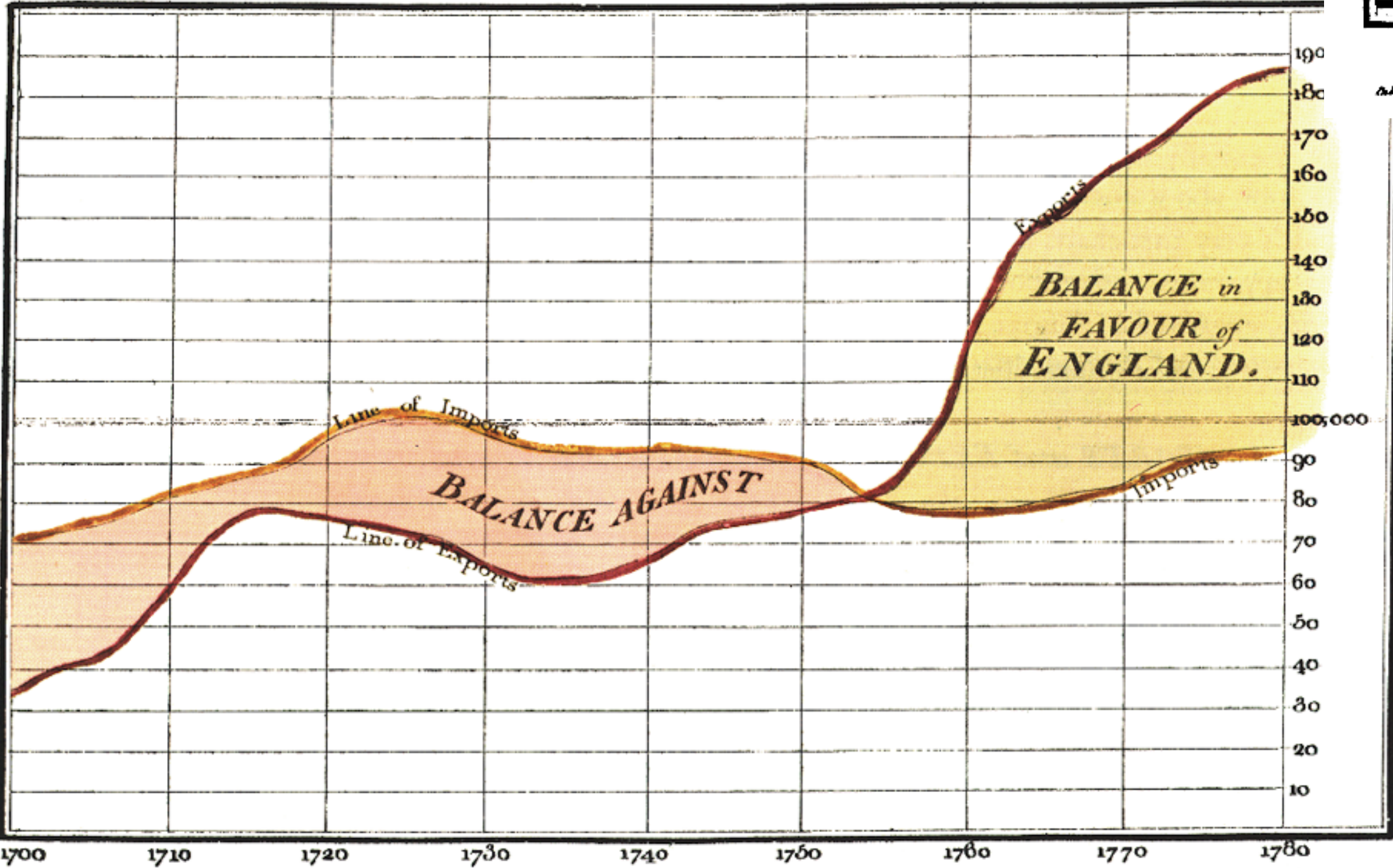
Analyze

Exports and Imports of SCOTLAND to and from different parts for one Year from Christmas 1780 to Christmas 1781.

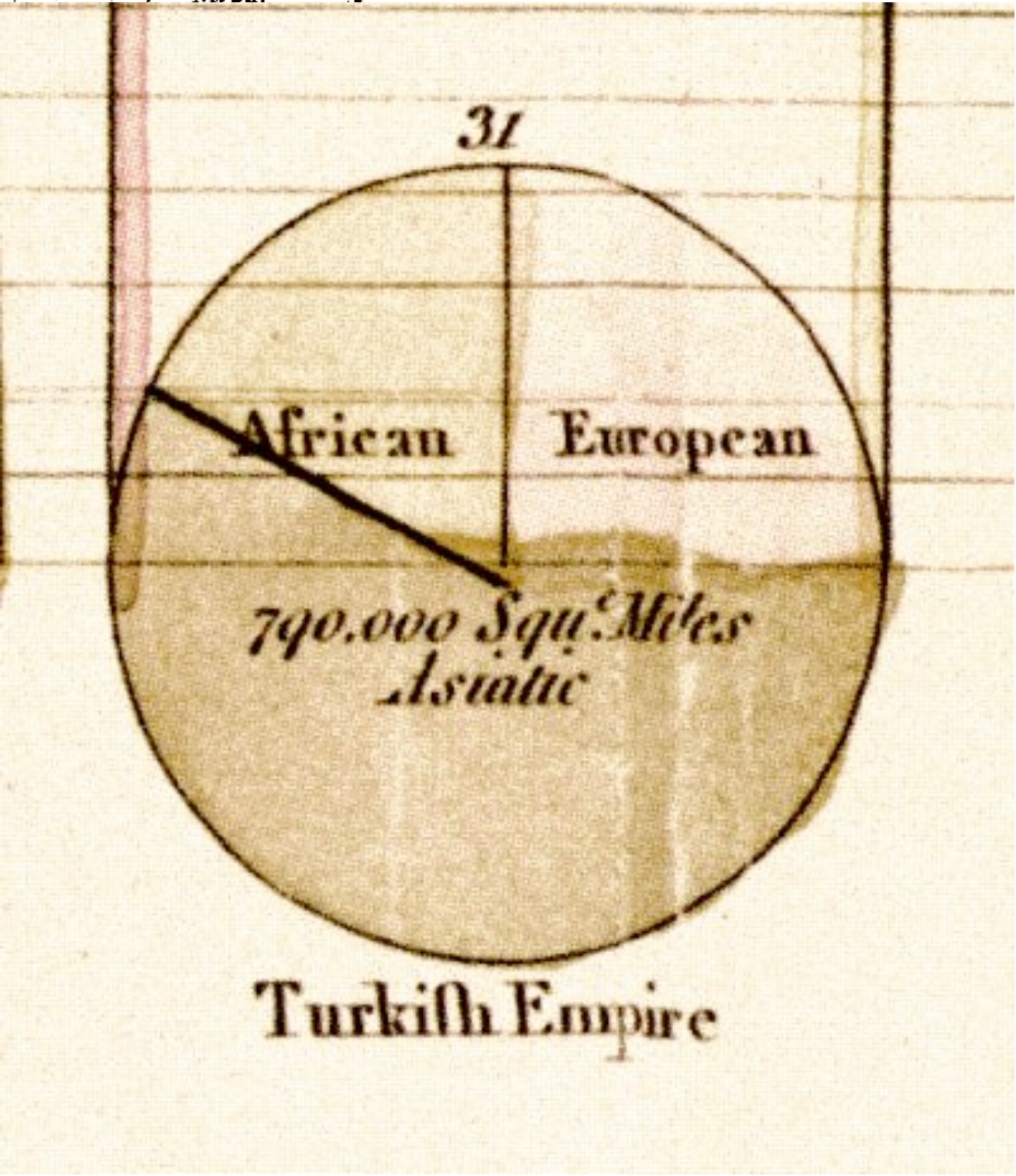


The upright divisions are Ten Thousand Pounds each. The Black Lines are Exports
Published in the *Edinburgh Journal* 9th 1781 by W. Playfair

