DATA VISUALIZATION AND DATA EXPLORATION





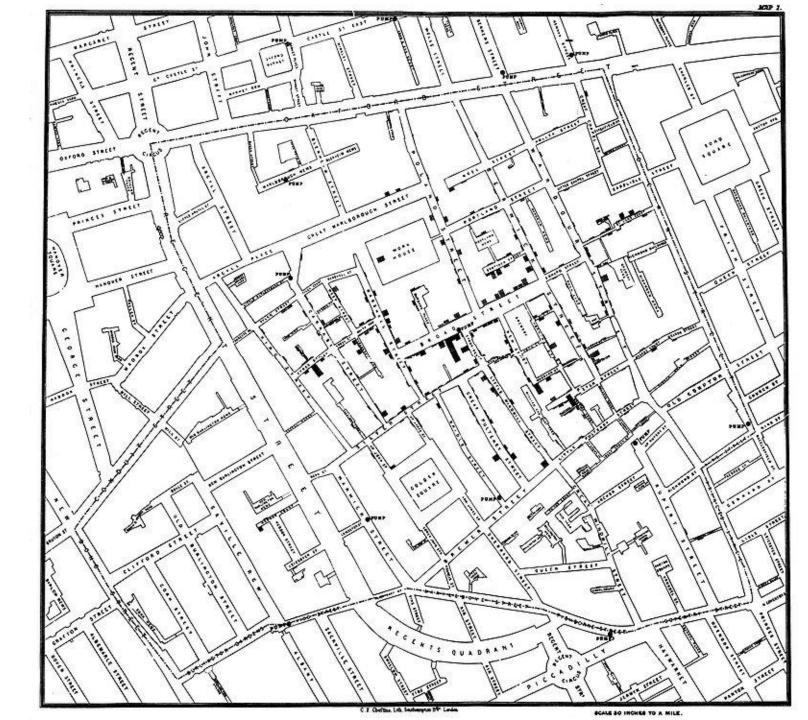
"Discovery is no longer limited by the collection and processing of data, but rather management, analysis, and visualization."

