

Fake News
Detection
using NLP

Problem Statement:

Begin building the fake news detection model by loading and preprocessing the dataset. Load the fake news dataset and preprocess the textual data.

Data Cleaning:



Data cleaning is a process of removing inconsistencies in the dataset and incorrect values .It also in involves handling missing values either by removing them or assigning them average values. It helps to improve the efficiency of the model. In the first step, we will only remove the unnecessary data points from the dataset which does not helps in improving the model performance.

Initially we import the necessary packages for our data cleaning process and also in the future purposes,

```
[] import numpy as np
  import pandas as pd
  import matplotlib.pyplot as plt
  import seaborn as sns
  import nltk
  import re
  from wordcloud import WordCloud
  from tensorflow.keras.preprocessing.text import Tokenizer
  from tensorflow.keras.preprocessing.sequence import pad_sequences
  from tensorflow.keras.models import Sequential
  from tensorflow.keras.layers import Dense, Embedding, LSTM, Conv1D, MaxPool1D
  from sklearn.model_selection import train_test_split
  from sklearn.metrics import classification_report, accuracy_score
  import numpy as np
  import pandas as pd
```

we use these packages in various stages of our cleaning process and also in the future in which we need to build models.

Here, we read the .csv files of true and fake news and then explore the count values of their subjects

```
import os
for dirname, _, filenames in os.walk('/content/drive/MyDrive/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

/content/drive/MyDrive/input/True.csv
/content/drive/MyDrive/input/Fake.csv
```

```
[ ] fake_news = pd.read_csv('/content/drive/MyDrive/input/Fake.csv')
    fake_news.head()
```

	title	text	subject	date
0	Donald Trump Sends Out Embarrassing New Year'	Donald Trump just couldn t wish all Americans	News	December 31, 2017
1	Drunk Bragging Trump Staffer Started Russian	House Intelligence Committee Chairman Devin Nu	News	December 31, 2017
2	Sheriff David Clarke Becomes An Internet Joke	On Friday, it was revealed that former Milwauk	News	December 30, 2017
3	Trump Is So Obsessed He Even Has Obama's Name	On Christmas day, Donald Trump announced that	News	December 29, 2017
4	Pope Francis Just Called Out Donald Trump Dur	Pope Francis used his annual Christmas Day mes	News	December 25, 2017

```
[ ] fake_news.columns
```

Index(['title', 'text', 'subject', 'date'], dtype='object')

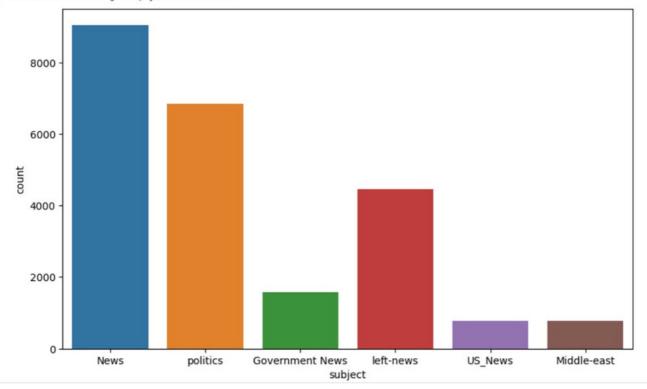
[] fake_news['subject'].value_counts()

News 9050 politics 6841 left-news 4459 Government News 1570 US News 783 Middle-east 778

Name: subject, dtype: int64

plt.figure(figsize=(10,6)) sns.countplot(x='subject',data=fake_news)

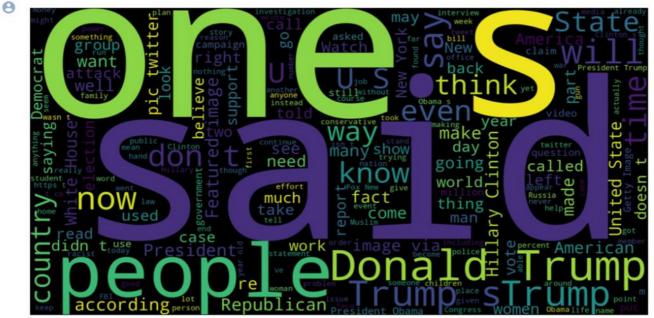




Here, we have used wordcloud to see that which word has mostly used for the fake news. By seeing that we can make a conclusion that which topic(about a person, event or anything) is mostly contains fake news). We also do the same for true news.

