

Fake News Petection 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
```

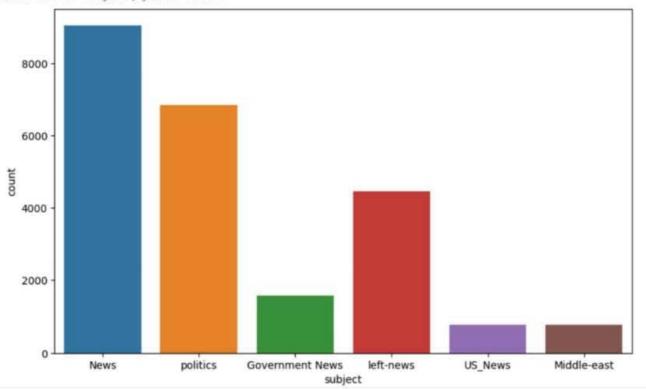
Index(['title', 'text', 'subject', 'date'], dtype='object')

[] 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)

Axes: xlabel='subject', ylabel='count'>



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:

```
XXtime
wordcloud = Wordcloud(width=1920, height=1800).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 66.4 s 66.4 s

Word cloud for True News:

real_news = pd.read_csv('/content/drive/MyDrive/input/True.csv') real news.head() title text subject date As U.S. budget fight looms, Republicans flip t ... WASHINGTON (Reuters) - The head of a conservat... politicsNews December 31, 2017 U.S. military to accept transgender recruits o... WASHINGTON (Reuters) - Transgender people will... politicsNews December 29, 2017 Senior U.S. Republican senator: 'Let Mr. Muell. WASHINGTON (Reuters) - The special counsel inv. politicsNews December 31, 2017 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

```
[ ] text = ' '.join(real_news['text'].tolist())

[ ] Xtime
    wordcloud = Wordcloud(width=1920, height=1800).generate(text)
    fig = plt.figure(figsize=(10,10))
    plt.imshow(wordcloud)
    plt.axis('off')
    plt.tight_layout(pad=0)
    plt.show()
```



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

1.1	real_news.sample(5)	
	1 cer mensionship refol	

	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 Kim	KUALA LUMPUR (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

```
[ ] len(unknown pubishers)
     35
     real_news.iloc[unknown_pubishers].text
              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. ...
    6660
              Republican members of Congress are complaining...
    6823
              Over the course of the U.S. presidential campa...
              After going through a week reminiscent of Napo...
    7922
             The following timeline charts the origin and s...
    8194
    8195
             Global health officials are racing to better u...
    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...
    8506
             The following timeline charts the origin and s...
    8771
             In a speech weighted with America's complicate...
    8970
    9998
             The following timeline charts the origin and s...
    9009
             Global health officials are racing to better u...
    9307
             It's the near future, and North Korea's regime...
              GOP leaders have unleashed a stunning level of...
    9618
              Caitlyn Jenner posted a video on Wednesday (Ap...
    9737
    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[θ].strip())
            tmp_text.append(record[1].strip())
  ] real_news['publisher'] = publisher
     real_news['text'] = tmp_text
[ ] real_news.head()
```

```
[ ] real_news.head()
                                                  title
                                                                                                               subject
                                                                                                                                                                   publisher
                                                                                                     text
                                                                                                                                      date

    As U.S. budget fight looms, Republicans flip t...

                                                             The head of a conservative Republican faction ... politicsNews December 31, 2017
                                                                                                                                                      WASHINGTON (Reuters)
             U.S. military to accept transgender recruits o...
                                                               Transgender people will be allowed for the fir... politicsNews. December 29, 2017.
                                                                                                                                                      WASHINGTON (Reuters)
      2
           Senior U.S. Republican senator: 'Let Mr. Muell...
