

Predicting Political Affiliation from Tweets

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Motivation

Knowledge of public opinion is a powerful tool for governments and companies to have, but is difficult to measure on a large scale. Currently, opinion polls are used, but these only sample a small portion of the population. However, deep learning can potentially be used on social media, such as Twitter, to gauge public opinion on a large scale.

For this project, we develop a classification algorithm to predict political party affiliation from tweets.

Data & Features

We used Kyle Pastor's "Democrat vs. Republican Tweets" dataset from Kaggle [1]. The dataset contains 86,460 labeled tweets from the 200 most recently elected congressmen (49% Democrat, 51% Republican). The dataset was split 80/10/10 between train/dev/test.

Sample Republican Tweet

Rep. Lou Barletta O Why are American citizens and legal immigrants always put at the back of the line by the Nancy Pelosi Democrats?

We are at the table to find a #DACA solution, but "demanding" from the American people the "right" to citizenship is a perversion of the law and the Constitution.

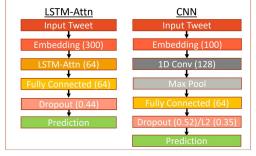
Sample Democrat Tweet

In a desperate effort to appear fiscally responsible, @POTUS is urging his #GOP allies to sacrifice health care for children from low-income homes. It's an attack on the most vulnerable among us. We must fight to save #CHIP from this cruel chopping block.

To preprocess the raw tweets, we removed hyperlinks, symbols, punctuation, and twitter handles. We then padded each tweet to the maximum tweet length and embedded each tweet using the Keras embedding layer.

Model

We created and optimized two models: a LSTM-Attention and a CNN model. We also experimented with other types of recurrent network types. The models were coded in Keras with Tensorflow backend.



Results

Model	Train Acc (%)	Valid Acc (%)	Test Acc (%)
CNN*	96.9	84.5	82.4
LSTM-Attn*	91.2	85.2	<u>83.9</u>
LSTM	91.3	84.6	83.3
GRU	88.6	84.8	83.6
LSTM (Bi-Dir)	90.0	84.6	83.3
GRU (Bi-Dir)	90.6	84.4	83.6

Confusion Matrix

LSTM-Attn Model	Predicted Dem.	Predicted Rep.
Labeled Dem.	3442 (39.9%)	787 (9.1%)
Labeled Rep.	608 (7.0%)	3808 (44.0%)

*performed random hyperparameter search

Discussion

Model Performance

- · LSTM-Attn. performs better than CNN due to its ability to learn long-term dependencies
- CNN still performs remarkably well for having such a simple architecture

Model Confidence (LSTM-Attn) Predicted Democrat with high confidence (p = 0.99):

Predicted Republican with high confidence (p = 0.04):

Unable to discern (p = 0.5):

Tonight I'll vote for the Regulatory Accountability Act to put a stop to Obama's expensive regulations. Cong. Mike Simpson O

Error Analysis (LSTM-Attn)

Tweets with no political content, short tweets (3-6 words), and tweets with multiple people mentioned were commonly misclassified

Future Work

- Character-level Embeddings: Twitter language is often quite different from standard English, so character-level embeddings have been suggested to work better for tweets [3].
- Different NN Architectures: RCNN's [4], Bi-directional LSTM's with max-pooling [5], and CNN-Attention [6] show promise in sentence classification tasks.
- Further hyperparameter tuning, text preprocessing, use existing embeddings (i.e. GloVe) for unseen words.