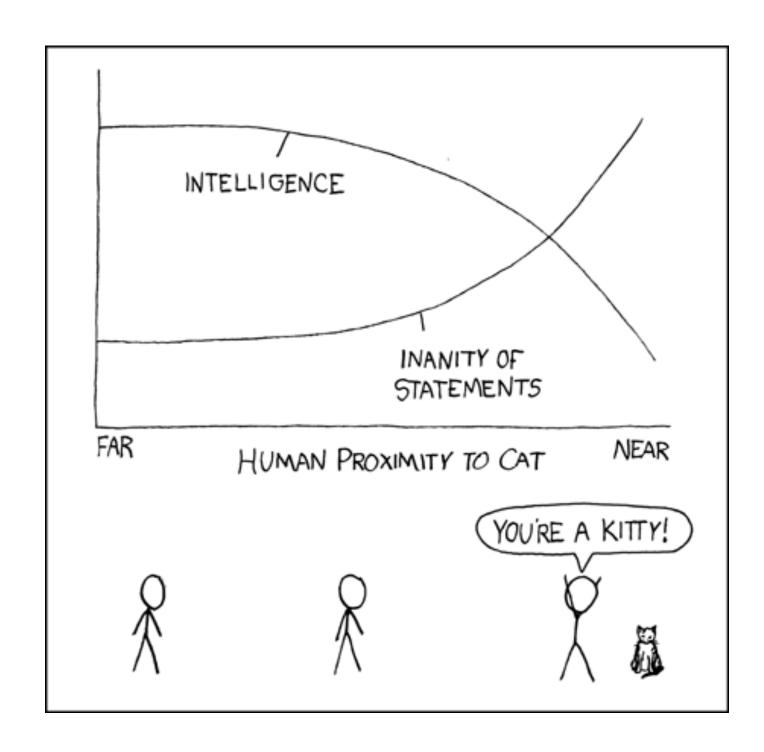
### CS-5630 / CS-6630 Uisualization for Data Science

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

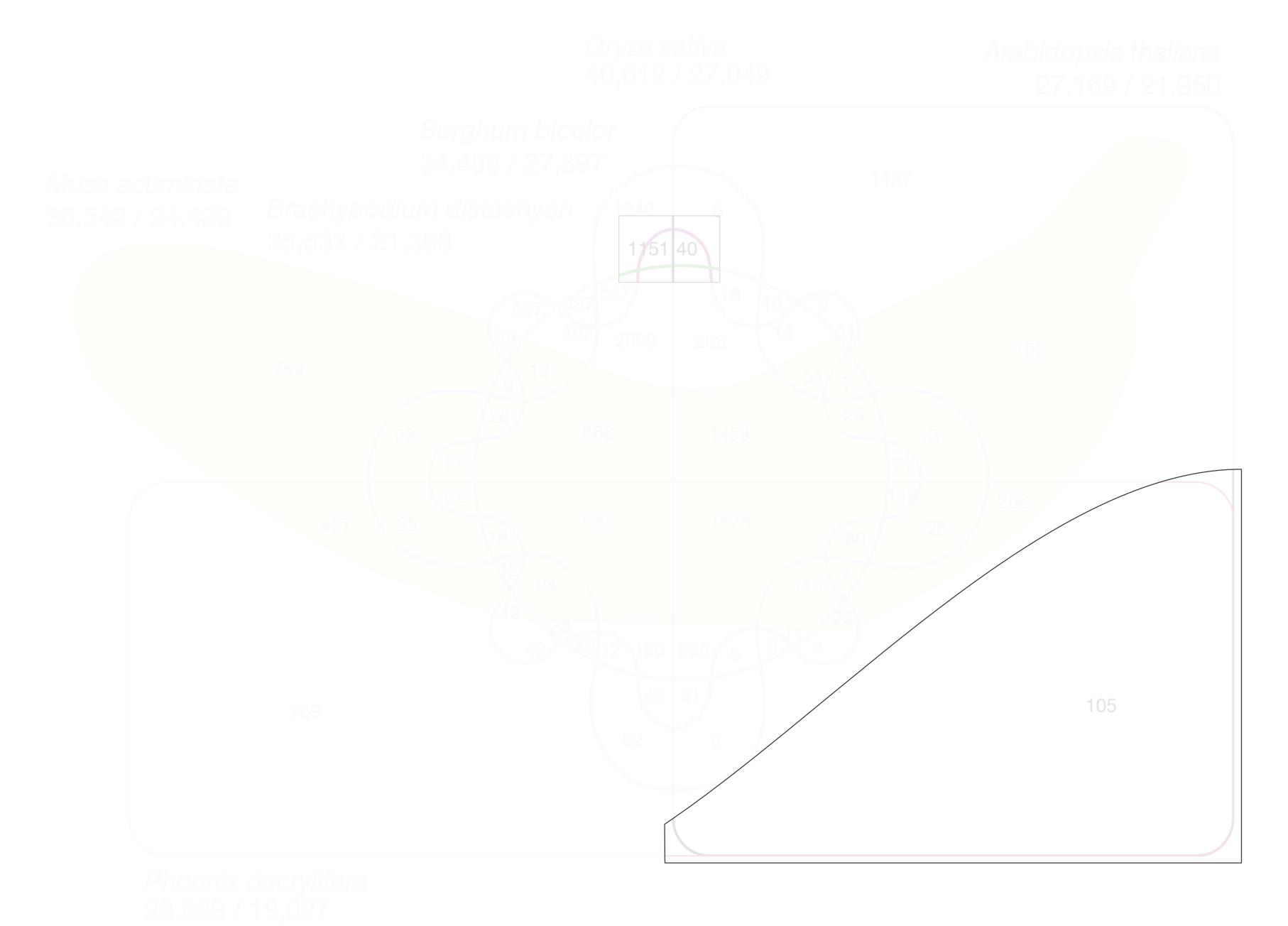
Date P. dactylifera

Cress Arabidopsis thaliana

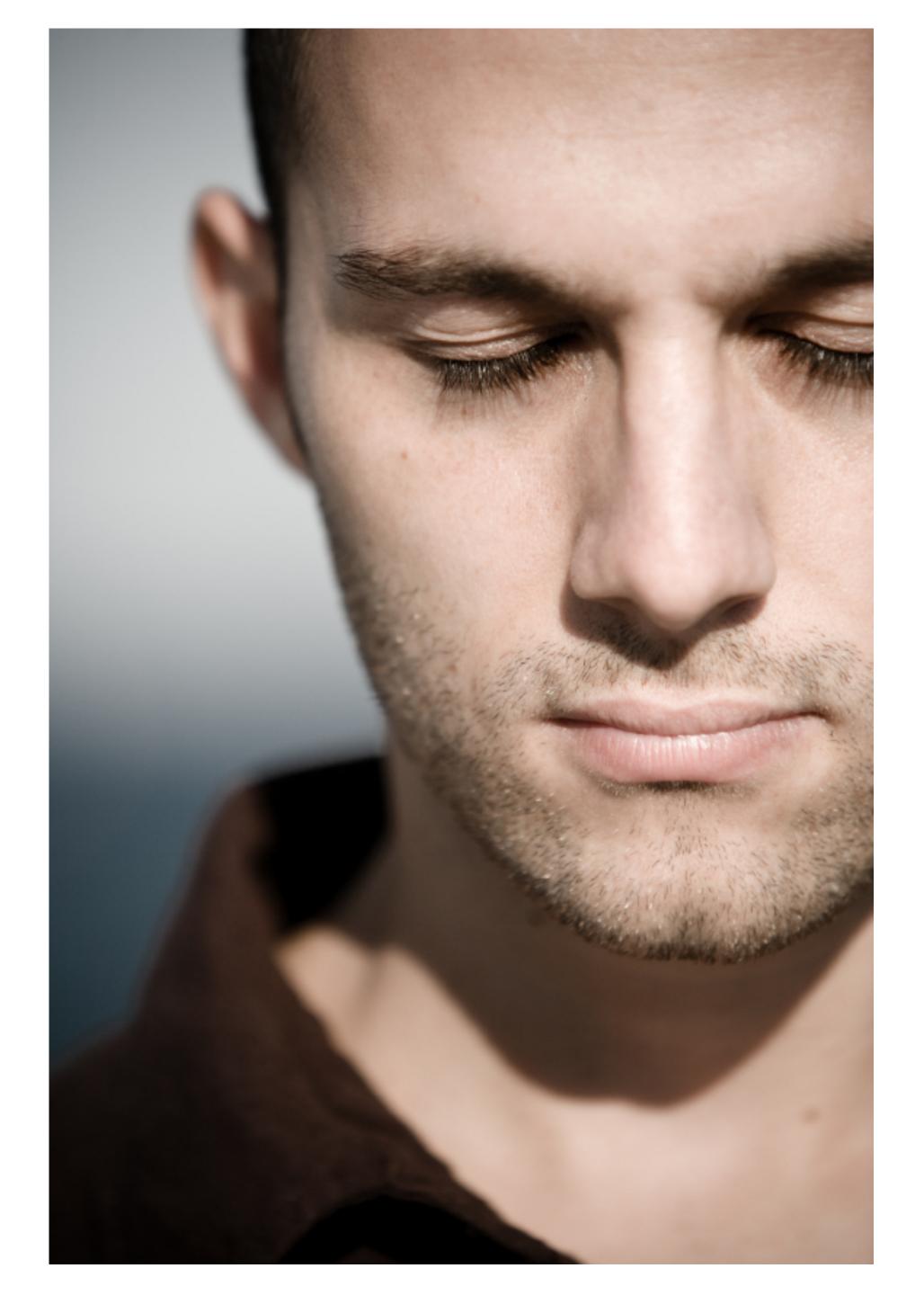
Rice Oryza sativa

Sorghum Sorghum bicolor

Brome Brachypodium distachyon



D'Hont et al., Nature, 2012



- vi · su · al · i · za · tion
- I. 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

### Uisualization

"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?

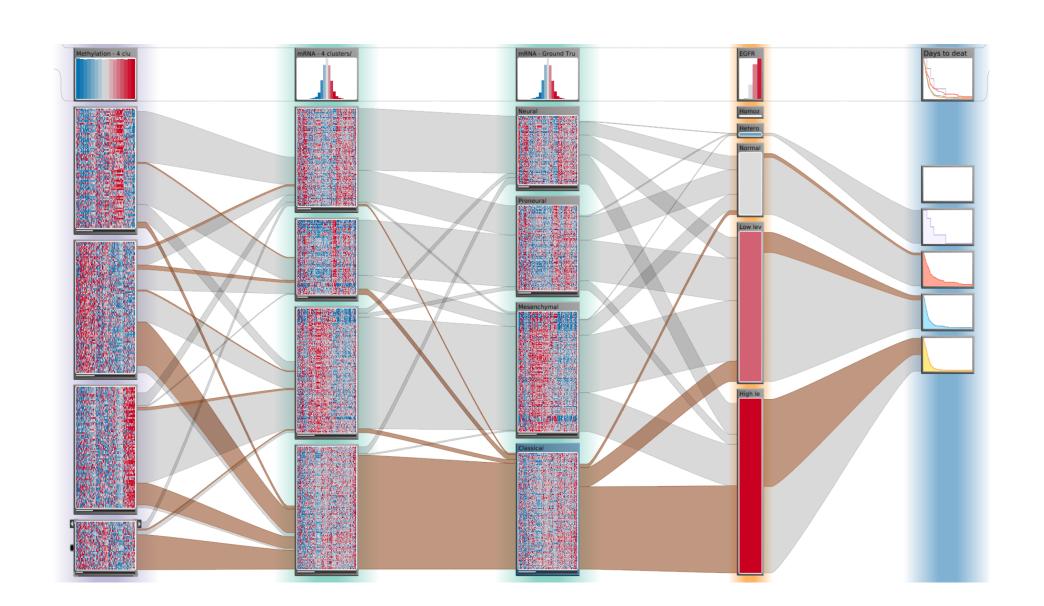
# When questions are not well defined: Exploration

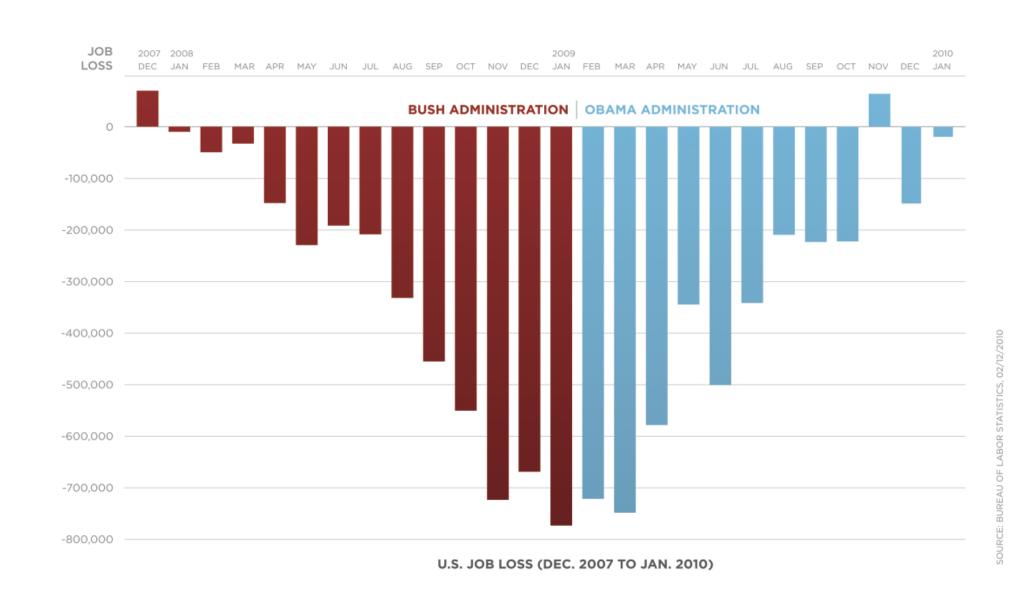
What is the structure of a terrorist network?

Which drug can help patient X?

### Purpose of Visualization

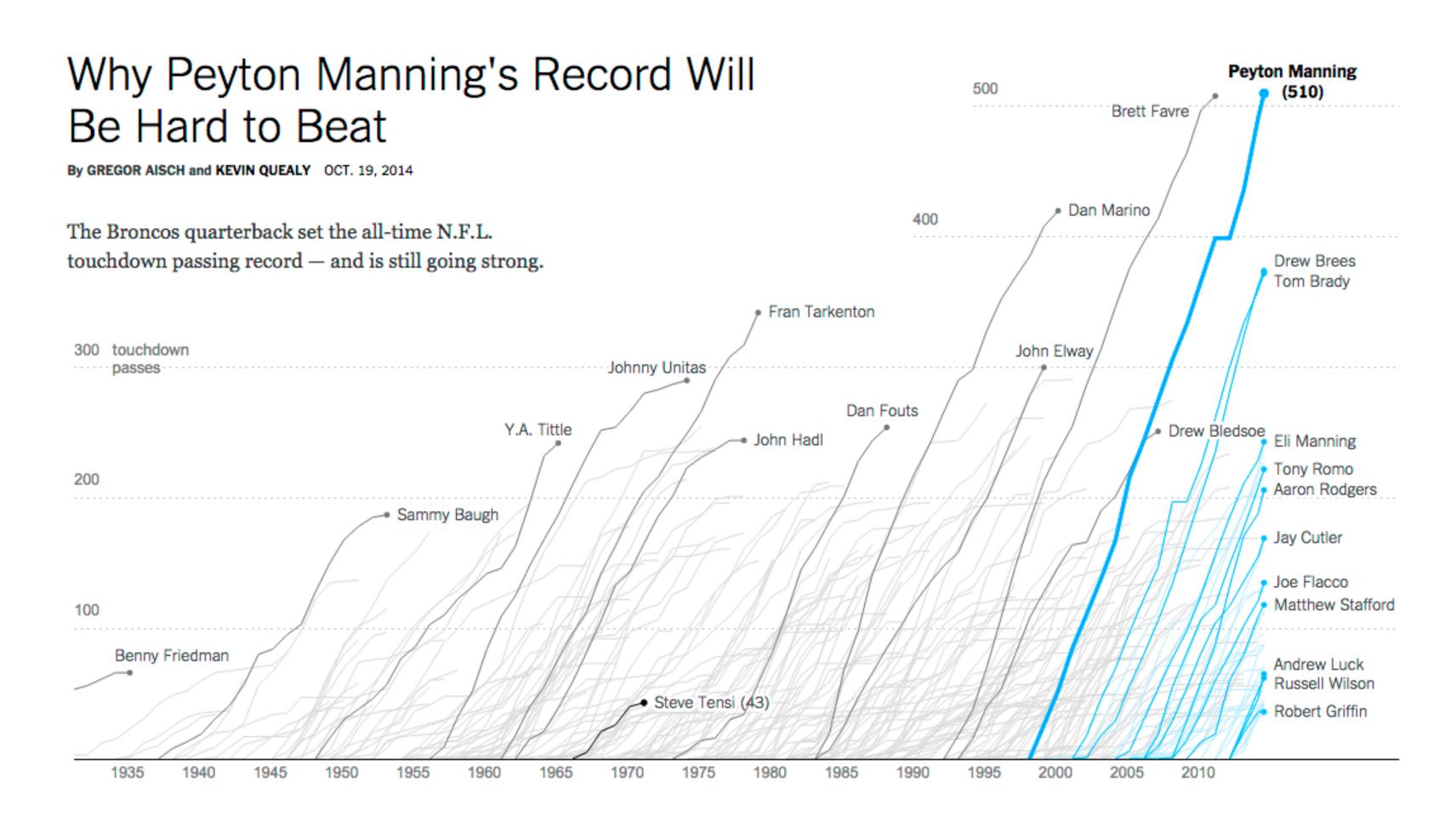
#### [Obama Administration]