Word Cloud for Fake News:

```
%*time
wordcloud = WordCloud(width=1920, height=1000).generate(text)
fig = plt.figure(figsize=(10,10))
plt.imshow(wordcloud)
plt.axis('off')
plt.tight_layout(pad=0)
plt.show()
```

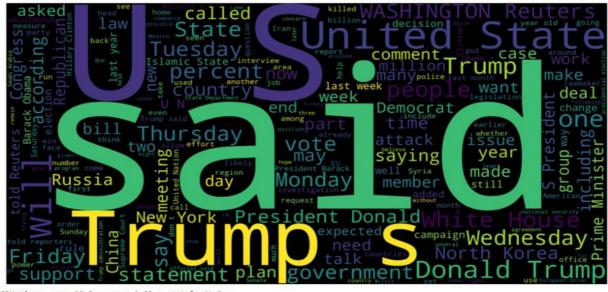


CPU times: user 43 s, sys: 3.33 s, total: 46.4 s Wall time: 50.4 s

Word cloud for True News:

real_news = pd.read_csv('/content/drive/MyDrive/input/True.csv')
real_news.head()

	title	text	subject	date
0	As U.S. budget fight looms, Republicans flip t	WASHINGTON (Reuters) - The head of a conservat	politicsNews	December 31, 2017
1	U.S. military to accept transgender recruits o	WASHINGTON (Reuters) - Transgender people will	politicsNews	December 29, 2017
2	Senior U.S. Republican senator: 'Let Mr. Muell	WASHINGTON (Reuters) - The special counsel inv	politicsNews	December 31, 2017
3	FBI Russia probe helped by Australian diplomat	WASHINGTON (Reuters) - Trump campaign adviser	politicsNews	December 30, 2017
4	Trump wants Postal Service to charge 'much mor	SEATTLE/WASHINGTON (Reuters) - President Donal	politicsNews	December 29, 2017



CPU times: user 38.5 s, sys: 2.83 s, total: 41.3 s Wall time: 41.4 s

[] real_news.sample(5)

	title	text	subject	date
5933	Peru and Colombia vow to stand with Mexico aft	LIMA (Reuters) - Peru and Colombia vowed to st	politicsNews	January 27, 2017
15390	North Korean embassy official in focus at $\operatorname{Kim} \dots$	$\ensuremath{KUALA}\xspace \ensuremath{LUMPUR}\xspace$ (Reuters) - Three men wanted for	worldnews	November 8, 2017
6088	Trump's exit from Pacific trade deal opens doo	BERLIN (Reuters) - Germany would take advantag	politicsNews	January 23, 2017
8915	Albanian town backs Clinton with bronze bust	SARANDE, Albania (Reuters) - Whatever the outc	politicsNews	June 30, 2016
1319	House Republicans to take up disaster funding	WASHINGTON (Reuters) - U.S. House of Represent	politicsNews	October 11, 2017