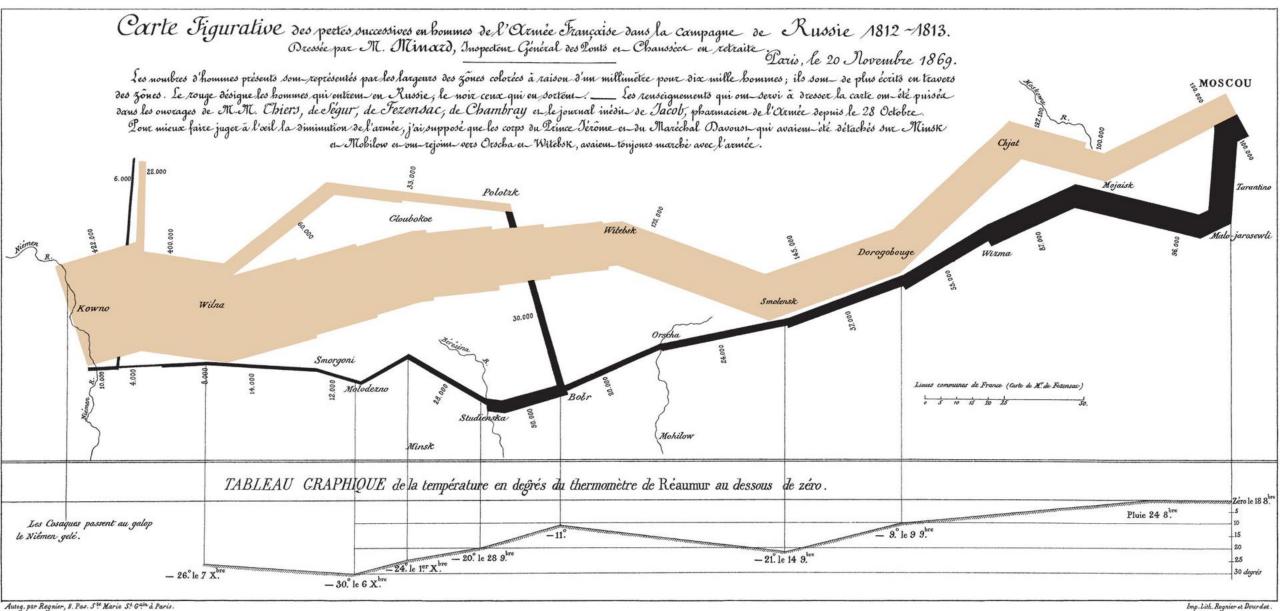
@DamianMingle



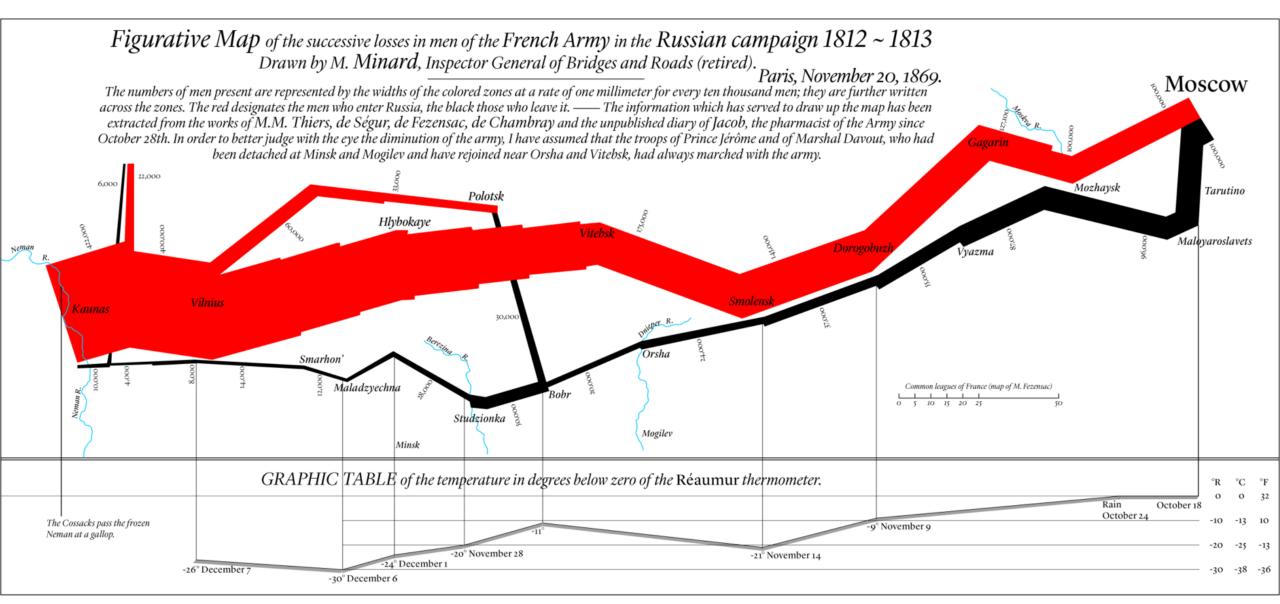
London's Cholera Outbreak of 1854

Physician John Snow links the outbreak to a contaminated well by plotting number of cases on a map, jump-starting the science of epidemiology.





Minard's March to Moscow



Minard's March to Moscow

INFOGRAPHICS

Created for **story-telling** purposes (**subjective**)

Intended for a **specific** audience

Self-contained and discrete

Graphic design aspect is key

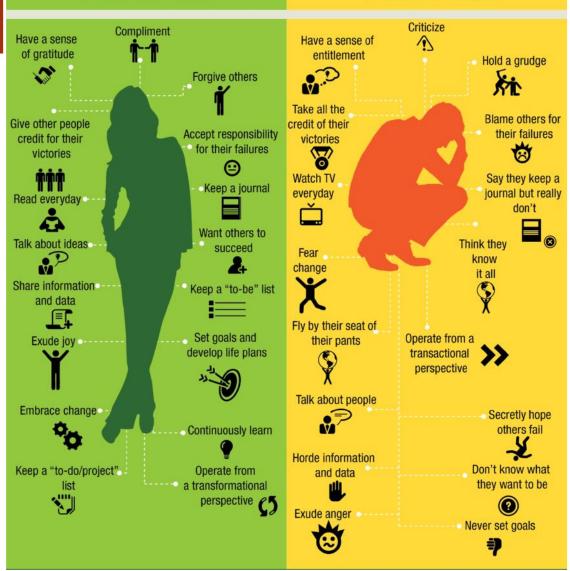
Cannot usually be re-used with other data

