



# Processing and exploring data

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# Challenge of The Course

- Every Week:  
Understanding complex social phenomena  
using big data
- This Week:  
Where do we get ‘big’ social data?  
What does it look like?  
How can we visualise it?



# Overview

- Data collection
- Data sources
- File formats and data types
- Descriptive statistics
- Visualisations



# Basic Research Framework

- Research Question & Hypothesis
  - RQ: What makes persuasion effective?
  - Hypothesis: Evidence makes persuasion effective
- Data Collection
  - Download Posts & Replies from Change My View Subreddit
- Sample Population
  - Redditors in Change My View as proxy for “people engaging in persuasion online”
- Methods & Analysis
  - Identify “Evidence” in replies (e.g., search for “http” links)
- Measure & Report Outcome
  - Persuasion rate of replies with evidence vs. without evidence



# Data Collection

- Asking
  - Surveys
  - Interviews
- Observing
  - Experiments
  - Big Data Analysis



# Data sources

Where do we get ‘big’ social data?

- APIs
- Scraping
- Existing datasets
  - Github, OSF, Kaggle, government open data portals



# Data sources: Existing data sets

- Most data sets available online were collected **ad hoc** for a specific purpose – useful for reproducing research, not so much for planning new research
- Some datasets can be used to answer a wider set of RQs
  - Historical example: MyPersonality data  
<https://sites.google.com/michalkosinski.com/mypersonality>
  - Example: Stanford Large Network Dataset Collection  
<https://snap.stanford.edu/data/>

By Jure Leskovec

STANFORD UNIVERSITY



**Stanford Large Network Dataset Collection**

- **Social networks** : online social networks, edges represent interactions between people
- **Networks with ground-truth communities** : ground-truth network communities in social and information networks
- **Communication networks** : email communication networks with edges representing communication
- **Citation networks** : nodes represent papers, edges represent citations
- **Collaboration networks** : nodes represent scientists, edges represent collaborations (co-authoring a paper)
- **Web graphs** : nodes represent webpages and edges are hyperlinks
- **Amazon networks** : nodes represent products and edges link commonly co-purchased products

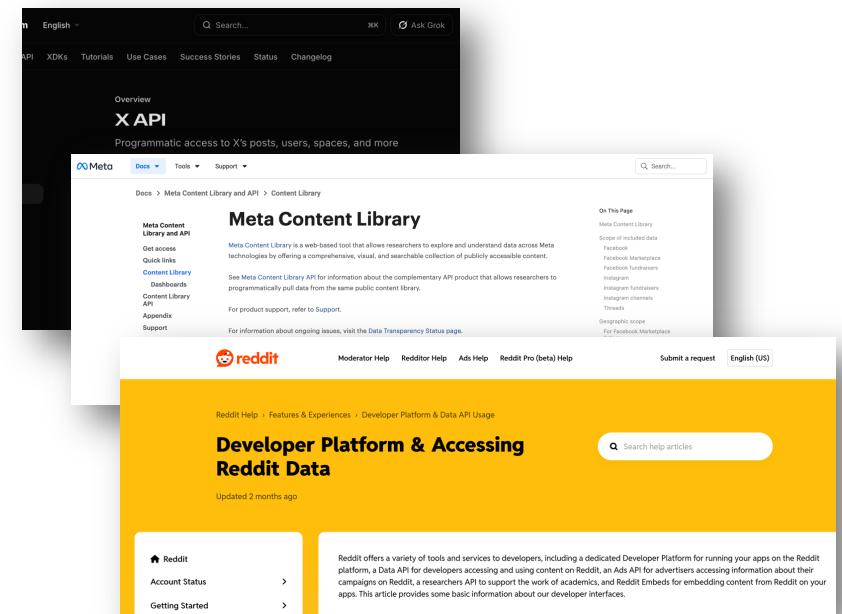
# Data sources: APIs

Often the first choice for collecting “new” data

APIs differ in

- Pricing (free / paid)
- Rate limiting (e.g. number of requests per minute)
- Scope of data available (comprehensive / limited)
- Exclusivity (access for everyone / selected applicants only)
- Documentation and support
- Reliability
- ...

API = application programming interface



# Data sources: Scraping

Often a “last resort” for obtaining data, but issues with

- Reliability
- Technical barriers
- Ethical considerations; ‘politeness’ (e.g. robots.txt)
- Data quality and consistency
- Possibly legal concerns (terms of service, privacy regulations e.g. GDPR, copyright, ..)

**-> Do not scrape without explicit permission from the website owner!**



# File formats

- CSV
- JSON
- XML
- HTML



# CSV (Comma-Separated Values)

```
id,neighbourhood,avg_rent
1,Old Town,1200
2,New Town,1350
3,Leith,900
4,Stockbridge,1100
5,Morningside,1300
6,Bruntsfield,1250
7,Marchmont,1150
```

- Tabular data: Columns and rows, like an Excel spreadsheet
- No hierarchical structure, or nested data



# JSON (JavaScript Object Notation)

```
[  
  {  
    "id": 1,  
    "neighbourhood": "Old Town",  
    "averageRent": 1200  
  },  
  {  
    "id": 2,  
    "neighbourhood": "New Town",  
    "averageRent": 1350  
  },  
  {  
    "id": 3,  
    "neighbourhood": "Leith",  
    "averageRent": 900  
  }]
```

- Key-value pairs
- Objects {} and arrays []
- Supports hierarchical structures
- Commonly used in data storage and exchange in web programming (e.g. between web server and client)



# XML (eXtensible Markup Language)

```
<Rents>
  <Neighbourhood id="1">
    <Name>Old Town</Name>
    <AverageRent>1200</AverageRent>
  </Neighbourhood>
  <Neighbourhood id="2">
    <Name>New Town</Name>
    <AverageRent>1350</AverageRent>
  </Neighbourhood>
  <Neighbourhood id="3">
    <Name>Leith</Name>
    <AverageRent>900</AverageRent>
  </Neighborhood>
</Rents>
```

- Tags and elements (root, parent elements, child elements)
- Hierarchical (tree structure)
- Commonly used in data storage and exchange



# HTML (Hypertext Markup Language)

```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <title>Average Rent in Edinburgh</title>
  <style>
    table {
      width: 50%;
      border-collapse: collapse;
      margin: 20px auto;
    }
    th, td {
      border: 1px solid #ddd;
      padding: 8px;
      text-align: left;
    }
    th {
      background-color: #f2f2f2;
    }
  </style>
</head>
<body>
  <h1>Average Rent in Edinburgh per Neighbourhood</h1>
  <table>
    <thead>
      <tr>
        <th>ID</th>
        <th>Neighbourhood</th>
        <th>Average Rent (£)</th>
      </tr>
    </thead>
    <tbody>
      <tr>
        <td>1</td>
        <td>Old Town</td>
        <td>1200</td>
      </tr>
      <tr>
        <td>2</td>
        <td>New Town</td>
        <td>1350</td>
      </tr>
      <tr>
        <td>3</td>
        <td>Leith</td>
        <td>900</td>
      </tr>
    </tbody>
  </table>
</body>
</html>
```

Tree structure (like XML) but

- Designed for creating web pages
- Focuses on presentation of content
- Predefined tags
- Can include CSS and JavaScript
- Not designed as a format for data storage / exchange



# Units of analysis

- Users
- Individual messages (e.g. Instagram posts, TikTok videos, ..)
- Interactions (e.g. friendship ties, retweets, replies, ..)
- Groups (e.g. subreddits, Facebook groups, ..)
- Geographical areas (neighbourhoods, countries, ..)

...

**Choose your unit of analysis wisely!**



# Data cleaning

- Handling missing data (removal, imputation..)
- Handling duplicates
- Standardisation; transforming variables
- Error correction; handling outliers
- Validation



# Data types

- Numerical data
  - Of a person: Age, income
  - Of a tweet: Number of retweets, ..
- Text data
  - Of a person: Name, Occupation, ..
  - Of a tweet: Text
- Network/relational data
  - Family relationships
  - Friendships
  - Follower / followee relationships
  - ...



# Descriptive statistics for numerical variables

$X = (1, 2, 7, 18, 23, 456, 1234)$

- Means

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad \bar{x} = \frac{1+2+7+18+23+456+1234}{2} \sim 248.71$$

- Medians

$$M = P_{50} = 23$$

1, 2, 7, 18, 23, 456, 1234  
~50% of data fall below this number

- Percentiles

$$P_{15} = 2$$

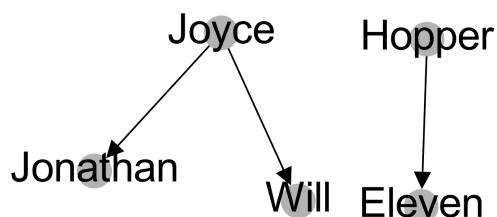
1, 2, 7, 18, 23, 456, 1234  
~15% of data fall below this number



# Representing network data as tabular data

## Edge list

parent,child  
Joyce,Will  
Joyce,Jonathan  
Hopper,Eleven



## Adjacency matrix

	JC	WI	JN	HO	EL
JC	0	1	1	0	0
WI	0	0	0	0	0
JN	0	0	0	0	0
HO	0	0	0	0	1
EL	0	0	0	0	0

- Each node is a row and a column
- “1” indicates a directed edge from row node to column node

## Incidence matrix

JC	-1	-1	0
WI	1	0	0
JN	0	1	0
HO	0	0	-1
EL	0	0	1

- Each node is a row
- Each edge is a column
- “-1” for outgoing edges, “1” for incoming edges



# Turning text data into numerical data

Five documents...