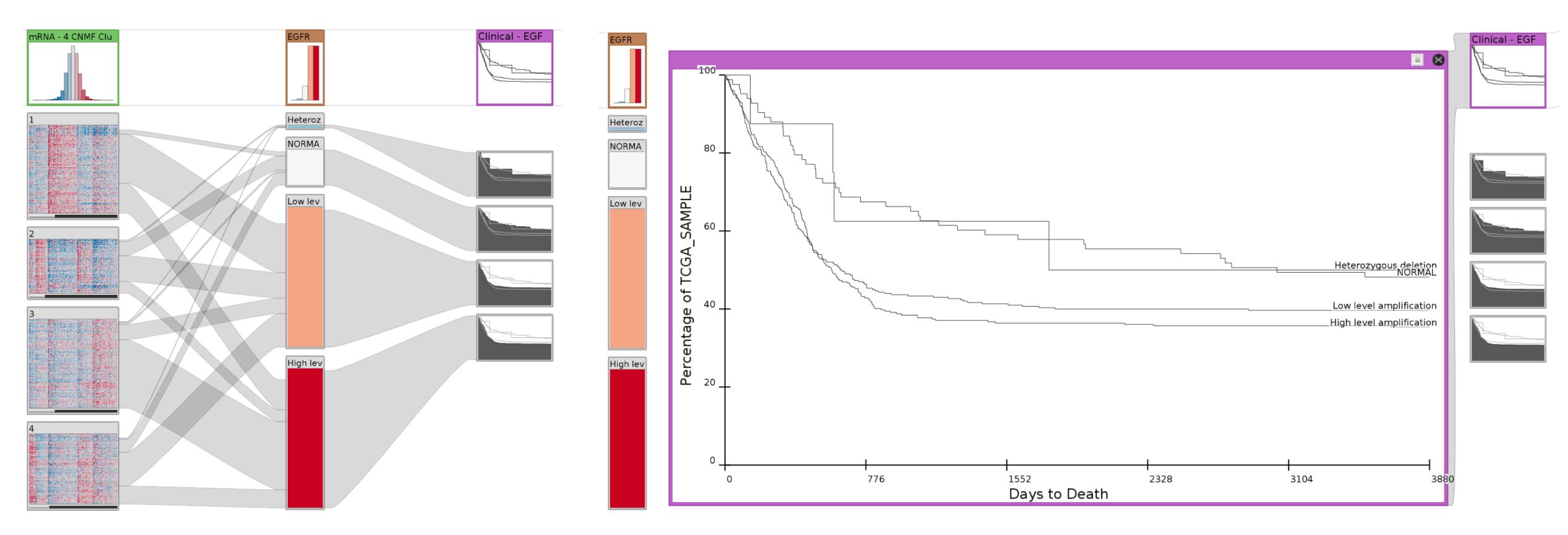


Open Exploration Confirmation Communication

### Example Communication



# **Example Exploration: Cancer Subtypes**



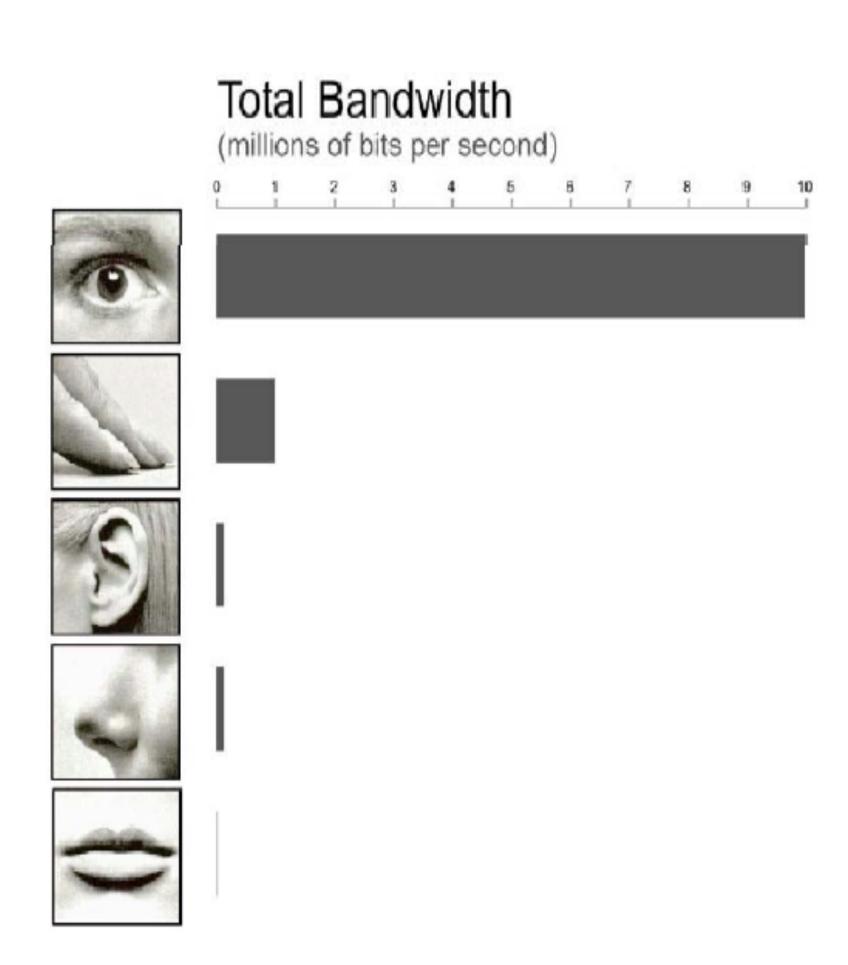
### Why Graphics?

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]



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

OV-

for

are

he

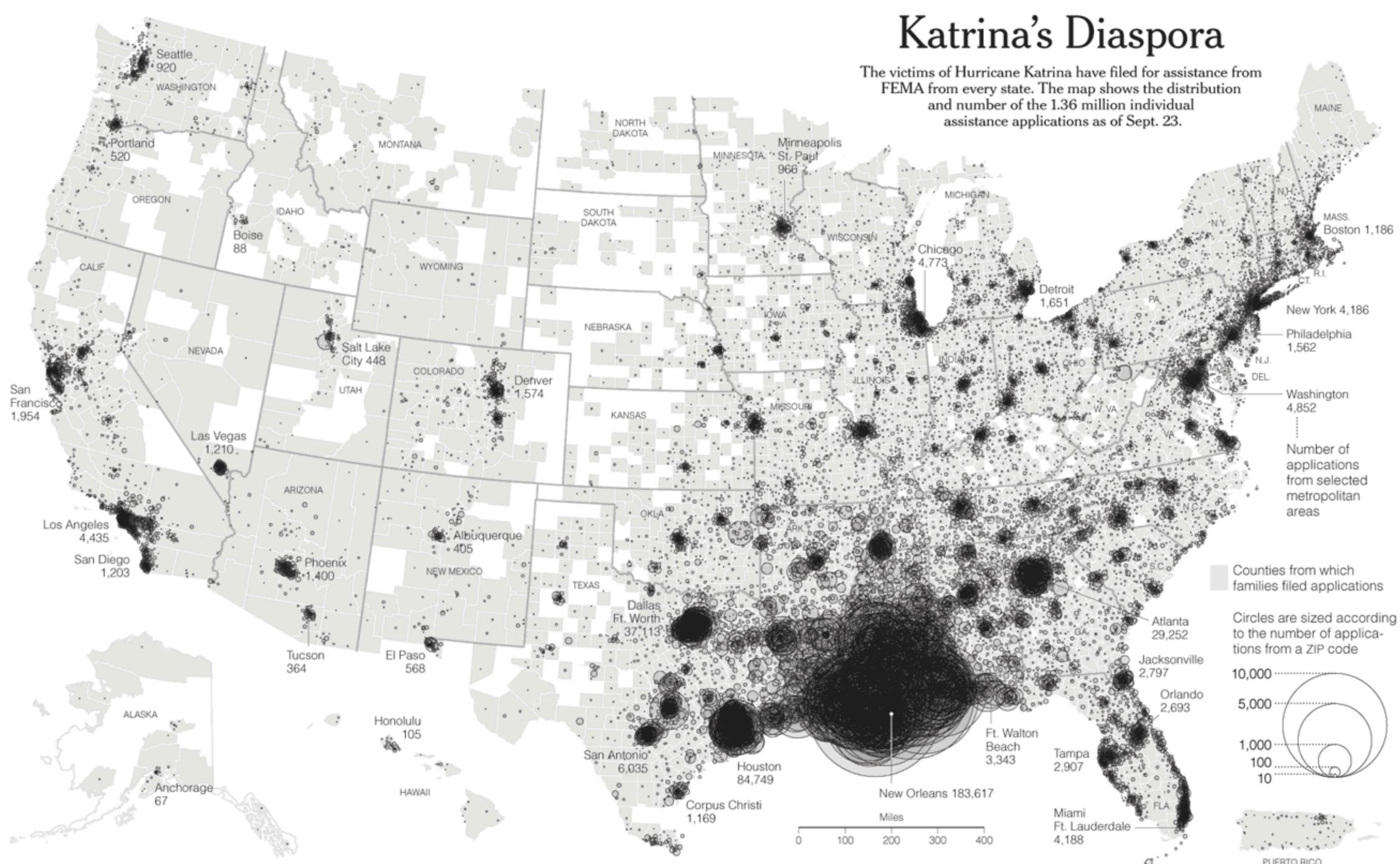
ne

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

Ctober 2, 2005



# 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?

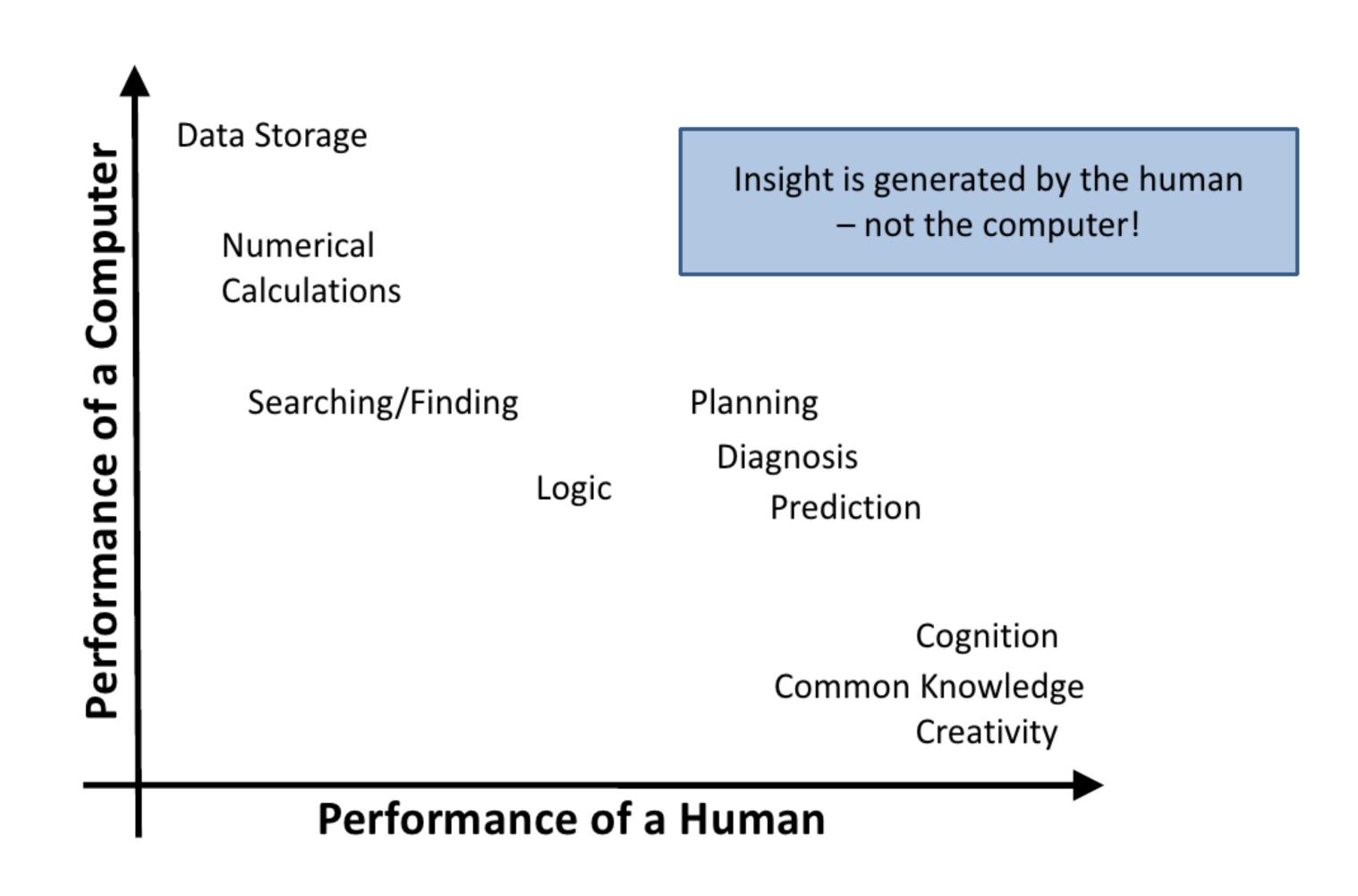
Manufacturing: is bottle broken?

Impractical for human to be involved

Automatic data products



### The Ability Matrix



### Why Use Computers?

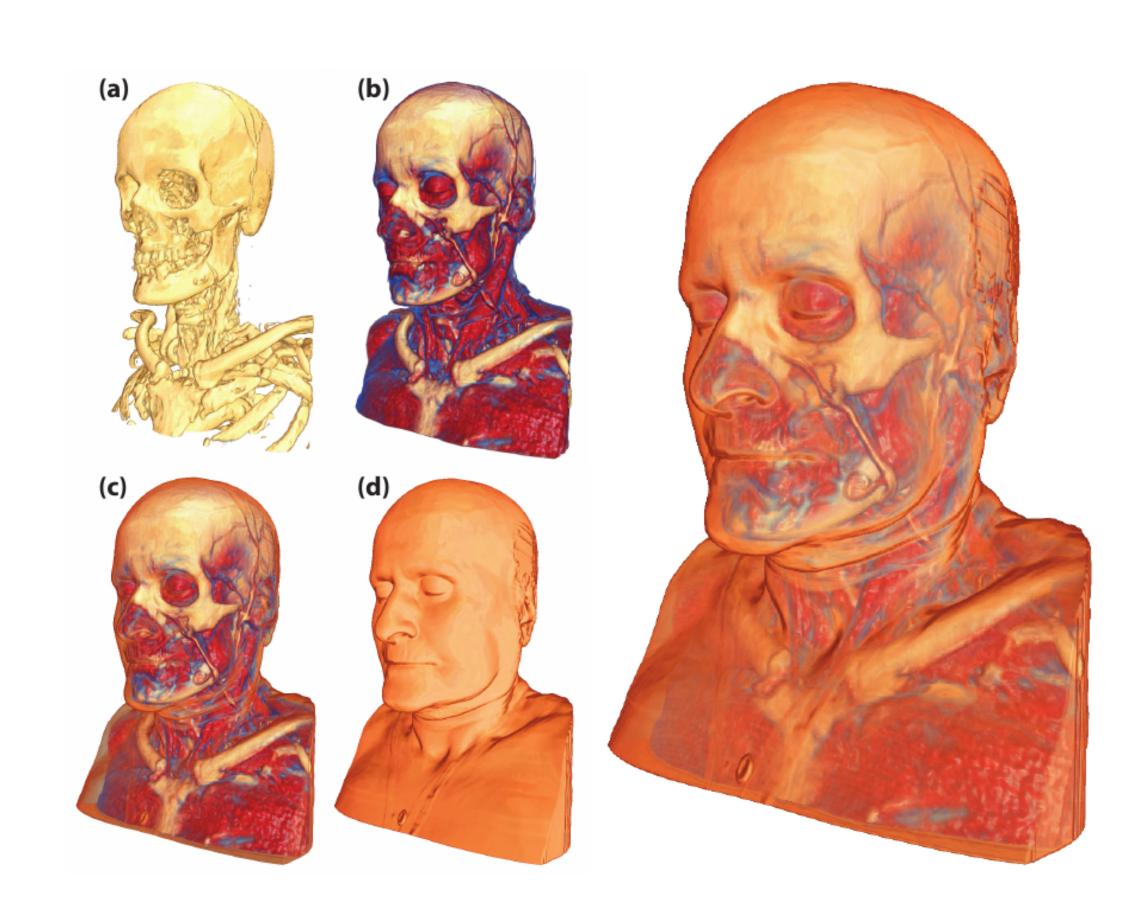
#### Scale

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



### Why User Computers?

#### Efficiency

Re-use charts / methods for different datasets

#### Quality

Precise data driven rendering

#### Storytelling

Use time

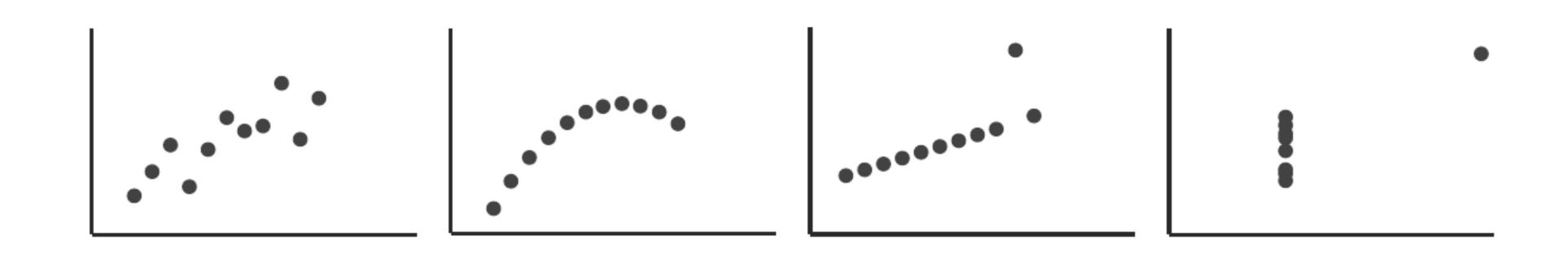
### Tell Stories



### Why not just use Statistics?

I	II	III	IV
X	X	<u> </u>	X
108.0	109.1	107.4	8 6.5
8 6.9	8 8.1	86.7	8 5.7
13 7.5	138.7	13 12.	8 7.7
98.8	98.7	97.1	8 8 8
11 8.3	11 9.2	11 7.8	8 8.4
14 9.9	148.1	148.8	87.0
67.2	66.1	66.0	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 . 2	777	7 6 1	9 7.9
5 5 Mean x: 9 y: 7.50			6.8
Variance x: 11 y: 4.122			
Correlation x - y: 0.816			
Line	ar regression	y = 3.00 + 0.5	UUX

