fake-news-detection

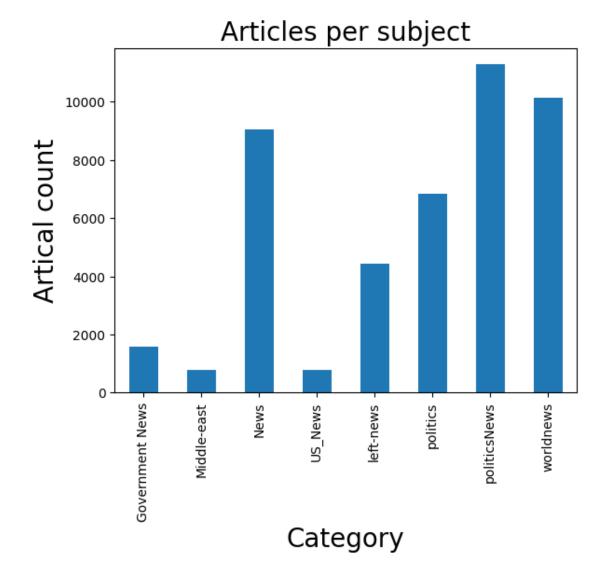
June 28, 2024

```
[]: import pandas as pd
     import numpy as nm
     from sklearn.model_selection import train_test_split as ttp
     from sklearn.metrics import classification_report
     import re
     import string
     import matplotlib.pyplot as plt
[]: data_true=pd.read_csv("/content/drive/MyDrive/ML_Team12/1SV21CS003/True.csv")
     data fake=pd.read csv("/content/drive/MyDrive/ML Team12/1SV21CS003/Fake.csv")
[]: data_true.shape,data_fake.shape
[]: ((21417, 4), (23481, 4))
[]: data_true["class"]=1
     data fake["class"]=0
[]: data_true_manual_testing=data_true.tail(10)
     for i in range(21416,21406,-1):
       data_true.drop([i],axis=0,inplace=True)
     data_fake_manual_testing=data_fake.tail(10)
     for i in range(21416,21406,-1):
       data_fake.drop([i],axis=0,inplace=True)
[]: data_manual_testing=pd.
      -concat([data_fake_manual_testing,data_true_manual_testing],axis=0)
     data_manual_testing.to_csv("manual_testing.csv")
[]: data_merge=pd.concat([data_true,data_fake],axis=0)
     data_merge.head(10)#displaying first 10 values after above operation
[]:
                                                    title \
     O As U.S. budget fight looms, Republicans flip t...
     1 U.S. military to accept transgender recruits o...
     2 Senior U.S. Republican senator: 'Let Mr. Muell...
     3 FBI Russia probe helped by Australian diplomat...
```

