

quanteda

Quantitative Analysis of Textual Data

An Introduction brought to you by:
Laura Menicacci & Dinah Rabe

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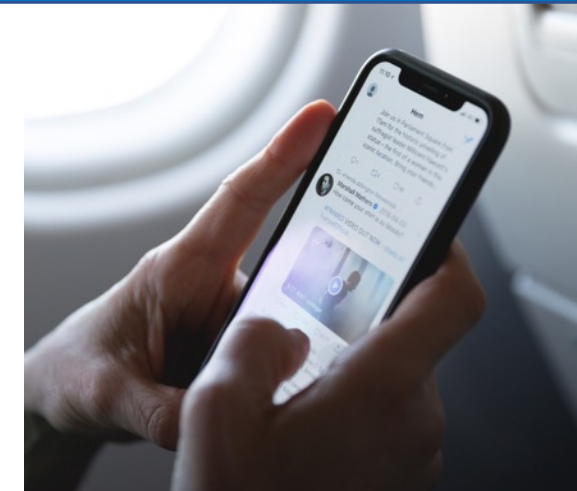
For the course: Intro to Data Science @ Hertie School of Governance

Overview

- Motivation
- Getting some buzzwords right
- Examples of quantitative text analysis
- Basics of quanteda
- The simplified workflow
- Main functions
- Some reasons to love quanteda
- Further resources and our references
- The dataset we will work with

Motivation

- Most of the data of the world exists in text form
- The volume of available textual data has increased dramatically
- A lot of data is generated as we speak, tweet or send messages
- But also for example archives are being digitalized (for all german speakers: the swiss news paper NZZ just finalized their digitalized archives with all their newspapers since 1780!)
- This data is highly unstructured in nature



NZZ Archiv 1780



Definition

Natural language = human language

Getting the buzzwords right

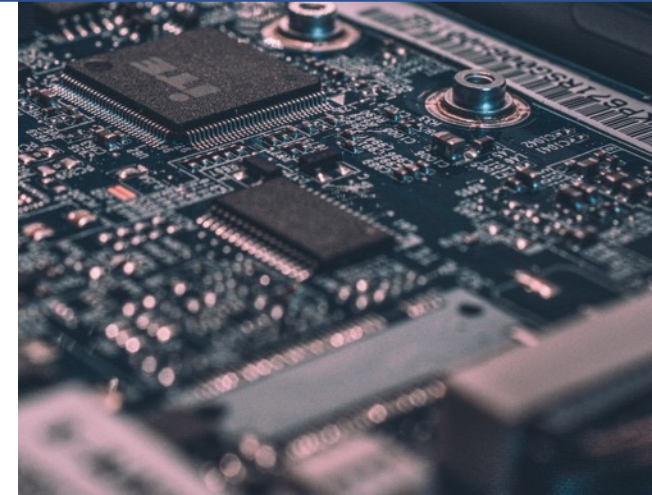
Natural Language Processing, also referred to as „Computational Linguistics“:

- program computers/machines to “read” text (or another input such as speech) by simulating the human ability to understand a *natural* language
- Any kind of computer manipulation of natural language

