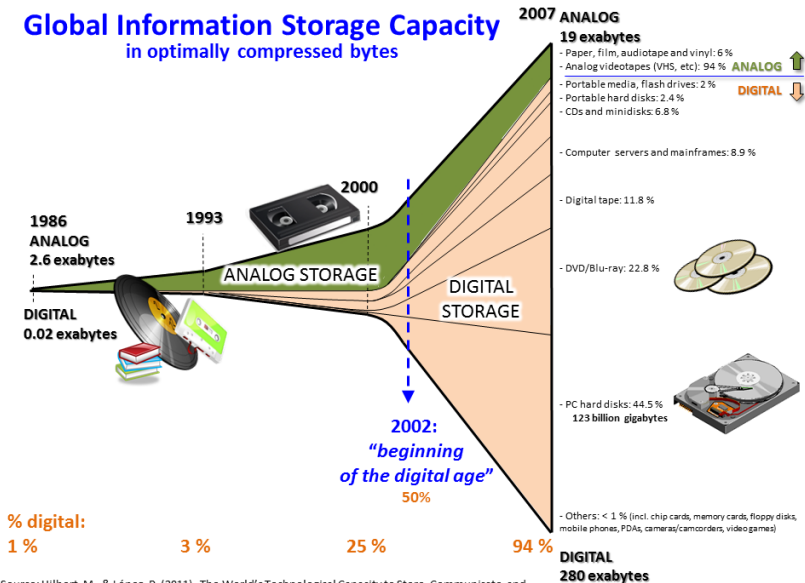


# Building a Robot Judge: Data Science for the Law

## 3. Text Data Essentials

Elliott Ash

# Global Information Storage Capacity in optimally compressed bytes



Source: Hilbert, M., & López, P. (2011). The World's Technological Capacity to Store, Communicate, and Compute Information. *Science*, 332(6025), 60 –65. <http://www.martinhilbert.net/WorldInfoCapacity.html>

# New Data, New Possibilities

## European Parliament Members' Twitter Networks *by country*

402 Twitter accounts of MEPs  
8,579 follower relations  
Node size = indegree  
Color = country

Accounts by country  
(in order of user numbers)

- France
- United Kingdom
- Germany
- Poland
- Italy
- The Netherlands
- Sweden
- Spain
- Belgium
- Portugal
- Romania
- Austria
- Other

CC BY-SA 4.0 — Axel Mairreder and Stephan Schlögl (University of Vienna)

universität  
wien



# Diversification of Text Data Methods

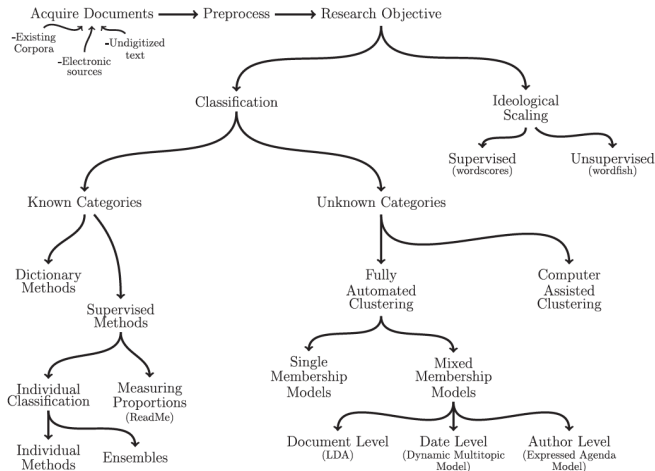


Fig. 1 An overview of text as data methods.

Source: Stewart and Grimmer (2013).

# Overview

- ▶ Input:
  - ▶ A set of documents (e.g. text files),  $D$ .
- ▶ Output:
  - ▶ A matrix,  $X$ , containing statistics about phrase frequencies in those documents.

# Text as Data

- ▶ Text data is a sequence of characters called **documents**.
- ▶ The set of documents is the **corpus**.
- ▶ Text data is **unstructured**:
  - ▶ the information we want is mixed together with (lots of) information we don't.
  - ▶ How to separate the two?
- ▶ All text data approaches will throw away some information:
  - ▶ The trick is figuring out how to retain valuable information.

