Fake News detection

1. Introduction

Fake news is a treat to the stability of the society as a whole. Currently, it has become a menace and had the capacity to sway elections, like the case of Donald Trump

2. Problem Definition

People often share information without verifying it. This can include political agendas or to sway opinions. Media outlets are fighting fake news so that it does not impact their credibility and revenue. Hence, it is important to detect fake news as it impacts the society as a whole.

3. Experimental Evaluation

3.1 Methodology

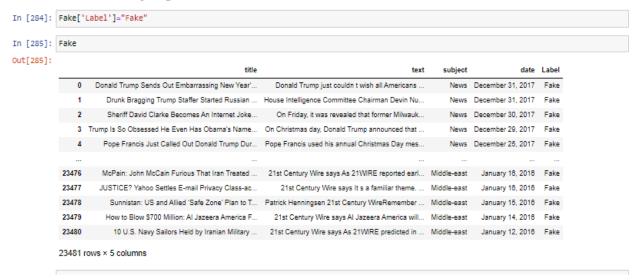
- Data cleaning and removing the unwanted columns from the data
- Combining the true and Fake into a single Label column and concating the 2 data sets with pandas and numpy
- Checking the length and count of the messages
- Used label encoding to add values to spam and ham messages
- Used stop words and word net lemmatizer to clean the data
- Used word cloud to analyse the most common words used in fake and True categories in label column
- Convert text into vectors using TF-IDF
- Instantiate MultinomiaLNB classifier
- Split feature and Label
- Train and split on the data set
- Train and predict
- Plot confusion matrix heatmap

3.2 Results

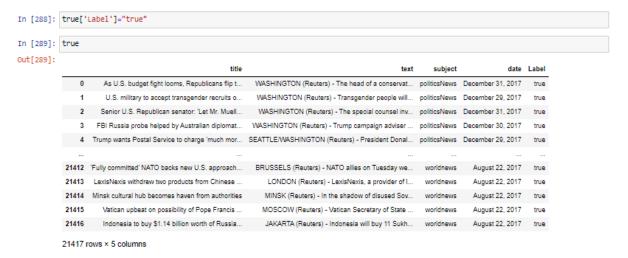
1. Found out the shape and size of the data



2. Added a Fake category under label to the Fake.csv data set



3. Added a true label to the true.csv data set



4. Used concat function to combine two data sets. And two labels: True and Fake

21417 rows × 5 columns

```
In [290]: #concatenation
               df_cat1 = pd.concat([Fake, true], axis=0, ignore_index=True)
               print (df_cat1)
                           Donald Trump Sends Out Embarrassing New Year'...
                           Drunk Bragging Trump Staffer Started Russian ...
Sheriff David Clarke Becomes An Internet Joke...
                            Trump Is So Obsessed He Even Has Obama's Name...
                           Pope Francis Just Called Out Donald Trump Dur...
                44893 'Fully committed' NATO backs new U.S. approach...
                44894 LexisNexis withdrew two products from Chinese ...
                44895 Minsk cultural hub becomes haven from authorities
               44896 Vatican upbeat on possibility of Pope Francis ... 44897 Indonesia to buy $1.14 billion worth of Russia...
                                                                                                           subject \
                          Donald Trump just couldn t wish all Americans ...
                                                                                                                News
                          House Intelligence Committee Chairman Devin Nu...
                          On Friday, it was revealed that former Milwauk...
On Christmas day, Donald Trump announced that ...
Pope Francis used his annual Christmas Day mes...
                                                                                                                News
                                                                                                                News
                                                                                                                News
               ... 44893 BRUSSELS (Reuters) - NATO allies on Tuesday we... worldnews
44894 LONDON (Reuters) - LexisNexis, a provider of l... worldnews
44895 MINSK (Reuters) - In the shadow of disused Sov... worldnews
44896 MOSCOW (Reuters) - Vatican Secretary of State ... worldnews
44897 JAKARTA (Reuters) - Indonesia will buy 11 Sukh... worldnews
                          date Label
December 31, 2017 Fake
                          December 31, 2017 Fake
                          December 30, 2017 Fake
                          December 29, 2017 Fake
December 25, 2017 Fake
                44893 August 22, 2017 true
                         August 22, 2017
August 22, 2017
August 22, 2017
                44894
                                                       true
                44895
                44896 August 22, 2017
```

