TEXT ANALYSIS

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AIMS

- · What is text analysis and why use it?
- · Human vs. machine coding
- · The general process
- · Specific issues with human coding
- · Specific issues with machine coding
- · Pros and Cons

DEFINING TEXT ANALYSIS

"When we perform textual analysis on a text, we make an educated guess at some of the most likely interpretations that might be made of that text." (McKee 2003, 1)

DEFINE CONTENT ANALYSIS

"Content Analysis is a research technique for making replicable and valid inferences from texts (or other meaningful matter [e.g. videos, audio]) to the Contexts of their use." (Krippendorff 2013, 24)

REPLICABLE

"Content Analysis is a research technique for making replicable and valid inferences from texts (or other meaningful matter [e.g. videos, audio]) to the contexts of their use."

Replicable: different researchers, independent of each other should get the same results when applying the same technique.

E.g. independent researchers come to the understanding of the text using the same method.

Replicable results are more reliable.

VALID

"Content Analysis is a research technique for making replicable and Valid inferences from texts (or other meaningful matter [e.g. videos, audio]) to the contexts of their use."

Valid: research is open to careful scrutiny and your claims can be upheld given independently available evidence.

"Content Analysis is a research technique for making replicable and valid inferences from texts (or other meaningful matter [e.g. videos, audio]) to the contexts of their use."

Texts: something that is produced by someone to have meaning for someone else.

E.g. newspaper articles, treaties, transcripts, tweets, maps, advertisements, press releases, movies, party manifestos.

In this course we focus exclusively on texts composed of words.

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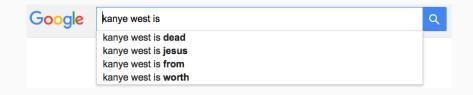
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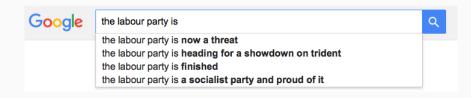
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- 5. Texts have meanings relative to particular contexts.
- Content analysts infer answers to particular research questions from their texts. Their inferences are merely more systematic, explicitly informed, and verifiable ...than what ordinary readers do.





You use and contribute to text analysis every day.





YOU

(Some of you) are building a data set that will be used for text analysis right now.



SOCIAL SCIENCE

People are creating increasingly more (machine accessible) texts.

Massive new source of data for social science analysis.

TEXT ANALYSIS IN SOCIAL SCIENCE (EXAMPLES)

We may have research questions where we conducted a survey with an open-ended question.

We need some systematic way to understand these texts and make comparisons across survey respondents.

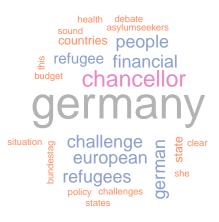
TEXT ANALYSIS IN SOCIAL SCIENCE (EXAMPLES)

We may have research questions where we want to interview a group of people that are hard to access, but who produce many texts.

For example, in an ideal world we may want to survey world leaders for their preferences to handling Syrian refugees. We may want to see how these preferences change over time.

World leaders don't given many interviews (especially not multiple interviews on the same topic), but they-often filtered through a press office-do create many texts.

GERMAN CHANCELLORY PRESS RELEASE SEPT. 2015



11. Januar 2016, 20:22 Uhr Ermittlungen zu den Übergriffen in Köln

Kölner Polizei: Vor allem Marokkaner fallen auf

- Die Ermittler haben bisher 19 Tatverdächtige identifiziert.
- Neun Verdächtige halten sich illegal in Deutschland auf, zehn Personen sind Asylsuchende.
- Von den 19 Verdächtigen kommen 14 aus nordafrikanischen Ländern, vor allem aus Marokko.
- Den Statistiken der Kölner Ermittler zufolge werden 40
 Prozent der nordafrikanischen Zuwanderer innerhalb eines Jahres straffällig.



