

Storytelling with Data

Nam Wook Kim

Mini-Courses — January @ GSAS
2018

Goal

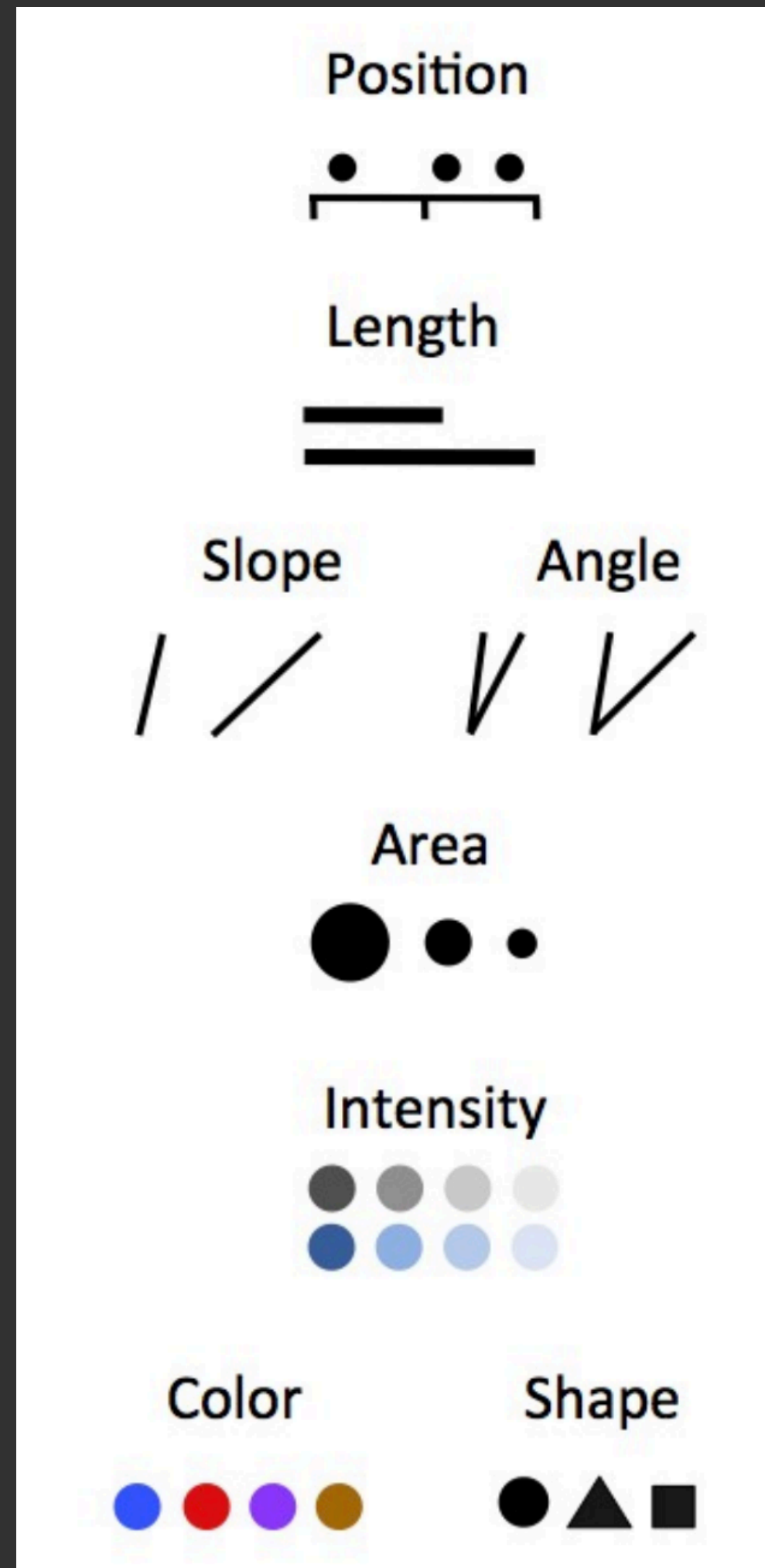
Beyond exploratory analysis:
visualization for communication

Data

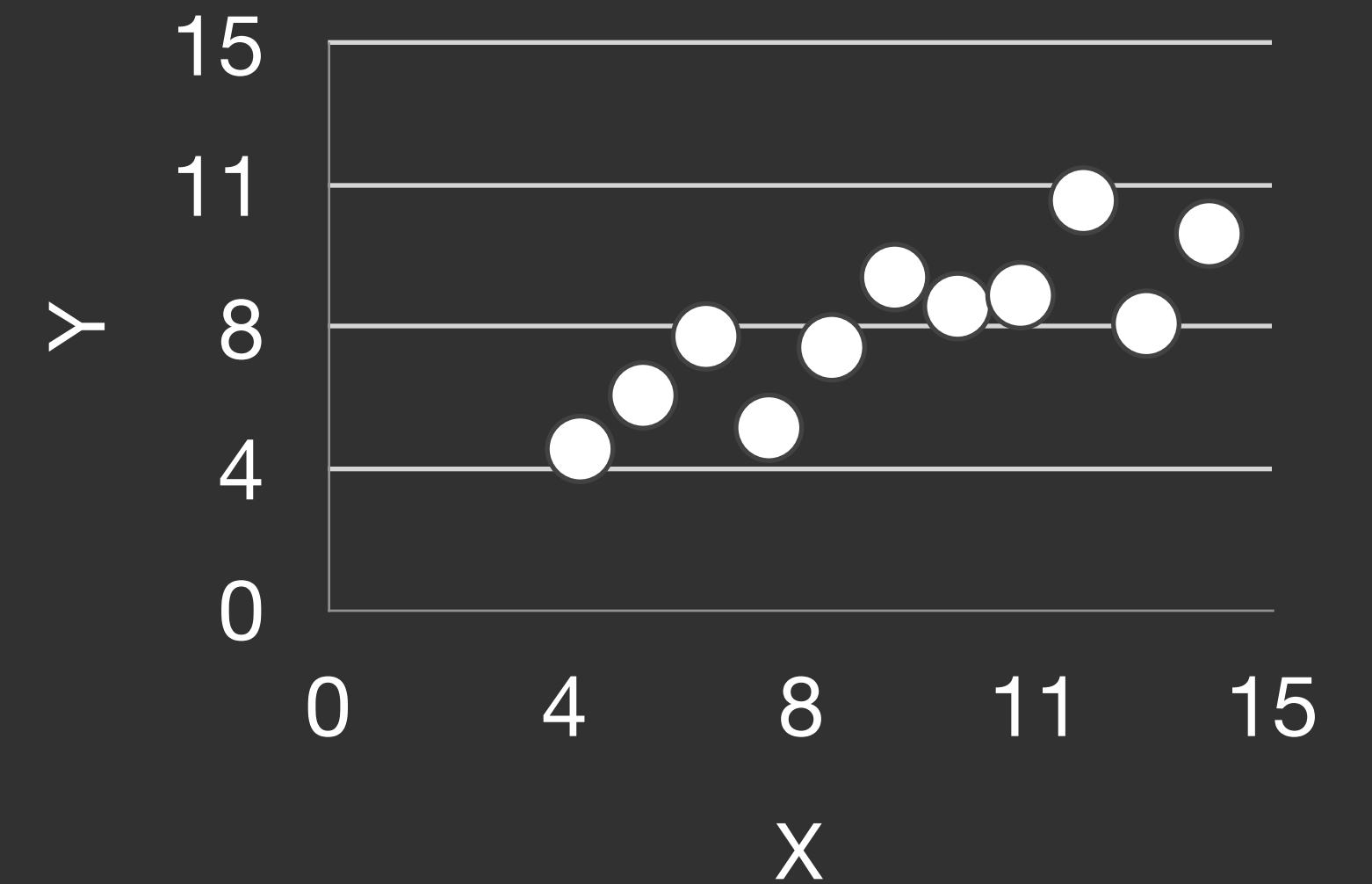
X	Y
10.0	8.04
8.0	6.95
13.0	7.58
9.0	8.81
11.0	8.33
14.0	9.96
6.0	7.24
4.0	4.26
12.0	10.84
7.0	4.82
5.0	5.68



Perceptual Variables



Visualization



Position (x, y)

Design Criteria

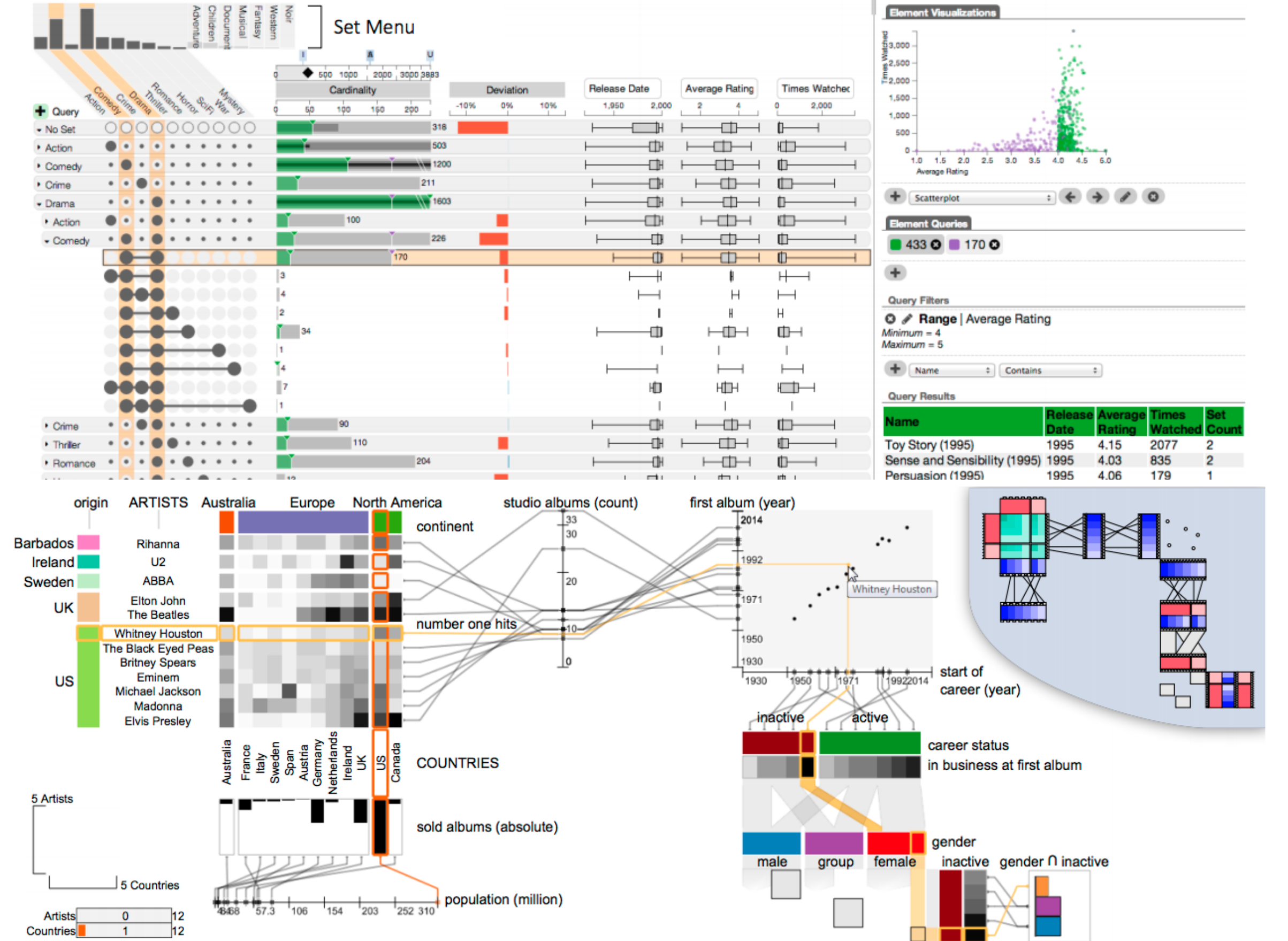
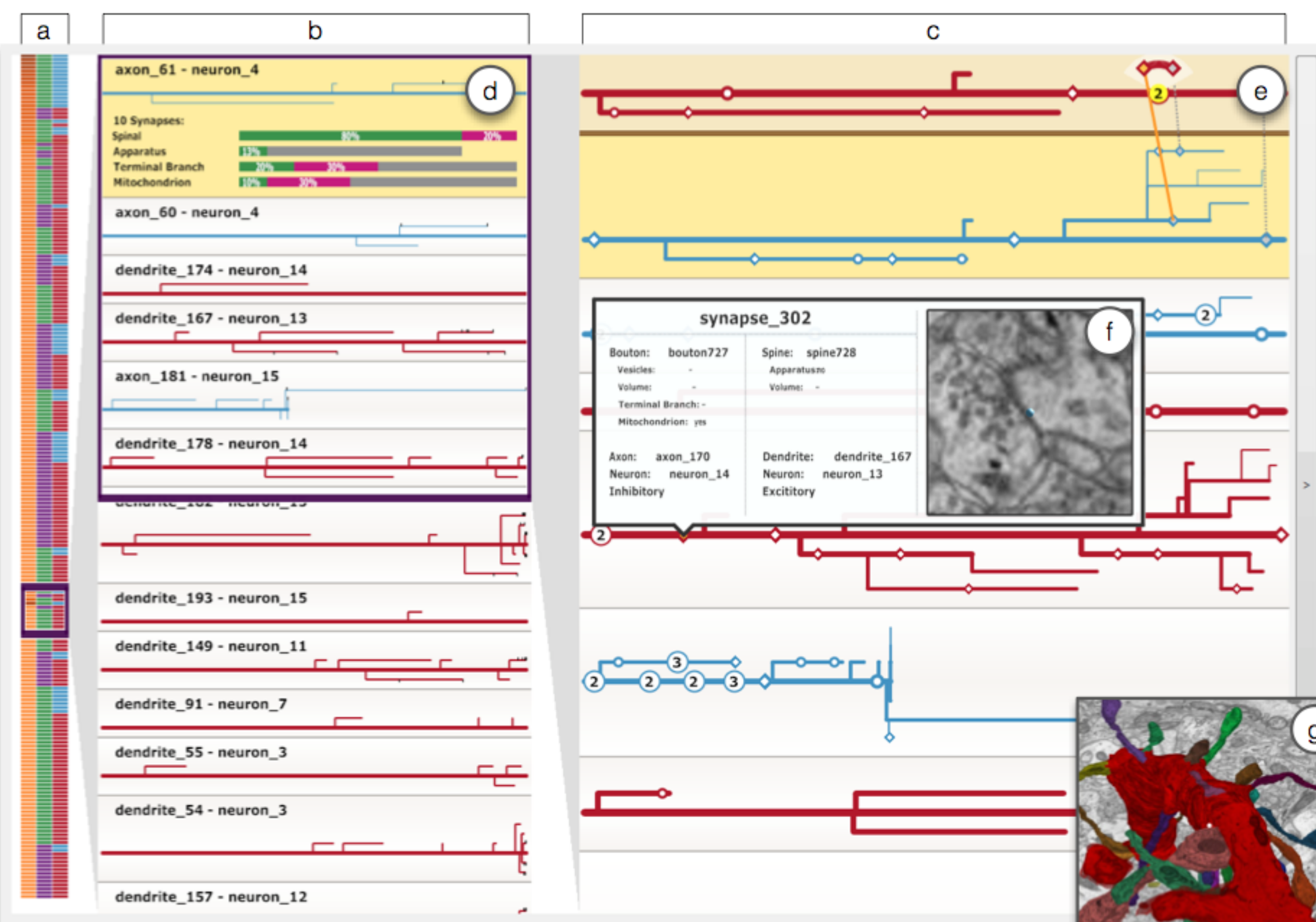
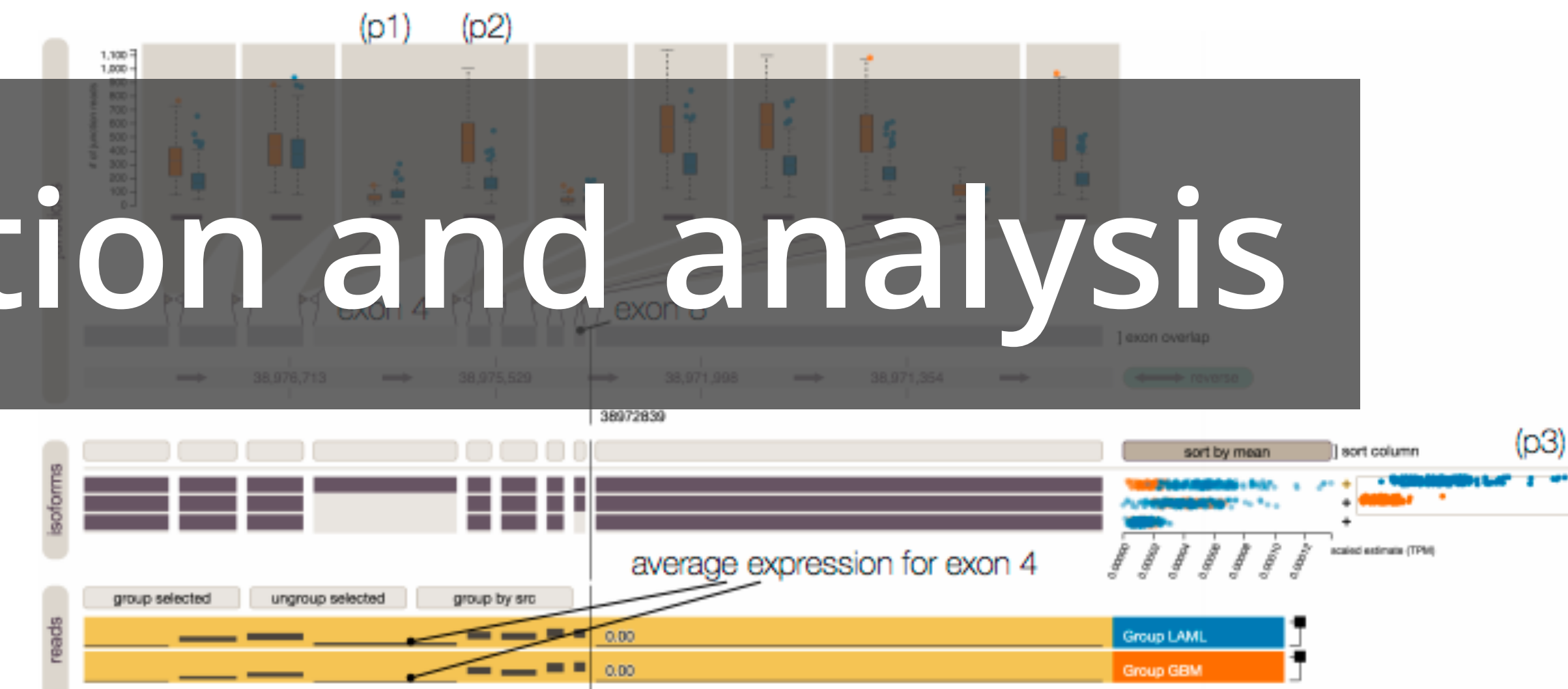
Expressiveness

A set of facts is expressible in a visualization if it expresses **all the facts** and **only the facts** in the data.

Effectiveness

A visualization is more effective than another one if the information conveyed is more **readily perceived**.

Rapid data exploration and analysis



Data ↔ Insights ↔ ?

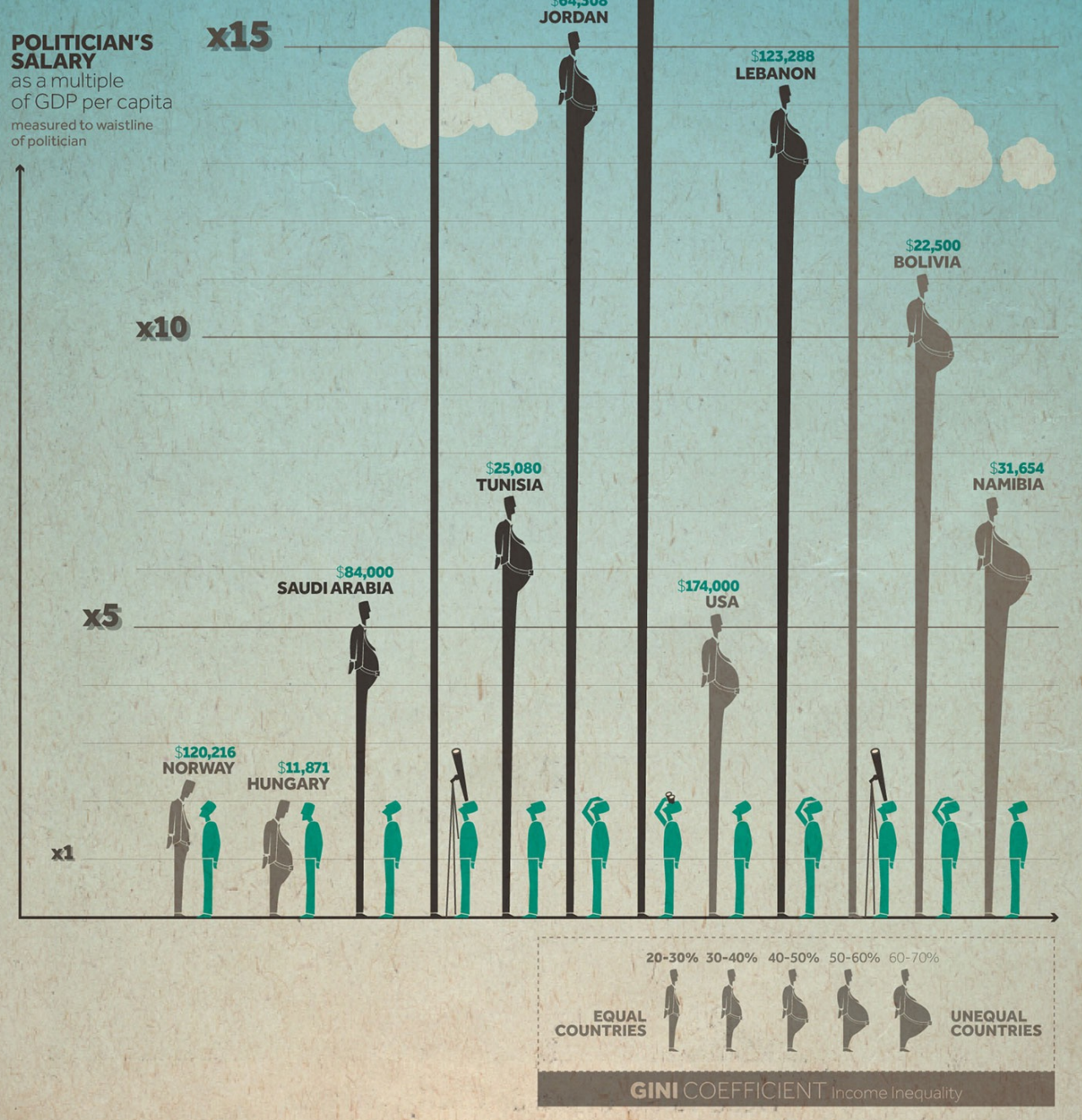
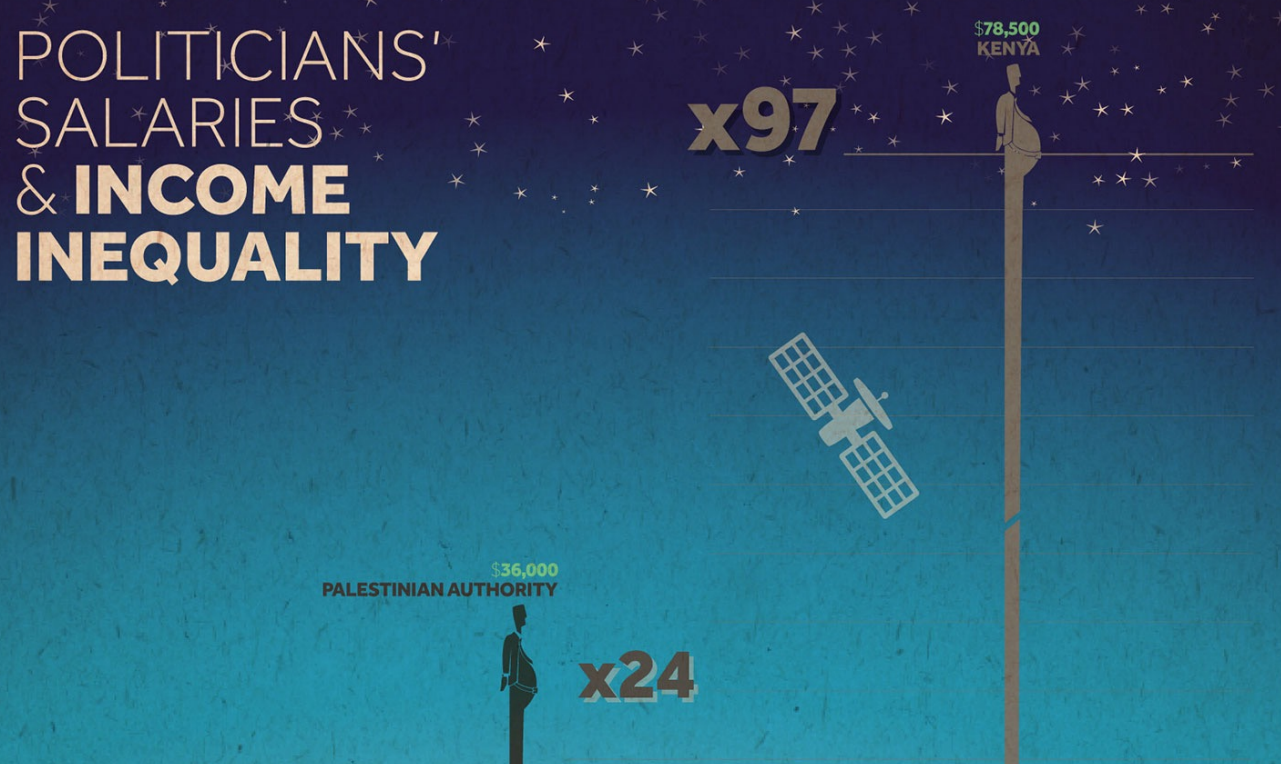
Exploratory

- Data Centered
- Domain Experts
- Analysis
- Desktop
- In-Lab

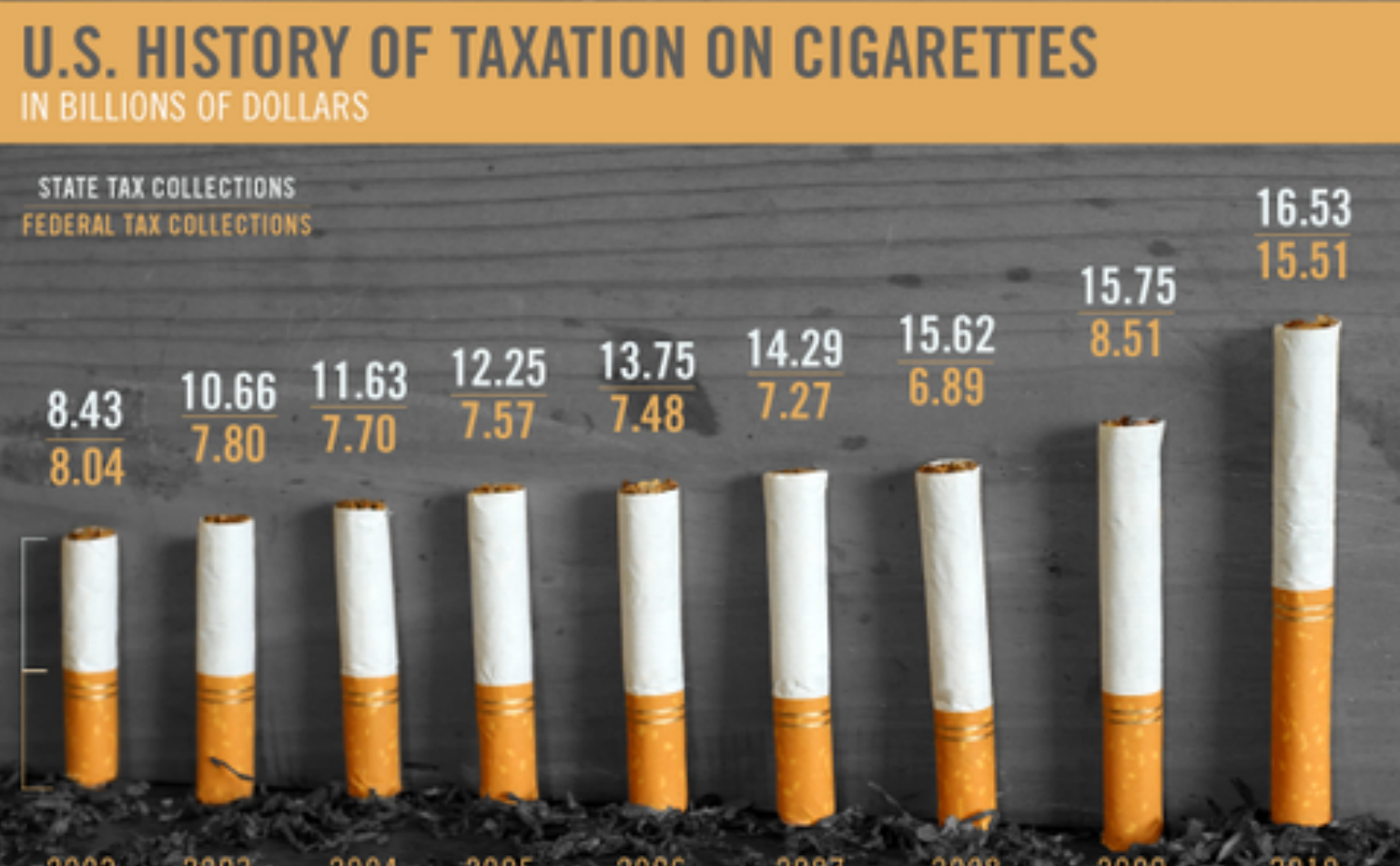
Need to **communicate** insights found from data



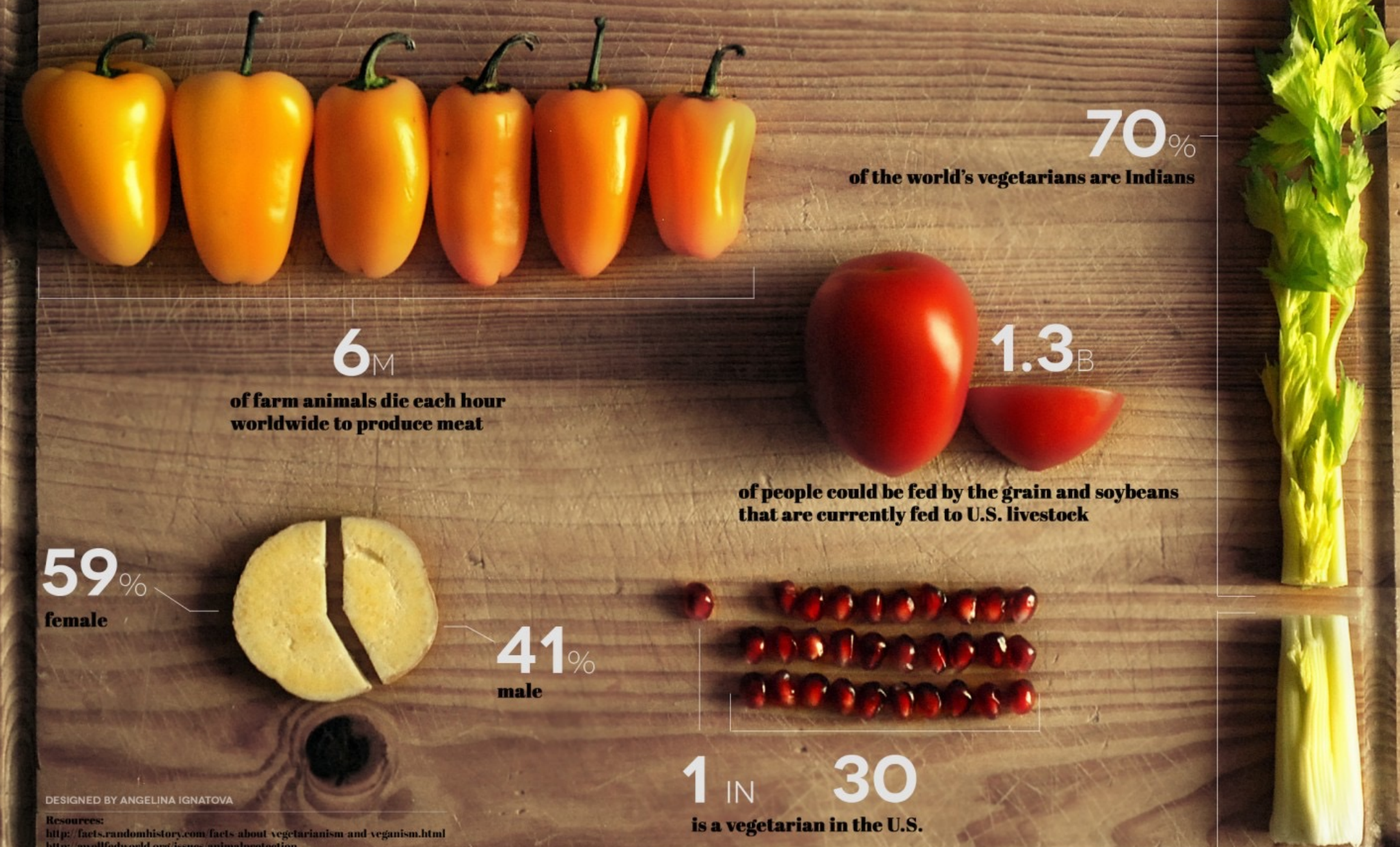
POLITICIANS' SALARIES & INCOME INEQUALITY



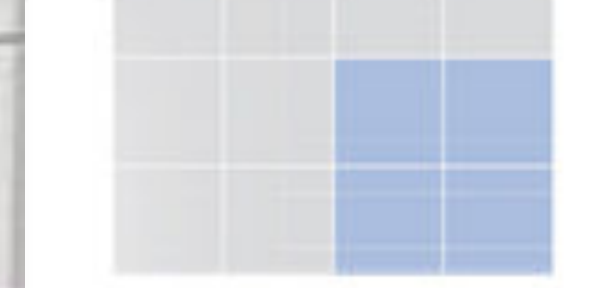
SOURCES: Politician salary data for MPs, deputies or equivalent. Data aggregated from multiple sources by Paiseef 22. Some data difficult to obtain and include all cases representative of total personal income or net worth. Bolivia <http://bit.ly/bolivia-sal> | France <http://bit.ly/france-sal> | Hungary <http://bit.ly/hungary-sal> | Jordan <http://bit.ly/jordan-sal> | Kenya <http://bit.ly/kenya-sal> | Namibia <http://bit.ly/namibia-sal> | Norway <http://bit.ly/norway-sal> | Saudi Arabia <http://bit.ly/saudi-sal> | USA <http://bit.ly/usa-sal> | Lebanon, Palestinian Authority, Tunisia data gathered from government sources and NGOs. GDP per capita, 2011 <http://bit.ly/gdp-capita> | Palestinian GDP per capita, 2011 <http://bit.ly/gdp-palestine> | GINI index income inequality data for most recent year available <http://bit.ly/gini-world> and <http://bit.ly/gini-arab>



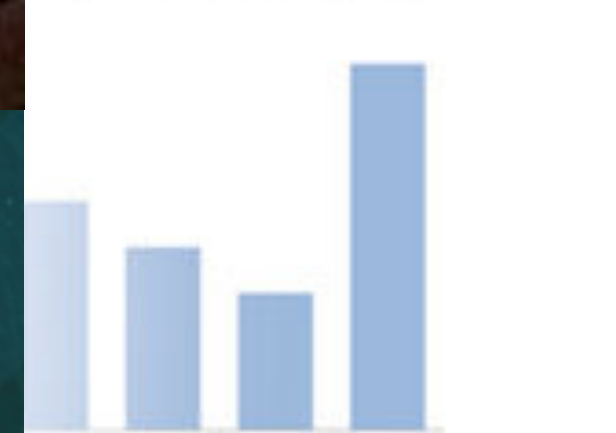
VEGETARIANISM IN NUMBERS



DESIGNED BY ANGELINA IGNATOVA
 Resources: <http://facts.randomhistory.com> facts about vegetarianism and veganism.html
<http://wellfedworld.org/issues/animalprotection>



	A	B	C
	15%	22%	42%
	40%	36%	20%
	35%	17%	34%
	30%	29%	26%
	55%	30%	58%
	11%	25%	49%



	A	B	C
Category 1	15%	22%	42%
Category 2	40%	36%	20%
Category 3	35%	17%	34%
Category 4	30%	29%	26%
Category 5	55%	30%	58%
Category 6	11%	25%	49%

cole nussbaumer knaflic

storytelling with data

a data visualization guide for business professionals

WILEY

Storytelling

Hans Rosling

tapestry

Storytelling is the most powerful way to put ideas into the world today.