                                                                                                                                                      WASHINGTON (Reuters)
                                                               The special counsel investigation of links bet. politicsNews December 31, 2017
       3 FBI Russia probe helped by Australian diplomat... Trump campaign adviser George Papadopoulos tol... politicsNews December 30, 2017
                                                                                                                                                      WASHINGTON (Reuters)
       4 Trump wants Postal Service to charge 'much mor...
                                                            President Donald Trump called on the U.S. Post... politicsNews December 29, 2017 SEATTLE/WASHINGTON (Reuters)
[] real_news.shape
      (21417, 5)
[ ] empty fake index = [index for index,text in enumerate(fake news.text.tolist()) if str(text).strip()==""]
[ ] fake_news.iloc[empty_fake_index]
                                                                   title text subject
                                                                                                   date
       10923
                       TAKE OUR POLL: Who Do You Think President Trum.
                                                                                   politics May 10, 2017.
       11041
                         Joe Scarborough BERATES Mika Brzezinski Over "...
                                                                                   politics Apr 26 2017
       11190
                                                                                   politics
                    WATCH TUCKER CARLSON Scorch Sanctuary City May
                                                                                             Apr 6, 2017
       11225
                     MAYOR OF SANCTUARY CITY: Trump Trying To Make ...
                                                                                   politics
                                                                                             Apr 2, 2017
       11236
                        SHOCKER: Public School Turns Computer Lab Info...
                                                                                   politics Apr 1, 2017
      21816 BALTIMORE BURNS MARYLAND GOVERNOR BRINGS IN N
                                                                                 left-news Apr 27 2015
                 FULL VIDEO: THE BLOCKBUSTER INVESTIGATION INTO.
                                                                                  left-news Apr 25, 2015
                                                                                  left-news Apr 25, 2015
      21827
                   (VIDEO) HILLARY CLINTON: RELIGIOUS BELIEFS MUS.
       21857
               (VIDEO)/CE PROTECTING OBAMA: WON'T RELEASE NAM...
                                                                                  left-news Apr 14, 2015
      21873
                  (VIDEO) HYSTERICAL SNL TAKE ON HILLARY'S ANNOU.
                                                                                  left-news Apr 12, 2015
      630 rows × 4 columns
[ ] real_news['text'] = real_news['title']+" "+ real_news['text']
[ ] fake_news['text'] = fake_news['title']+" "+ fake_news['text']
     real_news['text'] = real_news['text'].apply(lambda x: str(x).lower())
fake_news['text'] = fake_news['text'].apply(lambda x: str(x).lower())
     real_news['class']-1
U
      fake_news['class']=0
[ ] real_news.columns
      Index(['title', 'text', 'subject', 'date', 'publisher', 'class'], dtype='object')
     real - real_news[['text','class']]
      fake = fake_news[['text','class']]
     data = real.append(fake, ignore_index=True)
      data.head()
 🙆 cipython-input-33-8770c2a2a545>:1: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.
       data - real.append(fake, ignore_index=True)
                                              text class
            as u.s. budget fight looms, republicans flip t.
           u.s. military to accept transgender recruits o.
         senior u.s. republican senator. Yet mr. muell.
      3 fbi russia probe helped by australian diplomat.
      4 trump wants postal service to charge 'much mor...
```

```
o inin install spacy++2.2.3
           lpython -m spacy download en_core_web_sm
           lpip install beautifulsoup4--4.9.1
           lpis install textblobes#.15.1
           |pip install git+https://github.com/laxwimerit/preprocess kgptalkie.git --upgrade --force-reinstall
 Oute: This error originates from a subprocess, and is likely not a problem with pip.