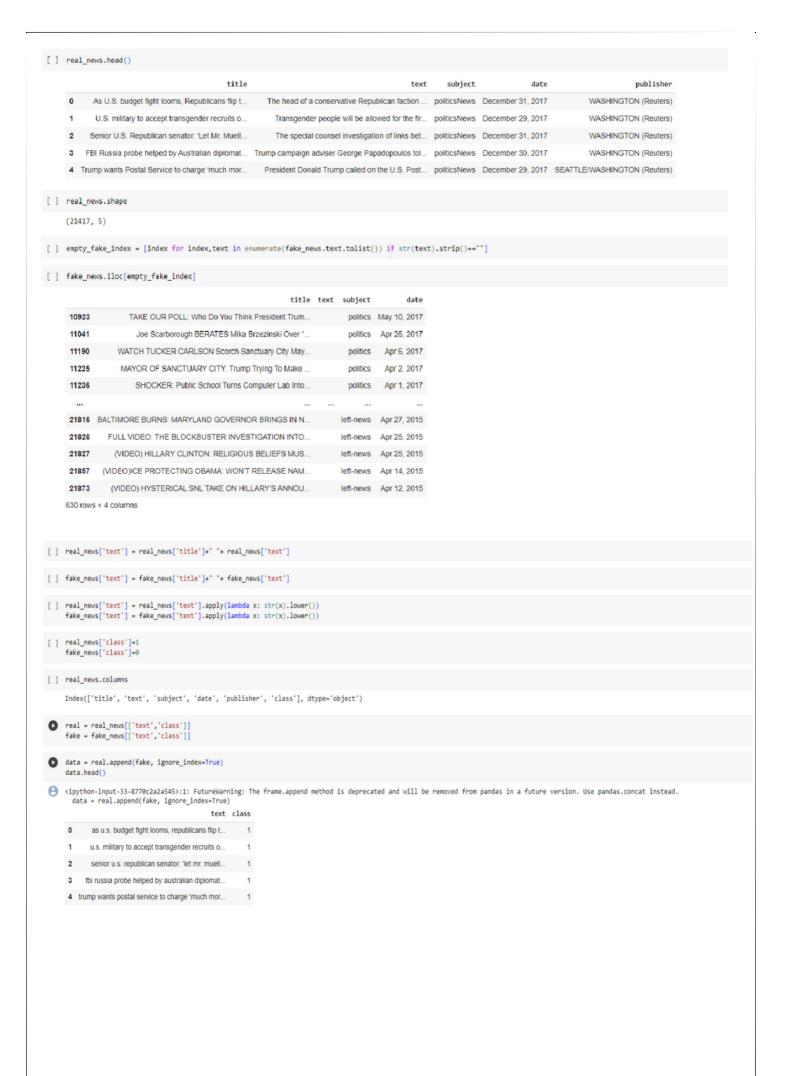
Let's create a list of news lists in real_news.csv with unknown publishers by using the following code snippets

```
[ ] unknown_pubishers = []
  for index, row in enumerate(real_news.text.values):
        try:
        record = row.split(' - ', maxsplit=1)
        record[1]

        assert(len(record[0])<260)
        except:
        unknown_pubishers.append(index)</pre>
```

```
[ ] len(unknown_pubishers)
     35
     real news.iloc[unknown pubishers].text
     2922
              The following statements were posted to the ve...
     3488
              The White House on Wednesday disclosed a group...
     3782
              The following statements were posted to the ve...
    4358
              Neil Gorsuch, President Donald Trump's appoint...
              WASHINGTON The clock began running out this we...
    4465
     5290
              The following statements were posted to the ve...
     5379
              The following statements were posted to the ve...
     5412
              The following statements were posted to the ve...
     5504
              The following statements were posted to the ve...
     5538
              The following statements were posted to the ve...
     5588
              The following statements were posted to the ve...
     5593
              The following statements were posted to the ve...
     5761
              The following bullet points are from the U.S. ...
     5784
              Federal appeals court judge Neil Gorsuch, the ...
              The following bullet points are from the U.S. ...
     6026
     6184
              The following bullet points are from the U.S. ...
     6669
              Republican members of Congress are complaining...
              Over the course of the U.S. presidential campa...
     6823
     7922
              After going through a week reminiscent of Napo...
    8194
              The following timeline charts the origin and s...
              Global health officials are racing to better u...
     8195
     8247
              U.S. President Barack Obama visited a street m...
     8465
              ALGONAC, MICH.-Parker Fox drifted out of the D...
     8481
              Global health officials are racing to better u...
     8482
              The following timeline charts the origin and s...
     8505
              Global health officials are racing to better u...
              The following timeline charts the origin and s...
     8506
     8771
              In a speech weighted with America's complicate...
     8970
     9008
              The following timeline charts the origin and s...
              Global health officials are racing to better u...
     9009
     9307
              It's the near future, and North Korea's regime...
              GOP leaders have unleashed a stunning level of...
     9618
    9737
              Caitlyn Jenner posted a video on Wednesday (Ap...
     10479
              The Democratic and Republican nominees for the...
    Name: text, dtype: object
    publisher =[]
     tmp_text = []
     for index, row in enumerate(real news.text.values):
         if index in unknown pubishers:
            tmp_text.append(row)
            publisher.append('Unknown')
         else:
            record = row.split('-', maxsplit=1)
            publisher.append(record[0].strip())
            tmp_text.append(record[1].strip())
 [ ] real_news['publisher'] = publisher
     real_news['text'] = tmp_text
```

