Lecture 5: Research on Digital Media

Seminar 'Foundations of Data Science'

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Course Outline

- Part 1: Foundations
 - Day 1: Mon. 14.06.2021: Information Coding & Data
 - Day 2: Tue. 15.06.2021: Programming & Algorithms
 - Day 3: Wed. 16.06.2021: Complexity & Efficiency
- Part 2: Applications
 - Day 4: Thu. 17.06.2021: Data Collection & Quality
 - Day 5: Fri. 18.06.2021: Research on Digital Media

Overview of this Session

- Illustrate real data science applications in research on digital media
 - Examples from my past and ongoing work
 - Background
 - Prior Research
 - WIN Project
 - Spark your curiosity for pursuing this kind of work
 - Many of you enrolled in "Political Data Journalism"
 - Media coverage more salient influence than ever
 - Role of digital media has only grown during COVID



Digital Society Initiative (DSI)

Interdisciplinary Competence Center on Digitalization



Communication

Democracy

DSI Communities

Health

Ethics

Work

Mobility

Digital Democracy Lab digdemlab.io



- Researchers from ETH, UZH, Hertie (Berlin)
- Research infrastructure for data collection and analysis

One of 16 new **DSI professors**

- Assistant Professor in the Department of Political Science (since April 2020)
- Affiliated with the DSI and its new graduate program

Research Area "Political Behavior and Digital Media"

- Substantive research and development of new methods for analysis of online behavior and implications for politics
- Details: www.ipz.uzh.ch/en/forschung/lehrstuehle/bdm.html

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Background



Framing of News Matters...

freedom fighters

Crimean crisis, 2015

terrorists

war <u>in</u> Iraq

Iraq war, 2003

war <u>on</u> Iraq



Two residents wade through chest-deep water after finding bread and soda from a local grocery store after Hurricane Katrina came through the area in New Orleans, Louisiana.

(AFP/Getty Images/Chris Graythen)



Hurricane Katrina, 2005



G7 summit, 2014

Jewel Samad/Getty

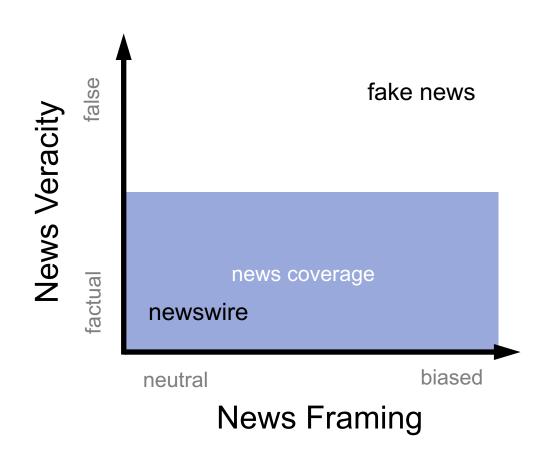


Jewel Samad/Getty

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Framing vs. Fake News



Fake News
Collins English Dictionary

"false, often sensational, information disseminated under the guise of news reporting"

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Framing vs. Opinions

- There is a vibrant current debate about the role of the media.
 - Media does not have to be neutral
 - Opinions on current events are valuable
 - Journalists can and should take a stance
 - Question is how it is done
 - Clearly marked opinion pieces
 - Stated opinions as part of coverage
 - Framing instead subtly transports opinion
 - Reflects own (un-)intentional biases
 - May not be clearly noticeable to readers but still influences them
 - Extreme case: "spinning" a (news) story

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Relevant Academic Research I