Robert McKee

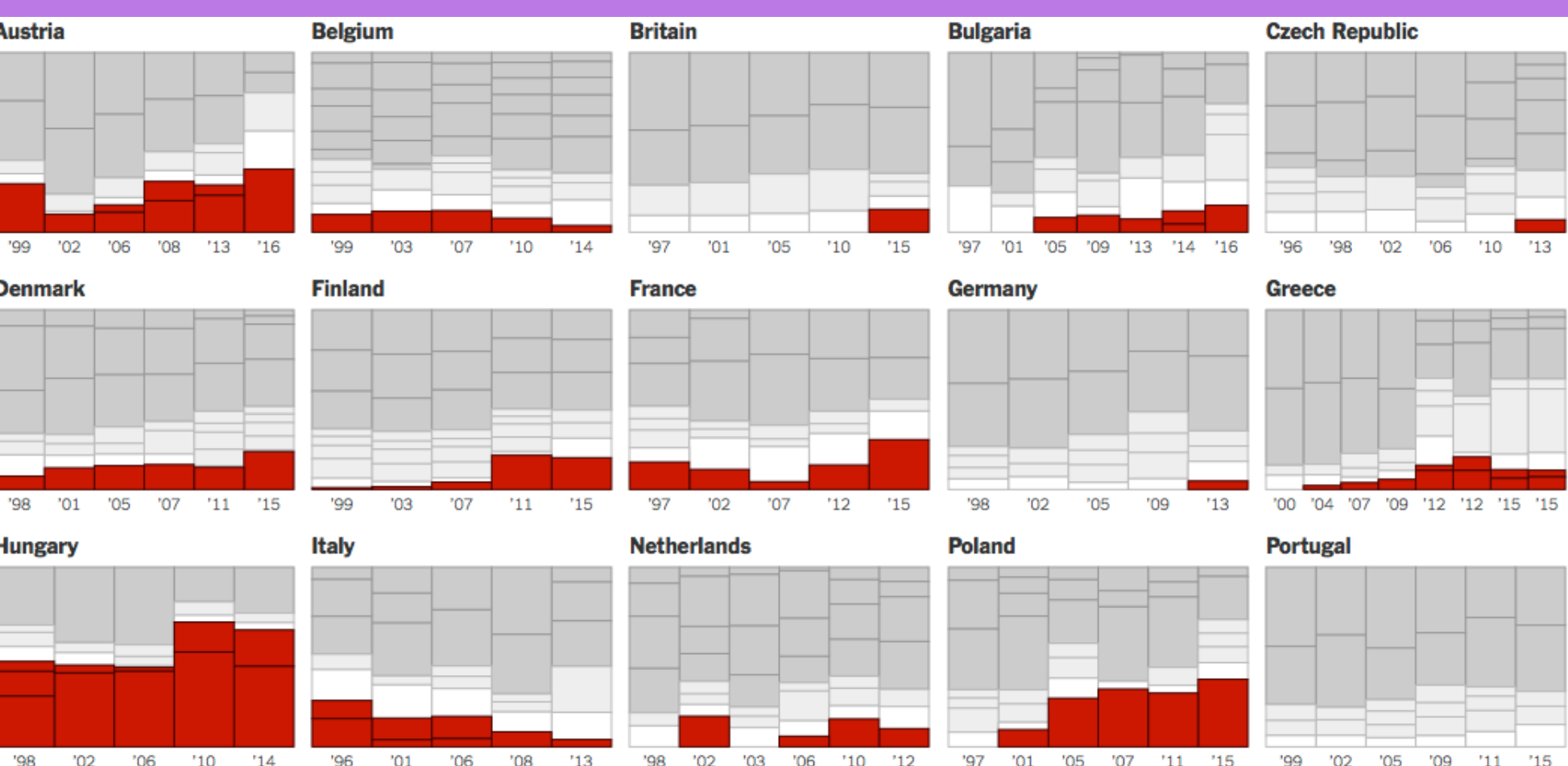
How Far Is Europe Swinging to the Right?

By GREGOR AISCH, ADAM PEARCE and BRYANT ROUSSEAU | UPDATED December 5, 2016

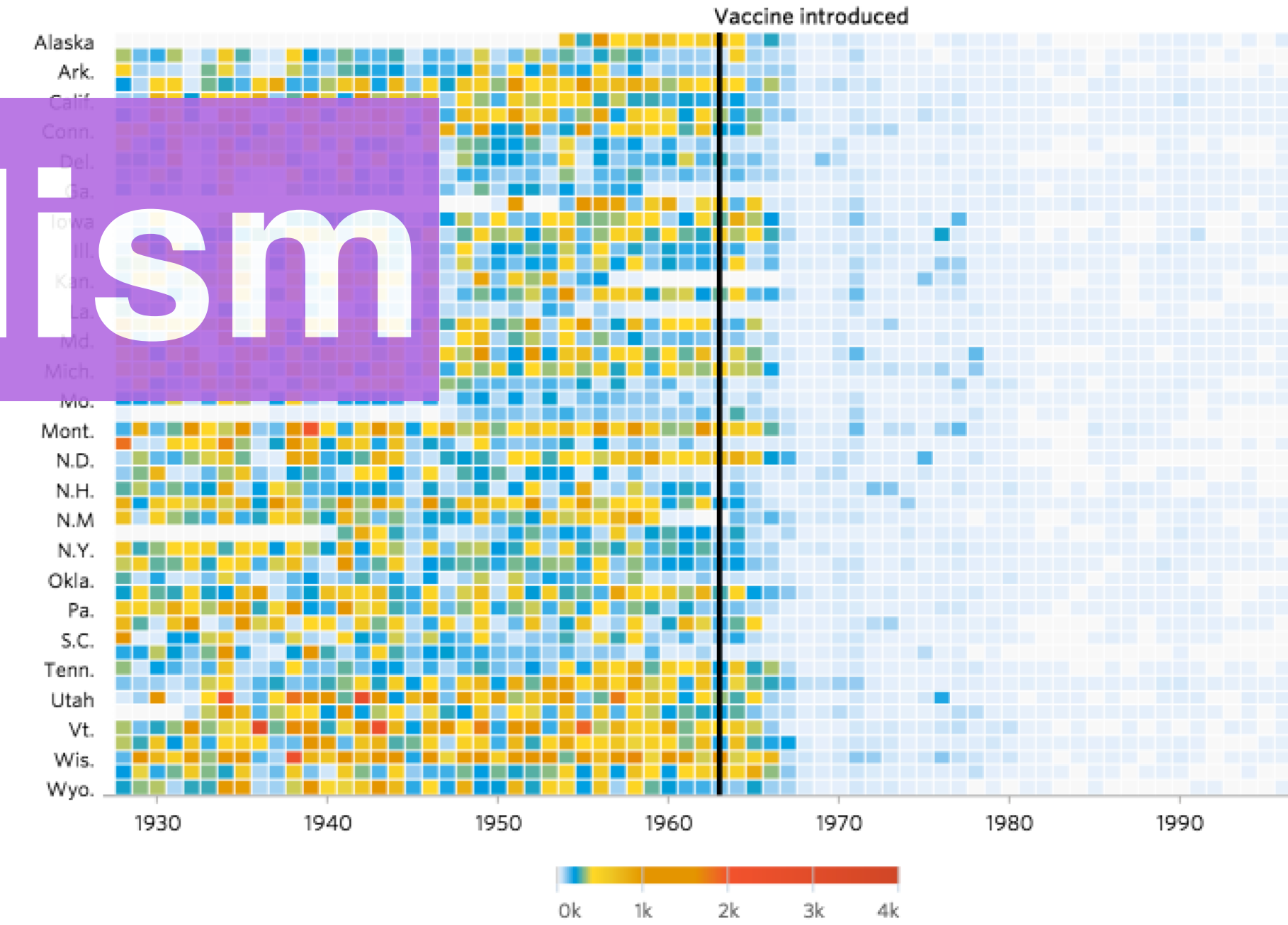
Amid a migrant crisis, sluggish economic growth and growing disillusionment with the European Union, right-wing parties in a growing number of European countries have made electoral gains. The right-wing parties included below range across a wide policy spectrum, from populist and nationalist to far-right neofascist.

Data Journalism

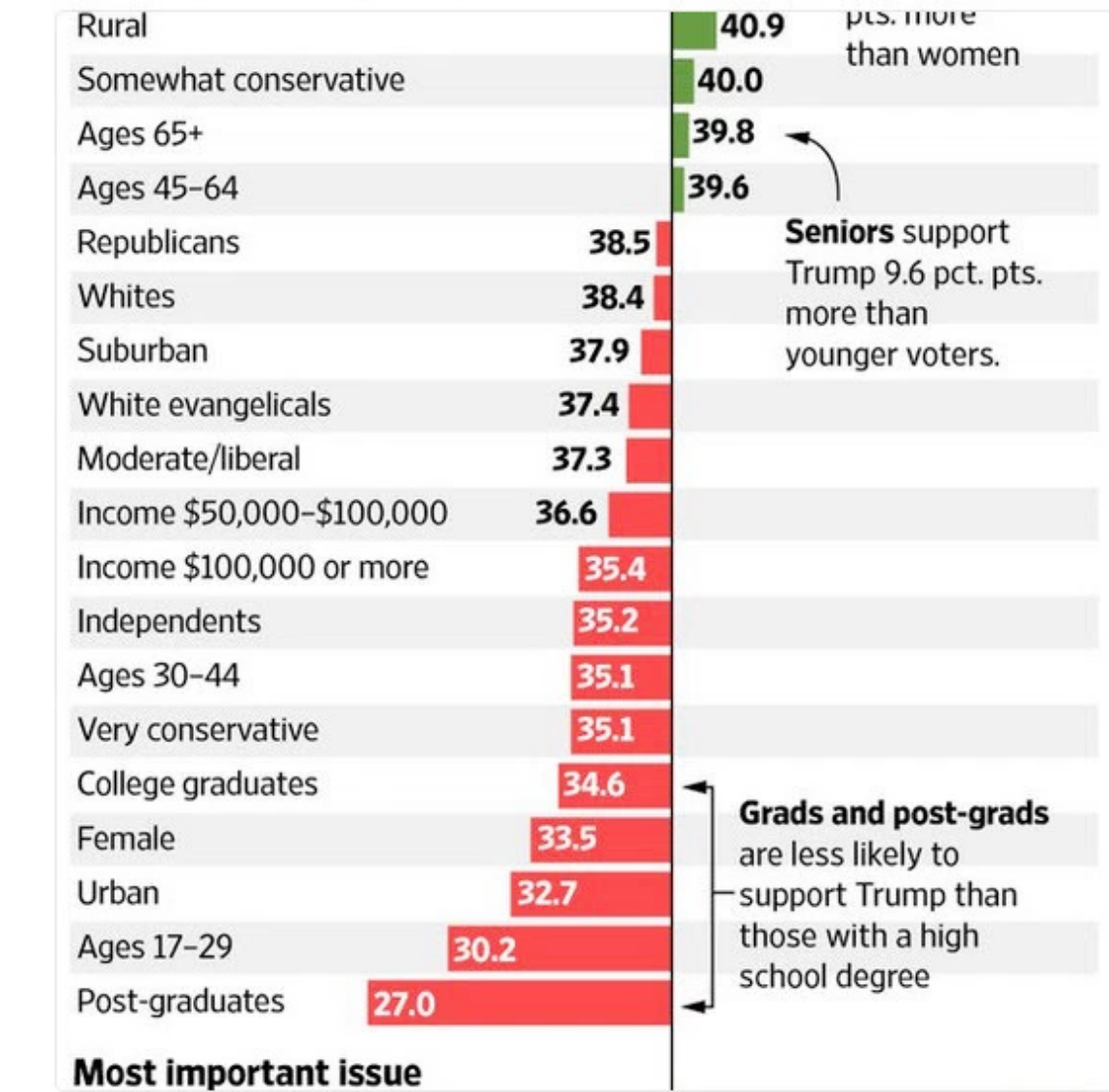
Party ideology in parliamentary elections* Center-left, center-right Other parties Right-wing and far-right



Measles



Inside the Trump coalition: How he performs among voter groups

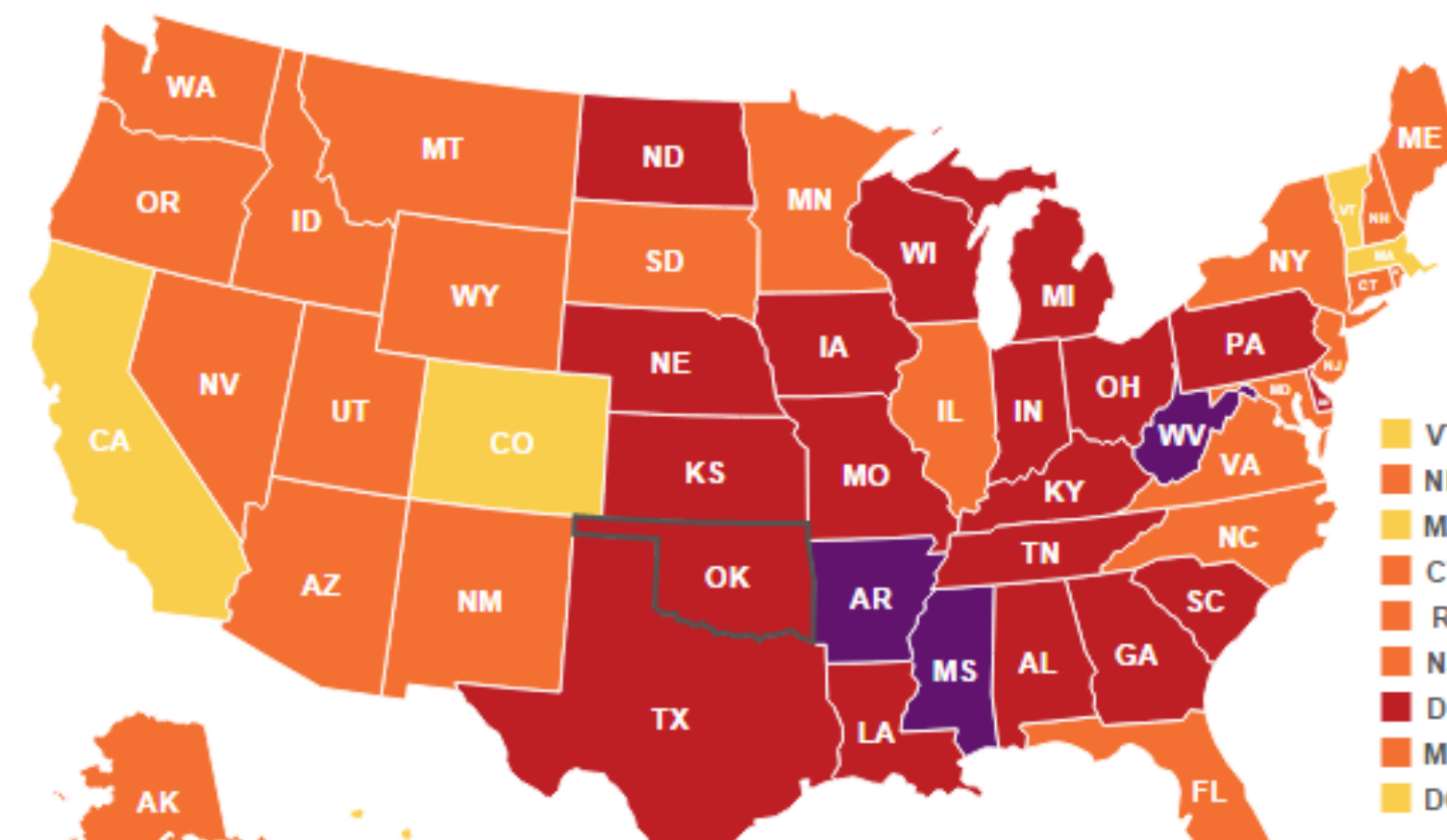


Adult Obesity Rate by State, 2014

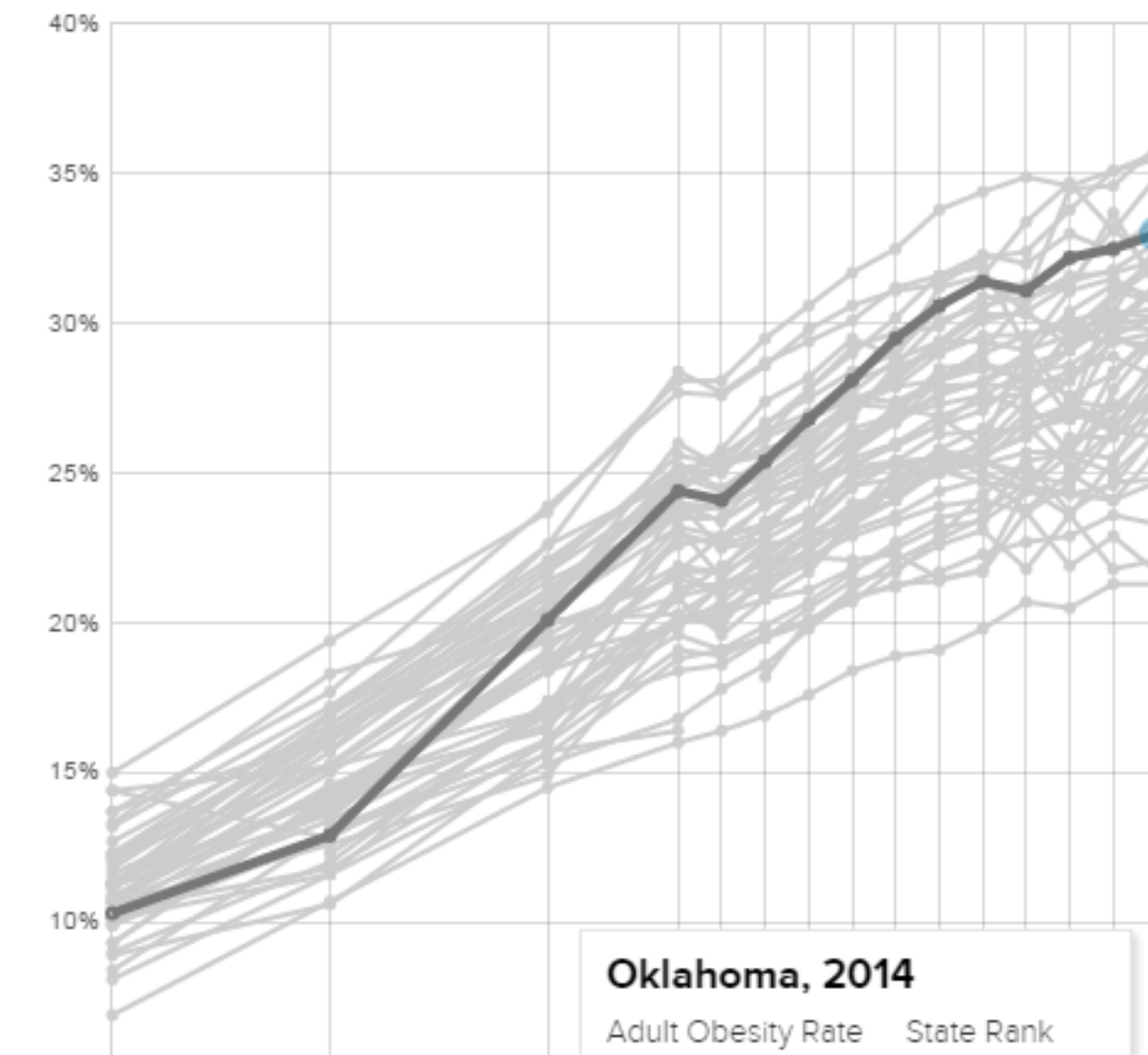
Select years with the slider to see historical data. Hover over states for more information. Click a state to lock the selection. Click again to unlock.

Percent of obese adults (Body Mass Index of 30+)

