Before the Workshop: go.unc.edu/topmod

A GENTLE INTRODUCTION TO

Text Analysis



What is text analysis and why do we use it?

What is Text Analysis

A Tool for Answering Research Questions

- What challenges do pediatric health providers experience in their work?
- How do people talk about vaccines on Twitter?
- Which laws in North Carolina are Jim Crow laws?
- How do posts on social media affect the impact of research articles?
- What information about cultural objects can be found in eBay listings?





What is Text Analysis

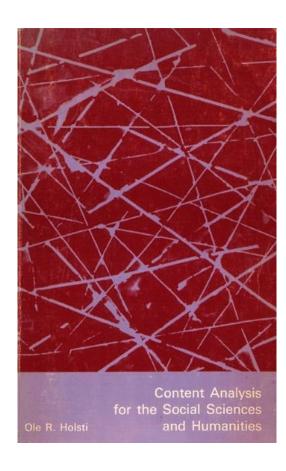
Text analysis is the process by which meaningful information is extracted from unstructured text data.



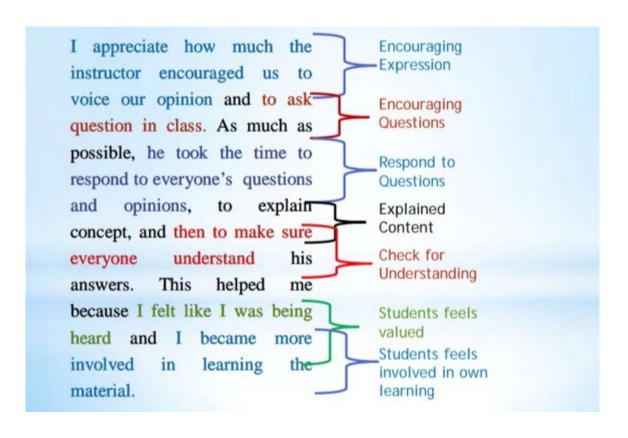
What is Text Analysis

Text Analysis is related to Content Analysis

- Used in Social Sciences and Humanities fields
- Technique for **systematically identifying patterns** in different types of communication
- **Difficult and time consuming** to do **manually** without the help of digital tools
- Manifest content what is said quantitative techniques work better
- Latent content what is meant qualitative techniques work better



Qualitative Text Analysis Technique: Thematic Coding



EXAMPLE

Speech-Language Pathologists in Pediatric Palliative Care: An International Study of Perceptions and Experiences

- Gathered information on the experiences and perceptions of pediatric speech language pathologists through in-depth interviews with professionals
- Interview transcripts were thematically coded using qualitative text analysis software

Krikheli, L., Erickson, S., Carey, L. B., Carey-Sargeant, C., & Mathisen, B. A. (2021). Speech-language pathologists in pediatric palliative care: An international study of perceptions and experiences. *American Journal of Speech - Language Pathology (Online), 30*(1), 150-168. doi:https://doiorg.libproxy.lib.unc.edu/10.1044/2020_AJSLP-20-00090

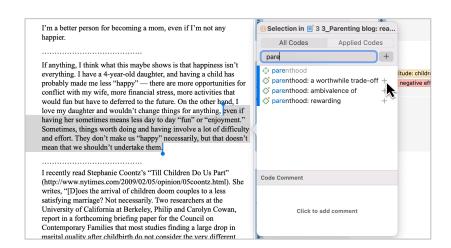
Major themes/codes	Subthemes/subcodes	Freq. of references
Workforce issues		215
	Advocacy for profession	27
	PPC SLP role clarification	56
	Professional challenges	49
	Personal challenges	16
	Workload or workforce	28
	Supervision	28
	Reflective practice	11
Training and resource development		112
	Tertiary education	48
	Professional development – providing	6
	Professional development - receiving	28
	Resources – desired	18
	Resources – existing	15
Governance		62
	Governance at any level	35
	Guidelines	27

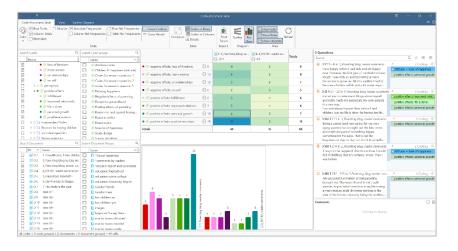
There was a focus on the patient's "voice." In the absence of the ability to efficiently communicate, SLPs discussed ensuring that the needs and desires of the child are known. This can have a greater impact on goal setting, making goals more functional and meaningful to the child and family.