### Anscombe's Quartett

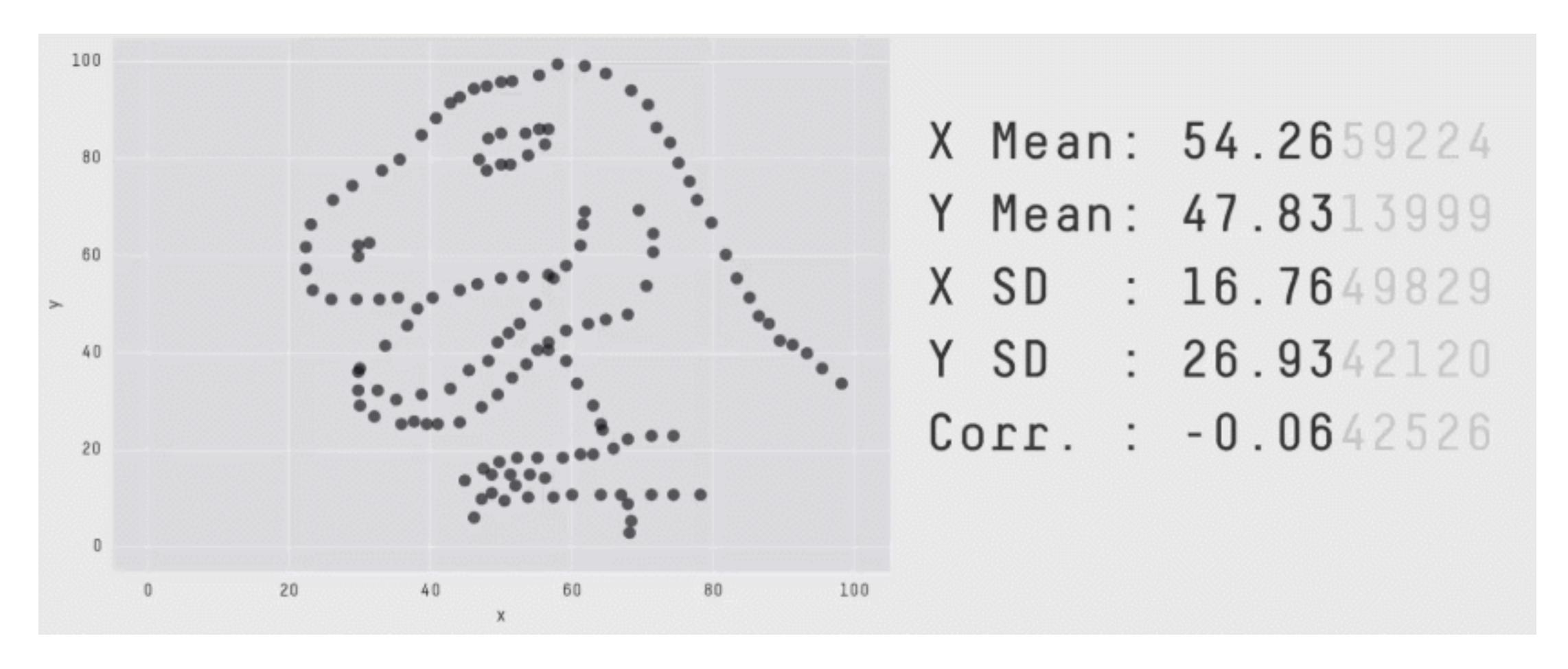


Mean x: 9 y: 7.50

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

#### Visualization in the Data Science Process

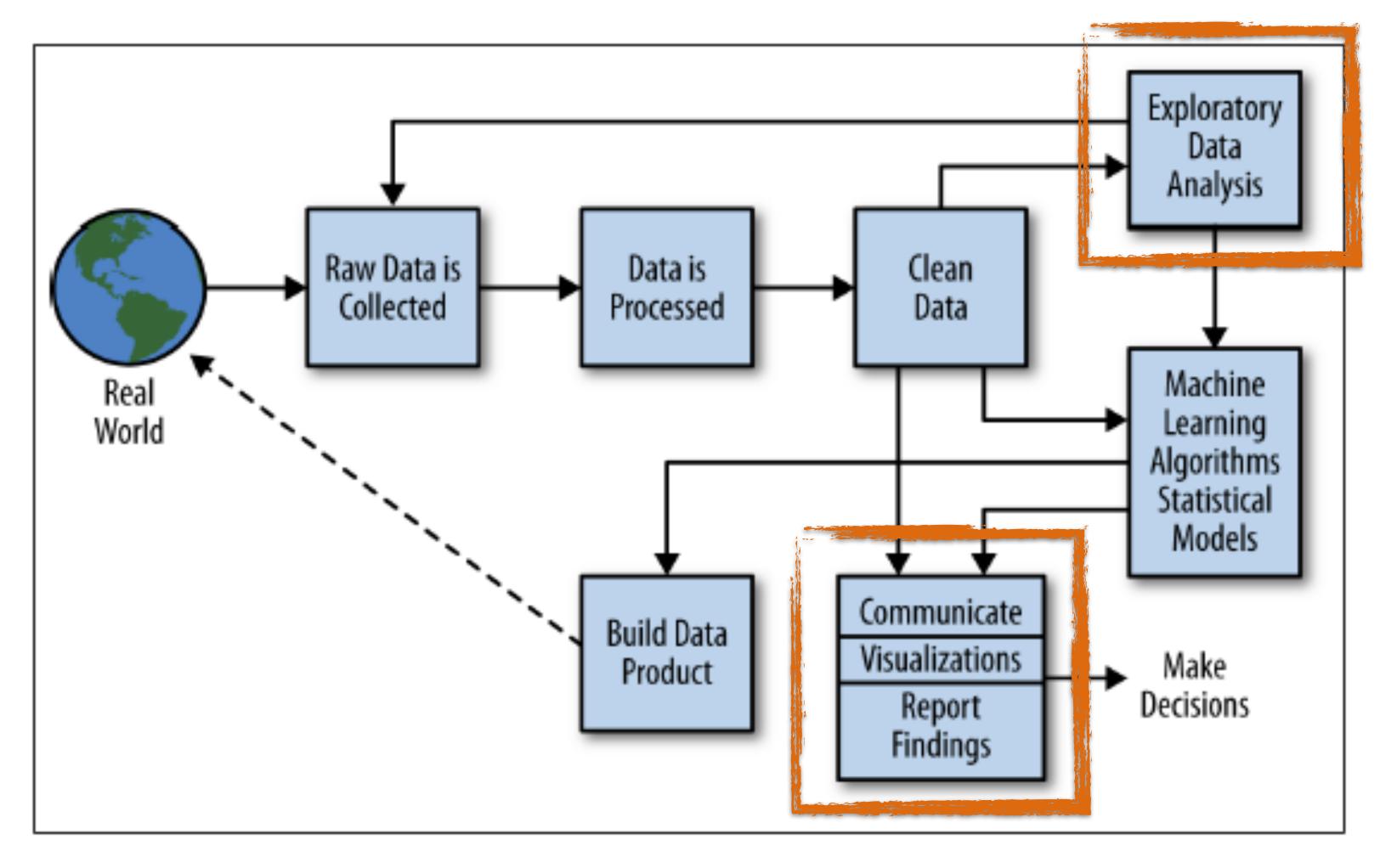


Figure 2-2. The data science process

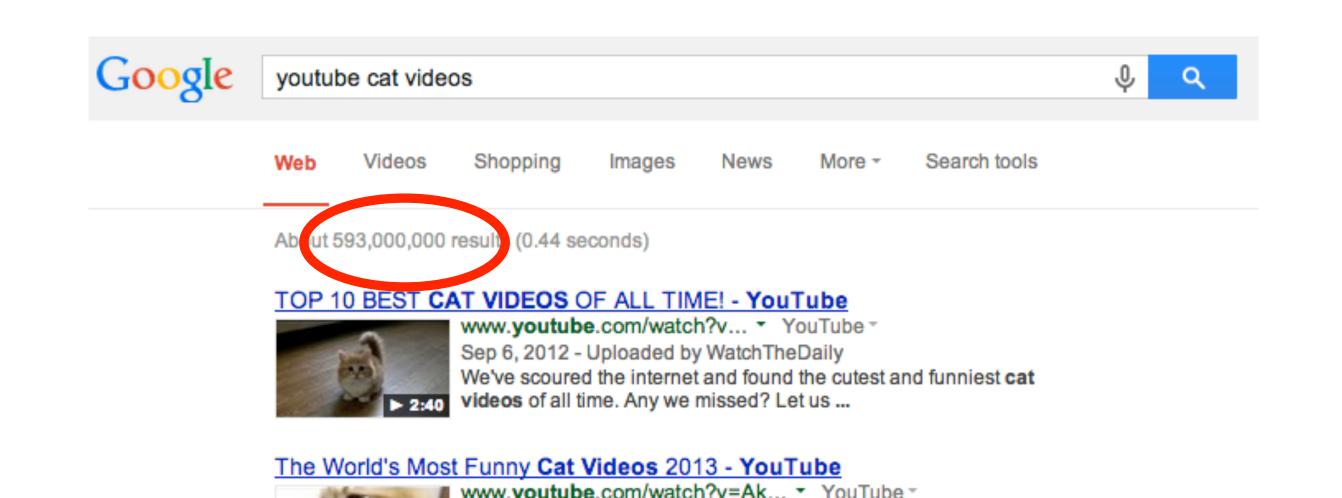
#### 15 Exabytes in Punch Cards:4.5 km over New England

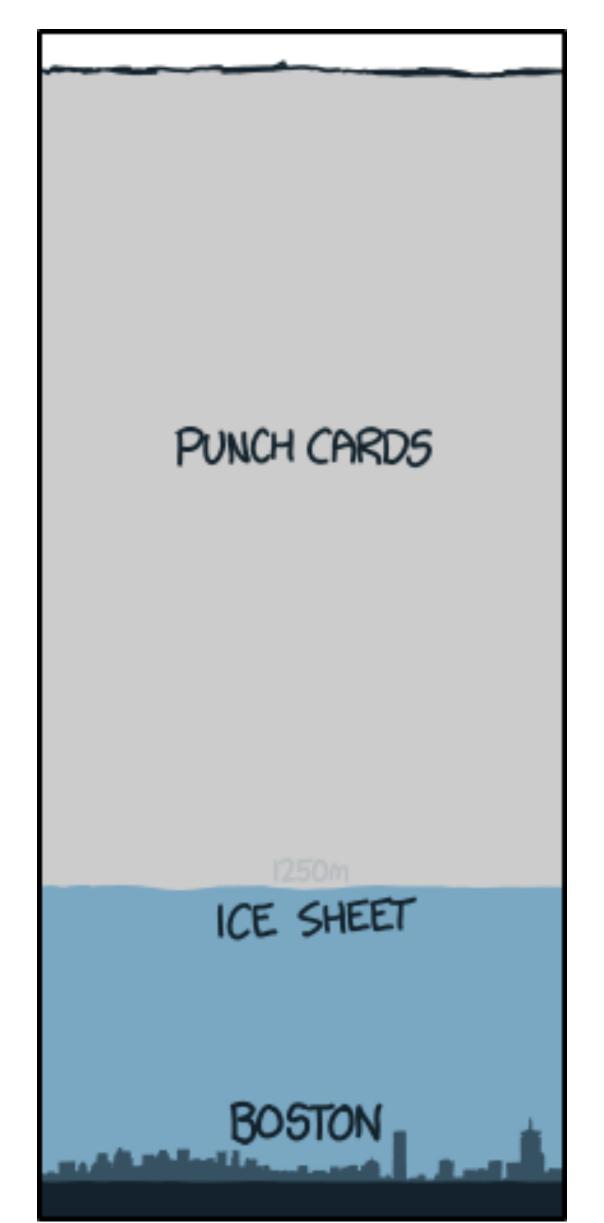
### Big Data

2017: 2.5 exabytes (quintillion bytes)

Source: IBM
of data per day, largely unstructured

90% of the data created in last two years





#### 798 Instagram photos uploaded in 1 second ( )

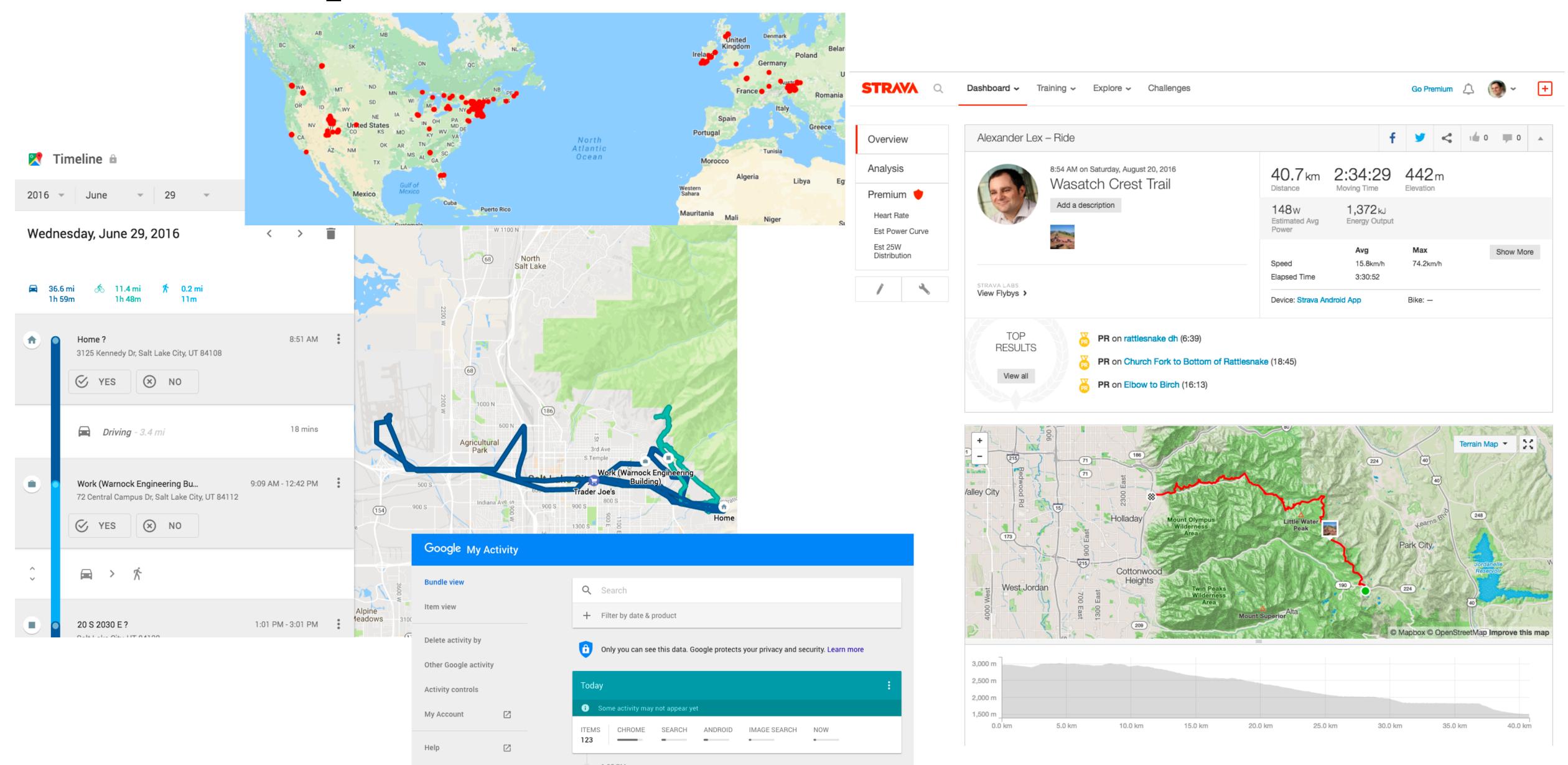


#### 1,277 Tumblr posts in 1 second



ttttttttttttttttttttttttttttt 

### Example: Personal Data



### Big Data in Science and Engineering

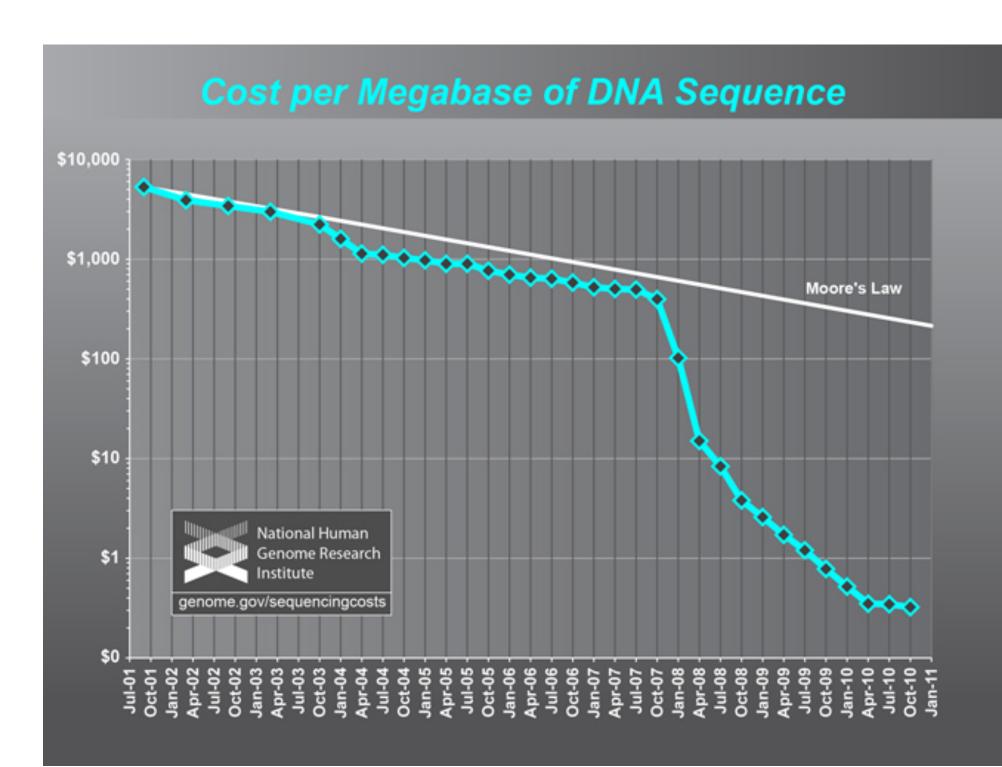
"Big Data" hasn't just transformed industry!

It's also transformed science and engineering. Cheap sensors (e.g. imaging) have changed the way science and engineering are done.

#### Examples:

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

Controversy: Hypothesis or data driven methods



#### **Example: CERN Large Hadron Collider Data**

CERN has publicly released over 300TB of data: CERN Open Data Portal

#### How much is that?

- A DVD-R holds 4.7 GB. You'd need 63,830 of them to hold 300 TB.
- It takes Pandora about a day and a half to burn through a gig of mobile data. So if the CERN data was an album, you could **stream it in just over 1,230 years**.
- At 350 MB per hour for 4K video streaming, so if the CERN data was a 4K movie it'd probably be about 857,142 hours, or about **98 years** long.
- But it ain't no thing compared to what the National Security Agency works with.
   Going by 2013 figures the agency released, the NSA's various activities "touch"
   300 TB of data every 15 minutes or so

(Popular Mechanics Article)

### Example: Genomics

**Eukaryotes** 

Archaea

Bacteria





**MinIO** 

Example TCGA: 1 Petabyte

#### NSA Utah Data Center (Bluffdale, Utah)

Storage Capacity?

estimates vary, but <u>Forbes</u> magazine estimates 12 exabytes (12,000 petabytes or 12 million terabytes)



"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

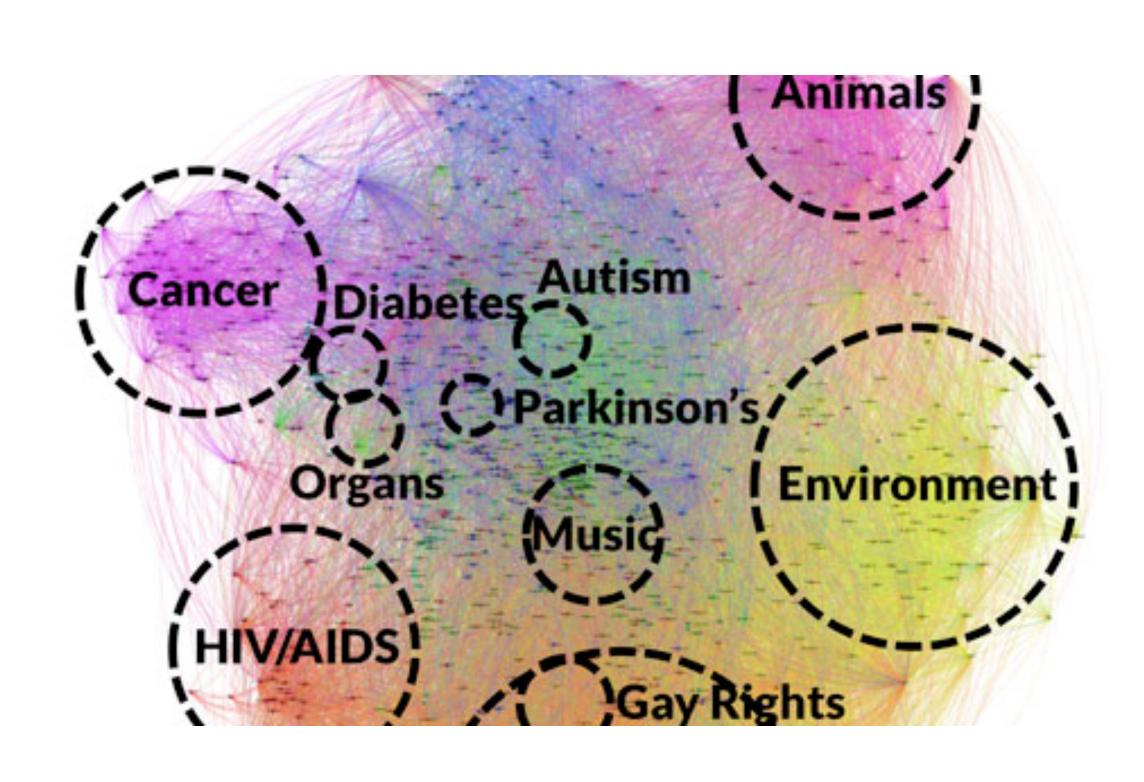
Pattern Discovery: clusters, outliers, trends