# Documents and metadata

- ▶ For small corpora, you might have the text and metadata together in a spreadsheet.
- ▶ For larger corpora, you might have:
  - ▶ A document is a text file (or an item in a relational database).
  - ▶ A corpus is a folder of text files.
  - ▶ The filenames for the text files should contain an identifier for linking to metadata.

# What counts as a document?

- ▶ The unit of document analysis will vary depending on your question.
- ▶ If you are looking at how judges decide different types of cases, then a case would be a document.
- ▶ If you are looking at how judges differ within a court, then you might aggregate all of a judge's cases as a document.
- ▶ If you are looking at the impact of court cases on crime in a year, you might aggregate all the cases in a single year as a single document.
- ▶ If you are looking at how different topics are discussed within single cases, then a document might be a section or a paragraph.



# Publicly Available Corpora

- ▶ There is already a vast amount of data out there that has already been compiled (e.g. CourtListener, Twitter, New York Times, Reuters, Google, Wikipedia).
- ▶ Chris Bail curates a list of these datasets:
  - ▶ <https://docs.google.com/spreadsheets/d/1I7cvuCBQxosQK2evTcdL3qtglaEPc0WFEs6rZMx-xiE/edit>
- ▶ Some interesting corpora described in NLTK Book Chapter 2.
- ▶ Many proprietary corpora are becoming available for research:
  - ▶ Lexis
  - ▶ Web of Science

# Screen Scraping

- ▶ A screen scraper is a computer program that:
  - ▶ loads/reads in a web page
  - ▶ finds some information on it
  - ▶ grabs the information
  - ▶ stores it in a dataset
- ▶ Once upon a time you could collect virtually any piece of information from the internet by screen scraping.
  - ▶ But now web sites make it difficult with restrictive terms of use, bot-blockers, javascript, etc.
  - ▶ Still, a little creativity goes a long way.

# What a web site looks like to us



WIKIPEDIA  
The Free Encyclopedia

[Main page](#)  
[Contents](#)  
[Featured content](#)  
[Current events](#)  
[Random article](#)  
[Donate to Wikipedia](#)  
[Wikipedia store](#)

Interaction

[Help](#)  
[About Wikipedia](#)  
[Community portal](#)  
[Recent changes](#)  
[Contact page](#)

Tools

[What links here](#)

Create account Log in

Article [Talk](#)

Read [Edit](#) [View history](#)

## World Health Organization ranking of health systems in 2000

From Wikipedia, the free encyclopedia

The **World Health Organization (WHO)** **ranked the health systems** of its 191 member states in its [World Health Report](#)<sup>[1]</sup> 2000. It provided a framework and measurement approach to examine and compare aspects of [health systems](#) around the world.<sup>[2]</sup> It developed a series of performance indicators to assess the overall level and distribution of [health](#) in the populations, and the responsiveness and financing of [health care](#) services. It was the organization's first ever analysis of the world's health systems.<sup>[3]</sup>

**Contents** [\[hide\]](#)

- [1 Ranking](#)
- [2 Methodology](#)
- [3 Criticism](#)
- [4 See also](#)
- [5 References](#)