[] real_news.head()



```
!pip install spacy==2.2.3
            !python -m spacy download en_core_web_sm
            !pip install beautifulsoup4==4.9.1!
            !pip install textblob==0.15.3
            !pip install git+https://github.com/laxmimerit/preprocess_kgptalkie.git --upgrade --force-reinstall

    note: This error originates from a subprocess, and is likely not a problem with pip.
2023-10-16 16:41:23.058566: W tensorflow/compiler/tf2tensorrt/utils/py_utils.cc:38] TF-TRT Warning: Could not find TensorRT
           Collecting en-core-web-sm==3.6.0

Downloading <a href="https://github.com/explosion/spacy-models/releases/download/en_core_web_sm=3.6.0/en_core_web_sm=3.6.0-py3-none-any.whl">https://github.com/explosion/spacy-models/releases/download/en_core_web_sm=3.6.0/en_core_web_sm=3.6.0-py3-none-any.whl</a> (12.8 MB)
                                                                                                 --- 12.8/12.8 MB 21.6 MB/s eta 0:00:00
           Requirement already satisfied: spacy<3.7.0,>=3.6.0 in /usr/local/lib/python3.10/dist-packages (from en-core-web-sm==3.6.0) (3.6.1)
Requirement already satisfied: spacy-legacy<3.1.0,>=3.6.0 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (3.0.12)
           Requirement already satisfied: spacy-loggers<2.0.0,>=1.0.0 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (1.0.5) Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (1.0.10)
           Requirement already satisfied: cymem(2.1.e,>=2.0.2 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.0.8) Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (3.0.9) Requirement already satisfied: thinc<8.2.0,>=8.1.8 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (8.1.12)
           Requirement already satisfied: wasabi<1.2.0,>=0.9.1 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (1.1.2) Requirement already satisfied: srsly<3.0.0,>=2.4.3 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.4.8)
           Requirement already satisfied: catalogue(2.1.0,>=2.0.6 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.0.10)
Requirement already satisfied: typer(0.10.0,>=0.3.0 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (0.9.0)
           Requirement already satisfied: pathy>=0.10.0 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (0.10.2)
           Requirement already satisfied: smart-open(7.0.0,>=5.2.1 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (6.4.0) Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (4.66.1)
           Requirement already satisfied: numpy=1.15.0 in /usr/local/lib/python3.10/dist-packages (from spacy(3.7.0,>3.6.0-ben-core-web-sm=3.6.0) (%.00.1)

Requirement already satisfied: requests(3.0.0,>2.13.0 in /usr/local/lib/python3.10/dist-packages (from spacy(3.7.0,>3.6.0-ben-core-web-sm=3.6.0) (2.31.0)

Requirement already satisfied: requests(3.0.0,>2.13.0 in /usr/local/lib/python3.10/dist-packages (from spacy(3.7.0,>3.6.0-ben-core-web-sm=3.6.0) (2.31.0)

Requirement already satisfied: judaritie=1.8,!=1.8.1;(3.0.0,>1.7.4 in /usr/local/lib/python3.10/dist-packages (from spacy(3.7.0,>3.6.0-ben-core-web-sm=3.6.0) (1.10.13)

Requirement already satisfied: judaritie=1.0,!=1.8.1;(3.0.0,)=1.7.4 in /usr/local/lib/python3.10/dist-packages (from spacy(3.7.0,>3.6.0-ben-core-web-sm=3.6.0) (6.7.2)

Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-packages (from spacy(3.7.0,>3.6.0-ben-core-web-sm=3.6.0) (6.7.2)
           Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-packages (from spacys3.7.0, >=3.6.9-en-core-web-sm==3.6.0) (3.7.2)

Requirement already satisfied: langcodes<a.0.0, >=3.2.0 in /usr/local/lib/python3.10/dist-packages (from spacys3.7.0, >=3.6.9-en-core-web-sm==3.6.0) (3.3.0)

Requirement already satisfied: typing-extensions>=4.2.0 in /usr/local/lib/python3.10/dist-packages (from pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4.>spacys3.7.0, >=3.6.9-en-core-web-sm==3.6.0) (4.5.0)

Requirement already satisfied: charset-normalizer<a.yes=2.7.0 in /usr/local/lib/python3.10/dist-packages (from requests<3.0.0, >=2.13.0.>spacys3.7.0, >=3.6.0-en-core-web-sm==3.6.0) (4.5.0)

Requirement already satisfied: charset-normalizer<a.yes=2.7.0 in /usr/local/lib/python3.10/dist-packages (from requests<3.0.0, >=2.13.0.>spacys3.7.0, >=3.6.0-en-core-web-sm==3.6.0) (3.3.0)

Requirement already satisfied: idna<a.yes=2.5.1.0 in /usr/local/lib/python3.10/dist-packages (from requests<3.0.0, >=2.13.0.>spacys3.7.0, >=3.6.0-en-core-web-sm==3.6.0) (3.3.0)
           Requirement already satisfied: unlibis(3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests(3,0.0,>=2.13.0-spacy(3.7.0,>=3.6.0-en-core-web-sm=-3.6.0) (2.0.6) Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests(3,0.0,>=2.13.0-spacy(3.7.0,>=3.6.0->en-core-web-sm=-3.6.0) (2.0.6) Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from thinc(0.2.0,>=3.18.0-spacy(3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.0.6) Requirement already satisfied: blis(0.8.0,>=0.7.8 in /usr/local/lib/python3.10/dist-packages (from thinc(0.2.0,>=3.18.0-spacy(3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (0.1.3) Requirement already satisfied: confection(1.0.0,>=0.11 in /usr/local/lib/python3.10/dist-packages (from thinc(0.2.0,>=0.3.0-spacy(3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (0.1.3) Requirement already satisfied: click(9.0.0,>=7.1.1 in /usr/local/lib/python3.10/dist-packages (from typer(0.10.0,>=0.3.0-spacy(3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (0.1.3)
           Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2->spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.1.3)
 [ ] import preprocess_kgptalkie as ps
 [ ] data['text'] = data['text'].apply(lambda x: ps.remove_special_chars(x))
[ ] from google.colab import drive drive.mount('/content/drive')
         Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True)
 import gensim
[ ] y = data['class'].values
[ ] X = [d.split() for d in data['text'].tolist()]
[ ] type(X[0])
         ['as', 'us', 'budget', 'fight', 'looms', 'republicans', 'flip', 'their', 'fiscal', 'script', 'the', 'head', 'of', 'a', 'conservative', 'republican', 'faction', 'in', 'the', 'us', 'congress', 'who', 'voted', 'this', 'month', 'for', 'a', 'h
         4
              w2v_model = gensim.models.Word2Vec(sentences=X, vector_size=DIM, window =10, min_count=1)
[ ] w2v model.wv['india']
               array([-1.717186 , 0.09980671, -0.54519814, 2.873516 , 1.1529748 ,
                                -1.612415
                                                               -0.4013639 ,
1.0115958 ,
                                                                                                 1.5510874 , -1.9637966 , -1.7413545 , 2.1367438 ,
                                                                                                                                                                   1.7516707
                                  2.3293521 ,
                                                                                                                                                                   0.30087247.
                                                                0.7790712 ,
1.4577612 ,
                                  1.1306514 ,
                                                                                                                                                                   0.8482319
                                                                                                  1.4371244 .
                                                                                                                                  -2.7274785 ,
                                -0.56985384.
                                                                  0.39158794, -0.17285948,
                                                                                                                                 -1.384397
                                                                                                                                                                   -0.29015785.
                               -0.56985384, 0.39158794, -0.17285948, 3.9678284, -0.66227573, -0.49198876, 4.286491, -1.8598944, 0.12558945, -0.06563634, -2.224771, 2.0822936, -0.83319783, 1.6255909, 0.84546375, -0.26788828, 1.3658859, 3.4065766, 3.244361
                                                                                                                                   1.5287828 ,
                                                                                                                                                                   -0.3802559 ,
2.274344 ,
                                                                                                 2.0082936 , -0.8159753 ,
0.84546375, -2.1749024 ,
                                                                                                                                                                  -2.254789
                                                                                                                                                                   -0 4050005
                                -0.26/8822, 1.7698163 , 0.5977933 , 0.32413623, 1.7693163 , 0.5977933 , 0.9586846 , 1.2754706 , -2.076492 , 2.8482974 , 0.03156389 , 0.2842725 , 0.6990299 , -3.034646 , 1.9252772 , 0.19032483 , -0.4042304 , 0.23345727,
                                                                                                                                   0.2685584
                                                                                                                                                                 -1.3846186 ,
                                                                                                                                   0.3734416 ,
                                                                                                                                                                   1.1651148
                                                                                                                                   2.050075 ,
1.1805288 ,
                                                                                                                                                                   2.0976923 ,
                                                                                                                                   0.96000504, -1.2318734 .
                                                                                                                                  -0.28300276,
                                                                1.195374 ,
0.61264 ,
