

# A Transnational Analysis of News and Tweets about *Nuclear Phase-Out* in the Aftermath of the Fukushima Incident

Philipp Heinrich<sup>1</sup>, Christoph Adrian<sup>3</sup>, Olena Kalashnikova<sup>2</sup>, Fabian Schäfer<sup>2</sup>, and Stefan Evert<sup>1</sup> Chair of CCL, <sup>2</sup> Chair of Japanese Studies, <sup>3</sup> Chair of Communication Science Friedrich-Alexander University of Erlangen-Nuremberg
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# Introduction







#### **Background**

- Exploring the Fukushima Effect
  - identification and analysis of the tempo-spatial propagation of discourses in the transnational algorithmic public sphere
  - case study: Fukushima Effect
  - · data: mass and social media (German, Japanese)
- Emerging Fields Initiative of FAU
- Team:
  - Chair of Corpus and Computational Linguistics
     Prof. Dr. Stefan Evert, Philipp Heinrich
  - Chair of Japanese Studies
     Prof. Dr. Fabian Schäfer, Olena Kalashnikova
  - Chair of Communication Science
     Prof. Dr. Christina Holtz-Bacha, Christoph Adrian
  - Chair of Computer Graphics
     Prof. Dr.-Ing. Marc Stamminger, Jonas Müller

#### Introduction

#### Methodology

Keywords, Collocations, and Discourse Corpora

#### Case Study: Fukushima Effect

Mass Media Social Media

Instead of a Conclusion: Work in Progress



# Methodology







#### From Text to Discourse

- Foucault (1969): statements in conversation
- interpretation of text means categorizing
  - utterances
  - sentences
  - paragraphs
  - tweets
- categories
  - not known a priori
  - must be made up *on the fly* by the hermeneutic interpreter





#### From Text to Discourse

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#### Case Study "Exploring the Fukushima Effect"

- √ discourse(s): attitudes and opinions towards energy sources
- √ intra- and transmedial and -national
- √ "edited mass communication" vs. "mass self-communication"





#### Corpus-Based Discourse Analysis (CDA)

- CDA means analyzing and deconstructing concordance lines (Baker, 2006)
  - concordances are the essence of discourses
- finding discourses: nodes + attitudes
  - (topic) nodes: defined by keywords or (more generally) corpus queries
  - attitudes: collocates that are retrieved by statistical methods
- examples
  - "refugees as victims" (Baker, 2006)
  - "Fukushima as worst case scenario"

#### in practice:

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- think of categories
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#### **Collocates and Keywords**

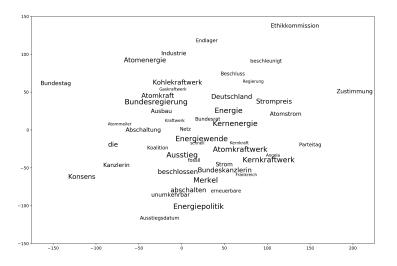
#### keywords

- given two frequency lists of lexical items
- · perform statistical tests
  - likelihood
  - log ratio
  - frequency filter

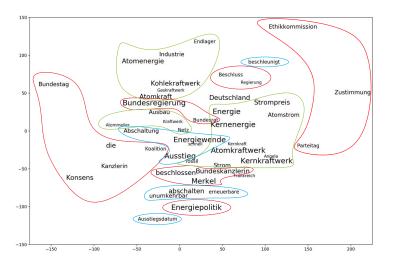
#### collocates

- given a definition of a subcorpus
- rate lexical items according to association strength
  - window size
  - segment-based
  - association measure

#### **Visualizing Semantics**



#### **Visualizing Semantics**







#### Corpora – mass media

#### Frankfurter Allgemeine Zeitung (2011–2014)

- statistics:
  - 306,580 articles, 1,656,372 paragraphs
  - 145,055,523 tokens (1,981,726 types)
- linguistic annotation:
  - TreeTagger (tokenization, POS-tagging, lemmatization)

#### Yomiuri Shimbun (2011–2015)

- statistics:
  - 1,688,435 articles, 12,757,433 paragraphs
  - 580,518,367 tokens (392,971 types)
- linguistic annotation:
  - MeCab (SUWs)





#### Corpora – social media (Twitter)

#### **German Twitter**

- 10,266,835 original posts
- linguistic annotation:
  - tokenization: SoMaJo (Proisl and Uhrig, 2016)
  - POS-tagging: SoMeWeTa (Proisl, 2018)
  - · lemmatization: work in progress