... represented as fixed-length document vectors



# Turning text data into numerical data

**Term vectors** represent **words** as fixed-length vectors:

- Sparse (most values are 0)
- Incidentally capture semantics (similar vectors are terms that appear together)

he	drink	ink	likes	pink	think	wink	
2	1	0	2	0	0	1	← D1: He likes to wink, he likes to drink
1	3	0	1	0	0	0	← D2: He likes to drink, and drink, and drink
1	1	1	1	0	1	0	← D3: The thing he likes to drink is ink
1	1	1	1	1	0	0	← D4: The ink he likes to drink is pink
1	1	1	1	1	0	1	← D5: He likes to wink, and drink pink ink



# Word embeddings

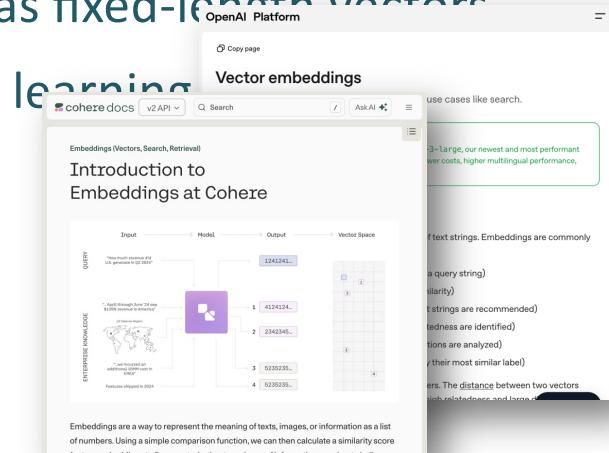
More complex representation of **words** as vectors

- Dense (all entries are non-zero)
- Capture semantics even better (similar words have similar vectors)

he	drink	ink
0.123	0.521	0.313
0.451	0.987	0.812
0.938	0.141	0.411
...	...	...

(many dimensions e.g. 300)

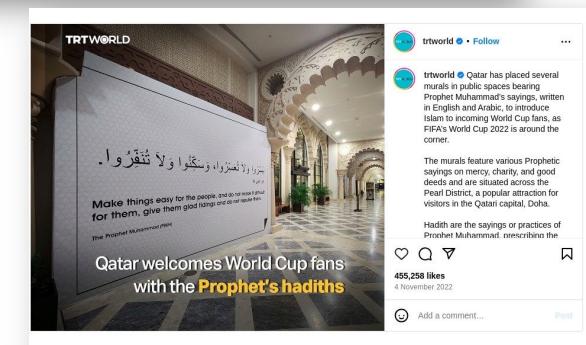
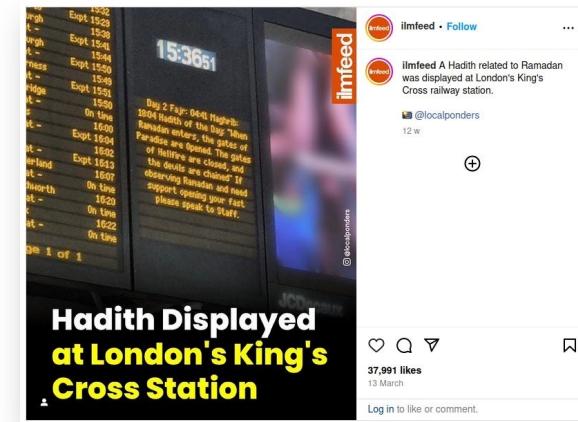
- Modern embeddings can represent entire **sentences, paragraphs or documents** as fixed-length vectors
- Obtained through machine learning collection of training data
- Pre-trained embeddings available online:



# Case study: Religious discussions online

- Hadiths are the **recorded actions and words of the prophet Muhammad**
- One of the main sources of religious knowledge in Islam
- Widely used in daily life, also by politicians
- We wanted to know: **Who is using hadith on social media, how and why?**

Mahmoud Fawzi, Walid Magdy, and Björn Ross. 2025. "The Prophet said so!": On Exploring Hadith Presence on Arabic Social Media. In *Proceedings of ACM on Human-Computer Interaction (CSCW '25)*. ACM, New York, NY, USA, Article CSCW192, 23 pages. <https://doi.org/10.1145/3711090>

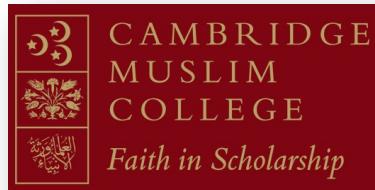


# Religious discussions online: Research framework

- Research Questions
  - Which hadiths are most frequently shared by users of Arabic social media?
  - What topics are they about? Are they authentic or fabricated? When do they share them?
- Data Collection
  - Existing tweet dataset from archive.org (originally obtained from Twitter API)
  - Hadiths with topic categories scraped with permission from [www.sonnaonline.com](http://www.sonnaonline.com)
  - Authenticity data from existing dataset (LK Corpus + MAHADDAT)
- Sample Population
  - Tweets in Arabic that contain specific phrase
- Methods & Analysis
  - Match tweets to hadiths (Jaccard similarity)
  - Calculate seasonality (Gini coefficient)
- Measure & Report Outcome
  - Compare topics in hadiths shared on social media with topics in all hadiths
  - Report most seasonal hadiths, authenticity distribution



# Religious discussions online: Data collection



ABOUT PROGRAMMES RE



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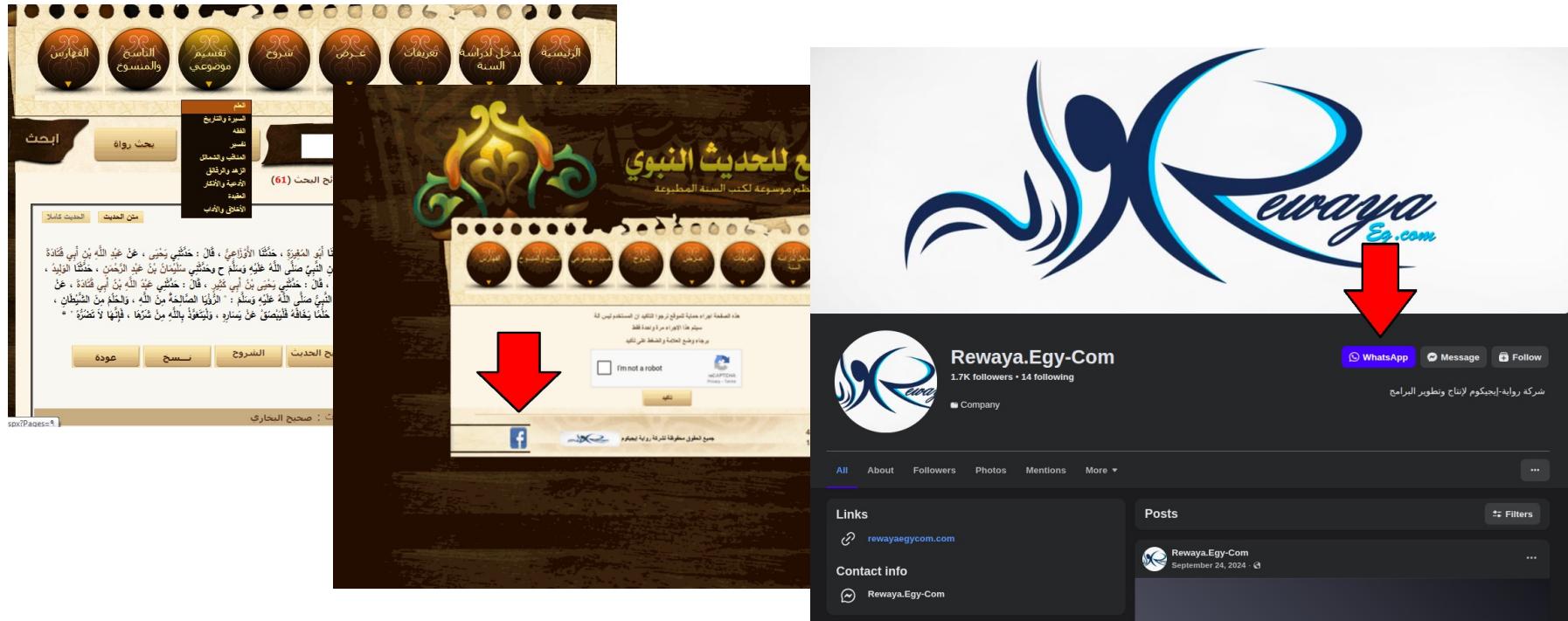
b.a@cambridgemuslimcollege.ac.uk

Website



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# Religious discussions online: Data collection



# Data visualisation

- Data visualization is often the first step in a project
  - What kind of **variation** is present in the data?
  - What are key **comparisons** between groups?
  - What kinds of **problems** can we anticipate (e.g. missing data, data not what we expected, selection effects)



Image: [thenewstack.io](http://thenewstack.io)



# Why visualize?

- The table shows the trade balance between England and Denmark/Norway in the 1700s
- In what year did the trade balance begin to favour England (exports higher than imports)?