                                                                                                -0.26830855, -0.28300276, 1.791177
0.73491406, -1.7531322 , 1.2770014
                                 -0.84461105,
                                 -1.8392042 ,
                                                                0.61264 , 0.73491406, -1.7531322 , 1.2770014 , 2.2037764 , -0.4719132 , 1.6767381 , 2.088745 , 0.39926797, -2.281819 , -1.1530005 , 1.840919 ],
                                  3.557539 ,
                                  0.7665555
                             dtype=float32)
```

```
[ ] w2v_model.wv.most_similar('india')
          [('pakistan', 0.7414124011993408),
('malaysia', 0.6891069412231445),
            ('malaysia', 0.6891069412231445),

('china', 0.6626362204551697),

('australia', 0.645916759967804),

('beijings', 0.6376063227653503),

('norway', 0.6274385452270508),

('japan', 0.611946702003479),

('controlchina', 0.6110749244689941),

('indian', 0.6049240827560425),

('indias', 0.5988717079162598)]
[ ] w2v_model.wv.most_similar('china')
         [('beijing', 0.8647976517677307),
('taiwan', 0.8008958101272583),
('chinas', 0.7648460268974304),
('pyongyang', 0.6972832679748535),
('chinese', 0.6958582401275635),
('india', 0.6626362204551697),
('japan', 0.6597095131874084),
('boijings', 0.44402401255575)
              ('beijings', 0.6444934010505676),
            ('xi', 0.6359792947769165),
('waterway', 0.6162828803062439)]
[ ] w2v_model.wv.most_similar('usa')
          [('mcculloughthis', 0.5617169141769409),
('wirecom', 0.5184991955757141),
                'nl2n1gc0i1', 0.510539710521698),
'pacsharyl', 0.4913540482521057),
'pictwittercomsfe6zfdoli', 0.48563042283058167),
                'orgs', 0.4720892906188965),
'pictwittercomzkutv76jll', 0.4677456021308899),
'biz', 0.4658149182796478),
            ('flopped', 0.4636586606502533),
('gospel', 0.4619619846343994)]
[ ] w2v_model.wv.most_similar('gandhi')
            [('rahul', 0.7698065638542175),
                 '75yearold', 0.6625608801841736),
'cristina', 0.6558746099472046),
             ('cristina', 0.6558746099472046),

('ozawa', 0.6513022184371948),

('tounes', 0.641105592250824),

('sobotka', 0.6337205171585083),

('grillo', 0.6289705938070679),

('loyalist', 0.6274853944778442),

('mediashy', 0.6266793012619019),

('pastrana', 0.6204155683517456)]
[ ] tokenizer = Tokenizer()
            tokenizer.fit_on_texts(X)
[ ] X = tokenizer.texts_to_sequences(X)
  plt.hist([len(x) for x in X], bins =700)
            plt.show()
  0
               1200
               1000
                  800
                  600
                  400
                 200
                      0
                                                           2000 3000 4000 5000 6000 7000 8000
                                 0
                                             1000
```

```
len(nos[nos>1000])
     1580
[ ] maxlen = 1000
     X = pad\_sequences(X, maxlen=maxlen)
[ ] len(X[101])
     1000
[ ] vocab_size = len(tokenizer.word_index)+1
     vocab =tokenizer.word_index
[ ] def get_weight_matrix(model):
         weight_matrix = np.zeros((vocab_size, DIM))
        for word, i in vocab.items():
            weight_matrix[i] = model.wv[word]
        return weight_matrix
[ ] embedding_vectors = get_weight_matrix(w2v_model)
[ ] embedding_vectors.shape
     (231850, 100)
[ ] model = Sequential()
    model.add(Embedding(vocab_size, output_dim=DIM, weights = [embedding_vectors], input_length=maxlen, trainable = False))
model.add(LSTM(units=128))
    model.add(bense(1, activation='sigmoid'))
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['acc'])
[ ] model.summary()
    Model: "sequential"
    Layer (type)
                             Output Shape
                                                    Param #
                .....
     embedding (Embedding)
                             (None, 1000, 100)
                                                   23185000
    1stm (LSTM)
                           (None, 128)
                                                   117248
                           (None, 1)
    Total params: 23302377 (88.89 MB)
Trainable params: 117377 (458.50 KB)
Non-trainable params: 23185000 (88.44 MB)
[ ] X_train, X_test, y_train, y_test = train_test_split(X,y)
model.fit(X_train, y_train, validation_split=0.2, epochs=1)
    [ ] y_pred = (model.predict(X_test) >= 0.5).astype(int)
     351/351 [-----] - 8s 23ms/step
[ ] accuracy_score(y_test, y_pred)
     0.9824498886414254
[ ] print(f"accuracy_score : {accuracy_score(y_test, y_pred).round(4)*100}%")
     accuracy_score : 98.24000000000001%
print(classification_report(y_test, y_pred))
0
                  precision recall f1-score support
                       0.99 0.98
                                           0.98
                       0.97
                                          0.98
                                                    5259
        accuracy
                                          0.98
                                                   11225
    macro avg
weighted avg
                     0.98
                               0.98
                                          0.98
                                                    11225
                      0.98
                                0.98
                                                    11225
```

