



# **Data Analysis 2 Project**

**Duration: 1st Semester** 

Task 2 (Fake news)

Submitted By:

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# **Introduction:**

Text analysis involves a collection of techniques and algorithms designed to derive meaningful information from text data. Rather than being a single approach, it encompasses a variety of methods that share a common goal: to transform unstructured text into structured data, enabling further analysis. To begin, let's consider a dataset of textual information.

As one of the most straightforward yet effective approaches, text analysis facilitates the rapid development of machine learning models that can process and interpret large volumes of text data efficiently. This process is highly valuable for various applications such as sentiment analysis, topic modeling, information extraction, and document classification.

In text analysis, the data typically involves high dimensionality, as each word or phrase can represent a unique feature. These methods are extensively used for tasks like spam filtering, sentiment detection, and categorizing reviews or articles. The advantage of text analysis lies in its ability to quickly process and analyze large datasets, making predictions and extracting insights with ease, even when the data contains thousands of features.

Text analysis models work by estimating the likelihood that a specific word or set of words is associated with a particular category or sentiment. They use probabilistic and statistical techniques to identify patterns, relationships, and trends within the text data. Although it is common to assume independence between words or phrases, in practice, this is rarely the case. However, by leveraging probabilistic algorithms and machine learning models, text analysis remains a robust and efficient solution for processing complex and high-dimensional text data.



# **Data Features:**

	id	title	author	text	label
0	0	House Dem Aide: We Didn't Even See Comey's Let	Darrell Lucus	House Dem Aide: We Didn't Even See Comey's Let	1
1	1	FLYNN: Hillary Clinton, Big Woman on Campus	Daniel J. Flynn	Ever get the feeling your life circles the rou	0
2	2	Why the Truth Might Get You Fired	Consortiumnews.com	Why the Truth Might Get You Fired October 29,	1
3	3	15 Civilians Killed In Single US Airstrike Hav	Jessica Purkiss	Videos 15 Civilians Killed In Single US Airstr	1
4	4	Iranian woman jailed for fictional unpublished	Howard Portnoy	Print \nAn Iranian woman has been sentenced to	1
5	5	Jackie Mason: Hollywood Would Love Trump if He	Daniel Nussbaum	In these trying times, Jackie Mason is the Voi	0
6	6	Life: Life Of Luxury: Elton John's 6 Favorite	NaN	Ever wonder how Britain's most iconic pop pian	1
7	7	Benoît Hamon Wins French Socialist Party's Pre	Alissa J. Rubin	PARIS – France chose an idealistic, traditi	0
8	8	Excerpts From a Draft Script for Donald Trump'	NaN	Donald J. Trump is scheduled to make a highly	0
9	9	A Back-Channel Plan for Ukraine and Russia, Co	Megan Twohey and Scott Shane	A week before Michael T. Flynn resigned as nat	0



#### **Explore the Dataset:**

We reviewed the size and dimensions of the data so that we knew the number of articles or news, the title, the content, etc. and the names of the columns present, and we concluded that the data contains 5000 news items and each news item contains four columns: title, text, author, date. Then we split the text in each row into a list of words using the str.split() function in the text column.

Then we counted the number of words in each row, i.e. in each article, using the str.len() function. The final result will be a string containing the number of words for each article in the text column. Then we calculated the descriptive statistics for the length of the texts (i.e. the number of words) in the text column, such as: count, mean, std, min, max. They were also applied to calculating the length of the titles.