Examples: Chatbots, Speech Recognition, Google Translate

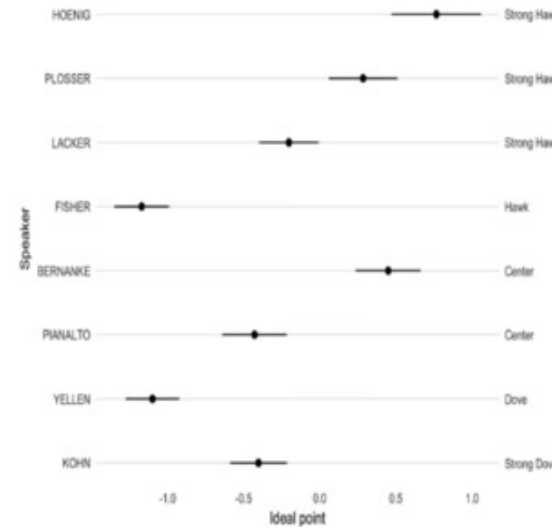
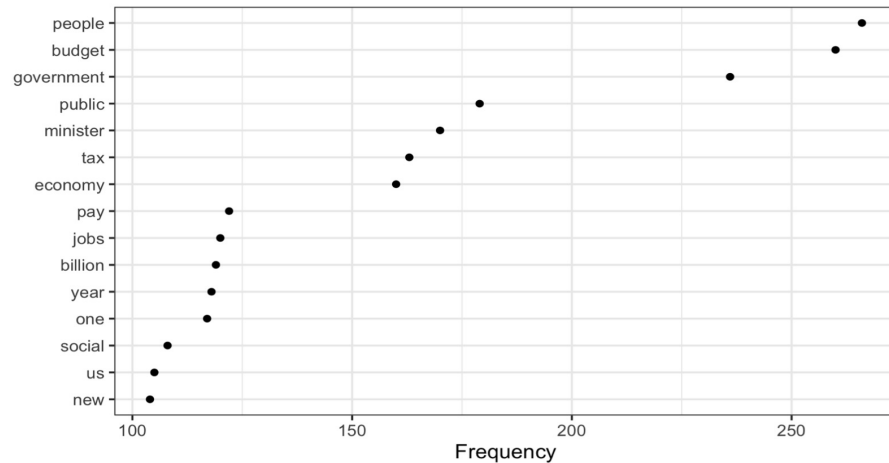
Quantitative Text Analysis/Text Analytics

- is the process of deriving meaningful information from natural language text
- it is expressly quantitative, meaning representing textual content numerically but also analysing it as such using computation and statistical methods



Examples of quantitative text analysis

Descriptive statistics of words

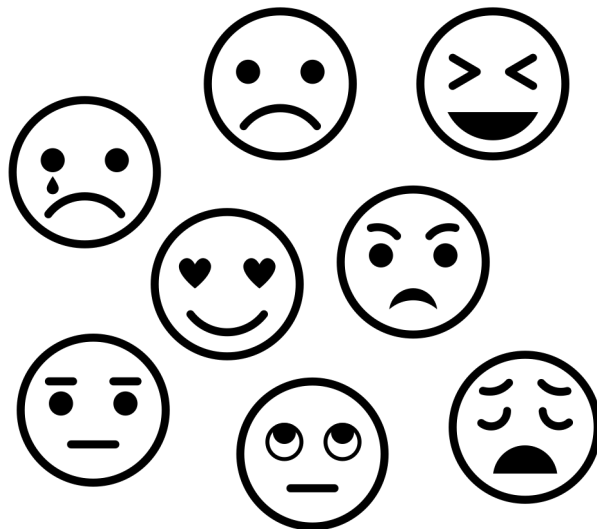


By Hertie Staff:

Central Banker's „positions“ on inflation: Comparing text based positions to „expert“ placements

by Baerg and Lowe (2018)

→ Will Lowe is part of the Hertie Faculty!



Sentiment analysis

AJPS AMERICAN JOURNAL
of POLITICAL SCIENCE

Measuring and Explaining Political Sophistication
through Textual Complexity

Kenneth Benoit London School of Economics and Political Science
Kevin Munger Pennsylvania State University
Arthur Spirling New York University

By the creator of
the quanteda
package:

Kenneth
Benoit

Abstract: Political scientists lack domain-specific measures for the purpose of measuring the sophistication of political communication. We systematically review the shortcomings of existing approaches, before developing a new and better method along with software tools to apply it. We use crowdsourcing to perform thousands of pairwise comparisons of text snippets and incorporate these results into a statistical model of sophistication. This includes previously excluded features