Can incorporate unquantifiable information



SUCCESSFUL PEOPLE

UNSUCCESSFUL PEOPLE



DATA VISUALIZATION

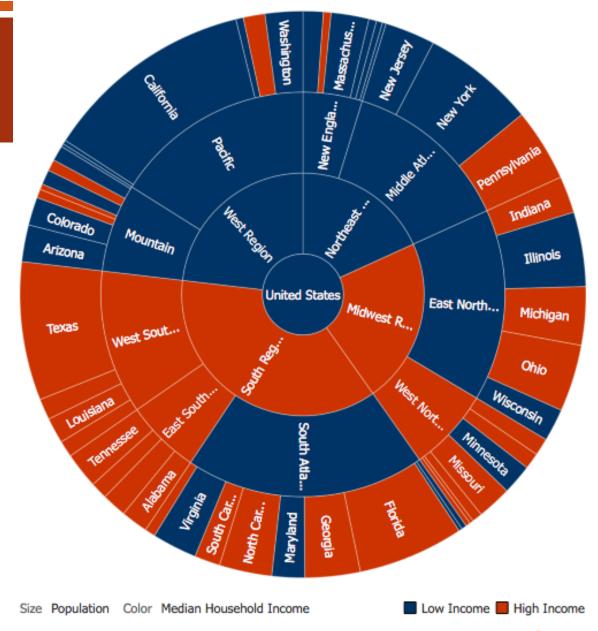
A **method**, as well as an item (**objective**)

Typically focuses on the quantifiable

Used to make sense of the data or to make it accessible (datasets can be massive and unwieldy)

May be generated automatically

The look and feel are less important than the insights conveyed by the data





DATA UP TO THE 20TH CENTURY

In the 20th century, data problems were mostly related to

- engineering (design of machines)
- sciences (formulation of theories)

Problems were solved **empirically**, **theoretically**, or through **computation**.



DATA UP TO THE 20TH CENTURY

Engineers equipped machines with sensors \Rightarrow used data to assess if the machines behaved as expected & to improve designs.

Scientists set up experiments \Rightarrow used data to test the validity of theories.

Experiments are expensive; relatively few data points are generated.

Data contained additional information which is often ignored.

Example: Mendel's experimental data, analyzed by Fisher, found to be too good to be true.





DATA IN THE 21ST CENTURY

In the 21st century, there is:

- there is more data
- it's mostly digital
- it's mostly **observed** (rather than generated by designed experiment)

Problems are solved **empirically**, **theoretically**, through **computation** and/or **data exploration/visualization**.



DATA IN THE 21ST CENTURY

Empirically: observe and describe what happens

Theoretically: generalize and build models and generalizations to understand what happens

Computationally: design computer simulations to better understand what happens

Data Exploration/Visualization: the new approach to understanding



EXERCISE

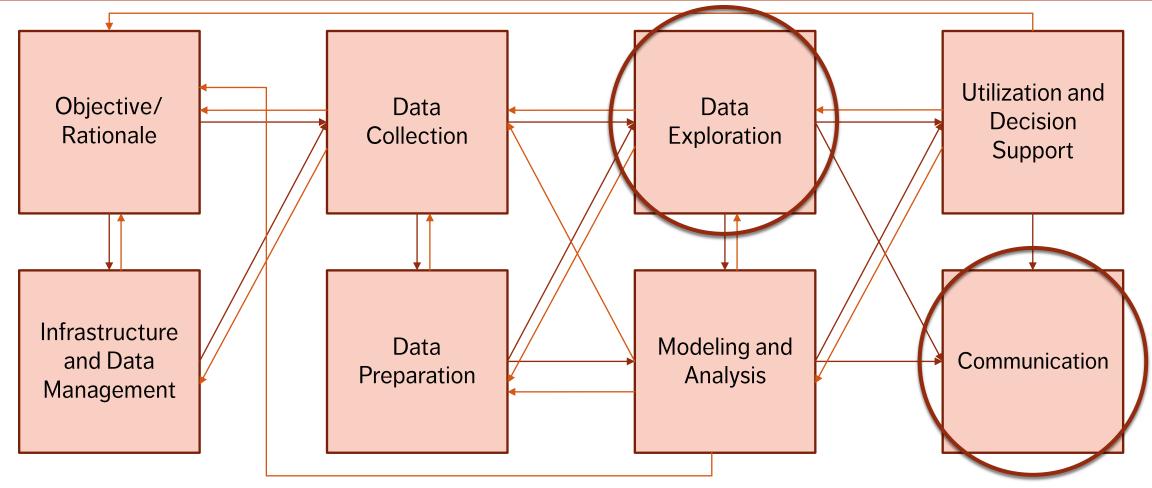
In teams or individually, identify a few data visualizations that appeal to you (professionally, esthetically, or both).

