

Word embeddings for predicting political affiliation based on Twitter data

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Motivation

- Plethora of means to communicate political alignment
- Twitter, one of the main source for political opinion expression
- Quantitative analysis of German party affiliations of a Twitter user not explored

Approach proposed: deep learning based classification model using state-of-the-art word embeddings

Methodology

- Using a CNN to learn features from different word-lengths for classification
- Training done by feeding data using one-hot-encoded labels
- Cross Entropy to define the loss and also to optimize the network

Results

- Accuracy
 - per single tweet for person low accuracy 0.55
 - Insufficient number of words
 - Retweets
 - Non-German messages
 - Low numbers of Tweets.
 - per multiple tweets high accuracy 0.817
- Visualization through a political compass
 - Users appear close to their respective political party
 - Parties with similar Ideologies appear close to each other

Conclusion and Future Work

- CNN is able to nicely separate political figures concerning party affiliations
- CNN gives better results when used for short sentences such as tweets
- Further research may tackle:
 - How the analysis would be affected if the embeddings were to be taken from intrinsically political data samples
 - Consider RNNs or other topologically different architectures

Data Sets and Feature Extraction

Dataset Collection

- Tweets made by political figures on Twitter from the 7 biggest parties: CDU, CSU, SPD, FDP, Grüne, Linke, AFD
- Categorized data set taken from www.wahl.de/politiker
- Dividing data into train set and test set for supervised learning

Feature Extraction

- Obtain vector representation of words under some similarity metrics (word embeddings)
- Word2Vec model - pre-trained on 200 million German Tweets
- Each word of the Tweet is represented as a 200-dimensional vector