Making sure that the patient's voice is heard. It could be different for children who have cognition that is impacted, but I think unless you're a teenager who is quite outspoken, quite often decisions might be made around what parents hope for or want. [P02]

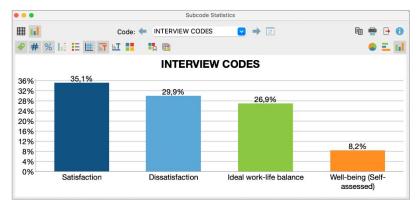
The Coding Process

- Decide what is to be coded entire response? Individual sentences?
- Decide what codes to use deductive or inductive?
- Select text and apply codes
- Analyze and visualize results

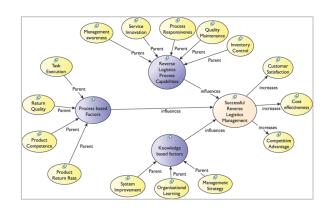


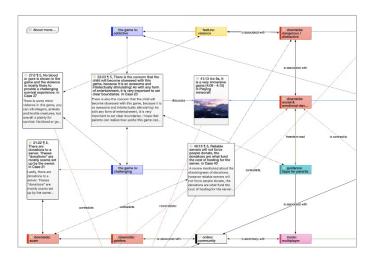


Visualization Examples









Tools

NVivo

Word

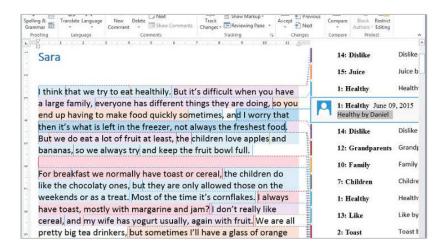
• ATLAS.ti

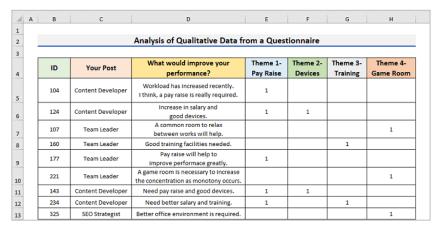
Excel

- MAXQDA
- Dedoose



odum.unc.edu/qualitative-research





Quantitative Text Analysis Techniques

- Bag of Words
- Classification
- Named Entity Recognition
- Sentiment Analysis

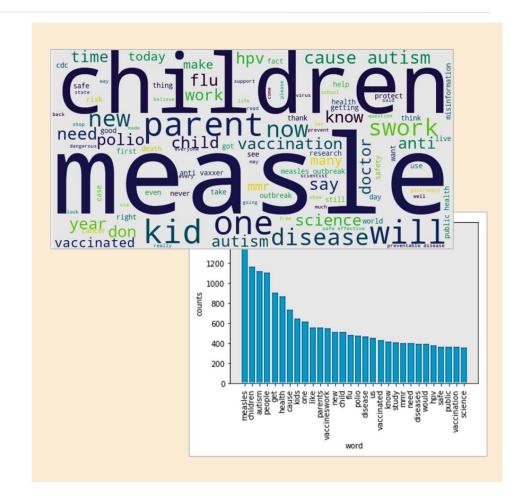


EXAMPLE

Studying Public Perception about Vaccination: A Sentiment Analysis of Tweets

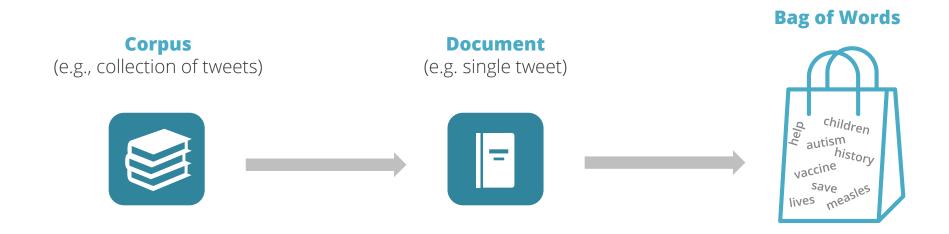
- Examined **vaccine-related tweets** from a period in 2019 when a measles outbreak occurred in the U.S.
- Used bag of words methods that focus on how often words occur in the text, a.k.a. term frequency