Source: <https://data.europa.eu/apps/data-visualisation-guide/data-visualisation-is-accessibility>

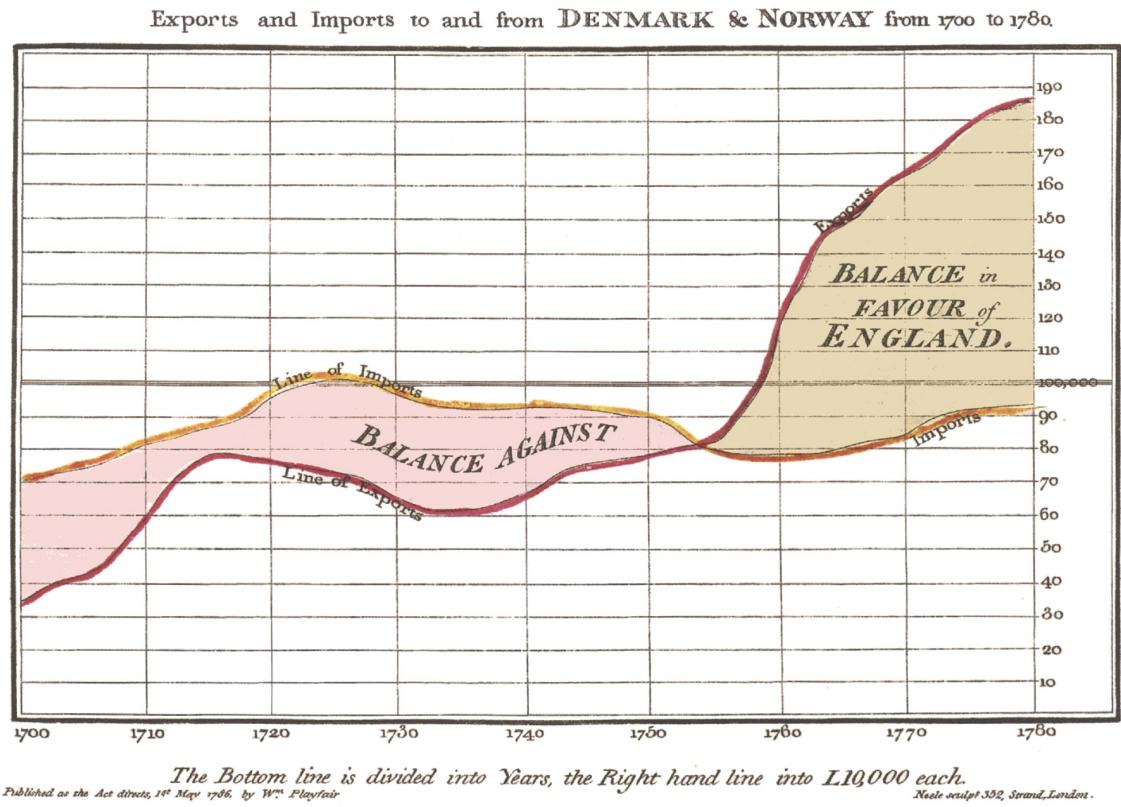
Year	Imports	Exports
1700	71.1	32.8
1705	74.5	40.9
1710	82.6	59
1715	87.2	77.9
1720	96.8	75.2
1725	102.6	71.3
1730	96.4	64.7
1735	93.7	60.5
1740	92.9	65.1
1745	92.5	74.3
1750	90.1	77.4
1755	79.9	82.8
1760	76.6	117.5
1765	79.6	151.8
1770	83.8	163.8
1775	90.4	175.7
1780	92.7	185.4



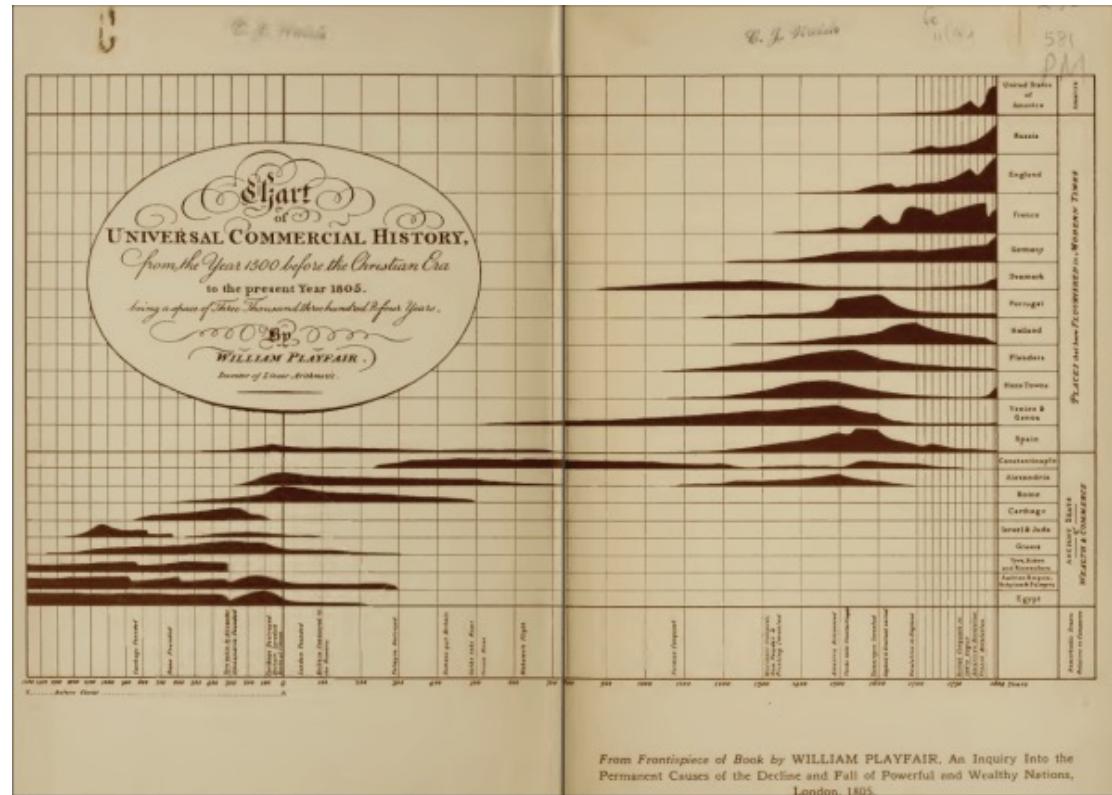
# Why visualize?

- Patterns in data are often much easier to see graphically than numerically.
- Famous early data visualization by Scottish Engineer William Playfair

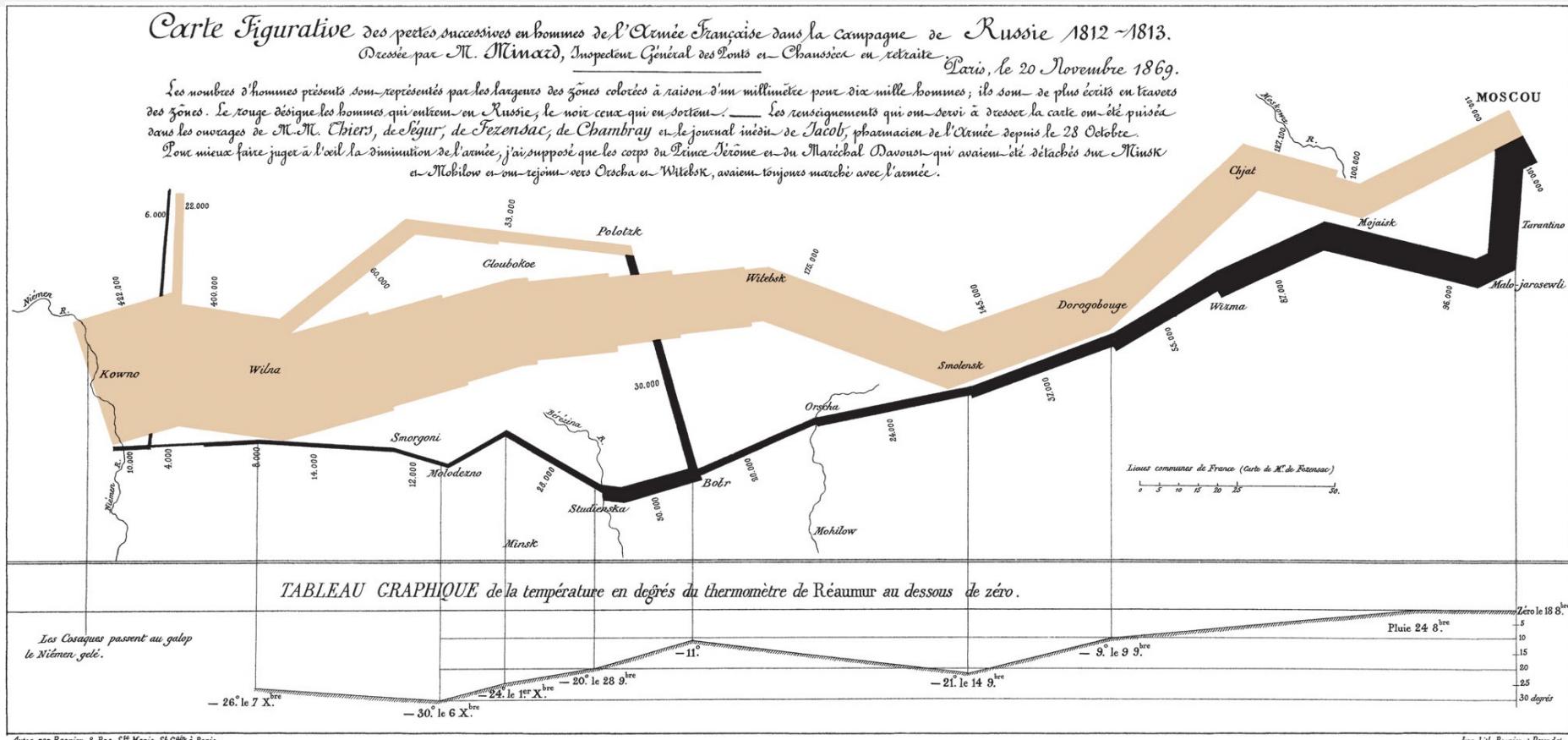
Image source:  
[https://commons.wikimedia.org/wiki/File:Playfair\\_TimeSeries-2.png](https://commons.wikimedia.org/wiki/File:Playfair_TimeSeries-2.png)



- Playfair's visualisations exemplify many aspects of modern data visualization best practice.
- Here, the idea of **small multiples**.