5. Used label encoder to add 0 and 1 values to label column, (0 being Fake and 1 is true)

Out[292]:

	title	text	subject	date	Label
0	Donald Trump Sends Out Embarrassing New Year'	Donald Trump just couldn t wish all Americans	News	December 31, 2017	0
1	Drunk Bragging Trump Staffer Started Russian	House Intelligence Committee Chairman Devin Nu	News	December 31, 2017	0
2	Sheriff David Clarke Becomes An Internet Joke	On Friday, it was revealed that former Milwauk	News	December 30, 2017	0
3	Trump Is So Obsessed He Even Has Obama's Name	On Christmas day, Donald Trump announced that	News	December 29, 2017	0
4	Pope Francis Just Called Out Donald Trump Dur	Pope Francis used his annual Christmas Day mes	News	December 25, 2017	0
4893	'Fully committed' NATO backs new U.S. approach	BRUSSELS (Reuters) - NATO allies on Tuesday we	worldnews	August 22, 2017	1
4894	LexisNexis withdrew two products from Chinese	LONDON (Reuters) - LexisNexis, a provider of I	worldnews	August 22, 2017	1
4895	Minsk cultural hub becomes haven from authorities	MINSK (Reuters) - In the shadow of disused Sov	worldnews	August 22, 2017	1
4896	Vatican upbeat on possibility of Pope Francis	MOSCOW (Reuters) - Vatican Secretary of State	worldnews	August 22, 2017	1
4897	Indonesia to buy \$1.14 billion worth of Russia	JAKARTA (Reuters) - Indonesia will buy 11 Sukh	worldnews	August 22, 2017	1

44898 rows × 5 columns

6. Finding out any null values

7. Finding out if there are any special characters in the object data type

8.I have dropped the date column as it is a challenge to deal with and there are many special character and unique values to calculate

```
In [298]: # dropping the date column
df_cat1.drop('date',axis=1,inplace=True)

In [299]: df_cat1.shape

Out[299]: (44898, 4)
```

9.replaced any missing values with blank values

```
Subject TT000
                                         politicatvews 11212
In [301]: df_cat1 = df_cat1.fillna(' ')
In [302]: df_cat1.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 44898 entries, 0 to 44897
         Data columns (total 4 columns):
          # Column Non-Null Count Dtype
                     -----
                     44898 non-null object
          a title
                     44898 non-null object
             text
          2 subject 44898 non-null object
          3 Label
                      44898 non-null int32
         dtypes: int32(1), object(3)
         memory usage: 1.2+ MB
```