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



W. Playfair, 1786



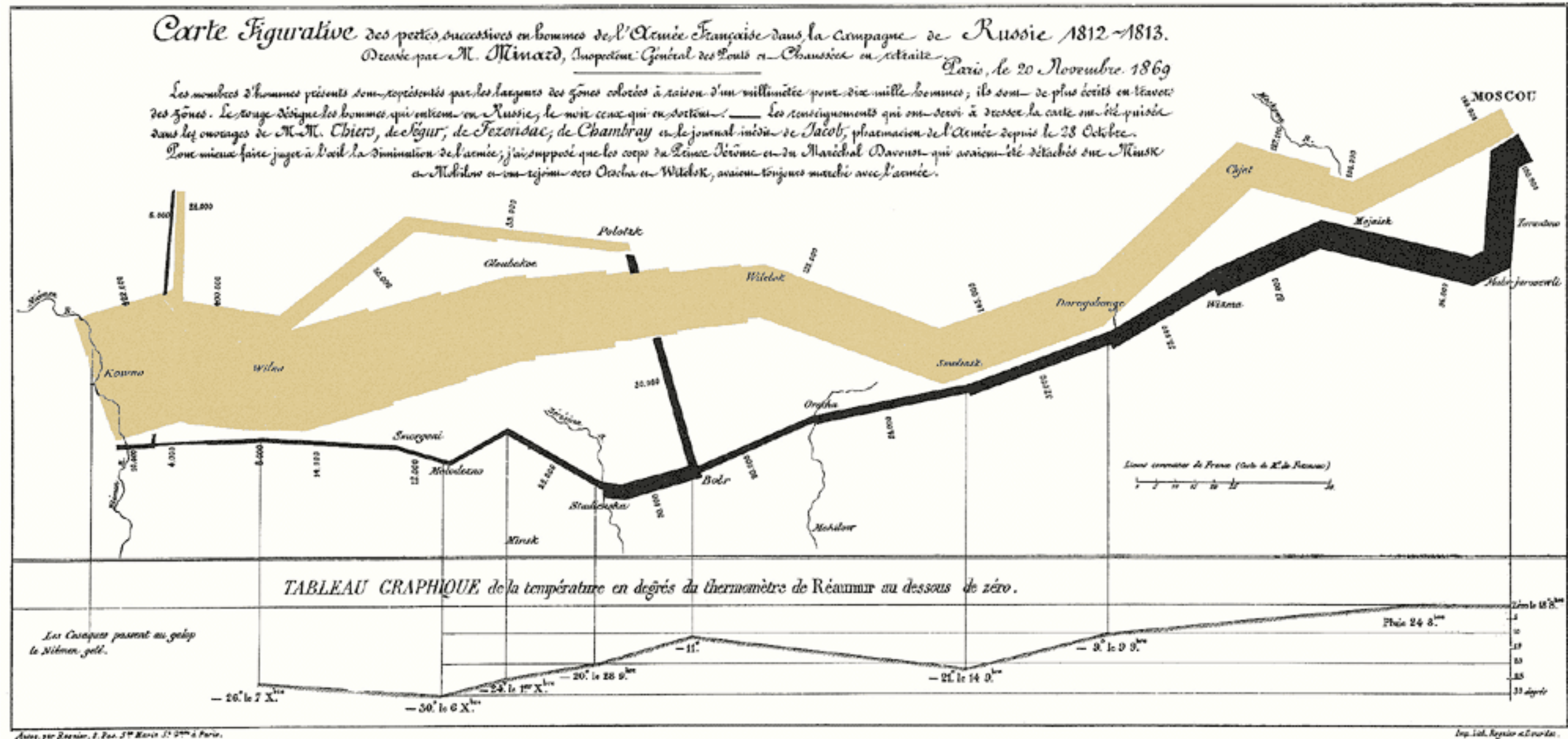
proportions of the Turkish Empire located in Asia, Europe and Africa before 1789