What is the story being told by the visualization?

What kind of data is needed to build these visualizations?



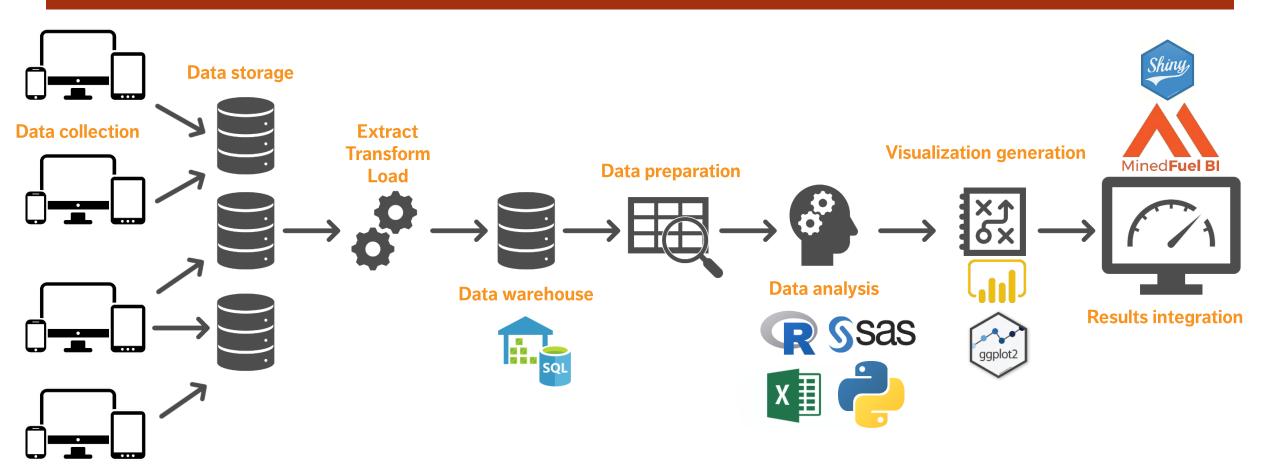
THE (MESSY) ANALYSIS PROCESS







DATA ENVIRONMENT



EXERCISE

In teams or individually, identify work scenarios for which data visualization could prove useful.

What insight could be drawn from such visualizations?

Would such visualizations get a buy-in from your supervisors/employers?

How much work would be required to get from design to completion? Are the obstacles mostly of a technical nature? Related to data procurement?

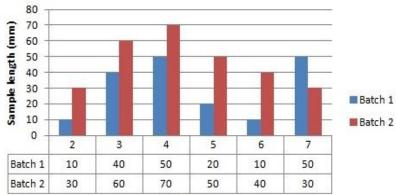


OVERVIEW

The past is **data-driven**:

- mostly Excel (or reporting tools like Cognos)
- mostly numbers, tables and non-interactive graphs
- distributed on desktop computers, by email, in PowerPoint presentation
- static, mostly backwards looking (lagging indicators)
- KPIs and dashboards were somewhat contrived

Region	Jan-06	Feb-06	Mar-06	Apr-06	May-06	Jun-06	Jul-06
			Actuals	;			
attle	111	653	1,598	3,411	3,972	5,092	5,2
ise	26,779	27,867	29,153	30,557	33,402	35,400	35,4
rtland	33,078	34,401	37,535	39,916	41,357	45,306	46,6
okane	25,417	26,669	28,092	29,020	29,674	30,501	30,8
orth Region	199,841	211,053	226,789	242,957	256,605	273,640	277,7
			Plan				
attle	693	468	790	1,383	2,205	3,180	4,2
ise	29,525	26,062	27,088	28,269	29,536	30,821	32,1
rtland	32,276	34,708	36,737	38,857	41,066	43,364	45,7
okane	30,500	26,644	27,987	29,430	30,994	32,594	34,2
orth Region	191,783	203,916	216,524	230,474	246,390	263,378	281,2
			Varianc	e			
attle	-582	185	808	2,029	1,767	1,912	1,0
ise	-2,746	1,805	2,064	2,288	3,866	4,578	3,2
rtland	802	-307	798	1,059	291	1,942	9
okane	-5,082	25	105	-410	-1,320	-2,093	-3,3
orth Region	8.057	7.137	10.265	12.483	10.215	10.261	3.6