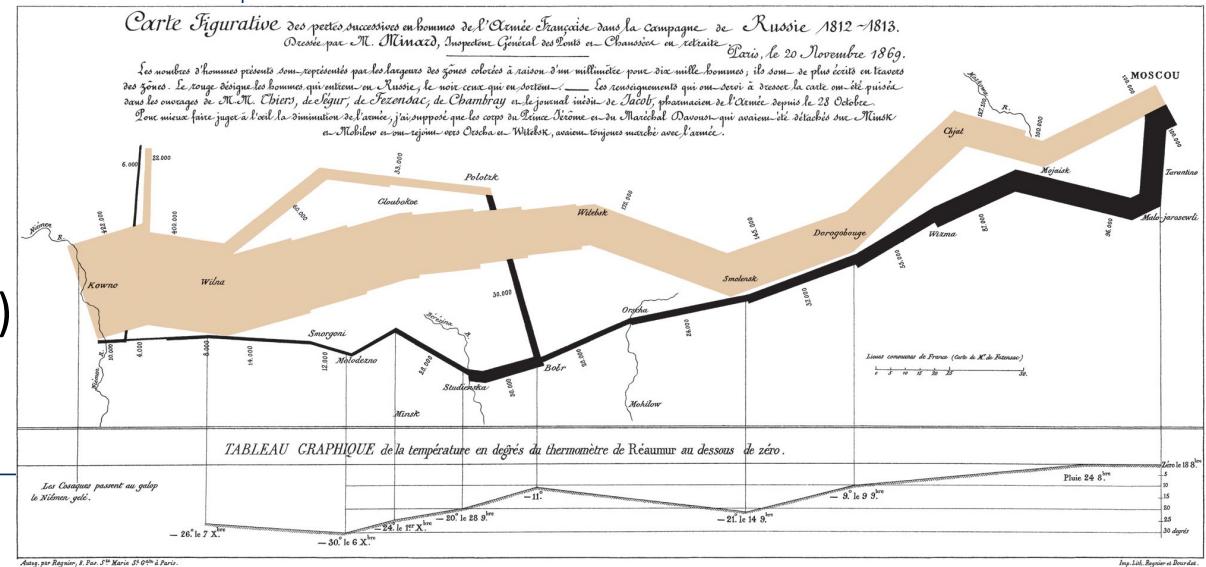


# Napoleon's march



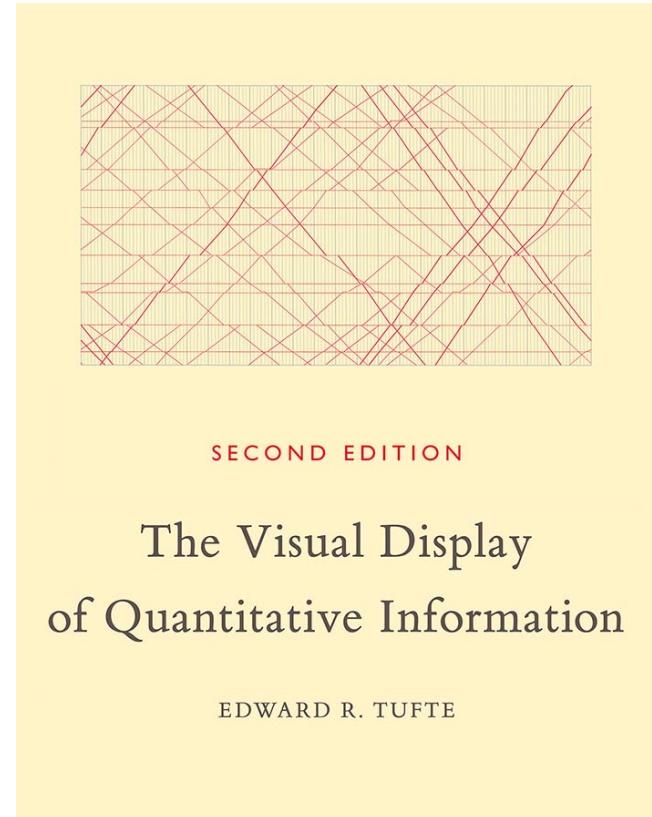
# 'The best statistical graphic ever drawn'

- Pierre Minard's depiction of Napoleon's retreat shows several variables:
  - Size of the army (width)
  - Location of the army (position)
  - Direction of movement (colour)
  - Temperature (secondary line plot)



# The science of data visualisation

- Data visualisation can look pretty, but it is not just aesthetic
- There are principles of clear, effective, honest visualization
- Data-ink ratio:
  - Show the data
  - Minimise not-data
  - Minimise redundancy



# Failure to minimize not-data

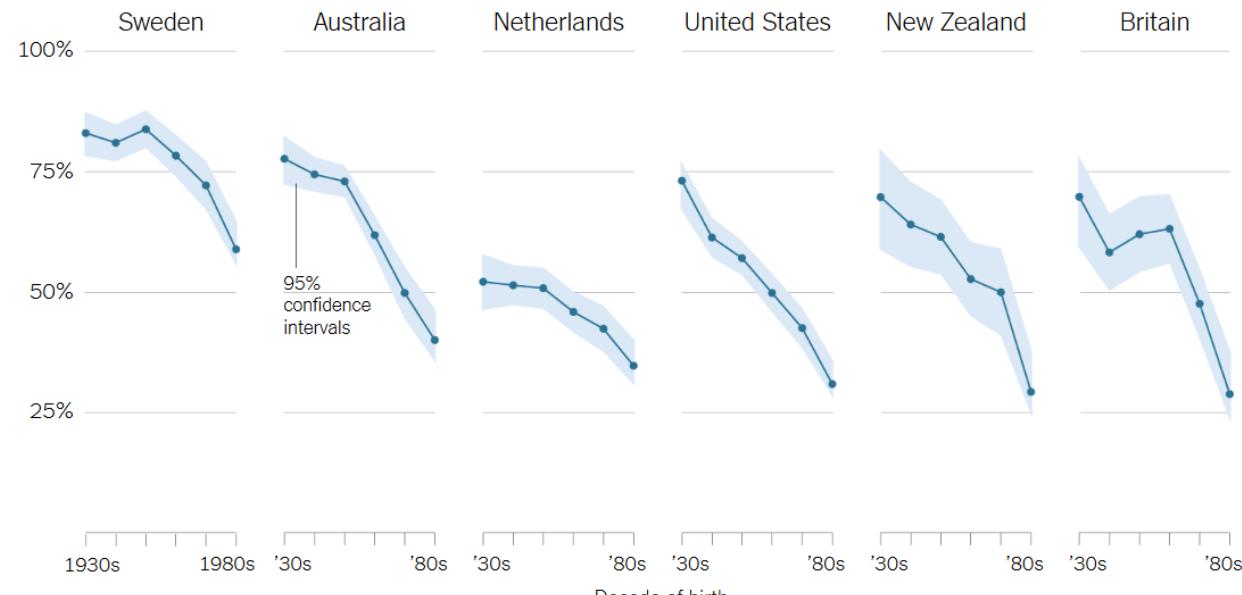
- Tufte argues against gimmicky graphics with a low data-ink ratio ('chartjunk')
- How else could we represent these data?

Example: Healy, Data Visualisation



# What does this graph show?

Percentage of people who say it is “essential” to live in a democracy



Source: Yascha Mounk and Roberto Stefan Foa, “The Signs of Democratic Deconsolidation,” Journal of Democracy | By The New York Times

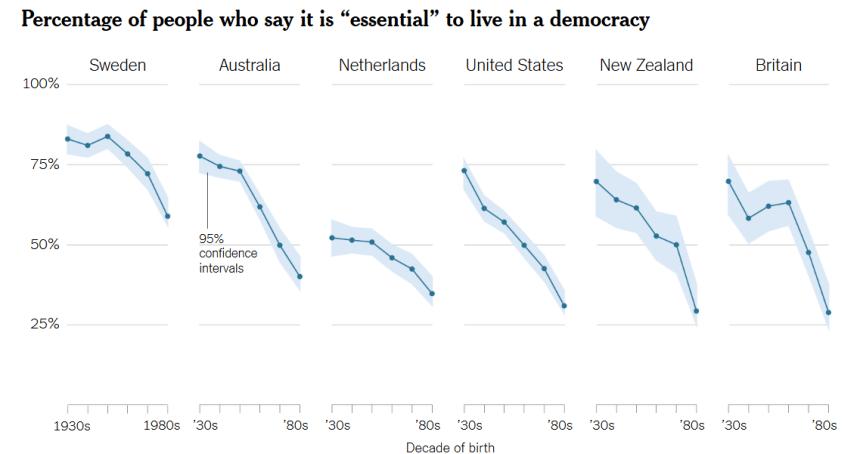
This and the following draw on Healy, Data Visualization



# Always read the fine print

- The figure shows the proportion of people who answered “10” to the question: “How important is it to you to live in a country that is governed democratically”
- The horizontal (x) axis shows year of birth, not year of survey
  - Differences between age cohorts, not change over time

