## Fake News Prediction

November 30, 2022

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[]: About the Dataset:
     1. id: unique id for a news article
     2. title: the title of a news article
     3. author: author of the news article
     4. text: the text of the article; could be incomplete
     5. label: a label that marks whether the news article is real or fake:
                1: Fake news
                0: real News
     Dataset Link: https://www.kaggle.com/c/fake-news/data?select=train.csv
    Importing the Dependencies
[]: import numpy as np
     import pandas as pd
     import re
     from nltk.corpus import stopwords
     from nltk.stem.porter import PorterStemmer
     from sklearn.feature_extraction.text import TfidfVectorizer
     from sklearn.model selection import train test split
     from sklearn.linear_model import LogisticRegression
     from sklearn.metrics import accuracy_score
[]: import nltk
    nltk.download('stopwords')
    [nltk_data] Downloading package stopwords to /root/nltk_data...
                  Package stopwords is already up-to-date!
    [nltk_data]
[]: True
[]: # printing the stopwords in English
     print(stopwords.words('english'))
    ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're",
    "you've", "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he',
    'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's",
    'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves', 'what',
    'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'is',
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'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
    'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or',
    'because', 'as', 'until', 'while', 'of', 'at', 'by', 'for', 'with', 'about',
    'against', 'between', 'into', 'through', 'during', 'before', 'after', 'above',
    'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under',
    'again', 'further', 'then', 'once', 'here', 'there', 'when', 'where', 'why',
    'how', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some',
    'such', 'no', 'nor', 'not', 'only', 'own', 'same', 'so', 'than', 'too', 'very',
    's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now',
    'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', "aren't", 'couldn',
    "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn',
    "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't",
    'mustn', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn',
    "shouldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'wouldn',
    "wouldn't"]
    Data Pre-processing
[]: # loading the dataset to a pandas DataFrame
     news_dataset = pd.read_csv('/content/train.csv')
[]: news_dataset.shape
[]: (20800, 5)
[]: # print the first 5 rows of the dataframe
     news_dataset.head()
[]:
        id ... label
        0 ...
         1
                  0
     1
           ...
     2
        2 ...
                  1
        3 ...
     3
                  1
         4 ...
                  1
     [5 rows x 5 columns]
[]: # counting the number of missing values in the dataset
     news_dataset.isnull().sum()
[]: id
                  0
     title
                558
     author
               1957
     text
                 39
```

label

dtype: int64

0

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[]: # replacing the null values with empty string
     news_dataset = news_dataset.fillna('')
[]: # merging the author name and news title
     news_dataset['content'] = news_dataset['author']+' '+news_dataset['title']
[]: print(news_dataset['content'])
    0
             Darrell Lucus House Dem Aide: We Didn't Even S...
    1
             Daniel J. Flynn FLYNN: Hillary Clinton, Big Wo...
    2
             Consortiumnews.com Why the Truth Might Get You...
    3
             Jessica Purkiss 15 Civilians Killed In Single ...
             Howard Portnoy Iranian woman jailed for fictio...
             Jerome Hudson Rapper T.I.: Trump a 'Poster Chi...
    20795
    20796
             Benjamin Hoffman N.F.L. Playoffs: Schedule, Ma...
             Michael J. de la Merced and Rachel Abrams Macy...
    20797
    20798
             Alex Ansary NATO, Russia To Hold Parallel Exer...
    20799
                        David Swanson What Keeps the F-35 Alive
    Name: content, Length: 20800, dtype: object
[]: # separating the data & label
     X = news_dataset.drop(columns='label', axis=1)
     Y = news dataset['label']
[]: print(X)
     print(Y)
              id ...
                                                                 content
    0
               0
                 ... Darrell Lucus House Dem Aide: We Didn't Even S...
    1
                  ... Daniel J. Flynn FLYNN: Hillary Clinton, Big Wo...
                  ... Consortiumnews.com Why the Truth Might Get You...
    2
    3
                     Jessica Purkiss 15 Civilians Killed In Single ...
    4
                      Howard Portnoy Iranian woman jailed for fictio...
           20795 ... Jerome Hudson Rapper T.I.: Trump a 'Poster Chi...
    20795
    20796 20796 ... Benjamin Hoffman N.F.L. Playoffs: Schedule, Ma...
    20797
           20797 ... Michael J. de la Merced and Rachel Abrams Macy...
           20798 ... Alex Ansary NATO, Russia To Hold Parallel Exer...
    20798
    20799
           20799
                                David Swanson What Keeps the F-35 Alive
    [20800 rows x 5 columns]
    0
             1
    1
             0
    2
             1
    3
             1
    4
             1
```

```
20795
             0
    20796
    20797
    20798
             1
    20799
             1
    Name: label, Length: 20800, dtype: int64
    Stemming:
    Stemming is the process of reducing a word to its Root word
    example: actor, actress, acting -> act
[ ]: port_stem = PorterStemmer()
[]: def stemming(content):
         stemmed_content = re.sub('[^a-zA-Z]',' ',content)
         stemmed_content = stemmed_content.lower()
         stemmed_content = stemmed_content.split()
         stemmed_content = [port_stem.stem(word) for word in stemmed_content if not__
      →word in stopwords.words('english')]
         stemmed_content = ' '.join(stemmed_content)
         return stemmed content
[]: news_dataset['content'] = news_dataset['content'].apply(stemming)
[]: print(news_dataset['content'])
    0
             darrel lucu hous dem aid even see comey letter ...
    1
             daniel j flynn flynn hillari clinton big woman...
    2
                         consortiumnew com truth might get fire
             jessica purkiss civilian kill singl us airstri...
    4
             howard portnoy iranian woman jail fiction unpu...
    20795
             jerom hudson rapper trump poster child white s...
    20796
             benjamin hoffman n f l playoff schedul matchup...
             michael j de la merc rachel abram maci said re...
    20797
    20798
             alex ansari nato russia hold parallel exercis ...
    20799
                                      david swanson keep f aliv
    Name: content, Length: 20800, dtype: object
[]: #separating the data and label
     X = news_dataset['content'].values
     Y = news_dataset['label'].values
[]: print(X)
    ['darrel lucu hous dem aid even see comey letter jason chaffetz tweet'
     'daniel j flynn flynn hillari clinton big woman campu breitbart'
     'consortiumnew com truth might get fire' ...
```

```
new york time'
     'alex ansari nato russia hold parallel exercis balkan'
     'david swanson keep f aliv']
[]: print(Y)
    [1 0 1 ... 0 1 1]
[]: Y.shape
[]: (20800,)
[]: # converting the textual data to numerical data
     vectorizer = TfidfVectorizer()
     vectorizer.fit(X)
     X = vectorizer.transform(X)
[]: | print(X)
      (0, 15686)
                     0.28485063562728646
      (0, 13473)
                     0.2565896679337957
      (0, 8909)
                     0.3635963806326075
      (0, 8630)
                     0.29212514087043684
      (0, 7692)
                     0.24785219520671603
      (0, 7005)
                     0.21874169089359144
      (0, 4973)
                     0.233316966909351
      (0, 3792)
                     0.2705332480845492
      (0, 3600)
                     0.3598939188262559
      (0, 2959)
                     0.2468450128533713
      (0, 2483)
                     0.3676519686797209
      (0, 267)
                     0.27010124977708766
      (1, 16799)
                     0.30071745655510157
      (1, 6816)
                     0.1904660198296849
      (1, 5503)
                     0.7143299355715573
      (1, 3568)
                     0.26373768806048464
      (1, 2813)
                     0.19094574062359204
      (1, 2223)
                     0.3827320386859759
      (1, 1894)
                     0.15521974226349364
      (1, 1497)
                     0.2939891562094648
      (2, 15611)
                     0.41544962664721613
      (2, 9620)
                     0.49351492943649944
      (2, 5968)
                     0.3474613386728292
      (2, 5389)
                     0.3866530551182615
      (2, 3103)
                     0.46097489583229645
      (20797, 13122)
                             0.2482526352197606
      (20797, 12344)
                             0.27263457663336677
```