OVERVIEW

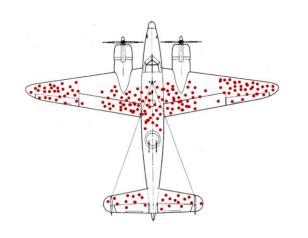
The future is **story-driven:**

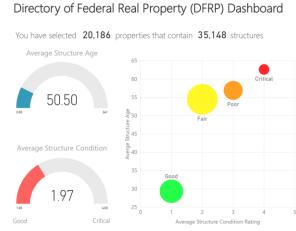
- new tools: Power BI, R, Qlickview etc.
- mostly visualizations, occasional numbers and tables
- distributed on the web (internal and external)
- dynamic and both backwards and forwards looking (leading and lagging indicators)
- data for everyone

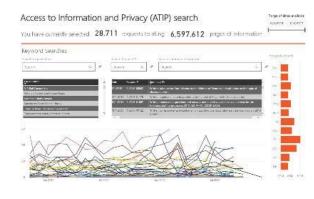


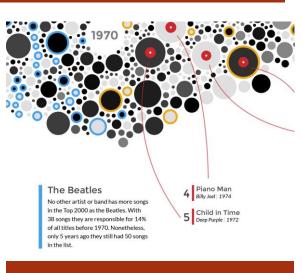


DEFINING CONTEXT









Seconds

Minutes

Fraction of Hour

Hours

Infographics/Data Viz

── Dashboards ── →

Reports ----



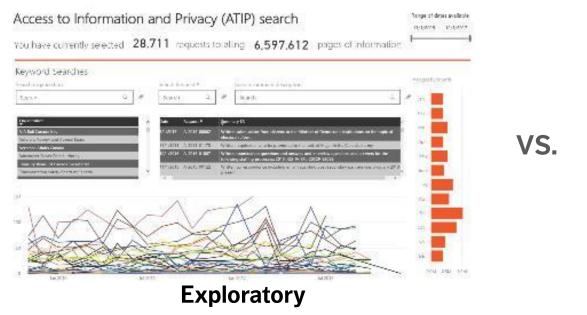


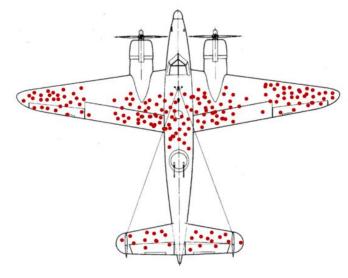


EXPLORATORY VS. EXPLANATORY ANALYSIS

Exploratory: understanding the **DATA** (associated with reports)

Explanatory: communicating a **STORY** (associated with dashboards and data viz)





Explanatory

SOME BASIC QUESTIONS

What system does your data represent – objects, attributes, relationships?

How does it represent this system – i.e. the data model?

Who made this dataset? When? For what purpose?

Assuming a flat file – what do the rows represent? What do the columns represent?

Do you even have enough information (e.g. **metadata**) to answer these questions? Where can you find more information?

NON-VISUALIZATION BASED SUMMARIES OF YOUR DATASET

Cl	N03	NH4
Min. : 0.222	Min. : 0.000	Min. : 5.00
1st Qu.: 10.994	1st Qu.: 1.147	1st Qu.: 37.86
Median : 32.470	Median : 2.356	Median : 107.36
Mean : 42.517	Mean : 3.121	Mean : 471.73
3rd Qu.: 57.750	3rd Qu.: 4.147	3rd Qu.: 244.90
Max. :391.500	Max. :45.650	Max. :24064.00
NA's :16	NA's :2	NA's :2