Selection/omission of news topics

News framing and its impact

Political relevance of news coverage

Social Sciences

Computer Science

Reader

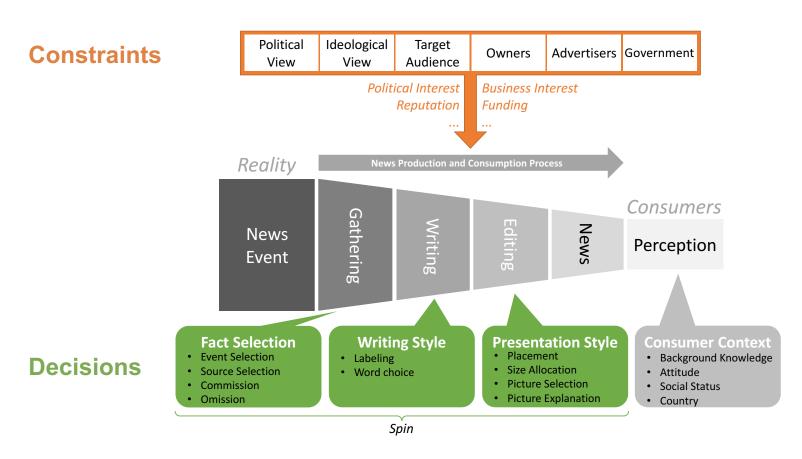
perceptions

Social influence online/offline

Information echo chambers



Conceptual Model of News Production Process



Hamborg, Donnay, Gipp (2019) IJDL

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Relevant Academic Research II

Selection/omission of news topics

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Reader perceptions

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Information echo chambers

Machine Learning/ Deep Learning

Natural language processing

Computer vision

Computer Science

Automated data collection

Topic modeling and event detection

Text similarity and reuse (plagiarism detection)



Prior Research

Biased Information Environments

- News Media
 - Framing affects how we perceive factual information
 - Look at two key dimensions
 - Tone of coverage
 - Choice of images
- Social Media
 - Information sharing online can be partisan and biased
 - Are social media users aware of those potential biases?
- Biased information environments affect political behavior
 - Degree of political engagement
 - Voting behavior, policy implementation etc.



Media Bias

Framing by word choice (= article tone) and labeling

freedom fighters

Crimean crisis, 2015 terrorists

immigrant

European refugee crisis, 2015 economic migrant

undocumented immigrants

US President Trump ended DACA, 2017 illegal aliens

labels

war <u>in</u> Iraq Iraq war, 2003

war on Iraq

U.N. arms inspectors said they had withdrawn two U-2 reconnaissance planes over Iraq for safety reasons

| Iraq war, 2003

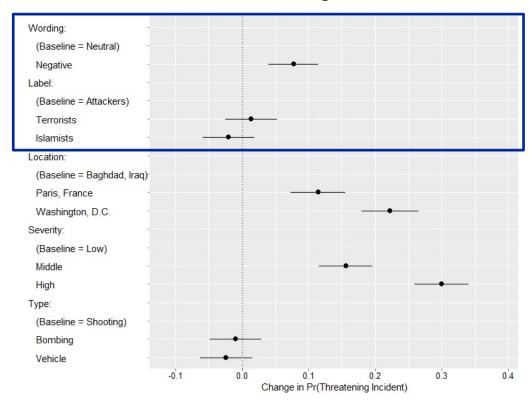
Iraqi <u>fighter</u> jets <u>threatened</u> two American U-2 <u>surveillance</u> planes, <u>forcing</u> them to abort their mission and to return.