# What a web site looks like to a computer

```
1 <!DOCTYPE html>
2 <html lang="en" dir="ltr" class="client-nojs">
3 <head>
4 <meta charset="UTF-8" />
5 <title>World Health Organization ranking of health systems in 2000 - Wikipedia, the free encyclopedia</title>
6 <meta name="generator" content="MediaWiki 1.26wmf10" />
7 <link rel="alternate" href="android-
  app://org.wikipedia/http/en.m.wikipedia.org/wiki/World_Health_Organization_ranking_of_health_systems_in_2000"
  />
8 <link rel="alternate" type="application/x-wiki" title="Edit this page" href="/w/index.php?
  title=World_Health_Organization_ranking_of_health_systems_in_2000&action=edit" />
9 <link rel="edit" title="Edit this page" href="/w/index.php?
  title=World_Health_Organization_ranking_of_health_systems_in_2000&action=edit" />
10 <link rel="apple-touch-icon" href="/static/apple-touch/wikipedia.png" />
11 <link rel="shortcut icon" href="/static/favicon/wikipedia.ico" />
12 <link rel="search" type="application/opensearchdescription+xml" href="/w/opensearch_desc.php" title="Wikipedia
  (en)" />
13 <link rel="EditURI" type="application/rsd+xml" href="//en.wikipedia.org/w/api.php?action=rsd" />
14 <link rel="alternate" hreflang="x-default"
  href="/wiki/World_Health_Organization_ranking_of_health_systems_in_2000" />
15 <link rel="copyright" href="//creativecommons.org/licenses/by-sa/3.0/" />
16 <link rel="alternate" type="application/atom+xml" title="Wikipedia Atom feed" href="/w/index.php?
  title=Special:RecentChanges&feed=atom" />
17 <link rel="canonical"
  href="https://en.wikipedia.org/wiki/World_Health_Organization_ranking_of_health_systems_in_2000" />
18 <link rel="stylesheet" href="//en.wikipedia.org/w/load.php?
  debug=false&lang=en&modules=ext.uls.nojs%7Cext.visualEditor.viewPageTarget.noscript%7Cext.wikihihero%7C
  mediawiki.legacy.commonPrint%2Cshared%7Cmediawiki.sectionAnchor%7Cmediawiki.skinning.interface%7Cmediawiki.ui.
  button%7Cskins.vector.styles%7Cwikibase.client.init&only=styles&skin=vector&*" />
19 <meta name="ResourceLoaderDynamicStyles" content="" />
20 <link rel="stylesheet" href="//en.wikipedia.org/w/load.php?
  debug=false&lang=en&modules=sites&only=styles&skin=vector&*" />
https://en.wikipedia.org/w/index.php?title=World_Health_Organization&a:lang(mzn),a:lang(ps),a:lang(ur){text-decoration:none}
```

# Browser Automation

- ▶ Many web sites are designed to be difficult to scrape.
- ▶ Python has solutions for simulating a human browser:
  - ▶ selenium (chromedriver, phantomjs)
- ▶ Other solutions if all else fails:
  - ▶ DownThemAll! plug-in for Firefox
  - ▶ Hire mechanical turkers to manually download data.

# API's

- ▶ API = Application Programming Interface
  - ▶ These are developer-oriented tools that provide access to cleaner data.
- ▶ Chris Bail's list of API's that could be interesting for research:
  - ▶ <https://docs.google.com/spreadsheets/d/1ZEr3okdlb0zctmX0MZKo-gZKPsq5WGn1nJ0xPV7a1-Q/edit>

## Other Languages

- ▶ All of the tools that we discuss in this class are available in many languages.
- ▶ spaCy has full functionality in English, German, Spanish, Portuguese, French, Italian, and Dutch.
  - ▶ beta functionality in dozens of other languages including Chinese and Arabic
  - ▶ See <https://spacy.io/usage/models>.
- ▶ The machine learning models are language-independent.

# Character Encodings

The screenshot shows the macOS Character Viewer interface. The sidebar on the left lists various symbol categories, with 'Unicode' currently selected. The main panel displays the 'Latin Extended-B' font family, showing a grid of characters. The character 'b' is highlighted, and its variations are displayed in a grid. The top right shows the character's Unicode (U+0180) and UTF-8 (C6 80) encoding. The bottom right shows the character's font variation options.



# Corpus cleaning

- ▶ What we've already done:
  - ▶ removed HTML markup, extra white space, and unicode
- ▶ But HTML markup is often valuable:
  - ▶ HTML markup for section header names.
  - ▶ Legal database web sites often have HTML tags for citations to other cases.
- ▶ Other cleaning steps:
  - ▶ page numbers
  - ▶ hyphenations at line breaks
  - ▶ table of contents, indexes, etc.
- ▶ These are all corpus-specific, so inspect ahead of time.

# Regular Expressions

- ▶ Regular Expressions, implemented in the Python package **re**, provide a powerful string matching tool.
  - ▶ A systematic string matching protocol – can match arbitrary string patterns
  - ▶ e.g., use `utilit*` to match `utility`, `utilities`, `utilitarian`, ...
  - ▶ Important for identifying speaker names (in political documents) section headers (in statutes), citations (in judicial opinions), etc.
- ▶ Also quite tedious, so we will not cover it here.
  - ▶ See NLTK book Chapter 3.4-3.5 for an introduction.