10. Counts in the label column

11. Finding the value counts in text

```
...... ...., ..,,...
In [304]: df_cat1['text'][5]
Out[304]: 'The number of cases of cops brutalizing and killing people of color seems to see no end. Now, we have another case that needs
                    to be shared far and wide. An Alabama woman by the name of Angela Williams shared a graphic photo of her son, lying in a hospit
                    al bed with a beaten and fractured face, on Facebook. It needs to be shared far and wide, because this is unacceptable.It is un clear why Williams son was in police custody or what sort of altercation resulted in his arrest, but when you see the photo yo u will realize that these details matter not. Cops are not supposed to beat and brutalize those in their custody. In the post y
                    ou are about to see, Ms. Williams expresses her hope that the cops had their body cameras on while they were beating her son, but I think we all know that there will be some kind of convenient malfunction to explain away the lack of existence of dash or body camera footage of what was clearly a brutal beating. Hell, it could even be described as attempted murder. Something tells me that this young man will never be the same. Without further ado, here is what Troy, Alabama s finest decided was appropri ate treatment of Angela Williams son:No matter what the perceived crime of this young man might be, this is completely unacceptable. The copy who did this capacity and the process who have they get a paid very time.
                   table. The cops who did this need to rot in jail for a long, long time but what you wanna bet they get a paid vacation while the force investigates itself, only to have the officers returned to duty posthaste? This, folks, is why we say BLACK LIVES MA TTER. No way in hell would this have happened if Angela williams son had been white. Please share far and wide, and stay tuned to Addicting Info for further updates. Featured image via David McNew/Stringer/Getty Images'
In [305]: df_cat1.value_counts()
                            according to an email seen by Reuters. Trump will seek to rebuild the U.S. relationship with Egypt at a meeting on Monda
                    y with Egyptian President Abdel Fattah al-Sisi focused on security issues and military aid, a senior White House official sa
ys. Trump will host Jordan's King Abdullah at the White House next week to discuss the fight against Islamic State militant
                          the Syria crisis and advancing peace between Israelis and Palestinians, the White House says. A U.S. judge approves a $25
                    million settlement to resolve a class action lawsuit that claimed fraud against Trump and his Trump University real estate s
                    eminars.
                    politicsNews 1
                     Highlights: The Trump presidency on April 26 at 9:12 P.M. EDT/0112 GMT on April 27
                    ers) - Highlights for U.S. President Donald Trump's administration on Wednesday: Trump proposes slashing tax rates for busin esses and on overseas corporate profits returned to the country in a plan greeted as an opening gambit by his fellow Republi
                    cans in Congress. Trump's plan could shift the U.S. economy into higher gear but could have one effect the White House would
                    not welcome — interest rates ratcheted higher than expected by a wary central bank. The Trump tax cut will generate growth b ut not nearly enough to replace trillions of dollars in lost revenues, while rising deficits could even take back some of th
                    e economic gains, fiscal experts say. Congress inches toward a deal to fund the government through September but is preparin
                   g to possibly extend a midnight Friday deadline in order to wrap up negotiations and avoid an imminent government shutdown. Trump is considering issuing an executive order to pull the United States from the North American Free Trade Agreement, an a dministration official says, a move that could unravel one of the world's biggest trading blocs. Trump and Canadian Prime Mi nister Justin Trudeau discuss bilateral trade in their second conversation in as many days amid strains over softwood lumbe
                    r and dairy. The Trump administration says it aims to push North Korea into dismantling its nuclear and missile programs thr
```

12. Checked for 209 duplicate value and removed them

```
In [306]: #checking for Duplicates and Dropping them
df_cat1.duplicated().sum() #there are 209 duplicate rows

Out[306]: 213

In [307]: #dropping duplicate rows
df_cat1.drop_duplicates(inplace=True)
df_cat1.duplicated().sum()

Out[307]: 0
```

13. Working with regular expressions and removing characters

```
In [309]: import re

In [310]: # expanding English Language contractions: https://stackoverfLow.com/a/47091490/4084039

def decontracted(phrase):
    # specific
    phrase = re.sub(r"won\'t", "will not", phrase)
    phrase = re.sub(r"can\'t", "can not", phrase)

# general
    phrase = re.sub(r"n\'t", " not", phrase)
    phrase = re.sub(r"\'re", " are", phrase)
    phrase = re.sub(r"\'re", " is", phrase)
    phrase = re.sub(r"\'1", " will", phrase)
    phrase = re.sub(r"\'1", " will", phrase)
    phrase = re.sub(r"\'t", " not", phrase)
    phrase = re.sub(r"\'t", " not", phrase)
    phrase = re.sub(r"\'re", " have", phrase)
    phrase = re.sub(r"\'re", " have", phrase)
    return phrase
```

14. Importing stopwords and word net lemmatizer from nltk

```
import nltk
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer

inltk.download('stopwords')

[nltk_data] Downloading package stopwords to
[nltk_data] C:\Users\Office\AppData\Roaming\nltk_data...
[nltk_data] Package stopwords is already up-to-date!

inltk_data] nltk.download('punkt')

[nltk_data] Downloading package punkt to
[nltk_data] C:\Users\Office\AppData\Roaming\nltk_data...
[nltk_data] Package punkt is already up-to-date!

inltk_data] Package punkt is already up-to-date!

stop_words=stopwords.words('english')

ilemmatizer=WordNetLemmatizer()
```