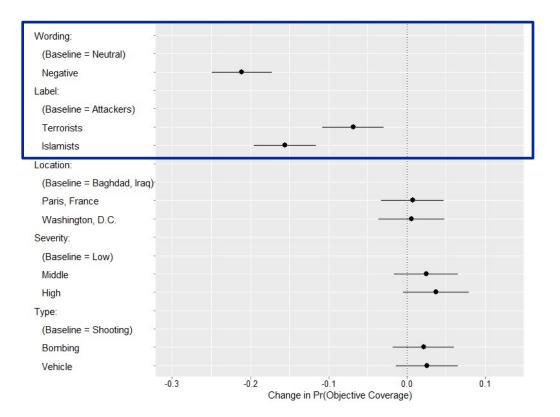
word choice

Hamborg, Zhukova, Gipp (2019) JCDL

Media Bias

Subtle word choice vs. recognizable labels

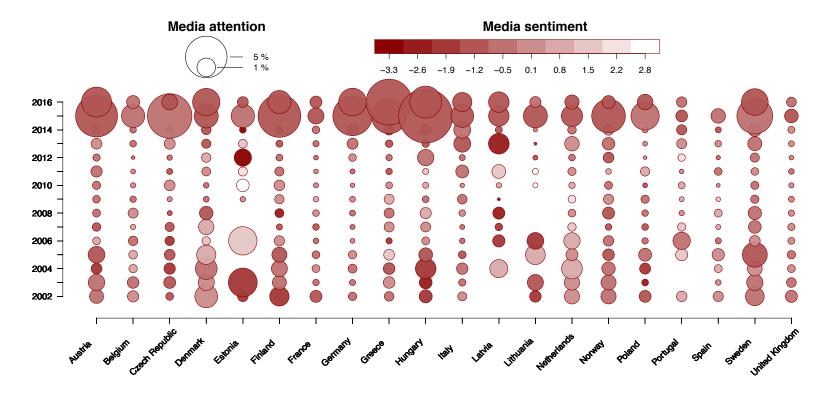




Feick, Donnay, McCabe (2021) APR

Media Bias

Negative article tone can affect policy





18.06.2021

Department of Political Science

Media Bias

- Choice of Images Matters: On-the-ground vs. media coverage of Charlottesville on left and right
 - August 11, 2017 Unite the Right rally in Charlottesville, Virginia
 - James Fields murders Heather Heyer in car attack on protestors that injures 28 more

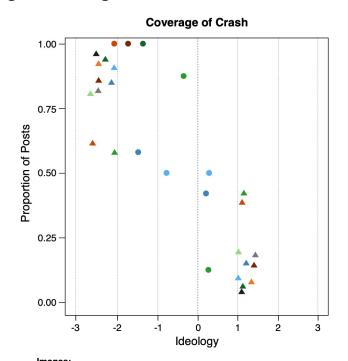


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Media Bias

Usage of images from car attack: media vs. social media, left vs. right



- Coverage of media vs. social media differs
 - Media crash coverage selectively omits
 - Chaos in crowd (image 6)
 - Rescue workers and injured (image 7)
 - Differences greater for news on the right
 - Show even less images
 - Pattern holds for broader coverage, largely showed official shots of president
- Coverage on social media also differs
 - Image sharing heavily dominated from left

Image 1 Image 2 Image 3 Image 4 Image 5 Image 6 Image 7 Image 8 Image 9 Image 10

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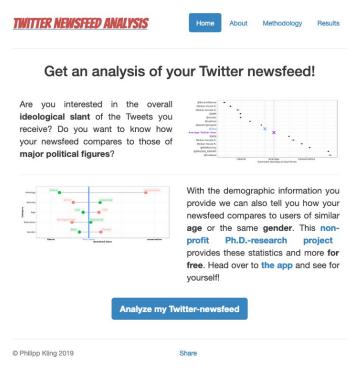


Awareness of Bias

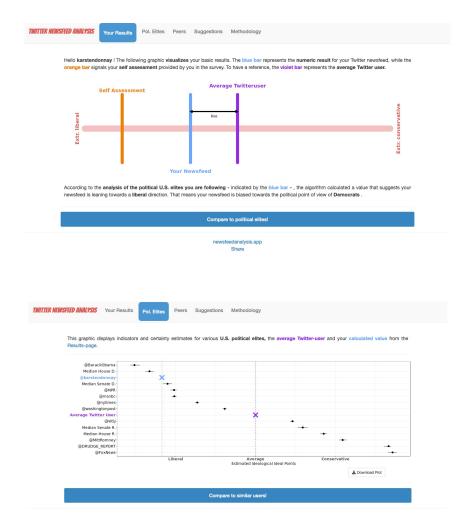
- Information sharing online is partisan
 - Whom you get information from matters because of the implicit selection of what information they share/ you see
 - Echo chambers, i.e., information environment is biased towards your own views
 - Knowing how your information environment looks like is crucial
- Social Media (Twitter)
 - Whom you follows, defines your information environment
 - Idea: measure your own perceptions of political leanings of your friends vs. reality
 - Linked survey and Twitter API measurement: www.newsfeedanalysis.app



Awareness of Bias



www.newsfeedanalysis.app



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WIN Project

WIN Project

- "Fake News and Collective Decision Making Rapid Automated Assessment of Media Bias"
 - Three-year project funded by Heidelberger Akademie der Wissenschaften
 - Interdisciplinary collaboration with team of computer scientists
 - Project team:
 - Gipp Team (Wuppertal) <u>dke.uni-wuppertal.de</u>
 Felix Hamborg, Anastasia Zhukova, Bela Gipp, Moritz Bock
 - Donnay Team (Zürich/Konstanz) <u>ipz.uzh.ch</u>
 Timo Spinde, Angelica Becerra, Marc Lüttecke
 - More information: https://www.karstendonnay.net/research/projects/win/