Find Patterns



John Snow, 1854

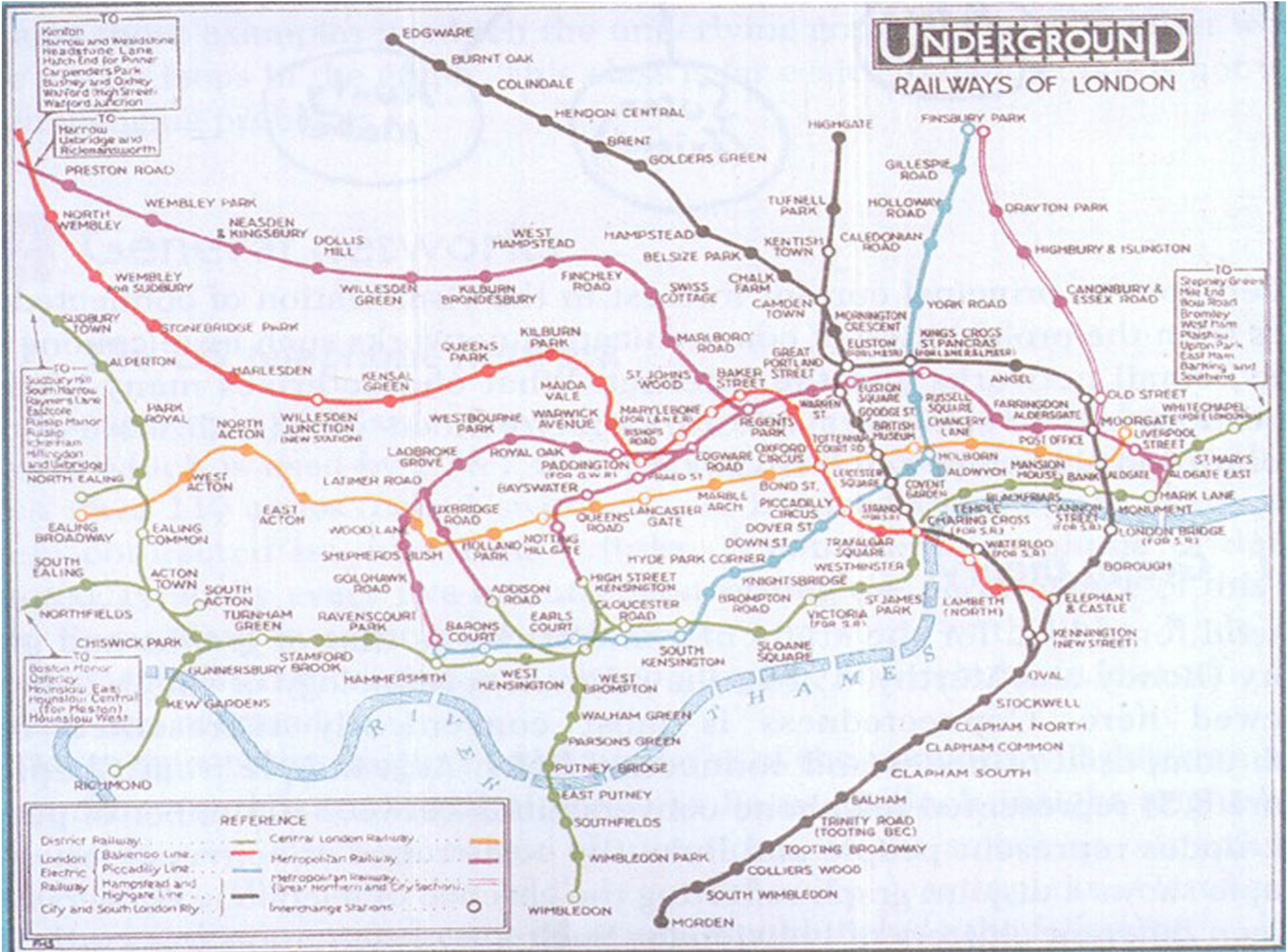
Communicate



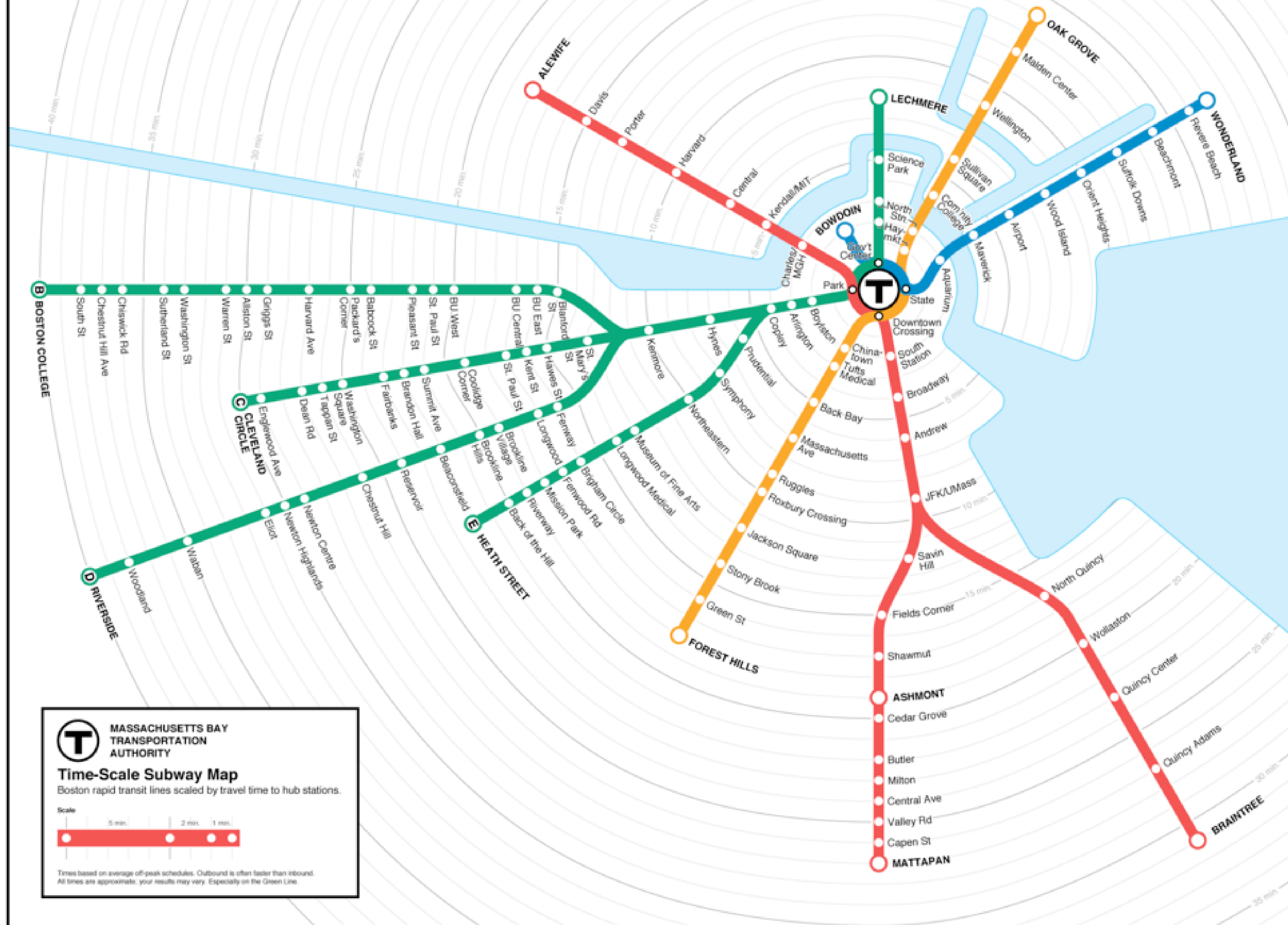


<https://history.infowetrust.com/>

Communicate



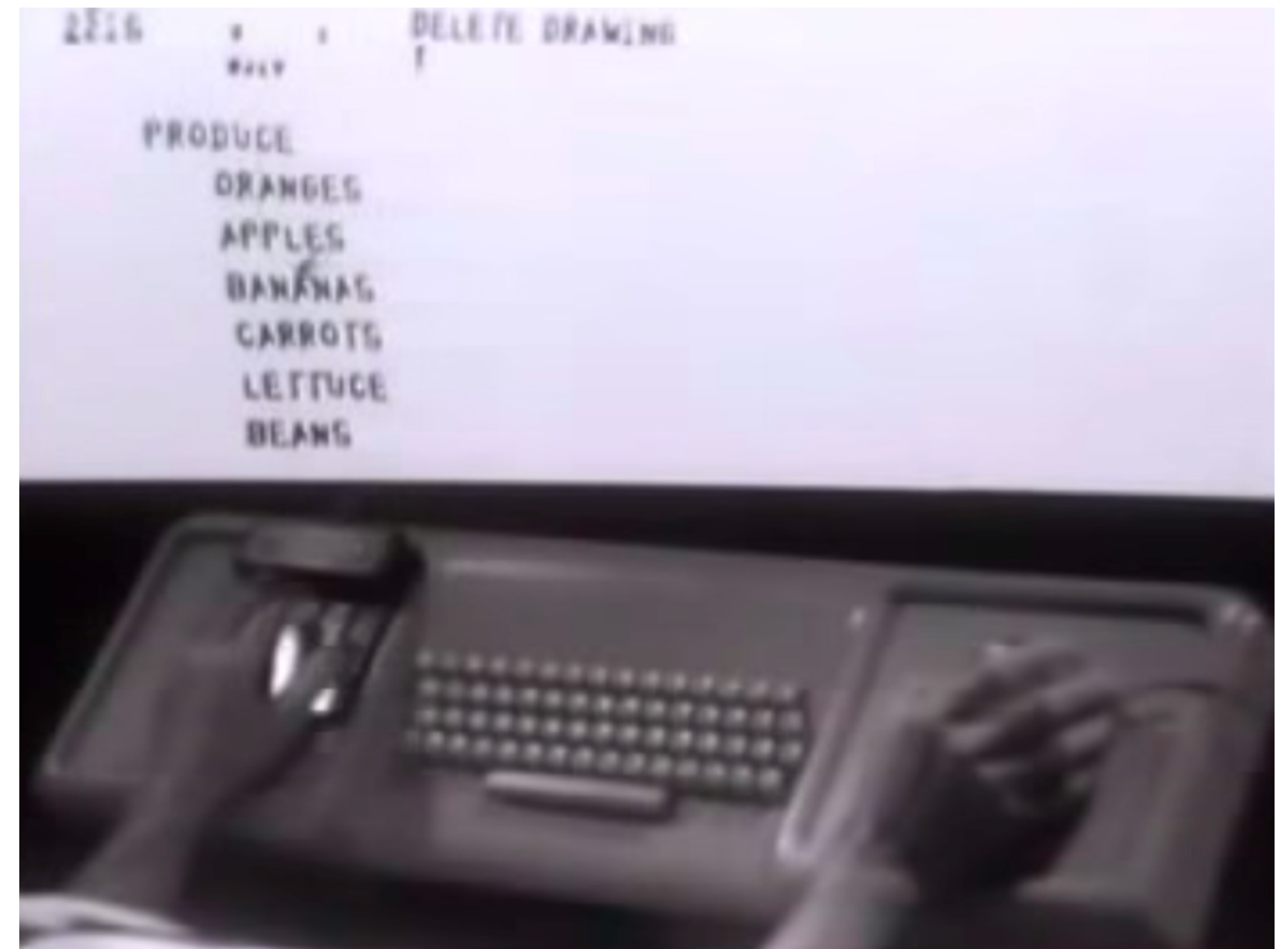
London Subway Map, 1927



Interact



Ivan Sutherland, Sketchpad, 1963



Doug Engelbart, 1968

Modern Examples

Analyze

jan

Starts with

Ends with

Contains

Exact Match

Girls

Boys

Both



Total



Compare

Jan M

Jan F

Janet F

Jane F

Janice F

Janie F

Janelle F

Jana F

Janiyah F

Janis F

Janessa F

Janae F

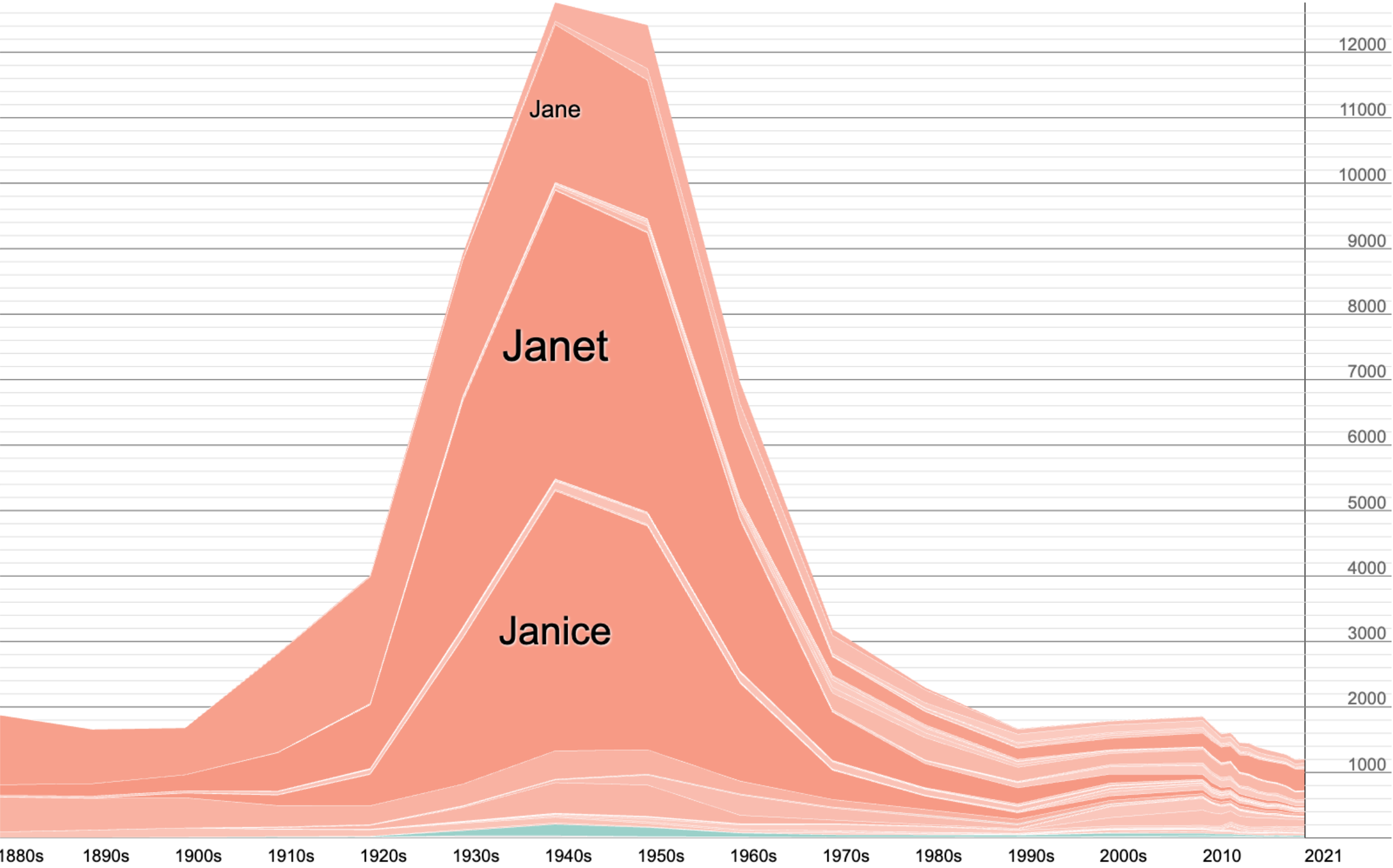
Janette F

Janiya F

Janine F

And 38 more...

per million births



Communicate



Hans Rosling, TED 2006

Who is CS-5630 / CS-6630?