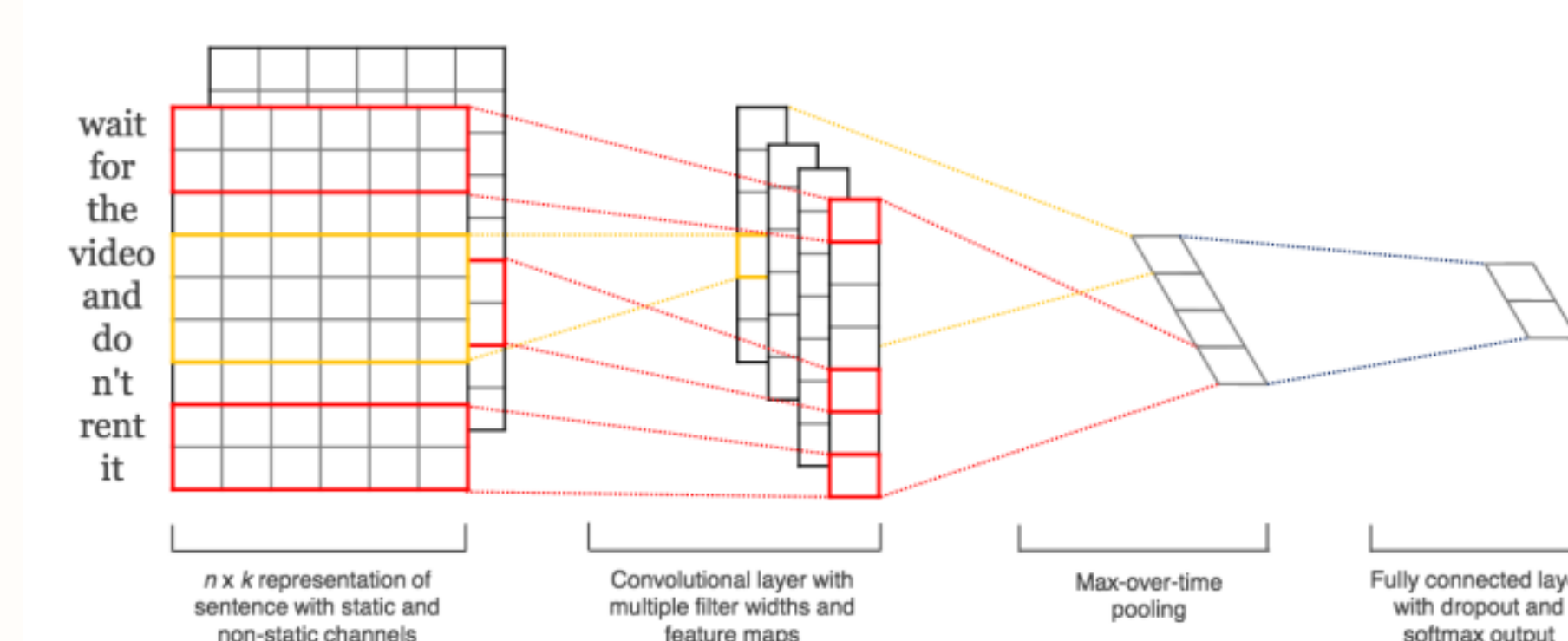
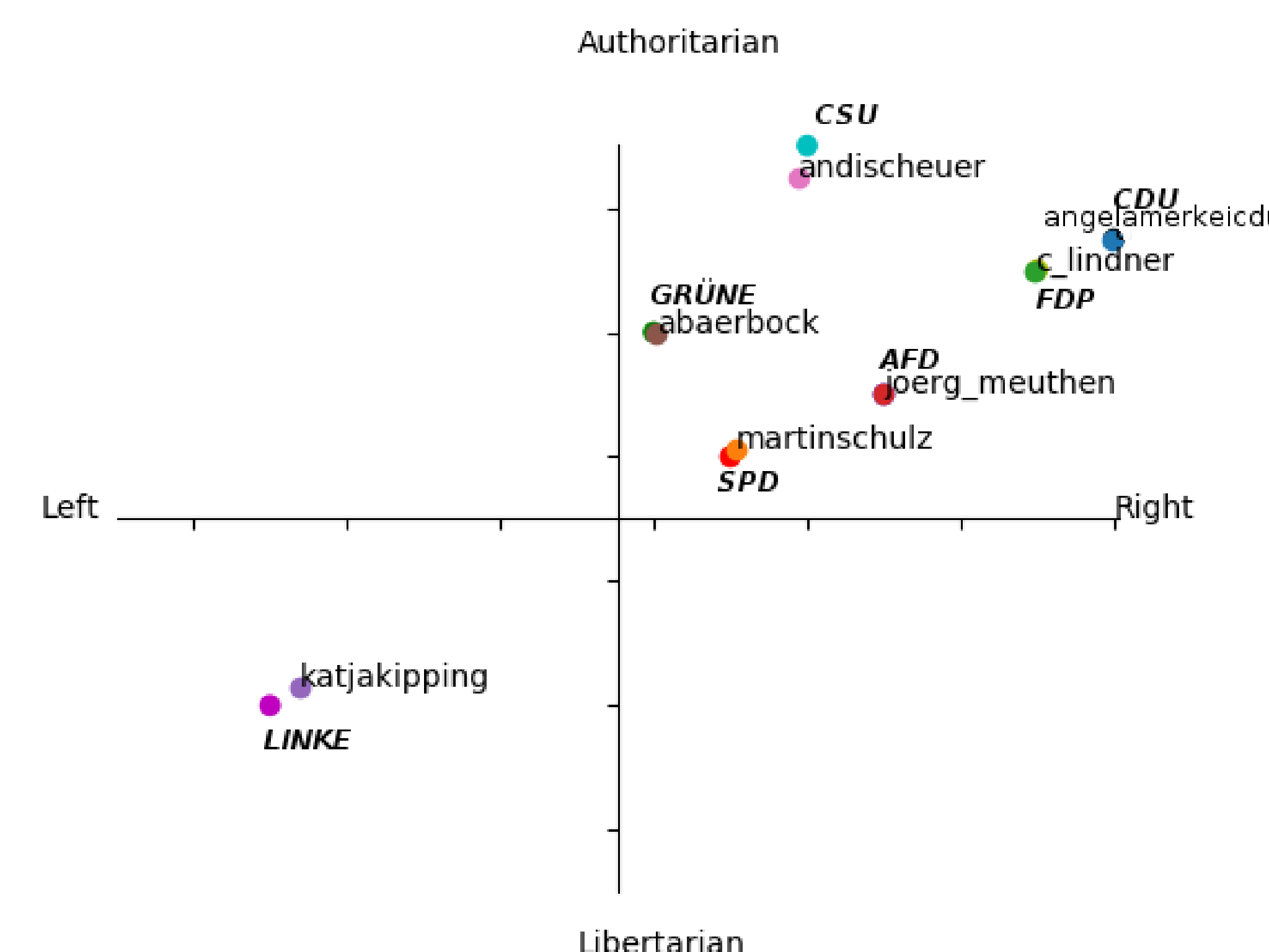


Figure 1: Model Architecture [Britz, 2015]



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