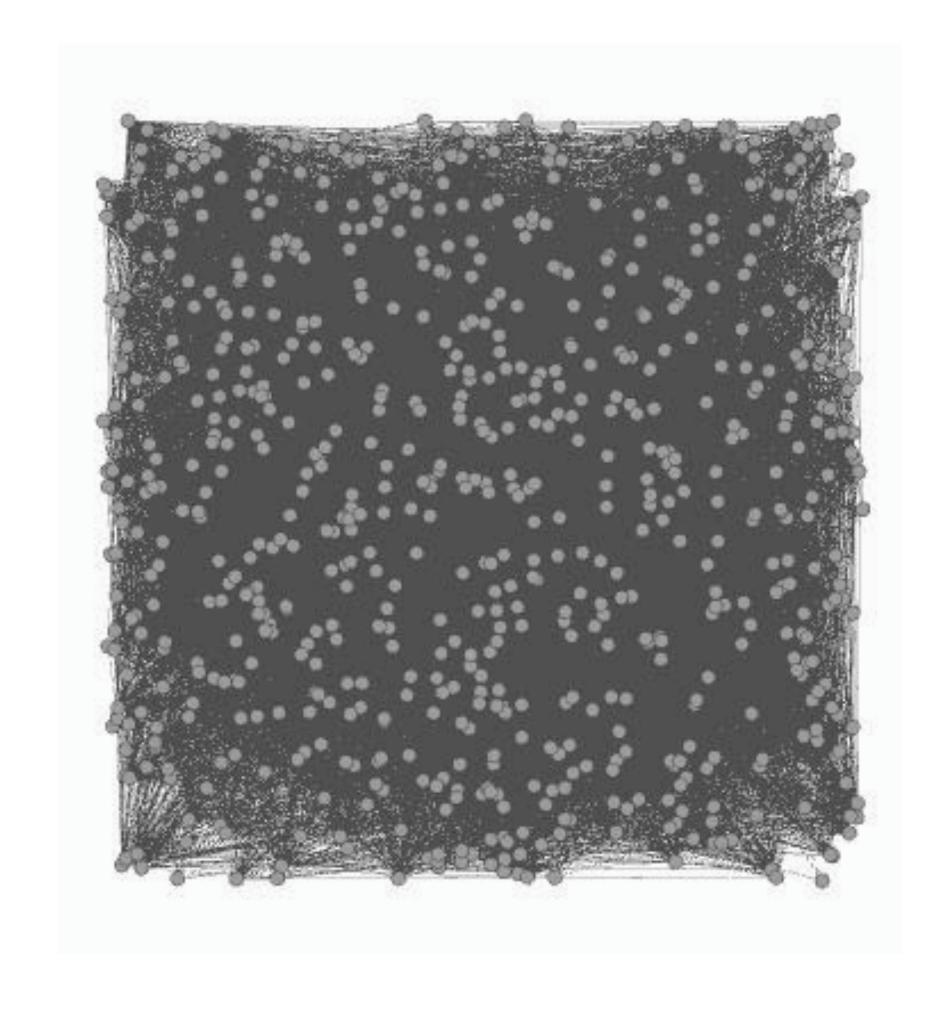
Contextual Knowledge: expectations for dataset, explanations for patterns

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

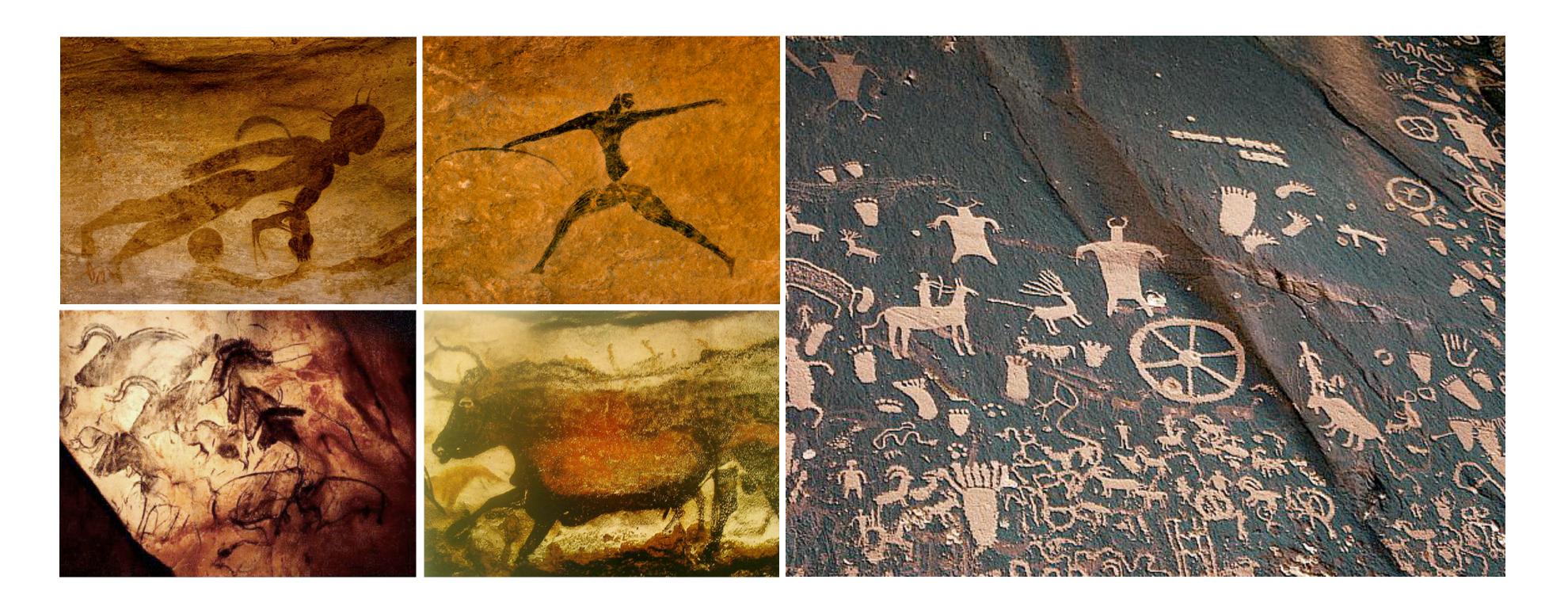


## How did we get here?

A bit of history

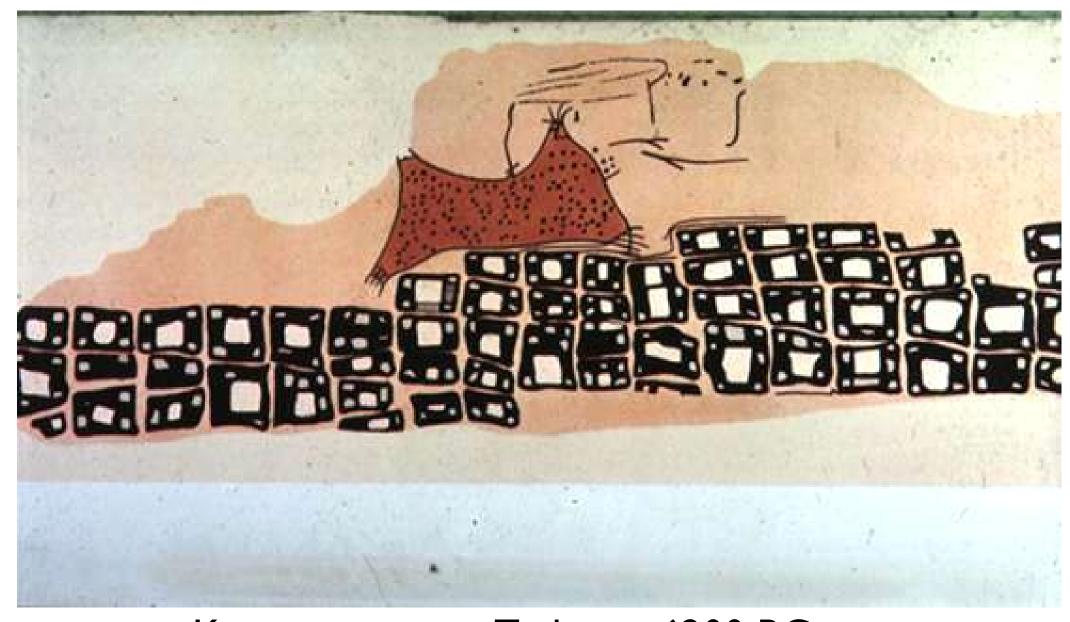
# "It is things that make us smart"

Donald A. Norman

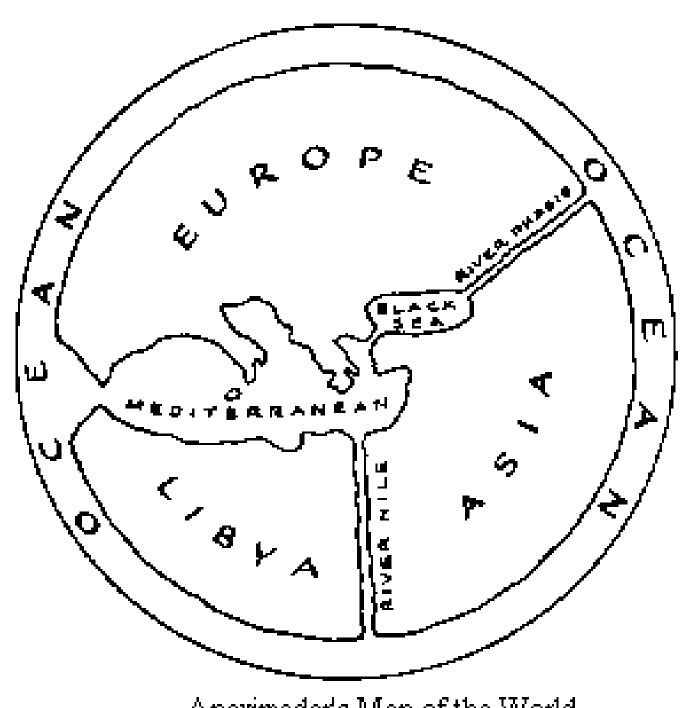




#### Record



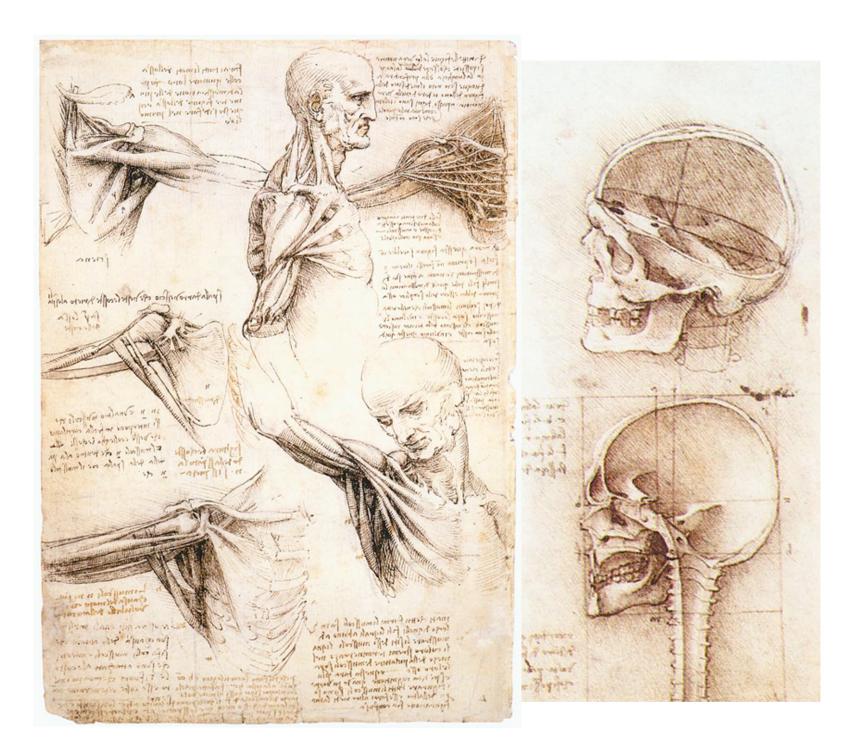
Konya town map, Turkey, c. 6200 BC



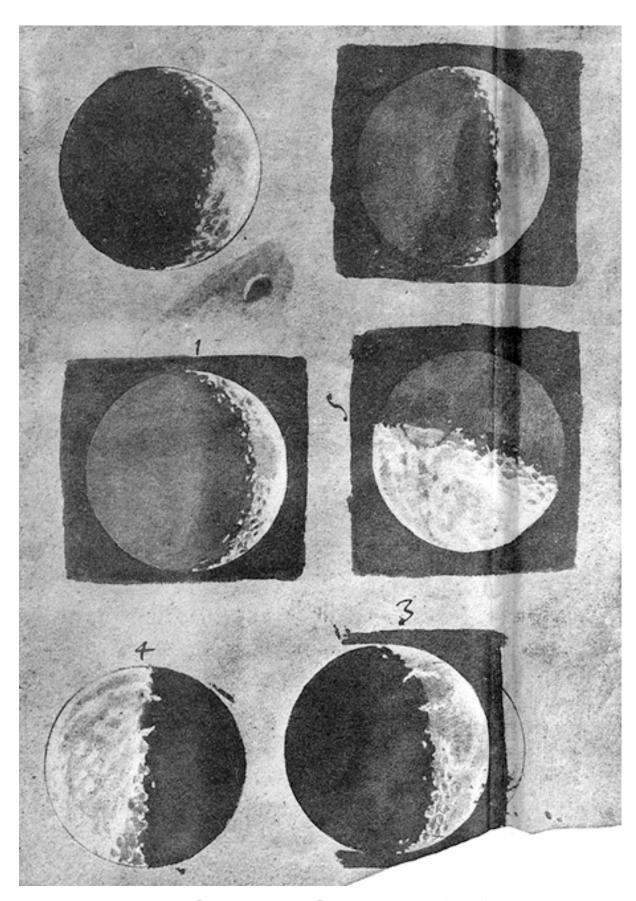
Anaximader's Map of the World

Anaximander of Miletus, c. 550 BC

#### Record



Leonardo Da Vinci, ca. 1500



Galileo Galilei, 1616

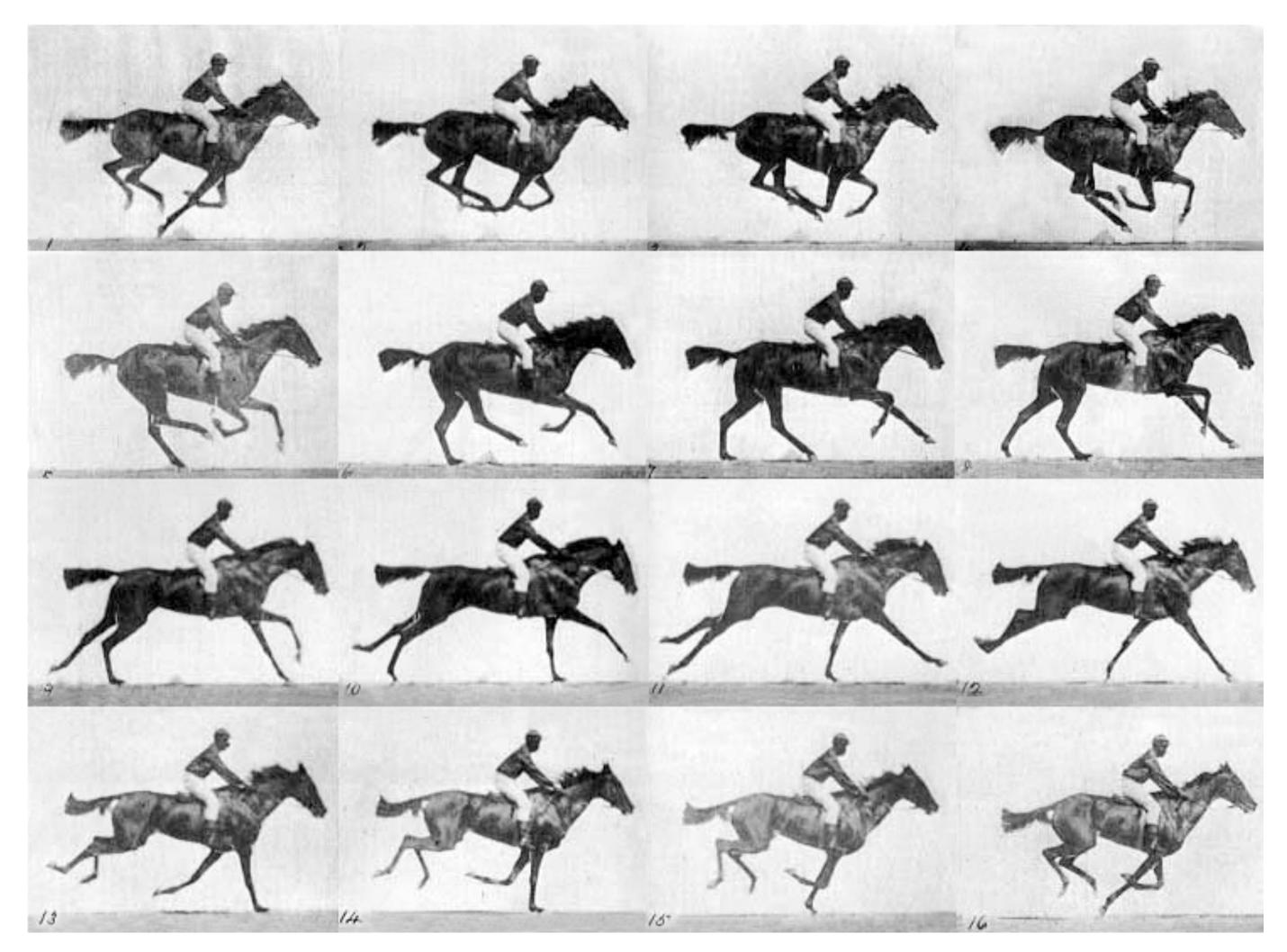
Donald Norman

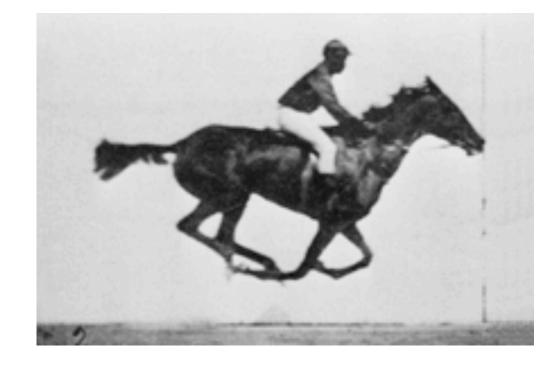


William Curtis (1746-1799)