Course Staff



Haihan Lin
Teaching Assistant



Kiran Gadhave
Teaching Assistant



Pranav Rajan
Teaching Assistant



Tripti Agarwal
Teaching Mentee

Alexander Lex

@alexander_lex
<http://alexander-lex.net>



Associate Professor, Computer Science

Before that: Lecturer, Postdoctoral Fellow, Harvard

PhD in Computer Science, Graz University of Technology





visualization
design lab

<http://vdl.sci.utah.edu/>



SCI Institute

Scientific Computing and Imaging Institute

Scientific Computing

Biomedical Computing

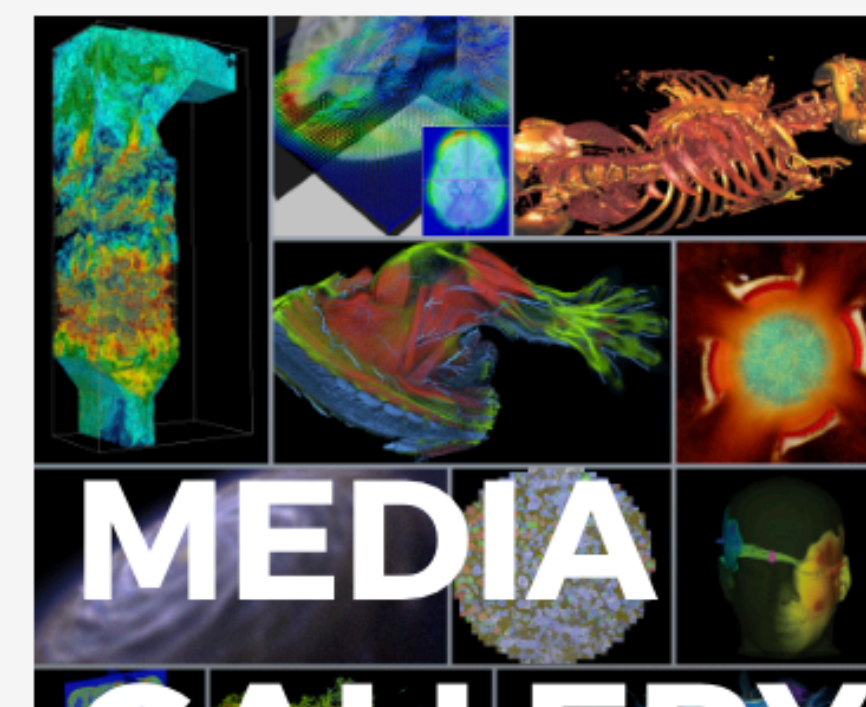
Scientific Visualization

Information Visualization

Image Analysis



<http://sci.utah.edu>



TECHNICAL CONTRIBUTIONS

**Novel Visualization
Techniques**

**Visualization Process
Innovations**

**Data Wrangling
Methods**

DOMAIN DRIVEN TECHNIQUES

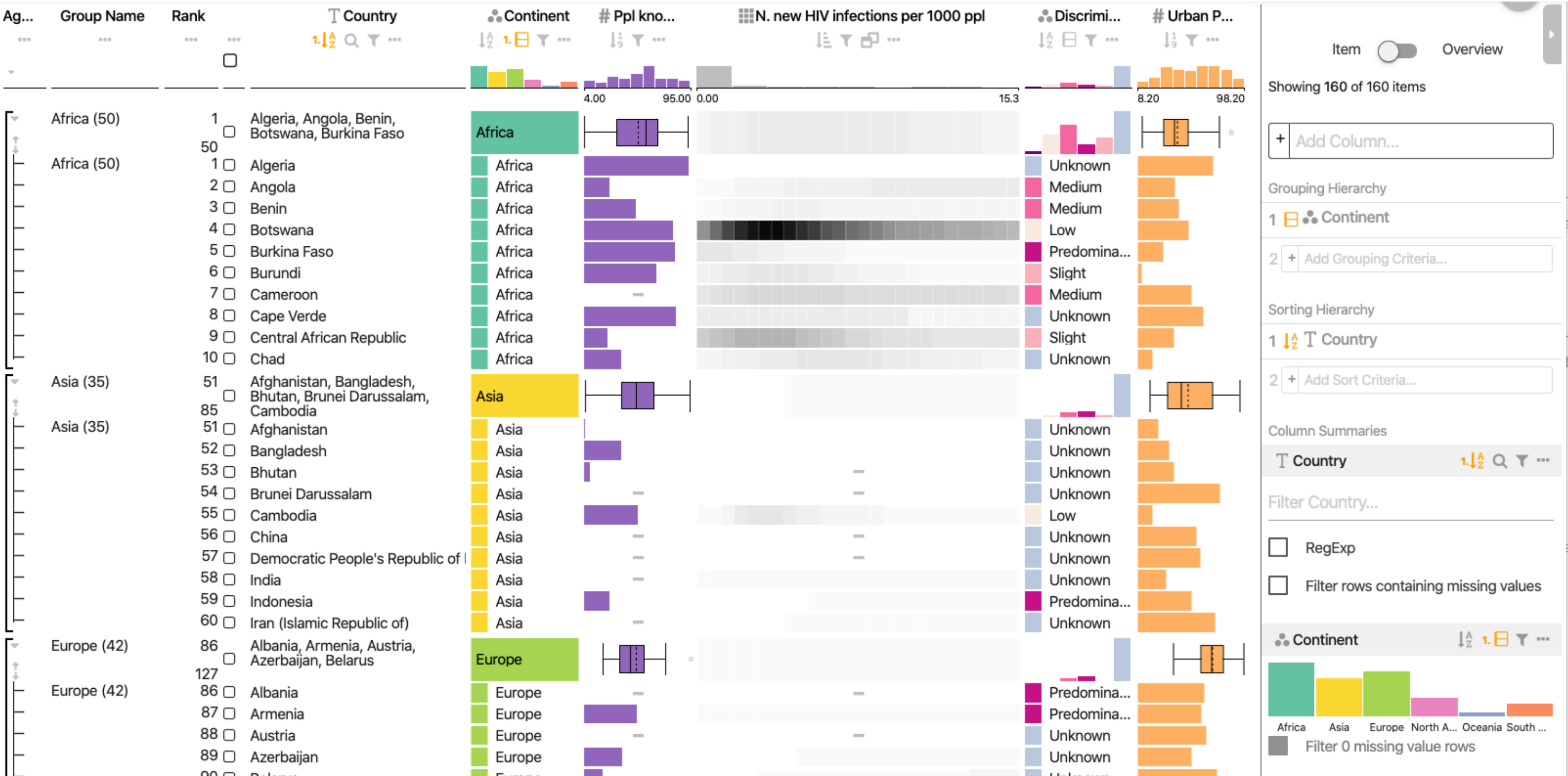
**Tailored Methods and
Systems for High
Impact Science
Problems**