season

Length: 340

Class: character autumn spring summer winter Mode: character 80 84 86 90





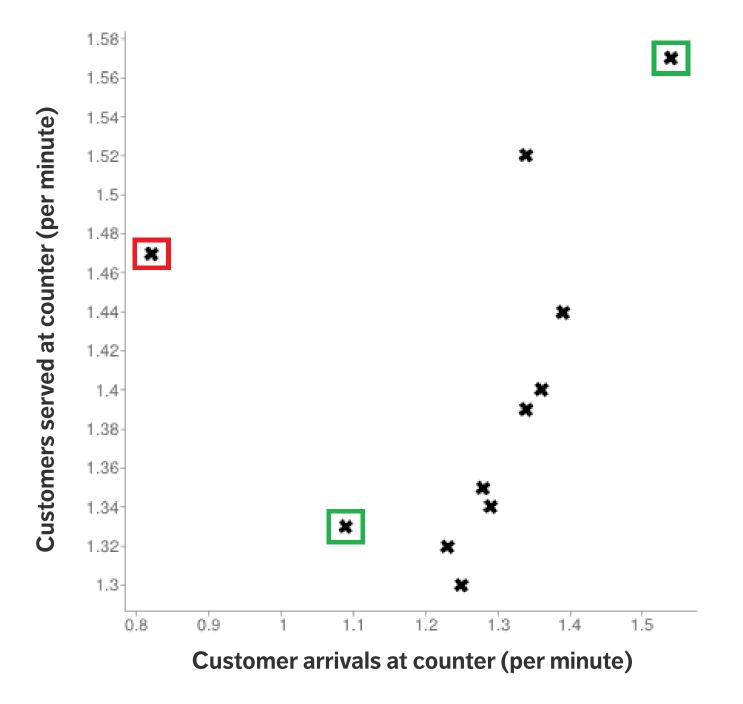
PRE-ANALYSIS USE

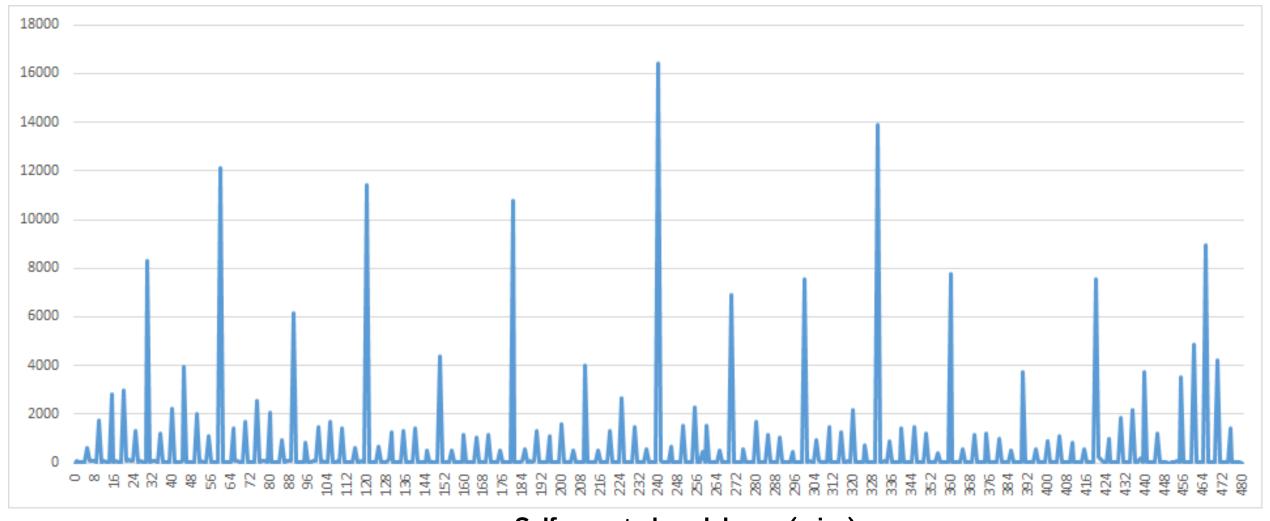
Data visualization can be used to set the stage for analysis:

- detecting anomalous entries invalid entries, missing values, outliers
- shaping the data transformations binning, standardization, Box-Cox transformations, PCA-like transformations
- getting a sense for the data data analysis as an art form, exploratory analysis
- identifying hidden data structure clustering, associations, patterns informing the next stage of analysis







