# Fake news Detection

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Abstract- Large amounts of false information are created online for a variety of reasons, including financial and political benefits, due to its ease of spread online. The widespread dissemination of false information can have a serious negative impact on people's lives and society as a whole by (i) upsetting the ecosystem's authenticity balance, (ii) purposefully influencing consumers to adopt biased or false beliefs, and (iii) altering how people perceive and react to actual news and information. This fake news detection exercise attempts to categorize text information regarding spreading rumors into four separate labels (FALSE, MISLEADING, TRUE, UNPROVEN). The claim and the articles must be preprocessed before being compared for classification.

#### I. INTRODUCTION

Newspapers, tabloids, and magazines have given way to online news sources, blogs, social media feeds, and other digital media formats as the news medium changed. Consumers now have more access to the most recent news at their fingertips. In their current form, these social media platforms are very effective and helpful for enabling users to debate, share, and discuss topics like democracy, education, and health. However, some organizations also utilize these platforms negatively, frequently to obtain financial advantage, and occasionally to sway public opinion, influence people's attitudes, or propagate satire or absurdity.

What percentage of the news we read on social media and on purportedly "reliable" news websites can we trust? It is very simple for anyone to post whatever they want, and while that may be acceptable, there is the possibility of going too far. Examples of this include posting false information online to incite panic, telling lies to influence another person's decision, or essentially anything else that could have long-lasting effects. The amount of information available online makes it difficult to distinguish between true and untrue. Due to the aforementioned reasons, it is important to detect fake news.

### II. RELATED WORKS

# 1) Fake News Detection Using Machine Learning Ensemble Methods -

This study investigates many textual characteristics that can be utilized to identify between true and false contents. These characteristics are used to train a variety of machine learning algorithms using various ensemble approaches, and four real-world datasets are used to assess their performance. The suggested ensemble learner strategy outperforms the individual learner approach, according to experimental evaluation.

# 2) Fake news detection based on news content and social contexts: a transformer-based approach -

In order to identify false news, the suggested approach uses data from news articles and social situations. The model is built on a Transformer architecture, which consists of two parts: an encoder to extract meaningful representations from the fake news data and a decoder to forecast behavior based on historical data. To further aid in the classification of the news, a number of characteristics from the social contexts and news content in our model are used. In addition, a successful labeling method to solve the label shortage issue is proposed. The algorithm can detect fake news more accurately and quickly than baselines within a few minutes of it spreading (early detection), according to experimental results using real-world data.

### III. MODEL DESCRIPTION

# 1) Data Scraping using Newspaper3k

Newspaper3k is a Python package for extracting content from the web. While parsing for lxml, it uses the requests library and depends on BeautifulSoup. Newspaper3k package web scrapes websites using advanced algorithms to extract all of the helpful text. On websites for online newspapers, it works incredibly well. Newspaper3k can scrape the complete article text for you as well as other types of information like the publish date, author(s), URL, photos, and video, to mention a few.

Image 1: Summary Generated for label False (0)

G6	i	*	1	×	(	<b>✓</b>	fx	An image shared on Facebook claims customers will be required to provide ID and proof of COVID-19 vaccination to Nov. 1.	o enter all \	Walmar	t stores ir	Can	ada starting	3
								Verdict: FalseWhile proof of vaccination is required for certain venues in parts of Canada, Walmart Canada has not vaccination for entry to all of its stores.	announce	d a poli	cy requiri	ng ID	and proof	of
								Now, Facebook posts containing a supposed notice from Walmart Canada claim it will be implementing a new pand Canadian media outlets such as The Globe and Mail, National Post and CBC also haven't reported that Walmart requirement that customers must provide proof of vaccination.			Š. a.			
4	А		В					C	D	E	F		G	Н
1	Country	(rR	eviev	/ Da	Clair	m			Source	Label	Fact-	heck	summary	
2	German	у С	ct 15	, 20	A vi	deo i	circu	ulating on the Internet claiming that masks used to contain the corona pandemic are harmful to health (archived here)	.person		0 https	://dp	«Durch	
3	United S	ta S	ep 17	, 20	Vice	Pres	ident	Kamala Harris "admits" that COVID vaccines don't work.	website		0 https	://lea	No Vax Is	
4	United S	ta S	ep 17	, 20	US t	o wit	hhold	benefits from unvaccinated veterans	multiple p	)	0 https	://fa	US not	
5	United S	ta S	ep 17	, 20	U.S.	Pres	ident	Joe Biden said that the COVID-19 vaccine would protect people against hurricanes.	No data		0 https	://w	Advertis	
6	Canada	S	ep 16	, 20	cust	ome	rs will	be required to provide ID and proof of COVID-19 vaccination to enter all Walmart stores in Canada starting Nov. 1	facebook		0 http:/	//che	all of its	
7	South Af	ri S	ep 16	, 20	List	of en	ploye	ers where Covid shot is not mandatory	multiple s		0 https	://fa	List of	
8	United S	ta S	ep 16	, 20	Post	tal wo	rkers	s aren't included in the Biden administration vaccine mandate	social me	c	0 https	://w\	The post,	
9	Australia	S	ep 16	, 20	Aus	tralia	n MP	Craig Kelly has leaked a secret list of the real side effects of the COVID-19 vaccine.	multiple s	ı	1 https	://w	ΣÎμ	
10	German	v S	en 16	20	Rert	ram	HÃgii	ssler said that Corona was not the cause of death in 80 nercent of the official Covid deaths.	multiple s	ı	0 https	1/00	"Coron	