EMPIRICAL & THEORETICAL WORK

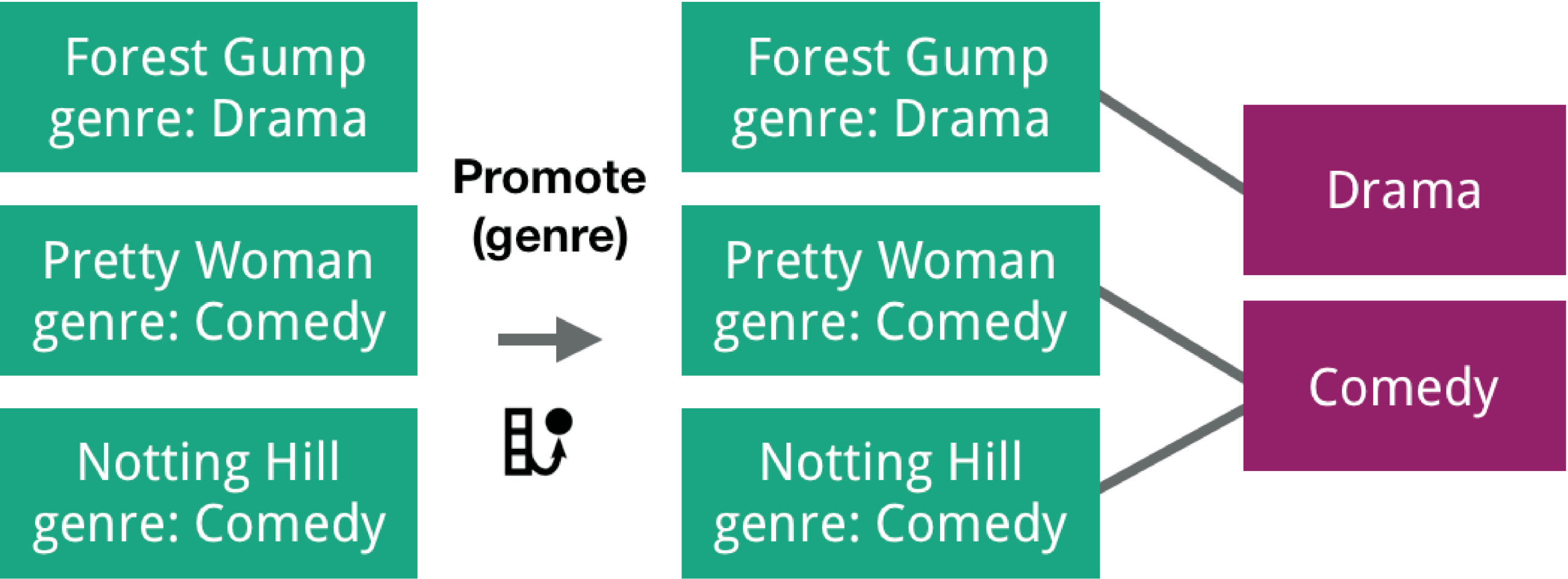
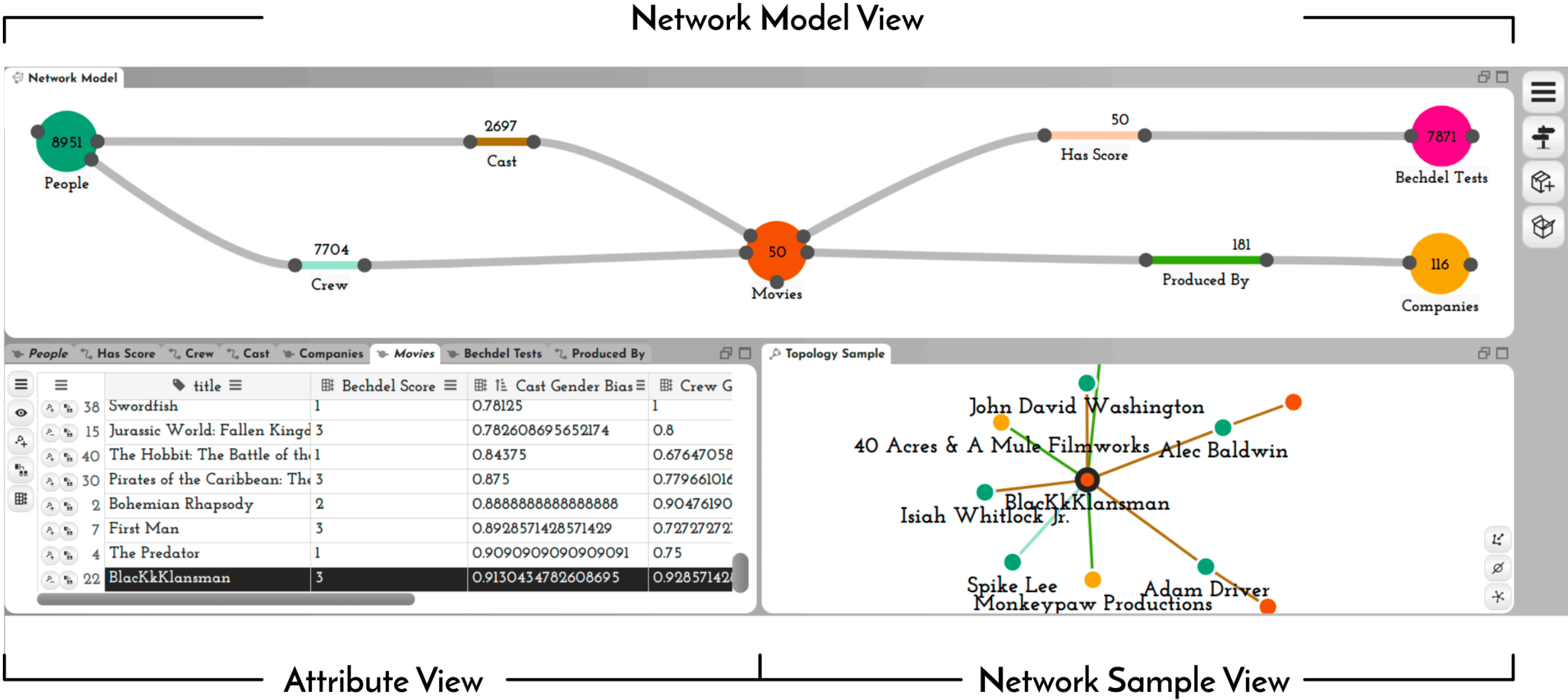
**Evaluation
Methodology**