2023-10-16 16:41:23.050566: W tensorflow/compiler/tf2tensorrt/utils/py_utils.cc:30] TF-TRT Warning: Could not find TensorRT
             Downloading https://eithub.com/explosion/spacy-models/releases/Nomiced/en_core_web_sm-1.6.0/em_core_web_sm-1.6.0-pyl-nome-any.uml (12.6 PM)
           12.8/12.8 MB 21.6 MB/3 eta 0:00:00

Requirement already satisfied: spacy3.7.8,>=3.6.8 in /usr/local/lib/python3.18/dist-packages (from en-core-web-im==3.6.8) (3.6.1)
          Requirement already satisfied: spacy-legacy3.1.0,><3.0.11 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-ueb-sm==1.6.0) (1.0.12)
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-ueb-sm==3.6.0) (1.0.5)
           Requirement already satisfied: murmurhusht1.1.0,>-0.28.0 in /usr/local/llb/python3.10/dist-packages (from spacyc3.7.0,>=3.6.0->en-core-ueb-sm==2.6.0) (1.0.10)
Requirement already satisfied: cymemc2.1.0,>=2.0.2 in /usr/local/llb/python3.10/dist-packages (from spacyc3.7.0,>=3.6.0->en-core-ueb-sm==2.6.0) (2.0.0)
         Requirement already satisfied: cymeecl.1.0,>-2.0.2 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (3.0.0)
Requirement already satisfied: preshedcl.1.0,>-3.0.2 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (3.0.0)
Requirement already satisfied: thinc(S.2.0,>-3.1.8 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (3.1.12)
Requirement already satisfied: smashicl.2.0,>-0.0.1 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (2.1.2)
Requirement already satisfied: smalyC.8.0,>-2.4.3 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (2.4.8)
Requirement already satisfied: cataloguec2.1.0,>-2.6.5 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (2.0.10)
Requirement already satisfied: pathy-sh.2.0 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (0.0.0)
Requirement already satisfied: smart-openC.0.0,>-5.2.1 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (0.0.0)
Requirement already satisfied: smart-openC.0.0,>-5.2.1 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (4.06.1)
Requirement already satisfied: tupboS.0.0,>-4.3.0.0 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (4.06.1)
Requirement already satisfied: tupboS.0.0,>-2.13.0 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (4.06.1)
Requirement already satisfied: tupboS.0.0,>-2.13.0 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (2.13.0)
Requirement already satisfied: smart-openC.0.0,>-2.13.0 in /usr/local/lib/python3.10/dist-packages (from spacyG.7.0,>-3.6.0-sen-cone-usb-ins-3.6.0) (2.13.0)
Requirement already satisfied: smart-openC.0.0,>-2.13.0 in /usr
           Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0-zen-core-web-sm=2.6.0) (3.1.2)
Requirement already satisfied: satuptools in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0-zen-core-web-sm=2.6.0) (67.7.1)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0-zen-core-web-sm=2.6.0) (23.2)
           Sequirement aiready satisfied: langcodes(4.8.8,>=3.2.0 in /usr/local/lib/python3.10/dist-packages (from spacy(3.7.0,>=3.6.0-sen-core-ueb-in=3.6.0) (3.3.0) Requirement aiready satisfied: typing-extensions>=4.2.0 in /usr/local/lib/python2.10/dist-packages (from pydamticl=1.0,|=1.8.1,(3.0,0,)=1.7.4-spacy(3.7.0,>=2.6.0-sen-core-ueb-sn=3.6.0) (4.5.0)
           Requirement already satisfied: churset-normalizer4,>+2 in /usr/local/lib/python3.18/dist-packages (from requests<0.8.8,>+2.13.8->spacy<1.7.8,>+3.6.8->en-core-web-sm+3.6.8) (3.3.8)
Requirement already satisfied: idnac4,>+2.5 in /usr/local/lib/python3.18/dist-packages (from requests<0.8.8,>+2.13.8->spacy<1.7.8,>+3.6.8->en-core-web-sm+3.6.8) (3.4.8)
           Requirement already satisfied: unlibb(3,>=1.21.1 in /usr/local/lib/python3.08/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/python3.08/dist-packages (from requests(3.0.0,)==2.13.0->spacy(3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2023.7.22)
Requirement already satisfied: bits(0.0.0,>=0.7.8 in /usr/local/lib/python3.08/dist-packages (from thinc(0.2.0,>=0.1.0->spacy(3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (0.7.11)
           Requirement already satisfied: confection(1.0.0,>=0.0.1 in /usr/local/lib/python3.10/dist-packages (from thince3.2.0,>=0.1.3->spacy<3.7.0,>=3.6.0->en-core-seb-sm=3.6.0) (0.1.3)
Requirement already satisfied: click(5.0.0,>=7.1.1 in /usr/local/lib/python3.20/dist-packages (from typer/0.10.0,>=0.1.0->spacy<3.7.0,>=3.6.0->en-core-seb-sm=3.6.0) (0.1.7)
           Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-puckages (from jinja2->spacy<3.7.0,>=3.5.0->en-core-web-in==3.6.0) (2.1.3)
 [ ] import preprocess_kgptalkie as ps
 [ ] data['text'] = data['text'].apply(lambda x: ps.remove_special_chars(x))
[ ] from gongle.colab import
         drive.mount('/content/drive')
         Grise already mounted at /content/drive; to attempt to forcifily remount, call drive.minent("/content/drive", force remount-Troe)
[ ] y - data['class'].values
[ ] X = [d.split() for d in data['text'].tolist()]