The statistics for the training and testing sets are as follows

The text attribute has a higher word count with an average of 760 words and 75% having more than 1000 words. The title attribute is a short statement with an average of 12 words, and 75% of them are around 15 words. Our experiment would be with both text and title together





The news is classified into different categories including (1: Unreliable, 0: Reliable) print(news\_d.label.value\_counts()) This function counts and displays the number of cases (number of rows) that contain each category (label) in the label column

We have imported NLTK, which is a famous platform for developing Python applications that interact with human language. Next, we import re for regex. We import stopwords from nltk.corpus. When working with words, particularly when considering semantics, we sometimes need to eliminate common words that do not add any significant meaning to a statement, such as "but", "can", "we", etc. PorterStemmer is used to perform stemming words with NLTK. Stemmers strip words of their morphological affixes, leaving the word stem solely. We import WordNetLemmatizer() from NLTK library for lemmatization. Lemmatization is much more effective than stemming. It goes beyond word reduction and evaluates a language's whole lexicon to apply morphological analysis to words, with the goal of just removing inflectional ends and returning the base or dictionary form of a word, known as the lemma. stopwords.words('english') allow us to look at the list of all the English stop words supported by NLTK. remove\_unused\_c() function is used to remove the unused columns. We impute null values with None using the null process() function. Inside the function





clean\_dataset(), we call remove\_unused\_c() and null\_process() functions. This function is responsible for data cleaning. To clean text from unused characters, we have created the clean\_text() function. For preprocessing, we will use only stop word removal. We created the nltk\_preprocess() function for that purpose

()ps = PorterStemmer is the process of converting words to their roots or base forms. WordNetLemmatizer is also used in the lemmatization process, which is a process similar to Stemming but more precise. It converts words to their base forms based on linguistic rules (such as converting "best" to "good").

Stop words are also used to remove unnecessary words during processing. A counter is used to create a dictionary of stop words and is used to count the frequency of these words or to quickly check for their presence





# **Modification:**

Some data cleaning functions are used such as def remove\_unused\_c(df,column\_n=remove\_c) is used: to remove unimportant columns from the data frame and def null\_process(feature\_df) is used to process missing values. Text cleaning functions are used including:def clean\_text(text) is used to clean the text from unnecessary elements such as links, punctuation marks and unwanted characters.

NLTK text processing functions include:

def nltk\_preprocess(text) This function performs the following steps to process text using NLTK:

Clean text using clean\_text(text)

Split text into a list of words using text).split()

Remove unwanted words: Words in stopwords\_dict are removed





# Dataset after cleaning and preprocessing step:

	title	text	label
0	house dem aide didnt even see comeys letter ja	house dem aide didnt even see comeys letter ja	1
1	flynn hillary clinton big woman campus breitbart	ever get feeling life circle roundabout rather	0
2	truth might get fired	truth might get fired october 29 2016 tension	1
3	15 civilian killed single u airstrike identified	video 15 civilian killed single u airstrike id	1
4	iranian woman jailed fictional unpublished sto	print iranian woman sentenced six year prison	1
5	jackie mason hollywood would love trump bombed	trying time jackie mason voice reason week exc	0
6	life life luxury elton john 6 favorite shark p	ever wonder britain iconic pop pianist get lon	1
7	benoît hamon win french socialist party presid	paris france chose idealistic traditional cand	0
8	excerpt draft script donald trump qampa black	donald j trump scheduled make highly anticipat	0
9	backchannel plan ukraine russia courtesy trump	week michael flynn resigned national security	0





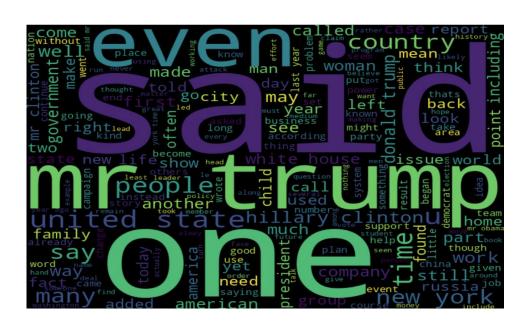
Here the WordCloud library is imported to create a word cloud. STOPWORDS is also imported, which is a list of common words that can be excluded from the word cloud (such as "the", "is", "and"), and matplotlib.pyplot is imported, which is a library used to create graphs and charts and will be used here to display the word cloud. text\_cloud = wordcloud.generate(' '.join(df['text'])) Here the word cloud is generated using the texts in the text column of the df dataset, all the texts in the text column are combined into a single text string using ' '.join(), where a space is added between each text to form a large string containing all the words, wordcloud.generate() analyzes the most frequent words in it to create a word cloud, where the most frequent words are displayed in a larger size, plt.imshow(text\_cloud) Here we display the word cloud using imshow() to display it as an image true\_n = ' '.join(df[df['label']==0]['text']) Here the dataset df is filtered to select only trustworthy news, which have label = 0 df[df['label']==0]: This statement selects all rows in the dataframe df where label = 0 (i.e. trustworthy news) ['text']: After filtering out trustworthy news, only the text column of these rows is accessed.