**Design Spaces /
Taxonomies**

Novel Visualization Techniques



Data Wrangling
Methods



TECHNICAL CONTRIBUTIONS

**Data Wrangling
Methods**

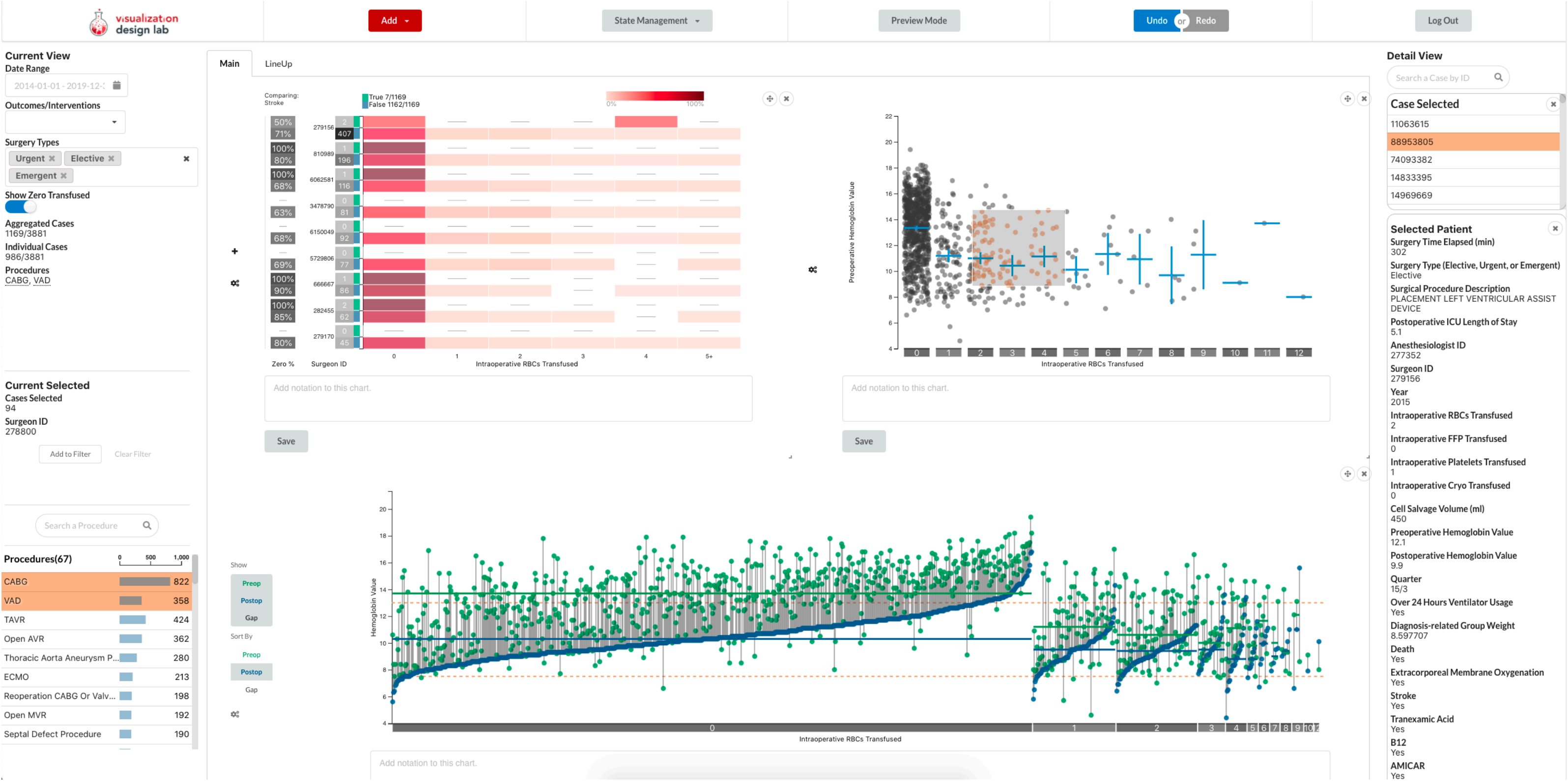
DOMAIN DRIVEN TECHNIQUES

**Tailored Methods and
Systems for High
Impact Science
Problems**

DOMAIN DRIVEN TECHNIQUES

Tailored Methods and Systems for High Impact Problems

Electronic Health Records



DOMAIN DRIVEN TECHNIQUES

**Tailored Methods and
Systems for High
Impact Science
Problems**

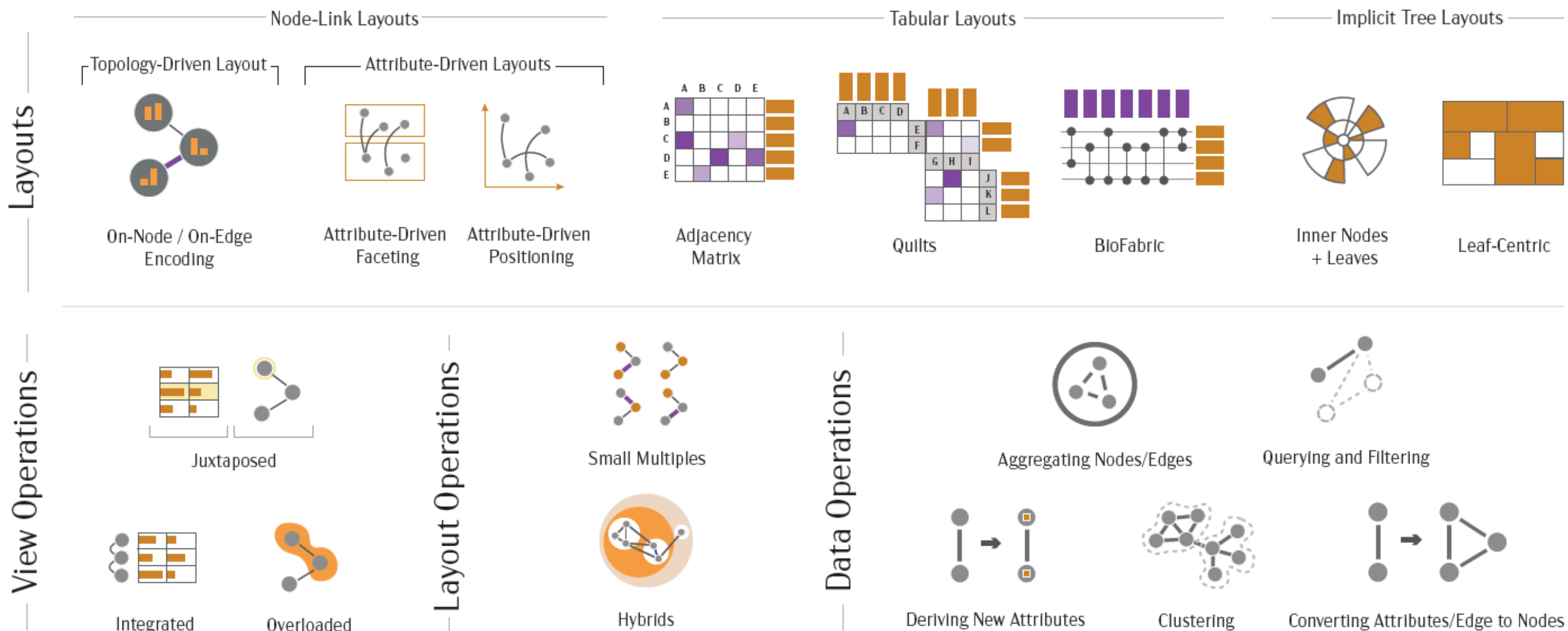
EMPIRICAL & THEORETICAL WORK

**Evaluation
Methodology**

**Design Spaces /
Taxonomies**

Evaluation
Methodology

Design Spaces /
Taxonomies



Structure & Goals

Course Goals. You will learn:

How to **efficiently visualize data**

Evaluate and **critique** visualization designs

Apply fundamental principles & techniques

Design visual data analysis solutions

Implement interactive data visualizations

Web development skills

Course Components

Theory

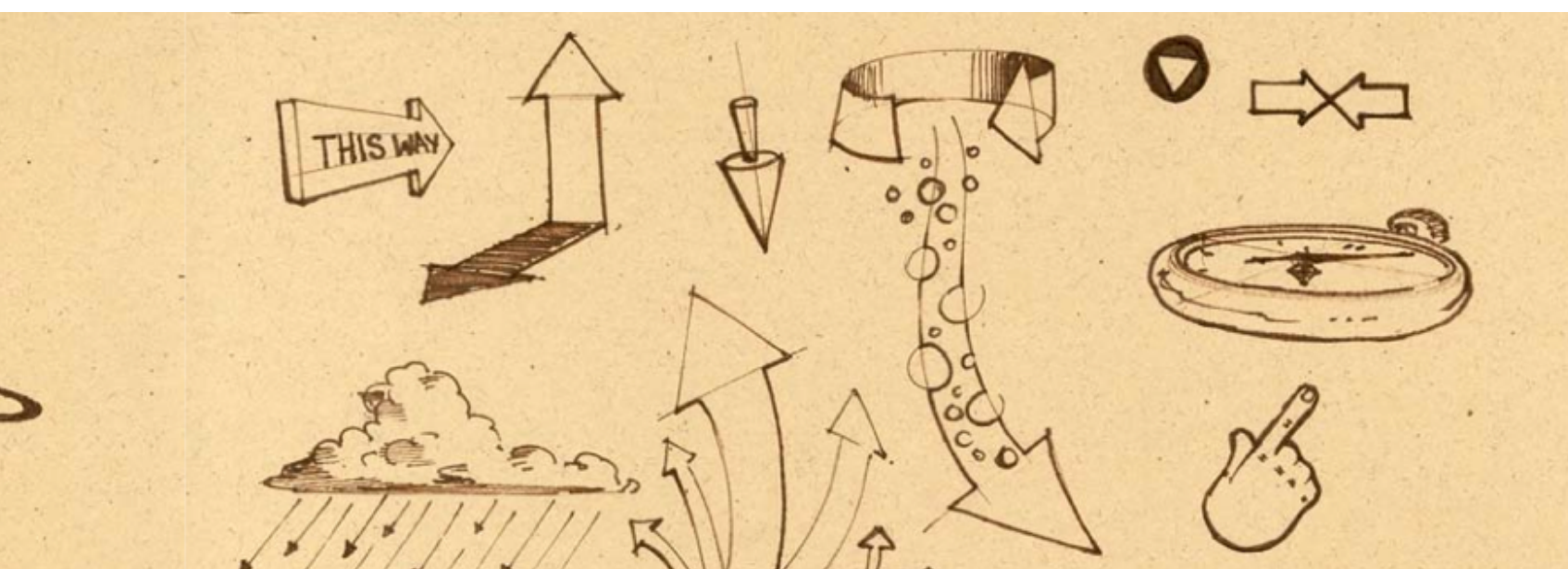
Lecture
Reading
Discussion

Design Critiques
Redesigns

Labs
D3 reading
Self-study
Office hours

Design Skills

Coding Skills



```
<!DOCTYPE html>
<meta charset="utf-8">
<style>

text {
  font: 10px sans-serif;
}

</style>
<body>
<script src="http://d3js.org/d3.v3.min.js"></script>
<script>
```


Course Components

Lectures: introduce theory

Design Critiques / Redesigns: develop “an eye” for vis design, critique, learn by example;

Will happen in breakout groups,

Submit 5 critiques, within 2 days after posting

Labs: short coding tutorials, examples

Based on a published script on website

Strongly related to homework assignments

Homeworks help practice specific skills

Final Project gives you a chance to go through a complete vis project

Schedule

Lectures: Tuesday and Thursday 2:00-3:20 pm

Labs: Mondays, 5:00-6:00 pm, L110 (scheduled on demand)

Review Lectures:
[YouTube Channel](#)

Three Parts:

I. Technical Foundations

HTML, Javascript, D3

II. Visualization Fundamentals

Perception, Visual encodings, Design Guidelines, Tasks..

III. Abstract Data Visualization

Tables, Graphs, Maps

Schedule

CS 5630/6630

Today◀▶September 2022

PrintWeekMonthAgenda

Mon	Tue	Wed	Thu	Fri	Sat	Sun
29 Week 2	30 10:00 Tripti's Office H 14:00 Vis for Data Sc 16:00 Alexander Lex	31 09:30 Kiran Office H	Sep 1 10:00 Pranav Rajan (C 14:00 Vis for Data Sc	2 Drop Deadline HW1 Due 10:00 Haihan's Office	3	4
5 Week 3	6 10:00 Tripti's Office H 14:00 Vis for Data Sc 16:00 Alexander Lex	7 09:30 Kiran Office H	8 10:00 Pranav Rajan (C 14:00 Vis for Data Sc	9 HW2 Due 10:30 Haihan's Office	10	11
12 Week 4	13 10:00 Tripti's Office H 14:00 Vis for Data Sc 16:00 Alexander Lex	14 09:30 Kiran Office H	15 10:00 Pranav Rajan (C 14:00 Vis for Data Sc	16 HW3 Due 10:00 Haihan's Office	17	18
19 Week 5	20 10:00 Tripti's Office H 14:00 Vis for Data Sc	21 09:30 Kiran Office H	22 10:00 Pranav Rajan (C 14:00 Vis for Data Sc	23 10:00 Haihan's Office	24	25
26 Week 6	27 10:00 Tripti's Office H 14:00 Vis for Data Sc 16:00 Alexander Lex	28 09:30 Kiran Office H	29 10:00 Pranav Rajan (C 14:00 Vis for Data Sc	30 HW4 Due 10:00 Haihan's Office	Oct 1	2