Raghupathi, V., Ren, J., & Raghupathi, W. (2020). Studying public perception about vaccination: A sentiment analysis of tweets. *International Journal of Environmental Research and Public Health*, 17(10), 3464. doi:https://doi.org/10.3390/ijerph17103464



What does "Bag of Words" Mean?

Nearly **ALL innate structure is removed** from the text including punctuation, white space and paragraphs, the order in which words occur, etc., so we end up with an **unstructured** "bag" full of jumbled-up words.

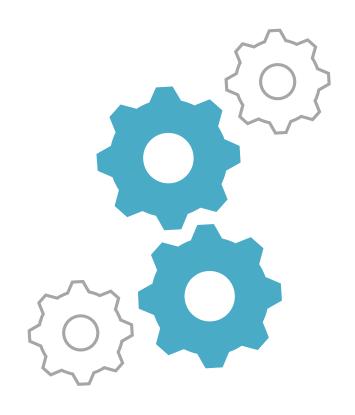


Preprocessing

 Refers to transformations that must be performed on the text before analysis can occur

Common Preprocessing Steps

- Tokenization
- Removal of stop words
- Stemming/Lemmatization



Tokenization

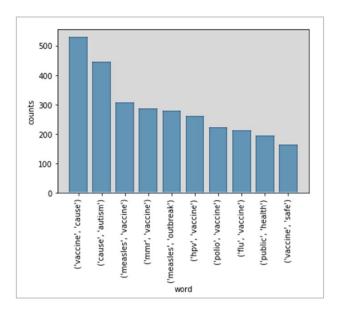
 Breaks a document or corpus into separate clusters of text that become the unit of analysis

Types of Tokenization

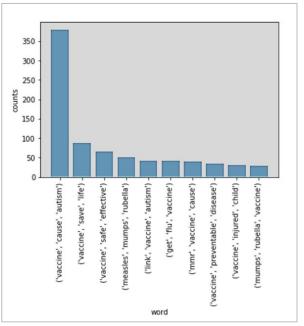
- Word
- N-gram

This is a sentence. this is a sentence

bigrams



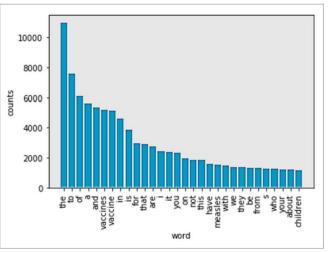
trigrams



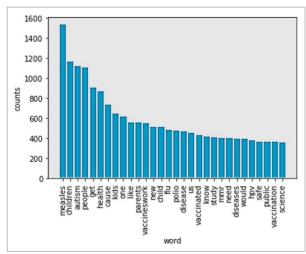
Stop Words

- List of words removed from the corpus
- Include extremely common words that aren't usually very informative
- Articles, pronouns, "to be" verbs, etc.

with stop words



without stop words

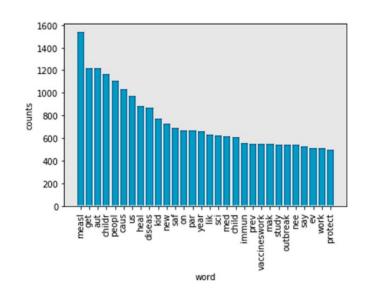


Stemming

- Algorithm removes certain word endings so similar words can be counted together (cars -> car)
- Can result in inaccurate word frequencies (caring -> car)

Lemmatization

- Takes linguistic morphology into account, providing better accuracy
- Algorithm relies on detailed dictionaries