Source: http://www.sueddeutsche.de/panorama/ermittlungen-zu-den-uebergriffen-in-koeln-vor-allem-marokkaner-fallen-auf-1.2814336

COLOGNE NEW YEARS EVE ASSULTS

January 11, 2016, 20:22 Investigations on the attacks in Cologne

Cologne police: Especially Moroccans to fall

- Investigators have identified so far 19 suspects.
- Nine suspects keep illegal in Germany, ten people are asylum seekers.
- 14 Of the 19 suspects come from North African countries, mainly from Morocco.
- Statistics of Cologne investigators According to 40 percent of North African immigrants become delinquent within a year.

Source: http://www.sueddeutsche.de/panorama/ermittlungen-zu-den-uebergriffen-in-koeln-vor-allem-marokkaner-fallen-auf-1.2814336

via Google Translate

GERMAN CHANCELLORY PRESS RELEASE JANUARY 2015



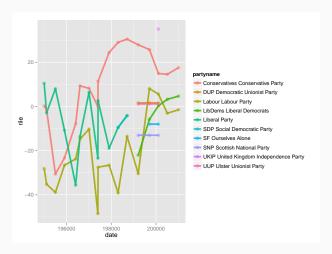
TEXT ANALYSIS IN SOCIAL SCIENCE (EXAMPLES)

We may have research questions about units that are not able to be surveyed, but which produce texts.

E.g. International organisations, political parties, neighbourhood groups.

COMPARATIVE PARTY MANIFESTOS PROJECT

Left-Right Position of UK Parties Based on their Party Manifestos

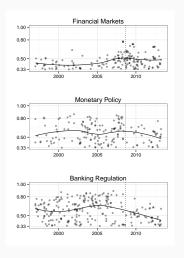


TEXT ANALYSIS IN SOCIAL SCIENCE (EXAMPLES)

We may have research questions about how actors communicate to achieve goals.

For example, what topics do monetary policy bureaucrats talk about more when there is a financial crisis?

TOPICS OF US FEDERAL RESERVE GOVERNOR SPEECHES



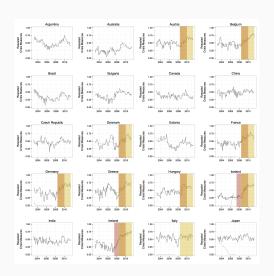
Source: Young and Gandrud (2016)

TEXT ANALYSIS IN SOCIAL SCIENCE (EXAMPLES)

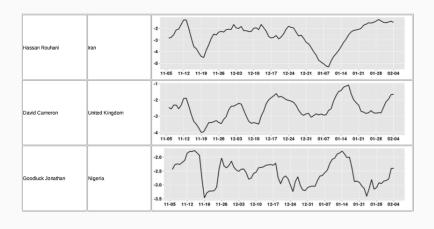
We may have research questions about widely held beliefs across time, where a survey would be too costly to run.

For example, if we wanted to study monthly perceptions of financial market stress across 180 countries.

REAL-TIME PERCEPTIONS OF FINANCIAL MARKET STRESS



GDELT GLOBAL LEADERS PRESS COVERAGE



Source: http://data.gdeltproject.org/worldleadersindex/GDELT_Leaders_Index-2016-02-05.pdf

COMMON OUTPUT

Text analysis can create data that is:

- · Discrete: e.g. the mains topic of your texts.
- · Continuous: e.g. proportion of a document dedicated to a specific word or words, scale (negative to positive, left-right).

The choice largely depends on what is relevant for you research question.

HUMAN AND MACHINE CODING

HUMAN VS. MACHINE CODING

You can analyse texts either by relying exclusively on human coders or primarily rely on machine-assistance.

HUMAN VS. MACHINE CODING

Note: you should never exclusively rely on machine coding. At a minimum, you need to check the validity of your machine assigned codes.

Do the assigned codes make sense in relation to the context?

HUMAN VS. MACHINE CODING

Machine coding has the advantage of being much more efficient for large numbers of texts.

