DEBUNKING FAKE NEWS BY LEVERAGING SPEAKER CREDIBILITY AND BERT BASED MODEL

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INTRODUCTION

FAKE NEWS

Fake news is a type of propaganda that consists of deliberate misinformation or hoaxes spread via traditional print and broadcast news media or online social media.

INTRODUCTION

A new study kills the notion that fake news swung the US election to Trump

Singapore Study auggeste that exposure to memorination vaccines

A team of Howard University researchers wants to know how disinformation impacts Black people

Facebook, Google extend political ad ban amid misinformation rise

Trump may owe his 2016 victory to

'fake news,' new study suggests

MOTIVATION

MOTIVATION

- Most of the fake news content is created by fabricating true news so it is more reasonable to perform multi-class classification that can help us assess the level of falsification of the fake news.
- Adding additional details about the subject can make the prediction model more efficient by assisting it in making more informed decisions.

FEATURE EXTRACTION

DATASETS

LIAR	Speaker2Credit
Statement	
Topic	
Speaker Profile	Global Credit History
Credibility Count	
Six different labels	

The choice of a political news dataset was due to the fact that false political news travels nearly three times faster and appeals to a larger audience than news of any other category.

FEATURES

Speaker Statement

Speaker Profile

Attributes

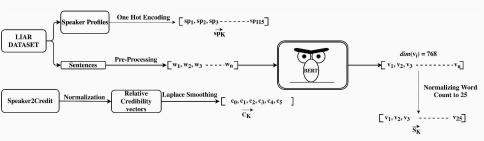
job of the speaker, political party affiliation of the speaker, state of residency of the speaker, topic on which statement is made, and the location of the speech

Speaker Credibility Speaker2Credit

CREATING CREDIT VECTORS FROM CREDIT HISTORY

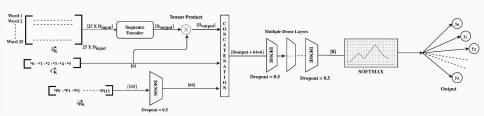
$$credit_{i_{\alpha}} = \left(\frac{count(\mu_0|i) + \alpha}{\sum_{\mu} count(i) + 5\alpha} \cdots \frac{count(\mu_5|i) + \alpha}{\sum_{\mu} count(i) + 5\alpha}\right)$$

FEATURE GENERATION





PROPOSED ARCHITECTURE





EXPERIMENTAL SETUP

- Dataset: LIAR (labeled articles from Politifact with speaker credentials)
- · Baselines:
 - Hybrid-CNN (Proposed in LIAR dataset paper) [5]
 - Multi-source multi-class fake news detection (MMFD) [2]
 - LSTM-Attention [4]
 - Bi-LSTM [1]
 - CreditLSTM [3]
- Evaluation Metric: Accuracy (Fraction of dataset correctly labeled)

RESULTS

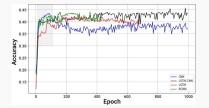


Figure 1: Accuracy Curves for Different Sequence Models

Model	Accuracy	Batch size
LSTM	0.427827	64
CNN	0.442835	32
LSTM-CNN	0.468040	128
RCNN	0.443182	128

Table 1: Accuracy comparison among different sequence encoding models.

Features	Accuracy
Word2Vec	0.2636
BERT	0.2718
Word2Vec + MetaData	0.2741
BERT + MetaData	0.3099
BERT + MetaData + Credit	0.4680

Table 2: Performance analysis of different feature models

Model	Accuracy
Hybrid-CNN [<u>5</u>]	0.274
MMFD [<u>2</u>]	0.348
LSTM-Attention [<u>4</u>]	0.415
Bi-LSTM [<u>1</u>]	0.415
CreditLSTM [<u>3</u>]	0.457
BERT + MetaData + Credit	0.4680

Table 3: Performance of previous works on the LIAR dataset

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QUESTIONS?

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