Connecting the Dots: NLP support for Analyzing Arguments, Evidence and Translations in the Humanities



Acknowledgements to Dr. Johannes Daxenberger and Chris Stahlhut, M.Sc. for the slides

Hamburg, 21. October 2019



Humanities Research Needs Search



- 1. Knowledge-Based Search: Retrieving Translations
- 2. Information Retrieval and Classification: Arguments
- 3. Interactive Machine Learning: Evidence Detection



How to Computationally Approach Extinct Languages: A Case Study on Hittite



- BMBF-funded "Centre for the Digital Foundation of Research in the Humanities, Social, and Educational Sciences"
- Coordination: TU Darmstadt (Iryna Gurevych), partners: Goethe-Universität Frankfurt a.M. and Leibniz Institute DIPF Frankfurt a.M.
- CEDIFOR intends to contribute to bridging the gap between research in the humanities and computer based methods, and help researchers to master the characteristic problems in this process



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Parts of this work are accepted to be published at the DHd conference 2017:

Daxenberger, J., Görke, S., Siahdohoni, D., Gurevych, I., & Prechel, D. (2017). Semantische Suche in Ausgestorbenen Sprachen: Eine Fallstudie für das Hethitische. In Proceedings of the DHd 2017 (p. to appear).



CEDIFOR Pilot Project: Building a Framework for Semantic Search on Cuneiform Scripts



- Goal: Make the translations of Hittite cuneiform documents more accessible by adding an additional layer of information
- Methods: Lexical-semantic methods based on Natural Language
 Processing technology for a deeper exploration of the documents
- Cooperation with Prof. Doris Prechel and Susanne Görke,
 Altorientalische Philologie, Johannes Gutenberg-Universität Mainz





The Cuneiform Script



- Cuneiform Script (4 cent. BC) is documented on clay tablets only
- Well-Documented Cuneiform Languages include
 - Akkadian (Semitic)
 - Sumerian (Isolated)
 - Hittite (Indo-European)



Digitizing Cuneiform Clay Tablets...



```
7 -- [... GÙ]B-laza [...]
 7 A Vs. II 6' [ ... GÙ]B-la-za (Rasur) Vs. II 7' [ ... ]
 8 -- [... <sup>H</sup>]<sup>I.A</sup> adanna [...]
    A vs. II 7' [ ... <sup>H</sup>]I.A a-da-an-na vs. II 8' [ ... ]
      -- [nu=tta <sup>DUG</sup>KA.G]AG gulšantan [ ... ]
      A Vs. II 8' [ ... DUGKA.G]AG gul-ša-an-ta-an Vs. II 9' [ ... ]
 9
10 -- [ ... walh]it? [ ... ]
10 A vs. II 9' [ ... wa-al-h]i?-it vs. II 10' [ ... ]
```



Digitizing Cuneiform Clay Tablets...



```
-- [... li]nks [...]
    -- [...] zu Essen [...]
    -- [ ... dir] den eingeritzten [Bech]er [ ... ]
        [ ... ] mit [walḫ]i-Getränk<sup>?</sup> [ ... ]
10
11
        Ich ließ [ ... ]
        [Iss Erstklassiges]
12
13
     -- [und] stille deinen Hunger!
14
        [Trink Erstklassiges]
        [und] stille deinen Durst!
15
16
         [ \dots ]
17
    -- [Der Bedrü]kte aber [ ... ]
         [ ... ] komm [wieder] in Ordnung!
18
```



Motivation: Accessing the Content of Digitized Cuneiform Documents



- Digitalization and Accessibility are highly desirable goals
- Currently no means to automatically process the transliterations
- But we can work with the translations (which are a bit special...)

Semantic Search

- Assume we have only a rough idea of what we want to search/find, e.g. a king, a god or a city
- There are plenty of named entities in the document falling into these categories

Primary requirements for the search engine

- Return both exact matches as well as results motivated by semantic relatedness
- Intuitive usage by non-experts



Current Data



- Original documents: Hittite and Akkadian
 - about 700 documents and growing...
 - approx. 400.000 translated words
- Translations in German, Italian, French, and English
- Content: rituals, myths, state treaties, prayers
- Parallel Text: Transliterations and Translations
 - Mapping on the level of meaning units
- Title and introductory comments for each document



Particularities of the Data



- Comments: in-line annotations and footnotes
- Mixed dialects/languages
- Normalized translations vs. text evidences
- Language experts/translators have diverging conventions (e.g. meaning of special characters)
- Damaged and thus fragmentary text (to the extent of illegibleness of entire documents)