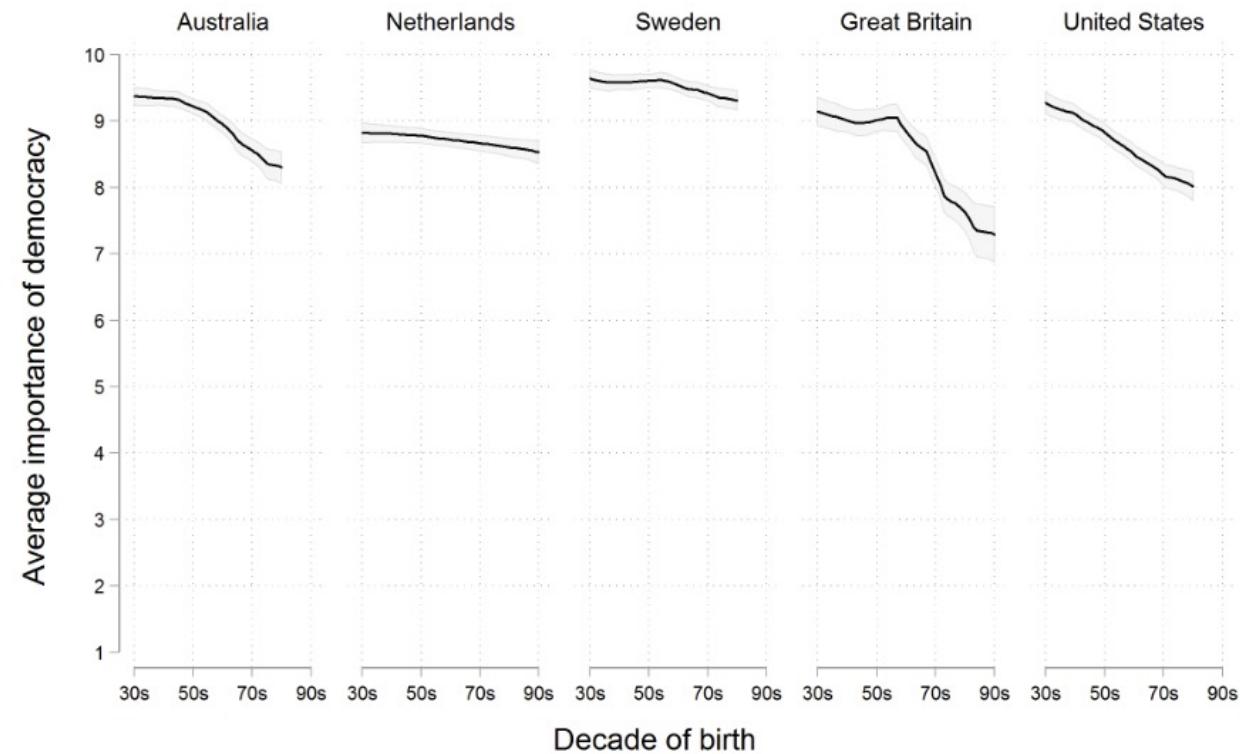
Example from Healey, Data Visualisation



Source: Yascha Mounk and Roberto Stefan Foa, “The Signs of Democratic Deconsolidation,” Journal of Democracy | By The New York Times



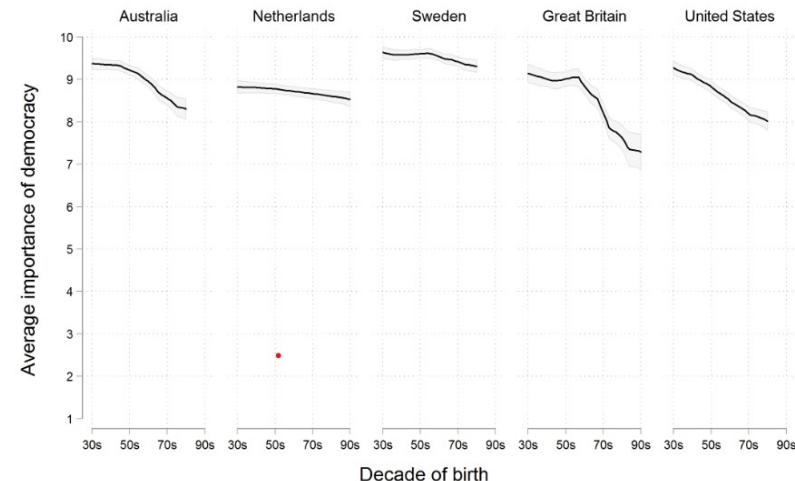
# How NOT to lie with graphics



Graph by Erik Voeten, based on WVS 5

# Visualisation choices matter

- The alternative plot shows the same data, but plots the average value of the 10-point scale (rather than the proportion of 10s)
- Suggests decline of importance of democracy, but not as much.
- Respondents still overwhelmingly say it is important to live in a democracy.

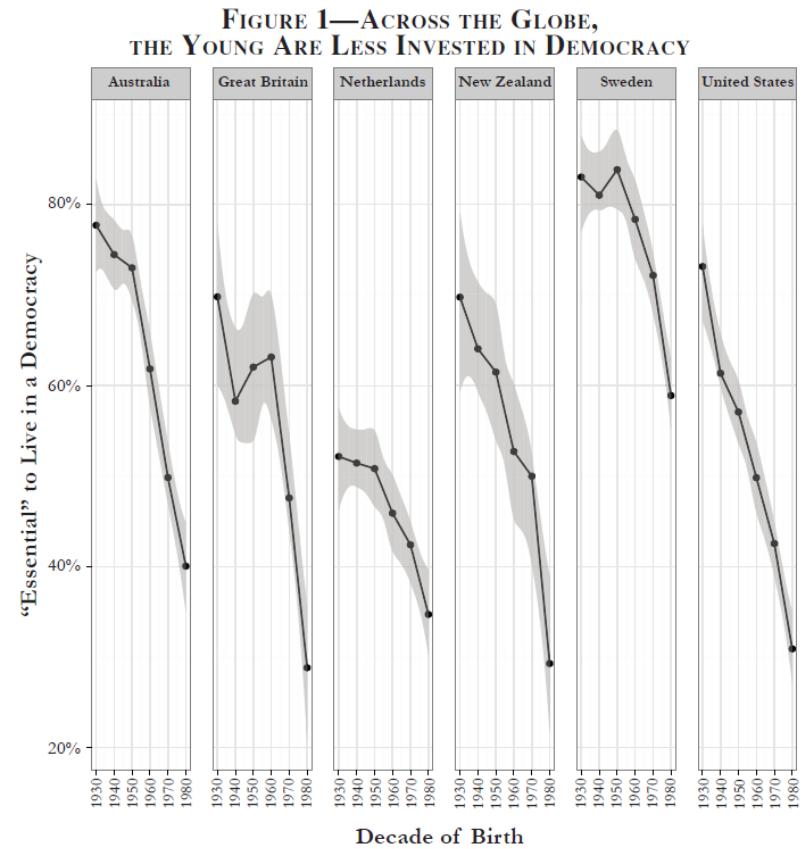


Graph by Erik Voeten, based on WVS 5



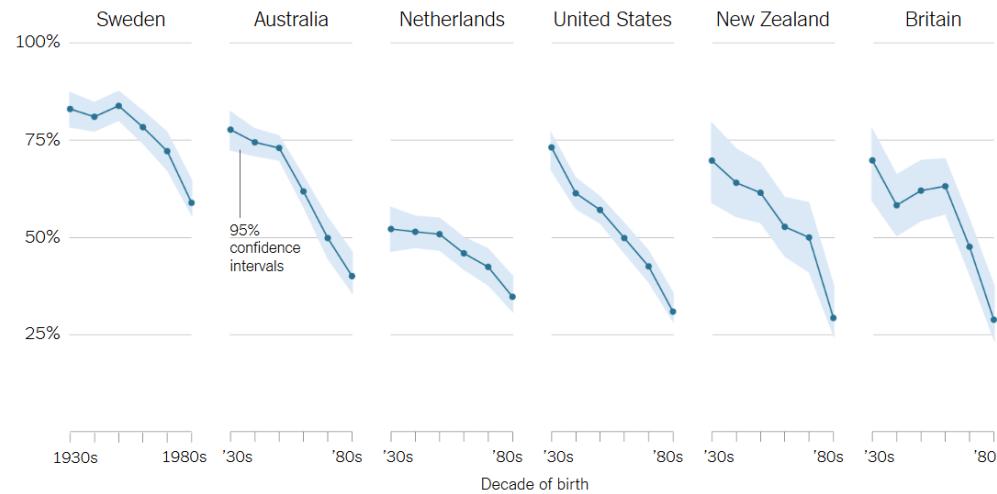
# Coordinate space matters

- The figure in the originally published article had another problem: a compressed **aspect ratio** accentuated the impression of decline
- Too much vertical space, not enough horizontal space



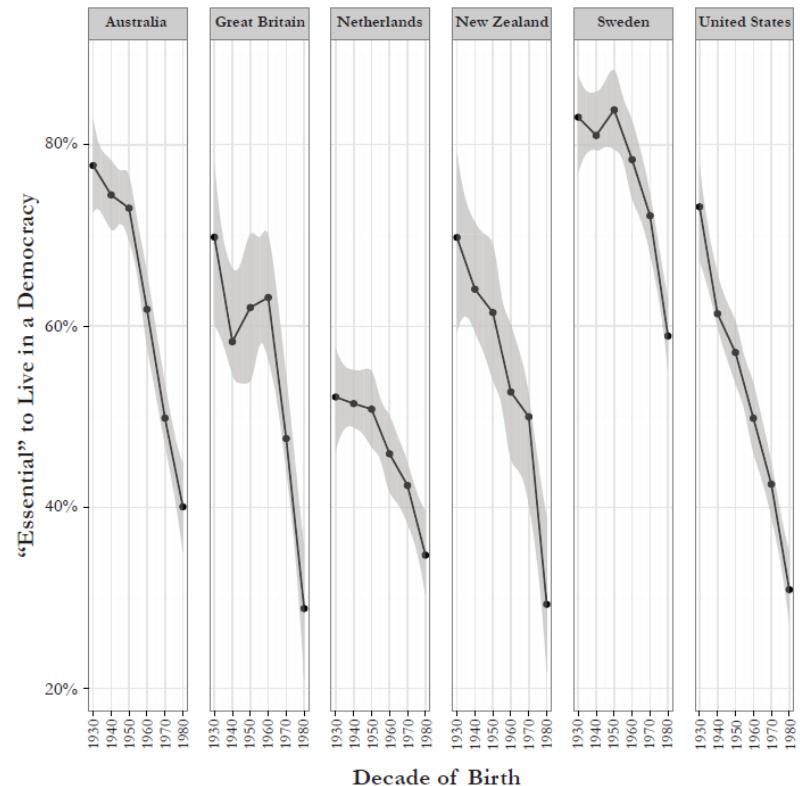
# Two versions of the same plot

Percentage of people who say it is “essential” to live in a democracy



Source: Yascha Mounk and Roberto Stefan Foa, “The Signs of Democratic Deconsolidation,” *Journal of Democracy* | By The New York Times