[-] type(x(x))
I'll principles
        ['as', 'os', 'hodget', 'fight', 'loom', 'republican', 'flip', 'their', 'fiscal', 'script', 'the', 'of', 'a', 'concervative', 'republican', 'faction', 'the', 'us', 'congress', 'who', 'voted', 'this', 'script', 'the', 'a', 'ho
        . ___
              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 , -8.4013639 , 1.5510874 , -1.9637966
2.3293521 , 1.0115958 , -1.7413545 , 2.1367438
2.0449893 , 0.7700712 , 3.5431712 , -3.6178732
                                                                                                                                                         1,7516787
                                                                                                                           2.1367438 ,
                                                                                                                                                          0.30087247,
                                                                                                                          -3.6178732 , -2.512199
                                1.1306514 ,
0.56985384,
3.9678284 ,
                                                             1.4577612 , 1.4371244 ,
0.39158794, -0.17285948,
-0.66227573, -0.49190876,
                                                                                                                          -2.7274785 ,
                                                                                                                                                         0.8482319
                                                                                                                                                          0.29815785
                                                                                                                           2.384397
                                                                                                                           1.5287828
                                                                                                                                                         -0.3802559
                                                             -1.8598944 , 8.12558945, 2.3769717 , -2.254771 , 2.0082936 , -0.8159753 , 1.6255909 , 8.64546375, -2.1740024 ,
                                4.286491 .
                                                                                                                                                         2.274344
                                0.06563634,
                                                                                                                                                        -2.254789
                                0.83319783,
                                                             1.3658859 , 3.4065766 , -0.53475
1.7603163 , 0.5977033 , 0.268581
                               0.20788828.
                                                                                                                                                         0.80121416.
                                0.32413623,
                                                                                                                           0.2685584
                                                                                                                                                        -1.1846186
                                                                                                                           0.3734416 .
                                2.8482974 .
                                                                                            8.2842725 .
                                                             0.03156389.
                                                                                                                           2.050075
                                                                                                                                                         8.83186256,
                                0.00002000
                                                             -1.034545
                                                                                            1.9253772
                                                                                                                           1.1885288
                                                                                                                                                         2.8976923
                                                             1.195374 , -0.26830855, -0.28300270, 1.791177 
0.61264 , 0.73491406, -1.7531322 , 1.2770014 , 2.203764 , 0.4719132 , 1.6767381 , 2.088745 
0.39926797, -2.281819 , -1.1530005 , 1.840919 ],
                               -0.84461105,
                               -1.8392042 .
