

Interlinking Social Science Research Data and Social Media Data

RQ: How can survey data be *interlinked* and *complemented* with data from different sources?

The collection of social science research data is expensive. Therefore, secondary research and exploitation of new information sources are important. This work presents two machine learning based approaches to help overcome current challenges.

Finding and reusing survey questions and items

- Interlinking of social science survey items across surveys.
- Foster reuse of data by extending search applications and data catalogs with new features
- >100 000 suitable survey items from the GESIS archive
- Taxonomy and vocabulary for the description of survey items, so called question features
- Multiclass classification for the Information Type feature

Q: „How would you rate the current economic conditions in Germany?“

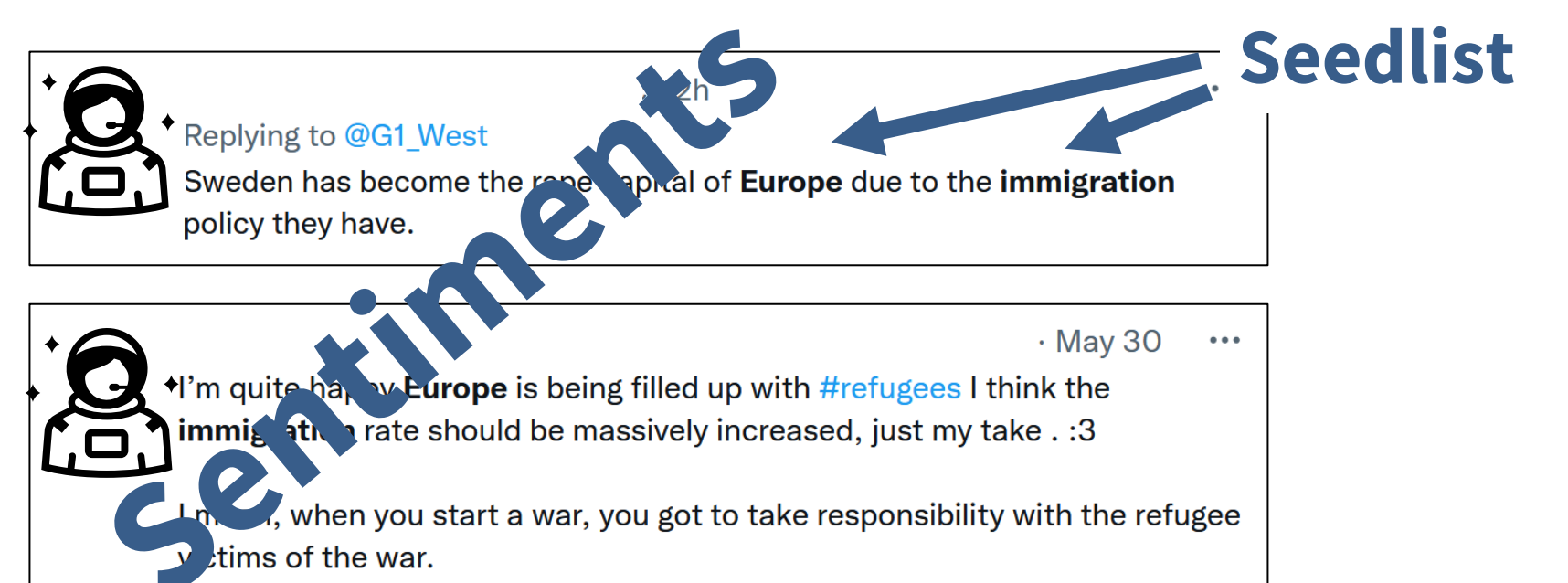
Information Type	<input checked="" type="checkbox"/> Evaluation	<input type="checkbox"/> Fact	<input type="checkbox"/> Cognition
Focus	<input type="checkbox"/> Self-focus	<input type="checkbox"/> Family-member	<input checked="" type="checkbox"/> Object-focus
Time Reference	<input type="checkbox"/> Past	<input checked="" type="checkbox"/> Present	<input type="checkbox"/> Future
Periodicity	<input checked="" type="checkbox"/> Point in time	<input type="checkbox"/> Time span	<input type="checkbox"/> Periodic p.
Relative Location	<input type="checkbox"/> Apartment	<input type="checkbox"/> Neighborhood	<input type="checkbox"/> Country
Geo. Location	<input type="checkbox"/> <Continent>	<input checked="" type="checkbox"/> <Country>	<input type="checkbox"/> <City>

Example from ALLBUS '18

Approach	L1 (3 classes)		L2 (9 classes)	
	micro-f1	macro-f1	micro-f1	macro-f1
LSTM	0.7644	0.7462	0.5764	0.5370
SVC	0.5934	0.4593	0.3991	0.2097
RF	0.5482	0.5241	0.3636	0.2423
LogReg	0.5416	0.5232	0.3368	0.2682
MNB	0.5207	0.5135	0.3373	0.2563