The History of Visual Communication
The Galileo Project, Rice University

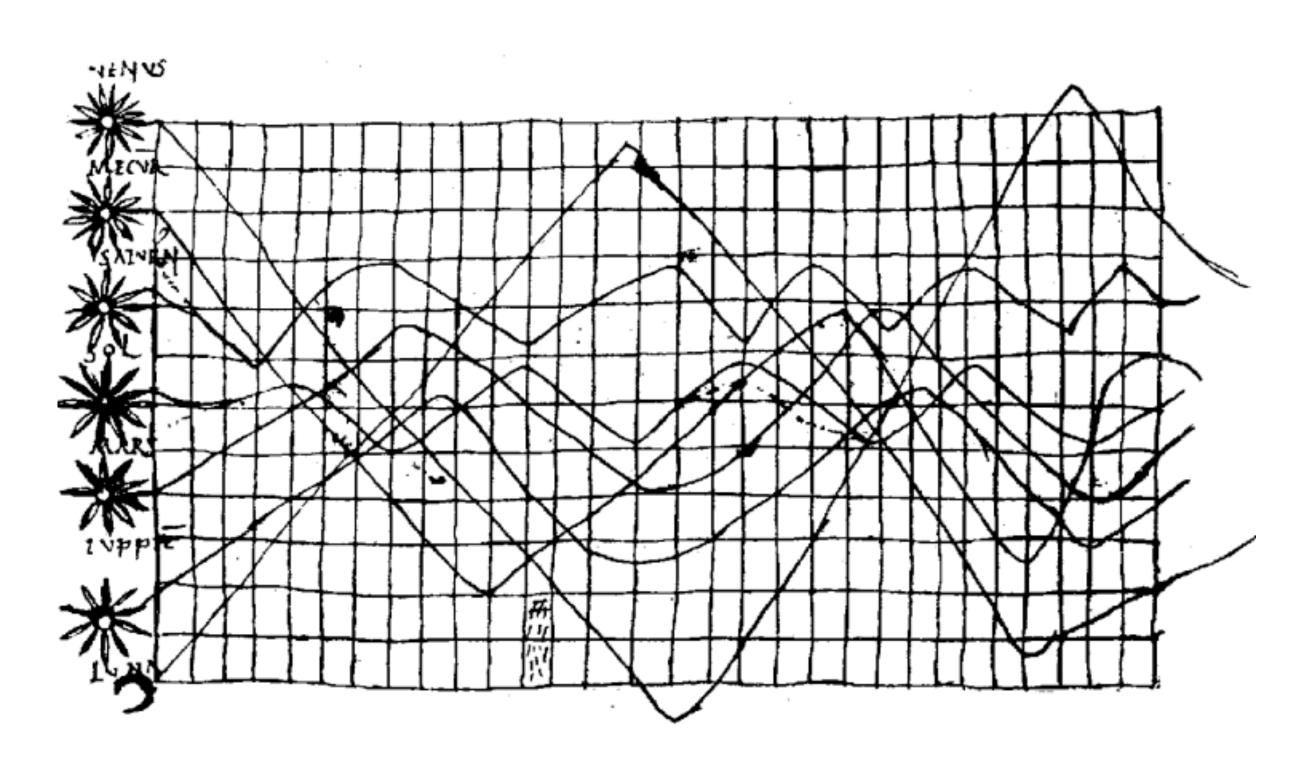
#### Record



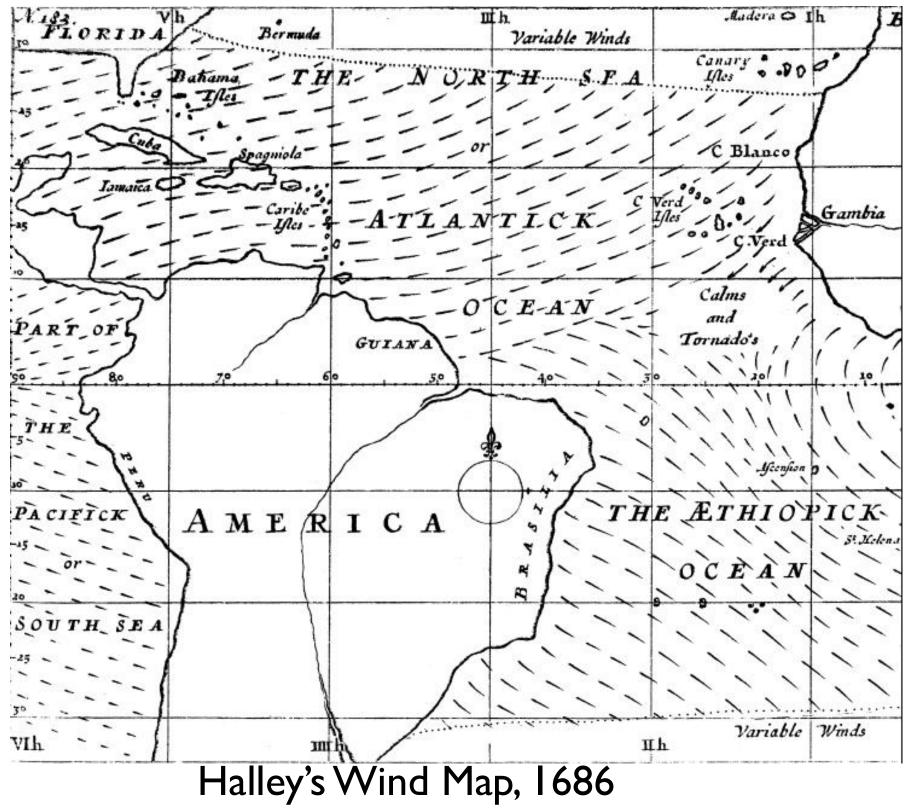


Eadweard J. Muybridge, 1878

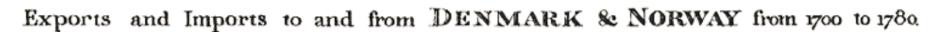
## Analyze

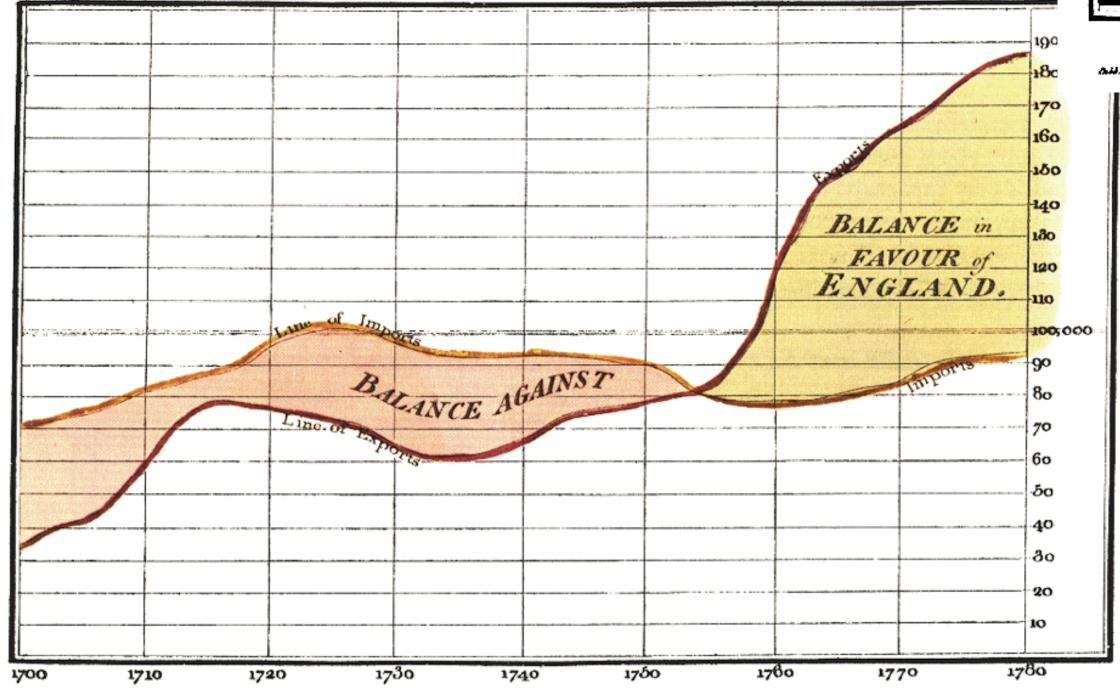




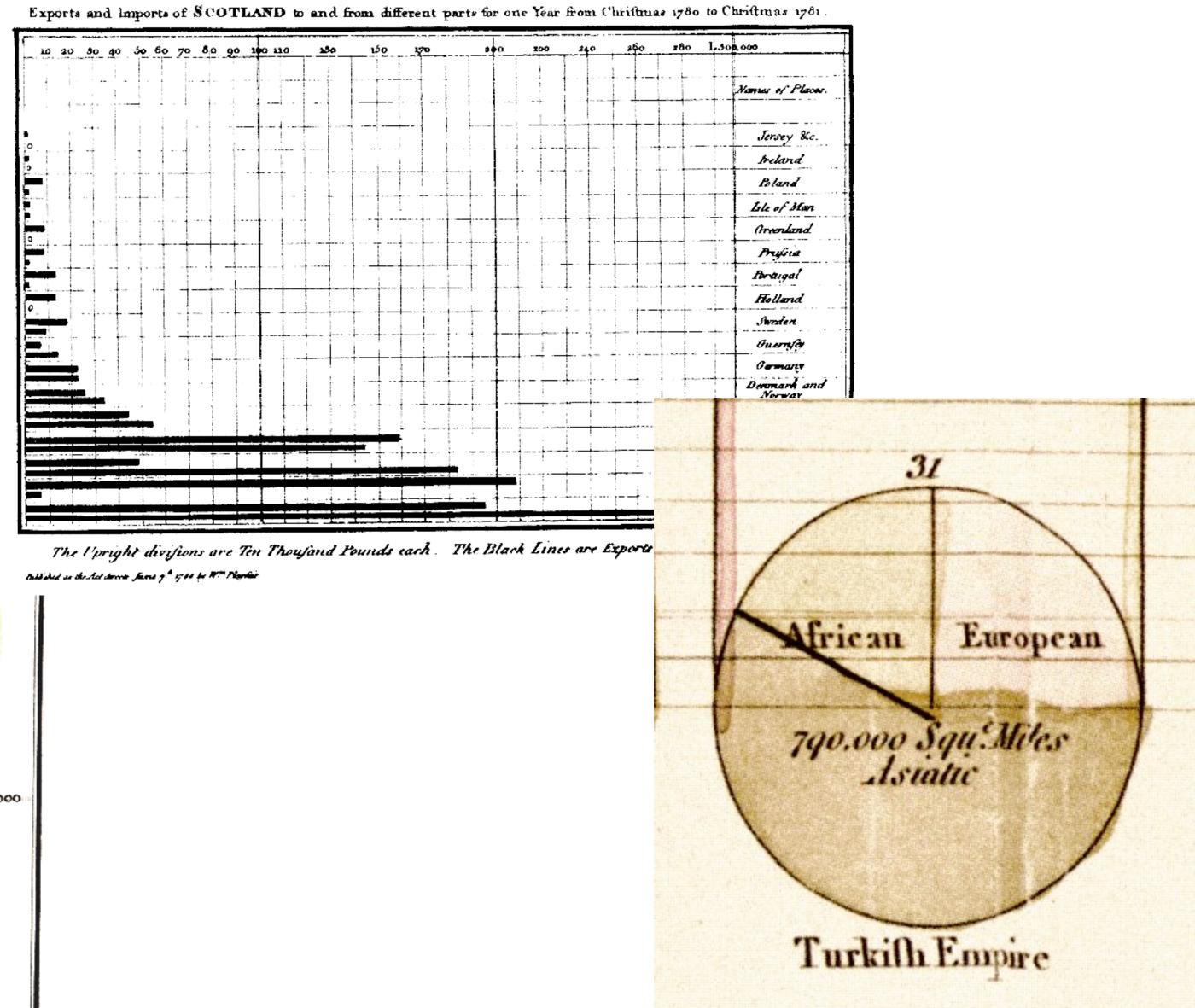


## Analyze



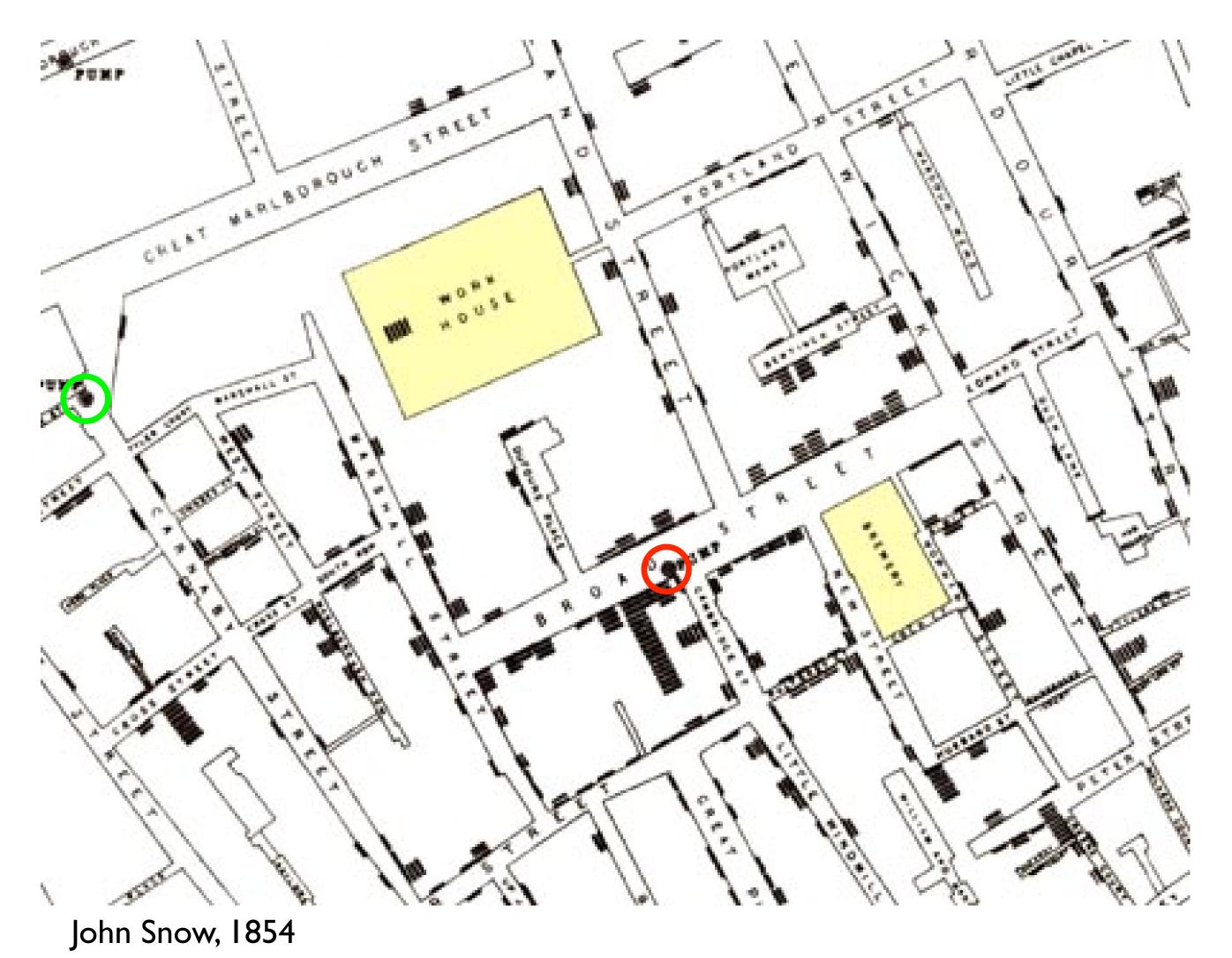


W. Playfair, 1786

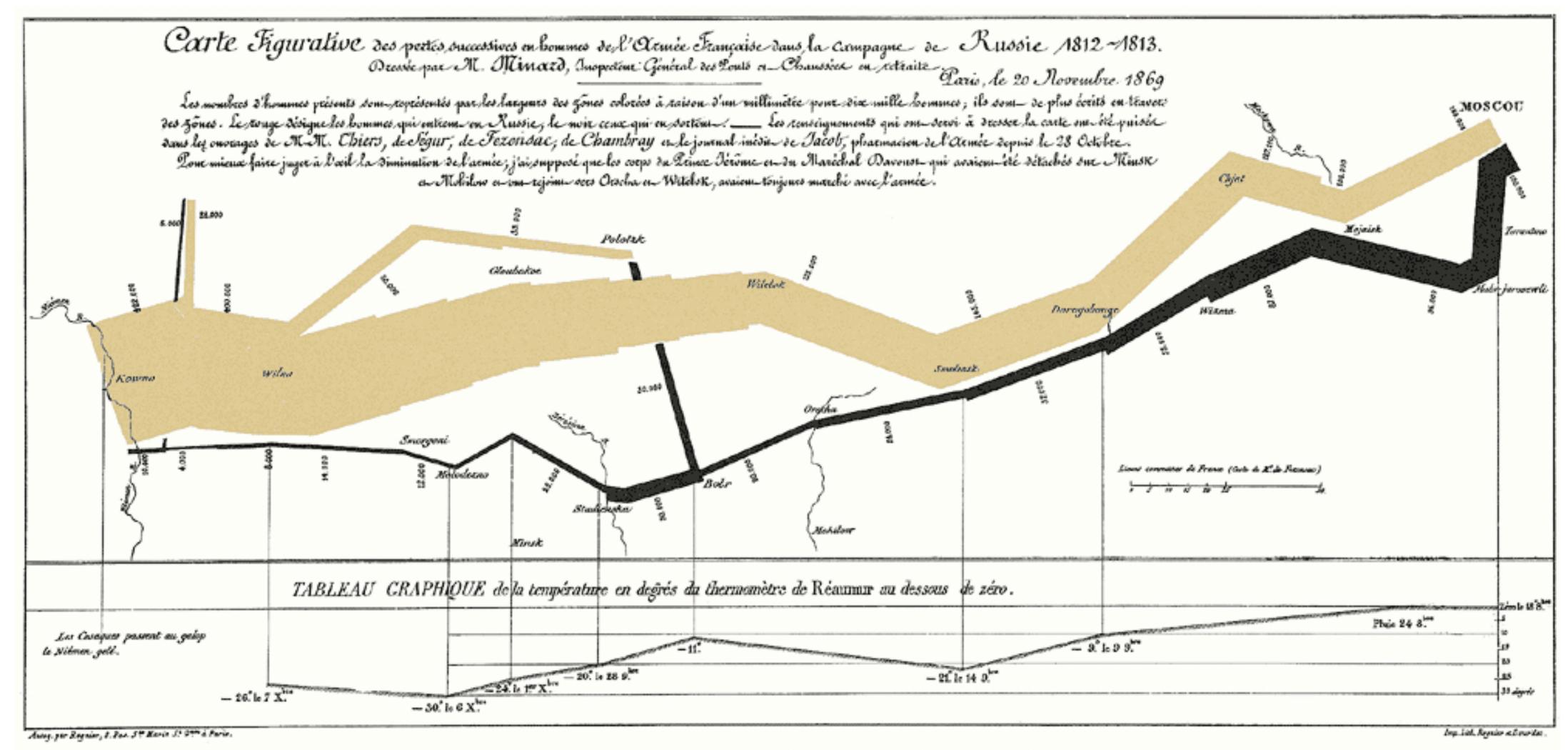


proportions of the Turkish Empire located in Asia, Europe and Africa before 1789

#### Find Patterns



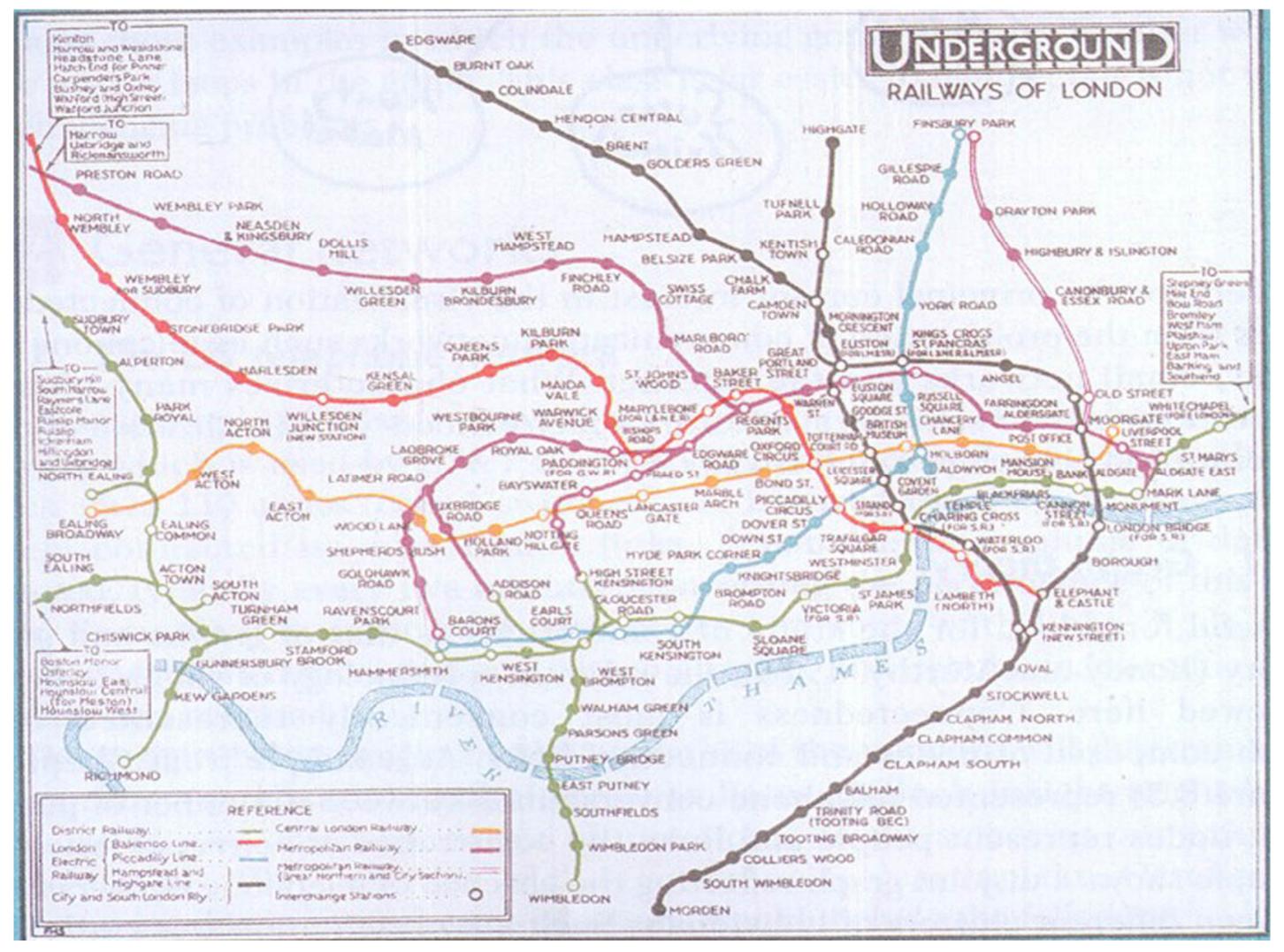
#### Communicate



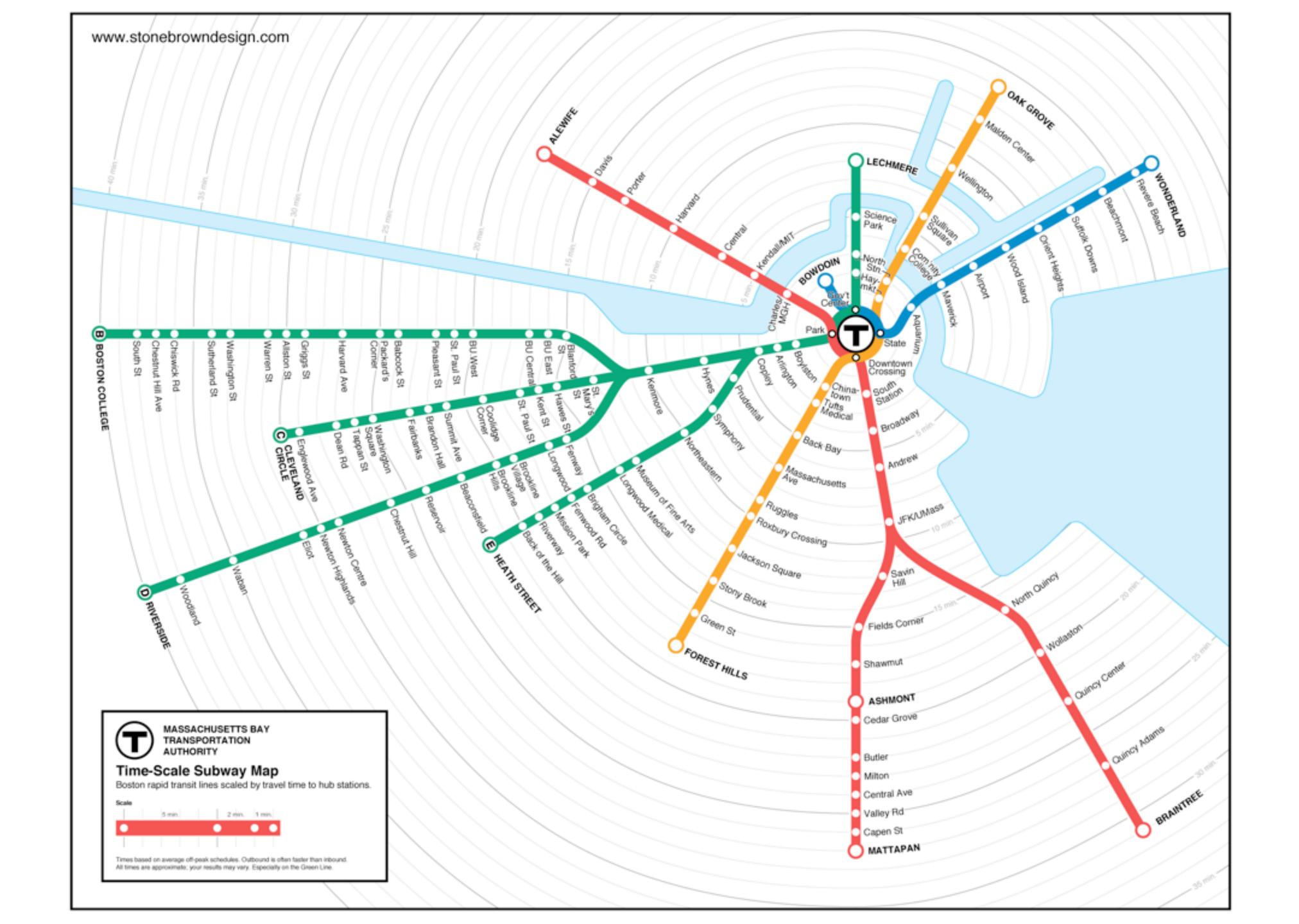