                                0.7665555
                           dtype=float32)
```

```
[ ] w2v_model.wv.most_similar('india')
         [('pakistan', 0.7414124811993488),
('malaysia', 0.6891869412231445),
            ('Balaysia', 0.009/1000012231403/,
('china', 0.6526162204551607),
('australia', 0.645916759967804),
('beijings', 0.6376063227653503),
('norway', 0.6274385452270508),
('japan', 0.611946702003479),
           ('controlchina', 0.6119749244689941),
('indian', 0.6049240827560425),
('indias', 0.5988717079162598)]
[ ] w2v_model.wv.most_similar('china')
          [('beljing', 0.8647976517677307),
              'taiwan', 0.8008958181272583),
'chinas', 0.7648460268974304),
             ('ipongyang', 0.6972832679748535),

('chinese', 0.6972832679748535),

('india', 0.6626362204551697),

('japan', 0.6597095131874084),
            ('beijings', 0.6444934010505676),
('xi', 0.6359792947769165),
            ('waterway', 0.6162828803062439)]
[ ] w2v_model.wv.most_similar('usa')
          [('mcculloughthis', 0.5617169141769409),
               wirecom', 8.5184991955757141),
              'nl2nigc011', 0.510539710521098),
'pacsharyl', 0.4913540482521057),
'pictwittercomsfe62fdoli', 0.48563042283858167),
             ("orgs", 0.4720892906188965),
("pictwittercomzkutv76jll", 0.4677456021308899),
("bir", 0.4658149182796478),
           ('flopped', 0.4636586606502533), ('gospel', 0.4619619846343994)]
 [ ] w2v_model.wv.most_similar('gandhi')
           [('rahul', 0.7698865638542175),
               "Syearold', 8.6625688881841736),
"Cristina', 8.6553746899472046),
"orawa', 8.6513022184371948),
"tounes', 8.641105592258824),
"sobotka', 8.6337265171585083),
             ('grillo', 8.6289785938878679),
('loyalist', 8.6274853944778442),
('mediashy', 8.6266793812619819),
('pastrana', 8.6284155683517456)]
 [ ] 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
                            0
                                      1000
                                                   2000
                                                                3000
                                                                            4000 5000
                                                                                                        6000
                                                                                                                     7000
                                                                                                                                   8000
```

```
[ ] nos = np.array([len(x) for x in X])
     len(nos[nos>1000])
[ ] maxlen + 1000
X = pad_sequences(X, maxlen-maxlen)
[ ] len(X[181])
     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, 1 in vocab.items():
             weight_matrix[1] = 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(Dense(), activation='sigmoid'))
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['acc'])
[ ] model.summary()
    Model: "sequential"
    Layer (type)
                              Output Shape
                                                     Paran #
     embedding (Embedding)
                             (None, 1888, 188)
                                                     23185000
                                                    117248
     1stm (LSTM)
                           (None, 128)
     dense (Dense)
                            (None, 1)
    Total params: 23302377 (88.89 PM)
Trainable params: 117377 (458.50 KR)
Non-trainable params: 23165000 (88.44 PM)
[ ] X_train, X_test, y_train, y_test = train_test_split(X,y)
o 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
                                0.98
                      0.97
                                          0.98
                                                     5259
                                           0.98
                                                    11225
         accuracy
                               0.98
                    0.98
                                         0.98
0.98
        macro avg
                                                     11225
     weighted avg
                                                     11225
```

```
x = ['this is a news']
      import tensorflow as tf
[ ] x = tokenizer.texts_to_sequences(x)
      x-pad_sequences(x, maxlen-maxlen)
[ ] (model.predict(x))
      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 Sims/step
      array([[0.00372225]], dtype=float32)
x = ["The heart and nourological 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 Davi Pressd Si
    "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
    Mounter, Dr Nitish Naik, Professor, Department of Cardiology, 4196, 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 palms, fatigue and palpitations during the recovery phase like after any viral illness.
    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")
      print("the input "x" is real news")
€ 1/1 [------] - 8s 60es/step
    [[0.98325056]]
    1/1 [------ - Bs 47ms/step
    the input 'x' is real news
```

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.





