```
**## The Code runtime is around 1 hours.**
```

#### Data sources:

https://www.kaggle.com/datasets/ruchi798/source-based-news-classification (https://www.kaggle.com/datasets/ruchi798/source-based-news-classification)

https://www.kaggle.com/datasets/clmentbisaillon/fake-and-real-news-dataset?resource=download (https://www.kaggle.com/datasets/clmentbisaillon/fake-and-real-news-dataset?resource=download)

https://www.kaggle.com/datasets/mrisdal/fake-news (https://www.kaggle.com/datasets/mrisdal/fake-news)

## **Import Libraries**

```
In [*]: import pandas as pd
    import numpy as np
    import seaborn as sns
    from sklearn.model_selection import train_test_split
    from sklearn.feature_extraction.text import TfidfVectorizer

from sklearn.metrics import accuracy_score
    from sklearn.metrics import confusion_matrix
    from sklearn.metrics import classification_report

from sklearn.linear_model import PassiveAggressiveClassifier
    from sklearn.ensemble import GradientBoostingClassifier
    from sklearn import svm
    from sklearn.linear_model import LogisticRegression
    import seaborn as sns

import matplotlib.pyplot as plt
    from sklearn.metrics import plot_confusion_matrix
```

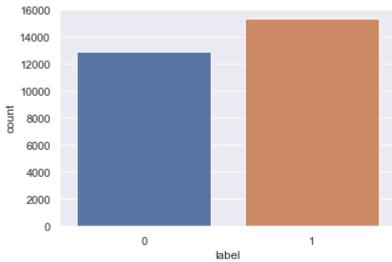
## Import Data Sets

Column: title, text, label

```
In [2]: df1=pd.read csv("Fake.csv")
        df1.drop(axis=0, index=df1.index[-15000:], inplace=True)
        df1 = df1[['title', 'text']]
        df1['label'] = 0
        print(df1.info())
        # print(df1.head(3))
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 8481 entries, 0 to 8480
        Data columns (total 3 columns):
             Column Non-Null Count Dtype
             -----
             title 8481 non-null
                                    object
                    8481 non-null
                                    object
         1
             text
                                    int64
         2
             label
                    8481 non-null
        dtypes: int64(1), object(2)
        memory usage: 198.9+ KB
        None
In [3]: df2=pd.read csv("True.csv")
        df2.drop(axis=0, index=df2.index[-10000:], inplace=True)
        df2 = df2[['title', 'text']]
        df2['label'] = 1
        print(df2.info())
        # print(df2.head(3))
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 11417 entries, 0 to 11416
        Data columns (total 3 columns):
         #
             Column Non-Null Count Dtype
             ----- ------ ----
             title 11417 non-null object
         0
                    11417 non-null object
         1
             text
             label 11417 non-null int64
         2
        dtypes: int64(1), object(2)
        memory usage: 267.7+ KB
        None
```

```
In [4]: df3=pd.read csv("news.csv")
        df3 = df3[['title', 'text', 'label']]
        df3['label'] = df3['label'].map({'FAKE':0 ,'REAL':1})
        print(df3.info())
        # print(df3.head(3))
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 6335 entries, 0 to 6334
        Data columns (total 3 columns):
             Column Non-Null Count Dtype
         0
             title
                    6335 non-null
                                    object
                    6335 non-null
         1
             text
                                    object
         2
             label
                    6335 non-null
                                    int64
        dtypes: int64(1), object(2)
        memory usage: 148.6+ KB
        None
In [5]: df4=pd.read_csv("news_articles.csv")
        df4 = df4.dropna()
        df4 = df4[['title', 'text', 'label']]
        df4['label'] = df4['label'].map({'Fake':0 ,'Real':1})
        print(df4.info())
        # print(df4.head(3))
        <class 'pandas.core.frame.DataFrame'>
        Int64Index: 2045 entries, 0 to 2045
        Data columns (total 3 columns):
             Column Non-Null Count Dtype
             _____
         0
             title
                    2045 non-null
                                    object
         1
             text
                    2045 non-null
                                    object
         2
             label
                    2045 non-null
                                    int64
        dtypes: int64(1), object(2)
        memory usage: 63.9+ KB
        None
```

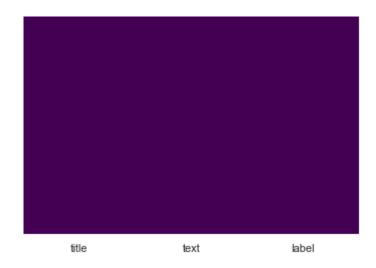
```
In [6]: # concatenating df1, df2, df3, df4 to df
        df = pd.concat([df1,df2,df3, df4], axis=0)
        print(df.info())
        print('\nTotal of Fake data:', (df['label'] == 0).sum())
        print('Total of Real data:', (df['label'] == 1).sum())
        <class 'pandas.core.frame.DataFrame'>
        Int64Index: 28278 entries, 0 to 2045
        Data columns (total 3 columns):
         #
             Column Non-Null Count Dtype
                     -----
             title
                     28278 non-null object
         1
                     28278 non-null object
             text
         2
                     28278 non-null int64
             label
        dtypes: int64(1), object(2)
        memory usage: 883.7+ KB
        None
        Total of Fake data: 12936
        Total of Real data: 15342
In [7]: sns.set theme(style="darkgrid")
        ax = sns.countplot(x='label', data=df)
           16000
           14000
```