Events shown in time zone: Mountain Time - Denver

+GoogleCalendar

Subject to change

Week 1

Introduction

Tuesday

What is visualization? Why is it important? Who are we? Course overview.
[Download slides](#)

Recommended reading

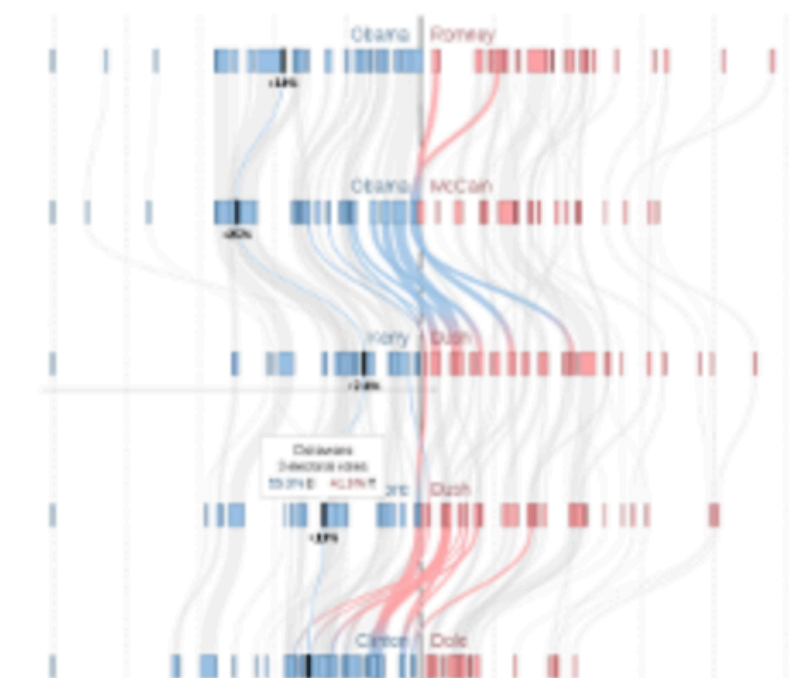
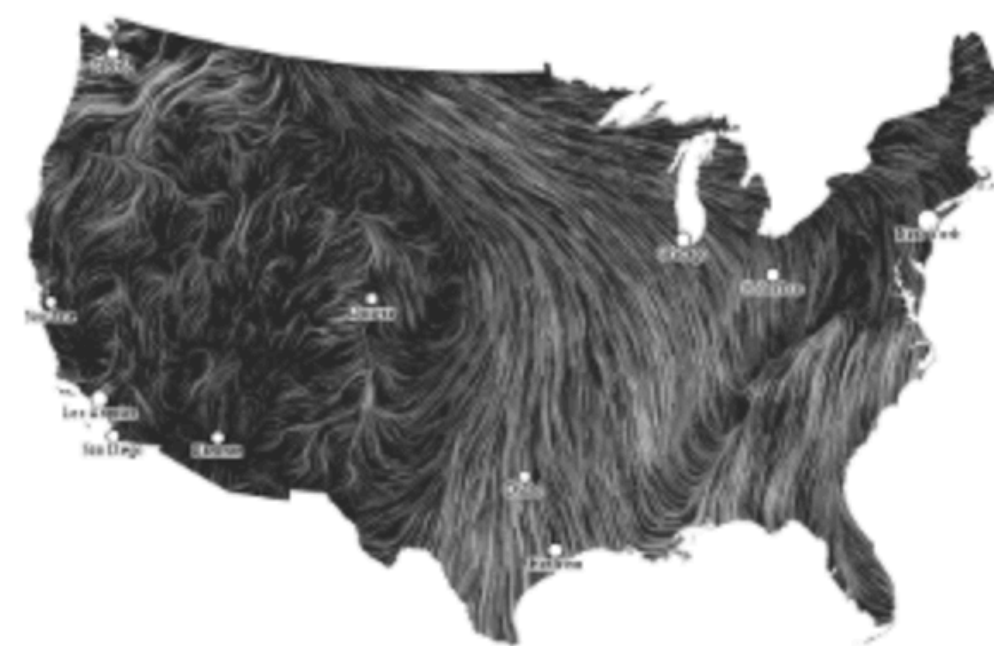
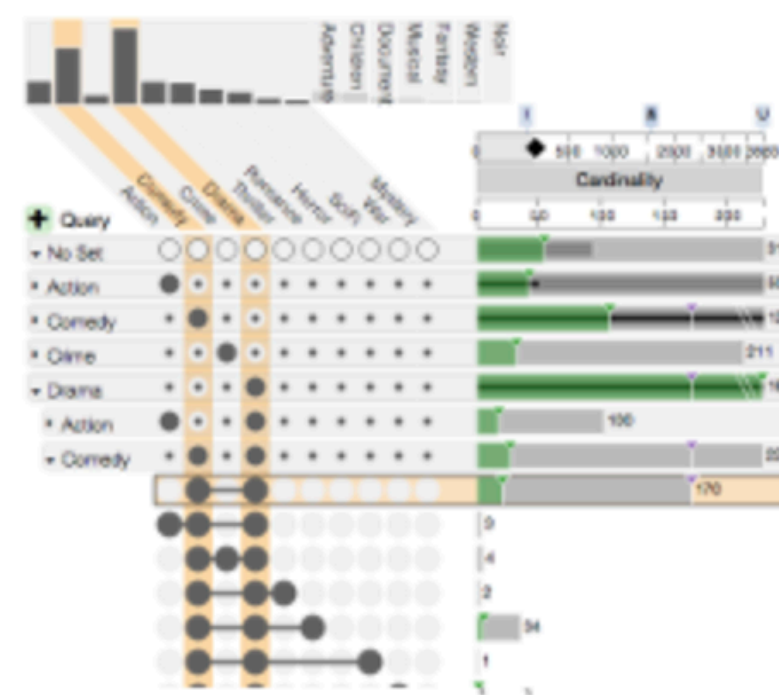
- [A Tour through the Visualization Zoo](#). Jeffrey Heer, Michael Bostock, Vadim Ogievetsky. Communications of the ACM, 53(6), pp. 59-67, Jun 2010.

Information <http://dataviscourse.net>

Visualization for Data Science CS-5630 / CS-6630



[Home](#) [Syllabus](#) [Schedule](#) [Project](#) [Resources](#) [Fame](#)



UpSet visualizing intersecting sets | Wind map | How states have shifted

The amount and complexity of information produced in science, engineering, business, and everyday human activity is increasing at staggering rates. The goal of this course is to expose you to visual representation methods and techniques that increase the understanding of complex data. Visualization for data discovery and communication is an important part of the data science pipeline. Good visualizations not only present a visual interpretation of data, but do so by improving comprehension, communication, and decision making.

In this course you will learn about the fundamentals of perception, the theory of visualization, good design practices for visualization, and how to develop your own web-based visualizations using HTML5, CSS, JavaScript, SVG, and [D3](#).

Communicate

Slack

<https://visfordatascience2022.slack.com/>

Please use slack for all general questions - code, concepts, etc.

Please don't use personal messages to me or TAs

Only use e-mail for personal inquiries

Canvas

<https://utah.instructure.com/courses/803252>

Homework submissions, Grades

Office Hours

Alex: Tuesdays, 4-5pm

TAs: starting next week

E-Mail

alex@sci.utah.edu

New U policy: must use utah.edu e-mail address to communicate (FERPA)

Lecture Recordings / Stream

In-Person Class

Lot's of discussion and in class activities; handouts, etc.