Word	Stemming	Lemmatization
information	inform	information
informative	inform	informative
computers	comput	computer
feet	feet	foot

Term Frequency Method

- Often needs to be normalized by the total number of words in a document
- Number of times a term appears divided by the total number of terms

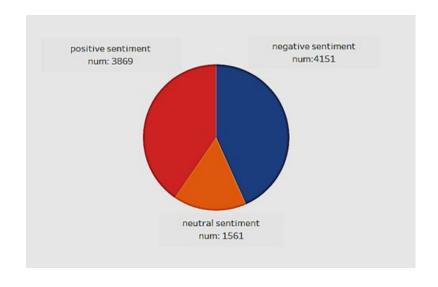
TF-IDF Method

- Uncommon terms are more informative than common ones
- TF-IDF takes into account both the frequency of a term in a document and its overall scarcity in the corpus

$$TF(t,d) = rac{number\ of\ times\ t\ appears\ in\ d}{total\ number\ of\ terms\ in\ d}$$

$$IDF(t) = log rac{N}{1+df}$$

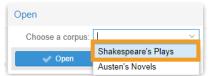
$$TF - IDF(t,d) = TF(t,d) * IDF(t)$$



EXERCISE

Shakespeare's works in Voyant

- 1. Go to voyant-tools.org
- 2. Click on 🗁 Open
- 3. Choose



Tips

- \square
- Open tool in a **new tab** (click **Export**)
- **Change tool**
- Change **options** (more at bottom of tool)
- Tool help

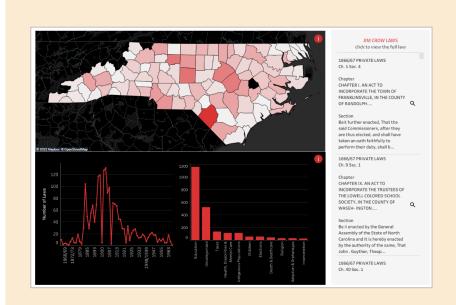
- What is the corpus? What are the documents?
 How many documents are in the corpus? (Hint: Summary tool)
- What is the **most common 4-gram** in the corpus? In what **document** does it have the **highest** relative **frequency**? (Hint: Phrases tool and Trends tool)
- Edit the **stop words** so that the **top 20 terms** are more **informative**. (Hint: Terms tool Options)
- Find a term that Shakespeare uses **more often** in his **early** works than his **later** works. (Hint: Terms tool and Trends tool)
- Look at the highest Significance scores for Document Terms. Why are they all character names? (Hint: Document Terms tool - Help)

EXAMPLE

On the Books: Jim Crow and Algorithms of Resistance

- Discovered Jim Crow and racially-based legislation signed into law in North Carolina between 1866 and 1967
- Used supervised and unsupervised classification techniques to identify laws with race-based language and the type of legislation covered by those laws

Henley, A., Jansen, M., Bruckner, L., Byers, N., & Dalwadi, R. (2020). *On the Books: Jim Crow and Algorithms of Resistance White Paper*. https://doi.org/10.17615/hvz4-sr14



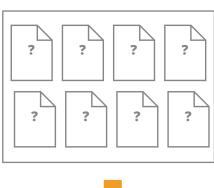
SEC. 9. That it shall be the duty of said School Commissioners to establish graded schools in said town, one for white children and one for colored children, and to appropriate the funds derived from said special taxes and from all other sources, for the maintenance of said schools so as to equalize the school facilities between the races.

Uses for Classification

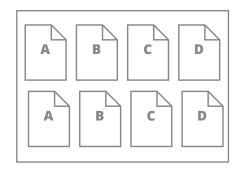
- Separate documents into different groups depending on the language used in the text
- Often called topic modeling when the groups represent topics