#### Japanese Twitter

- 275,689,087 original posts
- linguistic annotation:
  - special dictionary: ipadic-neologd (Sato et al., 2017)
    - + identification and removal of noise (Schäfer et al., 2017)





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# Case Study: Fukushima Effect







#### Mass media in the aftermath of the Fukushima incident

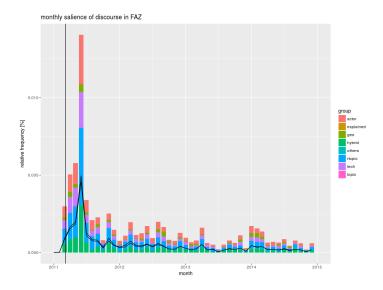
#### German (FAZ)

- salience of energy transition discourse relatively stable (2011–2014)
- nuclear phase-out (Atomausstieg) as part of this discourse: sparked shortly after 3/11
  - political actors and issues (Ethikkommission, electricity supply)
  - · economic actors (RWE)
  - technological issues (Stromnetz)

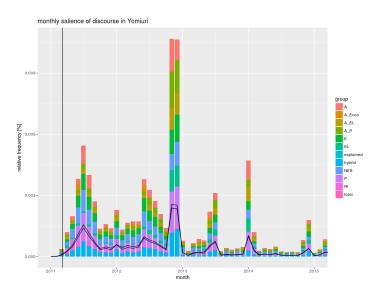
#### Japanese (Yomiuri)

- nuclear phase-out (脱原発) in 2011:
  - political actors (菅, 野田, 首相)
  - economic issues (発電, 稼働, 復興)
  - technological aspects (安全, 燃料)
- nuclear phase-out in 2014:
  - elections and politics (演説, as used in 街頭演説)
  - fewer words regarding economics (note  $\mathcal{T} \checkmark \mathcal{I} \stackrel{.}{\leq} \mathcal{I} \mathcal{I}$ )

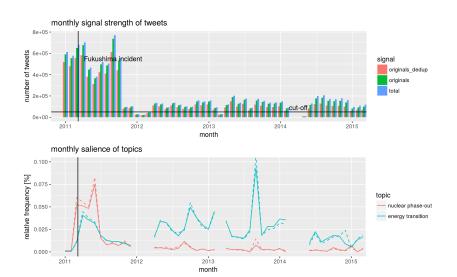
#### **FAZ**



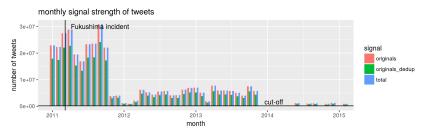
#### **Yomiuri**

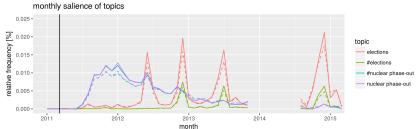


#### **German Twitter**



#### **Japanese Twitter**







## Instead of a Conclusion: Work in Progress



#### **Second-Order Collocates**

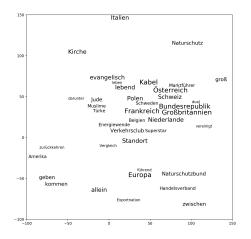


Figure: Paragraph-collocates of Germany in the FAZ corpus.

#### **Second-Order Collocates**

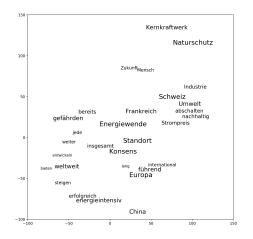
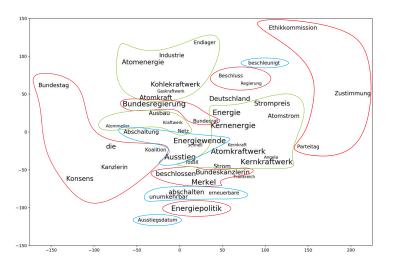


Figure: Collocates of Germany in energy-transition paragraphs.

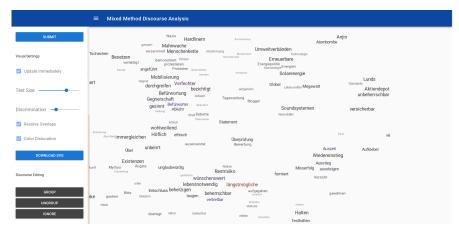
#### **Discourse Embeddings**







#### **Towards Mixed-Methods Discourse Analysis**







Thanks for listening. **Questions?** 



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