 For example, it would be basically impossible for GDELT to create a daily updated index of world leader press coverage with human coders.

Machine coding is often more easily reproducible and update-able.

HUMAN AND MACHINE CODING SIMILARITIES

Regardless of whether you use human or machine coding, the general text analysis process is the same.

GENERAL TEXT ANALYSIS STEPS

TEXT ANALYSIS STEPS

- 1. Define the population of texts you are interested in (e.g. press releases by a particular organisation, open-ended survey responses).
- 2. Gather your sample of texts
- 3. Develop a coding scheme and classify your texts.
- 4. Establish the reliability and validity of your classifications.

Modified from: http://psc.dss.ucdavis.edu/sommerb/sommerdemo/content/doing.htm

POPULATION

At least two items to consider when defining your population of texts:

- · Should be relevant for your research question.
- · Texts should be accessible.

GATHER YOUR SAMPLE

As with all data gathering, how you sample your texts can greatly affect your results.

For example, if you want to code press attitudes towards immigrants, but only gather articles from The Guardian, you will get much different results than if you only sampled The Daily Mail.

We will discuss sampling in more detail in Week 7.

GATHER YOUR SAMPLE

When you gather your sample (your Corpus), in order to enhance reproducibility, it should be well-organised and electronically available.

DEVELOP A CODING SCHEME

Always consider reliability when developing your coding scheme.

· Will another coder make the same choices given your coding scheme?

So, always fully document your coding scheme and explain your rationale.

DEVELOP A CODING SCHEME (1)

Determine if you want to create a discrete (e.g. main topic of the text) or continuous coding scheme (e.g. attitude scale). This should be based on relevance to your research question.

Skim a sub-sample of the texts to make a list of possible topics, words that would indicate a particular attitude, etc.

DEVELOP A CODING SCHEME (2)

From this initial list, create operational definitions of your topic categories or scale.

· In order to enable replication, make these definitions as clear and specific as possible.

Check that your definitions are comprehensive. Do they cover as many topics, words related to attitudes as possible?

Make sure that your definitions are mutually exclusive, i.e. there is no overlap.

Now, apply your coding scheme.

CHECK RELIABILITY

You should always have at least one other rater independently apply your coding scheme.

Then check the level of aggreement. Ideally, different coders will give the same codes to the same texts based on the same coding scheme.

· This is known as high inter-rater reliability

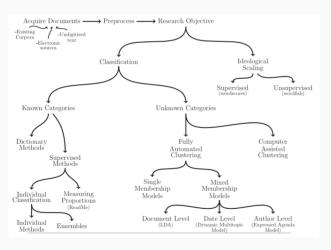
If there is considerable disagreement between raters, you need to re-evaluate your coding scheme and possibly recode your corpus.

CHECK MACHINE CODED RELIABILITY

If you used machine coding, then you should select a random sub-sample of the texts and check to see if the machine codes match your intended coding scheme.

SPECIAL ISSUES WITH MACHINE CODING

There are many different advanced techniques for machine coding:



THIS COURSE

In this course we will focus on:

- · The text preprocessing step.
- · Simple word count methods of text analysis.

Regardless of the type of machine coding you use, you need to preprocess your texts.

This can include...

Removing unnecessary white space (spacing between words), punctuation, capitalisation, numbers, etc.

Removing Stopwords: function words that do not convey meaning like "a" and "the".

Stem your words: reduce the ends of words to reduce the totla number of unique words.

- · For example: family, families, families', familial, are changed to their stem: famili.
- · Related to lemmatization.

Note: each preprocessing decision affects your results and so should be fully justified.

PROS AND CONS

SOME PROS AND CONS OF TEXT ANALYSIS

Cons
Results can be misleading if we don't appreciate the context within which the speech acts takes place.
Purely descriptive, need to do more work to understand why
Sampling bias (including if writers delete texts) can be a major challenge

 $Partially\ from\ http://psc.dss.ucdavis.edu/sommerb/sommerdemo/content/strengths.htm$