0 - 9.9% 10 - 14.9% 15 - 19.9% 20 - 24.9% 25 - 29.9% 30 - 34.9% 35%+



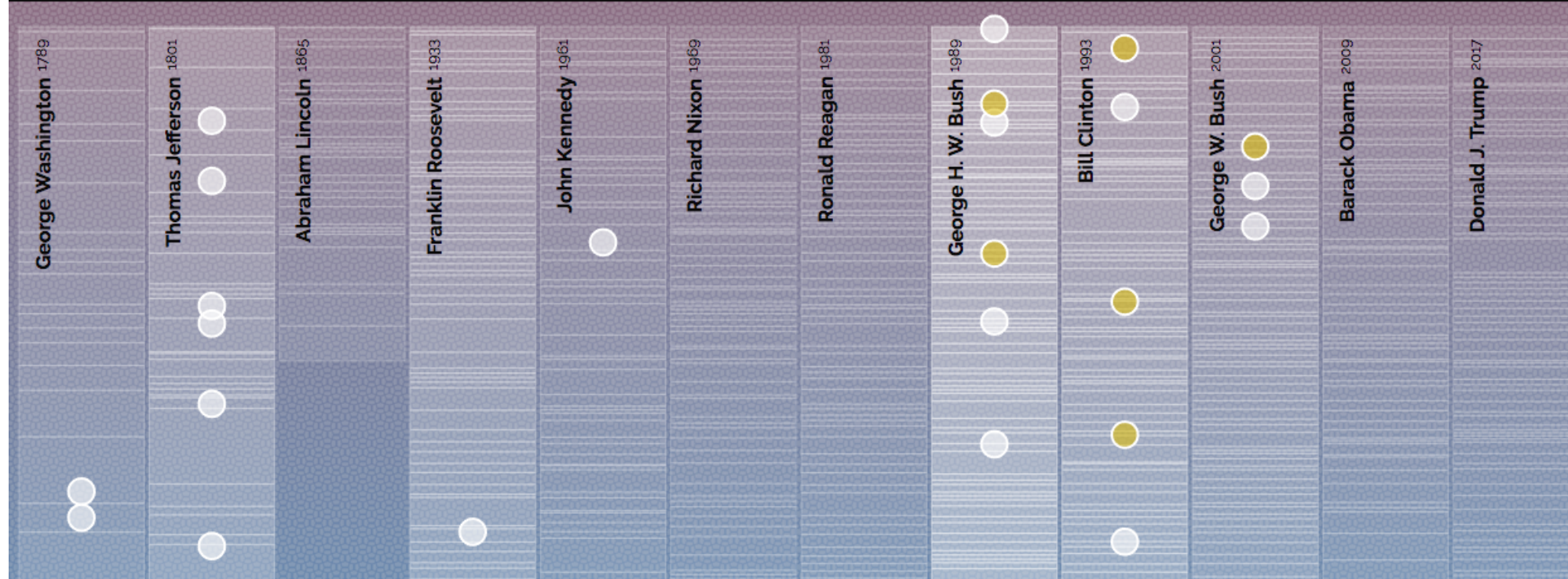
Adult obesity rates, 1990 to 2014



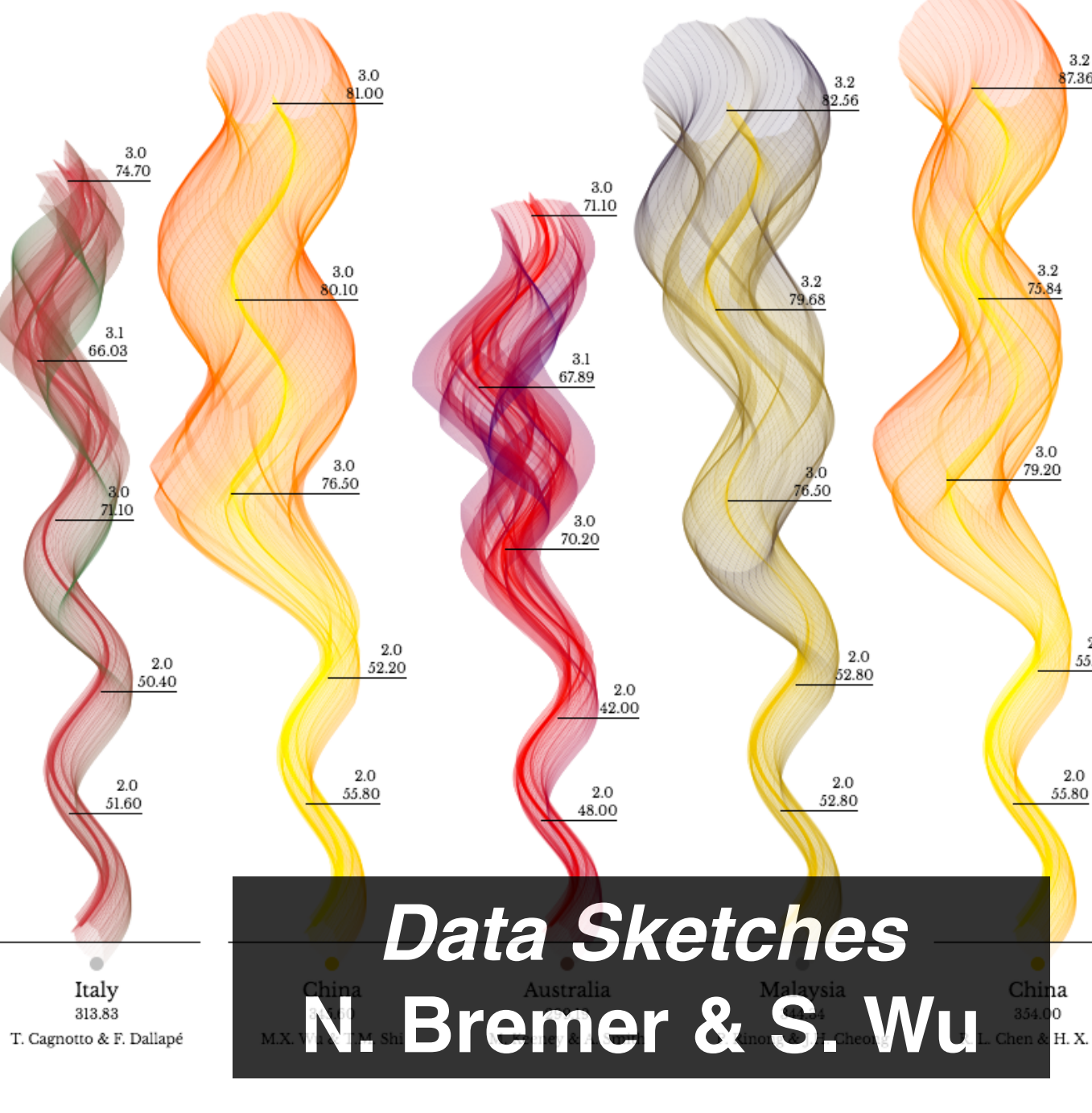
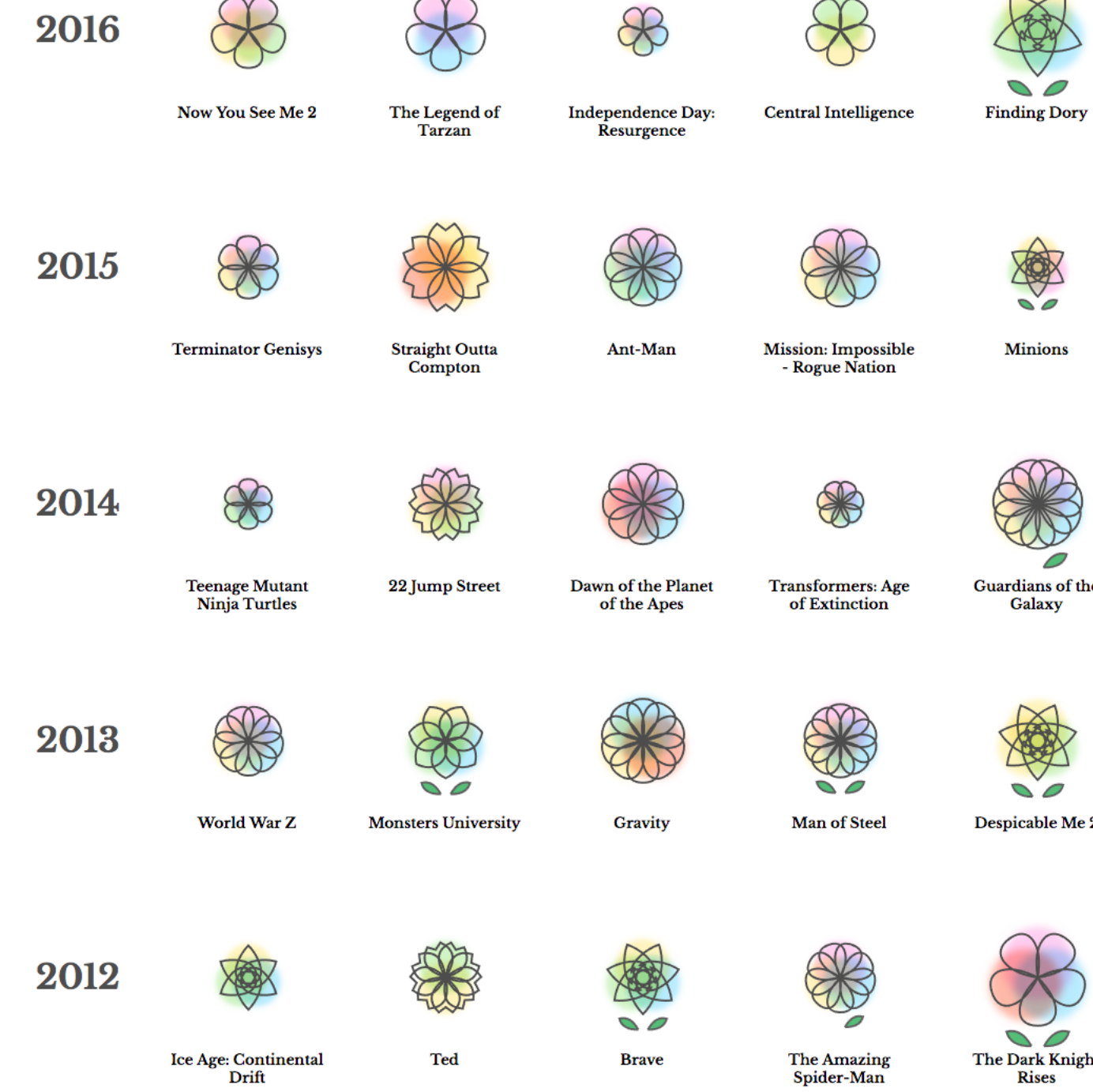
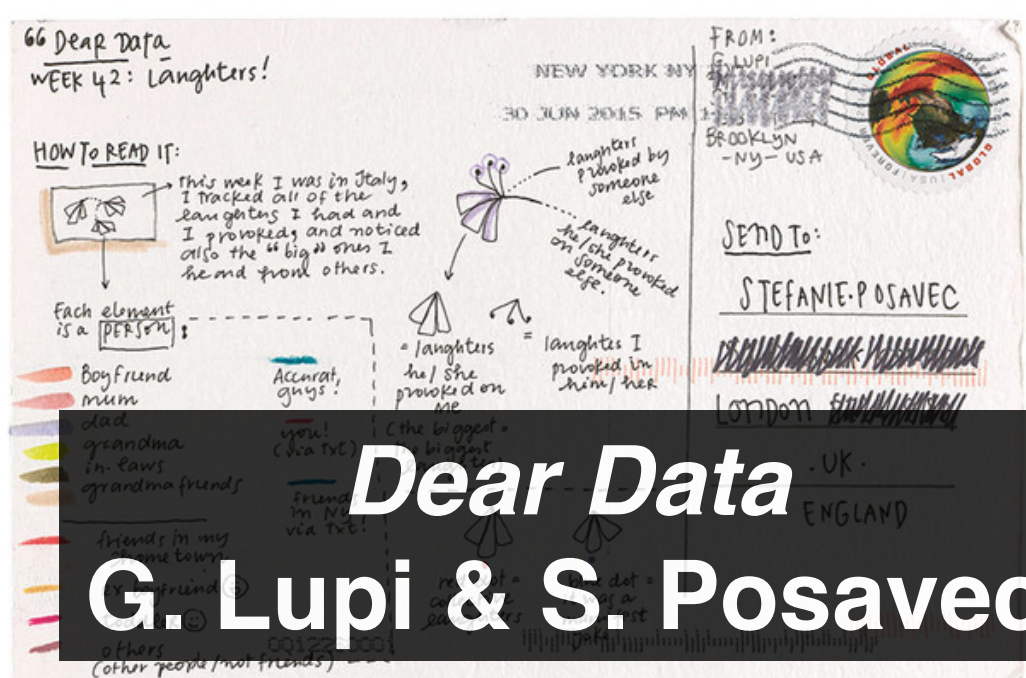
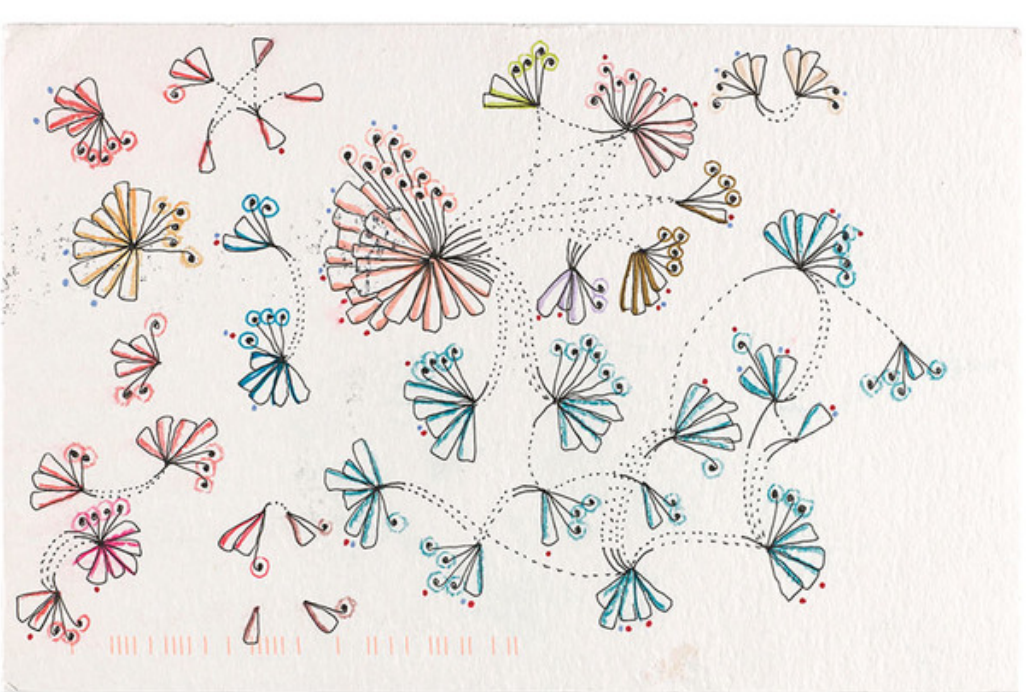
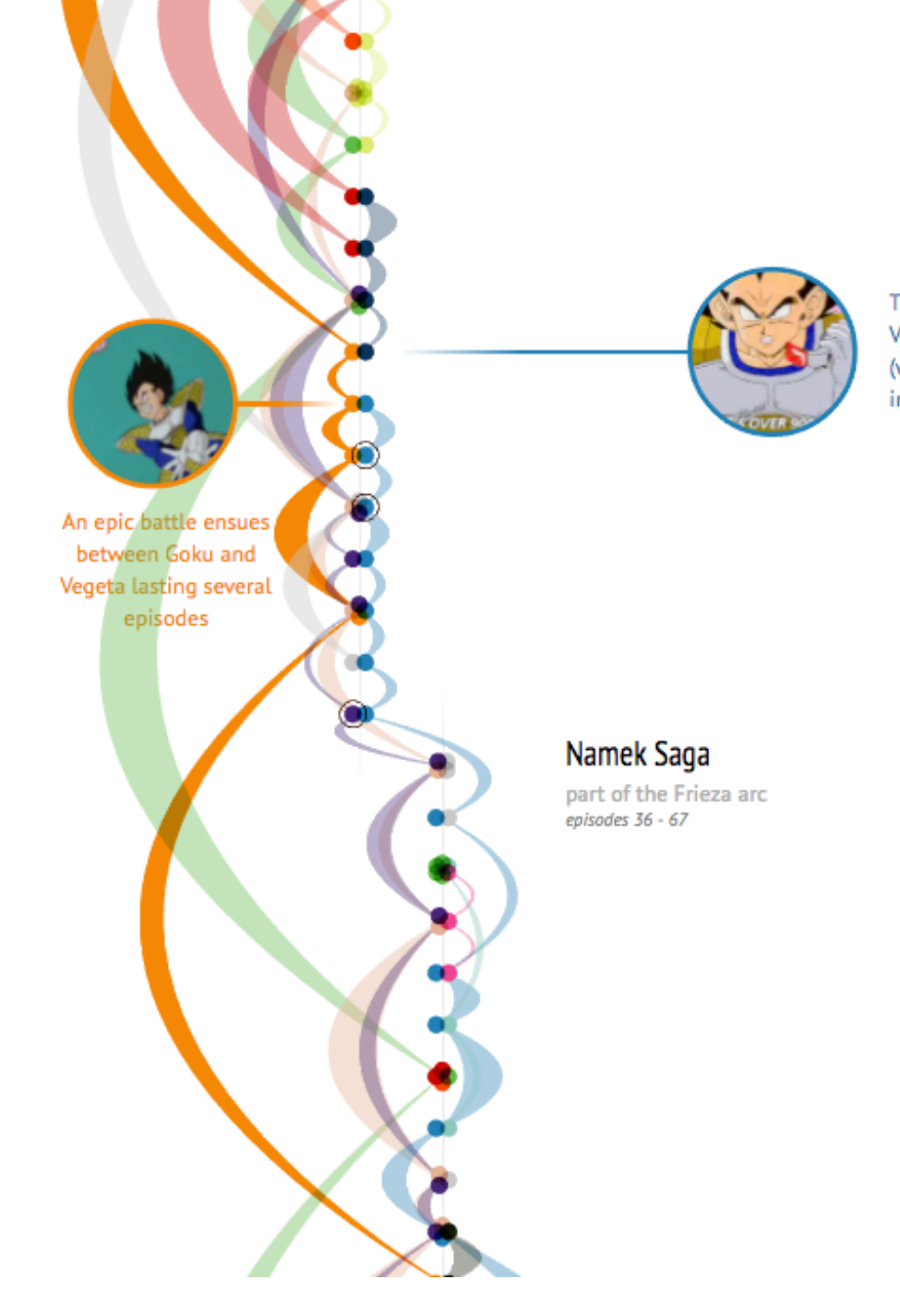
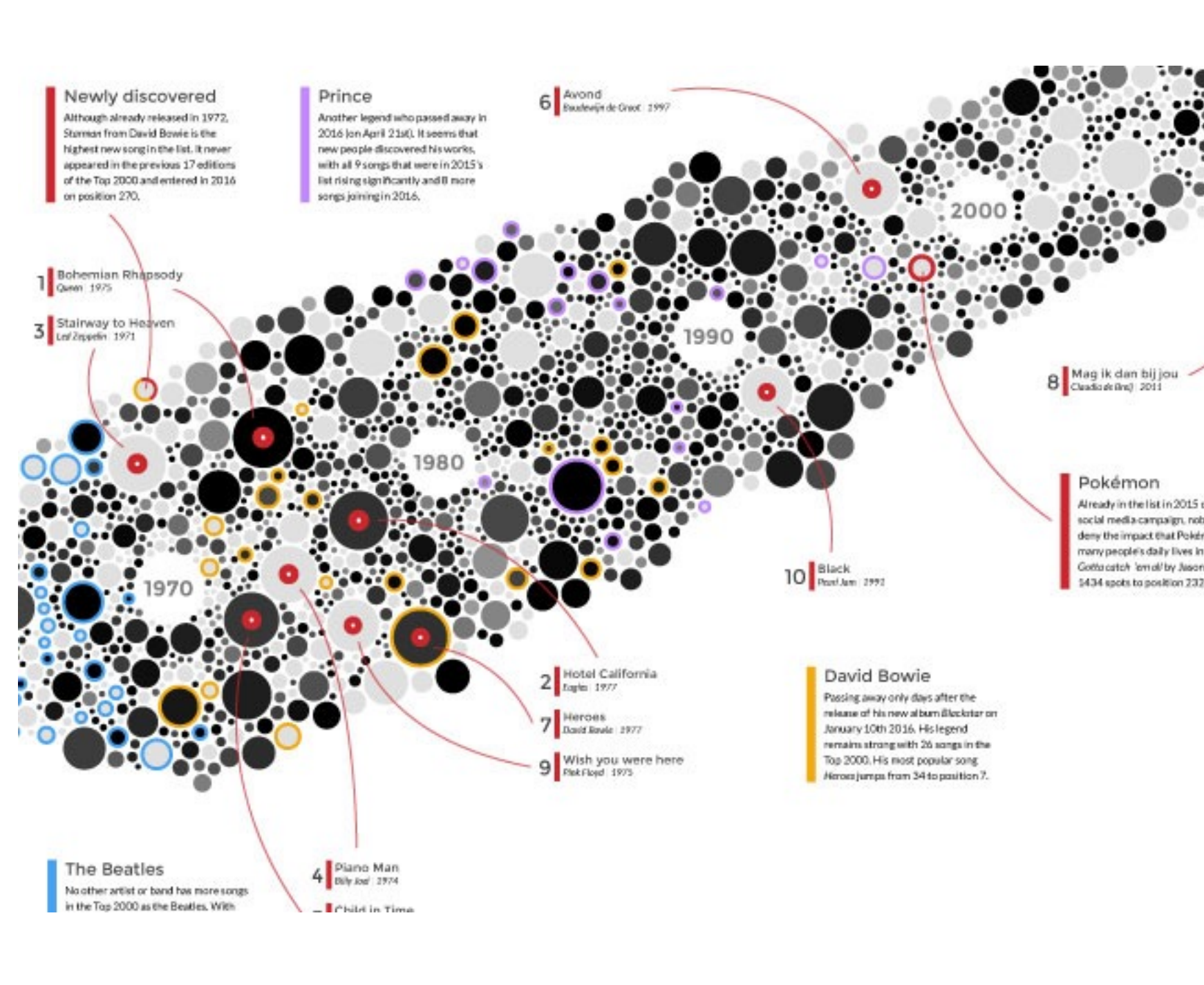
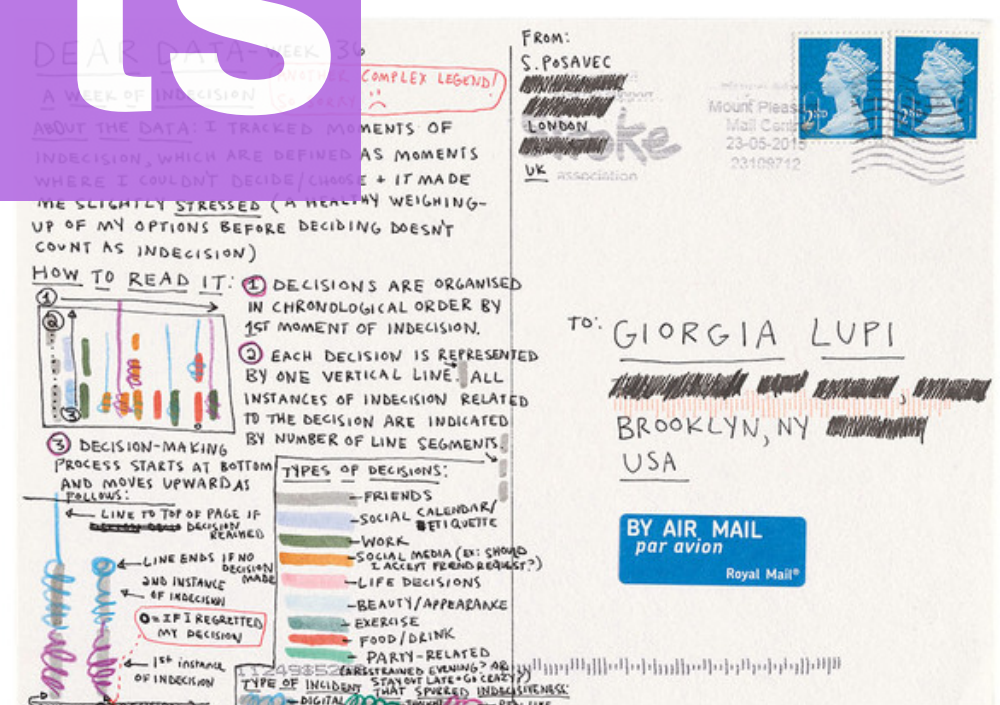
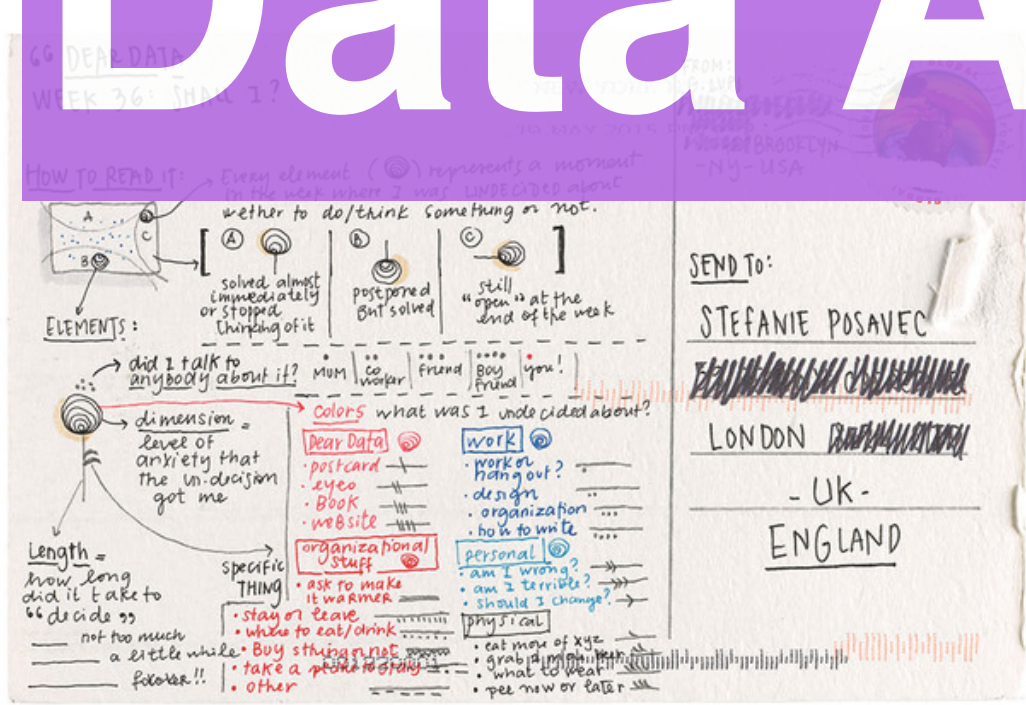
average U.S. search interest in 2016 in Politics

high low

Justice Constitution Democracy President of the U.S. Communism Republic U.S. Congress

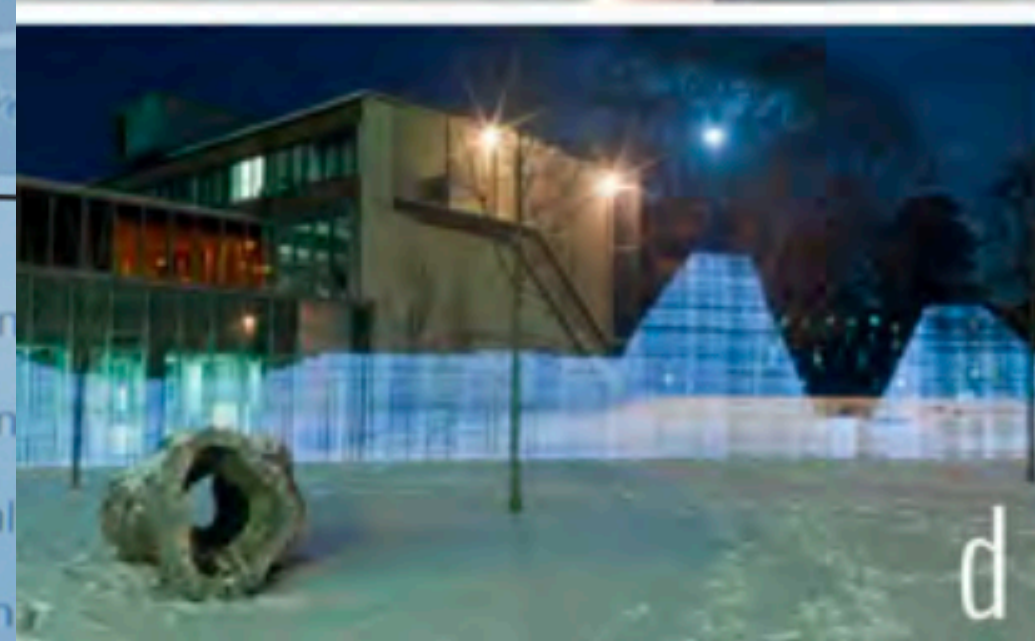
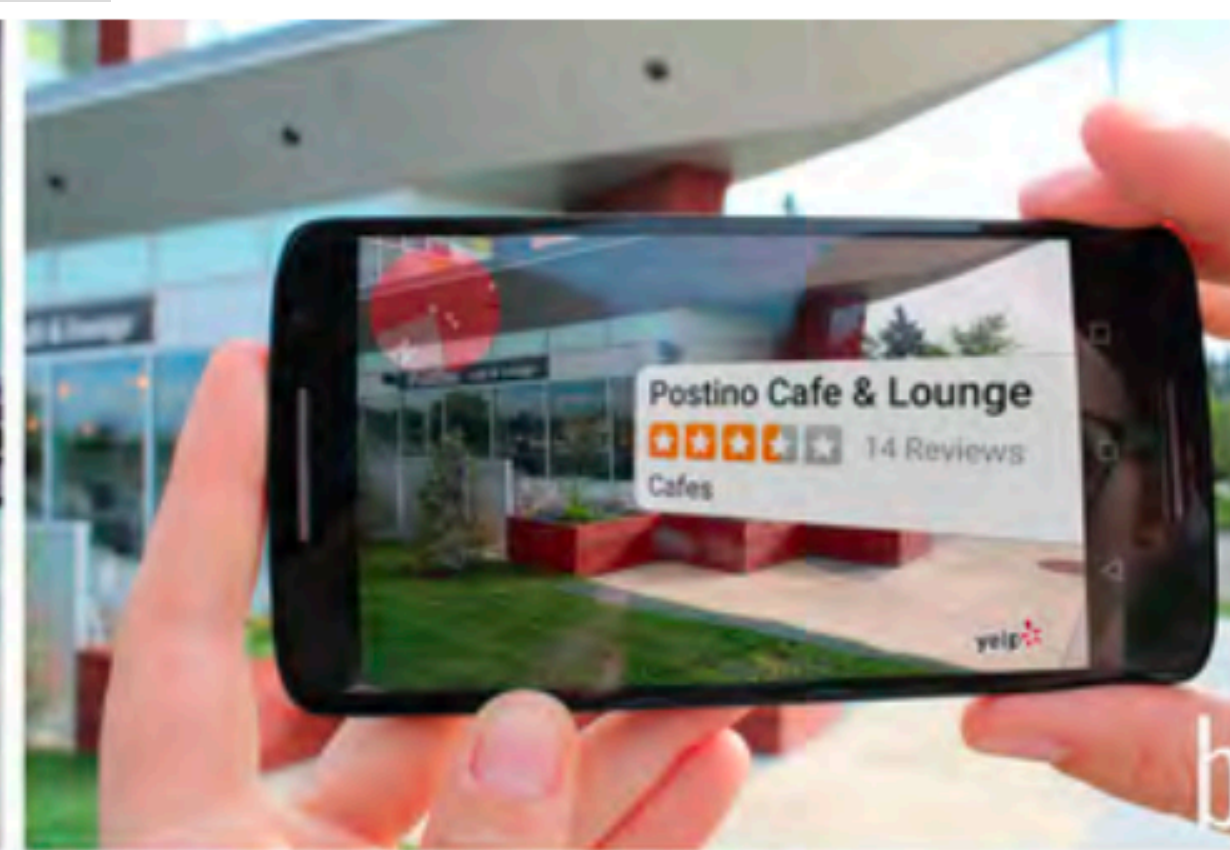
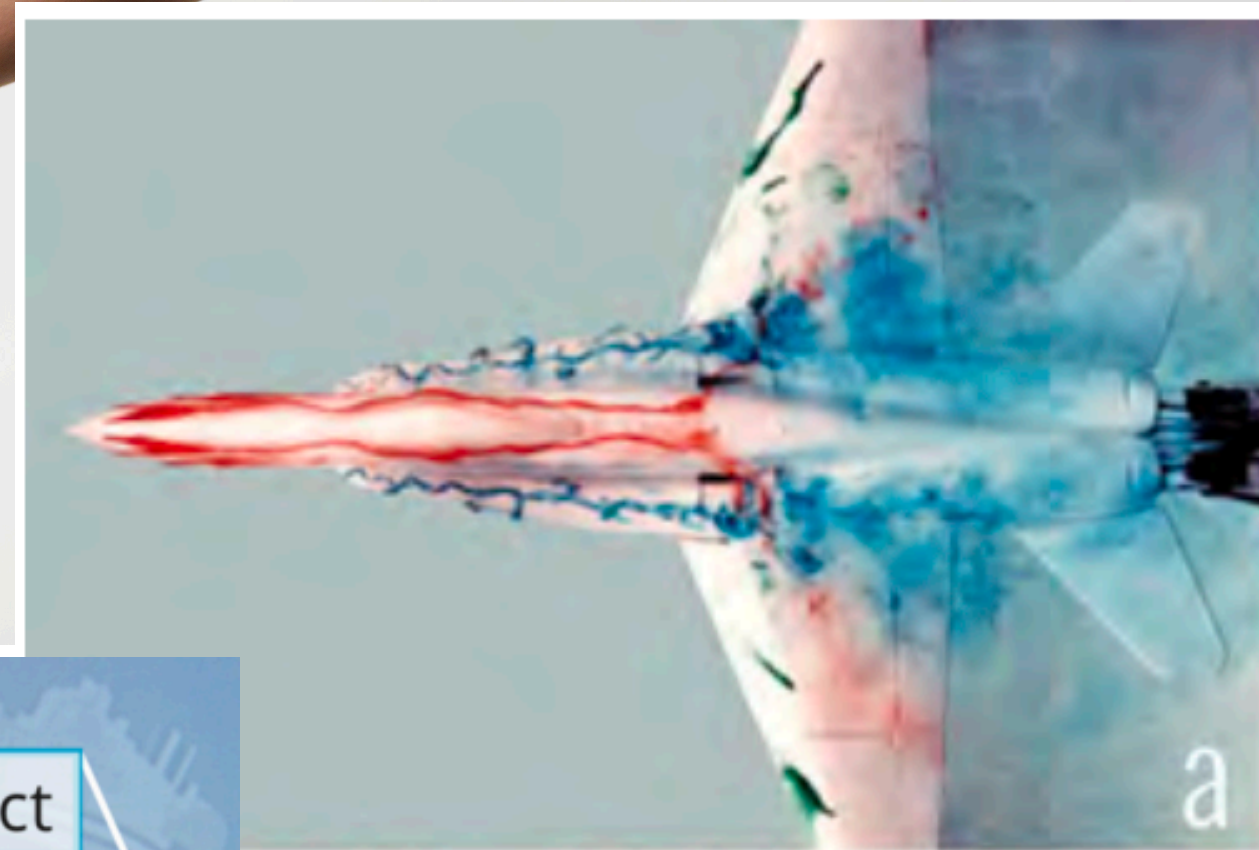


Data Arts



Data Sketches N. Bremer & S. Wu

New Environments



Data ↔ Insights ↔ Messages

Exploratory

- Data Centered
- Domain Experts
- Analysis
- Desktop
- In-Lab

Beyond Exploratory

- Human Centered
- General Audience
- Communication
- Off-Desktop
- In-the-Wild

Topics

- Visualization for communication
- Telling compelling stories with data
- Story Points and Dashboards in Tableau

Visualization for Communication

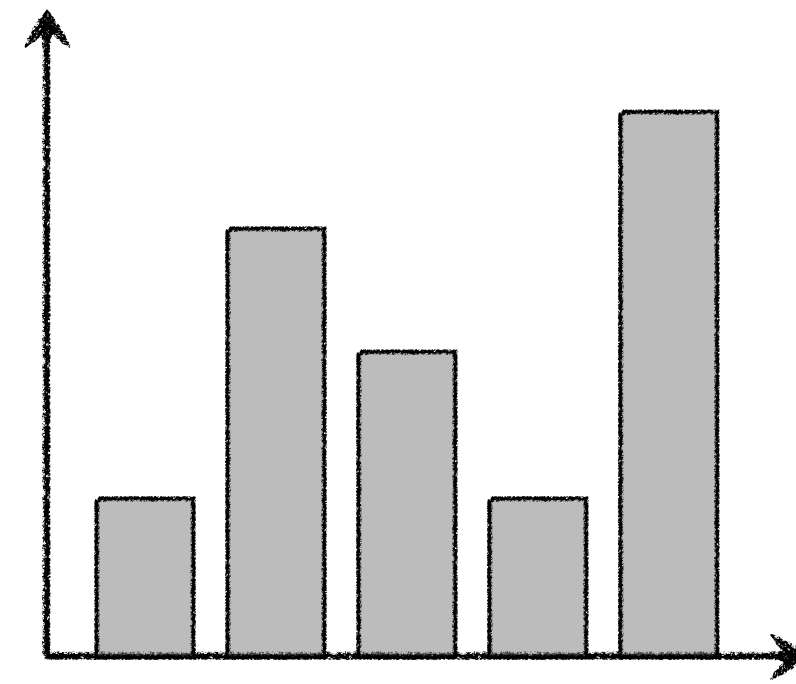
Visualization for Exploration

- Faster and accurate reading
- Rapidly generate many visualizations
- No titles, annotations, embellishments

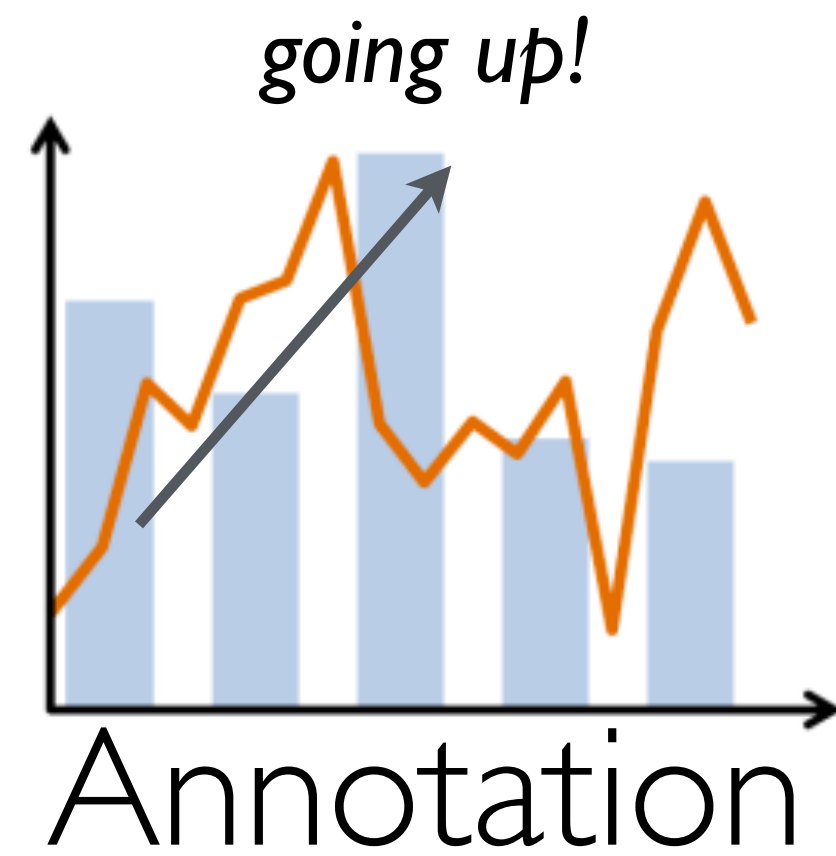
Visualization for Communication

- Good headlines
- Annotations and highlights
- Redundant encodings
- Pictograms and useful embellishments
- Explanations (e.g., legend, source)