#### Check null value

```
In [8]: sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='viridis')
```

Out[8]: <AxesSubplot:>



```
In [9]: df.isnull().sum()
Out[9]: title  0
    text  0
    label  0
    dtype: int64
```

# **Set Up Train/Test Data**

```
In [10]: labels = df['label']
    df = df.drop(['label'], axis=1)

In [11]: X_train,X_test,y_train,y_test=train_test_split(df['text'], labels, test_size=0.2,

In [12]: print("X_train shape: ", X_train.shape)
    print("X_test shape: ", X_test.shape)

    X_train shape: (22622,)
    X test shape: (5656,)
```

#### **Transform the Data Set**

```
In [13]: # Init the TfidfVectorizer function
    vectorizer = TfidfVectorizer()

# Transform training data set
    X_train = vectorizer.fit_transform(X_train)

# Transform testing data set
    X_test = vectorizer.transform(X_test)
```

### **Models**

Passive Aggressive Classifier

```
In [14]: # Set paramaters
         pac = PassiveAggressiveClassifier(max iter=50, C = 0.5)
         # training model
         pac.fit(X_train, y_train)
         # predicting
         pac_pred = pac.predict(X_test)
         # Display the accuracy score
         pac score = accuracy score(y test, pac pred)
         print('The confusion matrix is: ')
         print(confusion_matrix(y_test, pac_pred))
         print(f'\nThe accuracy score is: {round((pac_score)*100, 2)}%')
         print('\nThe classification report is:\n', classification_report(y_test,pac_pred)
         The confusion matrix is:
         [[2442 139]
          [ 132 2943]]
         The accuracy score is: 95.21%
         The classification report is:
                        precision recall f1-score
                                                         support
                    0
                             0.95
                                       0.95
                                                 0.95
                                                           2581
                    1
                             0.95
                                       0.96
                                                 0.96
                                                           3075
                                                 0.95
                                                           5656
             accuracy
                             0.95
                                       0.95
                                                 0.95
            macro avg
                                                           5656
                                                 0.95
         weighted avg
                             0.95
                                       0.95
                                                           5656
```

**Gradient Boosting Classifier** 

```
In [15]: #from sklearn.model selection import GridSearchCV
         gbrt = GradientBoostingClassifier(learning_rate=0.1, max_depth= 5, max_features=
         gbrt.fit(X_train, y_train)
         # predicting
         gbrt_pred = gbrt.predict(X_test)
         # Display the accuracy score
         gbrt_score = accuracy_score(y_test, gbrt_pred)
         print('The confusion matrix is: ')
         print(confusion_matrix(y_test, gbrt_pred))
         print(f'\nThe accuracy score is: {round((gbrt_score)*100, 2)}%')
         print('\nThe classification report is:\n', classification_report(y_test,gbrt_pred
         The confusion matrix is:
         [[2351 230]
          [ 208 2867]]
         The accuracy score is: 92.26%
         The classification report is:
                        precision
                                      recall f1-score
                                                         support
                    0
                            0.92
                                       0.91
                                                 0.91
                                                           2581
                            0.93
                                       0.93
                    1
                                                 0.93
                                                           3075
```

0.92

0.92

0.92

5656

5656

5656

Logistic Regression

0.92

0.92

0.92

0.92

accuracy

macro avg

weighted avg

	precision	recall	f1-score	support
0	0.92	0.95	0.94	2581
1	0.96	0.93	0.95	3075
accuracy			0.94	5656
macro avg	0.94	0.94	0.94	5656
weighted avg	0.94	0.94	0.94	5656

Support Vector Classifier

```
In [17]: SVM = svm.SVC(C=1.9, kernel='linear')
SVM.fit(X_train, y_train)

svm_prediction = SVM.predict(X_test)

# Display the accuracy score
svm_score = accuracy_score(y_test, svm_prediction)

print('The confusion matrix is: ')
print(confusion_matrix(y_test, svm_prediction))
print(f'\nThe accuracy score is: {round((svm_score)*100, 2)}%')
print('\nThe classification report is:\n', classification_report(y_test,svm_prediction))
The confusion matrix is:
[[2471 110]
[ 152 2923]]
```

The accuracy score is: 95.37%

The classification report is:

	precision	recall	f1-score	support
0	0.94	0.96	0.95	2581
1	0.96	0.95	0.96	3075
accuracy			0.95	5656
macro avg	0.95	0.95	0.95	5656
weighted avg	0.95	0.95	0.95	5656