Contribution of this Project

Selection/omission of news topics

Machine Learning/ Deep Learning

News framing and its impact Political relevance of news coverage

Natural language processing

Computer vision

Social Sciences ←



Computer Science

Reader perceptions

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Automated data collection

Topic modeling and event detection

Text similarity and reuse (plagiarism detection)

Information

echo chambers



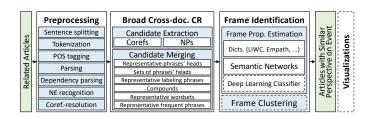
Contribution of this Project – Two Pillars

Testing the impact of readers' perceptions

Automated classification of degree of bias in coverage

Social Sciences ← Computer Science



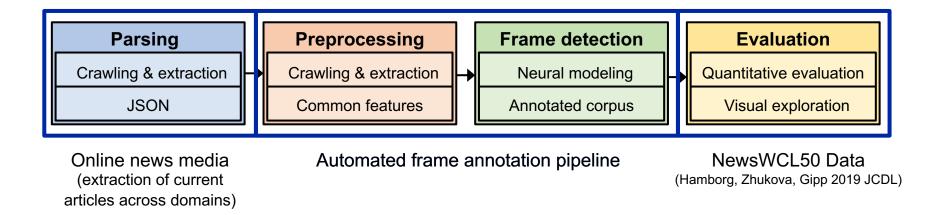


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Technical Approach

Overview of processing pipeline



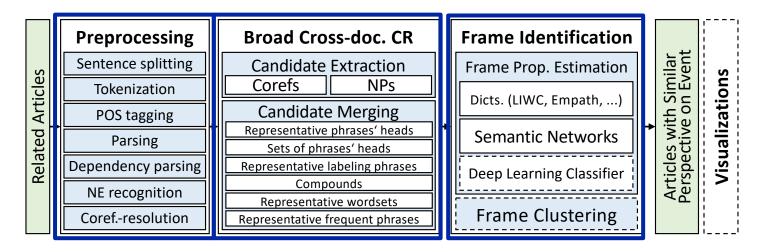
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Technical Approach – Details

Automated frame annotation pipeline



Stanford CoreNLP (neural preferred, else default) Own CDCR approach (candidate extraction + 6 step candidate merging heuristic)

Neural modeling (training data for frames at sentence/paragraph level)

Hamborg, Zhukova, Donnay Gipp 2020 JCDL

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Target-dependent Sentiment

NewsTSC data

In the sentence below, what do you think is the affection of the sentence's author towards the underlined target?

Consider the affection only towards the underlined target, not the event itself or other people.

Target: Mr. Turnbull

Pressed about Mr. Trump's tone, and whether the president ended the call by hanging up, Mr. Turnbull refused to comment.

The attitude of the sentence's author towards the underlined target is...

strong
neutral /
strong
dislike
can't tell
affection

Class frequencies of NewsTSC sets

3 ¹	negative	neutral	positive	total
training test	530 167	1600 487	171 47	2301 701
total	697	2087	218	3002

Hamborg, Donnay, Gipp (2021) iConf

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Target-dependent Sentiment

Fine-tuned BERT model for NewsTSC

LM	Method	AvgRec	acc	$\mathbf{F1_m}$	$\mathbf{F1_{pn}}$
base	AEN-BERT	59.7	62.9	55.0	47.3
	BERT-SPC	62.1	62.1	53.3	44.9
	LCF-BERT	67.3	61.3	54.4	46.5
news	AEN-BERT	59.8	62.9	54.5	46.2
	BERT-SPC	66.7	63.5	55.0	45.8
	LCF-BERT	69.8	66.0	58.8	51.4

State-of-the-art TSC methods:

AEN-BERT (Song et al., 2019)

BERT-SPC (Devlin et al., 2018)

LCF-BERT (Zeng et al., 2019)

(developed on Laptop, Restaurant, Twitter TSC data)

Hamborg, Donnay, Gipp (2021) iConf



Ground-truth Annotation

Structural Challenges

- High class imbalance
- No academic benchmark dataset in news domain
- Neural model requires very large datasets
- Proper sampling across domains and news sources

Coding Challenges

- Attitude of coder towards target might affect coding
- Low attention of coders on MTurk
- ICR not very high for all coding tasks
- Difficult to generalize beyond sentiment ("frame properties")