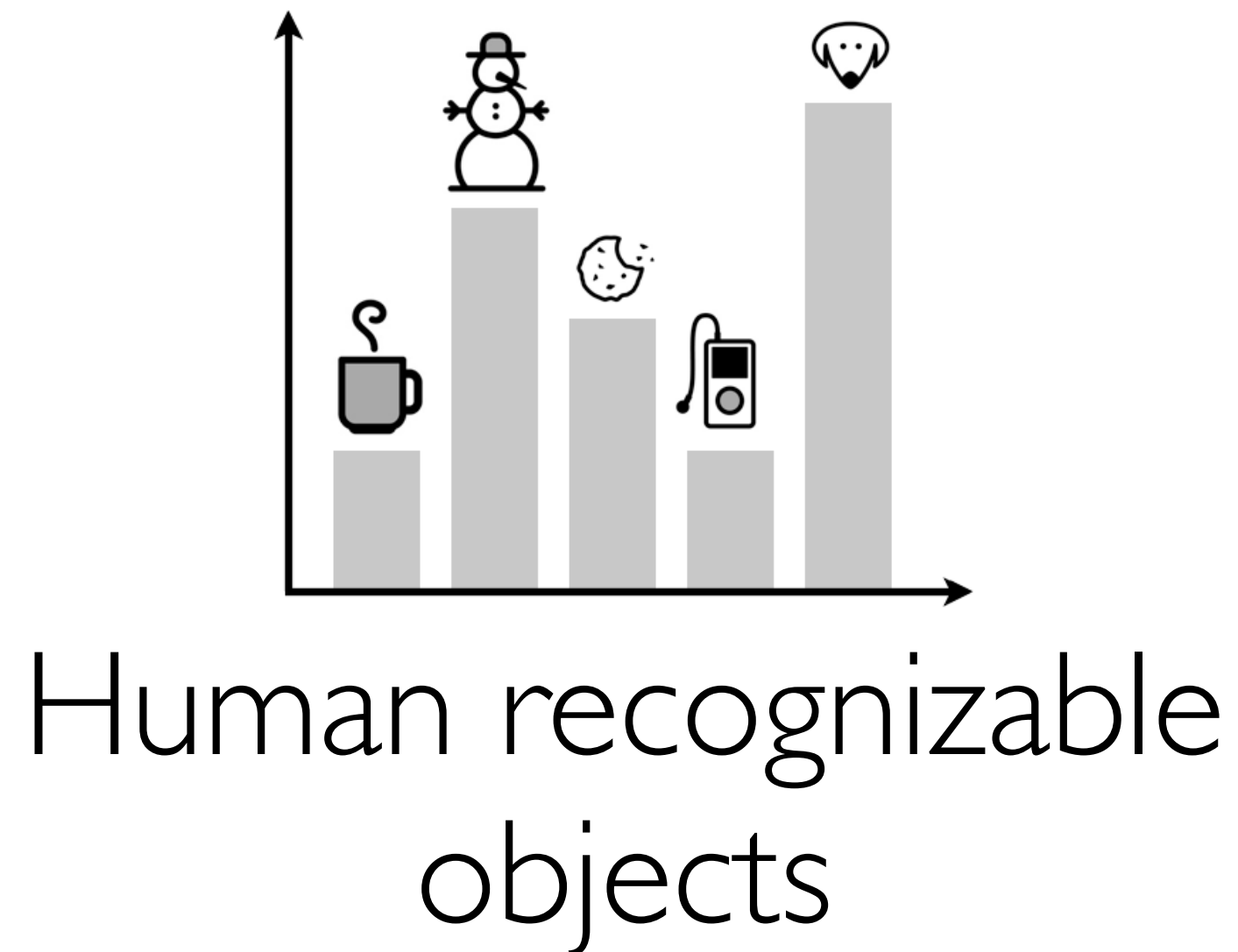
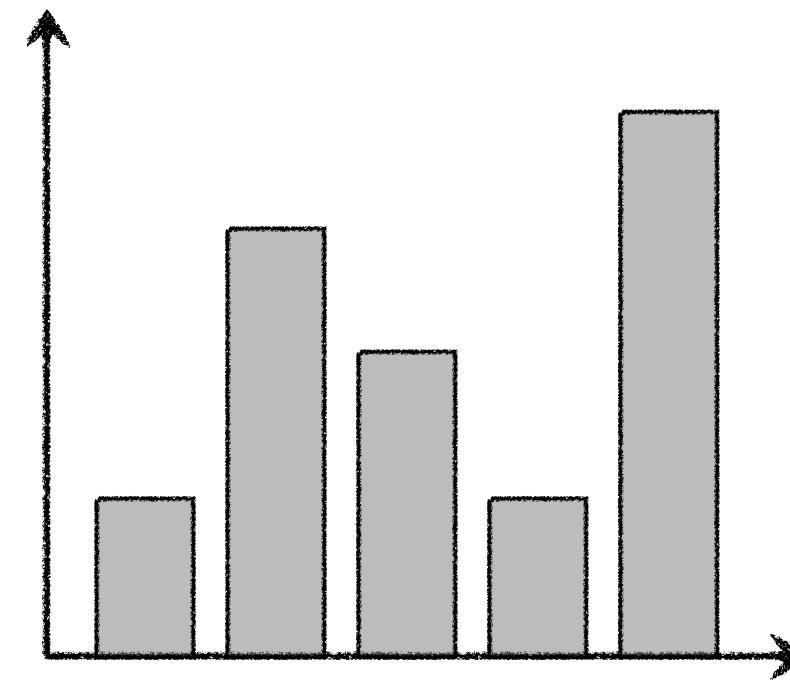
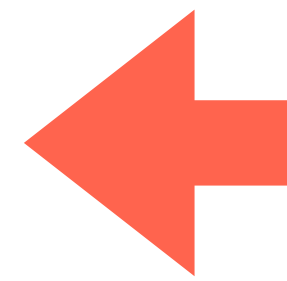
Examples of Redundant Encoding



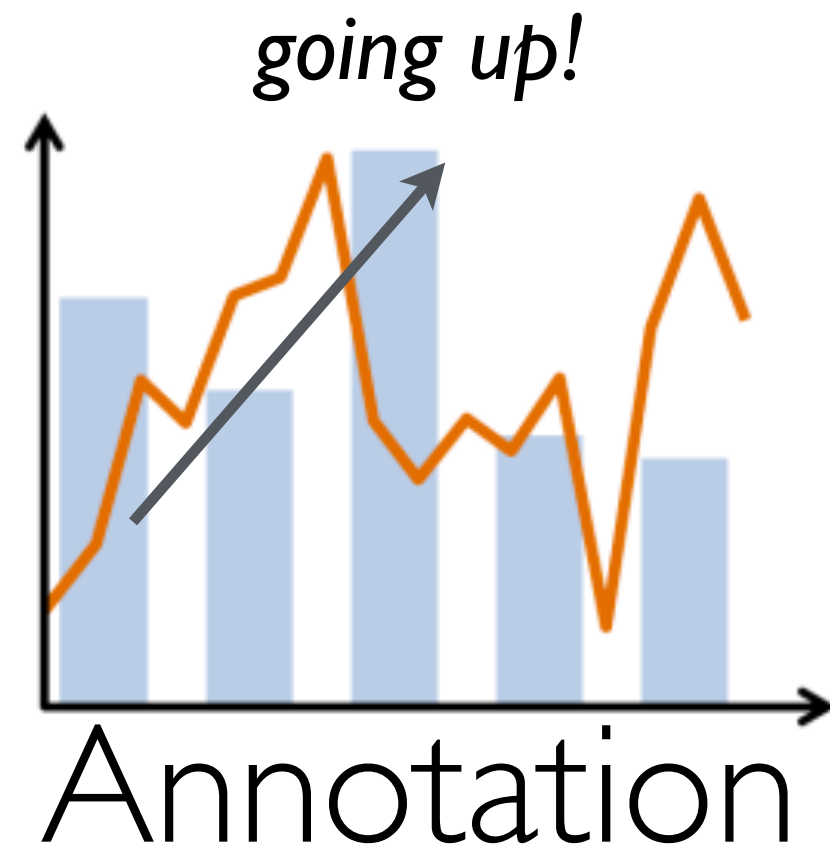
Examples of Redundant Encoding



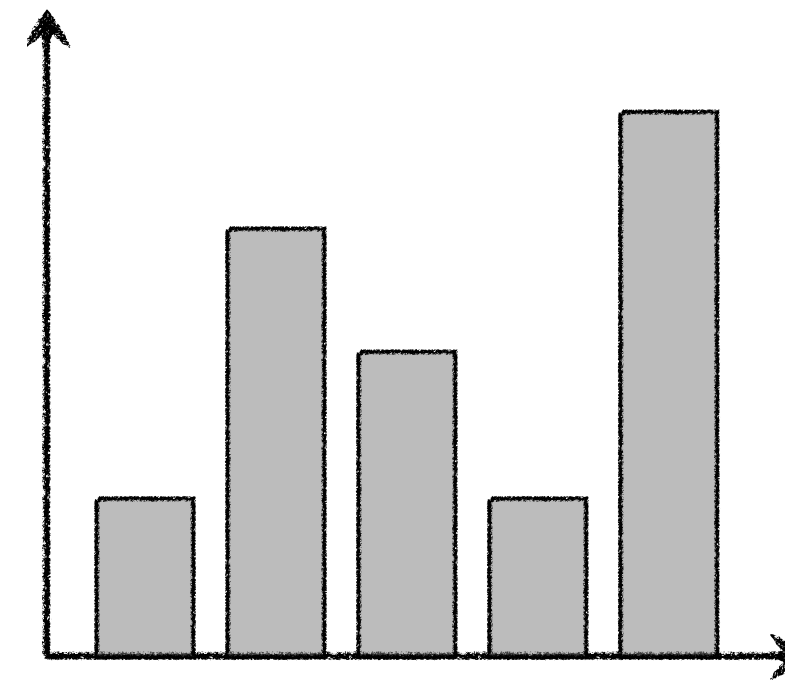
**Message
Redundancy**



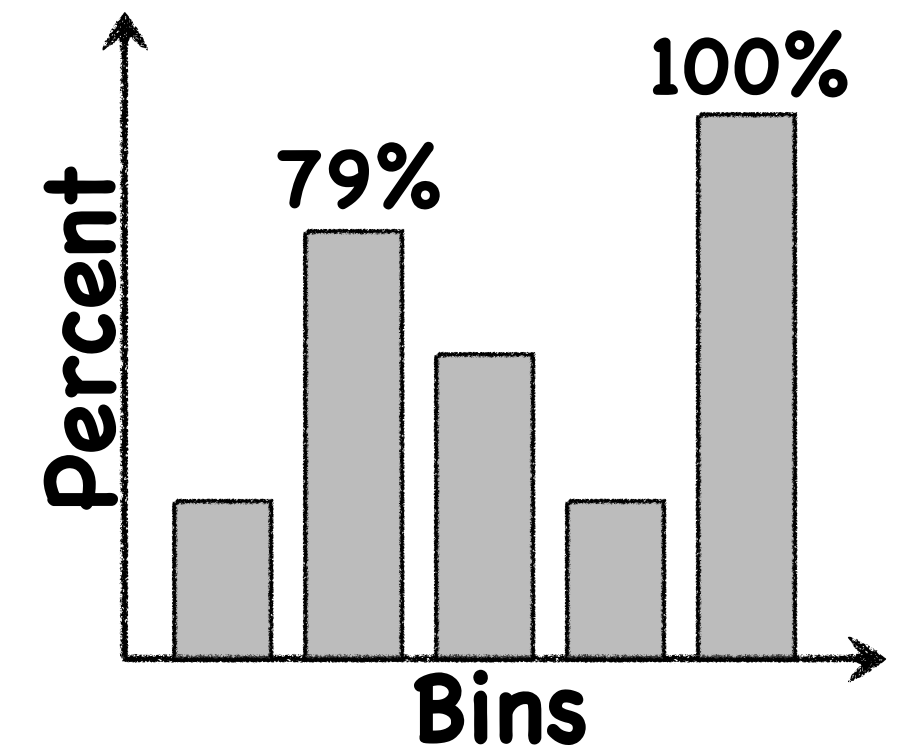
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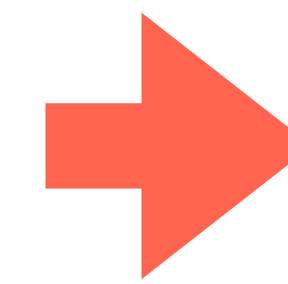
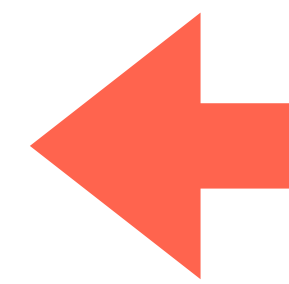
Message Redundancy



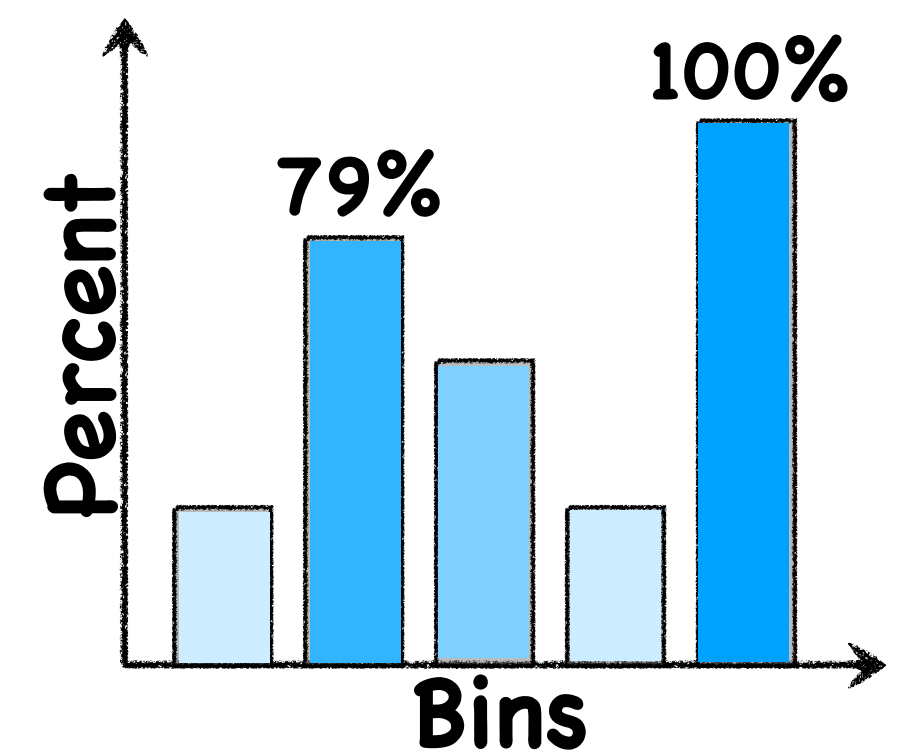
Data Redundancy



Data Labels



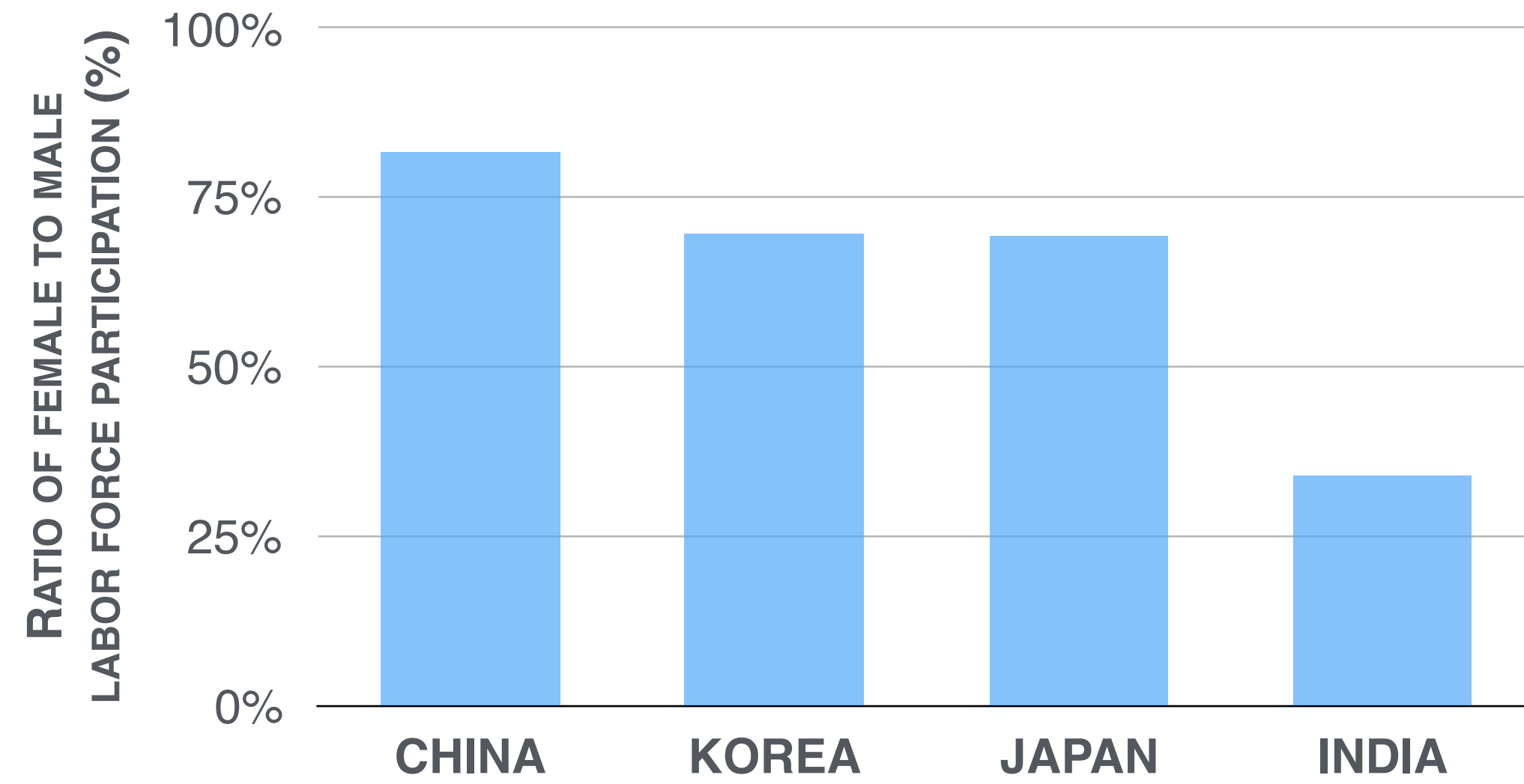
Human recognizable objects



Lightness

Examples

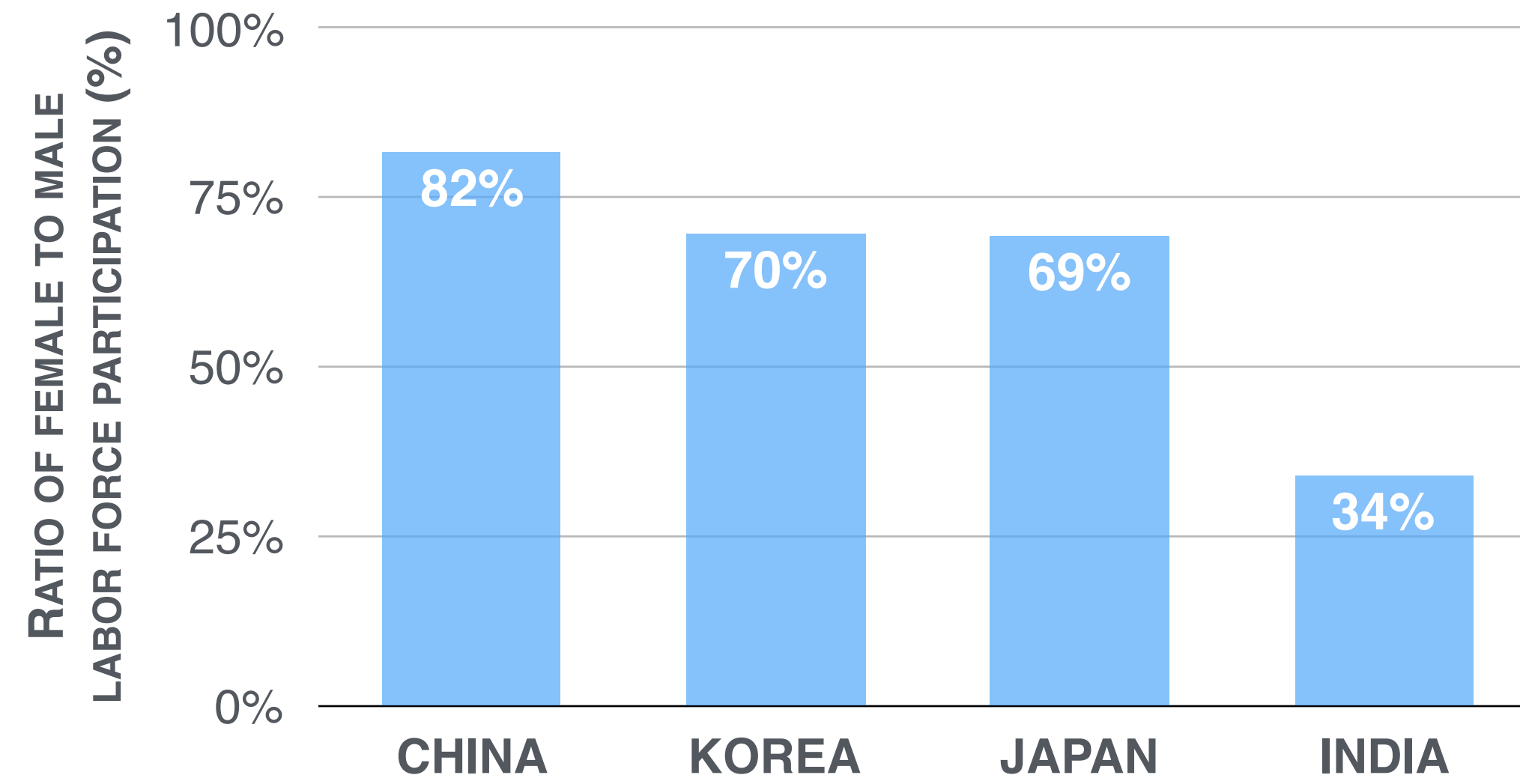
GENDER EQUALITY IN LABOR FORCE PARTICIPATION



Source: Gender Statistics 2013, World Bank

ORIGINAL

GENDER EQUALITY IN LABOR FORCE PARTICIPATION

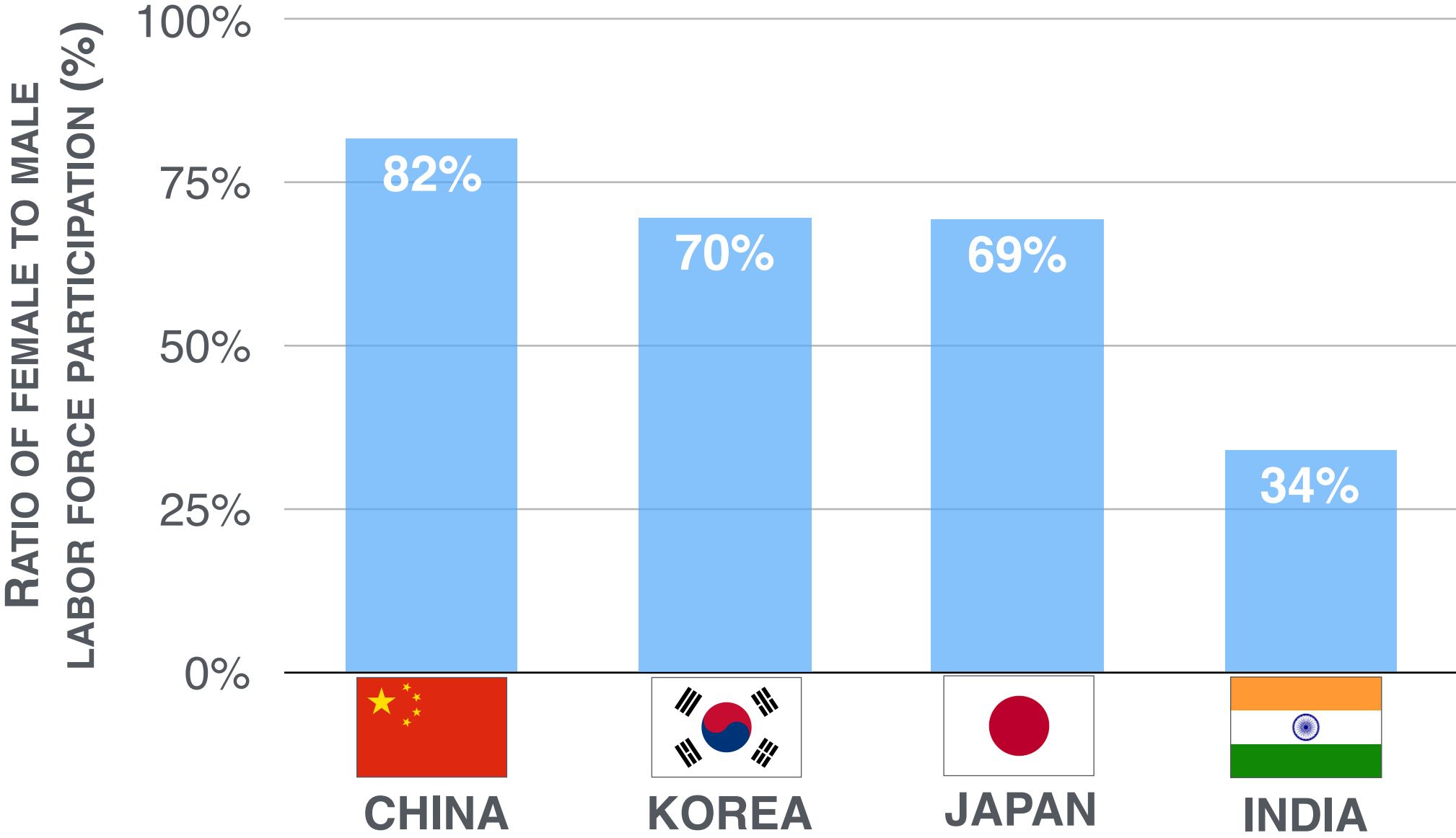


Source: Gender Statistics 2013, World Bank

DATA REDUNDANCY

GENDER EQUALITY IN LABOR FORCE PARTICIPATION

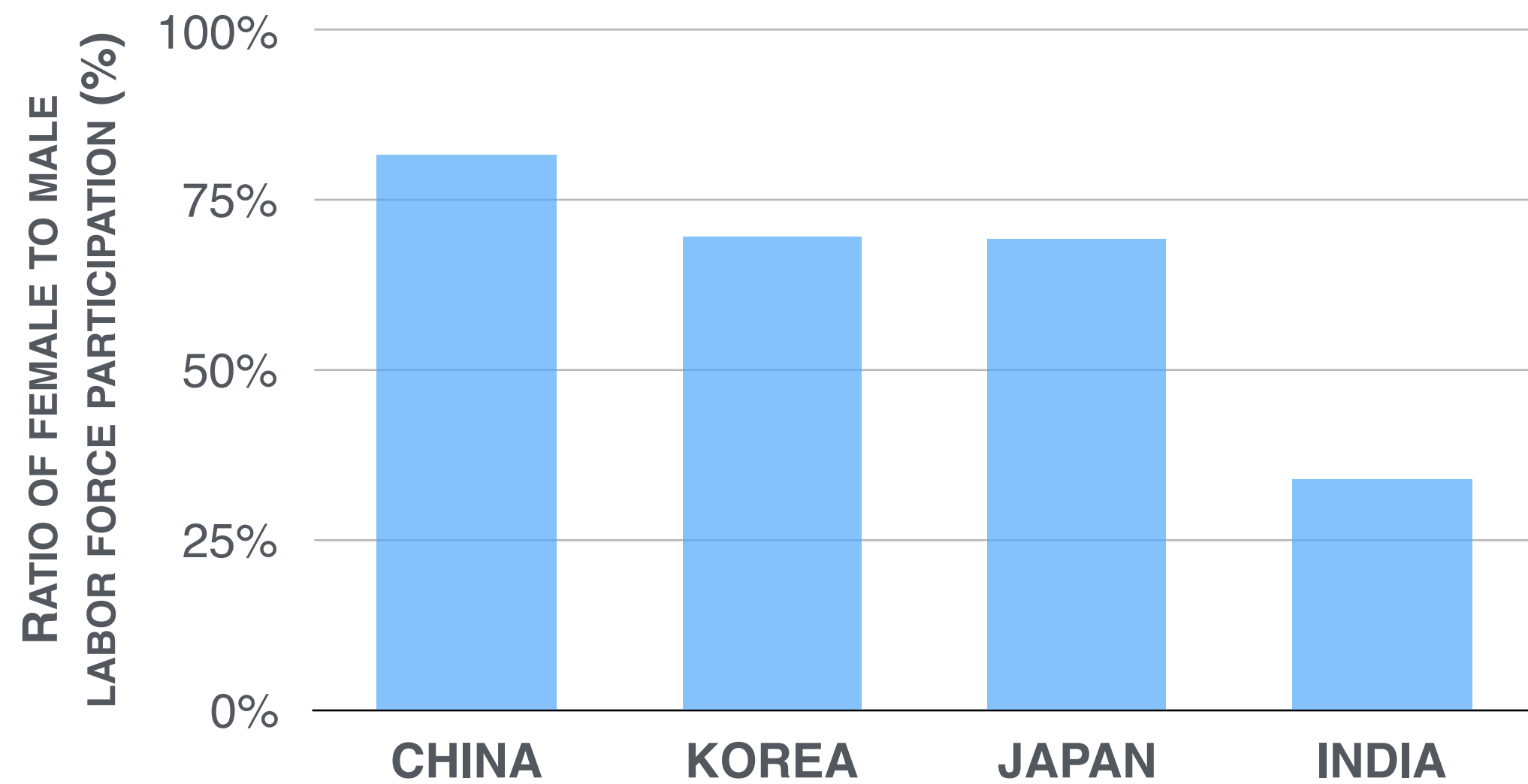
CHINA LEADS IN FEMALE LABOR FORCE PARTICIPATION WHEREAS INDIA LAGS SIGNIFICANTLY BEHIND AMONG OTHER ASIAN-PACIFIC COUNTRIES IN 2013.



Source: Gender Statistics 2013, World Bank

DATA & MESSAGE REDUNDANCY

GENDER EQUALITY IN LABOR FORCE PARTICIPATION

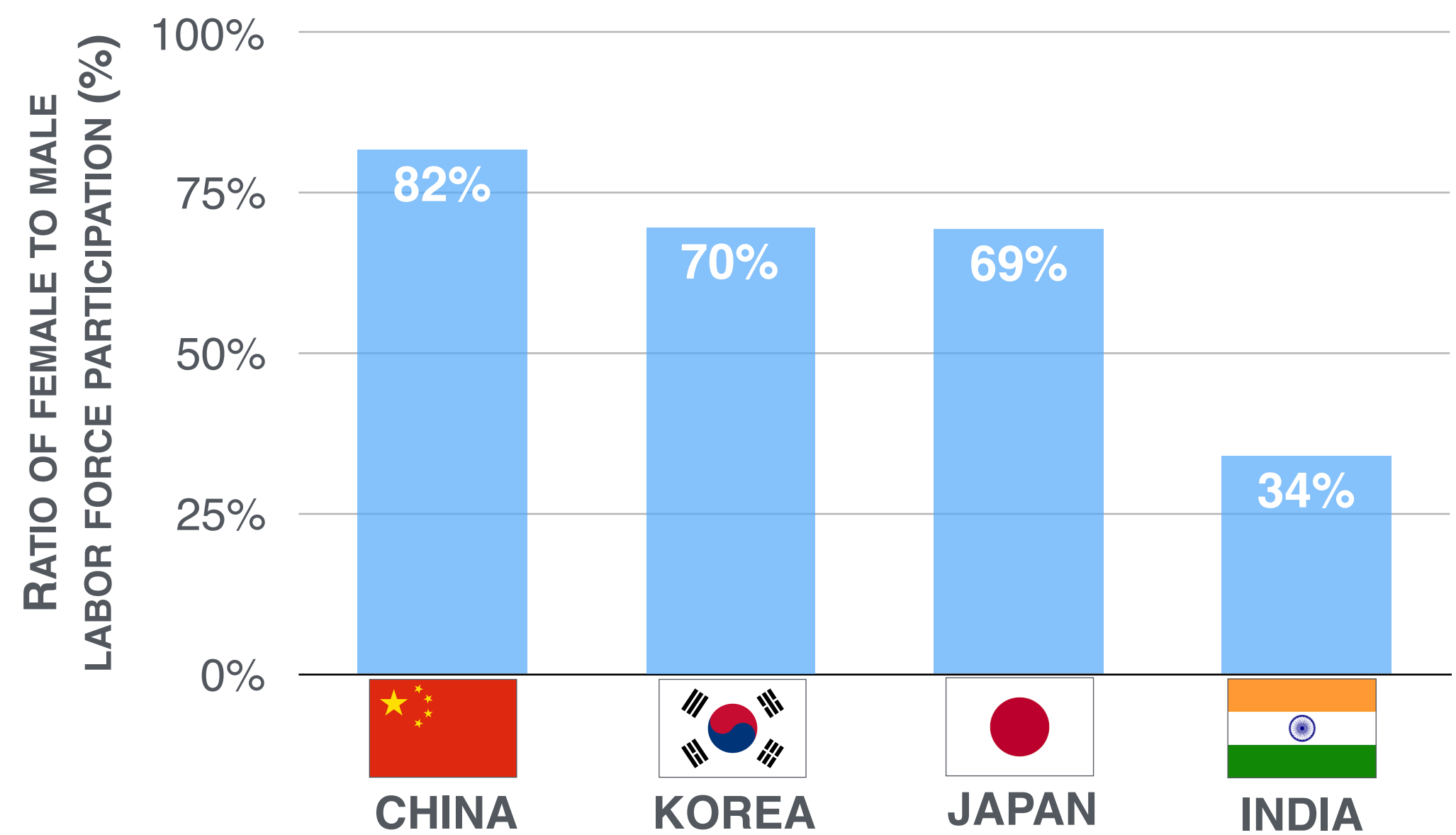


Source: Gender Statistics 2013, World Bank

ORIGINAL

GENDER EQUALITY IN LABOR FORCE PARTICIPATION

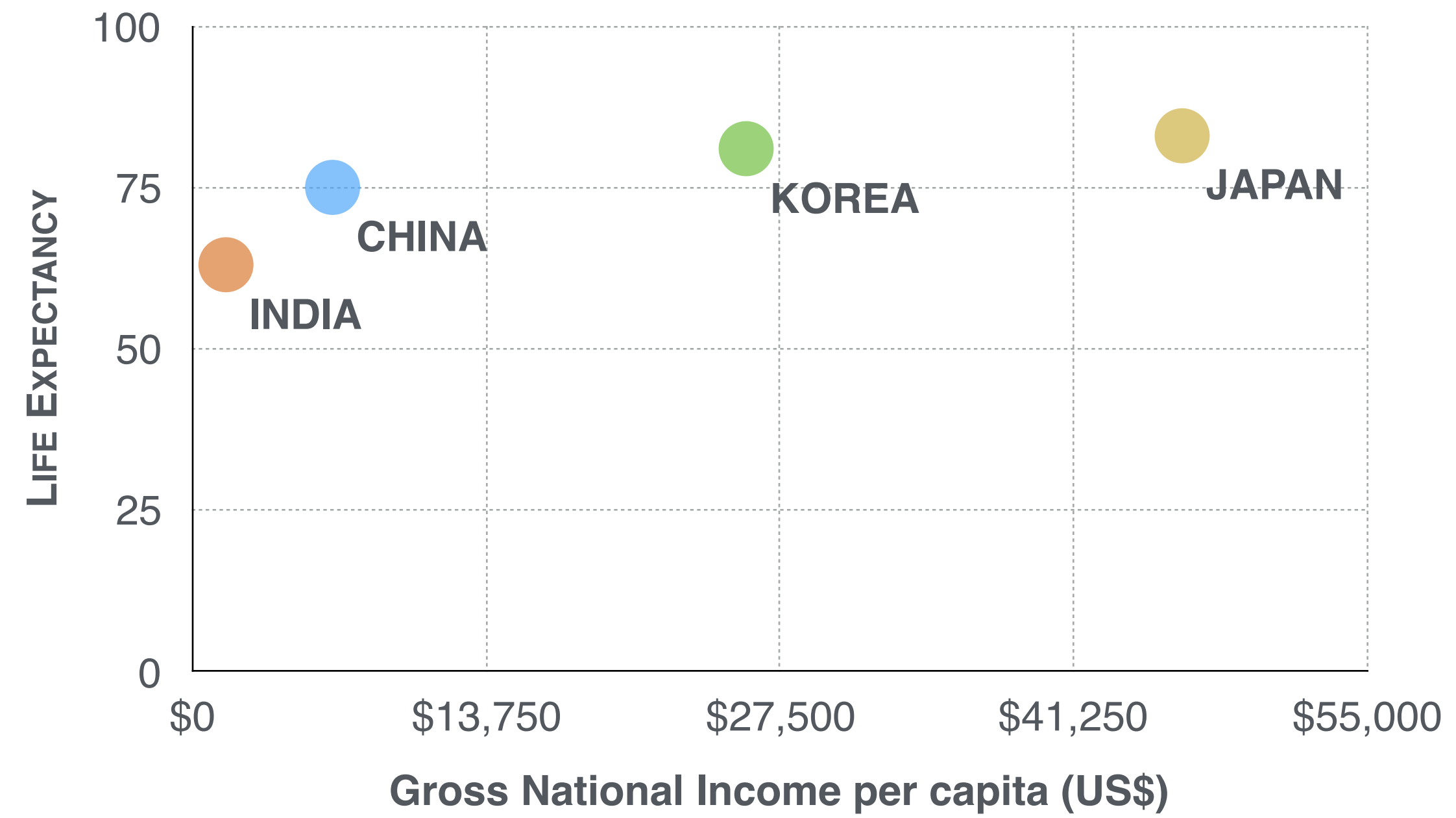
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DATA & MESSAGE REDUNDANCY