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5 White House, Congress prepare for talks on spe...
     6 Trump says Russia probe will be fair, but time...
     7 Factbox: Trump on Twitter (Dec 29) - Approval ...
               Trump on Twitter (Dec 28) - Global Warming
     8
     9 Alabama official to certify Senator-elect Jone...
                                                     text
                                                                 subject \
     O WASHINGTON (Reuters) - The head of a conservat... politicsNews
     1 WASHINGTON (Reuters) - Transgender people will... politicsNews
     2 WASHINGTON (Reuters) - The special counsel inv... politicsNews
     3 WASHINGTON (Reuters) - Trump campaign adviser ... politicsNews
     4 SEATTLE/WASHINGTON (Reuters) - President Donal... politicsNews
     5 WEST PALM BEACH, Fla./WASHINGTON (Reuters) - T... politicsNews
     6 WEST PALM BEACH, Fla (Reuters) - President Don... politicsNews
     7 The following statements were posted to the ve... politicsNews
     8 The following statements were posted to the ve... politicsNews
     9 WASHINGTON (Reuters) - Alabama Secretary of St... politicsNews
                      date class
     0 December 31, 2017
                                1
     1 December 29, 2017
                                1
     2 December 31, 2017
                                1
     3 December 30, 2017
                                1
     4 December 29, 2017
     5 December 29, 2017
     6 December 29, 2017
                                1
     7 December 29, 2017
                                1
     8 December 29, 2017
                                1
     9 December 28, 2017
                                1
[]: print(data_merge.groupby(['subject'])['text'].count())
     data_merge.groupby(['subject'])['text'].count().plot(kind="bar")
     plt.title("Articles per subject",size=20)
     plt.xlabel("Category",size=20)
     plt.ylabel("Artical count",size=20)
     plt.show()
    subject
    Government News
                        1570
    Middle-east
                         778
    News
                        9050
    US_News
                         783
    left-news
                        4449
    politics
                        6841
    politicsNews
                       11272
    worldnews
                       10135
```

4 Trump wants Postal Service to charge 'much mor ...

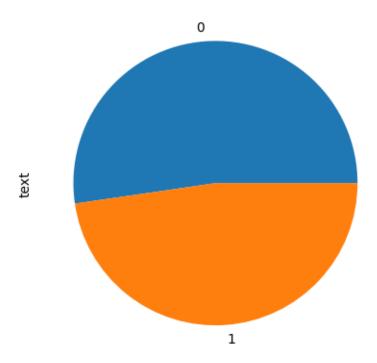
Name: text, dtype: int64



```
[]: print(data_merge.groupby(['class'])['text'].count())
    print("0 = Fake News\n1 = True News")
    data_merge.groupby(['class'])['text'].count().plot(kind="pie")
    plt.title("Fake News vs True News",size=20)
    plt.show()
```

class
0 23471
1 21407
Name: text, dtype: int64
0 = Fake News
1 = True News

Fake News vs True News



```
[]: text class
21837 Nothing but the best for the queen and the que... 0
13798 HARARE (Reuters) - New Zimbabwean President Em... 1
8201 WASHINGTON (Reuters) - Republican presidential... 1
```

data.head(10)

```
1566
            NEW YORK (Reuters) - Former U.S. Representativ...
                                                                  1
     13721 While researching for a wrap-up on the June 7 \dots
                                                                  0
     16952
           MANILA (Reuters) - Three Russian warships, inc...
                                                                  1
           BERLIN (Reuters) - German Chancellor Angela Me...
     12028
                                                                  1
     9428
            Trey Gowdy railed against the erosion of trust...
                                                                  0
           21st Century Wire says For many viewers, this ...
     23224
                                                                  0
     5894
            In the wake of the worst mass shooting in Amer...
                                                                  0
[]: data.isnull().sum()
              0
[]: text
     class
              0
     dtype: int64
[]: def filtering(data):
      text=data.lower()
       text=re.sub('\[.*?\]', '', text)
       text=re.sub("\\W"," ",text)
       text=re.sub('https?://\S+|www\.\S+', '', text)
       text=re.sub('<.*?>+', '', text)
       text=re.sub('[%s]' % re.escape(string.punctuation), '', text)
       text=re.sub('\n', '', text)
       text=re.sub('\w*\d\w*', '', text)
       return text
[]: data['text']=data['text'].apply(filtering)
     data.head(10)
[]:
                                                                class
     21837 nothing but the best for the queen and the que...
                                                                  0
     13798 harare reuters
                               new zimbabwean president em...
                                                                  1
     8201
            washington reuters
                                   republican presidential...
                                                                  1
     1566
            new york reuters
                                 former u s representativ...
                                                                  1
     13721 while researching for a wrap up on the june p...
                                                                  0
     16952 manila reuters
                              three russian warships inc...
                                                                  1
     12028 berlin reuters
                               german chancellor angela me...
     9428
           trey gowdy railed against the erosion of trust...
           century wire says for many viewers this came...
     23224
                                                                  0
     5894
            in the wake of the worst mass shooting in amer...
                                                                  0
[]: x=data['text']
     y=data['class']
[]: x_train,x_test,y_train,y_test=ttp(x,y,test_size=0.25,random_state=0)
[]: from sklearn.feature_extraction.text import TfidfVectorizer
```