# OCR (Optical Character Recognition)

- ▶ Your data might be in PDF's or images. Needs to be converted to text
- ▶ The best solution (that I know of) is ABBYY FineReader, which is expensive but might be available at your university library.
- ▶ My colleague Joe Sutherland at Columbia has a nice open-source package for OCR:
  - ▶ <https://github.com/jlsutherland/doc2text>

# Should you run a spell checker?

- ▶ The short answer is no:
  - ▶ Most corpora have important specialized vocabulary that would be flagged by standard spell-checkers.
  - ▶ They are also very slow to run on large corpora.
  - ▶ In most empirical contexts, it's safe to assume that spelling errors (especially OCR errors) are uncorrelated with treatment assignment.
- ▶ Better solutions:
  - ▶ drop short (one or two letters) and long words (over 12 letters).
  - ▶ get doc frequencies for each word and filter out rare words
    - ▶ or use word embeddings and trust that misspellings will be nearby the true word.
- ▶ But:
  - ▶ There are cases where spelling errors could be correlated with treatment (for example, increasing legislator salaries might change both policy priorities and spelling error rates)

# Measuring Judicial Output using Decision Texts

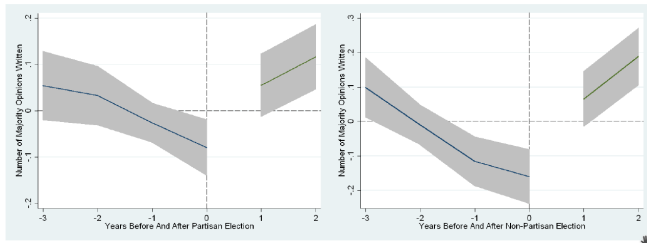
- ▶ The number of documents, and the length of those documents, already provide an interesting set of variables for analysis.
- ▶ For example:
  - ▶ How do electoral incentives affect judge effort?
  - ▶ How does the biological aging process affect effort and writing style?

# Empirical Setting

- ▶ The setting for Ash and MacLeod (2015, 2017, 2018):
  - ▶ State supreme courts: the highest appellate court for each of the 50 states in the USA.
  - ▶ Data set has 1.1 million judicial opinions for 1947-1994
- ▶ States are a nice place to look at natural experiments:
  - ▶ Unlike most jurisdictions, state judges are often elected, and the rules for election change over time.

# Elections Reduce Number of Opinions Written

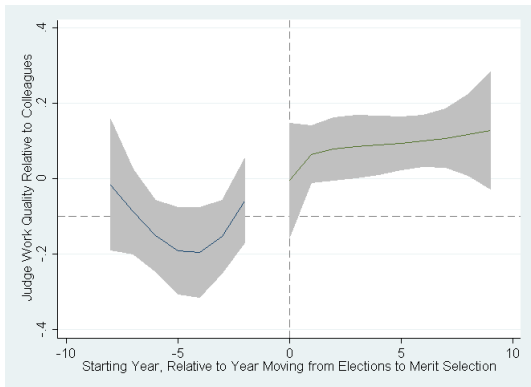
- ▶ Left panel: Partisan Elections, Right panel: Non-Partisan Elections



Fractional-polynomial prediction plots with  $y$  = outcomes and  $x$  = years before and after election year; outcomes residualized on judge and year fixed effects and standardized by judge; gray bars give 95% confidence intervals.