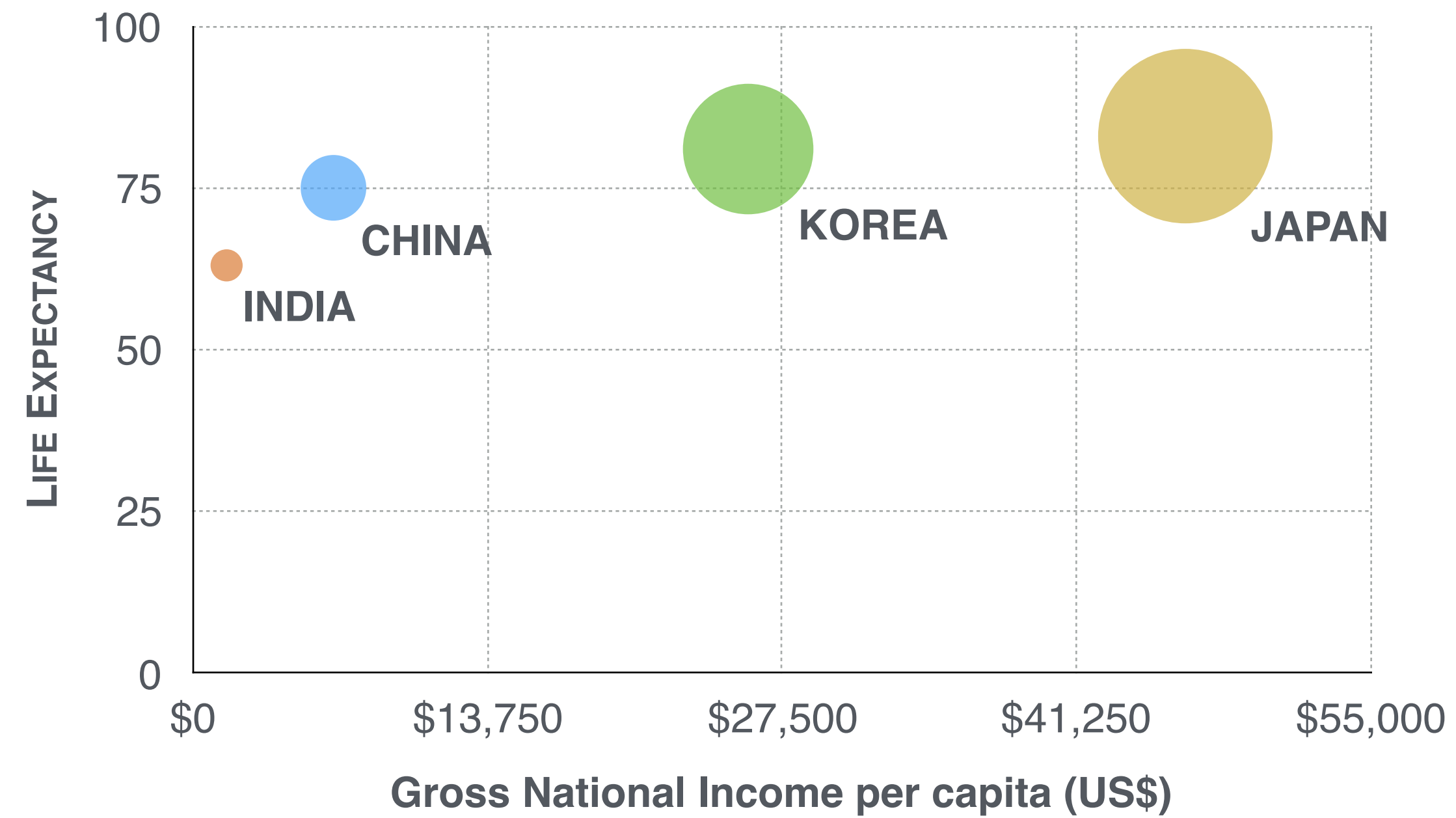
LIFE EXPECTANCY VS ECONOMIC GROWTH



Source: World Bank national accounts data, and
OECD National Accounts data files (2013).

ORIGINAL

LIFE EXPECTANCY VS ECONOMIC GROWTH

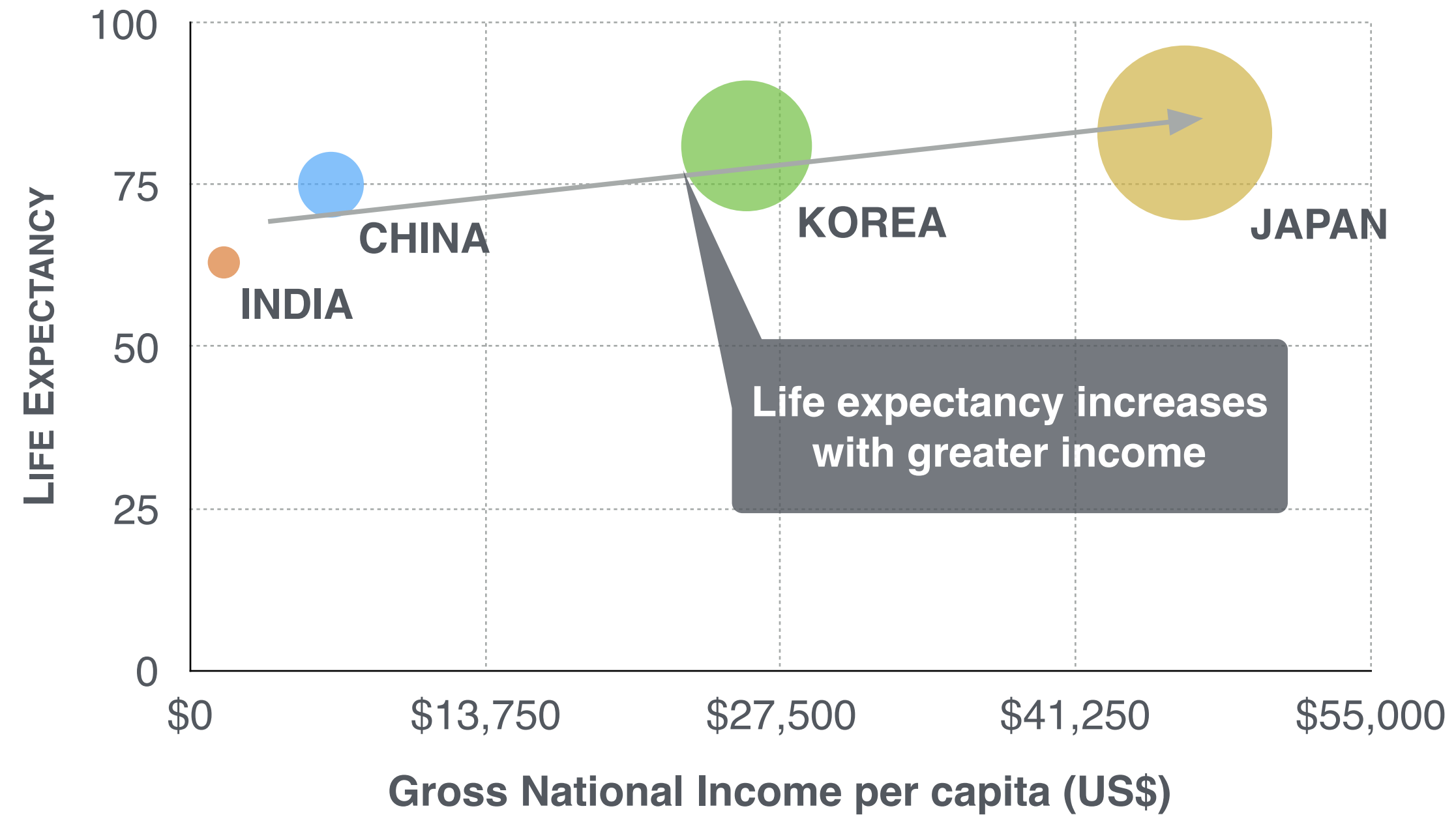


Source: World Bank national accounts data, and
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DATA REDUNDANCY

LIFE EXPECTANCY VS ECONOMIC GROWTH

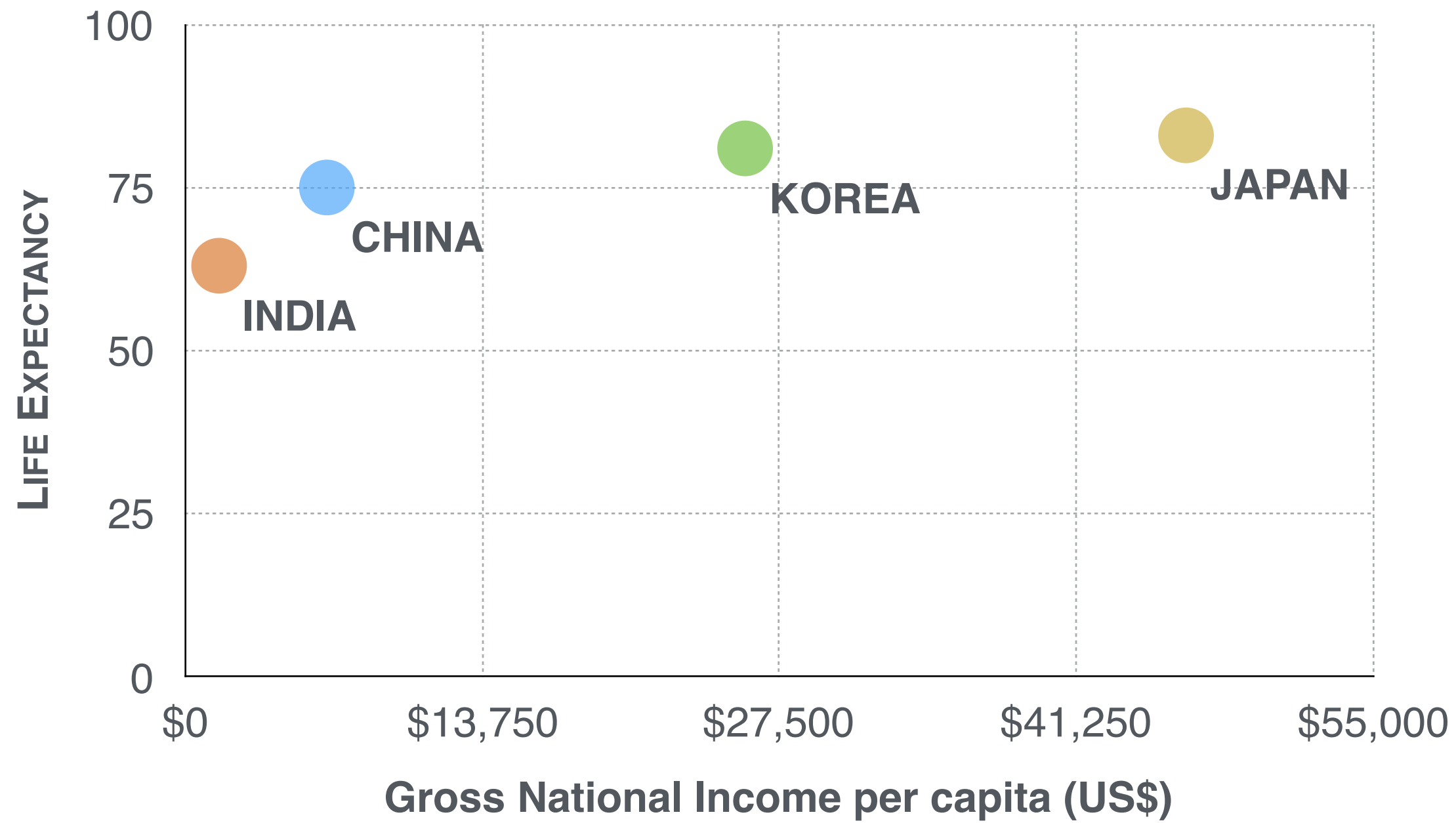
RELATION BETWEEN LIFE EXPECTANCY AND GROSS NATIONAL INCOME OF ASIAN-PACIFIC COUNTRIES IN 2013



Source: World Bank national accounts data, and OECD National Accounts data files (2013).

DATA & MESSAGE REDUNDANCY

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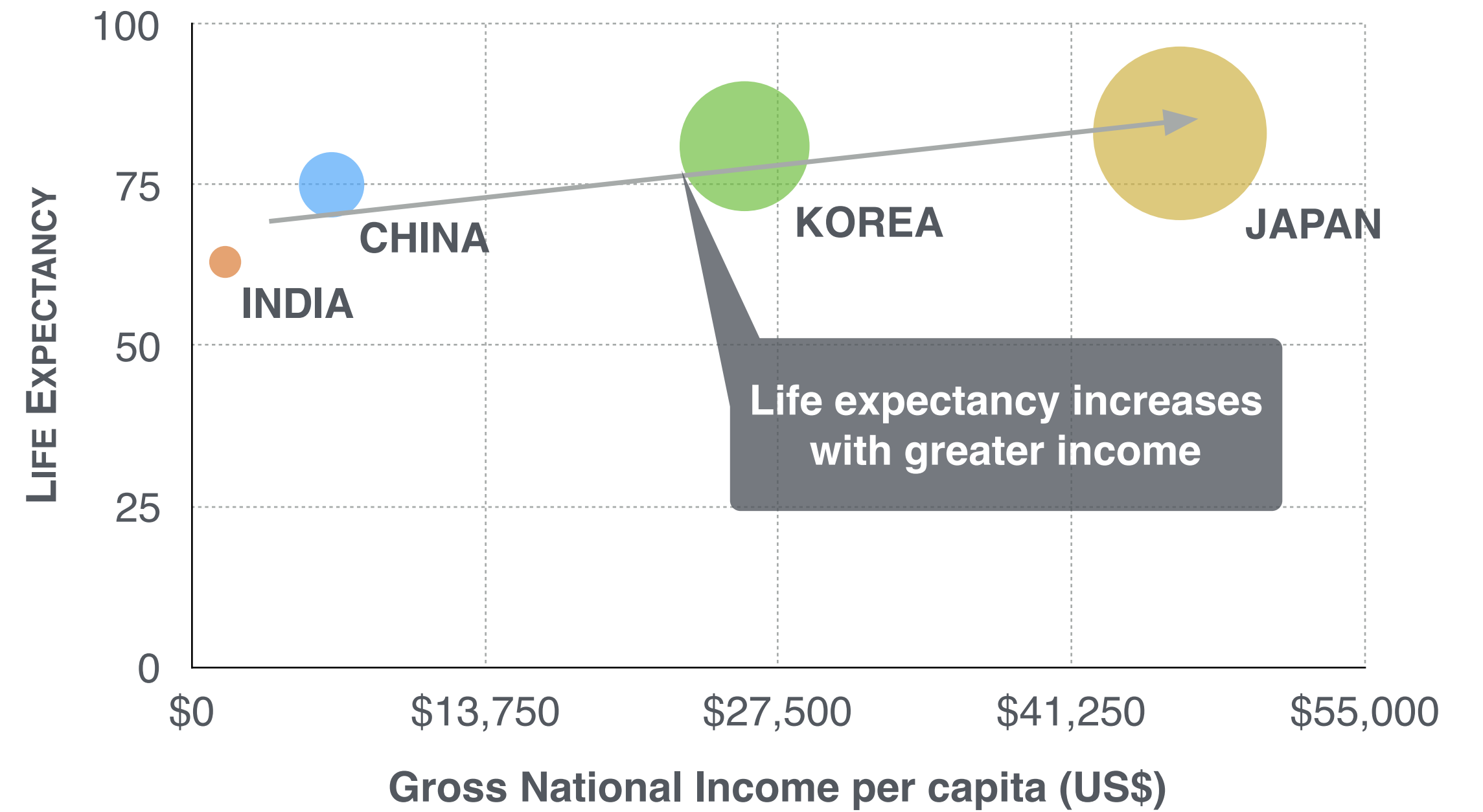


Source: World Bank national accounts data, and OECD National Accounts data files (2013).

ORIGINAL

LIFE EXPECTANCY VS ECONOMIC GROWTH

RELATION BETWEEN LIFE EXPECTANCY AND GROSS NATIONAL INCOME OF ASIAN-PACIFIC COUNTRIES IN 2013



Source: World Bank national accounts data, and OECD National Accounts data files (2013).

DATA & MESSAGE REDUNDANCY

VERY FEW WOMEN ARE STUDYING COMPUTER SCIENCE

% of bachelor degrees awarded for women

COMPUTER SCIENCE

18%

MATHEMATICS

42%

SOCIAL SCIENCES

53%

PSYCHOLOGY

72%

0%

25%

50%

75%

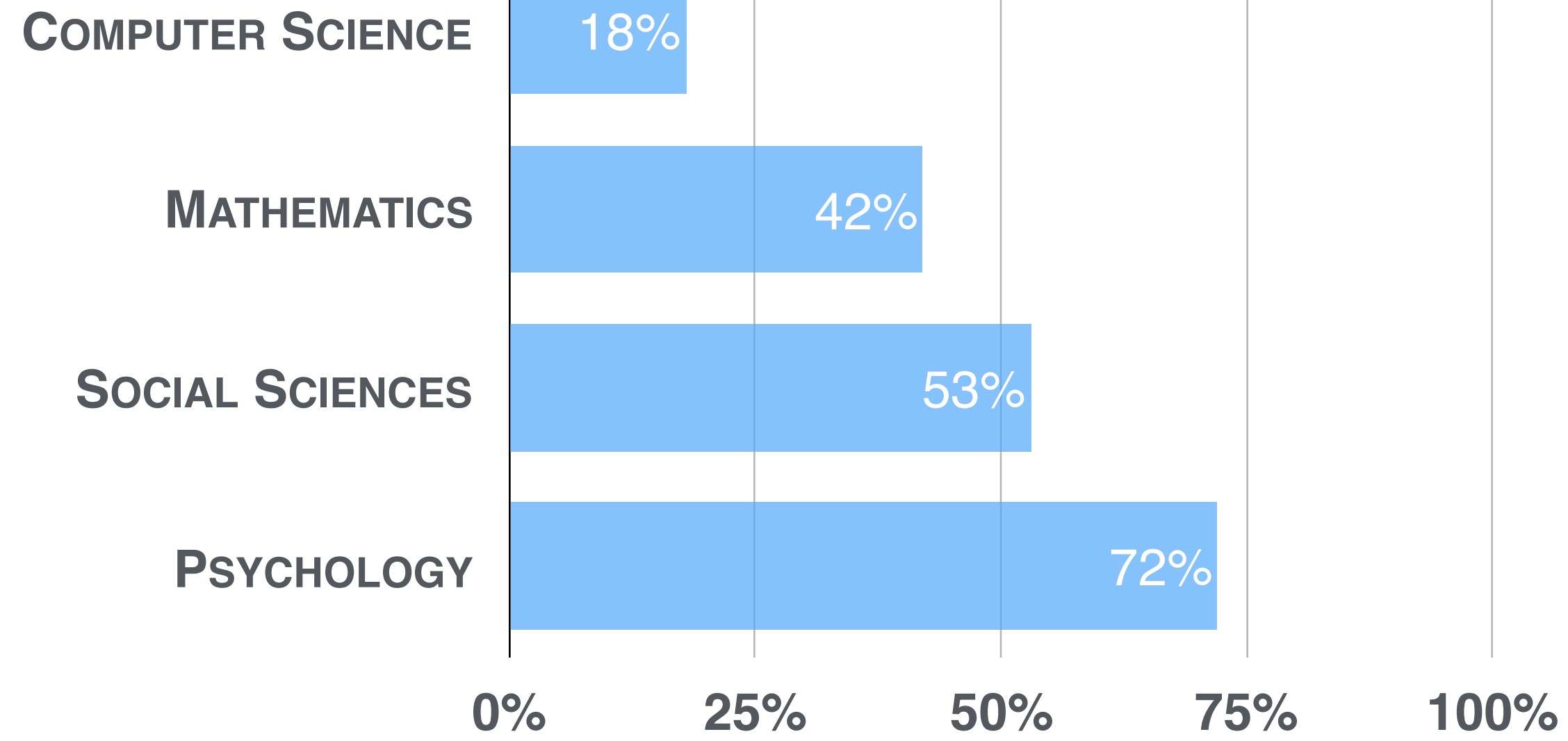
100%

Source: National Student Clearinghouse Research Center (2014).

WITHOUT PICTOGRAM

VERY FEW WOMEN ARE STUDYING COMPUTER SCIENCE

% of bachelor degrees awarded for women

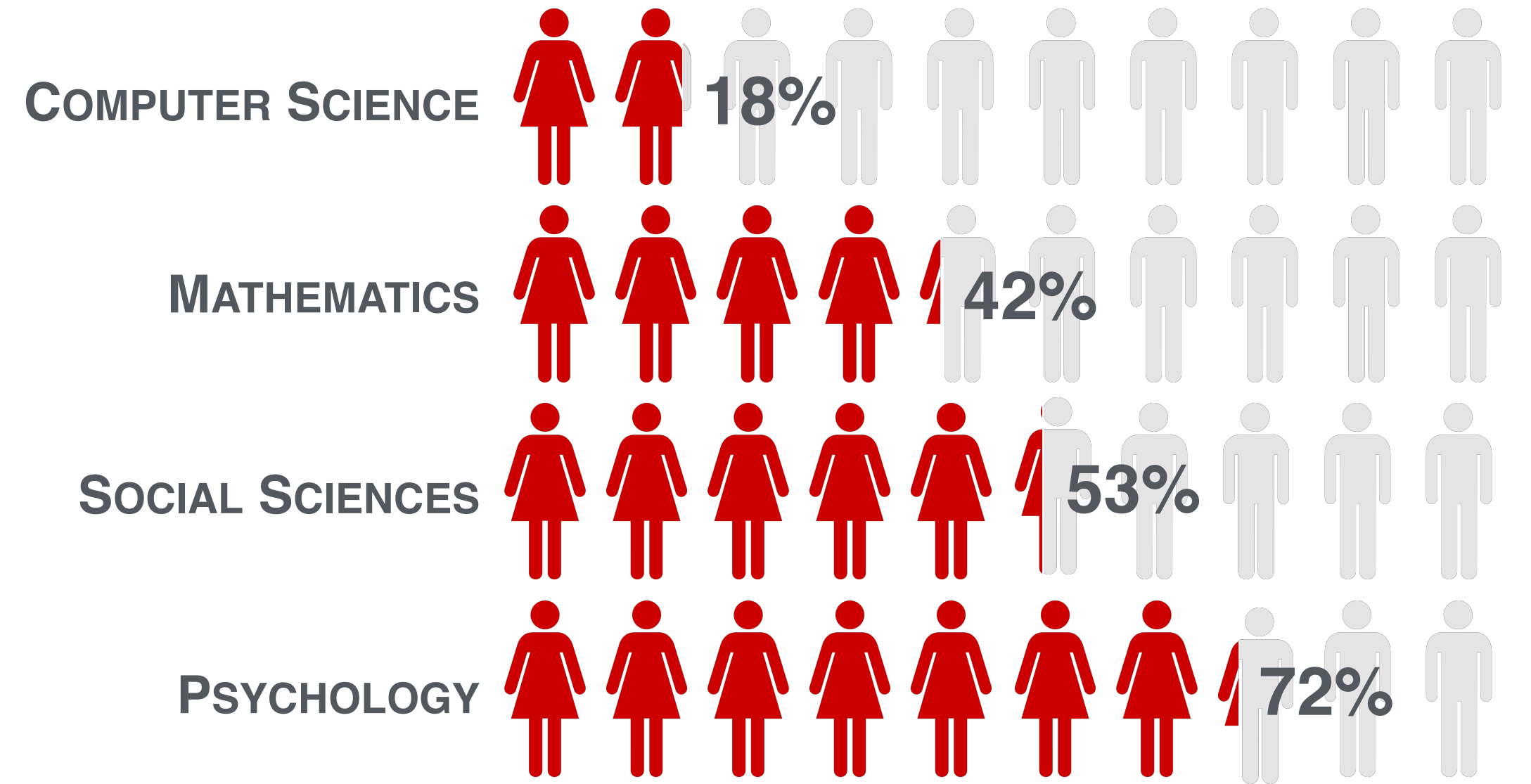


Source: National Student Clearinghouse Research Center (2014).

WITHOUT PICTOGRAM

VERY FEW WOMEN ARE STUDYING COMPUTER SCIENCE

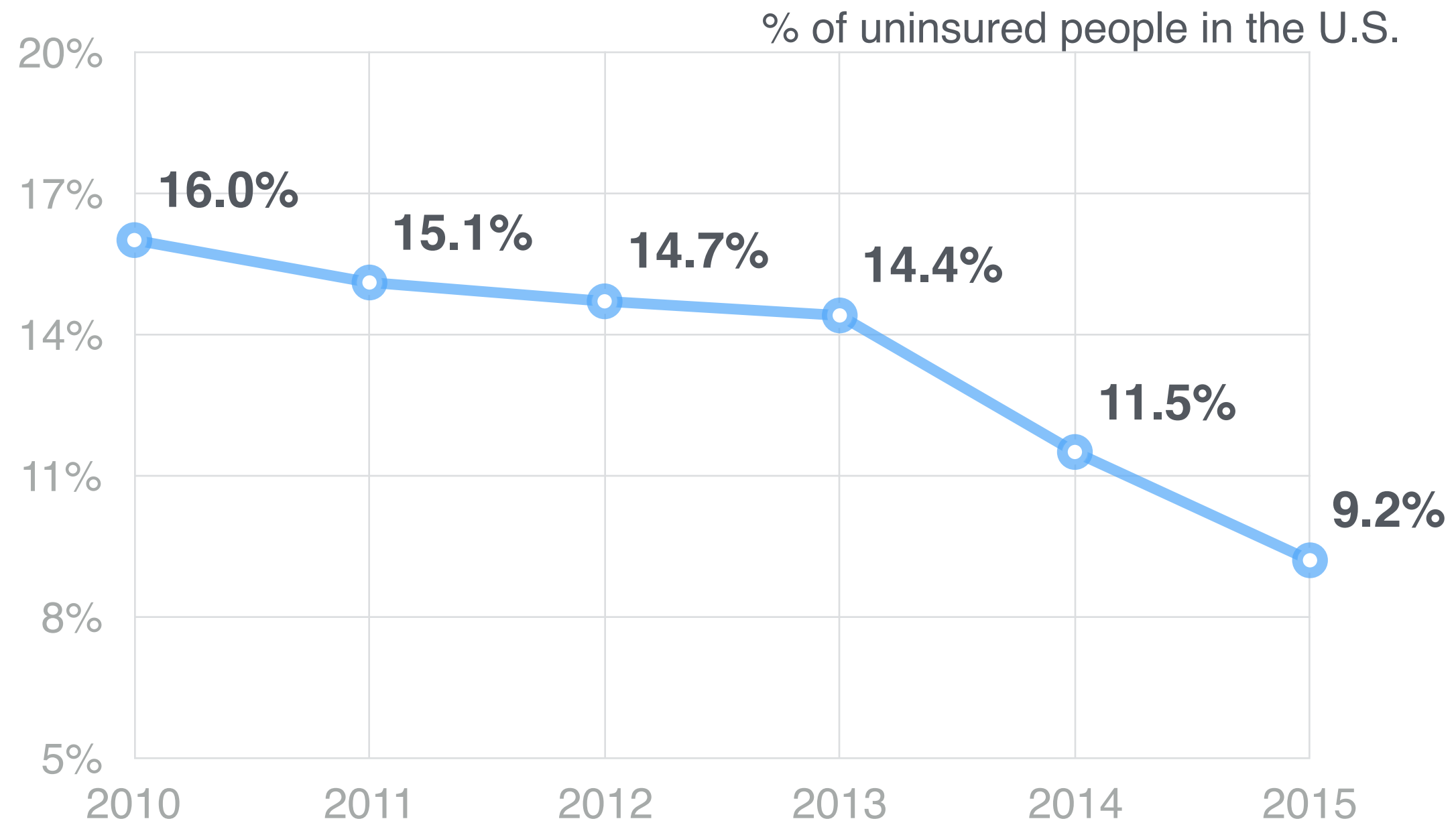
% of bachelor degrees awarded for women



Source: National Student Clearinghouse Research Center (2014).