```
vectorization = TfidfVectorizer()
     xv_train = vectorization.fit_transform(x_train)
     xv_test = vectorization.transform(x_test)
[]: from sklearn.linear_model import LogisticRegression
     LR = LogisticRegression()
     LR.fit(xv_train,y_train)
[]: LogisticRegression()
[]: pred_lr=LR.predict(xv_test)
     LR.score(xv_test, y_test)
     print(classification_report(y_test,pred_lr))
                  precision
                               recall f1-score
                                                   support
               0
                       0.99
                                 0.98
                                           0.99
                                                      5928
               1
                       0.98
                                 0.99
                                           0.98
                                                      5292
                                           0.99
                                                     11220
        accuracy
       macro avg
                       0.98
                                 0.99
                                           0.99
                                                     11220
    weighted avg
                       0.99
                                 0.99
                                            0.99
                                                     11220
[]: def predict_news(text):
         text_vectorized = vectorization.transform([filtering(text)])
         prediction = LR.predict(text_vectorized)
         if prediction == 1:
             return "This news is likely true."
         else:
             return "This news is likely fake."
     user_input = input("Enter news text: ")
     result = predict_news(user_input)
     print(result)
    Enter news text: modi is died
    This news is likely fake.
[]: from sklearn.tree import DecisionTreeClassifier
     DT = DecisionTreeClassifier()
     DT.fit(xv_train, y_train)
```

```
[ ]: DecisionTreeClassifier()
[]: DecisionTreeClassifier()
[ ]: DecisionTreeClassifier()
[]: pred_dt = DT.predict(xv_test)
[]: DT.score(xv_test, y_test)
[]: 0.9962566844919786
[]: print (classification_report(y_test, pred_lr))
                  precision
                               recall f1-score
                                                   support
               0
                       0.99
                                 0.98
                                           0.99
                                                      5928
                       0.98
                                 0.99
                                           0.98
                                                      5292
               1
                                           0.99
                                                     11220
        accuracy
       macro avg
                       0.98
                                 0.99
                                            0.99
                                                     11220
    weighted avg
                       0.99
                                 0.99
                                            0.99
                                                     11220
[]: def predict_news(text):
         text_vectorized = vectorization.transform([filtering(text)])
         prediction = DT.predict(text_vectorized)
         if prediction == 1:
             return "This news is likely true."
         else:
             return "This news is likely fake."
     user_input = input("Enter news text: ")
     result = predict_news(user_input)
     print(result)
    Enter news text: modi is dies
    This news is likely fake.
[]: from sklearn.ensemble import RandomForestClassifier
[]: RF = RandomForestClassifier(random_state = 0)
     RF.fit(xv_train, y_train)
[]: RandomForestClassifier(random_state=0)
[]: RandomForestClassifier(random_state=0)
```

```
[]: RandomForestClassifier(random_state=0)
[]: pred_rf = RF.predict(xv_test)
[]: RF.score(xv_test, y_test)
[]: 0.9883244206773618
[]: print (classification_report(y_test, pred_rf))
                  precision
                               recall f1-score
                                                   support
               0
                       0.99
                                 0.99
                                           0.99
                                                      5928
               1
                       0.99
                                 0.99
                                           0.99
                                                      5292
                                           0.99
                                                     11220
        accuracy
       macro avg
                                 0.99
                                           0.99
                                                     11220
                       0.99
    weighted avg
                       0.99
                                 0.99
                                           0.99
                                                     11220
[]: def predict_news(text):
         text_vectorized = vectorization.transform([filtering(text)])
         prediction = RF.predict(text_vectorized)
         if prediction == 1:
             return "This news is likely true."
         else:
             return "This news is likely fake."
     user_input = input("Enter news text: ")
     result = predict_news(user_input)
     print(result)
```

Enter news text: washington This news is likely fake.