15. Import tqm library

```
In [316]: #https://stackoverflow.com/a/5843547/4084039
                #TitLe
                from tqdm import tqdm
                preprocessed_titles = []
#tqdm is for printing the status bar
                for sentence in tqdm(df_cat1['title'].values):
                       sent = decontracted(sentence)
                      sent = decontracted(sentence)
sent = re.sub(r'https?:\/\/.*[\r\n]*', '', sent) # remove hyperlinks
sent=re.sub('[^A-Za-z]+', ', sent) #remove spacial character, numbers:
sent = ' '.join(e for e in sent.split() if e not in stop_words) #removing stop words
sent=' '.join(lemmatizer.lemmatize(e) for e in sent.split()) # Lemmatization
                       preprocessed_titles.append(sent.lower().strip())
                                                                                                                                 44685/44685 [00:15<00:00, 2812.51it/s]
In [317]: df_cat1['titles']=preprocessed_titles
In [318]: df cat1['titles']
                              donald trump sends out embarrassing new year e...
drunk bragging trump staffer started russian c...
sheriff david clarke becomes an internet joke ...
Out[318]: 0
                               trump is so obsessed he even has obama name co...
                              pope francis just called out donald trump duri...
                4
                           fully committed nato back new u s approach afg...
lexisnexis withdrew two product chinese market
                44893
                44894
                             minsk cultural hub becomes authority
vatican upbeat possibility pope francis visiti...
indonesia buy billion worth russian jet
                44895
                 44896
                44897
                Name: titles, Length: 44685, dtype: object
```

16. Working with title column

```
In [317]: df_cat1['titles']=preprocessed_titles
In [318]: df_cat1['titles']
Out[318]: 0
                             donald trump sends out embarrassing new year e... drunk bragging trump staffer started russian c...
                             sheriff david clarke becomes an internet joke ...
                             trump is so obsessed he even has obama name co...
                            pope francis just called out donald trump duri...
                44893
                           fully committed nato back new u s approach afg...
                44894
                              lexisnexis withdrew two product chinese market
                44895
                                                minsk cultural hub becomes authority
                           vatican upbeat possibility pope francis visiti...
indonesia buy billion worth russian jet
                44896
                44897
                Name: titles, Length: 44685, dtype: object
In [319]: #TEXT
                preprocessed_texts = []
                #tqdm is for printing the status bar
for sentence in tqdm (df_cat1['text'].values):
                     sent = decontracted(sentence)
sent = re.sub(r'https?:\\/.*[r\n]*', '', sent) # remove hyperlinks
sent=re.sub('[^A-Za-Z]+',' ', sent) #remove spacial character, numbers:
sent = ' '.join(e for e in sent.split() if e not in stop_words) #removing stop words
sent=' '.join(lemmatizer.lemmatize(e) for e in sent.split()) # Lemmatization
preprocessed_texts.append(sent.lower().strip())
                                                                                                     44685/44685 [03:40<00:00, 202.89it/s]
```

17. Working with text column

```
In [319]: #TEXT
                  preprocessed_texts = []
                  #tqdm is for printing the status bar
for sentence in tqdm (df_cat1['text'].values):
                         sent = decontracted(sentence)
                        sent = decontracted(sentence)
sent = re.sub(r'https?:\/\/.*[\r\n]*', '', sent) # remove hyperlinks
sent=re.sub('[^A-Za-z]+',' ', sent) #remove spacial character, numbers:
sent = ' '.join(e for e in sent.split() if e not in stop_words) #removing stop words
sent=' '.join(lemmatizer.lemmatize(e) for e in sent.split()) # Lemmatization
preprocessed_texts.append(sent.lower().strip())
                                                                                                                                            44685/44685 [03:40<00:00, 202.89it/s]
In [320]: df_cat1['text']= preprocessed_texts
In [321]: df_cat1['text']
Out[321]: 0
                                 donald trump wish americans happy new year lea...
                                 house intelligence committee chairman devin nu...
                                on friday revealed former milwaukee sheriff da...
on christmas day donald trump announced would ...
pope francis used annual christmas day message...
                 44893 brussels reuters nato ally tuesday welcomed pr...
44894 london reuters lexisnexis provider legal regul...
                  44895
                                 minsk reuters in shadow disused soviet era fac...
                                moscow reuters vatican secretary state cardina...
jakarta reuters indonesia buy sukhoi fighter j...
                  44896
                  44897
                  Name: text, Length: 44685, dtype: object
```