http://infowetrust.com/scroll/

#### Communicate



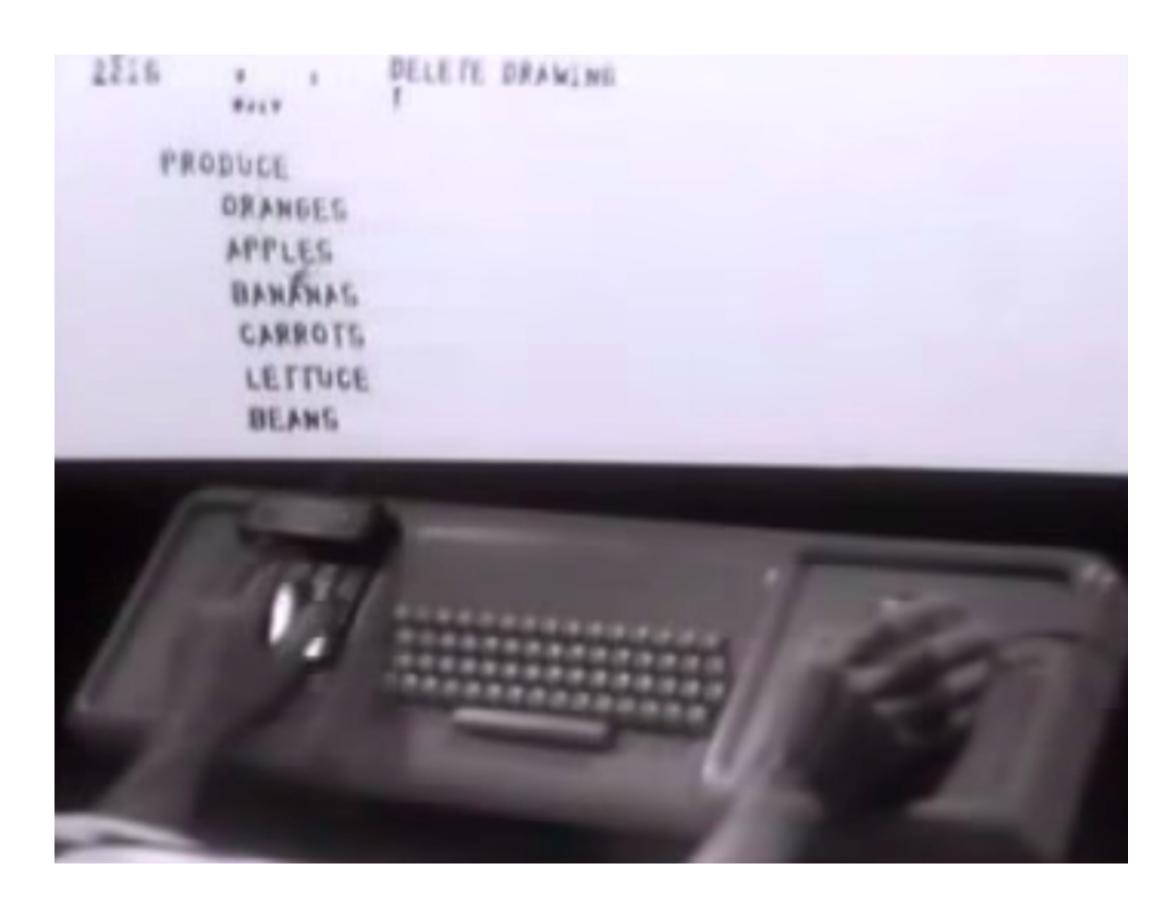
London Subway Map, 1927



#### Interact



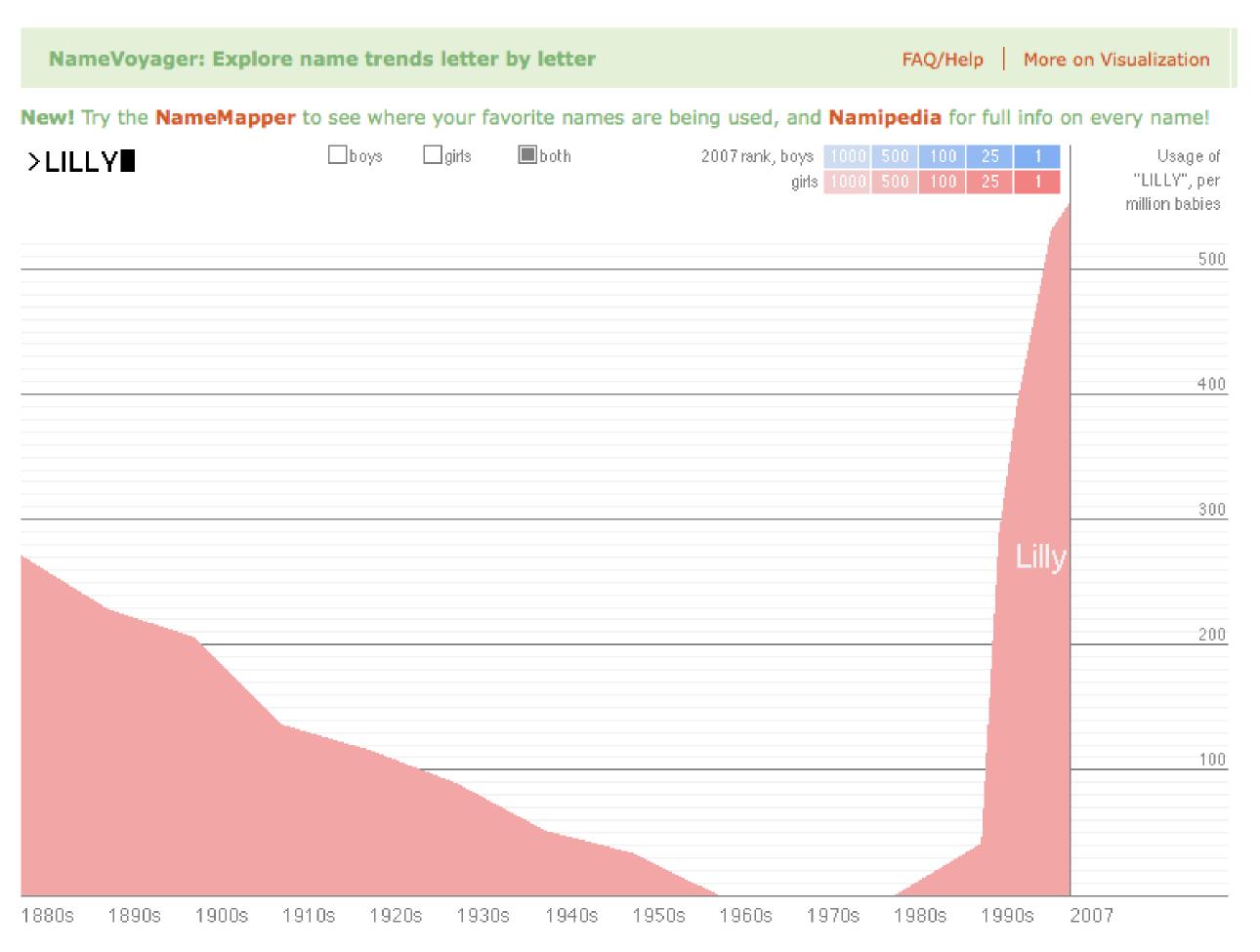
Ivan Sutherland, Sketchpad, 1963



Doug Engelbart, 1968

#### Modern Examples

## Analyze



M. Wattenberg, 2005

#### Communicate



Hans Rosling, TED 2006

#### Who is CS-5630 / CS-6630?

#### Course Staff



Jen Rogers
Teaching Mentee



Kiran Gadhave

Teaching Mentee



Ilkin Safarli
Teaching Mentee

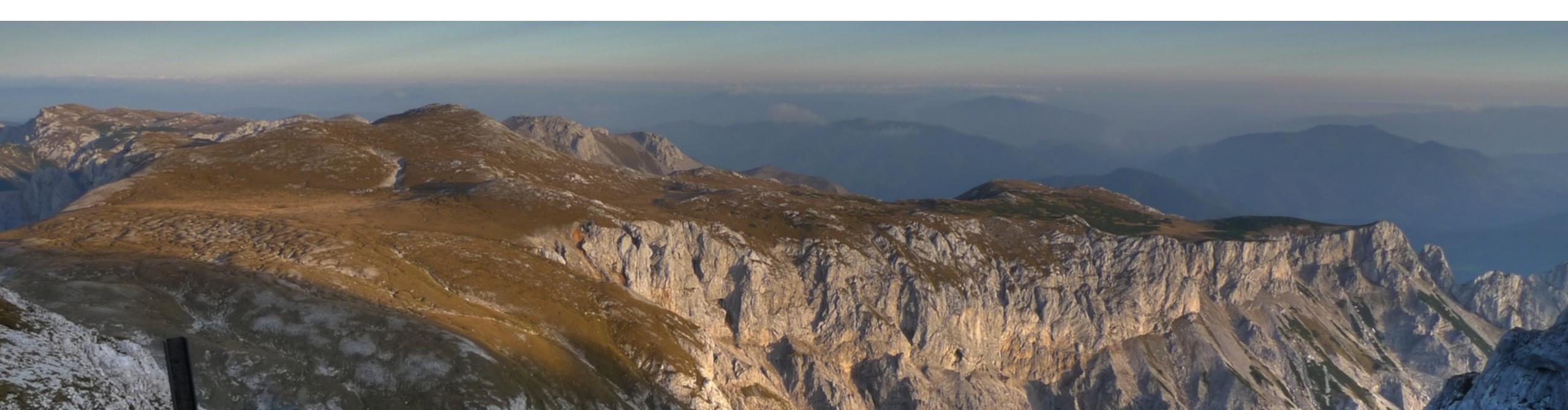
#### Alexander Lex



Assistant 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/





Miriah Meyer



Ilkin Safarli