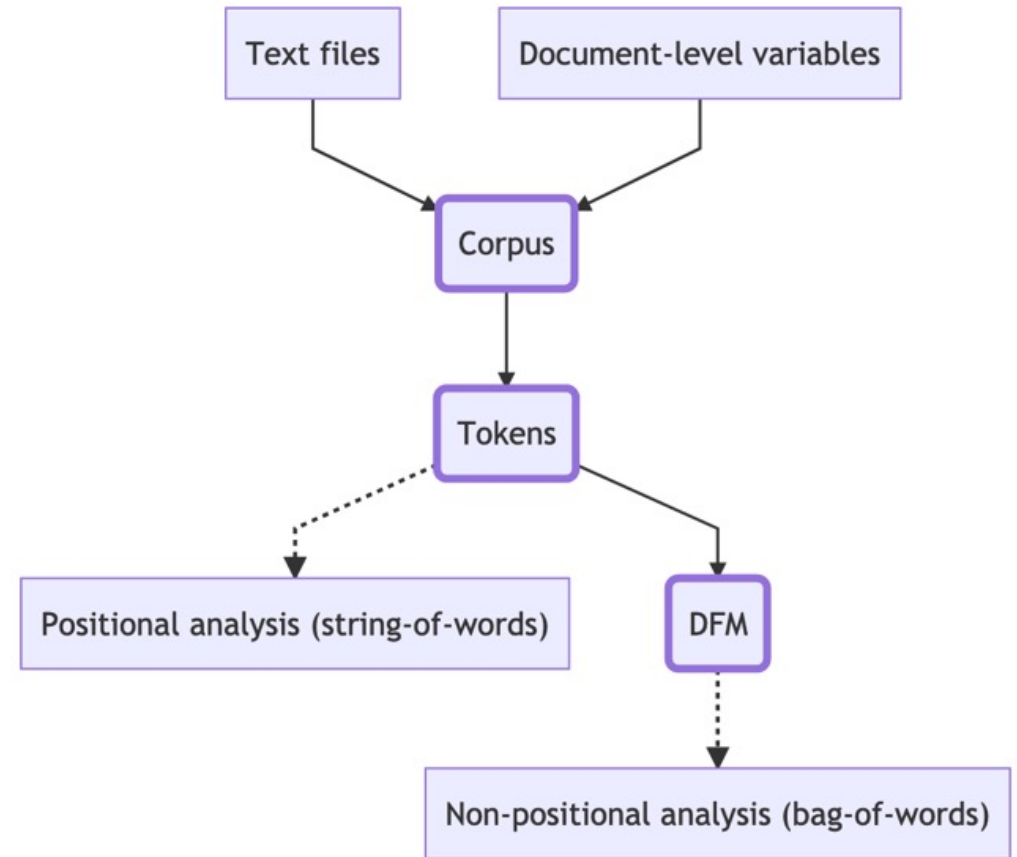
Basics of quanteda

Quanteda works with 3 main objects that you need to know:

Corpus

Tokens

Document Feature Matrix (DFM)



Corpus

- body/set of texts
- Similar to a dataframe
- Often contains document-level variables (*docvars*)
- Docvars are information associated with the documents

```
summary(corp_immig)
```

```
## Corpus consisting of 9 documents, showing 9 documents:
##
##      Text Types Tokens Sentences      party
##      BNP    1125   3280         88      BNP
##      Coalition 142    260          4      Coalition
##      Conservative 251   499         15      Conservative
##      Greens    322   677         21      Greens
##      Labour    298   680         29      Labour
##      LibDem    251   483         14      LibDem
##      PC        77   114          5      PC
##      SNP       88   134          4      SNP
##      UKIP     346   722         26      UKIP
```

Example

Sections of British election manifestos on the topics of immigration and asylum.

```
## Corpus consisting of 9 documents and 1 docvar.
## BNP :
## "IMMIGRATION: AN UNPARALLELED CRISIS WHICH ONLY THE BNP CAN S..."
##
## Coalition :
## "IMMIGRATION. The Government believes that immigration has e..."
##
## Conservative :
## "Attract the brightest and best to our country. Immigration h..."
##
## Greens :
## "Immigration. Migration is a fact of life. People have alway..."
##
## Labour :
## "Crime and immigration The challenge for Britain We will cont..."
##
## LibDem :
## "firm but fair immigration system Britain has always been an ..."
##
## [ reached max_ndoc ... 3 more documents ]
```

Token

- a sequence of characters that are grouped together as a useful semantic unit, often a word (could also be sentences)
- Tokenization is the process of splitting text into tokens
- In our example we will be working with words