Image 2: Summary Generated for label Misleading (1)

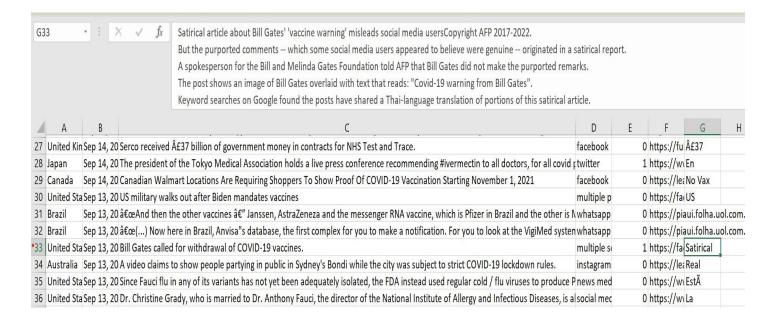
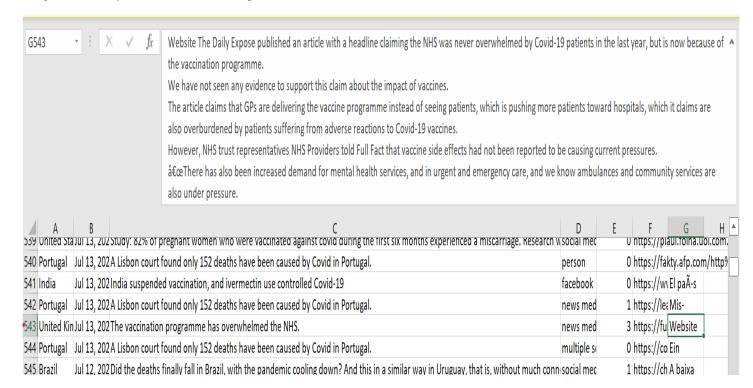


Image 3: Summary Generated for label True (2)

G42 • If you look now, there's 300 percent more cases in this country today than a year ago when we had no vaccines at all."						
		Ron DeSantis: "I think the problem I have with Joe Biden, more than anything, this guy doesn't take responsibility fo	r anything.			
		The same dataset shows that on September 10, 2020, the U.S. recorded 35,100 new cases of COVID.				
		The difference between these two figures marks a 312 percent increase, meaning DeSantis was correct in terms of a	average dail	y new cases.		
		The number of new COVID cases recorded in the U.S. on September 10, 2021, marked around a 300 percent increase	e from the	figure on Sept	ember 10, 2020.	
A	В	С	D	E	F G	Н
39 Spain	Sep 13, 20 Protest in Barce	elona against the covid vaccination passport	twitter	0 http	os://m:Se estÃj	
40 United	Sta Sep 13, 20 The Red Cross I	has banned plasma donations from COVID-19 vaccinated because their blood is contaminated.	Social Net	1 http	os://w\Ο	
41 Brazil	Sep 13, 20 Most serious ca	ases nowadays are vaccinated in Brazil. We see this on a daily basis over the last few weeks.	whatsapp	0 http	os://piaui.folha.	uol.com
42 United	Sta Sep 13, 20 There were 300	percent more cases recorded in the U.S. on September 10, 2021 than the same date a year ago.	person	2 http	os://w\"If you	]
43 Germa	ny Sep 13, 20 Germany halts	all COVID-19 vaccines, says they are unsafe & no longer recommended.	multiple so	0 http	os://facGermany	/
44 United	Sta Sep 13, 20 In September 2	1021, U.S. President Joe Biden directed the U.S. Department of Veterans Affairs to stop providing healthcare benefits t	news med	0 http	os://w\Advertis	