FIGURE 1—ACROSS THE GLOBE, THE YOUNG ARE LESS INVESTED IN DEMOCRACY

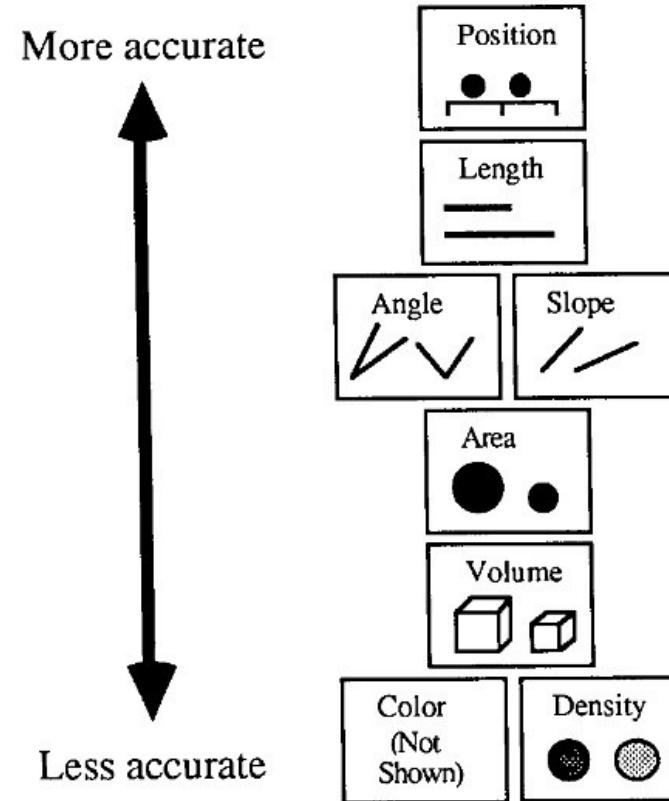


# Perceptual accuracy

Most to least accurate:

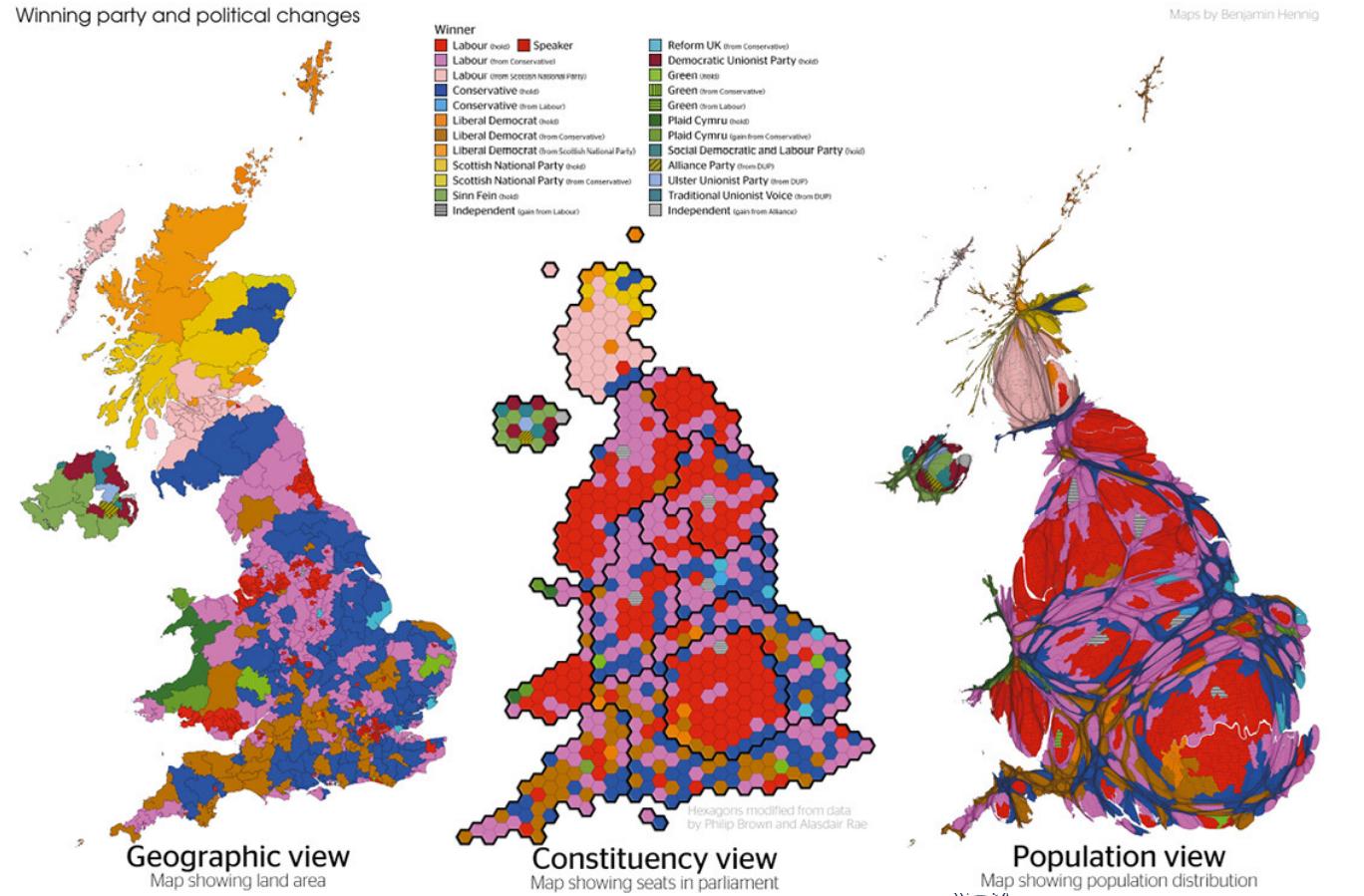
- Position on a common scale
- Length
- Direction
- Angle and slope
- Area
- Volume
- Density, color

Source: Mackinlay (1986); ucdavisdatalab.github.io



# Using colour

- Colour can convey key information, but is also hard to perceive clearly.



# The problem with colour

- But not everyone sees colour in the same way
- Do you need colour in your plot? If not, don't use it.

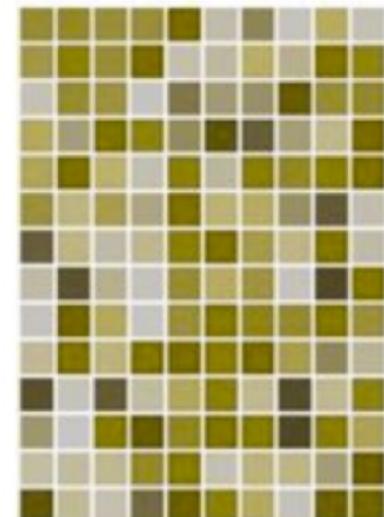
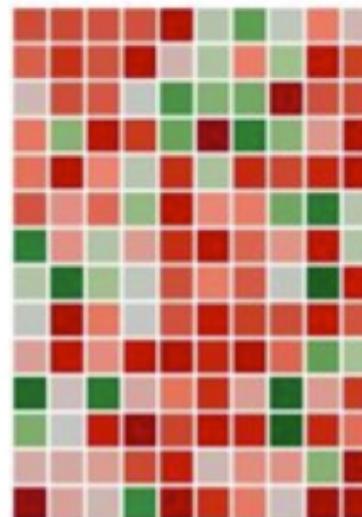
White and gold or blue and black? A 2015 meme →



# Colour blindness

- About 8% of men and 0.5% of women have some form of “colour-blindness”
- Most common form: red-green colour blindness
- Avoid red-green contrasts

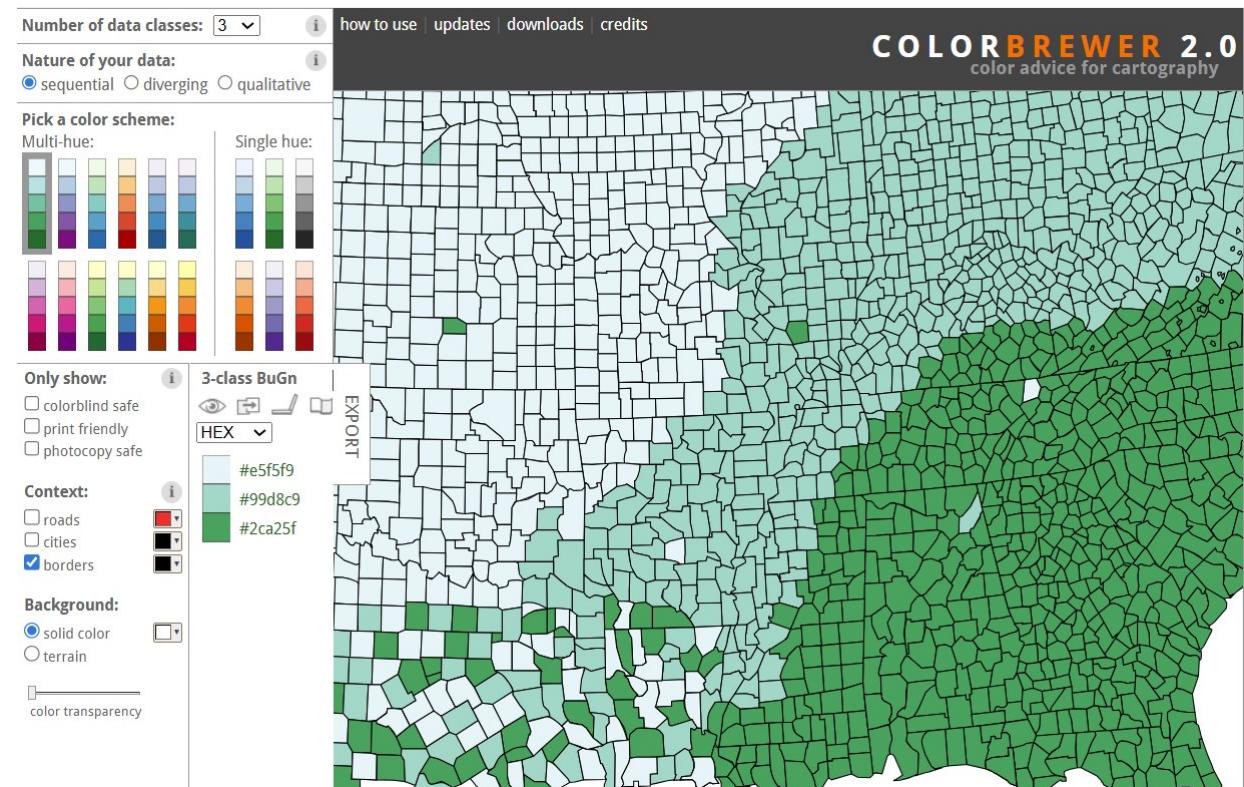
Source: <https://www.tableau.com/en-gb/blog/examining-data-viz-rules-dont-use-red-green-together>