Little definition heads up:
A *type* is a unique token

```
## Tokens consisting of 9 documents.
## BNP :
## [1] "IMMIGRATION" "AN" "UNPARALLELED" "CRISIS"
## "WHICH" "ONLY" "THE"
## [8] "BNP" "CAN" "SOLVE" "At"
## "current"
## [ ... and 2,839 more ]
##
## Coalition :
## [1] "IMMIGRATION" "The" "Government" "believes" "that"
## "immigration" "has"
## [8] "enriched" "our" "culture" "and"
## "strengthened"
## [ ... and 219 more ]
##
## Conservative :
## [1] "Attract" "the" "brightest" "and" "best"
## "to" "our"
## [8] "country" "Immigration" "has" "enriched" "our"
## [ ... and 440 more ]
##
## Greens :
## [1] "Immigration" "Migration" "is" "a" "fact"
## "of" "life"
## [8] "People" "have" "always" "moved" "from"
## [ ... and 598 more ]
##
## Labour :
## [1] "Crime" "and" "immigration" "The"
```


Document Feature Matrix

- Is constructed out of a tokens object
- Like a dataframe with documents in rows and “features” (of the token) as columns
- sparsity/sparseness = the proportion of cells that have zero counts

Example

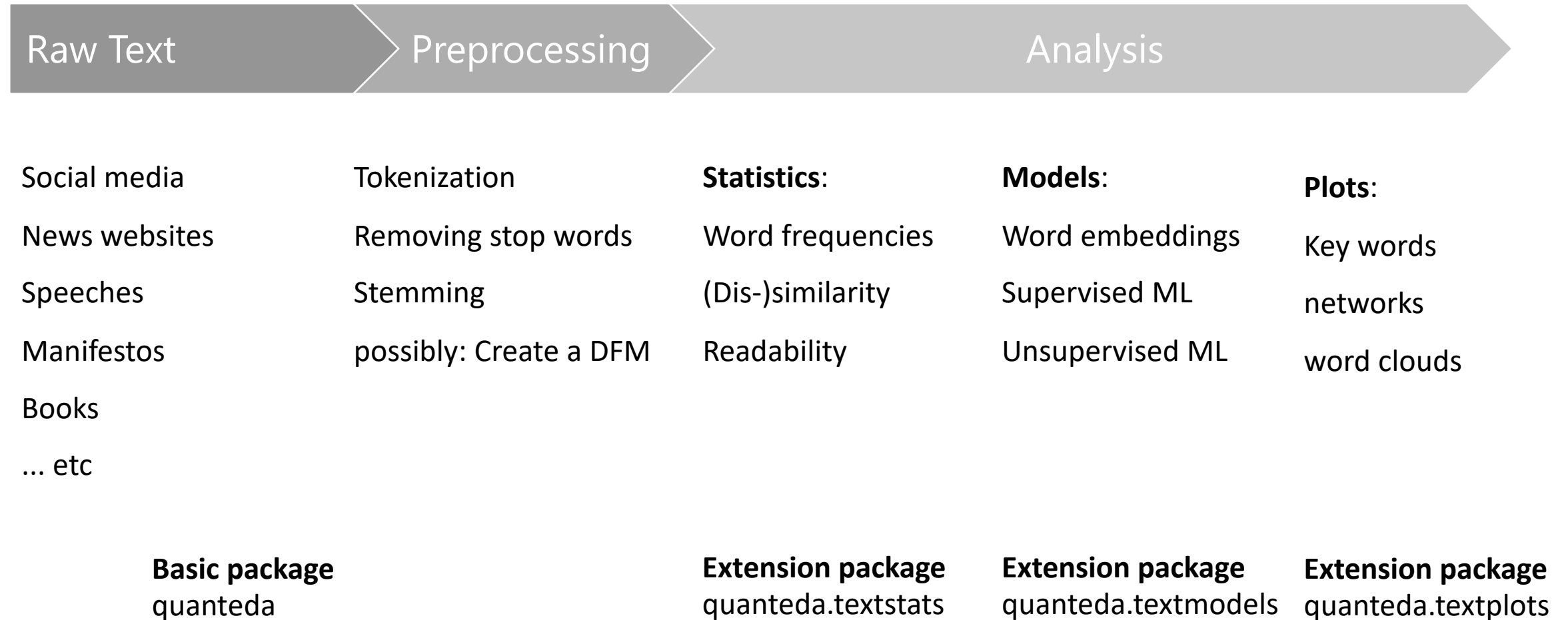
Inaugural speeches of American Presidents