Indexing: Technologies



- DKPro Core/UIMA as convenient NLP processing framework
 - Reader recognizes document structure and parallel translation/transliterations (e.g. footnotes)
 - Tokenizer: OpenNLP and Language Tool
 - Fixing token boundaries due to fragmentary text
 - POS-tagging: OpenNLP
 - Lemmatizing: MateTool, LanguageTool
 - Word Sense Disambiguation: DKPro WSD/Lesk
 - Semantic Preprocessing: Uby
 - Writer: MySQL writer (12 tables)



Indexing: Semantic Preprocessing



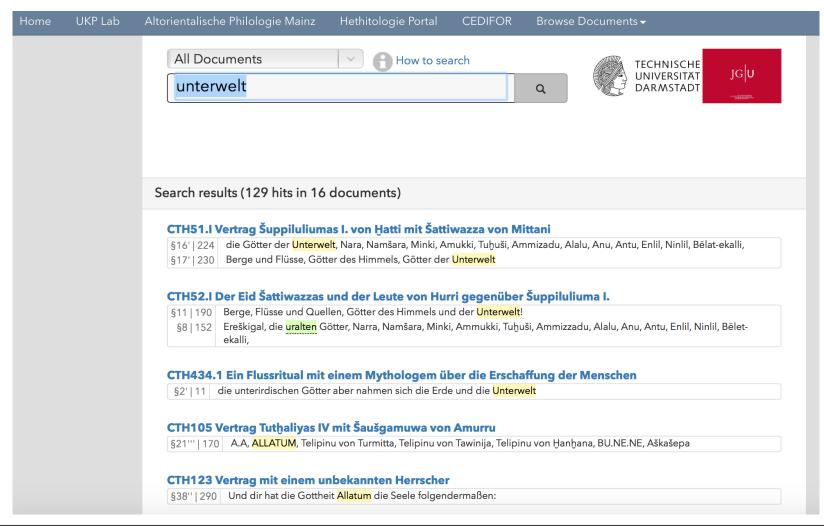
- Uby (combines sense information from WordNet, FrameNet, VerbNet, Wiktionary, Wikipedia and others)
 - Semantic Labels
 - Semantic Field (e.g. person)
 - Category (e.g. fruit)
 - Domain (e.g. sports)
 - Synonyms
- Custom Dictionaries (extensible)
 - Alternative Spellings
 - e.g. Hattuša, Hattusa, Hattuscha, Hattuşaş, Hattusha, Hatusha, ...
 - Hypernyms
 - e.g. Herrscher, Großkönig, König : Pitḫana, Anitta, Labarna, Ḫattušili, ...

https://dkpro.github.io/dkpro-uby/



Query Interface







Computationally Approaching Extinct Languages: Lessons Learned



- Typical issues when computationally approaching extinct languages:
 - Small data (not just small training data)
 - Non-standard vocabulary (translations)
 - Fragmentary text
- Some solutions
 - Implement language-independent, broad approaches
 - any available meta-information (commentaries, footnotes, etc.) should be used
 - Use NLP machinery off-the-shelf a far as possible, add custom solutions where required



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Argument Retrieval and Classification



Topic

Autonomous trucks

PRO: The use of such technology will reduce fatal truck and bus accidents by 20 per cent, resulting in 800 less deaths on European roads.

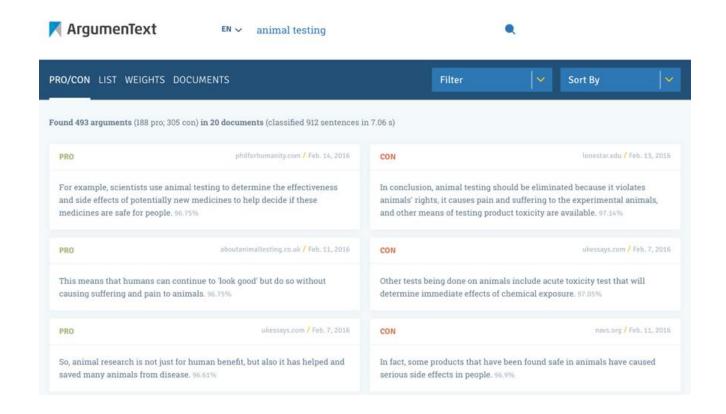
CON: These components are still expensive today; for example, the 600-rpm spinning lightimaging radar system that crowns most current autonomous vehicles costs upward of \$70,000.