# Choose and appropriate palette

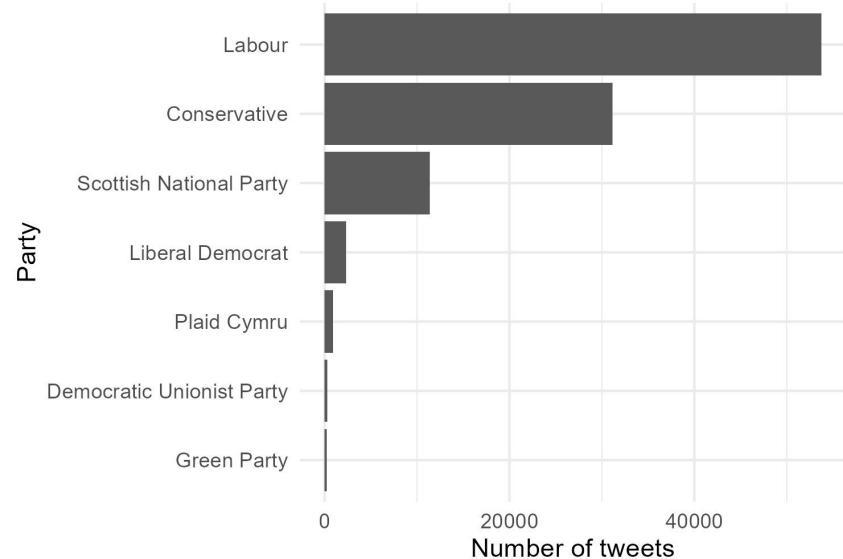
- When using color, pick an appropriate palette:
  - Sequential
  - Diverging
  - Qualitative

See [colorbrewer2.org](http://colorbrewer2.org)



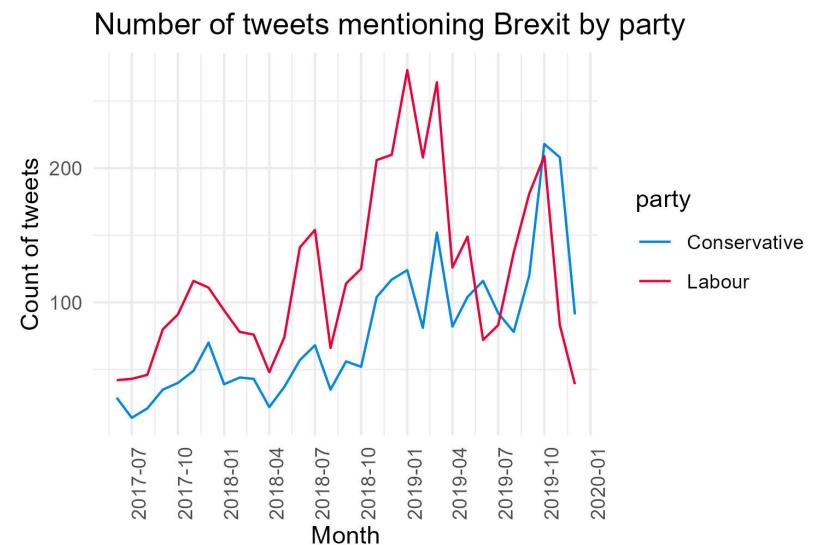
# Common plot types: bar charts

- Tried and true choice – if it works, use it
- Good for comparison between groups
- Works with more than one grouping factor (grouped bar chart)



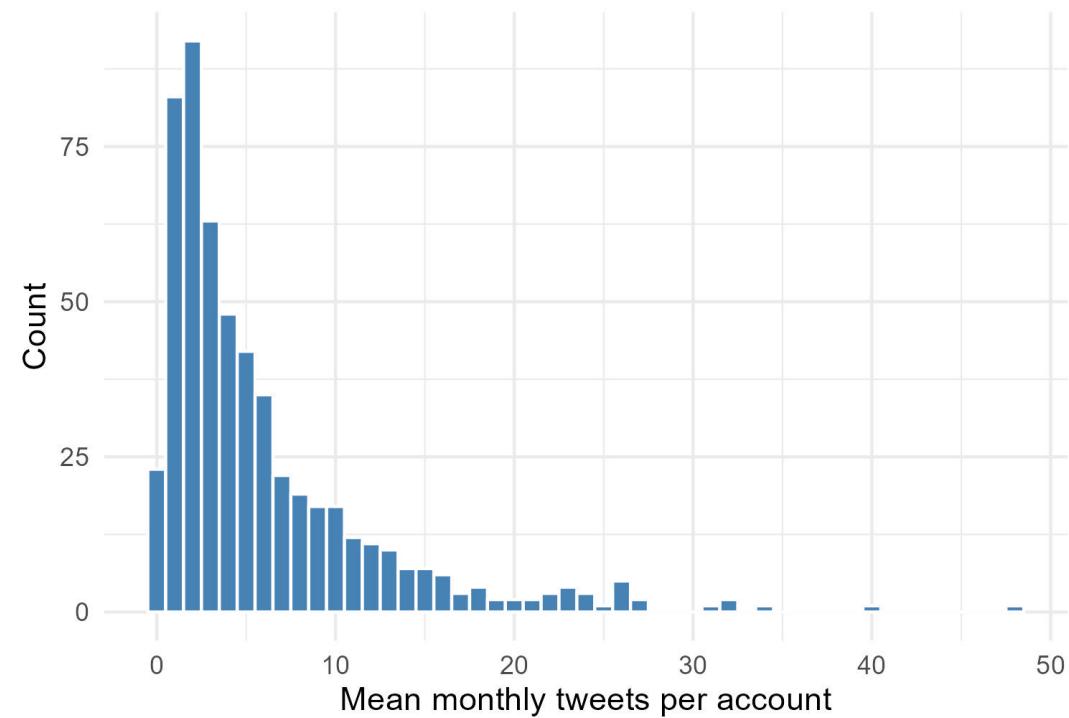
# Common plot types: line plots

- Good for time trends
- Can add group information using colour or other elements



# Common plot types: Distributions

- Visualising the distribution of key variables of interest can inform analysis
- A histogram shows the number of observations falling bins (small increments of the data)
- Alternatives: density plot, violin plot



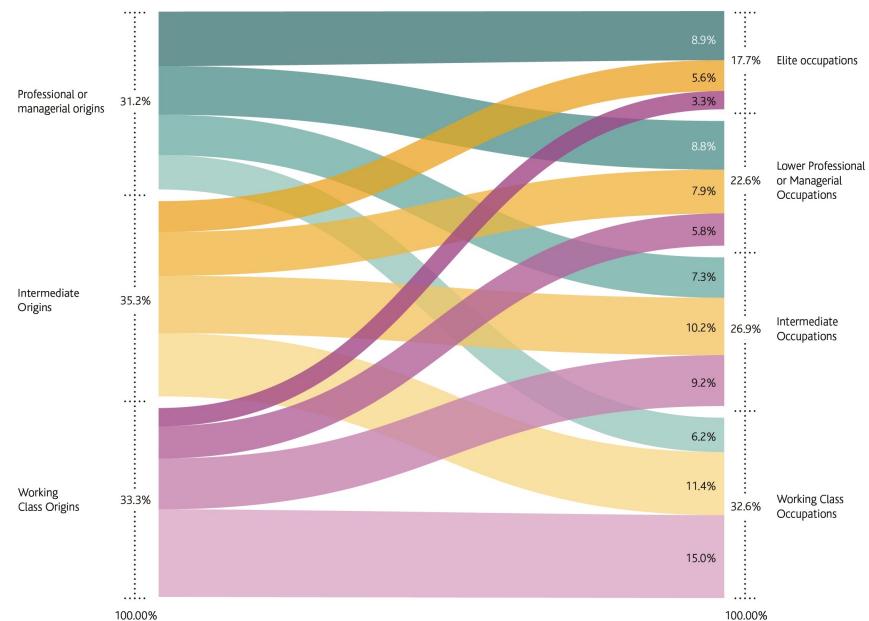
# Small multiples

- Tufte: “At the heart of quantitative reasoning is a single question: Compared to what?”
- Small multiple: a series of similar plots showing the difference (in trend, distribution, etc) by relevant group variables.