(this dataset is used everywhere in tutorials on text analysis)

```
## Document-feature matrix of: 59 documents, 9,423 features (91.89% sparse) and 4 docvars.
```

```
##               features
## docs               fellow-citizens  of the senate and house
representatives among vicissitudes incident
## 1789-Washington                1  71 116                1  48    2
2      1      1      1
## 1793-Washington                0  11  13                0   2    0
0      0      0      0
## 1797-Adams                    3 140 163                1 130    0
2      4      0      0
## 1801-Jefferson                2 104 130                0  81    0
0      1      0      0
## 1805-Jefferson                0 101 143                0  93    0
0      7      0      0
## 1809-Madison                  1  69 104                0  43    0
0      0      0      0
```

Reduce the magic to a typical workflow



Main function classes

Text corpus: `corpus()`

Tokenization: `tokens()`

Document-feature matrix: `dfm()`

Text statistics: `textstat_()`

Text models: `textmodel_()`

Text plots: `textplot_()`

Corpus functions

- `corpus()`
- `corpus_subset()`
- `corpus_reshape()`
- `corpus_segment()`
- `corpus_sample()`

Pre-existing corpora in the quanteda package:

- `data_corpus_inaugural`
- `data_corpus_irishbudget2010`

There is an entire package with corpora: `quanteda.corpora`

Tokens functions

- `tokens()`
- `tokens_tolower()/tokens_toupper()`
- `tokens_wordstem()`
- `tokens_compound()`
- `tokens_lookup()`
- `tokens_ngrams()`
- `tokens_skipgrams()`
- `tokens_select()/tokens_remove()/tokens_keep()/tokens_replace()`
- `tokens_sample()`
- `tokens_subset()`

Remember that you can use ?
to lookup the functions

Some additional terminology of quanteda

Stems = words with suffixes removed (using a set of rules)

Lemmas = canonical word form

Stop words = words that are designed for exclusion from any analysis of text

Parts of speech = linguistic markers indicating the general category of a word's linguistic property, e.g. noun, verb, adjective, etc.

Named entities = a real-world object, such as persons, locations, organizations, products, etc., that can be denoted with a proper name, often a phrase, e.g. "Hertie School" or "United Kingdom"

Multi-word expressions = sequences of words denoting a single concept, e.g. value added tax (in German: Mehrwertsteuer)

Why quanteda is amazing

- compability with other packages
- You can use a pipelined workflow using magrittr's %>%



THE QUANTEDA INITIATIVE

- UK non-profit organization devoted to the promotion of open-source text analysis software
- software, technical support, teaching and workshops: <https://quanteda.org/>

Further resources

Documentation:

- <https://quanteda.io>
- <https://readtext.quanteda.io>
- <https://spacyr.quanteda.org>
- <https://github.com/quanteda>

Tutorials:

- <https://tutorials.quanteda.io>

Cheatsheet:

- <https://www.rstudio.com/resources/cheatsheets/>
- <https://github.com/rstudio/cheatsheets/blob/master/quanteda.pdf>

Our references

- All the mentioned further resources
- Workshop presentation of Kenneth Benoit at the University of Münster (27–28 June 2019): https://www.uni-muenster.de/imperia/md/content/ifpol/grasp/2019-06-27_muenster.pdf
- <https://manifesto-project.wzb.eu/>

Our example for the tutorial

- The Manifesto Project collects and analyzes parties' electoral programs (manifestos)
- Its data collection is publicly available – data dates back to 1979
- Located at the WZB Berlin Social Science Center and funded by the German Research Foundation



The Manifesto Corpus is the digital text collection of the electoral programs

It contains three types of informations:

- machine-readable texts,
- meta-information for each document (such as language and title)
- annotations/codes on the quasi-sentence level(for some documents)

We will make use of these so called CMP codes; they classify sentences with regards to policy topics (isn't that amazing?!)

A few words about ManifestoR

To access the database through Rstudio, you need 2 things:

- The R package ManifestoR
- An API-key

ManifestoR

- facilitates downloading and processing the Manifesto Corpus
- it allows bulk downloading several documents at once and transforms the downloaded data into a corpus format

API-key

- You need login on the manifesto project website
- There you can create the key on your profile page