'michael j de la merc rachel abram maci said receiv takeov approach hudson bay

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(20797, 10306)
                            0.08038079000566466
      (20797, 9588) 0.174553480255222
      (20797, 9518) 0.2954204003420313
      (20797, 8988) 0.36160868928090795
      (20797, 8364) 0.22322585870464118
      (20797, 7042) 0.21799048897828688
      (20797, 3643) 0.21155500613623743
      (20797, 1287) 0.33538056804139865
      (20797, 699) 0.30685846079762347
      (20797, 43)
                    0.29710241860700626
      (20798, 13046)
                            0.22363267488270608
      (20798, 11052)
                            0.4460515589182236
      (20798, 10177)
                            0.3192496370187028
      (20798, 6889) 0.32496285694299426
      (20798, 5032) 0.4083701450239529
      (20798, 1125) 0.4460515589182236
      (20798, 588) 0.3112141524638974
      (20798, 350) 0.28446937819072576
      (20799, 14852)
                            0.5677577267055112
      (20799, 8036) 0.45983893273780013
      (20799, 3623) 0.37927626273066584
      (20799, 377) 0.5677577267055112
    Splitting the dataset to training & test data
[]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.2,__
      ⇔stratify=Y, random_state=2)
    Training the Model: Logistic Regression
[]: model = LogisticRegression()
[]: model.fit(X_train, Y_train)
[]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                        intercept scaling=1, 11 ratio=None, max iter=100,
                        multi_class='auto', n_jobs=None, penalty='12',
                        random_state=None, solver='lbfgs', tol=0.0001, verbose=0,
                        warm_start=False)
    Evaluation
    accuracy score
[]: # accuracy score on the training data
     X train prediction = model.predict(X train)
     training_data_accuracy = accuracy_score(X_train_prediction, Y_train)
[]: print('Accuracy score of the training data : ', training_data_accuracy)
```

0.24778257724396507

(20797, 12138)

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[]: # accuracy score on the test data
     X_test_prediction = model.predict(X_test)
     test_data_accuracy = accuracy_score(X_test_prediction, Y_test)
[]: print('Accuracy score of the test data : ', test_data_accuracy)
    Accuracy score of the test data : 0.9790865384615385
    Making a Predictive System
[ ]: X_new = X_test[3]
     prediction = model.predict(X_new)
     print(prediction)
     if (prediction[0]==0):
      print('The news is Real')
       print('The news is Fake')
    [0]
    The news is Real
[]: print(Y_test[3])
    0
[]:
```

Accuracy score of the training data: 0.9865985576923076