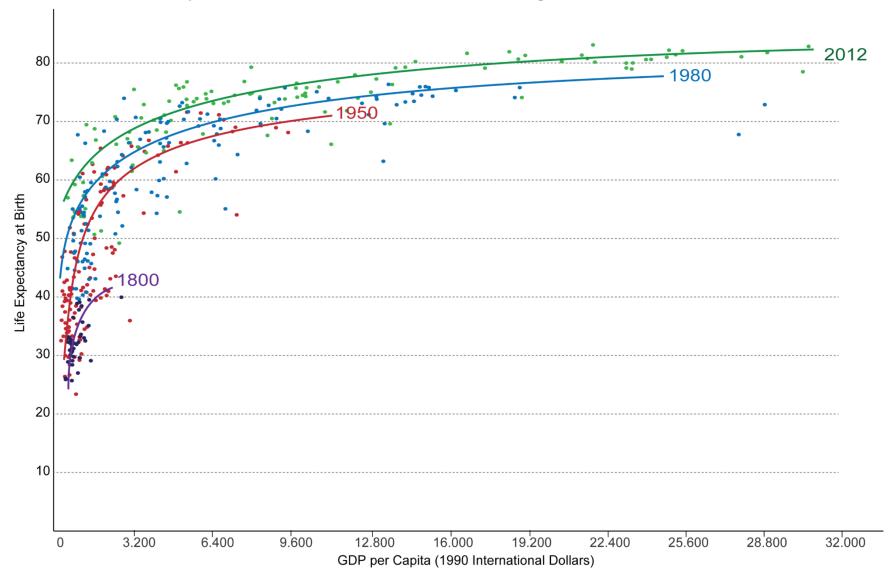


Self-reported work hours (mins)

in Data

Life Expectancy vs. GDP per Capita from 1800 to 2012 – by Max Roser GDP per capita is measured in International Dollars. This is a currency that would buy a comparable amount of goods and services a

U.S. dollar would buy in the United States in 1990. Therefore incomes are comparable across countries and across time.

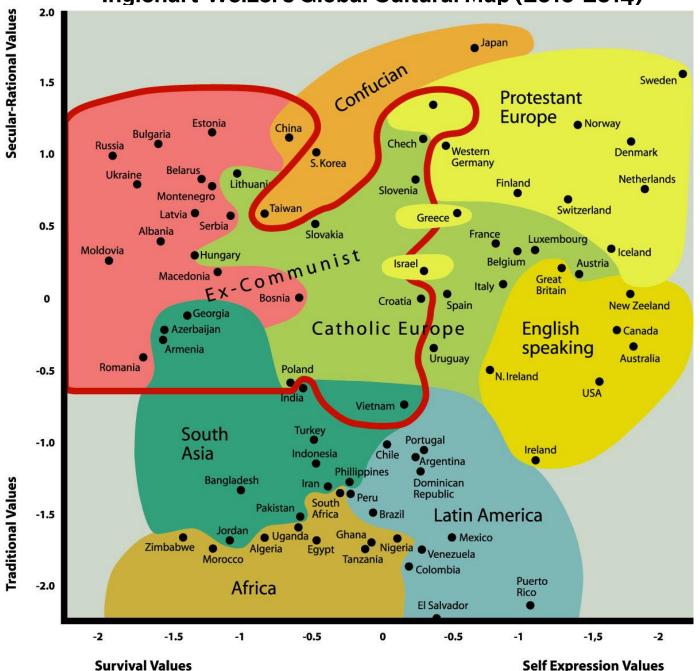


This graph displays the correlation between life expectancy and GDP per capita.

Countries with higher GDP have a higher life expectancy, in general.

The relationship seems to follow a logarithmic trend: the unit increase in life expectancy per unit increase in GDP decreases as GDP per capita increases.

Inglehart-Welzel's Global Cultural Map (2010-2014)



Factor Score

Traditional values

importance of religion, parent-child ties, deference to authority and traditional family values.

Secular-rational values

less emphasis on religion, traditional family values and authority.

Survival values

emphasis on economic and physical security.

Self-expression values

high priority to environmental protection, growing tolerance of foreigners, gays and lesbians and gender equality

DISCUSSION

Which of the pre-analysis uses of visualization is most relevant to your work?