# Effect of Merit-Selection Reform on Work Quality

- Quality of judges, residualized on state-year fixed effects, plotted by starting year, relative to merit reform:

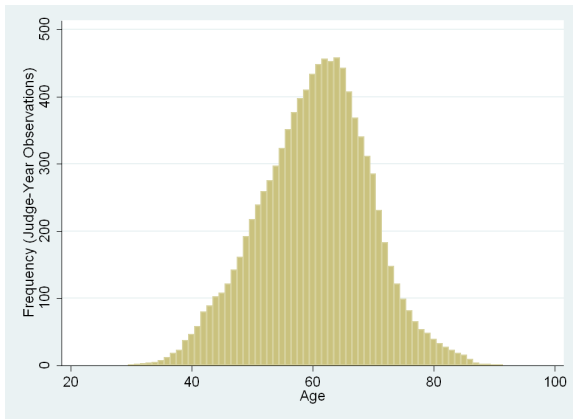


Fractional-polynomial prediction plots with  $y$  = judge quality and  $x$  = judge starting year - reform year; outcomes residualized on state  $\times$  year fixed effects and standardized by state  $\times$  year; gray bars give 95% confidence intervals.

- Judges selected after the reform write higher-quality decisions than judges selected before the reform.

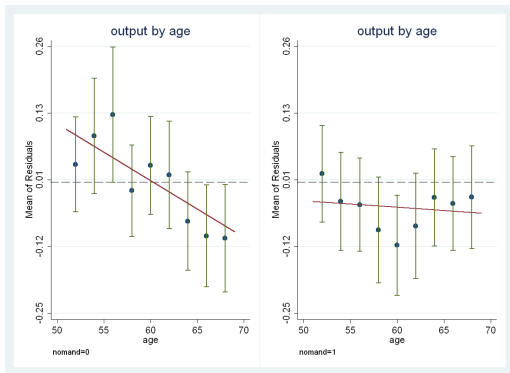


# Judge Age Distribution



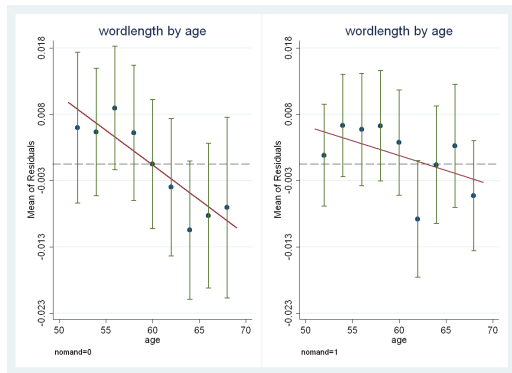
- State supreme court judges have a wide age range but all do the same work task.

# Judge Age and Output



- ▶ Judge output decreases with age, but only under mandatory retirement (left panel).
  - ▶ Consistent with an incentive rather than physiological effect on productivity.

# Characters-per-Word and Judge Age



- ▶ Older judges use shorter words (fewer characters per word).

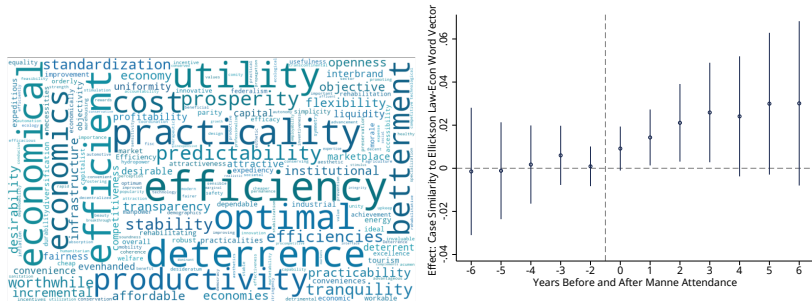
# Overview of Dictionary-Based Methods

- ▶ Dictionary-based text methods use a pre-selected list of words or phrases to analyze a corpus.
- ▶ Three major categories:
  - ▶ Corpus-specific (e.g., number of times a judge says “justice” vs “efficiency”)
  - ▶ General (e.g. LIWC)
  - ▶ Sentiment Analysis

# Corpus-specific words

- ▶ Sometimes counting sets of words or phrases across documents can provide useful evidence.
- ▶ Ash, Chen, and Naidu (2017):
  - ▶ We analyze the use of economics reasoning in the judiciary.
  - ▶ For example, use of the word “efficiency” or “deterrence” after attending a two-week intensive summer course in economics.