Carolina Nobre

Sam Quinan

Jimmy Moore

Kiran Gadhave

Alexander Lex

Alex Bigelow

Pascal Goffin

Nina McCurdy

Ethan Kerzner

Jennifer Rogers

Haihan Lin

#### **SCI Institute**

Scientific Computing and Imaging Institute

Scientific Computing

Biomedical Computing

Scientific Visualization

Information Visualization

Image Analysis































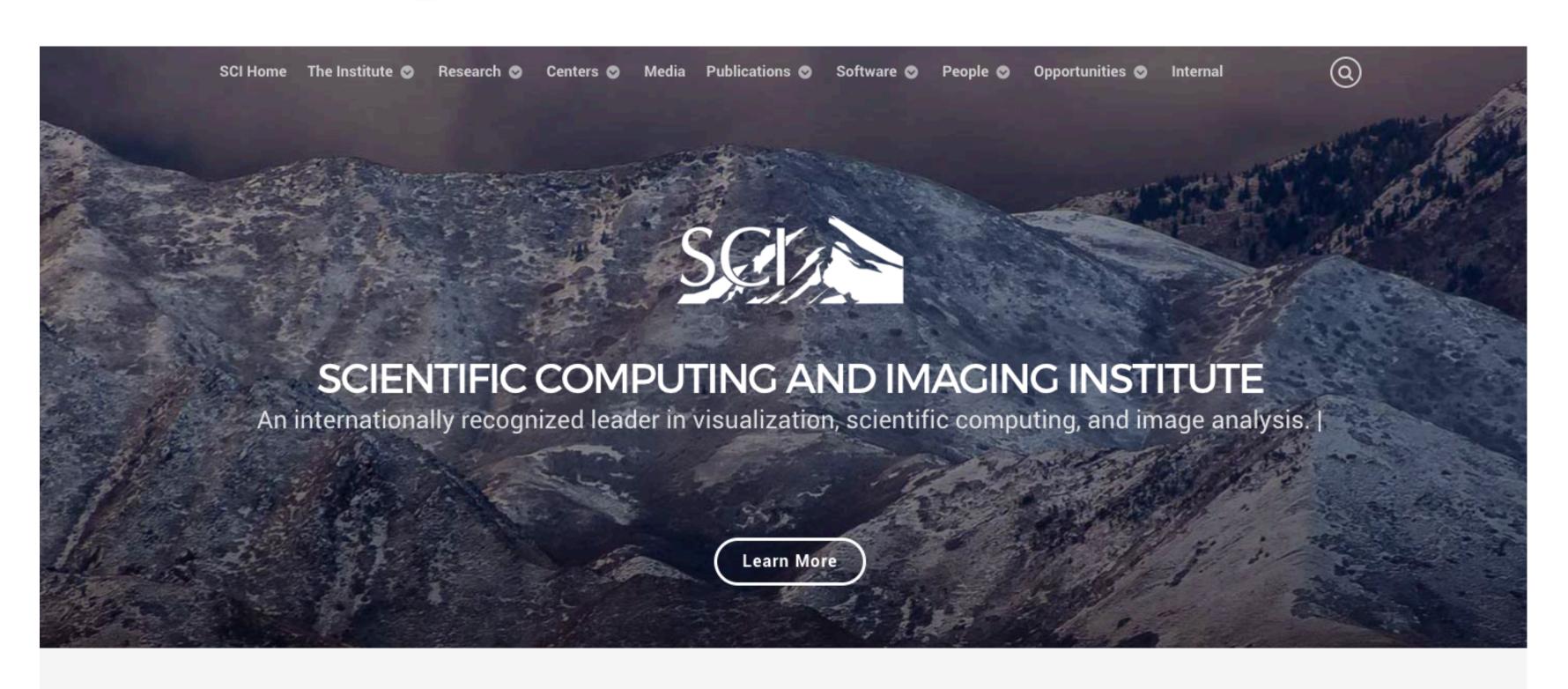






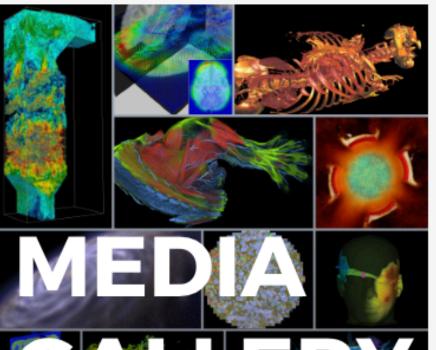


#### http://sci.utah.edu

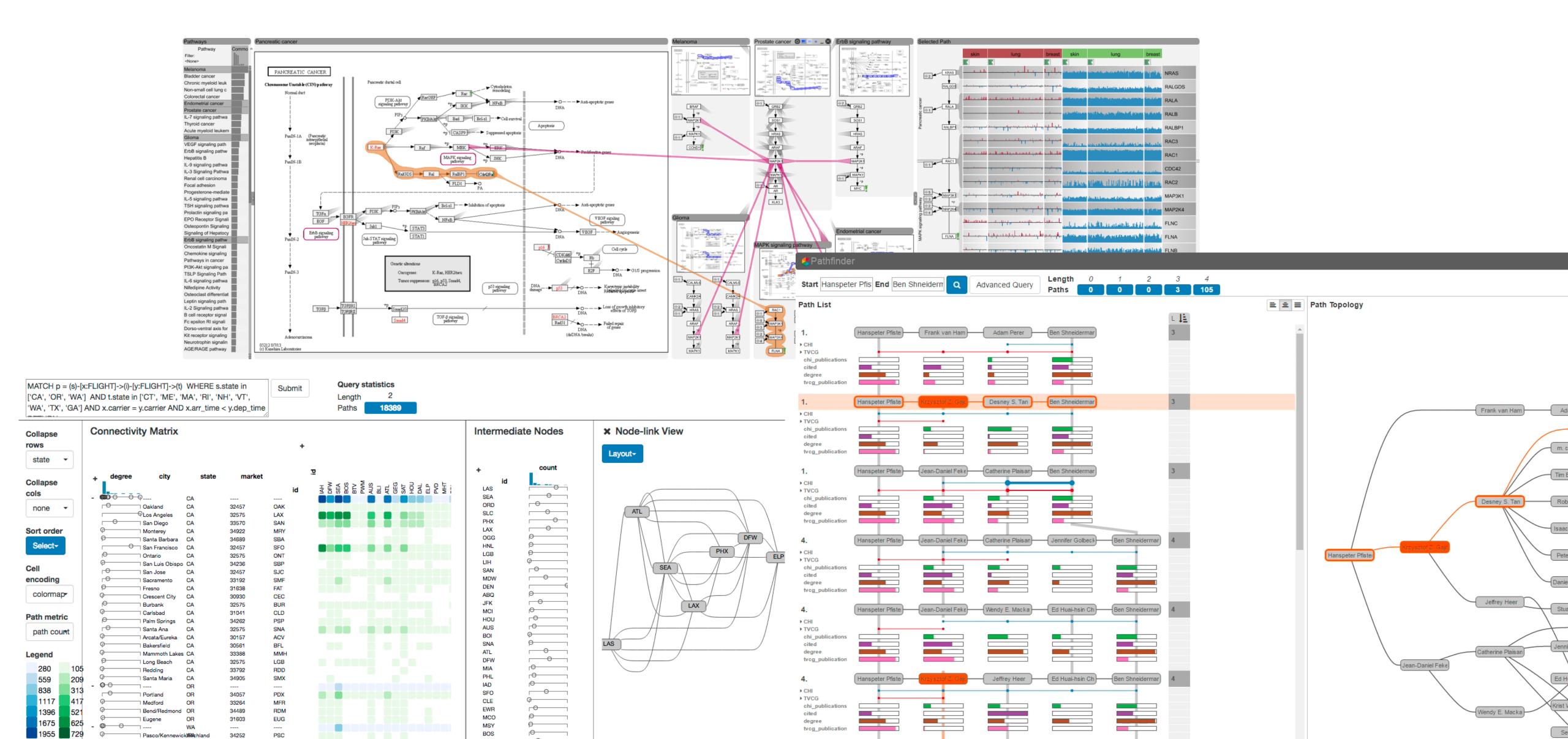




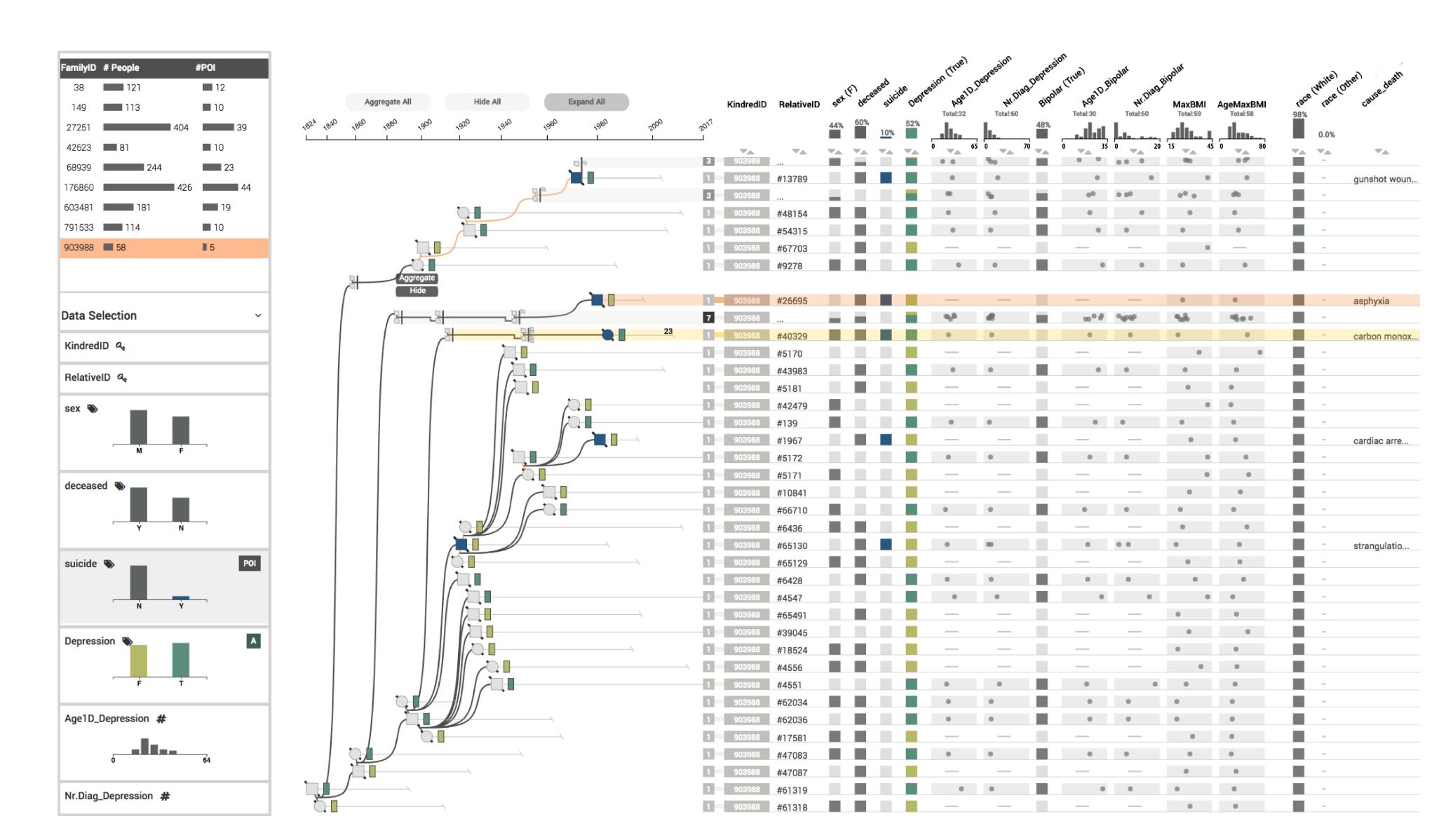




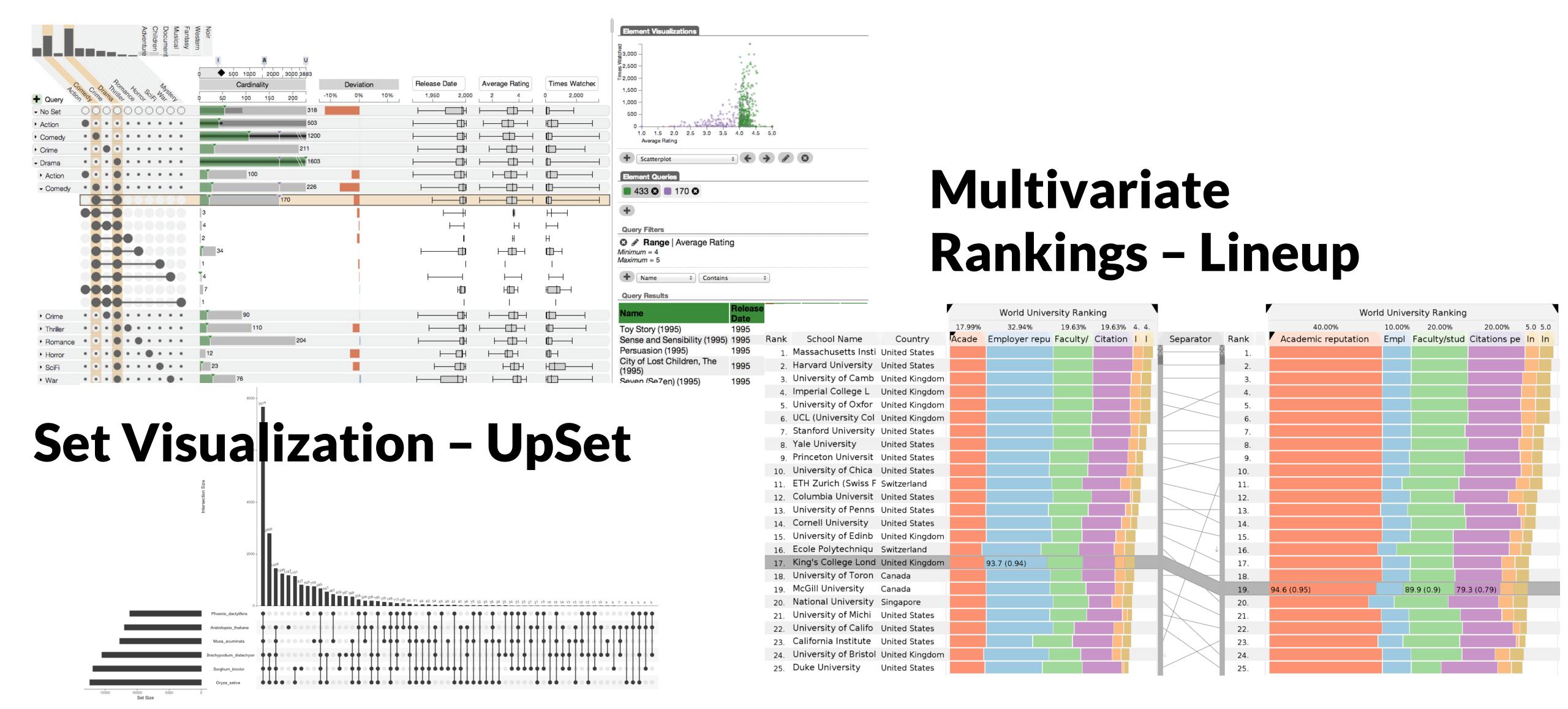
#### Large, Multivariate (Biological) Networks