http://www.strategy-business.com



Search Engine: Retrieval and Classification







Challenges in the Web Scenario



Heterogeneity

Robustness

Multiple text types

Different languages

Small training resources

Scalability

Expert annotations

Specific domains

Various applications



Training Data Generation



Data

- Heterogeneous text types
- Collected from web searches
- 100+ topics

Task

- Given a (controversial) topic
- Label text as pro or con argument

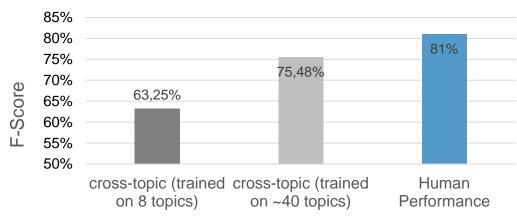
Crowdsourcing

Amazon Mechanical Turk



Results (BERT): Stance + Argument Detection





Reimers, N., Schiller, B., Beck, T., Daxenberger, J., et al. (2019)

- 3-class problem: pro, con, non-argument
- Approx. 5pp improvement over BiLSTM

Visualizing Argument Clusters



```
splito/large/cluster167
size:12
scientifically
detected Associate brain Week
becomesense feeling Mayren
bodily callege Adjunct reflex
system brain week
becomesense feeling Mayren
bodily callege Adjunct reflex
system brain week
becomesense feeling Mayren
concludes reflex
system brain week
becomesense feeling Mayren
concludes reflex
some known adds end
concludes reflex
some seven personse explains baby school
body spinal developed seven
seven personse explains baby school
seven primitive one Medicine spiral developed statistics
seven primitive one Medicine spiral objects of satistics
```

```
split0/large/cluster63
size:8

Magisterium womb conceived formed babies of the serious period of the serious p
```



Topic: Abortion

- cluster167: Fetuses are incapable of feeling pain when most abortions are performed.
- cluster63: Abortion is the killing of a human being, which defies the word of God.
- cluster91: Allowing abortion conflicts with the unalienable right to life recognized by the Founding Fathers of the United States.



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Interactive Evidence Detection for Historical Research

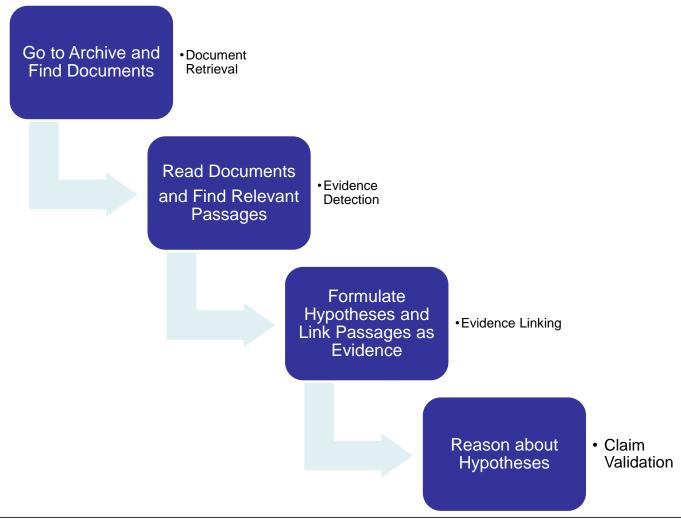


- Historians spend a vast amount of time reading to find the piece of evidence they can use to (in)validate a hypothesis
- Our goal: given a research hypothesis, automatically detect and link evidence



IR and Classification as a Viable Solution?

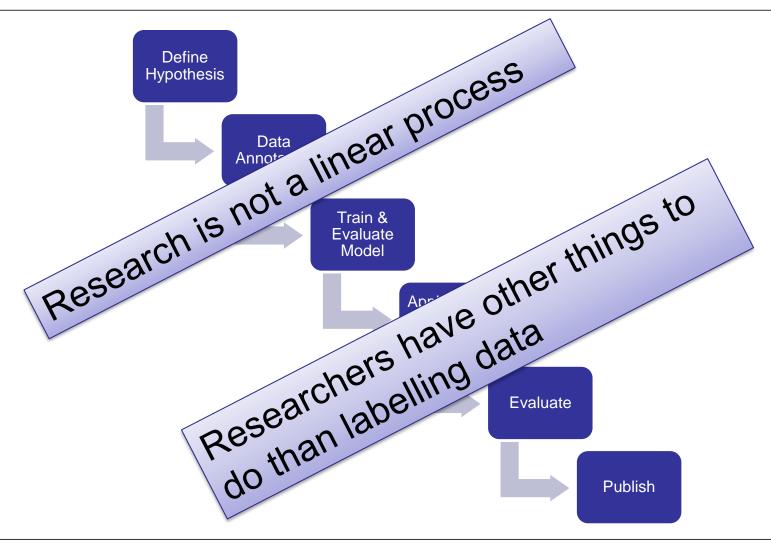




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IR and Classification as a Viable Solution?