Types of Classification

- Supervised
- Unsupervised
- Rule based

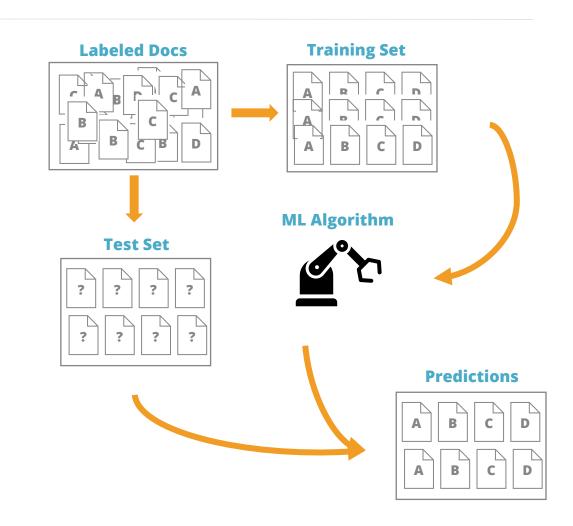






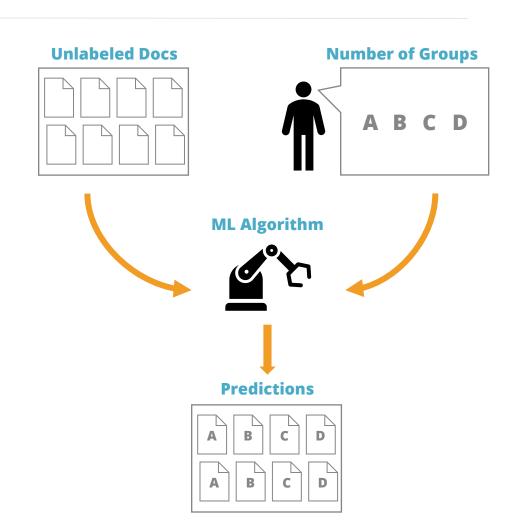
Supervised

- Portion of the corpus must be manually labeled to provide a training set (very time consuming)
- Training set serves as an example of correctly classified text (it's the supervisor)
- Machine learning algorithm uses the labels supplied by the training set to predict classifications for documents in a test set, and eventually unlabeled documents
- Often, many different algorithms are trained, tested and compared to find the best results, e.g., Naïve Bayes, Random Forest, Support Vector Machine, XGBoost



Unsupervised

- Does not require manually labeled text and is less time consuming
- Due to lack of labeling, results are often less accurate and harder to evaluate
- Researcher determines number of groups the algorithm will organize the data into
- Machine learning algorithm looks for similarities in the data and creates groups based on those similarities
- Many different types of algorithms including Latent Dirichlet allocation, K-Means clustering, word embeddings, etc.



EXERCISE

Topic Modeling Tool

- 1. Follow the steps at **go.unc.edu/topmod.**
- 2. Go to your **Desktop** and open the **output_html** folder.
- 3. Double-click on all_topics.html

Tips

To learn more about the topic modeling algorithm we are using, visit:

go.unc.edu/lda

(written by Google employee, Ria Kulshrestha)

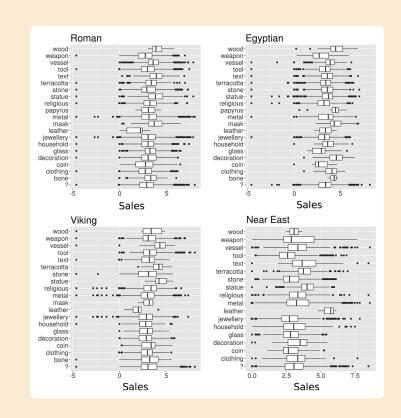
- If you had to come up with a name for each topic, what would it be?
- Click on each topic and look at the top five laws within it. Based on the text of those laws, how representative are the topic names you picked?
- How useful do you think the results of the topic modeling algorithm are?

EXAMPLE

The Market for Heritage: Evidence From eBay Using Natural Language Processing

- Examined the trade of cultural objects and antiquities through eBay listings
- Used Named Entity Recognition techniques to detect the names of cultures, items, and materials within the text of the listings

Altaweel, M. (2019). The Market for Heritage: Evidence From eBay Using Natural Language Processing. *Social Science Computer Review, 39*(3), 391-415. doi:https://doi.org/10.1177/0894439319871015

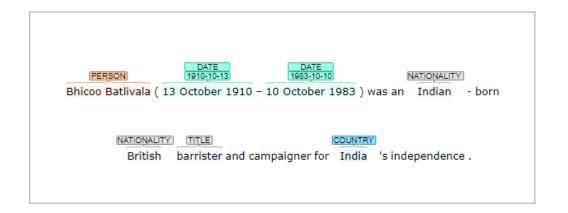


Detect Entities in Text

- NER tools label text with certain tags
- Tags refer to various entities that may represent important information such as people, places, times, etc.