Image 4: Summary Generated for label Unproven (3)



Newspaper3k also generates a summary of the article that gives us the main points without reading the complete thing. After the data has been extracted, it can be merged and saved in a variety of forms, including CSV, JSON, and even pandas. It is available in almost 30 languages in Newspaper3k.

We have primarily used the Newspaper3k package to generate a summary of the news article given as URLs in the training dataset. We added a column named "summary" and stored the output(concise short paragraph) of the entire news article in the respective rows.

The examples of summaries generated for each of the unique labels are as follows:

Summary generated when the label is 0 - In the Image1 we can see the summary for the Row Number 6 along with the claim that is given. Since it is not true, the label is 0

Summary generated when the label is 1 - In the Image2 we can see the summary for the Row Number 33 along with the claim that is given. The label here is 1 as the information is misleading.

Summary generated when the label is 2 - In the Image3 we can see the summary for the Row Number 42 along with the claim that is given. The label here is 2 and we can see that the summary supports the claim and thus the label is True.

Summary generated when the label is 3 - In the Image4 we can see the summary for the Row Number 543 along with the claim that is given. We can clearly see in the summary that there is no evidence to support the claim and thus the label is 3 that is unproven.

## 2) Data Preprocessing

Mapping of Source names

- a) The source column had many redundant values which were wrongly spelled. We have corrected them by mapping those values to their respective source. For example- the misspelled sources were "perseon", "facbook", etc. These values were correctly mapped under as person and social media.
- b) The source column had a lot of entries which were names of different persons. All those sources were mapped to the source as- person. Example- Names like anthony fauci, bernie sanders, beverley turner, boris johnson, etc were all mapped under the source person.
- c) All the social media websites like youtube, facebook, instagram, twitter, tiktok etc were all mapped to the source social media.
- d) Values like unknown, video, photo, study, meme etc were all mapped to a source as Others.

# data['Source'].value\_counts()

social media	3675
multiple sources	1089
person	930
others	266
news media	266
websites	158
Name: Source, dtype:	int64

### Mapping of Country names-

Considering only Top 10 countries which account for 65% of the data and binning the others countries together as "Others" -

#Top 10 contributing to 65 percent of the data..So using them
data['Country (mentioned)'].value\_counts()[:9]\*100/len(data)

United S	states	21.788847	
India		14.379699	
Brazil		10.197368	
United K	ingdom	6.876566	
Italy		3.947368	
China		3.493108	
France		2.662907	
Spain		2.474937	
Indonesi	.a	2.427945	
Name: Co	ountry (	mentioned), dtype:	float64

# 3) One-hot encoding of Countries and Sources columns

One hot encoding is a method that involves transforming categorical information into a format that is given to the Machine Learning algorithms to help them perform better at prediction.

Data that don't relate to one another can benefit from one hot encoding. The arrangement of numbers is treated as a significant characteristic by machine learning algorithms. This technique of encoding makes our training data more useful and expressive, and it can be rescaled easily. By using numeric values, we more easily determine a probability for our values. In particular, one hot encoding is used for our output values, since it provides more nuanced predictions than single labels. We have used the sklearn library in python to perform one-hot encoding.

## 4) Train-Test Split

Perform train-test split of our data- The train\_test\_split() method is used to split the data into train and test sets. We must first separate our data into features (X) and labels (y). The dataframe is split into the X train, X test, Y train, and Y test sections. The model is trained and fitted using the X train and y train sets. The model is tested to see if it correctly predicts the outputs and labels using the X test and y test sets.

The size of the train and test sets can be explicitly tested. In our split, the training data is 80% and the remaining 20% is testing data.