# Visualising flows

- A Sankey plot is a good way to visualize flows (changes from one state to another)
- This plot: visualising social mobility (change in class from parent's to children's generation)



# Animated plots

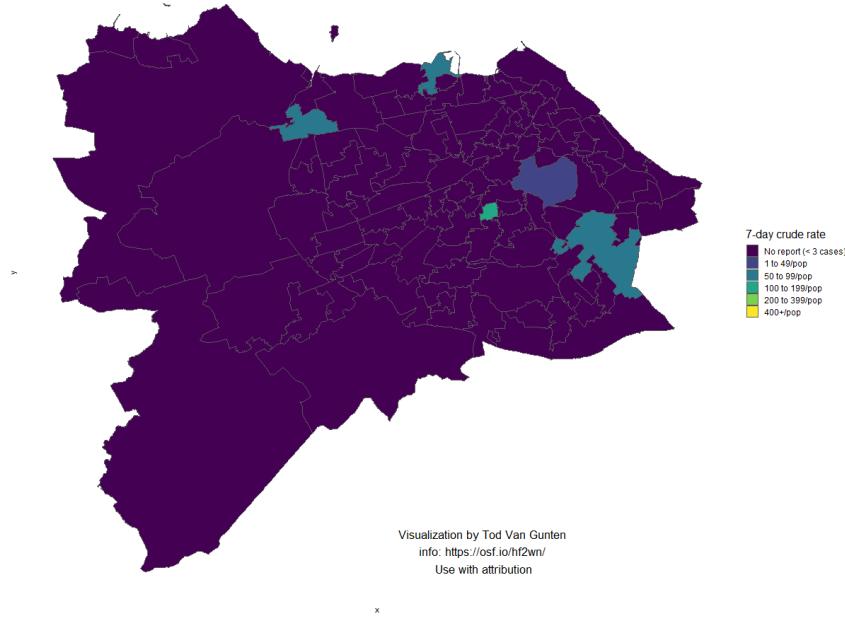
- Animated plots can make it easier to spot trends



# Animated plots

- Choropleth maps: represent spatially-distributed data using colour and familiar geographic outlines.
- Animated version can

Covid-19 rate by neighbourhood in Edinburgh  
2020-09-02



# Religious discussions online: visualisations

CSCW192:10

Mahmoud Fawzi, Walid Magdy, and Björn Ross

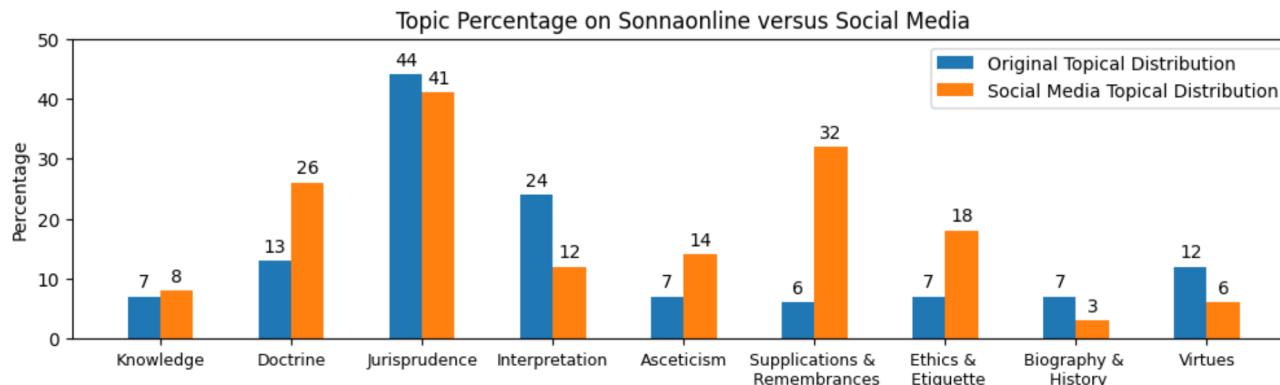


Fig. 4. The topical categories distribution of Hadith on Arabic social media from January 2019 to January 2023 versus on [sonnaonline.com](http://sonnaonline.com) (Note: The sum of percentages exceeds 100 because a hadith can belong to multiple categories)



# Religious discussions online: visualisations

Exploring Hadith Presence on Arabic Social Media

CSCW192:11

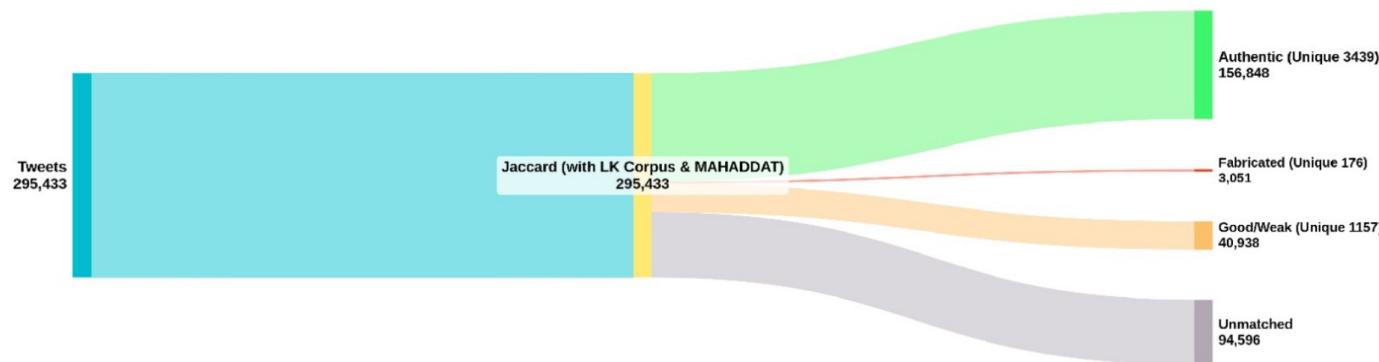


Fig. 5. The authenticity distribution of hadiths on Arabic social media from January 2019 to January 2023



# Religious discussions online: visualisations

Exploring Hadith Presence on Arabic Social Media

CSCW192:13

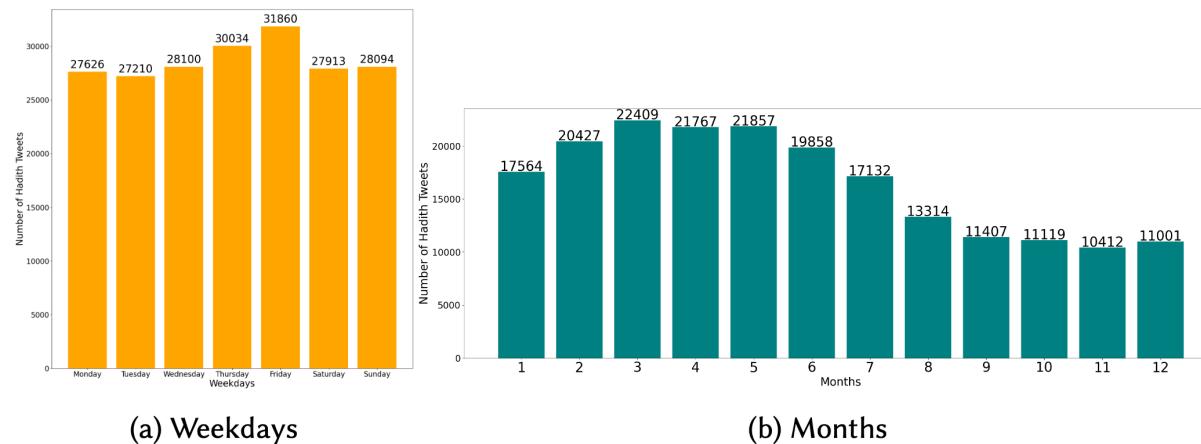
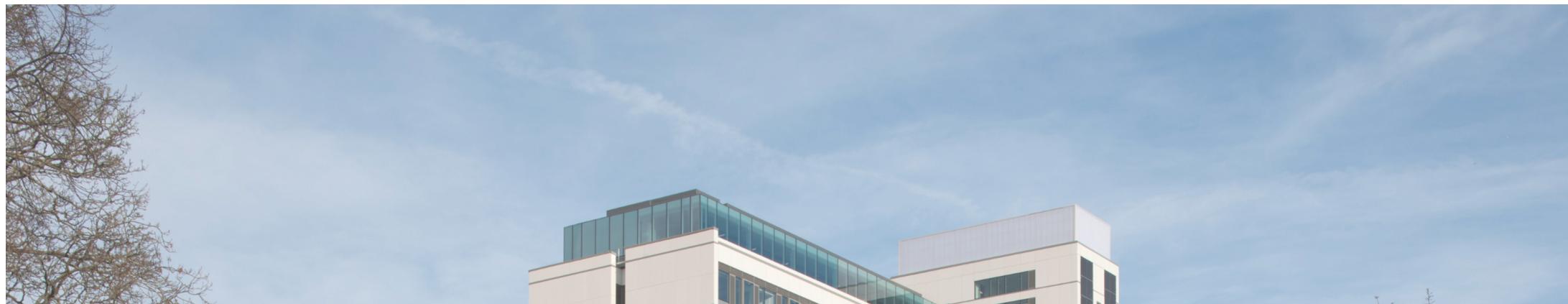


Fig. 6. The distribution of hadiths over weekdays and months. **Note:** Data for January 2023 is filtered for (b) so that all months have an equal number of occurrences.



# Questions?



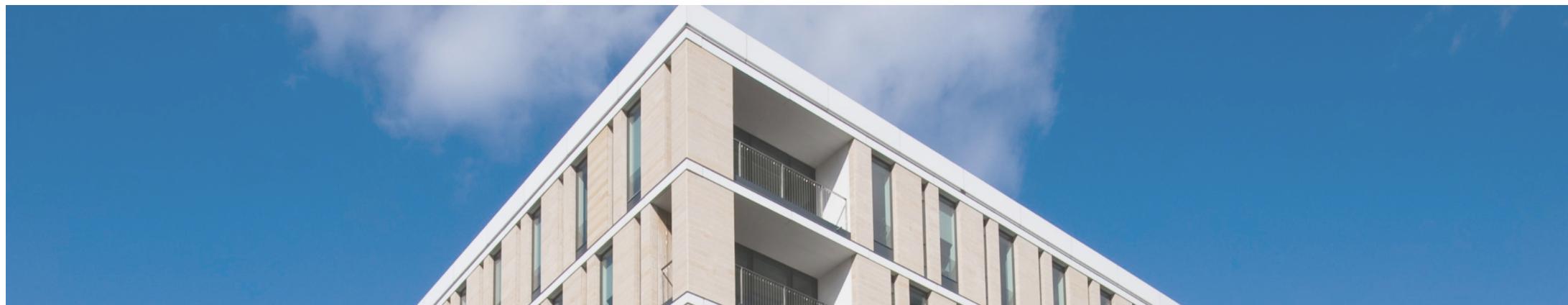


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