18. Processing for subject column

```
[223]: #subject
          preprocessed_subject = []
           #tqdm is for printing the status bar
           for sentence in tqdm (df_cat1['subject'].values):
                sent = decontracted(sentence)
sent = re.sub(r'https?:\/\.*[\r\n]*', '', sent) # remove hyperlinks
sent=re.sub('[^A-Za-z]+',' ', sent) #remove spacial character, numbers:
                sent = ' '.join(e for e in sent.split() if e not in stop_words) #removing stop words
sent=' '.join(lemmatizer.lemmatize(e) for e in sent.split()) # Lemmatization
preprocessed_subject.append(sent.lower().strip())
                                                                                  44685/44685 [00:02<00:00, 15252.64it/s]
 [224]: df_cat1['subject']= preprocessed_subject
 [225]: df_cat1['subject']
t[225]: 0
                             news
                             news
                             news
          44893
                      worldnews
          44894
                      worldnews
           44895
                      worldnews
           44896
                      worldnews
          44897
                      worldnews
          Name: subject, Length: 44685, dtype: object
```

19. Splitting the data into train and test

```
In [226]: #Splitting data into train, cv, test
from sklearn.model_selection import train_test_split
           y=df cat1["Label"]
           x = df_cat1.drop(columns={'Label'})
 stratify=y,
                                                                  random state=0)
           train_modified, cv, train_output_modified, cv_output = train_test_split(train, train_output,
                                                                                 test_size=0.25,
                                                                                stratify=train_output,
                                                                                random_state=0)
 In [228]: train.shape, cv.shape, test.shape
 Out[228]: ((33513, 4), (8379, 4), (11172, 4))
 In [229]: train_output.shape, cv_output.shape, test_output.shape
 Out[229]: ((33513,), (8379,), (11172,))
20. Using TDF vectorization
#data encoding
#title-TFIDF Vectorization
     from sklearn.feature extraction.text import TfidfVectorizer
     title_tfidf_vectorizer = TfidfVectorizer(min_df=5)
h1]: train_title_tfidf=title_tfidf_vectorizer.fit_transform(train['title'].values)
32]: cv_title_tfidf=title_tfidf_vectorizer.transform(cv['title'].values)
     test_title_tfidf=title_tfidf_vectorizer.transform(test['title'].values)
33]: #saving the tfidf vectorizer
     import pickle
     with open("title_tfidf_vectorizer.pickle", "wb") as fp:
        pickle.dump(title_tfidf_vectorizer, fp, protocol=pickle.HIGHEST_PROTOCOL)
        title_tfidf_vectorizer.get_feature_names() [:10]
     C:\Users\Office\anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87: FutureWarning: Function get_feature_names is depre
     cated; get_feature_names is deprecated in 1.0 and will be removed in 1.2. Please use get_feature_names_out instead.
      warnings.warn(msg, category=FutureWarning)
14]: title_tfidf_vectorizer.get_feature_names()[:10]
```