join(...): All trustworthy texts are combined into one large text string, with a space between each text Result: true\_n contains only trustworthy news texts in the form of a long text string , and has been applied to fake news as well

• Most frequently repeated words



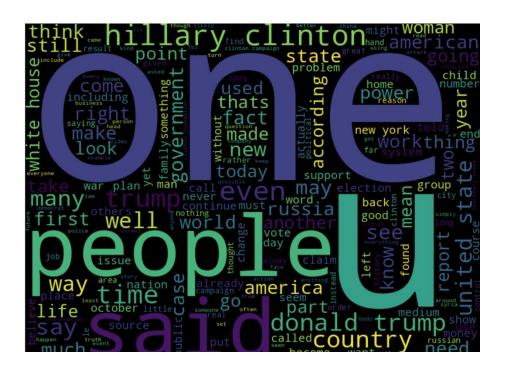




#### • Reliable news



Fake news







We have created a new dataframe named dt by copying data from the variable df which contains previously prepared or cleaned data.

dt['label'] = df['label']

This line resets or adds a label column in the new DataFrame dt If the label column already exists in dt, the values are updated, If it does not exist the column is added automatically.

We used lambda to remove Stopwords (common, unimportant words) from the text in the text column of the dt data frame, and then saved the resulting text (without Stopwords) in a new column called no\_sw.

(From import counter groups) It is a powerful tool used to count the frequency of items in a given list or text, (get\_most\_frequent\_words)

This function counts the most frequent words in a text string and makes a list of the most frequent words with the number of times they appear. Each text is split into words using split(), and then the words are added to the ()list all\_words using extend.

The final result is that all\_words contains all the words from the text.

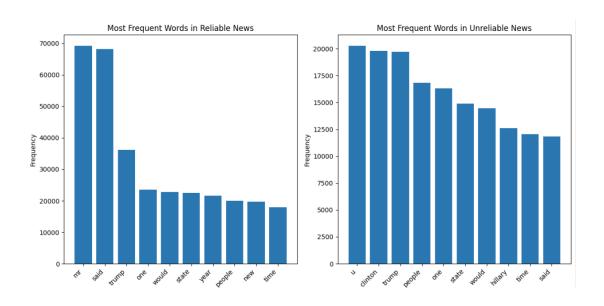




dt[dt['label'] == 0]['no\_sw']: This part filters only reliable news from the dt data frame, and then takes the text without stop words from the no\_sw column.

Result: The most frequent words in the reliable news are stored in the variable most frequent reliable.

We then drew two vertical charts: one to show the most frequent words in reliable news, and the other to show the most frequent words in unreliable news. The charts are displayed in parallel so that the most frequent words in each type of news can be compared, which helps in understanding the different linguistic patterns between reliable and unreliable news.







We remove the most frequent words from the texts in the no\_sw column of the dt data frame. The goal is to improve the quality of the texts by removing words that appear frequently and may not be useful for analysis, after previously removing the stopwords. A new column is created in the data frame called (wo\_stopfreq) that contains the texts after removing the most frequent words.

