STEPS

- 1. Importing Modules
- 2. Loading and Preprocessing the Dataset: Reading Data:

Loads a CSV file containing news articles and their labels (fake or reliable). Stores data in a pandas DataFrame.

Handling Missing Values:

Fills null values with empty strings.

Dropping Unnecessary Columns:

Keeps only the 'text' and 'label' columns.

Stemming:

Applies stemming (reducing words to their base forms) using a SnowballStemmer.

3. Building the Model: Splitting Data:

Divides data into training and testing sets.

Text Vectorization:

Uses TfidfVectorizer to convert text into numerical vectors for model training.

Model Training:

Trains a decision tree classifier on the training data.

Evaluation:

Assesses model accuracy on the testing set.

Saving Models:

Saves the trained model and vectorizer for future use.

4. Making Predictions: Defining Prediction Function:

Creates a function 'fake_news(news)' to predict the reliability of new articles. Stems the input news text.

Transforms text into a vector using the saved vectorizer.

Generates a prediction using the loaded model.

Example Prediction:

Passes a sample news article to the prediction function.
Classifies it as either 'reliable' or 'unreliable' based on the model's output.

In [1]: #import necessary modules

import pandas as pd #for data manipulation
from nltk.corpus import stopwords #natural language toolkit - collection of natural language dataset (such as writter
from nltk.stem.porter import PorterStemmer #porter stemmer for basic word stemming (removes suffixes from root words)
import re #used for regular expressions operations primarily for text cleaning
import nltk

from nltk.stem import SnowballStemmer #snowbal stemmer (aka porter2 stemming algorithm as it can handle more languages from sklearn.model_selection import train_test_split #provides tools for splitting data into training and testing sets from sklearn.feature_extraction.text import TfidfVectorizer #contains the TfidfVectorizer for text feature extraction from sklearn.tree import DecisionTreeClassifier #for model building and classification import pickle #used for saving and loading models and vectorizers



In [2]: #locate dataset and store it in a variable

dataset_path = 'C:\\Users\\DISHA\\Downloads\\News Authenticator\\train.csv'
df = pd.read_csv(dataset_path)

In [3]: #read top 10 values

df.head(10)

Out[3]:

	id	title	author	text	label
0	0	House Dem Aide: We Didn't Even See Comey's Let	Darrell Lucus	House Dem Aide: We Didn't Even See Comey's Let	1
1	1	FLYNN: Hillary Clinton, Big Woman on Campus	Daniel J. Flynn	Ever get the feeling your life circles the rou	0
2	2	Why the Truth Might Get You Fired	Consortiumnews.com	Why the Truth Might Get You Fired October 29, \dots	1
3	3	15 Civilians Killed In Single US Airstrike Hav	Jessica Purkiss	Videos 15 Civilians Killed In Single US Airstr	1
4	4	Iranian woman jailed for fictional unpublished	Howard Portnoy	Print \nAn Iranian woman has been sentenced to	1
5	5	Jackie Mason: Hollywood Would Love Trump if He	Daniel Nussbaum	In these trying times, Jackie Mason is the Voi	0
6	6	Life: Life Of Luxury: Elton John's 6 Favorite	NaN	Ever wonder how Britain's most iconic pop pian	1
7	7	Benoît Hamon Wins French Socialist Party's Pre	Alissa J. Rubin	PARIS — France chose an idealistic, traditi	0
8	8	Excerpts From a Draft Script for Donald Trump'	NaN	Donald J. Trump is scheduled to make a highly	0
9	9	A Back-Channel Plan for Ukraine and Russia, Co	Megan Twohey and Scott Shane	A week before Michael T. Flynn resigned as nat	0

In [4]: #read dataset description

df.describe()