21. Data encoding for text column with TDF

['000', '04', '10', '100', '100k', '101', '11', '12', '120', '13']

```
: #data encoding
#title-TFIDF Vectorization
   from sklearn.feature_extraction.text import TfidfVectorizer
  text_tfidf_vectorizer = TfidfVectorizer(min_df=10)
: train_text_tfidf=text_tfidf_vectorizer.fit_transform(train['text'].values)
: cv_text_tfidf=text_tfidf_vectorizer.transform(cv['text'].values)
test_text_tfidf=text_tfidf_vectorizer.transform(test['text'].values)
: #saving the tfidf vectorizer
  import pickle
  with open("text_tfidf_vectorizer.pickle", "wb") as fp:
pickle.dump(text_tfidf_vectorizer, fp, protocol=pickle.HIGHEST_PROTOCOL)
       text_tfidf_vectorizer.get_feature_names() [:10]
: text_tfidf_vectorizer.get_feature_names()[:10]
: ['00', '000', '0000', '001', '002', '003', '005', '005380', '00pm', '01']
22. Using TDF use multinomial classifier. The accuracy is 93%
5]:
    #1. Convert text into vectors using TF-IDF
    #2. Instantiate MultinomiaLNB classifier
    #3. Split feature and Label
    from sklearn.feature extraction.text import TfidfVectorizer
    from sklearn.naive_bayes import MultinomialNB
     from sklearn.model_selection import train_test_split
    from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
    tf_vec = TfidfVectorizer()
    naive=MultinomialNB()
    features=tf_vec.fit_transform(df_cat1['text'])
    x = features
    y = df_cat1['Label']
5]: # Train and predict
    x_train,x_test, y_train,y_test=train_test_split(x,y,random_state=42)
naive.fit(x_train,y_train)
    y_pred=naive.predict(x_test)
print('Final score => ', accuracy_score(y_test,y_pred))
     Final score = > 0.9327219901942981
7]: print (classification_report(y_test, y_pred))
                   precision recall f1-score support
                а
                         0.93
                                  0.94
                                             0.94
                                                         5792
                1
                         0.94
                                  0.92
                                              0.93
                                                         5312
        accuracy
                                              0.93
                                                       11014
                         0.93
                                   0.93
                                                        11014
        macro avg
                                              0.93
    weighted avg
                         0.93
                                   0.93
                                              0.93
                                                       11014
```

Conclusion:

1. Finding out the unique value in object type column

df_cat1.describe(include=['object','datetime']).transpose()