We load the WordNet database and it uses a lemmatization tool to trace the words in the wo\_stopfreq column back to their roots, then saves the results in a new column wo\_stopfreq\_lem.

wo_stopfreq	wo_stopfreq_lem
house dem aide	house dem aide didnt
didnt even see	even see comeys letter
comeys letter ja	ja
ever get feeling life	ever get feeling life
circle roundabout	circle roundabout
rather	rather
truth might get fired	truth might get fired
october 29 2016	october 29 2016
tension	tension
video 15 civilian killed single airstrike iden	video 15 civilian killed single airstrike iden





The text in wo\_stopfreq\_lem is split into distinct words (tokens) using word\_tokenize.

These lyrics are saved in a new column called tokenized text.

A RegexpTokenizer object is created with a regular expression (regex) that is used to extract words that contain only letters and numbers (a-z, A-Z, and 0-9). The regular expression r'[a-zA-Z0-9]+' means to search for a sequence of only letters or numbers.

fit\_transform: This function learns from texts to convert each text into a numerical representation using Bag of Words, where each word represents a column in an array, and each row represents a document or sentence. The value in the cell is the number of times the word appears in the text.

The result is a sparse array (an array containing many zero values) called text\_counts, which represents the frequency of words in the texts.





0.79

weighted avg

0.78

0.78

4160

We train a Complement Naive Bayes model to classify the data, and then evaluate the model using: Accuracy to measure the percentage of correct predictions.

A Confusion Matrix to illustrate the errors the model makes for each class.

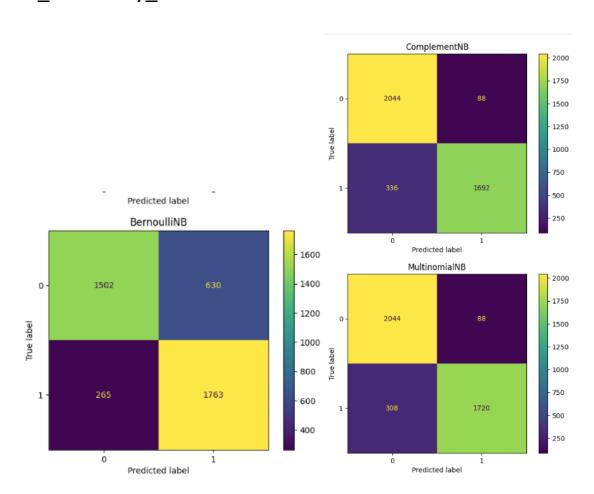
A Classification Report that contains precision, recall, and F1 score for each class, and was also applied to MultinomialNB() and BernoulliNB()

Confusion Matrix:					MultinominalNB model accuracy is 90.48%						
0 0 2044 1 336	1692					Confusion 0 0 2044 1 308	n Mat 1 88	rix:			
Classific		recision	recall :	F1-score	sunnort	61					
	,	51 60131011	recarr	1 30016	suppor c	Classiti	catio	n Report:		64	
	0	0.86	0.96	0.91	2132			precision	recall	f1-score	support
	1		0.83		2028		a	0.87	0.96	0.91	2132
	_								0.85		2028
accur	acv			0.90	4160		-	0.55	0.05	0.50	2020
	•	0.90	0.90		4160	accui	racv			0.90	4160
	_	0.90			4160			0.91	0.90		
							_	0.91			
Confusio 0 0 1502	 n Mat 1		acy = 78.4								
Classifi	catio	n Report:									
		precision	recall	f1-scor	e supp	ort					
	0	0.85	0.70	0.7	7 2	132					
	1	0.74	0.87	0.8	30 2	028					
accii	racy			0.7	'8 4	160					
accu											





It displays the confusion matrix for several models including k = [CNB, MNB, BNB] and each confusion matrix is generated based on the expected results of the model versus the actual results using the test data X\_test and y\_test.







ROC curves are calculated and displayed for three Naive Bayes models (CNB, MNB, BNB) and compare these models with the results of random predictions. The ROC curve measures the performance of models across a range of thresholds, measuring the rate of true positives versus the rate of false positives.