WITH PICTOGRAM

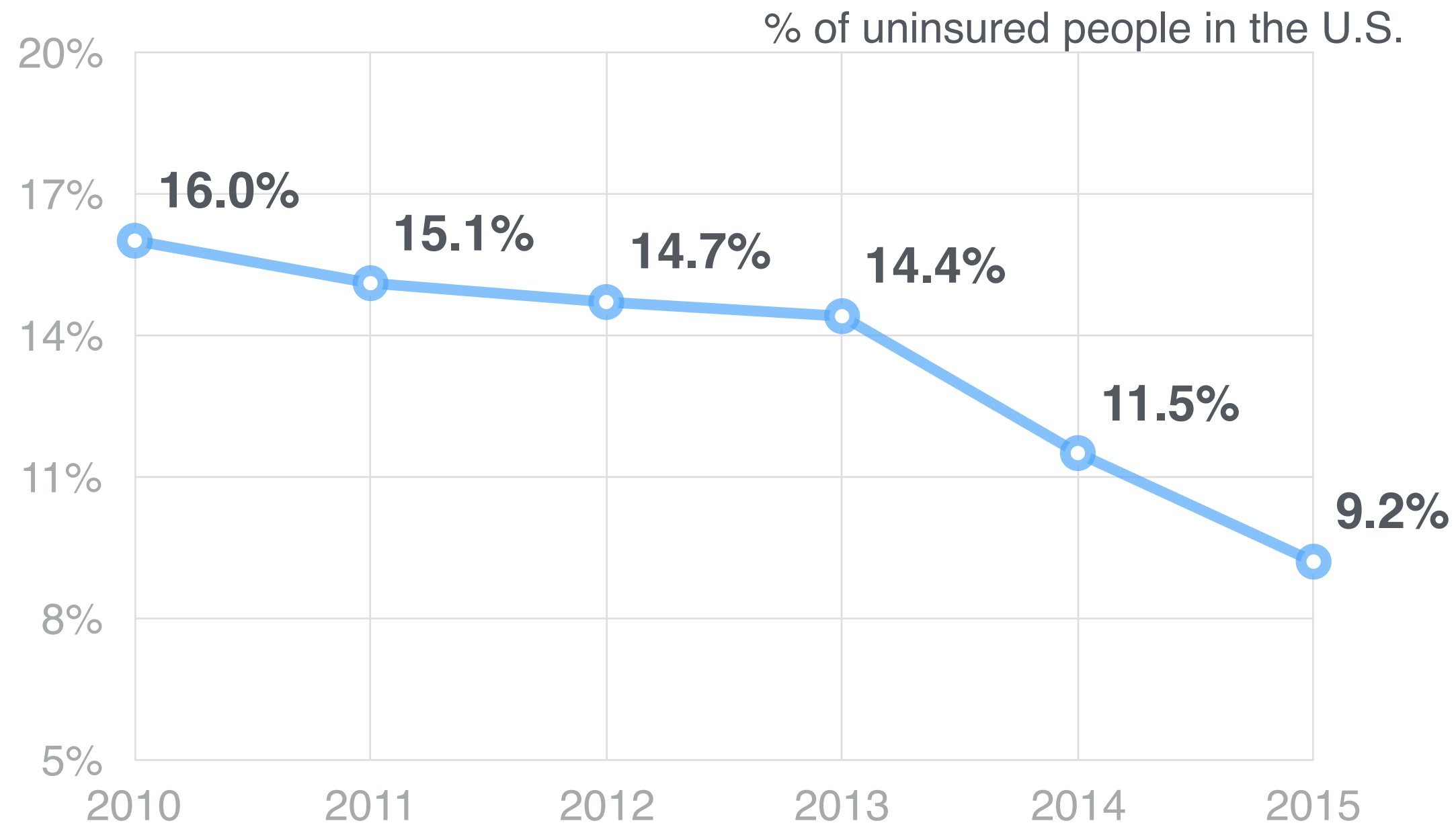
PERCENTAGE OF UNINSURED AMERICANS



Source: CDC/NCHS, National Health Interview Survey, 2010–2015

BAD TITLE

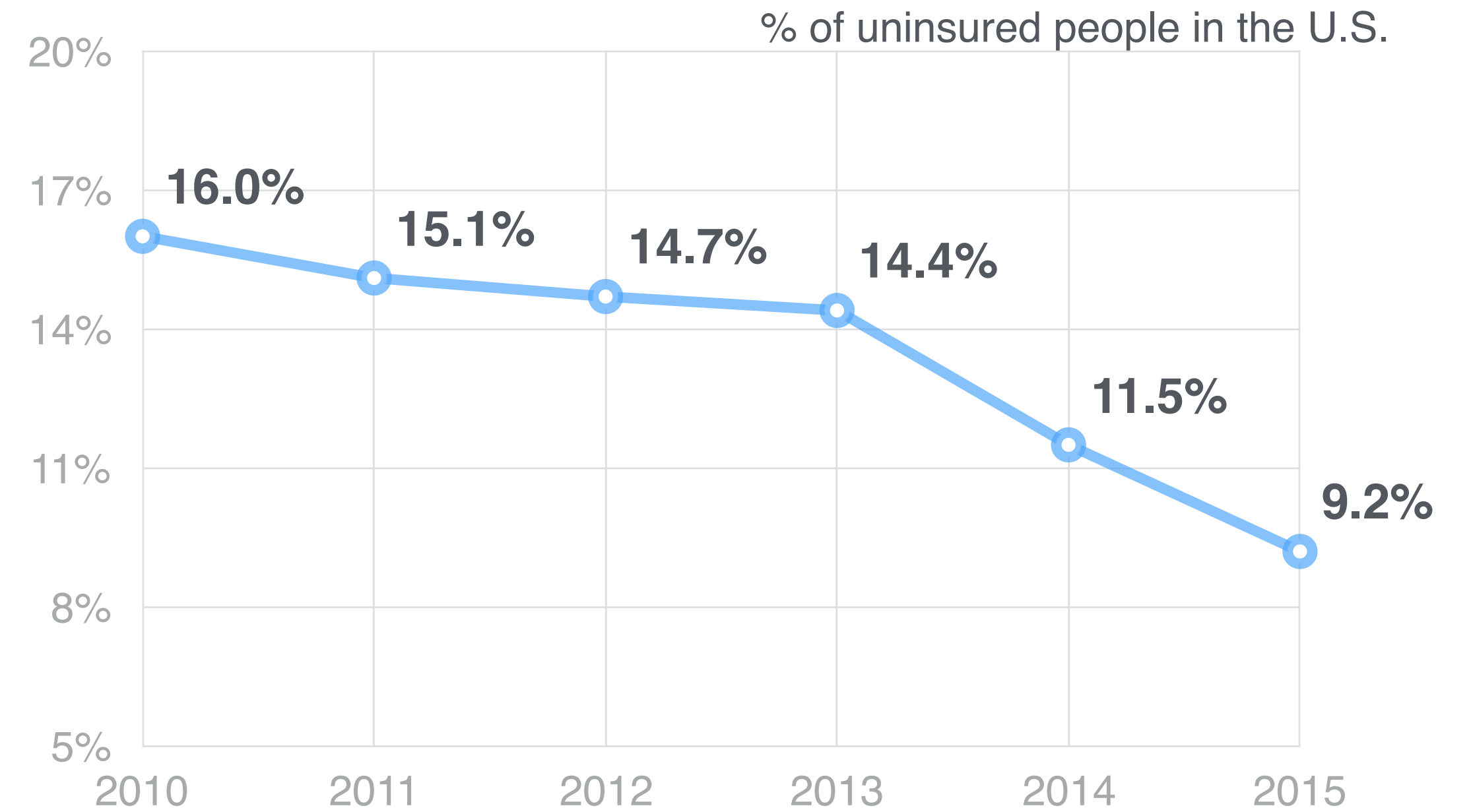
PERCENTAGE OF UNINSURED AMERICANS



Source: CDC/NCHS, National Health Interview Survey, 2010–2015

BAD TITLE

AMERICA'S UNINSURED RATE DIPS BELOW 10%



Source: CDC/NCHS, National Health Interview Survey, 2010–2015

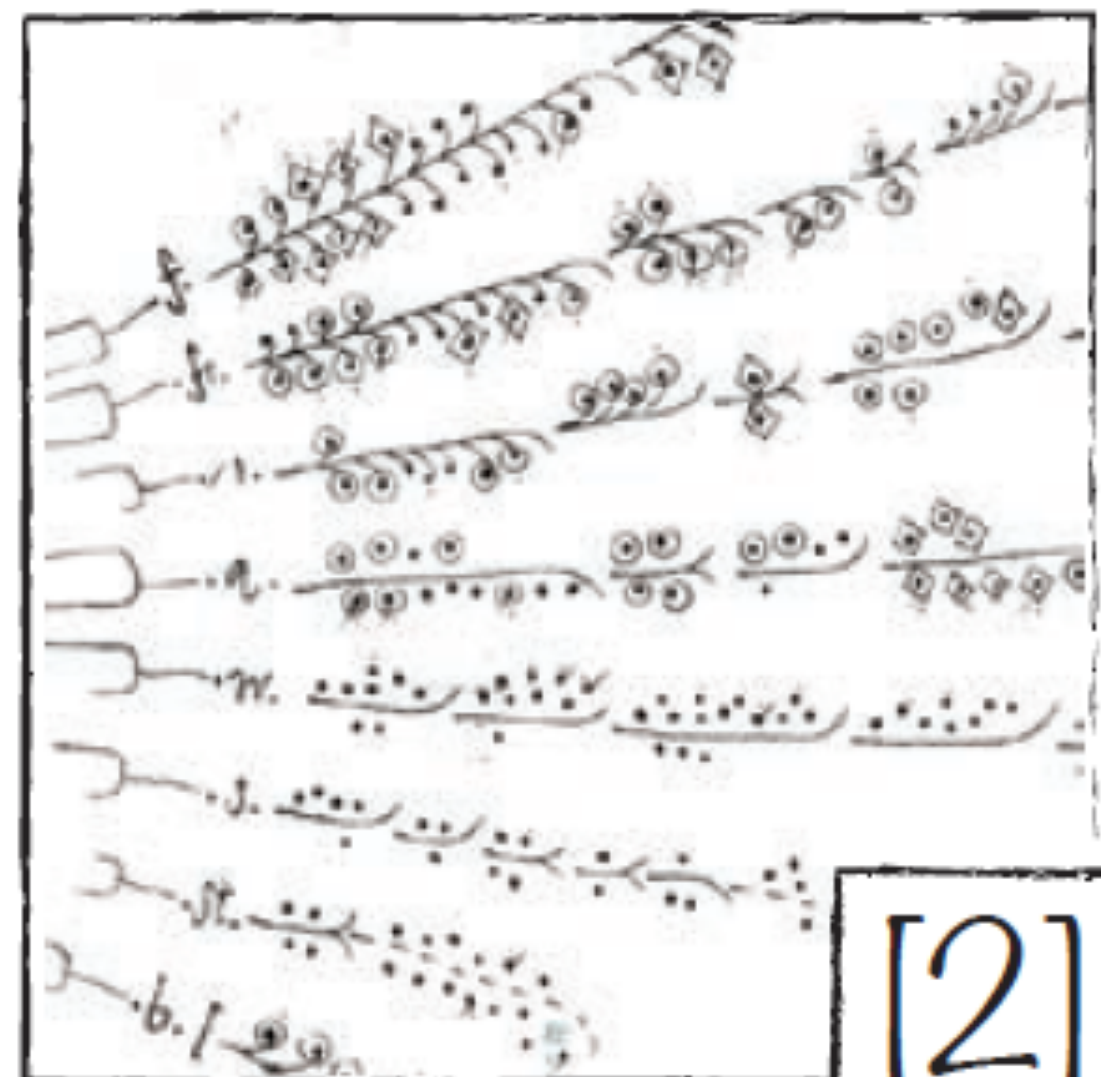
GOOD TITLE

Telling Stories with Data

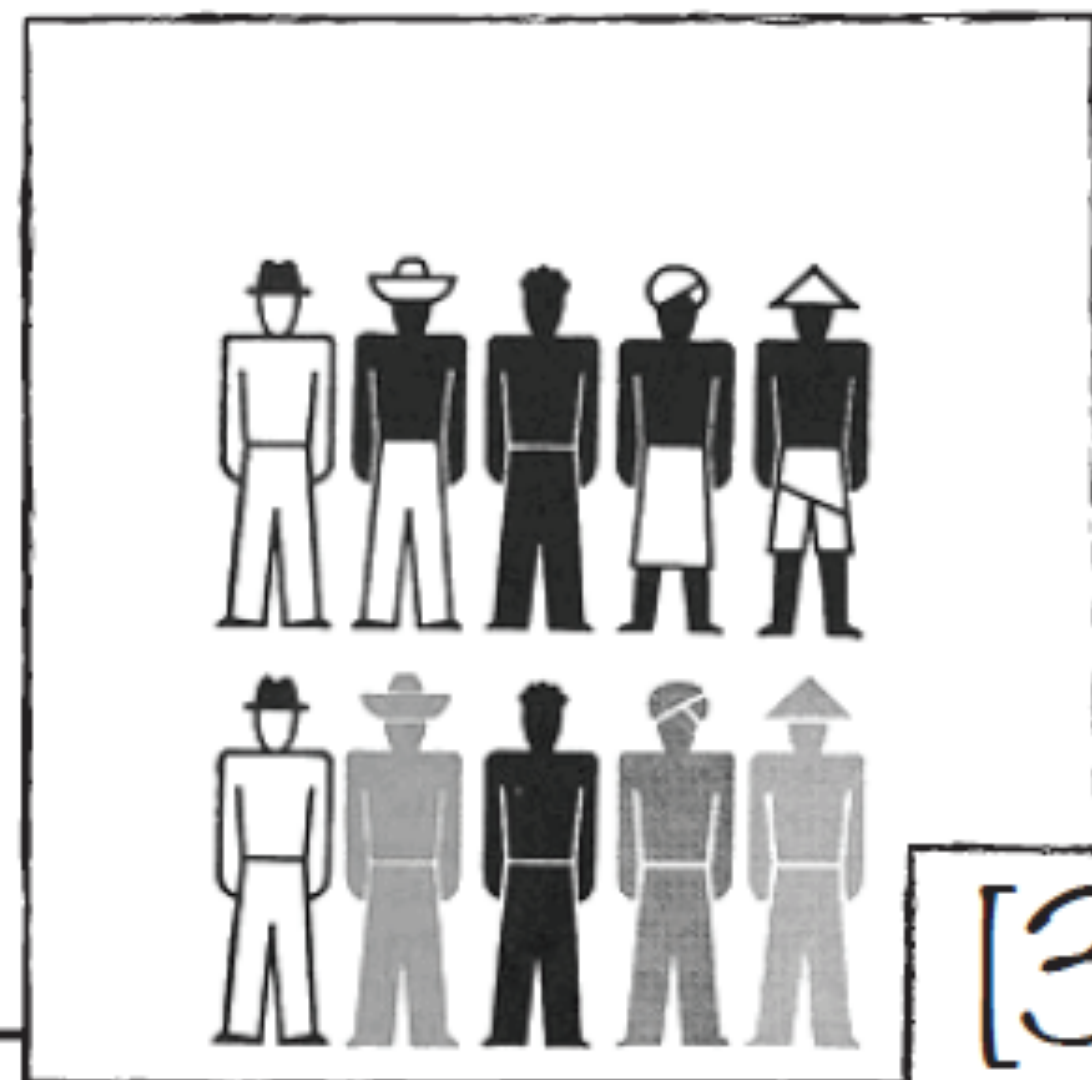
Visualizations can be effective in showing data,



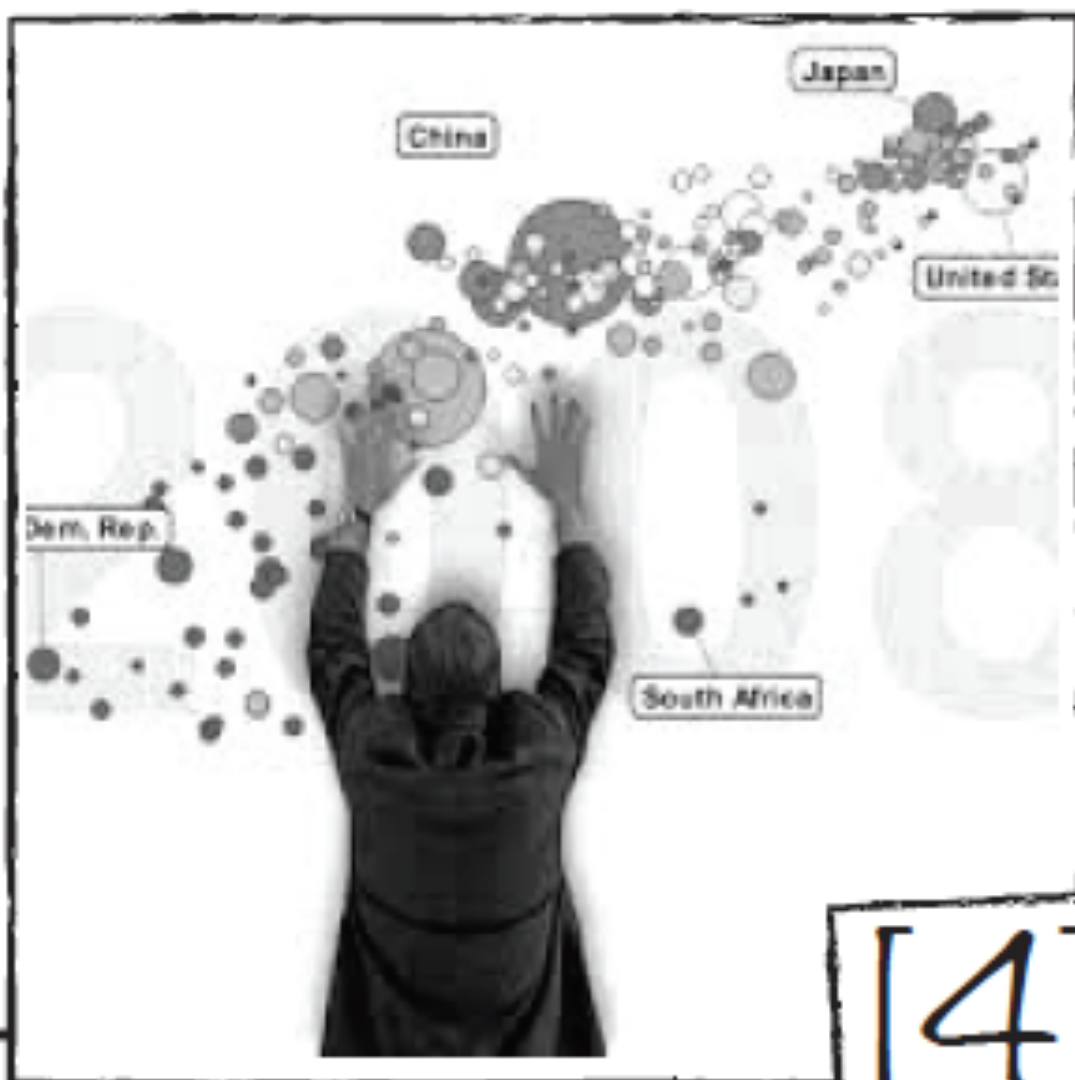
[1]



[2]

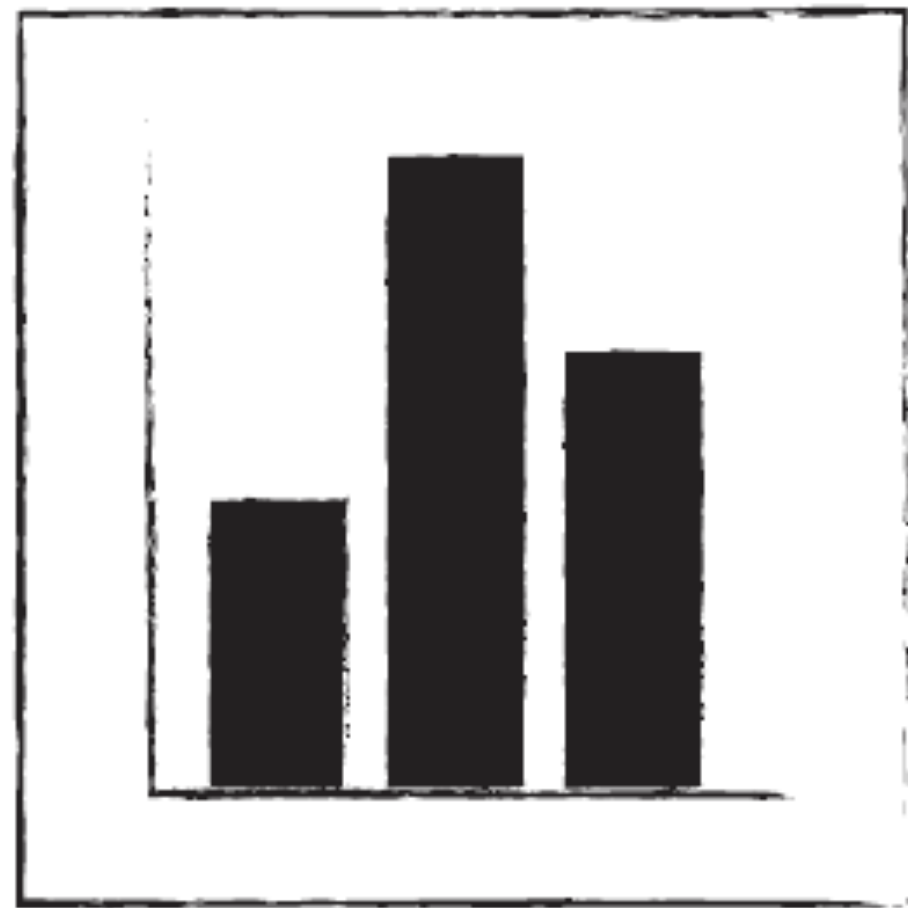


[3]



[4]

but a single picture alone may not be able to explain nor to engage an audience in decoding a message.



How do we engage an audience?


How do we break down complexity?

How do we guide the audience?



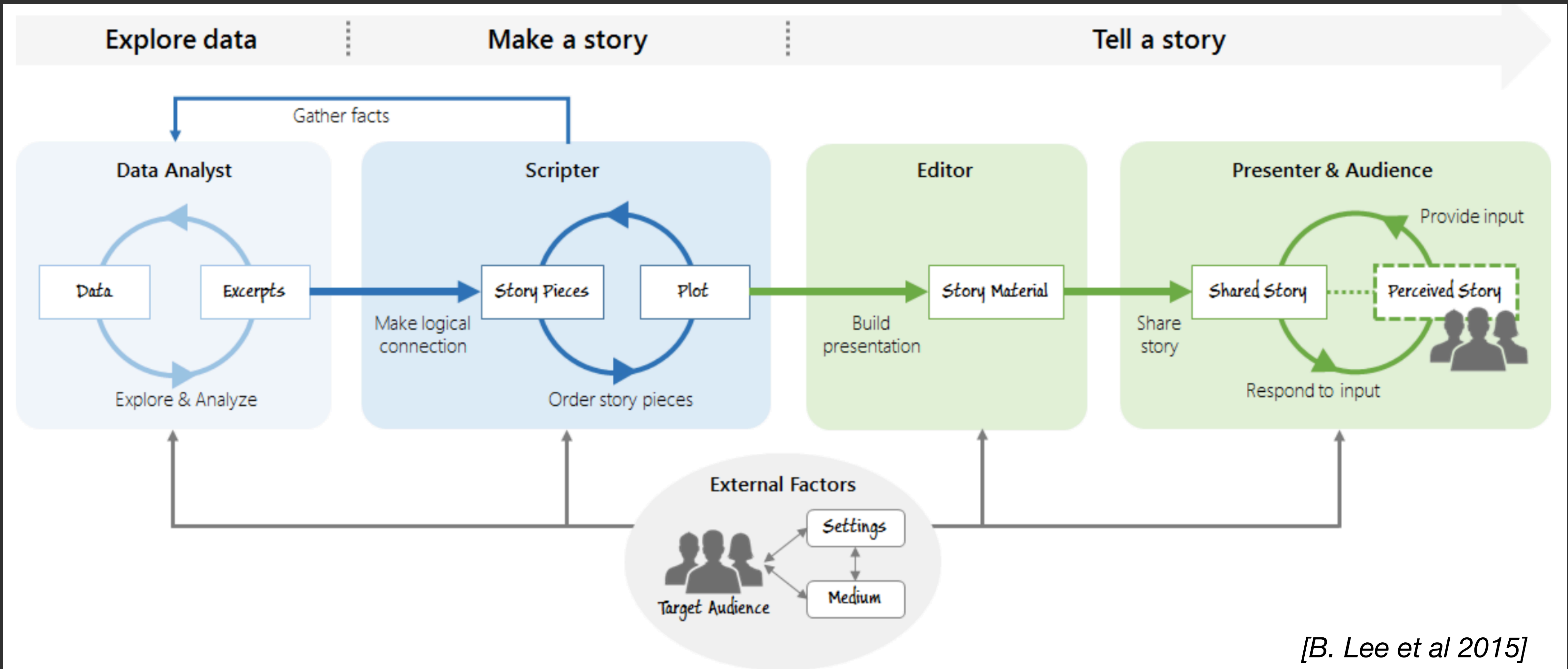
Stories

Stories are the most powerful delivery tool for information, more powerful and enduring than any other art form

A man in a dark jacket is seen from the back, reaching up with both hands towards a large digital display. The display features a grid with various colored points (red, blue, green, yellow) and a large red oval. The background is a light-colored wall with a grid pattern. The text "Use of elements from storytelling to convey compelling data stories" is overlaid on a dark rectangular background in the center of the image.

Use of **elements from storytelling**
to convey compelling data stories

Data Storytelling Process: transforming data into visual stories.

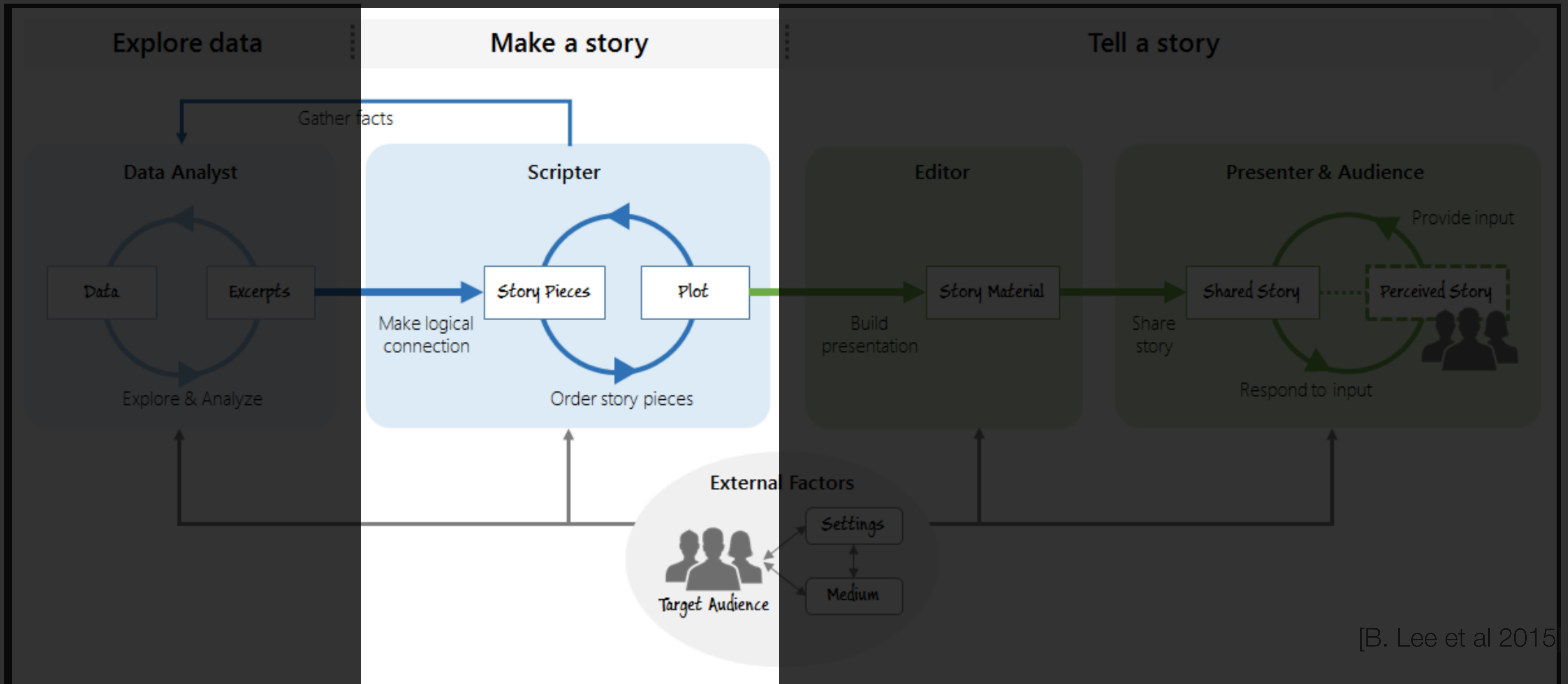


[B. Lee et al 2015]

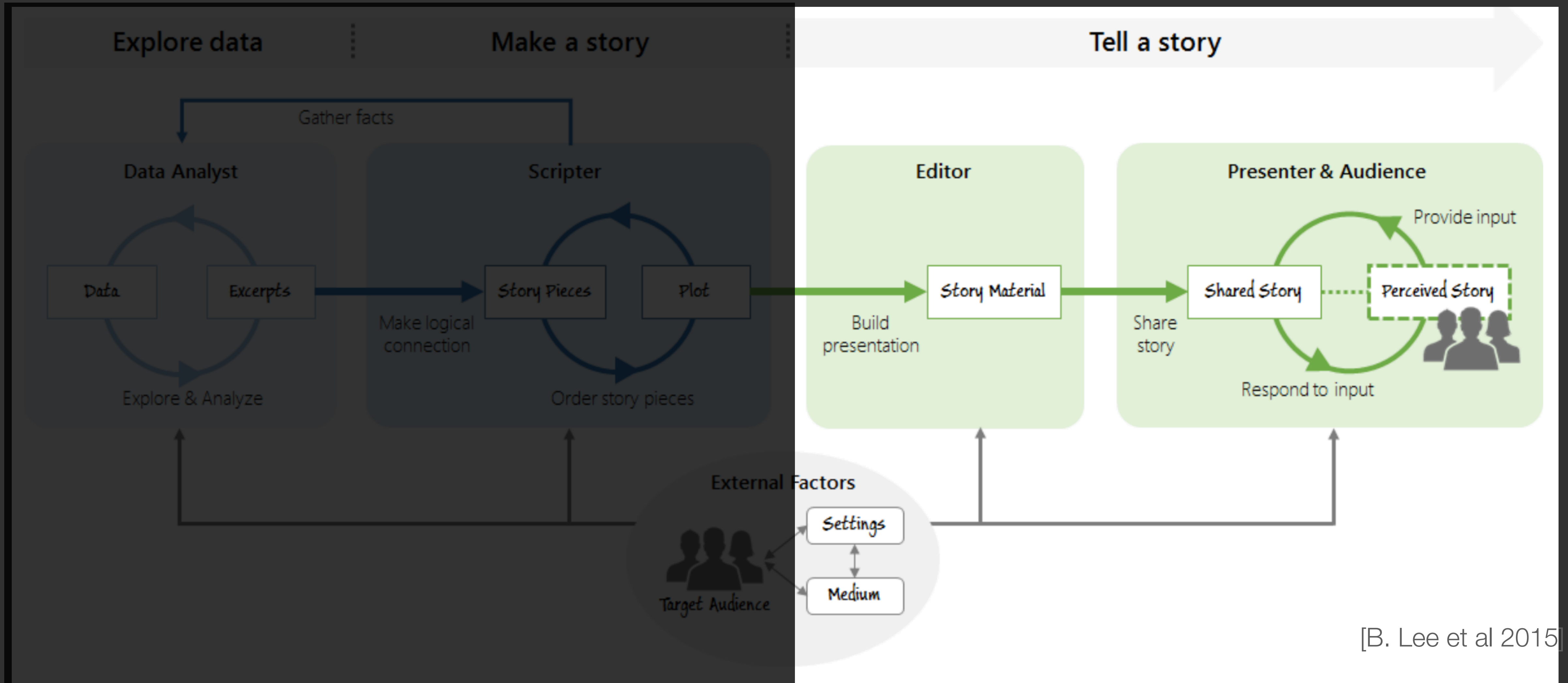
Data Storytelling Process: transforming data into visual stories.



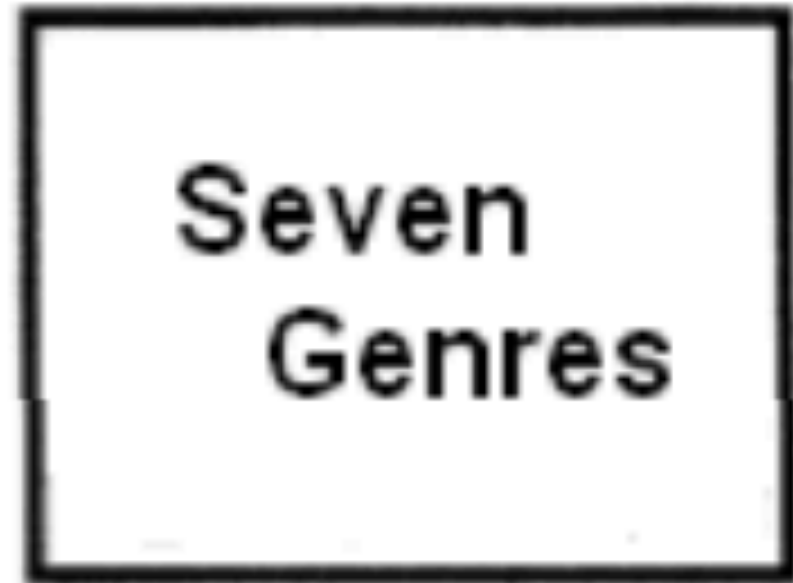
Data Storytelling Process: transforming data into visual stories.



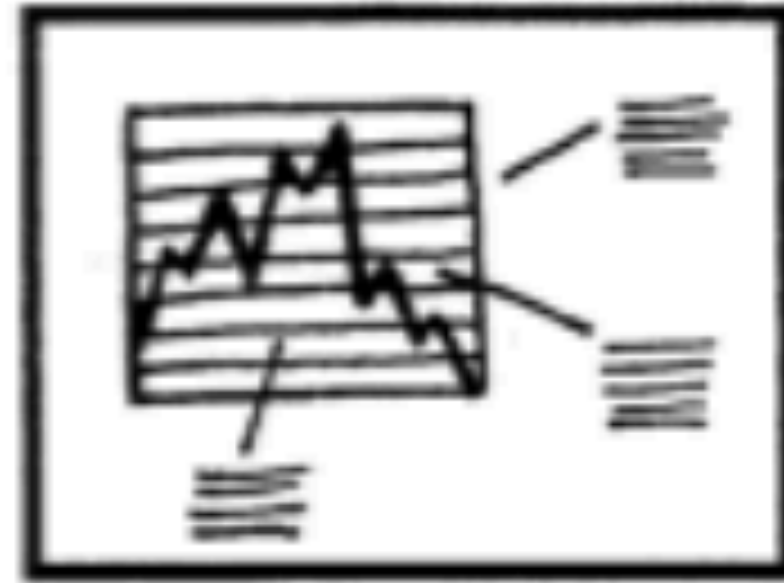
Data Storytelling Process: transforming data into visual stories.