# Impact of Economics Training on Economics Language



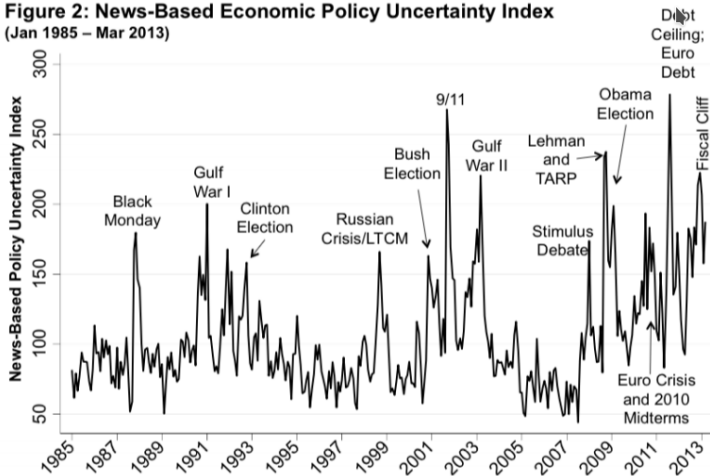
After attendance, Economics Trained Judges increase use of a selection of terms related to law and economics

# Measuring uncertainty in macroeconomy

- ▶ Baker, Bloom, and Davis measure economic policy uncertainty using Boolean search of newspaper articles. (See <http://www.policyuncertainty.com/>).
- ▶ For each paper on each day since 1985, submit the following query:
  - ▶ 1. Article contains “uncertain” OR “uncertainty”, AND
  - ▶ 2. Article contains “economic” OR “economy”, AND
  - ▶ 3. Article contains “congress” OR “deficit” OR “federal reserve” OR “legislation” OR “regulation” OR “white house”
- ▶ Normalize resulting article counts by total newspaper articles that month.

# Measuring uncertainty in macroeconomy

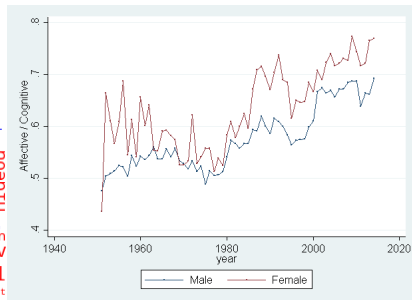
**Figure 2: News-Based Economic Policy Uncertainty Index**  
(Jan 1985 – Mar 2013)





- ▶ LIWC (pronounced “Luke”) stands for Linguistic Inquiry and Word Counts
  - ▶ Info and publications at [liwc.net](http://liwc.net)
  - ▶ Invented in 1980s, now in third version
- ▶ Word List Poster: <http://elliottash.com/wp-content/uploads/2017/07/LIWC2015-dictionary-poster.pdf>

# Emotive vs. Cognitive Processing in U.S. Congress



Source: Gennaro, Ash, and Loewen (2019)

# Sentiment Analysis in Python

- ▶ The vader class in nltk provides positive, negative, and neutral scores for a document, and a composite score that combines all three.
  - ▶ vader works best on raw text – capitalization and punctuation are used in the calculus.
- ▶ Designed for online writing – hard to say how well it works on legal text, for example.
  - ▶ Hamilton-Clark-Leskovec-Jurafsky (2016) provide a method for making domain-specific sentiment lexicons using word embeddings (more on this later).

# Limitations of sentiment analysis

## I'd hate to be the president

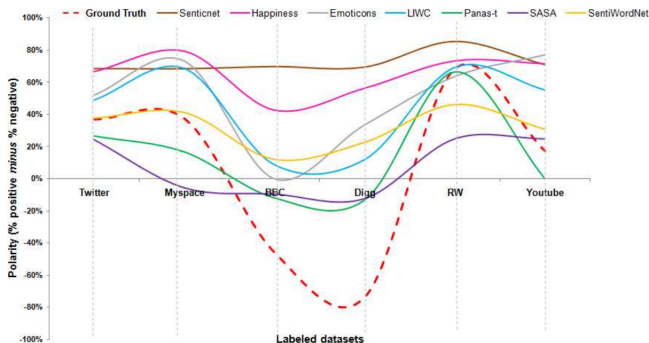


Figure 2: Polarity of the eight sentiment methods across the labeled datasets, indicating that existing methods vary widely in their agreement.