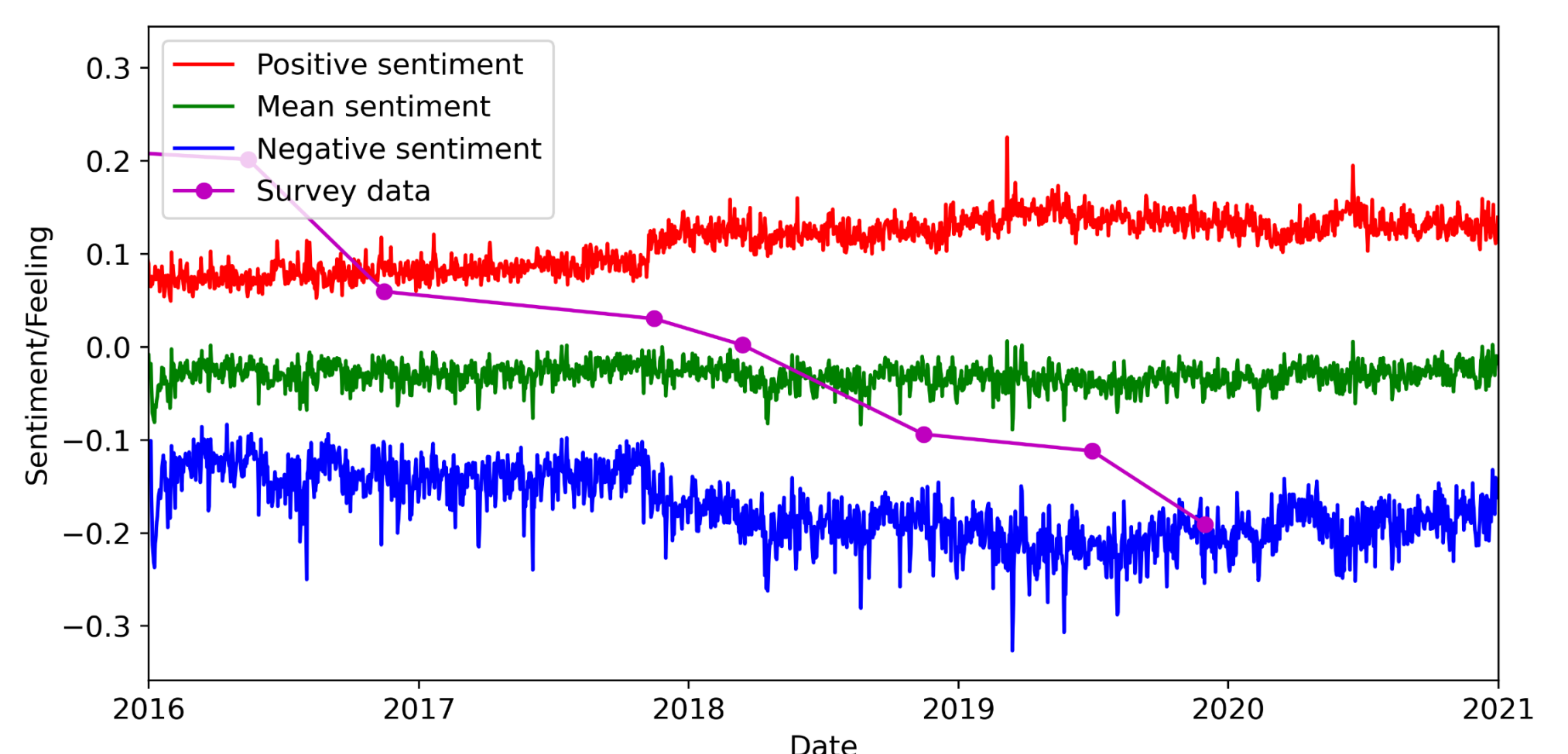
Example from ALLBUS '18

Extension with social media discourse data



Extraction of immigration related tweets based on seed terms

- Micropost retrieval from Twitter for a given topic
- Reduction of humanly-introduced bias (vocabulary mismatch)
- Systematic evaluation of methods in extraction pipeline
- Based on large scale data from the TweetsKB Twitter archive (10 billion tweets, since 2013)
- Generation of seedlists based on term frequency and semantic similarity (word embeddings)
- Operationalization through sentiment analysis
- Restriction to certain demographic groups



Sentiment of immigration tweets in the UK 2016-2021

Machine learning has many applications in the area of interlinking survey data. Thereby, a major challenge will be to preserve acceptance by social science researchers. Provenance and transparency are key aspects. Also, a comprehensive evaluation of the applied ML methods will be required.