count unique top freq

title 44898 38729 Factbox: Trump fills top jobs for his administ... 14

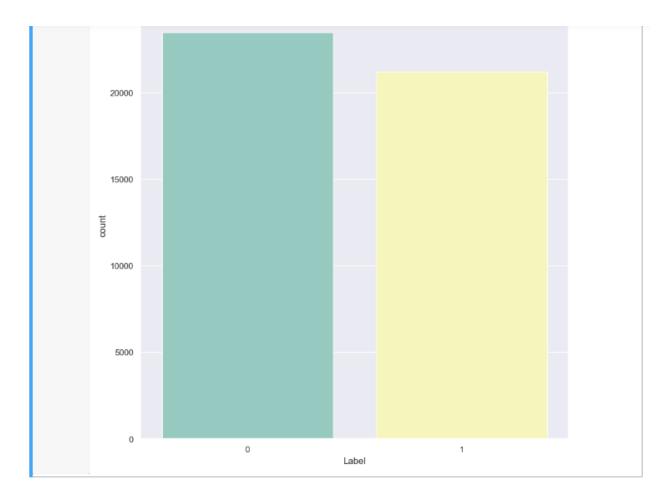
text 44898 38646 627

subject 44898 8 politicsNews 11272

There are 38729 unique values and they occur in frequency of 14 for title $\,$

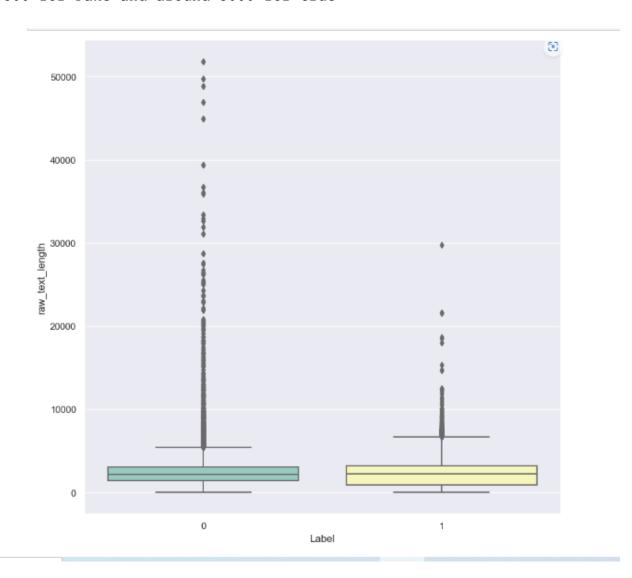
There are v

2. Dividing labels into 0 and 1. (0 being Fake and 1 is true). Around 2800 for Fake and 2300 for true



3. Raw text length for both labels 0 and 1 . (0 being Fake and 1 is true)

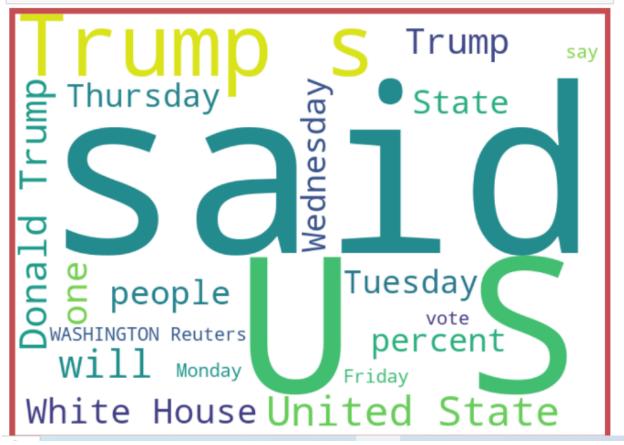
5500 for Fake and around 3000 for true



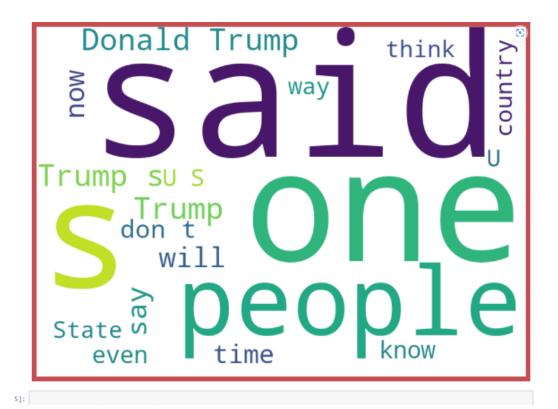
4. Using word cloud, some common word cloud words for true label ie 1 include:

Trump, said, Thursday, White House, United states, Vote, Donald Trump, State

```
#Getting sense of word cloud in spam
from wordcloud import Wordcloud
spams=df_cat1['text'][df_cat1['label']==1]
spam_cloud = Wordcloud(width=700,height=500, background_color='white', max_words=20).generate(' '.join(spams))
plt.figure(figsize=(10,8), facecolor='r')
plt.imshow(spam_cloud)
plt.axis('off')
plt.tight_layout(pad=0)
plt.show()
```



5. Using word cloud, some common word cloud words for Fake label ie 0 include: Said, One, people Trump, Donald Trump, dont, will, state, say, time, know



6. Accuracy with multinomial classifier. Accuracy is 93%

```
5]:
    #1. Convert text into vectors using TF-IDF
#2. Instantiate MultinomiaLNB classifier
    #3. Split feature and Label
    from sklearn.feature_extraction.text import TfidfVectorizer
    from sklearn.naive_bayes import MultinomialNB
    from sklearn.model_selection import train_test_split
    from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
    tf_vec = TfidfVectorizer()
    naive=MultinomialNB()
    features=tf_vec.fit_transform(df_cat1['text'])
    x = features
    y = df_cat1['Label']
5]: # Train and predict
    x_train,x_test, y_train,y_test=train_test_split(x,y,random_state=42)
naive.fit(x_train,y_train)
    y_pred=naive.predict(x_test)
print('Final score => ', accuracy_score(y_test,y_pred))
    Final score = > 0.9327219901942981
7]: print (classification_report(y_test, y_pred))
                   precision
                                recall f1-score support
                         0.93
                                   0.94
                                               0.94
                                                          5702
                0
                1
                         0.94
                                   0.92
                                              0.93
                                                          5312
        accuracy
                                               0.93
                                                        11014
                         0.93
                                   0.93
                                                         11014
       macro avg
                                               0.93
    weighted avg
                                              0.93
                                                         11014
                         0.93
                                    0.93
81. R alak and Contac and St. banks
```

7. Using the correlation matrix, plotting the True and predicted values