Uses Previously Trained Models

- NER tools typically come pre-trained through supervised machine learning
- Some tools also allow you to train the model with your own labeled data
- Current tools are commonly based on the Stanford Model



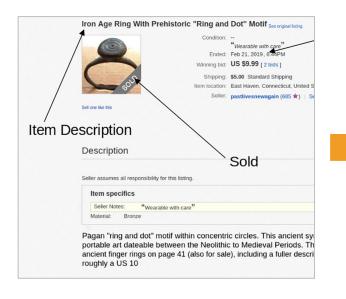
```
In fact, the Chinese NORP market has the three CARDINAL most influential names of the retail and tech space - Alibaba OPE ,

Baidu ORO , and Tencent PERSON (collectively touted as BAT ORO ), and is betting big in the global Al OPE in retail
industry space . The three CARDINAL giants which are claimed to have a cut-throat competition with the U.S. OPE (in terms of resources and capital) are positioning themselves to become the 'future Al PERSON platforms'. The trio is also expanding in other

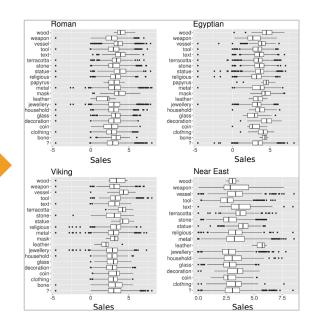
Asian NORP countries and investing heavily in the U.S. OPE based Al OPE startups to leverage the power of Al OPE .

Backed by such powerful initiatives and presence of these conglomerates, the market in APAC Al is forecast to be the fastest-
growing One CARDINAL , with an anticipated CAGR PERSON of 45% PERCENT OVER 2018 - 2024 DATE .
```

NER can turn unstructured text data into a structured dataset, allowing for exploratory analysis.



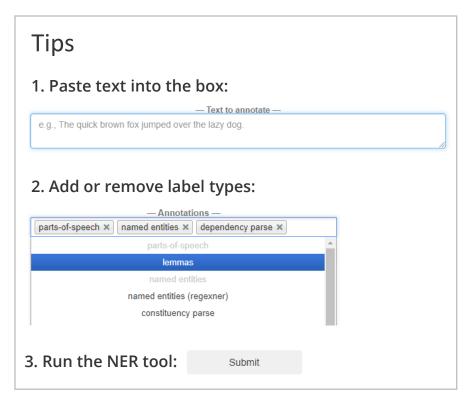
Туре	Total (US\$)	Mean
Total	US\$2,556,092.00	US\$46.7
Roman	US\$873,809.24	US\$46.94
Egyptian	US\$357,256.92	US\$47.93
Unknown culture	US\$299,434.10	US\$48.54
Viking	US\$273,632.65	US\$41.30
Near East	US\$232,599.42	US\$45.94
Greek	US\$178,332.60	US\$49.28
Islamic	US\$152,316.15	US\$156.06
Medieval	US\$139,970.33	US\$30.09



EXERCISE

NER with CoreNLP

1. Go to **corenlp.run**



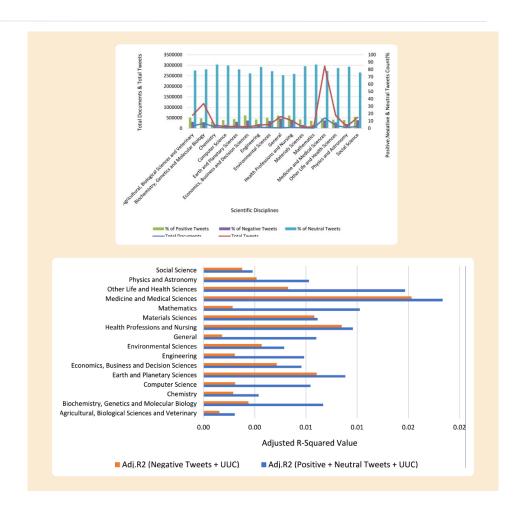
- Go to wikipedia.com and search for "Ynes Mexia"
- Copy the **first paragraph** of the article and paste it into the text box.
- Remove the parts-of-speech and dependency parse annotations so that only named entities is showing.
- Click Submit.
- Carefully read the text and the labels.
- Are there any entities that were missed?
- Are there any entities that were labeled incorrectly?