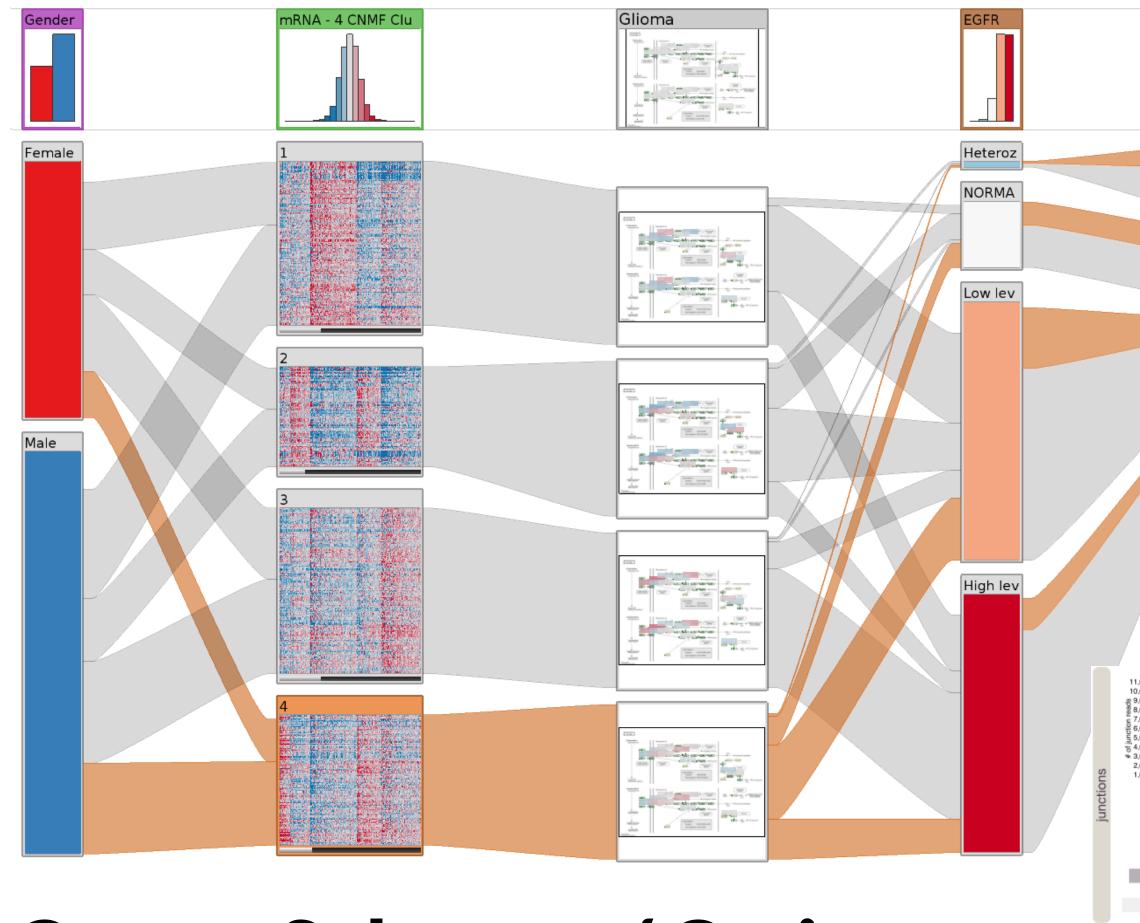


#### Genealogies & Clinical Data



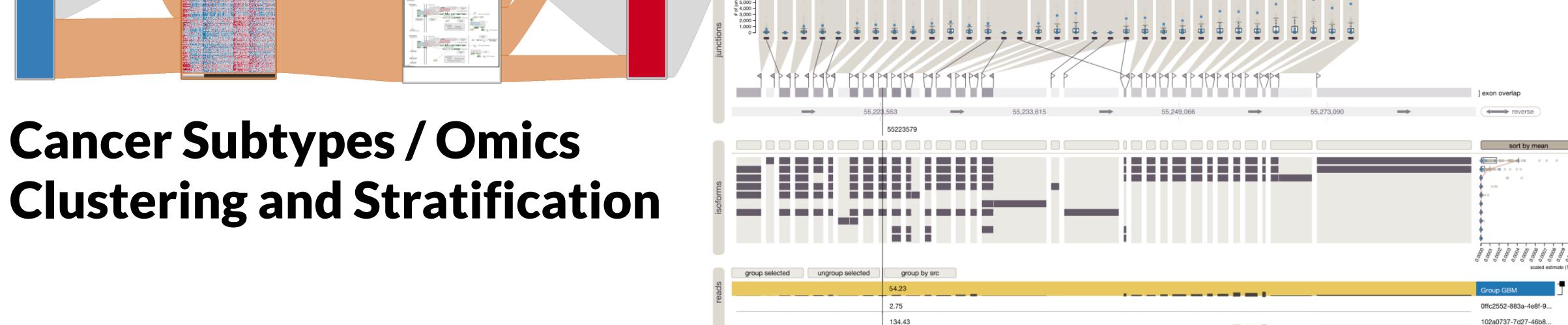
#### Multidimensional Data



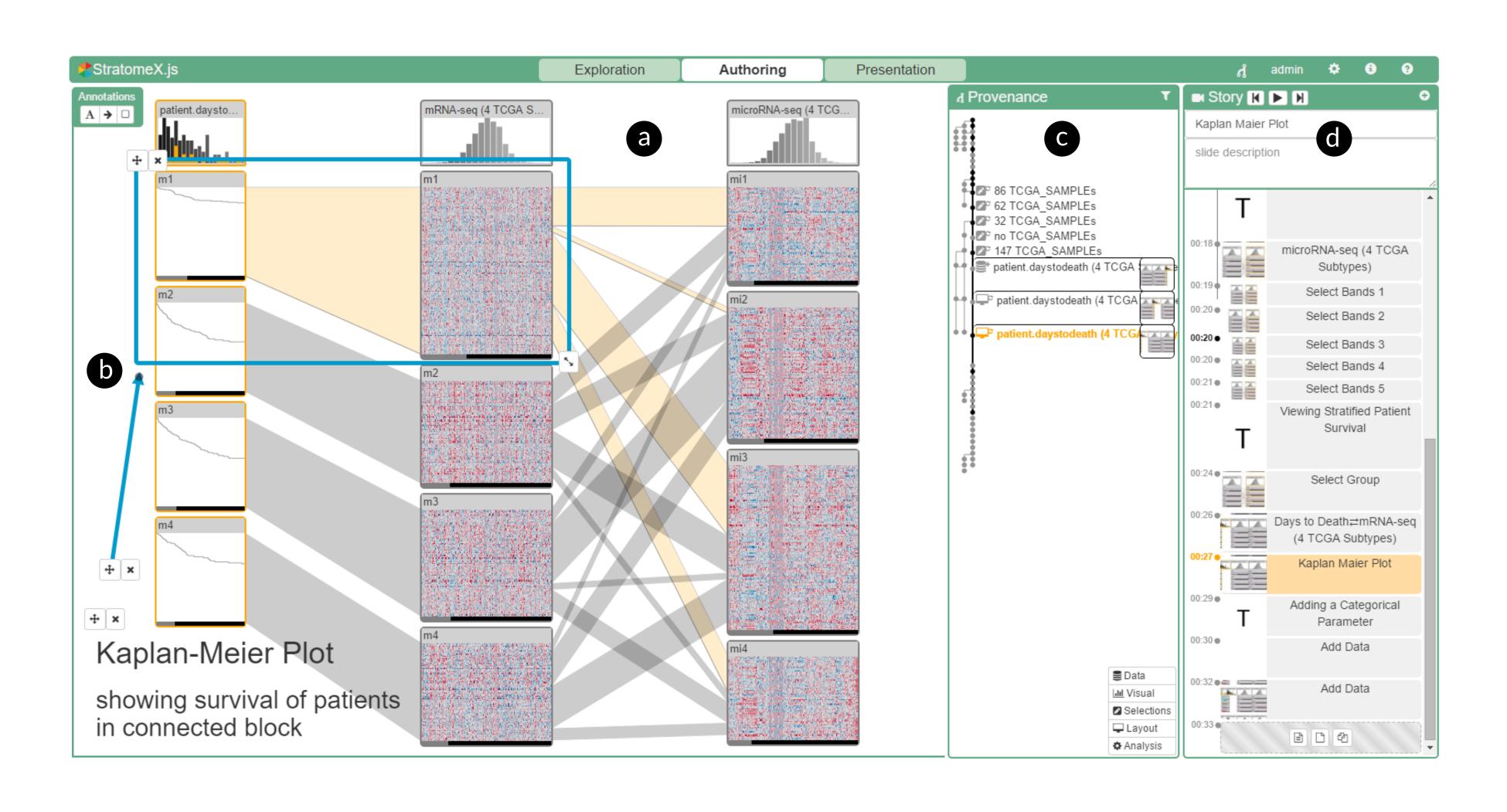


#### Genomic Data

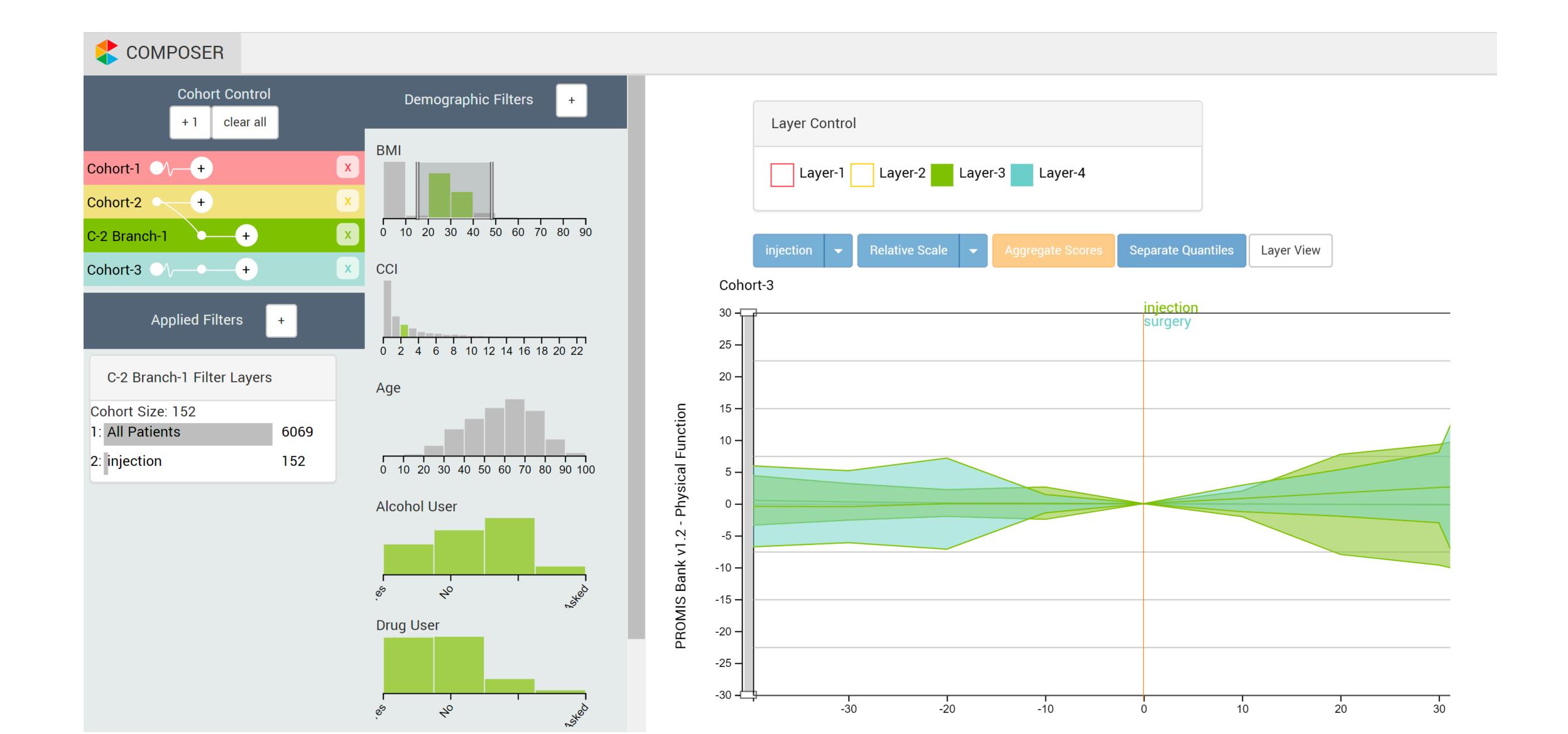
# Alternative Splicing / mRNA-seq



# Reproducibility, Storytelling, Annotation, and Integration in Computational Workflows



#### EHRS



#### Floout You

#### 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

Lectures: introduce theory

Design Critiques: develop "an eye" for vis design, critique, learn by example

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

# Course Components

Lecture Reading Discussion

# Theory

Design Lecture Design Studios



Labs D3 reading Self-study Office hours

# Design Skills - Coding Skills



```
<meta charset="utf-8">
text {
  font: 10px sans-serif;
</style>
<body>
<script src="http://d3js.org/d3.v3.min.js"></script>
<script>
```

## Schedule

Lectures: Tuesday and Thursday 12:25-1:45 pm, L103

**WEB** 

Labs: Wednesday, 6:00-7:00 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

### 

17:00 Kiran's Office I

17:00 Kiran's Office

Fall break

		17:00 Kiran's Office I	
Events shown in time zone: Mountain Time - Denver			
Subject to change			

12:25 Vis Lecture

#### Week 1

### Lecture 1: Introduction Tuesday, August 20 What is visualization? Why is it important? Who are we? Course overview.

#### Recommended reading

14:00 Alex Lex Office

14:00 Alex Lex Office

12:25 Vis Lecture

14:00 Alex Lex Office

Oct 1

- A Tour through the Visualization Zoo. Jeffrey Heer, Michael Bostock, Vadim Ogievetsky. Communications of the ACM, 53(6), pp. 59-67, Jun 2010.
- The Value of Visualization. Jarke van Wijk. Proceedings of the IEEE Visualization Conference, pp. 79-86, 2005.

# Information <a href="http://dataviscourse.net">http://dataviscourse.net</a>

### Uisualization for Data Science cs-5630 / cs-6630



Home Syllabus Schedule Project Resources Fame



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.

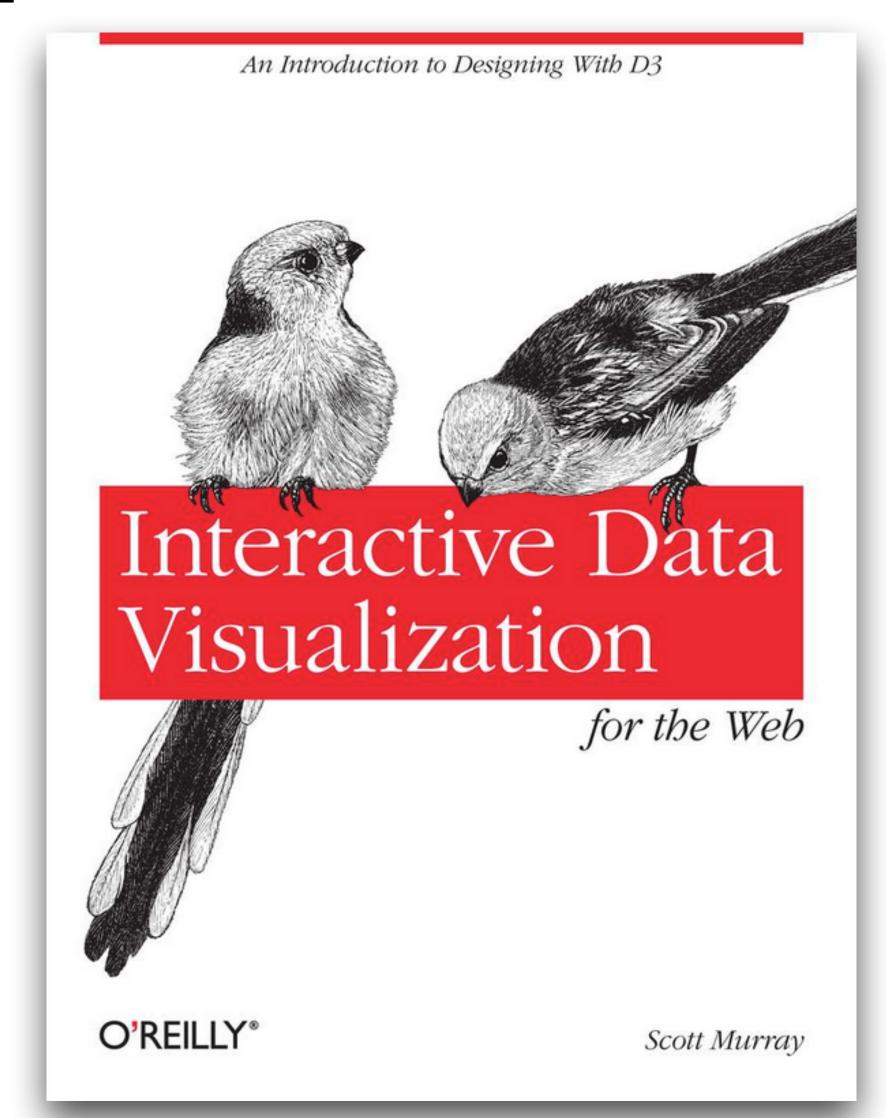
The course begins by bootstrapping your web development skills, moves on to fundamentals of perception, introduces data types you will encounter, and then focuses on visualization techniques and methods for a broad range of data types. An integral component of the course are regular design critiques and redesigns that will hone your skills in understanding, critiquing and developing visualization techniques.

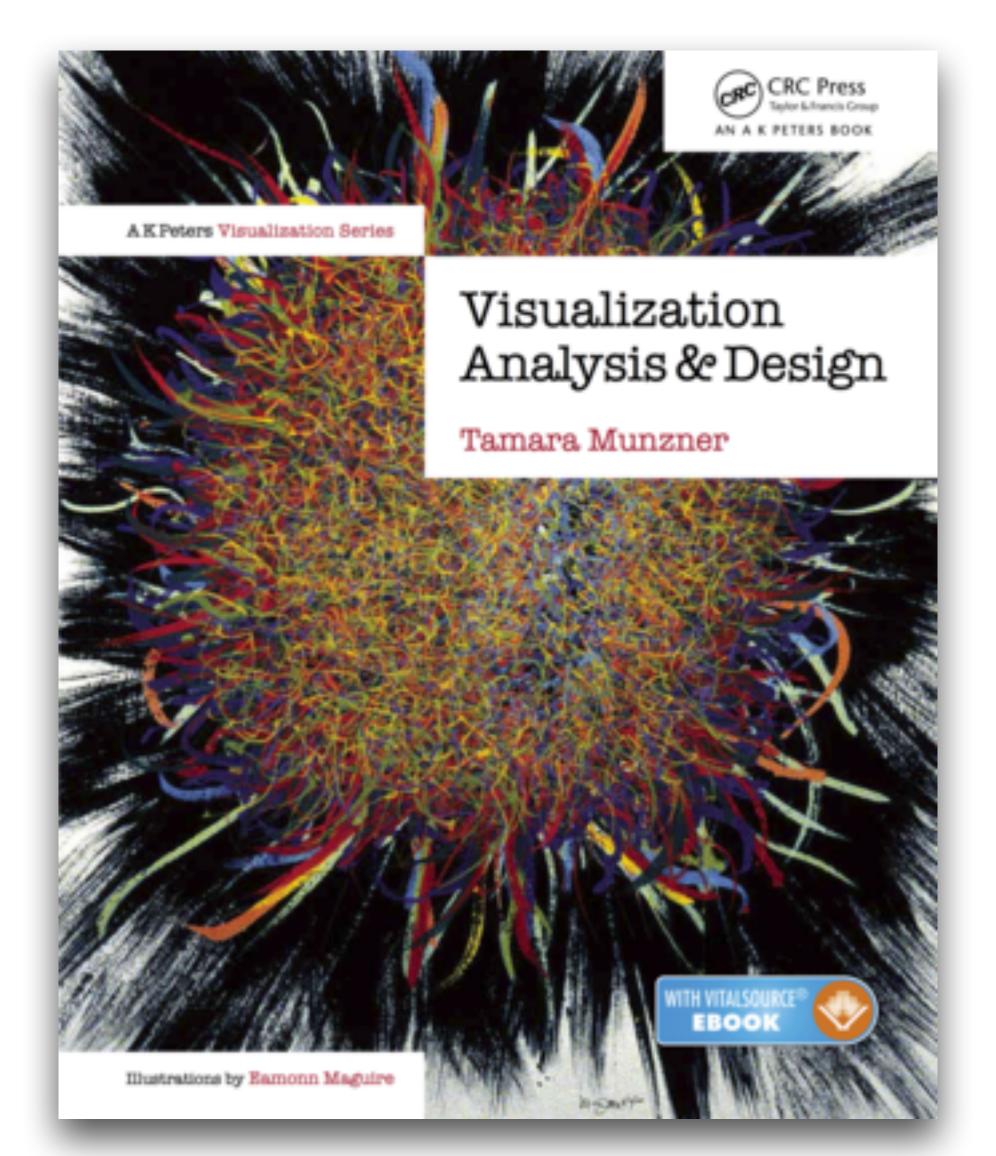
The course is offered in the fall term 2017 at the University of Utah in two variants: CS-5630 for undergraduates and CS-6630 for graduate students, with a special section of CS-6630 (002) designated for data certificate students. Classes start on Tuesday.

### Communicate

### Slack http://dataviscourse2019.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/574005 Homework submissions, Grades **Office Hours** Alex: Tuesdays after Class, WEB 3887 TAs: starting next week E-Mail alex@sci.utah.edu

# Required Books

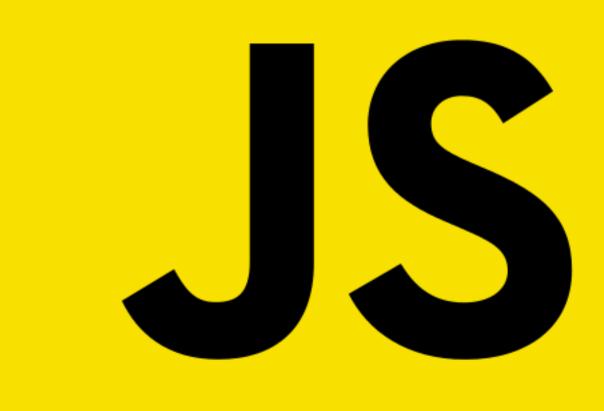




# Programming

### HTML









### Is this course for me???



# 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?

6 Homework Assignments: 40%

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

### 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 <a href="https://safeu.utah.edu/">https://safeu.utah.edu/</a>
- Please review the syllabus on these issues and the student code of conduct at <a href="https://regulations.utah.edu/academics/6-400.php">https://regulations.utah.edu/academics/6-400.php</a>

# 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: <a href="http://www.cs.utah.edu/wp-content/uploads/2014/12/cheating\_policy.pdf">http://www.cs.utah.edu/wp-content/uploads/2014/12/cheating\_policy.pdf</a>

You will fail the class if you cheat.

A "strike" will be recorded.

We will automatically check for plagiarism in all your submissions.

# No Device Policy

No Computers, Tablets, Phones in lecture hall except when used for exercises

Switch off, mute, flight mode

Why?

It's better to take notes by hand

Notifications are designed to grab your attention

Applies to theory lectures, coding along in technical lectures encouraged

### This Week

HW0, including course survey

Lecture on Perception

Readings

D3 Book, Chapters 1-3

VDA Book, Chapter 1

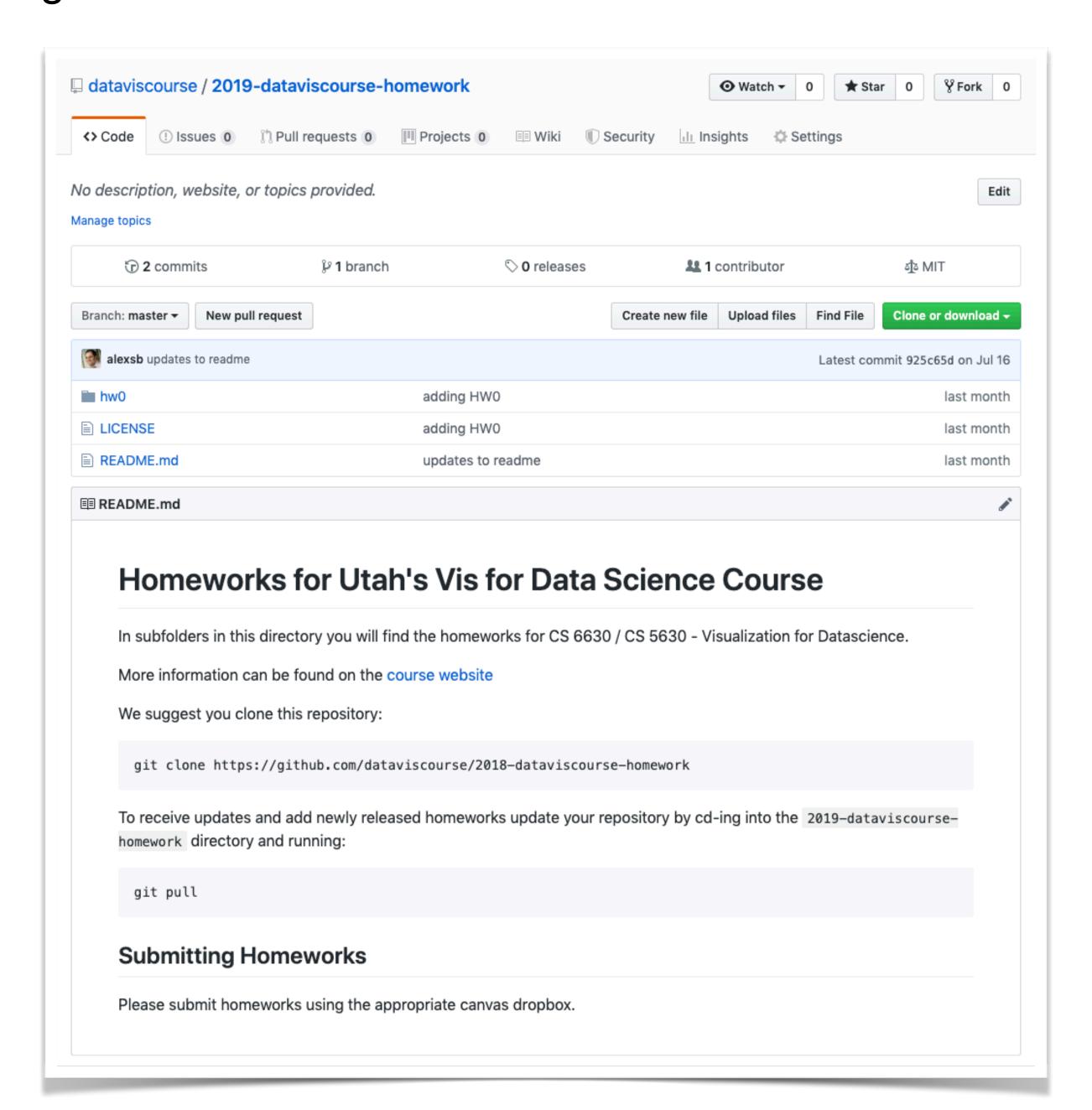
#### **Table of Contents**

1. Introduction	
Why Data Visualization?	
Why Write Code?	
Why Interactive?	
Why on the Web?	
What This Book Is	
Who You Are	
What This Book Is Not	
Using Sample Code	
Thank You	
2. Introducing D3	
What It Does	
What It Doesn't Do	
Origins and Context	
Alternatives	1
Easy Charts	1
Graph Visualizations	1
Geomapping	1
Almost from Scratch	1
Three-Dimensional	1
Tools Built with D3	1
3. Technology Fundamentals	1
The Web	1
HTML	1
Content Plus Structure	1

### Next Week

HW1 due
Introduction to Git, HTML, CSS
Office hours start!

### https://github.com/dataviscourse/2019-dataviscourse-homework



### Newish Track: Human Centered Computing NON-CS COURSES

#### REQUIRED COURSES

CS 6540 - HCI (humans + interfaces)

CS 6963 - Advanced HCI (humans + things)

CS 6630 - Visualization for Data Science (humans + data)

ED PS 6010: Introduction to Stats and Research Design (methods)

#### **ELECTIVES**

Pre-approved course list from within CS and across campus

Up to 3 electives can be taken from outside CS

Design

DES 5320 - Typographic Communication

DES 5370 - Digital Fabrication

DES 5710 - Product Design and Development

Ed Psych

ED PSY 6030 - Introduction to Research Design

Psych

PSY 6120 - Advanced Human Cognition

PSY 6140 - Cognitive Neuroscience Approaches to Research

PSY 6420 - Methods in Social Psychology

PSY 6700 - Neuropsychology

Anthropology

ANTH 6169 - Ethnographic Methods

Sociology

SOC 6110 - Methods of Social Research

EAE

EAE 6900 - Games User Research

EAE 6900 - A.I. For Games

# Companion Course: Visualization for Scientific Data



CS 5635 / CS 6635
Valerio Pascucci
Spring 2020