# 5) Similarity Calculation

### a) spaCy's Similarity score:

After doing the splitting, we use en\_core\_web\_sm -spaCy's similarity calculator.spaCy is a free and open-source library for Natural Language Processing (NLP) in Python, We are able to calculate the similarity between sentences using this and we can generate a score as well. The value ranges from 0 to 1, with 1 meaning both sentences are the same and 0 showing no similarity between both sentences

# b) Cross Encoder Semantic Similarity Score:

Since cosine similarity will not be able to capture the semantics or context of the sentences, we've also used a Cross-Encoder approach. Here, we pass both sentences simultaneously to the Transformer network. It gives an output value between 0 and 1 indicating the similarity of the input sentence pair. Cross-Encoders can be used whenever you have a predefined set of sentence pairs you want to score. For example, you have 100 sentence pairs and you want to get similarity scores for these 100 pairs.

#### 6) DecisionTreeClassifier

A decision tree is a flowchart-like tree structure in which an internal node represents an attribute or feature, the branch represents a decision rule, and each leaf node represents the outcome. The topmost node in a decision tree is known as the root node. It learns to partition on the basis of the attribute value.



We are using the following features one-hot encoded values of country column and source columns along with the similarity score. Next, we try to predict the labels for the test split from the train data by applying our trained model.

We got an accuracy of 77.36% for the test split using the spaCy's similarity score as feature

```
from sklearn import metrics
metrics.accuracy_score(y_test,dtree_predictions)
```

0.7736883320281911

We got an accuracy of 77.05% for the test split using the Cross encoder semantic similarity score as feature

- from sklearn import metrics
  metrics.accuracy\_score(y\_test,dtree\_predictions)
- 0.7705559906029757

# 7) Making Predictions

Similarly, we repeat the same steps for the testing data. Repeating the same preprocessing steps for test data - Summary generation, country and source binning and calculating the similarity scores. We're then using the trained Decision Tree Classifier model to predict for the test data points.

We got a test accuracy of 86.19% using the Semantic similarity score as a feature and 85.63 using the spaCy's similarity score as a feature.

$\odot$	Submission1_CrossEncoder.csv  Complete · R@HuL · 18h ago · Cross Encoder	0.86197	~
$\odot$	Submission2_CosSim.csv Complete · R@HuL · 18h ago · Spacy similarity	0.85633	

#### IV. EXPERIMENT

We tried out quite a few approaches to perform fake news detection before finalizing on our approach. Here is the list of approaches we tried.

- 1) Tf-IDF approach: We scraped the URLs provided in the Fact-checked Article column and performed a Tf-IDF vectorization to identify the critical keywords and extracted relevant sentences containing these keywords to form a summary and calculated a cosine similarity score between the claim and the extracted summary. Used the score along with Country (mentioned), Source as features for a multinomial logistic regression model. The cosine similarity scores could not capture the semantic similarity between the sentences and the Tf-IDF way of generating summary was not sufficient to decide on the labels.
- 2) BERT Question Answering Approach: We used the fine-tuned BERT model from the Hugging Face Transformers library to answer questions. It was trained using the CoQA dataset (Conversational Question Answering dataset). The model reads the user-provided text context and attempts to respond to any questions posed by that text context. We framed the problem in the following way: The scraped data being the user-provided text and questions being the rephrased versions of the claim asking if it is fake,true,misleading or unproven.

- The answer would be the label indicating the category of the news.
- 3) BERT Text Entailment Approach: We used the fine-tuned BERT model from the Hugging Face Transformers library to answer questions. It was trained using the MultiNLI dataset (Multi-Genre Natural Language Inference dataset). If a premise is true, then there is entailment. Simply put, a sentence Y is said to entail a sentence X if X is true and Y can be deduced logically from it. A pair of sentences in the dataset we used can either entail each other, be neutral, or contradict each other. Here, the premise is the generated summary (which is always true) while the conclusion being the claim (which can entail, contradict, or be neutral to the premise).

### V. FUTURE WORKS

- Translation: Some of the webpages had non-English content in them. We performed similarity checks between claims that were in English versus the summary that was generated in other languages. So, the scores were not accurate. So, In the future, we'd include language translation of the summary into English to get accurate measures of similarities.
- 2. Text preprocessing: Preprocessing of the scraped content would be something we might want to take up in the future for trying out other approaches. Examples being tokenization, stemming or lemmatization, removing stop words and punctuations, etc. Since, our current models don't require us to do that for similarity score calculations, we didn't do that for the final version.
- Cross Validation: Different cross validation techniques like Holdout Method, K-Fold Cross-Validation can be used to identify the best possible parameters for the Decision Tree Classifier model.

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