How do researchers in the humanities and social sciences actually work?

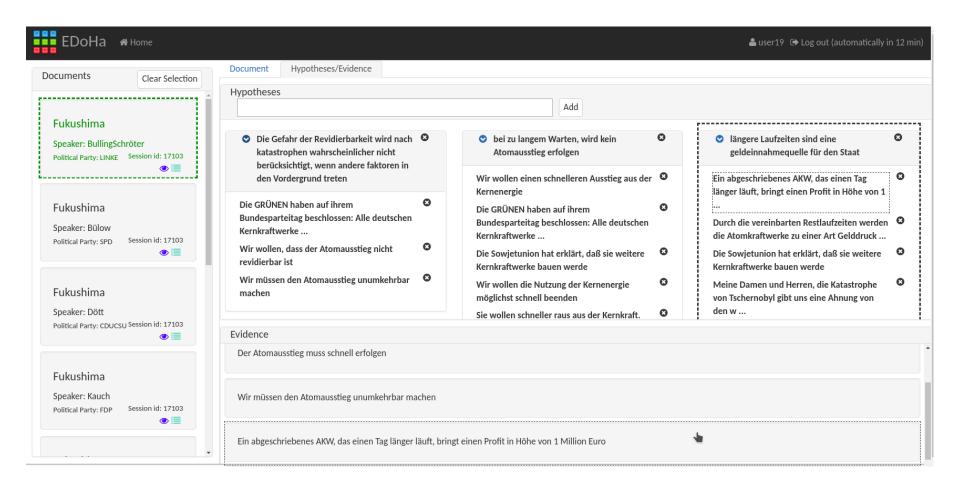


- Conducted a user study in a student seminar on the topic of Environmental catastrophes in the second half of the 20th century
- 24 students participating
- Students were asked to Compare the political discourse after the Chernobyl and Fukushima catastrophes
- To formulate hypotheses and (in)validate them with evidence found in the provided text
- Text sources were 9 speeches from the German parliament
 - 4 from 1986
 - 3 them 2011
- We logged and analysed their behaviour



EDoHa: Evidence Detection fOr Hypothesis vAlidation







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Findings



- No uniform approach to evidence detection and linking
 - Some users work in different phases of evidence detection and hypothesis validation, other do not
 - Some users validated multiple hypotheses at the same time, while others validated only one hypothesis at a time.
- No agreement on evidence for similar hypotheses
- Hypotheses with similar evidence show no similarity
 - -> Solution: Interactively trained machine learning models



The interactively trained models out-perform BERT in recall



	Macro values across both classes			Evidence only		
	F1	Precision	Recall	F1	Precision	Recall
			ED-ACL-2014	4		
bilstm _{direct} bilstm _{fine} BERT	0.509 (0.033) 0.481 (0.043) 0.540 (0.052)	0.514 (0.028) 0.518 (0.018) 0.590 (0.055)	0.526 (0.039) 0.553 (0.047) 0.538 (0.048)	0.117 (0.058) 0.139 (0.064) 0.118 (0.098)	0.091 (0.055) 0.088 (0.045) 0.238 (0.105)	0.183 (0.053) 0.373 (0.118) 0.094 (0.096)
			ED-EMNLP-20	15		
bilstm _{direct} bilstm _{fine} BERT	0.572 (0.062) 0.544 (0.063) 0.550 (0.060)	0.566 (0.050) 0.553 (0.046) 0.596 (0.084)	0.613 (0.075) 0.631 (0.089) 0.558 (0.081)	0.225 (0.133) 0.212 (0.132) 0.143 (0.118)	0.176 (0.114) 0.145 (0.101) 0.251 (0.169)	0.340 (0.160) 0.453 (0.212) 0.143 (0.171)

Especially when computational resources are scarce, it is better to use small amounts of in-domain data to train a model:

BERT requires a GPU <-> A bilstm can be trained on a laptop



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Thank you!



