Novel Approaches to Analyzing and Distinguishing Fake and Real News to Mitigate the Problem of Disinformation

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Overview

Identifying fake news has become an important challenge. Increasing usage of social media has led an increase in the number of people who can be influenced, thus the spread of fake news can potentially impact important events. Fake news has become a major societal issue and a technical challenge for social media companies to identify.

Our goal

Distinguish between the real and fake news.

What is disinformation?

False information deliberately and often covertly spread (as by the planting of rumors) in order to influence public opinion or obscure the truth.

Our approaches

We addressed the problem of fake news identification using three approaches to make it manageable and more accurate:

1. Sociocultural and textual approach. It allows us to identify the rhetorical and textual characteristics that distinguish “real” or “fake” information.
2. Data science approach. It helps to dig into the data analytic by building the words and phrases frequencies,
3. Deep learning approach. We built a binary classifiers that extract features from fake and real news using deep learning models, such as Long Short Term Memory (LSTM), Recurrent Neural Network (RNN) and Gated Recurrent Unit (GRU).

Dataset

We used the dataset extracted using the FakeNewsNet* tool. The final dataset contains both fake and real news in the political domain.

Total Number	Fake	Real
News articles	432	624
News articles with text	420	528
News articles with images	336	447

*https://github.com/KaiDMML/FakeNewsNet

Sociocultural textual analysis

HEADLINES

In both real and fake news articles, headlines will often begin with a word like “breaking” in all capital letters to catch a reader’s attention and communicate an urgency to the information in the article.

Fake news

Real news

Example of the use of all capital letters in both fake and real news story headlines to communicate urgency

SOURCE

Fake news will often report information without including a source, simply using phrases such as “told reporters” or “is being reported” to signify credibility rather than offering actual sourcing.

Real news

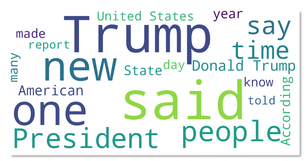
USE OF ADJECTIVES

WRITTEN INFORMATION

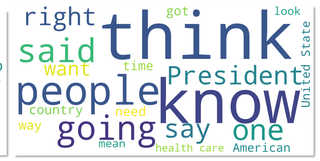
- Missing author biography
- Author biography that provides no information about their journalistic affiliation
- Irrelevant or content-less photos
- Missing publication date
- Erroneous metadata

Data Analysis

WORDCLOUDS



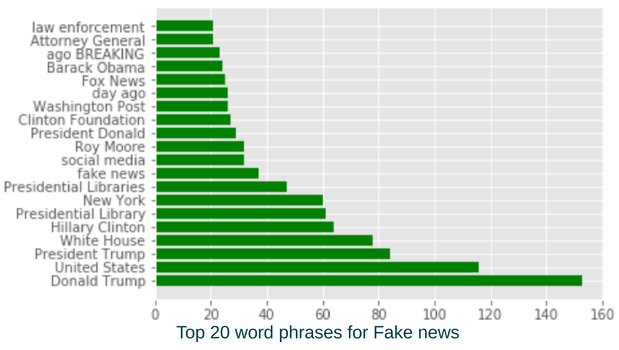
Fake news



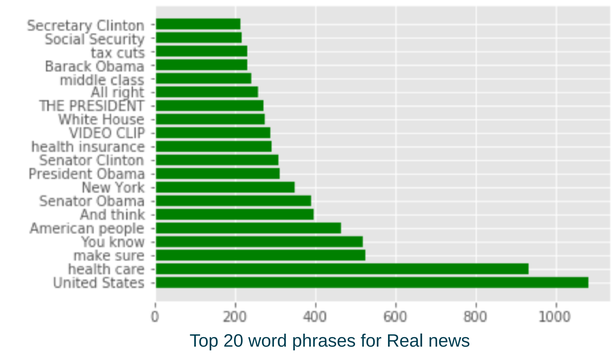
Real news

In fake news the word "Trump" word is more prevalent, while in real news "president" is used to address the key news maker. This distinction between the office of the president and the individual begins to show that fake news has a stronger investment in affective and ideological approaches than real news.

BIGRAMS



Top 20 word phrases for Fake news



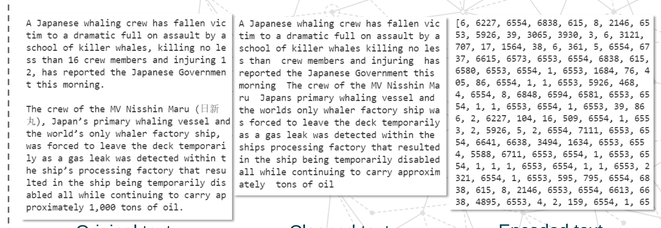
Top 20 word phrases for Real news

Similar to the wordcloud, the bigrams show in real news, public figures are often referred to by official titles: Secretary Clinton, Senator Obama, Senator Clinton. In fake news, however, first names are given instead, or in addition, such as “President Donald.”

Deep Learning

WORD EMBEDDING

We encoded the dataset using Byte Pair Encoding (BPE). BPE is a simple data compression technique that iteratively replaces the most frequent pair of bytes in a sequence with a single, unused byte.



DEEP LEARNING MODELS

We have applied three deep learning models to do feature extraction and perform binary classification.

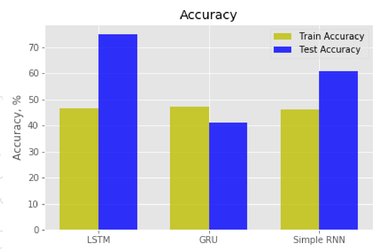
Hyperparameters for training the models

Parameter	LSTM	GRU	RNN
Layers	1	2	1
Activation	-	SeLu	SeLu
Activation output	Sigmoid	Sigmoid	Sigmoid
Optimizer	Adam	Adam	SGD
Epochs	3	1	5
Nodes 1-layer	15	15	15
Nodes 2-layer	-	1	5

RESULTS

Test Accuracy

LSTM: 75%
GRU: 41.07 %
RNN: 60.71%



Conclusion

We can notice one of characteristics of disinformation is its ideological context. This is the first time this kind of sociocultural textual analysis has been conducted using this dataset. Deep learning models showed reasonable results, but it might not be generalized to other types of datasets.

References

Alina Vereshchaka, Seth Cosimini, and Wen Dong. Novel Approaches to Analyzing and Distinguishing Fake and Real News to Mitigate the Problem of Disinformation, 2019

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