Ground-truth Annotation

- Building a new comprehensive gold standard dataset for frame detection
 - 10,000s of instances annotated on Amazon MTurk
 - Optimized coding protocol to minimize annotation bias
 - Only consider high quality coding (4/5 or 5/5 agreement)
 - Balanced across classes and news domains
 - Articles from across the US political spectrum
 - See also new POLUSA dataset: https://zenodo.org/record/3813664
 - Systematic sampling of paragraphs/sentences from many articles
 - Validation of MTurk coding against human coders
 - 5 IPZ students performing same task on subset of sentences
 - Provide benchmark comparison of online annotation quality

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Experimental Design I

News aggregator with bias indication similar to allsides.com



Noncitizen voter fraud allegations

According to a 2012 Pew Research Center (Pew) report, which Donald Trump has cited as proof of voter fraud, millions of voter registration records were out of date as people were either deceased or had moved.

From the left

FOLLOWING TRUMP VOTER FRAUD ALLEGATIONS, CLAIM THAT 5.7 MILLION NONCITIZENS VOTED IS WRONG

President Donald Trump's unfounded allegations that millions voted illegally in 2016 is back in the news, with...

Source: Tampa Bay Times

Strength of slanted language bar:

From the center

Report: Trump commission did not find widespread voter fraud

PORTLAND, Maine (AP) — The nowdisbanded voting integrity commission launched by the Trump ...

Source: Associated Press

Strength of slanted language bar:

From the right

Hillary Says Voter Registration Cost Her the Election

President Trump was criticized last year when he tweeted out an accusation that he would have won the popular vote were it ...

Source: National Review

Strength of slanted language bar:



Experimental Design II

Text level annotation of framing of targeted named entity

The boxes in the following article are showing how named entities (such as persons or institutions) are framed. Red color illustrates negative framing, green color illustrates positive framing. If you hover over the grey text, the box that it relates to is highlighted.

FOLLOWING TRUMP VOTER FRAUD ALLEGATIONS, CLAIM THAT 5.7 MILLION NONCITIZENS VOTED IS WRONG

President Donald Trump's unfounded allegations that millions voted illegally in 2016 is back in the news, with his supporters pointing to a new analysis that claims millions of undocumented immigrants voted in 2008. Fox and Friends cohost Ainsley Earhardt talked about it on the morning show recently." 5.7 million -that's how many illegal immigrants might have voted" in 2008, she said. Her comments referenced an article in the Washington Times, a conservative newspaper. Trump has made repeated claims about massive voter fraud and election rigging, which we've debunked again and again and again and again and again and again and again (and we debunked a claim by his spokesman Sean Spicer). The claim made on Fox and Friends is based on an extrapolation of a controversial study that relied on a very small number of responses. Researchers involved in the underlying survey of voters have cautioned against using their data to reach conclusions about noncitizen voters. We emailed a spokeswoman for Fox News and did not get a reply; however, the Washington Times article showed that the information came from Just Facts, a think tank that describes itself as conservative/libertarian and was founded by James D. Agresti, a mechanical engineer in New Jersey.

Online survey experiment

- Conjoint experimental design with variation of
 - Overview
 - Target annotation
 - Type of annotation
- Subjects see overview and then one article
- Questions about perceptions of news reporting after each article
- Clean inference from full randomization of variations

Spinde, Hamborg, Donnay, Becerra, Gipp (2020) JCDL



Course Administration

Reminder – Final Project

– Idea:

- Work on a topic you are interested in or that is related to your research/thesis
- Identify a concrete research question you are trying to address
- Apply the lessons-learned from this class in the context of this question

– Format:

- Full data science pipeline from initial data collection/processing to data wrangling, extracting relevant information, analysis and presentation of results
- Should also be done in Markdown and doubles as your project report; emphasis is on coding but we also expect
 - Motivation for your research question (& short overview of related research)
 - Text narrative that leads through all parts of the data science pipeline
 - Discussion of results and potential shortcomings

Questions & Feedback

- Please reach out to us if you have any questions (or use the forum)
 - Working on your final project next week
 - But also afterwards if you continue to work with data science approaches
- We are looking forward to your feedback and comments
 - The course evaluation is now live and you should have received an invitation
 - There will be time to fill this out during the exercise this afternoon but it also stays open until June 30, 2021
 - The evaluation form might be a bit different this term again given the online-only teaching format..



Up Next

- Exercise (this afternoon)
 - Course Evaluation
 - Accessing APIs
 - Q&A Final Project

18.06.2021