Genres of Data Stories



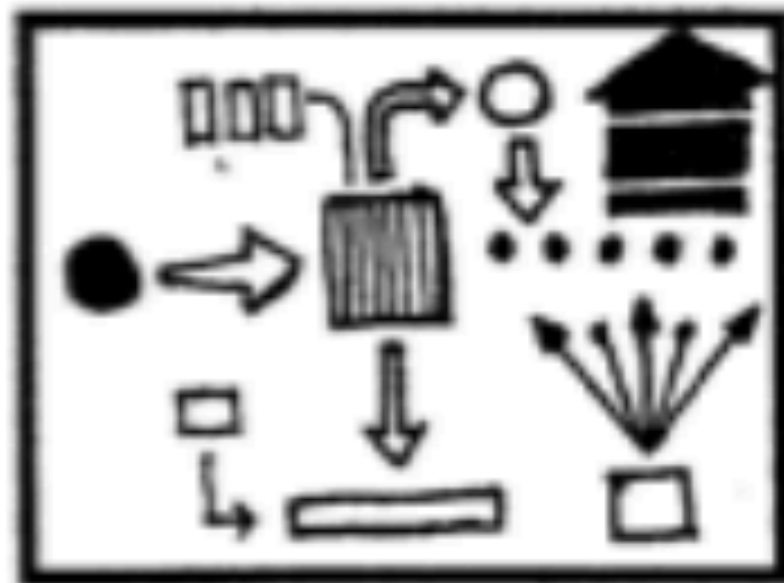
Magazine Style



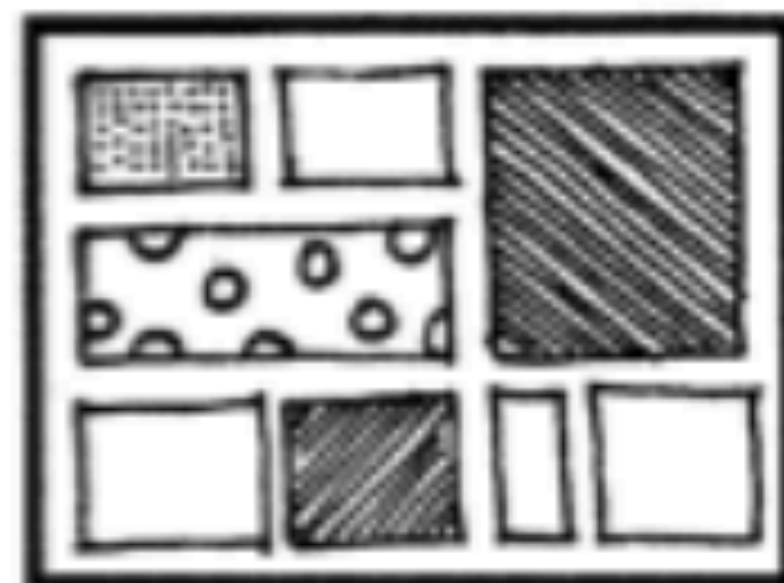
Annotated Chart



Partitioned Poster



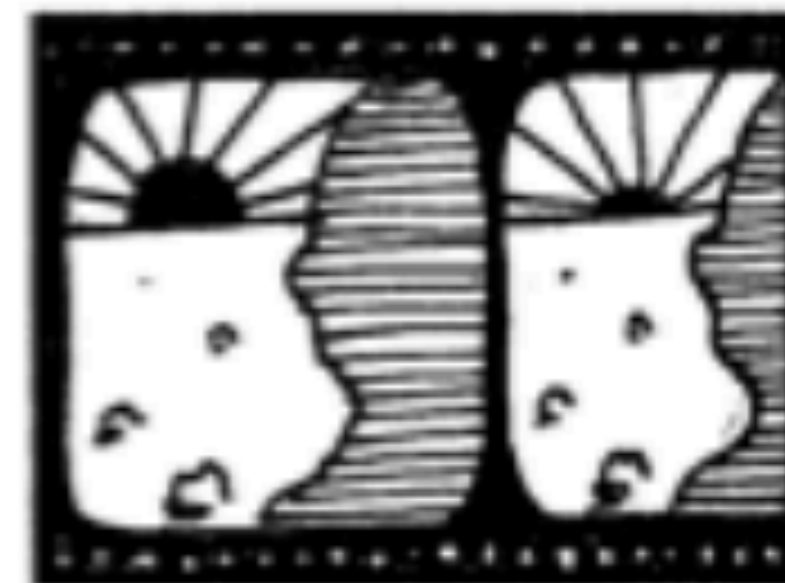
Flow Chart



Comic Strip



Slide Show



Film/Video/Animation

Narration Styles

Author-Driven

- Strict ordering of scenes
- Heavy messaging
- No interactivity

Reader-Driven

- No prescribed ordering
- No messaging
- Free interactivity

Narration Styles

Author-Driven + Reader-Driven

- Linear ordering of scenes
 - Heavy messaging
 - No interactivity
- No prescribed ordering
 - No messaging
 - Free interactivity

Author-Driven



Reader-Driven



Epoch
000,000

Learning rate
0.03

Activation
Tanh

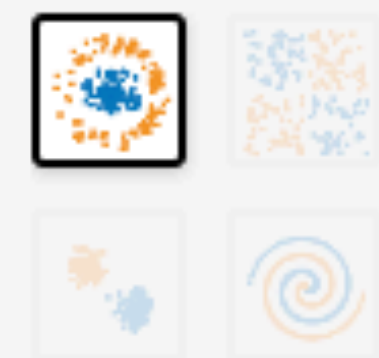
Regularization
None

Regularization rate
0

Problem type
Classification

DATA

Which dataset do you want to use?



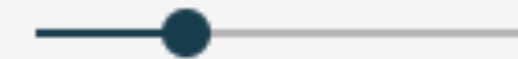
Ratio of training to test data: 50%



Noise: 0



Batch size: 10



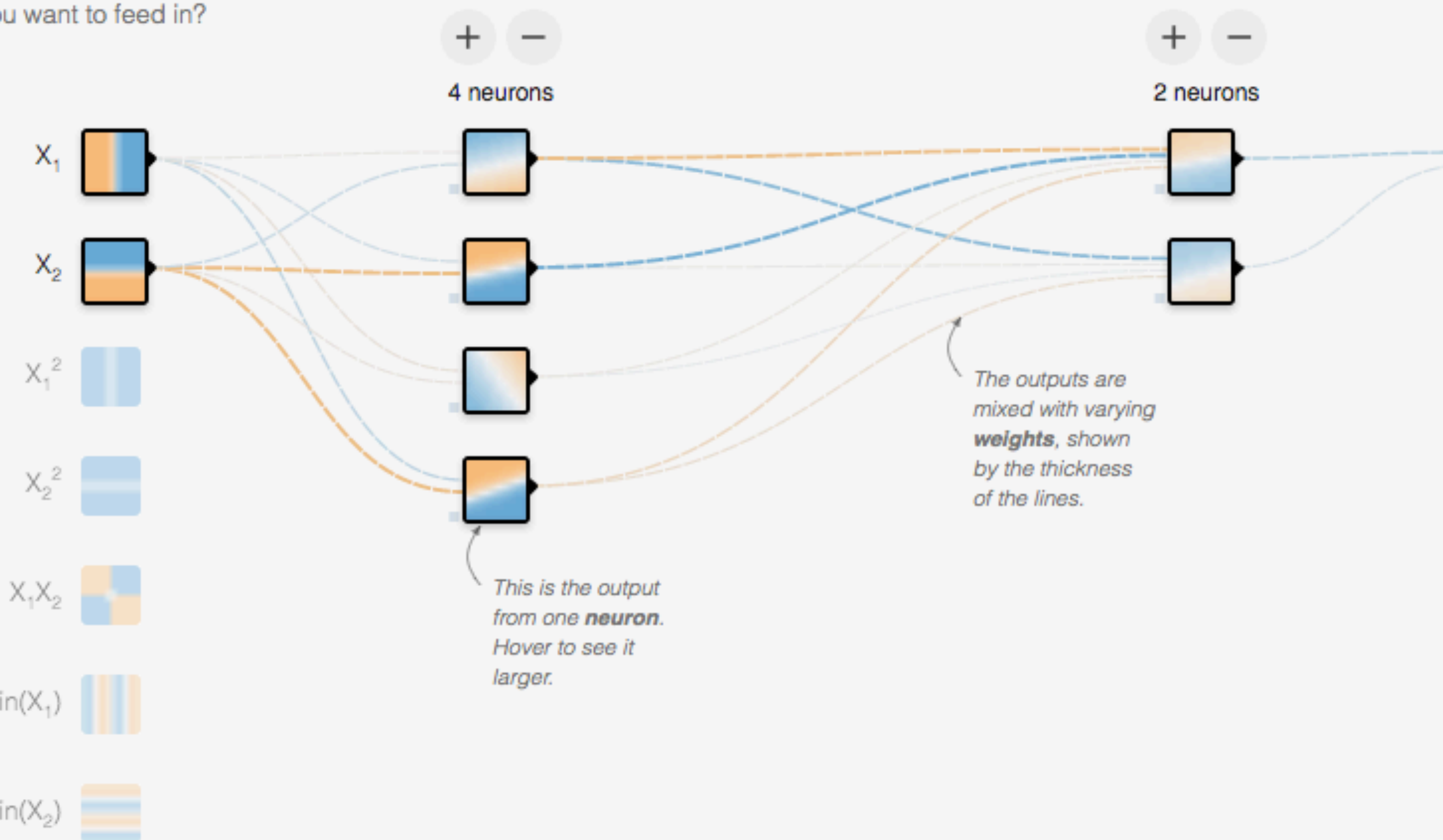
REGENERATE

FEATURES

Which properties do you want to feed in?

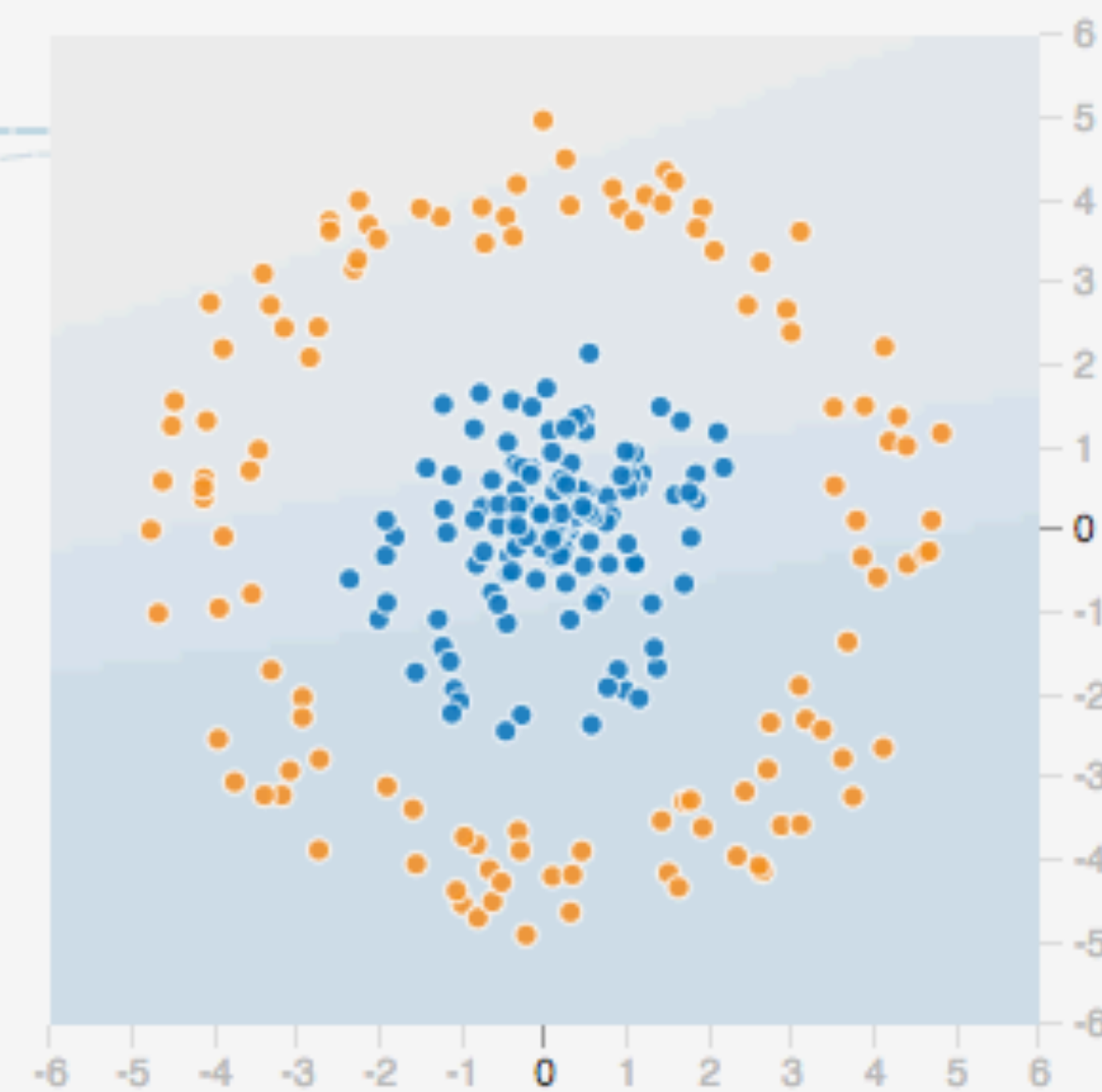
- X_1
- X_2
- X_1^2
- X_2^2
- X_1X_2
- $\sin(X_1)$
- $\sin(X_2)$

+ - 2 HIDDEN LAYERS



OUTPUT

Test loss 0.513
Training loss 0.500



Colors shows data, neuron and weight values.

Show test data Discretize output

Author-Driven + Reader-Driven

Budget Forecasts, Compared With Reality

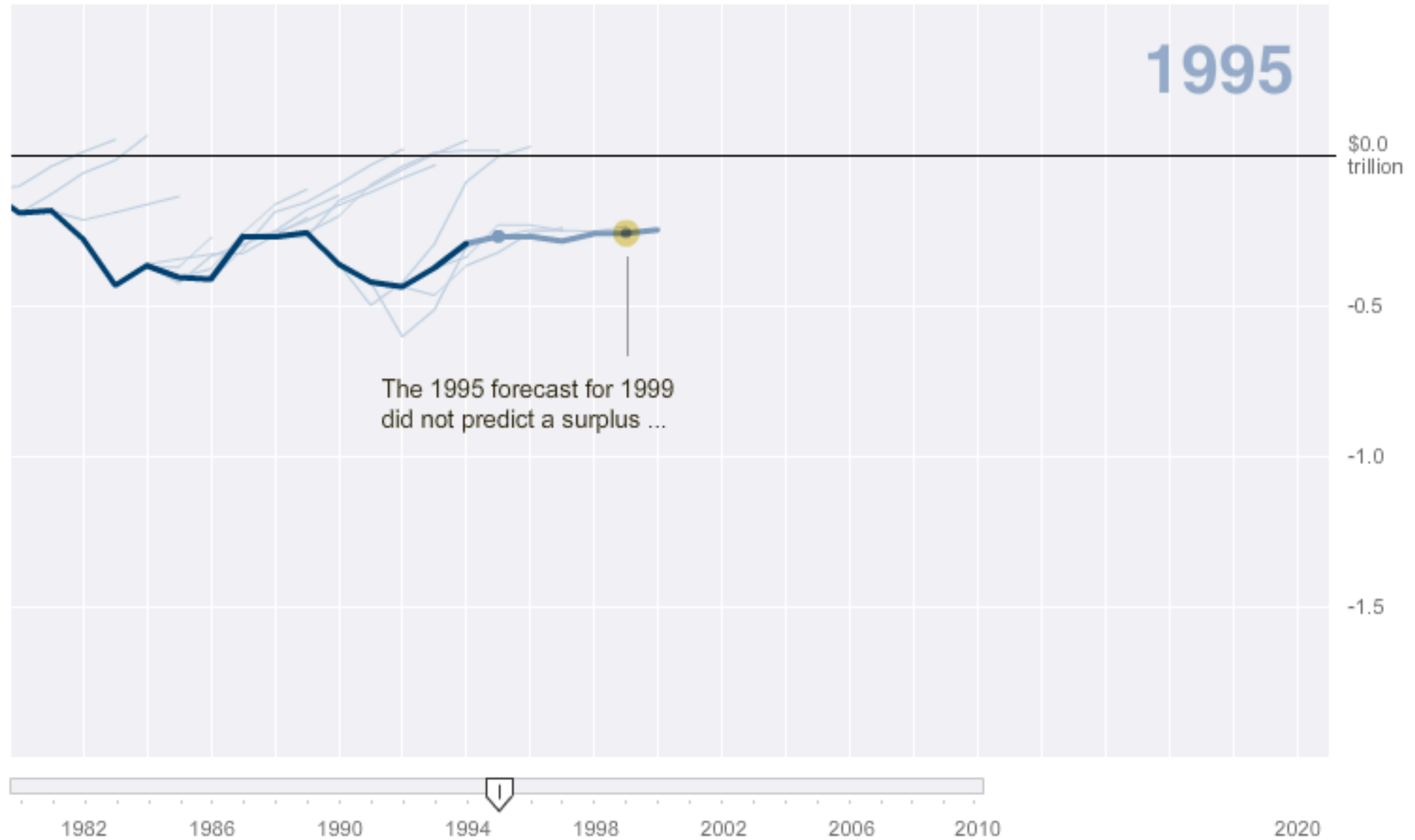
Just two years ago, surpluses were predicted by 2012. How accurate have past White House budget forecasts been?

1 2 3 4 5 6 NEXT ▶

Past forecasts

Even that may be an understatement. In the last 30 years, about 80 percent of four-year deficit forecasts have been too optimistic.

The early Clinton budgets — which failed to predict the surpluses that were generated, in part, by a stock market bubble — are the only major exception.



By AMANDA COX | [Send Feedback](#)

Source: Office of Management and Budget

[TWITTER](#) [LINKEDIN](#) [SHARE](#)

U.S. GUN DEATHS IN

2013 2010

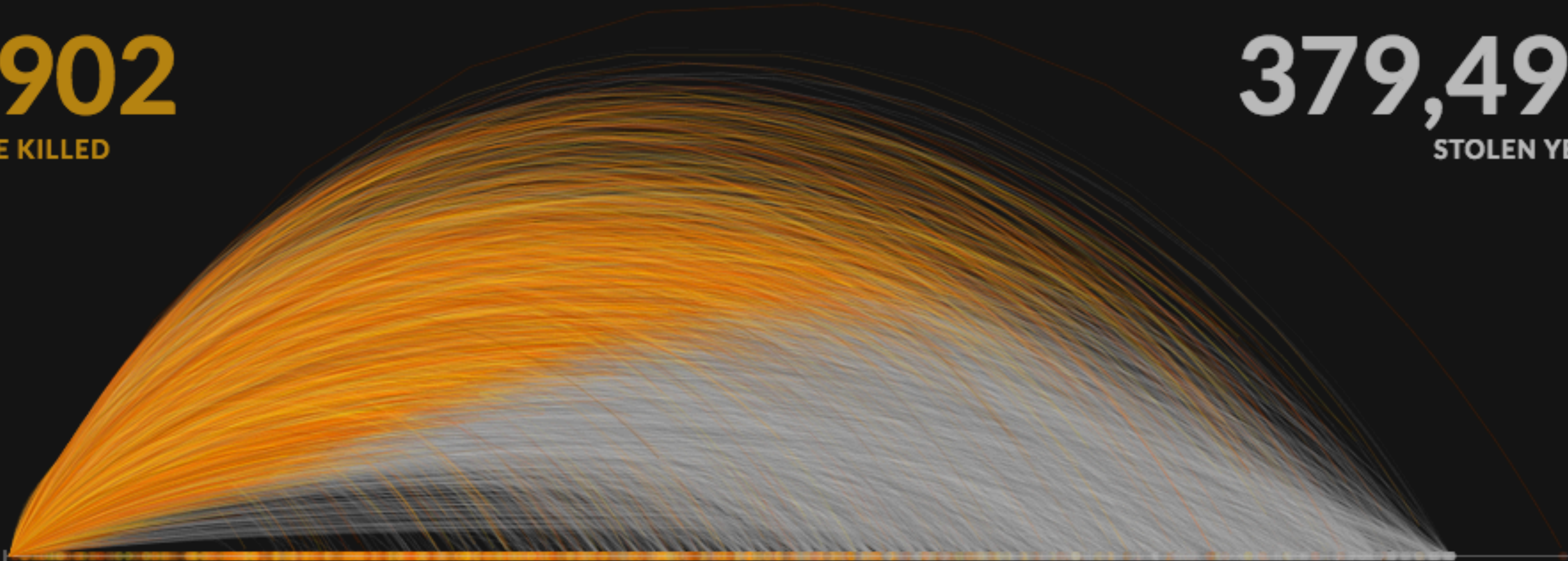
OCTOBER

8,902

PEOPLE KILLED

379,491

STOLEN YEARS



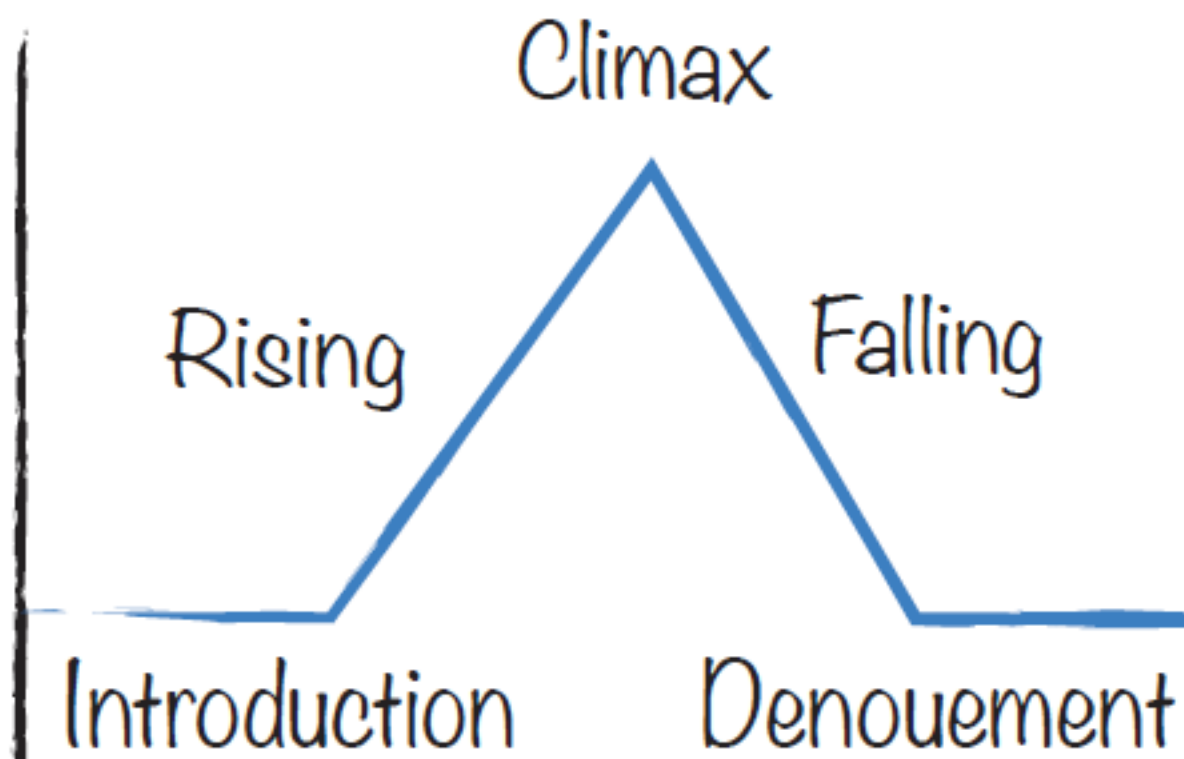
THE
Fallen
OF
World War II



Visual Narrative Design

Employ Narrative Structure

But storytelling is an old **art**
and learning from the masters



... can help to create truly

dramatic **walkthroughs.**



755



Steroids or Not, the Pursuit Is On

Barry Bonds is taking aim at the career home run record. He needs only six more to tie Babe Ruth and 47 to equal Hank Aaron.

Lines are cumulative home runs.

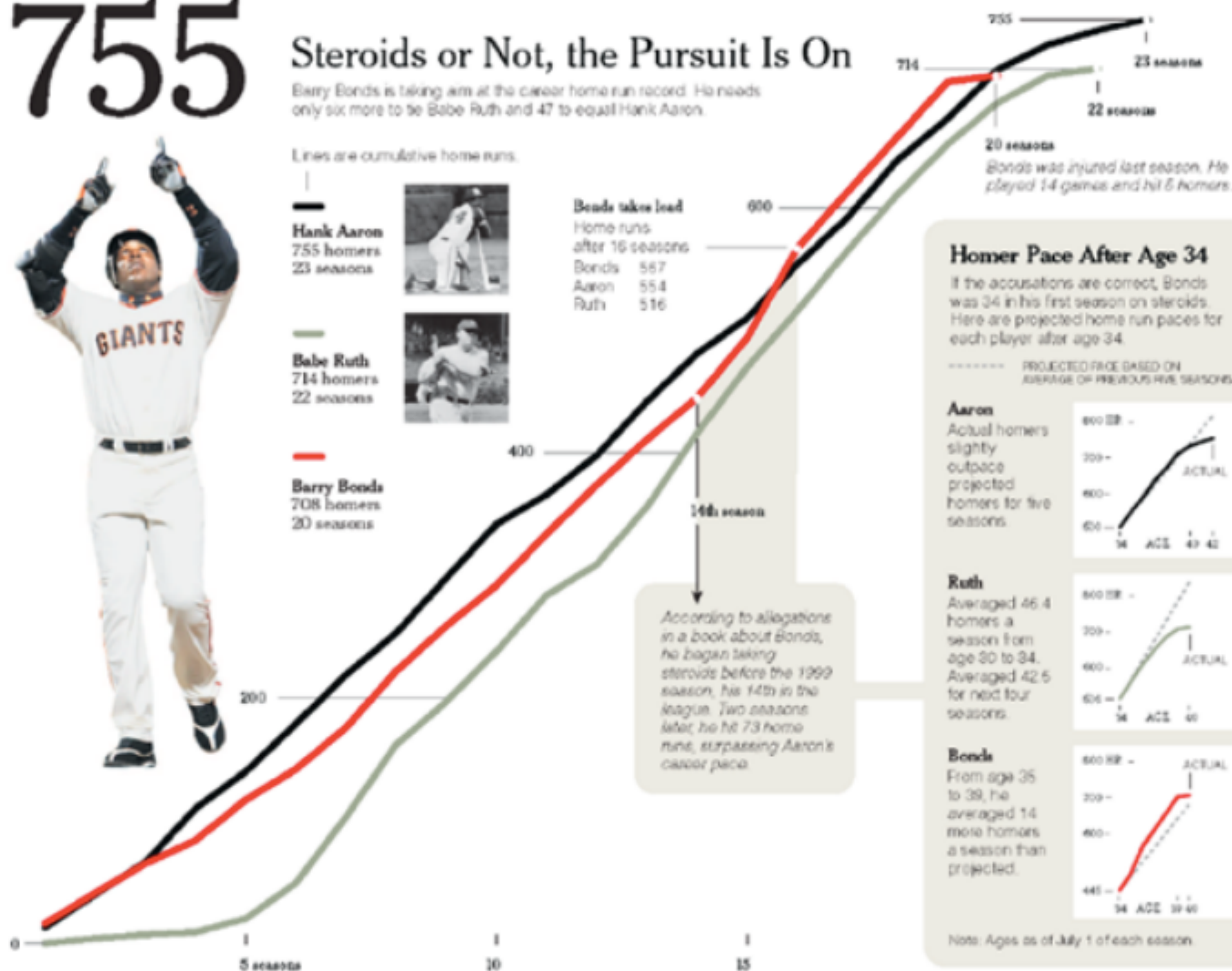
Hank Aaron
755 homers
23 seasons



Babe Ruth
714 homers
22 seasons



Barry Bonds
708 homers
20 seasons



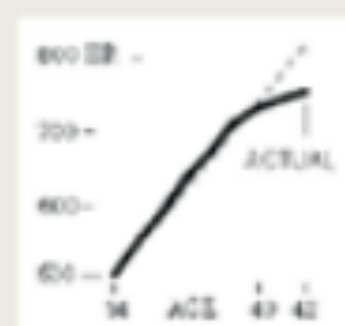
Homer Pace After Age 34

If the accusations are correct, Bonds was 34 in his first season on steroids. Here are projected home run paces for each player after age 34.

----- PROJECTED PACE BASED ON AVERAGE OF PREVIOUS FIVE SEASONS

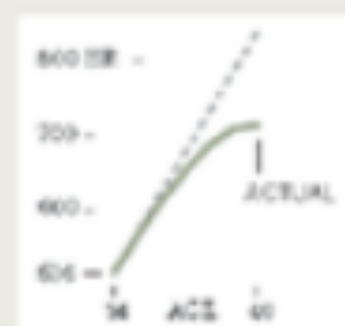
Aaron

Actual homers slightly outpace projected homers for five seasons.



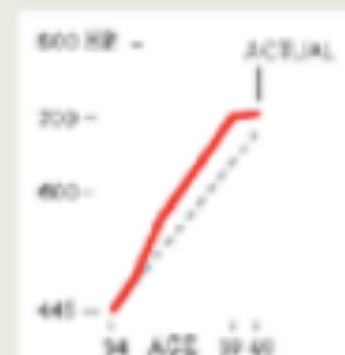
Ruth

Averaged 46.4 homers a season from age 30 to 34. Averaged 42.5 for next four seasons.



Bonds

From age 35 to 39, he averaged 14 more homers a season than projected.



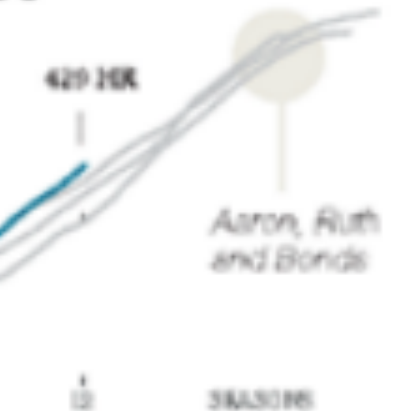
Note: Ages as of July 1 of each season.

Others Taking Aim



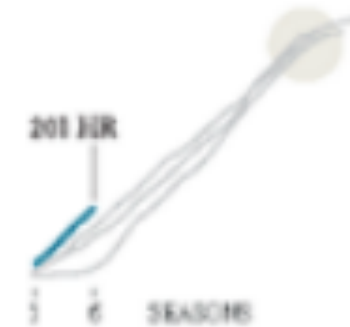
Alex Rodriguez

Is ahead of the pace set by all three home run leaders.



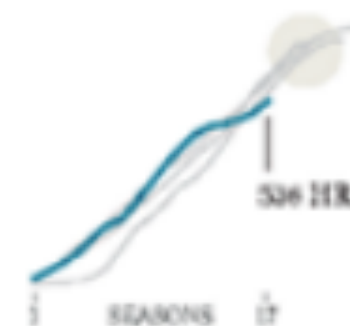
Albert Pujols

Averaging 40 homers a season, he has started stronger than the three leaders did.



Ken Griffey Jr.

Many thought he would be the first to catch Ruth and Aaron until injuries limited his output.



Differing Paths to the Top of the Charts

The top seven players on the career home run list, along with a look at Griffey (12th), Rodriguez (37th) and Pujols (56th).

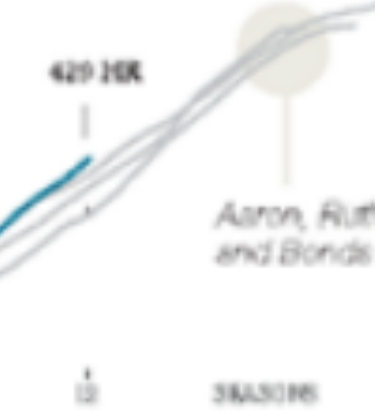




Others Taking Aim

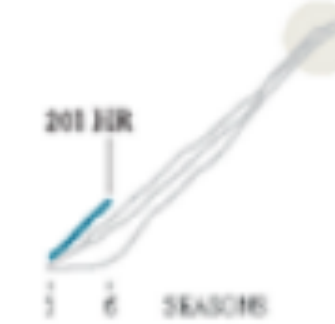
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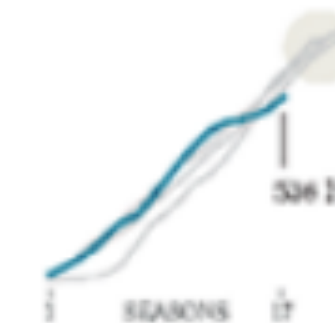
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Differing Paths to the Top of the Charts

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Illustration by Jeff Labrecque for The New York Times

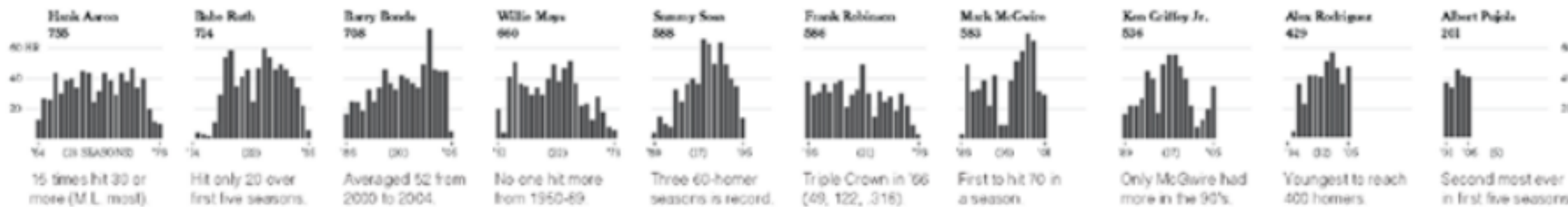


BEGINNING



MIDDLE

Differing Paths to the Top of the Charts The top seven players on the career home run list, along with a look at Griffey (12th), Rodriguez (37th) and Pujols (56th).