[] nos = np.array([len(x) for x in X])

```
x = ['this is a news']
      import tensorflow as tf
[ ] x = tokenizer.texts_to_sequences(x)
      x=pad_sequences(x, maxlen=maxlen)
[ ] (model.predict(x))
      1/1 [======] - 0s 31ms/step
      array([[0.00372225]], dtype=float32)
[ ] if (model.predict(x) >=0.5).astype(int)==0:
           print("the input 'x' is fake news")
          print("the input 'x' is real news")
      1/1 [======] - 0s 30ms/step
      the input 'x' is fake news
[ ] model.predict(x)
      1/1 [======] - 0s 51ms/step
      array([[0.00372225]], dtype=float32)
x = ['''The heart and neurological disorders have seen an uptick as a result of the post-COVID condition which reportedly began since the second wave of the virus, according to health experts. Speaking to ANI on Saturday, Dr Devi Prasad Sh
    "COVID patients especially during the second wave, there was definitely a slight increase in the incidence of COVID patients developing clot forms, and clots in the brain or in the heart. But that pattern we saw only during the second wave
   However, Dr Nitish Naik, Professor, Department of Cardiology, AIIMS, Delhi said that the study about the role of COVID in precipitating acute cardiac problems after recovery is still evolving. "All flu like illnesses have always been asso
   The expert explained that it can happen that some persons may experience persistent aches and pains, fatigue and palpitations during the recovery phase like after any viral illness."
   x = tokenizer.texts_to_sequences(x)
   x=pad_sequences(x, maxlen=maxlen)
   print((model.predict(x)))
   if (model.predict(x) >=0.3).astype(int)==0:
      print("the input 'x' is fake news")
[[0.98325956]]
   1/1 [-----] - 0s 47ms/step
```

Conclusion:

In conclusion, utilizing Natural Language Processing (NLP) techniques for fake news detection has proven to be a significant advancement in combating misinformation. The model developed demonstrates the potential of machine learning in identifying deceptive content, contributing to the ongoing efforts to maintain the integrity of information online. By leveraging NLP algorithms, the accuracy and efficiency of fake news detection have been greatly enhanced, empowering users to make informed decisions and fostering a more reliable digital information ecosystem. As we move forward, continued research and development in this field will play a pivotal role in ensuring the authenticity and trustworthiness of online content, thereby promoting a healthier and more informed society.