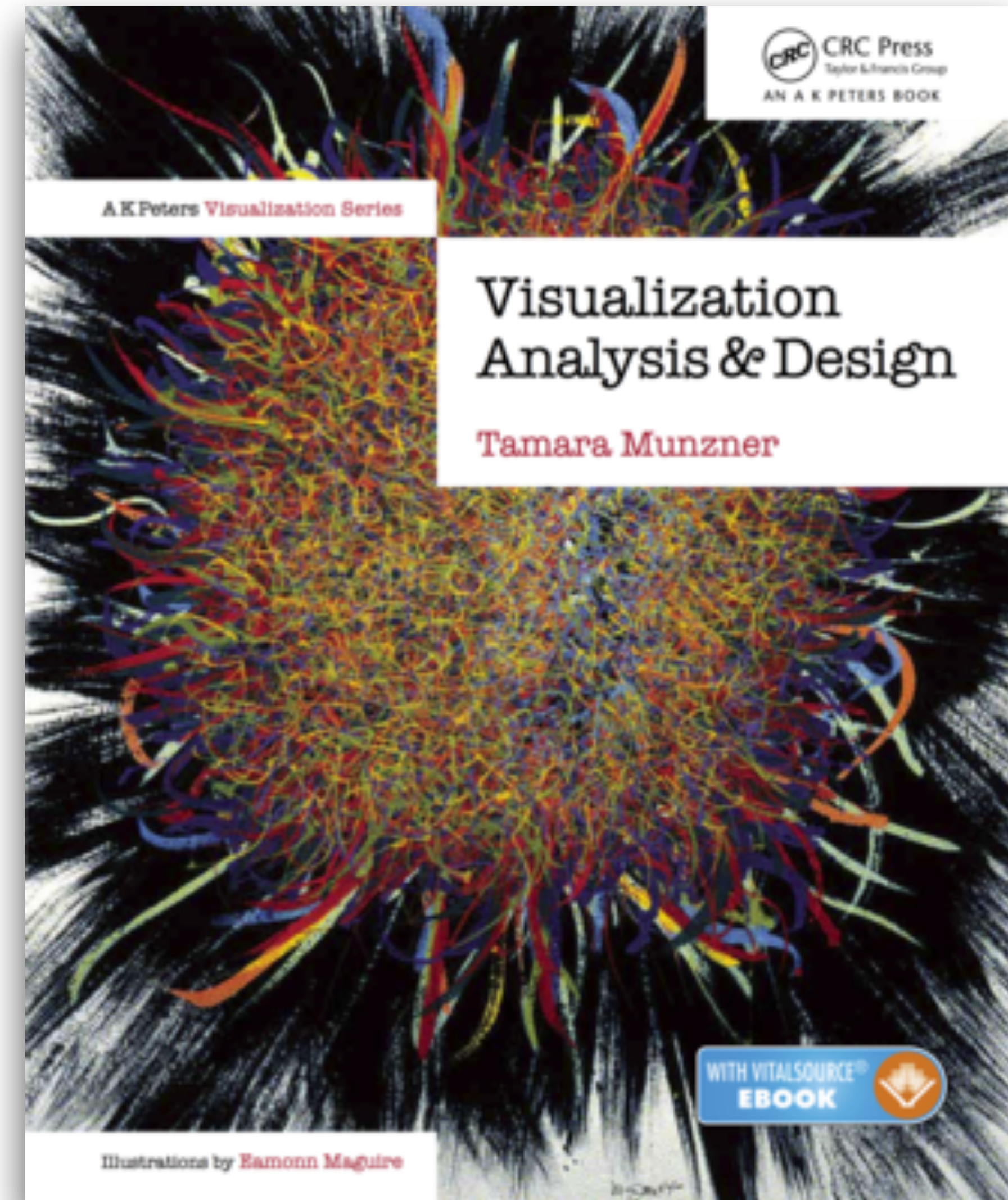
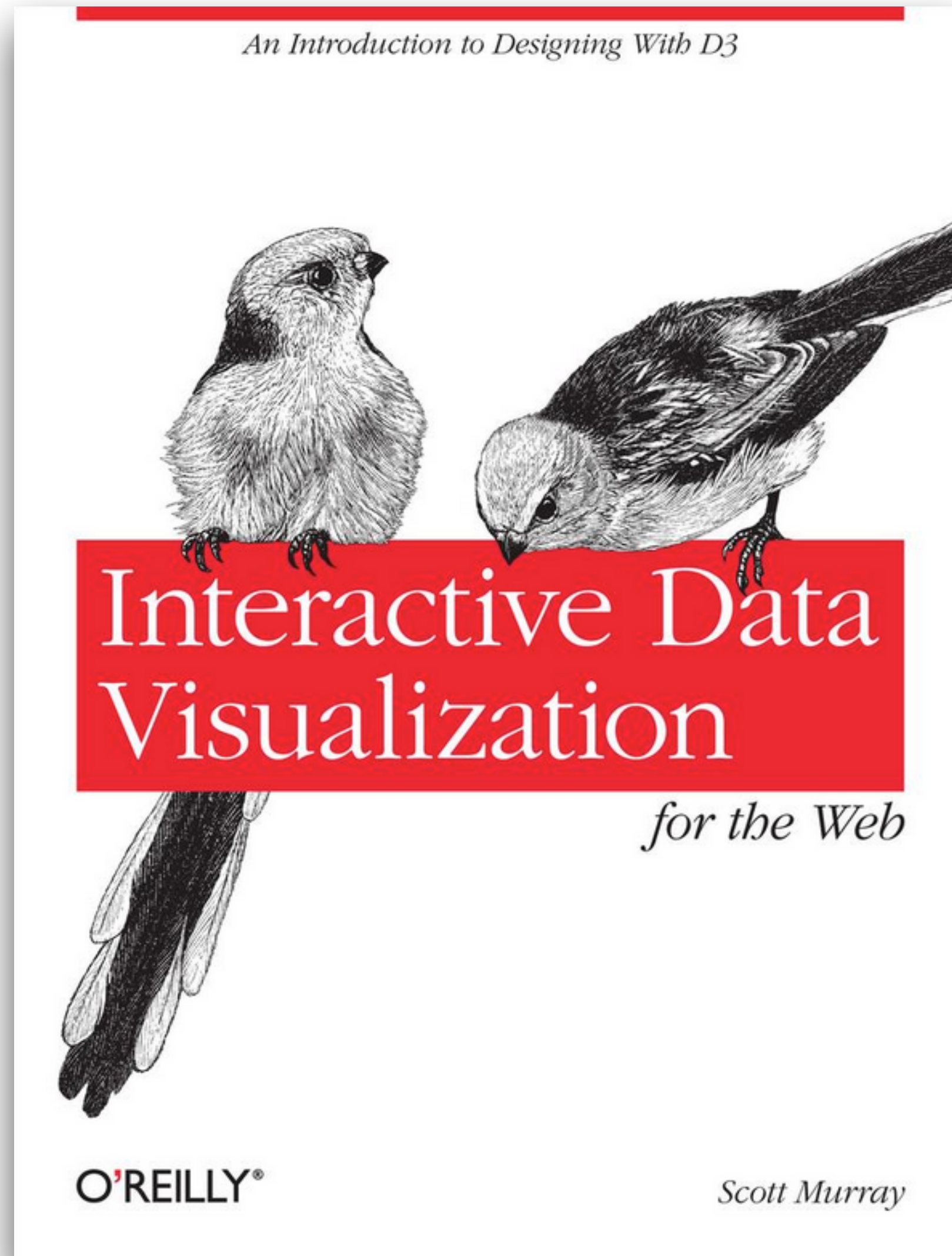
Recording as a back-up:

- If you're sick or have emergency travel

- To review lecture content

<https://www.youtube.com/channel/UCDUS80bdunpmvWVPyFRPqFQ/live>

Required Books



Programming

HTML



JS



Data-Driven Documents



Prerequisites

Programming experience

C, C++, Java, Python, etc.

Willingness to think about user-centered design

This is not your average CS course! We care about the human in the loop!

Willingness to learn new software & tools

This can be time consuming

You will need to build skills by yourself!

Formalities

How are you graded?

5-6 Homework Assignments: 35%

Varying value, 2%-10%, depending on length/difficult

Start early! Will take long if you don't know JS/D3 yet

Due on Fridays, late days: -10% per day, up to two days.

Final Project: 40%

Teams, proposal and two milestones

Exams: 20%

Two exams: last class before fall break and end of term

Group Activities: 5%

Code of Conduct

- We are committed to providing an inclusive and harassment-free environment in all interactions regardless of gender, sexual orientation, disability, physical appearance, race, or religion.
- We do not tolerate harassment in any form.
- Please report any harassment to me or the appropriate university office, which you can find at <https://safeu.utah.edu/>
- Please review the syllabus on these issues and the student code of conduct at <https://regulations.utah.edu/academics/6-400.php>

Cheating

You are welcome to **discuss** the course's ideas, material, and homework with others in order to better understand it, but **the work you turn in must be your own** (or for the project, yours and your teammate's). For example, you must **write your own code**, design your own visualizations, and critically evaluate the results in your own words.

You **may not submit the same or similar work** to this course that you have submitted or will submit to another. **Nor may you provide or make available solutions to homeworks to individuals** who take or may take this course in the future.

See also the SoC Academic Misconduct Policy:
https://www.cs.utah.edu/docs/misc/cheating_policy.pdf

You will fail the class if you cheat.

A “strike” will be recorded.

We will **automatically check for plagiarism** in all your submissions.

Cheating

Cheating is easy to catch

Do a cost-benefit analysis:

- loosing points on a homework vs failing a class (and loosing tuition)

Tools like MOSS make it easy to catch cheating

- If you have copied code in the past but have not been caught, it's likely no one checked, or they didn't want to bother with the hassle

This Week

HW0, including course survey

Lecture on Perception

Readings

D3 Book, Chapters 1-3

VDA Book, Chapter 1

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Next Week

HW1 due

Introduction to Git, HTML, CSS

Office hours start!

<https://github.com/dataviscourse/2022-homework>

 **dataviscourse** / **2022-homework** Public

 Edit Pins  Watch **3**

 **Code**

 Issues

 Pull requests

 Actions

 Projects

 Wiki

 Security

 main 

 1 branch

 0 tags

Go to file

Add file 

Code 

Alexander Lex initial release of readme / license / giti... 1fe7606 1 hour ago  1 commit

 hw0	initial release of readme / license / gitignore an...	1 hour ago
 .gitignore	initial release of readme / license / gitignore an...	1 hour ago
 LICENSE	initial release of readme / license / gitignore an...	1 hour ago
 README.md	initial release of readme / license / gitignore an...	1 hour ago

 README.md 

Homeworks for Utah's Vis for Data Science Course

In subfolders in this directory you will find the homeworks for CS 6630 / CS 5630 / DS 4630 – Visualization for Datascience.