BEGINNING



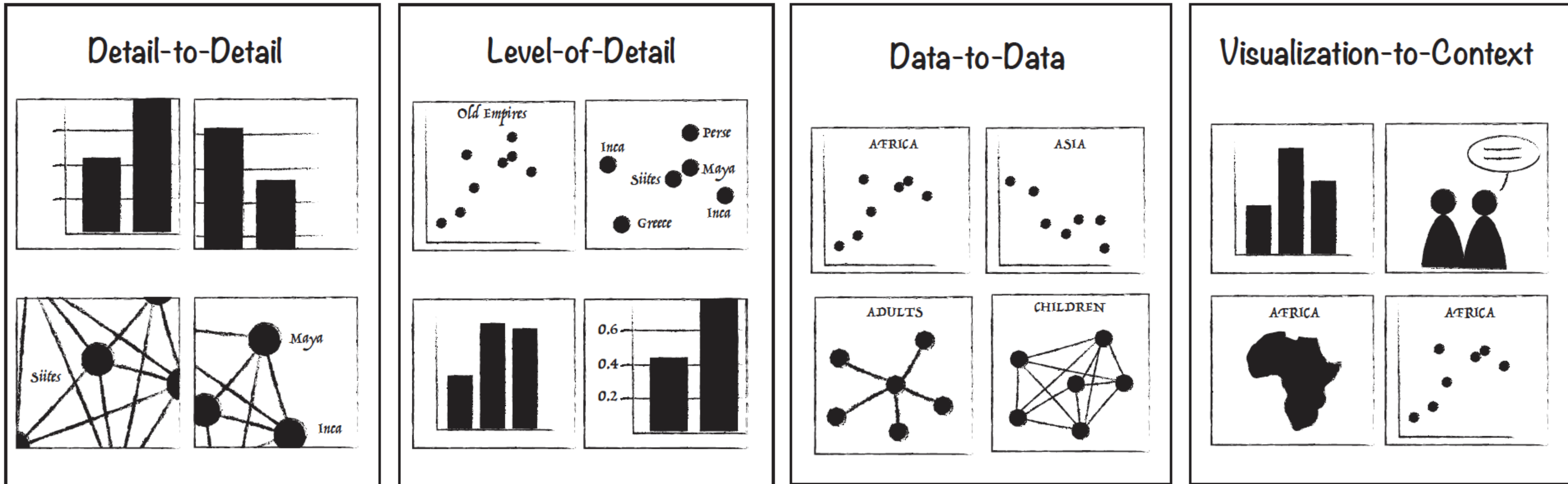
MIDDLE



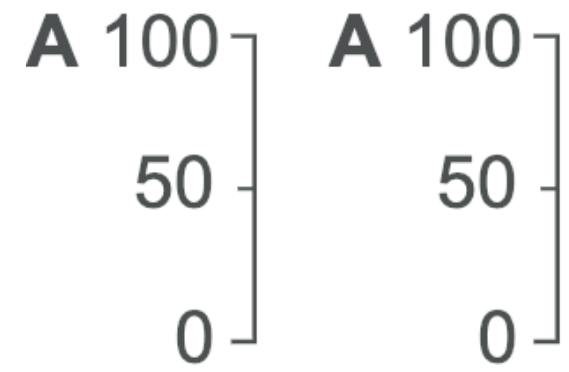
END

Use Multiple Charts

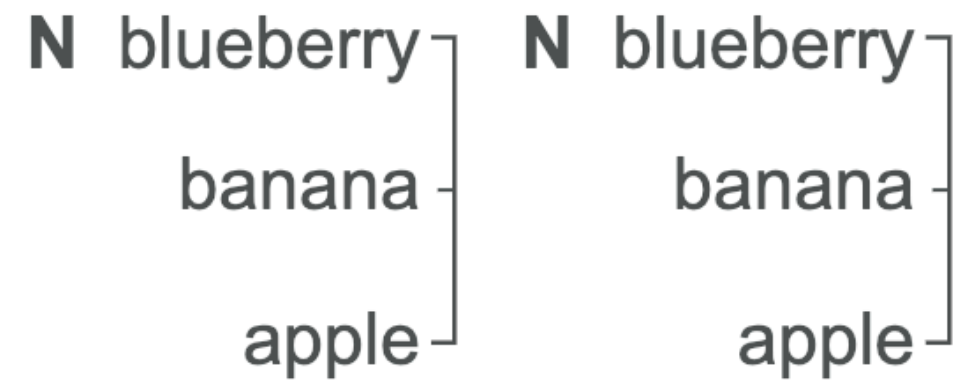
To break down the complexity of the story and progressively reveal different facets of the data.



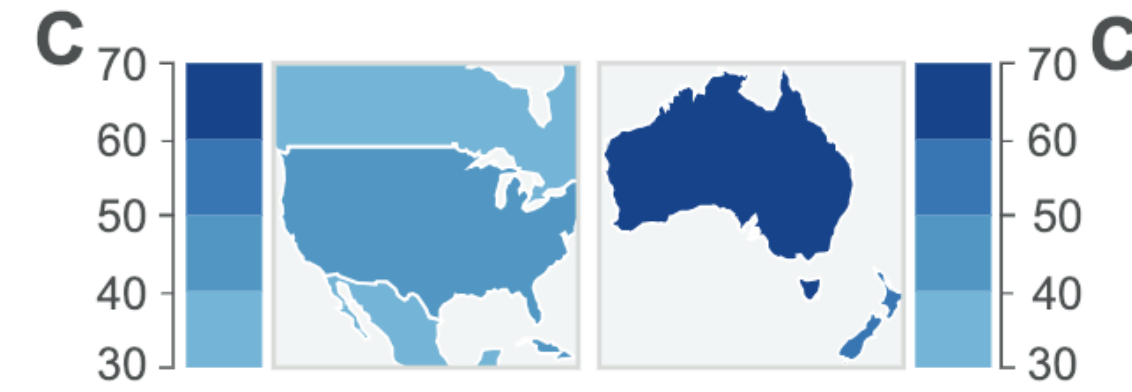
C1.1 Same Field, Same XY Scale



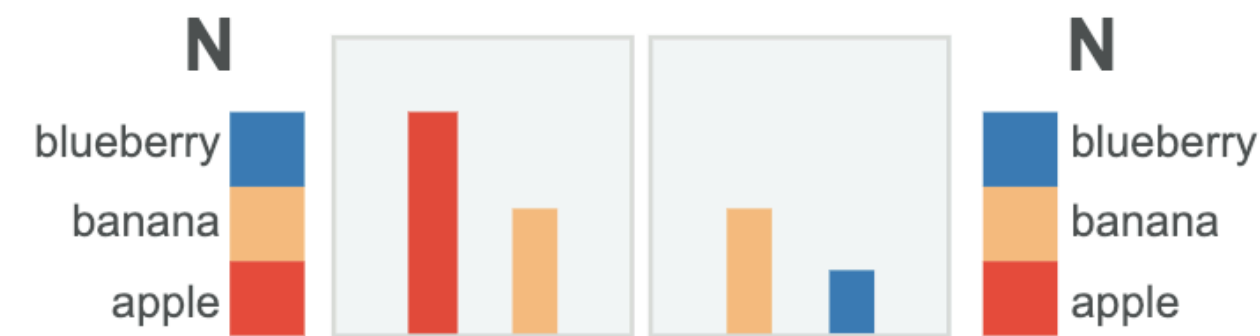
C1.2 Same Field, Same Values in Same Order



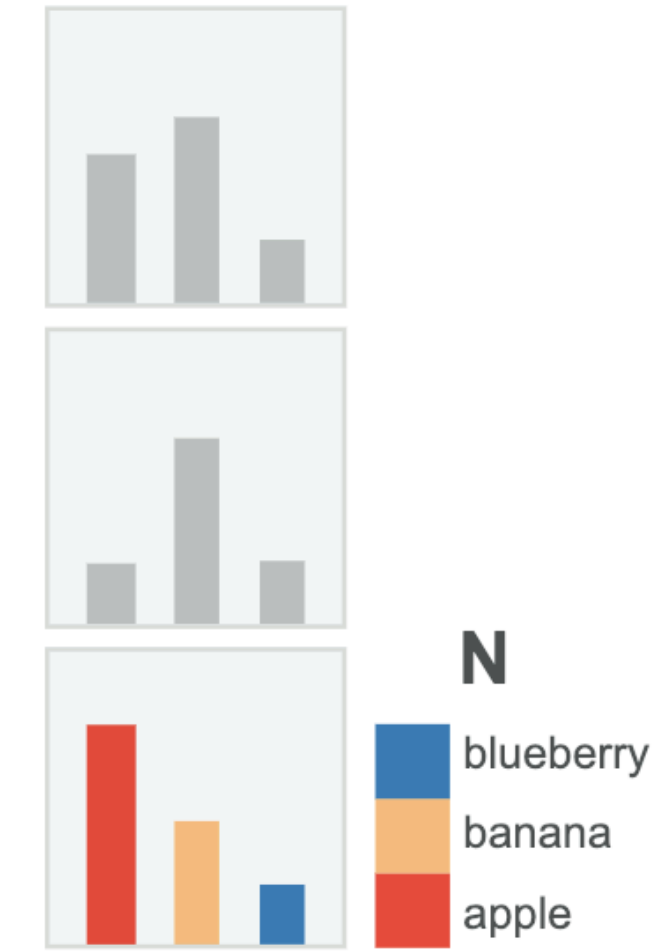
C1.3 Same Field, Same Quantitative Color Scale



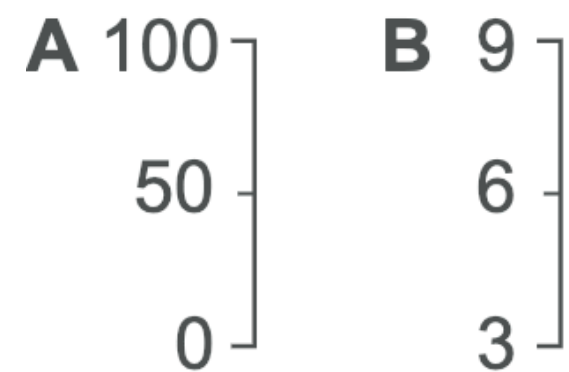
C1.4 Same Field, Same Value-Color Mapping



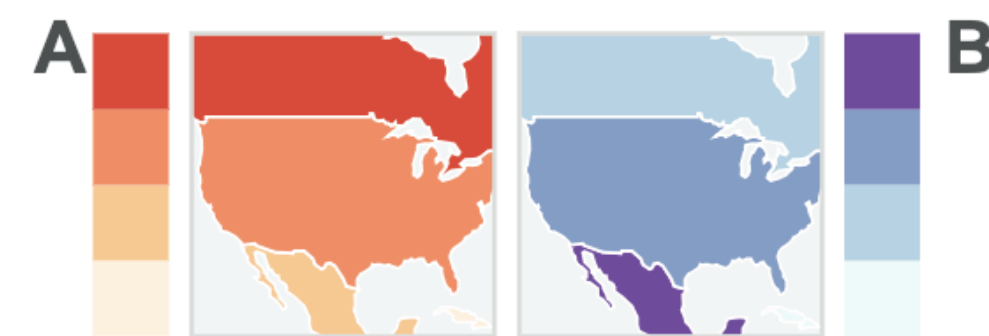
C2.5 Non-overlapping Nominal and Constant Colors



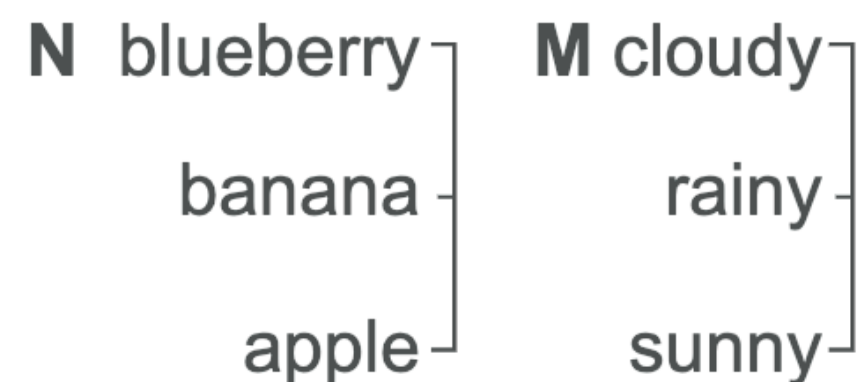
C2.1 Different Fields, Different XY Domains



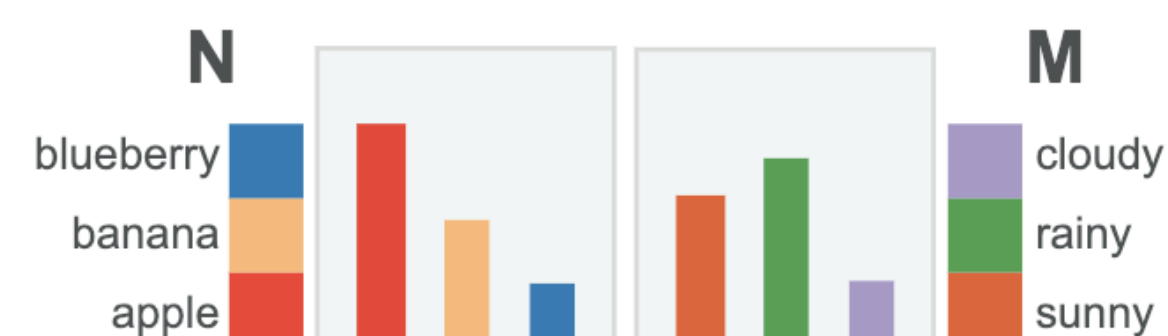
C2.3 Different Fields, Non-Overlapping Hues



C2.2 Different Fields, Different Nominal Values



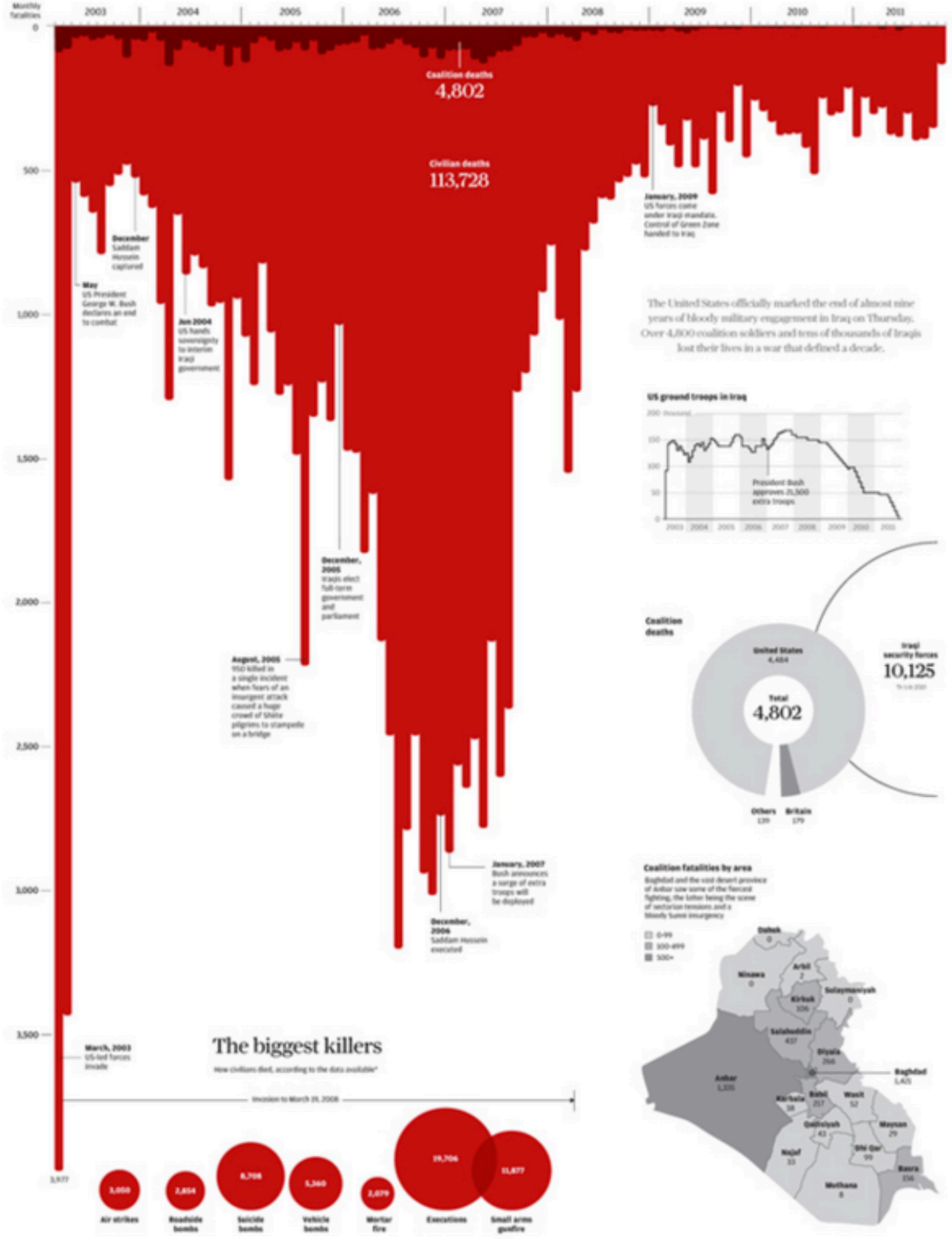
C2.4 Different Fields, Non-overlapping Palettes



Keeping Multiple Views Consistent

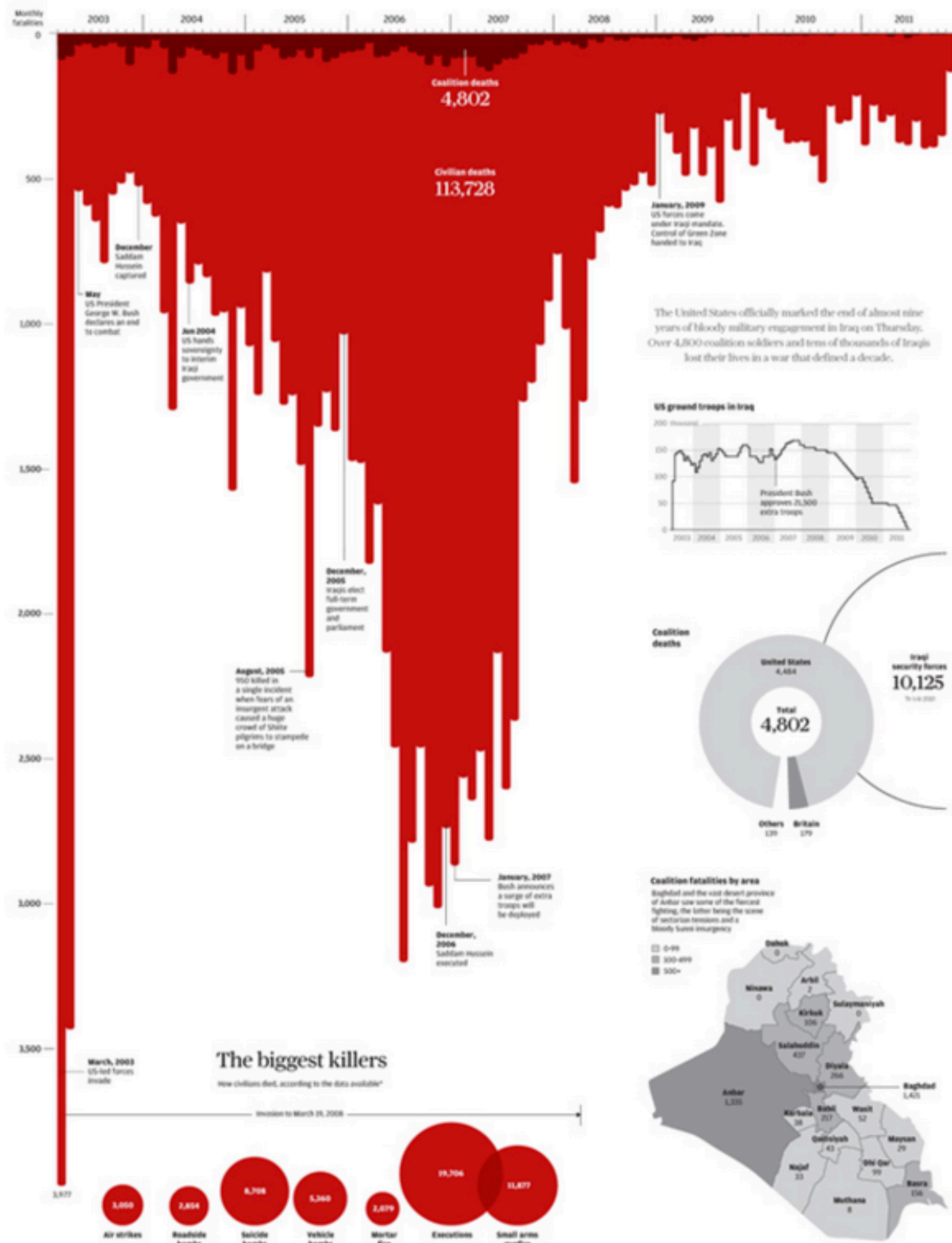
Storytelling, Double-edged Sword?

Iraq's bloody toll



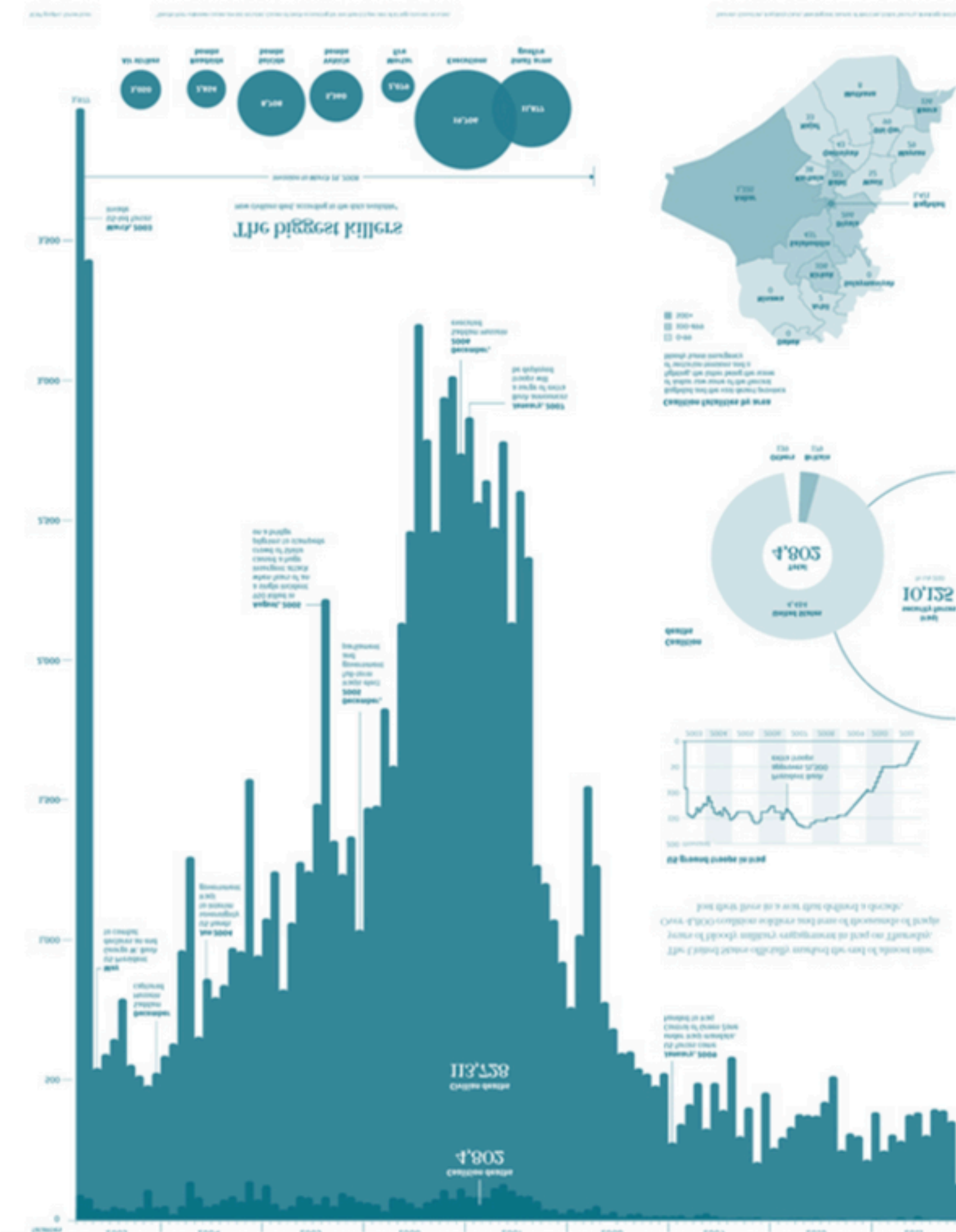
[South China Morning Post 2011]

Iraq's bloody toll



[South China Morning Post 2011]

Iraq: Deaths on the decline

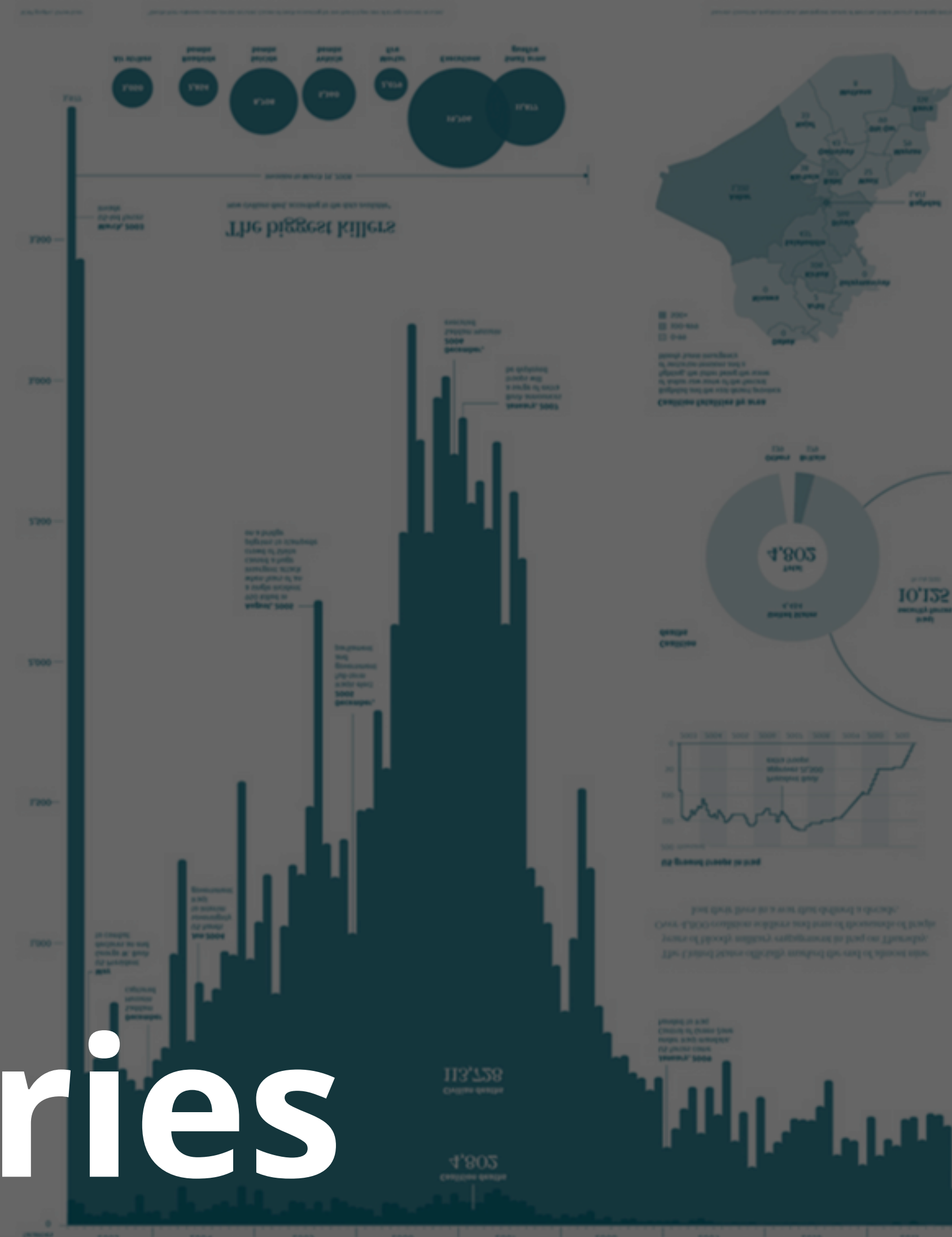


Flipped upside down....

Iraq's bloody toll



Iraq: Deaths on the decline



Same Data, Different Stories

[South China Morning Post 2011]

Flipped upside down....

Presentation & Storytelling in Tableau

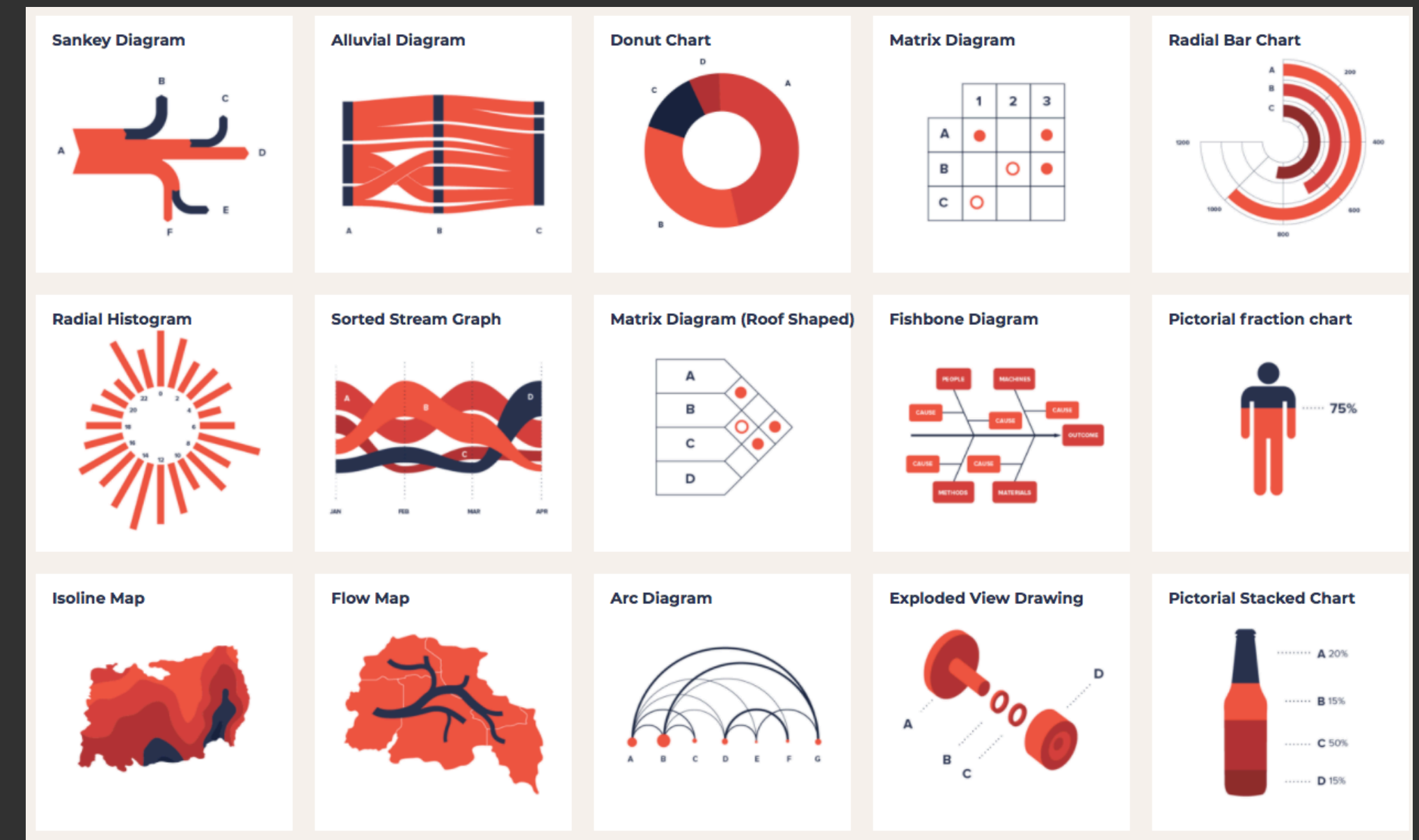
Dashboard

Annotation

Story Points

Next

Advanced visualizations



5 min break