EXAMPLE

Predicting literature's early impact with sentiment analysis in Twitter

- Examined how sentiment in posts on Twitter related to the early impact of research articles
- Used the SentiStrength lexicon to determine sentiment of tweets, then compared sentiment to citation counts

Hassan, S., Aljohani, N., Idrees, N., Sarwar, R. Nawaz, R., Martínez-Cámara, E., Ventura, S., & Herrera, F., (2020). Predicting literature's early impact with sentiment analysis in Twitter. *Knowledge-Based Systems*,192, 105383.doi:https://doi.org/10.1016/j.knosys.2019.105383.



More of a Goal Than a Technique

- Goal is usually to determine whether a given document ultimately maintains a positive, negative or neutral sentiment
- More recently has branched into detecting specific emotions, intentions, sarcasm, etc. (can be very challenging)

Techniques Used for Sentiment Analysis

- Thematic Coding (Qualitative Text Analysis)
- Lexicons
- Supervised Classification
- Unsupervised Classification



















Lexicons

- **Dictionaries with "scores"** that reflect the positive or negative sentiment of a word
- Documents are given a **total sentiment score** by adding up scores from words in the lexicon
- Often important to normalize sentiment by number of words in the document
- Results can be finely tuned by adjusting scores for negation ("not great"), amplifiers ("very good"), and de-amplifiers ("somewhat nice")
- <u>Benefits</u>: can be made **domain specific**; **programmatically easy** to use – **rule based**
- <u>Drawbacks</u>: difficult to accurately score words with multiple meanings (e.g., "kind"); rule based systems are often closely tied to the domain

AFINN

WORD	SCORE
outstanding	5
distressed	-2
fraud	-4
clever	2

SOCIALSENT

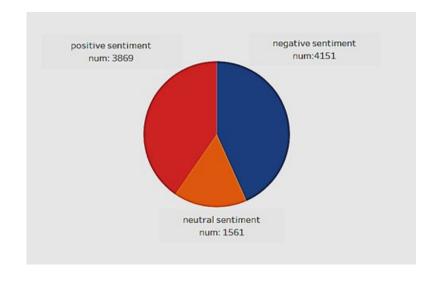
WORD	SCORE
#inspirational	0.984
cancelled	-0.406
#norespect	-0.563
cooperation	0.625

NRC EMOTION

WORD	EMOTION	SCORE
abomination	anger	1
abomination	anticipation	0
abomination	disgust	1
abomination	fear	1
abomination	joy	0
abomination	sadness	0
abomination	surprise	0
abomination	trust	0
abomination	negative	1
abomination	positive	0

Classification

- Documents are grouped by sentiment using classification methods previously discussed
- Can be supervised or unsupervised
- Clustering is one common method for unsupervised classification as is described in the Public Perception about Vaccination study



EXERCISE

SentiStrength Lexicon

1. Go to **sentistrength.wlv.ac.uk**



 Find the scaled sentiment of the following sentence:

This book is great.

- Without changing or removing words, add one word to the above sentence that changes the sentiment to -1.
- Without changing or removing words, add one word to the above sentence that changes the sentiment to 3.
- Change the word "great" in the above sentence so that the resulting sentiment is **0**.

Discussion: Problems & Ethics in Text Analysis

Problems in Text Analysis

Discussion

- What issues can lead to inaccurate results when using text analysis techniques?
- How can bias in data affect text analysis?
- How can bias in researchers affect text analysis?
- Could text analysis techniques cause harm to marginalized communities or people?