Out[4]:

	id	label
count	20800.000000	20800.000000
mean	10399.500000	0.500625
std	6004.587135	0.500012
min	0.000000	0.000000
25%	5199.750000	0.000000
50%	10399.500000	1.000000
75%	15599.250000	1.000000
max	20799.000000	1.000000

```
In [5]: #get dataset info
        df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 20800 entries, 0 to 20799
        Data columns (total 5 columns):
            Column Non-Null Count Dtype
             id
                     20800 non-null int64
            title 20242 non-null object
            author 18843 non-null object
            text 20761 non-null object
            label 20800 non-null int64
        dtypes: int64(2), object(3)
        memory usage: 812.6+ KB
In [6]: #to find no. of null values of each columns
        df.isnull().sum()
Out[6]: id
                     0
        title
                   558
        author
                 1957
        text
                    39
        label
        dtype: int64
In [7]: #to fill the null value with one empty string
        df=df.fillna('')
```

```
In [8]: #now all the null values are filled
          df.isnull().sum()
 Out[8]: id
                      0
          title
          author
          text
          label
          dtype: int64
 In [9]: #to see all the columns
          df.columns
 Out[9]: Index(['id', 'title', 'author', 'text', 'label'], dtype='object')
In [10]: #drop useless columns and keep columns like text and label from axis no. 1
          df=df.drop(['id', 'title', 'author'],axis=1)
In [11]: #to verify if it worked or not
          df.head()
Out[11]:
                                                     text label
           0 House Dem Aide: We Didn't Even See Comey's Let...
                     Ever get the feeling your life circles the rou...
           1
                 Why the Truth Might Get You Fired October 29, ...
           2
           3
                    Videos 15 Civilians Killed In Single US Airstr...
                Print \nAn Iranian woman has been sentenced to...
```

```
In [12]: #initialize porter stemmer
         port stem = PorterStemmer()
In [13]: #display the Porter Stemmer object
         port stem
Out[13]: <PorterStemmer>
In [14]: #stem a sample word
         port stem.stem("Hi * % %@@@")
Out[14]: 'hi * % %@@@'
In [15]: #define a function for stemming
         def stemming(content):
             con=re.sub('[^a-zA-Z]', ' ', content) #remove non-alphabetic characters
             con=con.lower() #convert to Lowercase
             con=con.split() #tokenize the text - process of breaking down a sequence of text into individual elements, called
             con=[port stem.stem(word) for word in con if not word in stopwords.words('english')] #stem each word and remove s
             con=' '.join(con) #join the stemmed words
             return con
In [16]: #test the stemming function
         stemming('Hi this is Disha')
Out[16]: 'hi disha'
In [17]: #initialize Snowball Stemmer for English
         stemmer = SnowballStemmer(language='english')
```

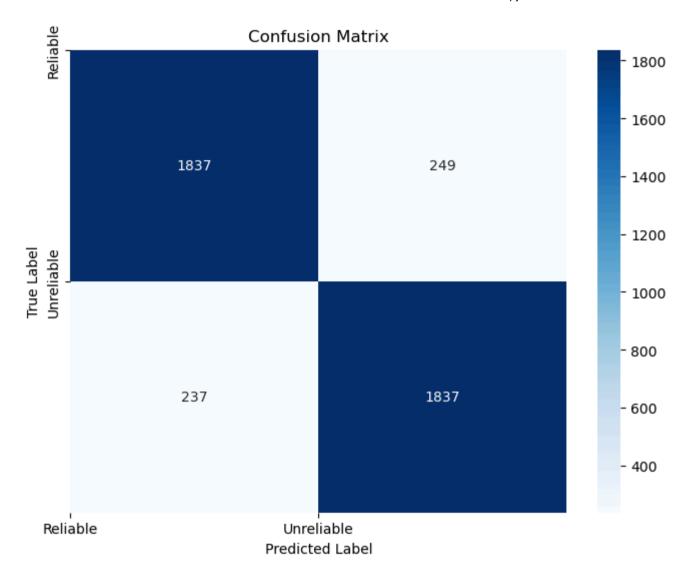
```
In [18]: #apply stemming to the 'text' column (takes time)
         df['text'] = df['text'].apply(lambda x: ' '.join([stemmer.stem(word) for word in x.split()]))
In [19]: #get features and target variable
         x=df['text']
         y=df['label']
In [20]: #check the shape of target variables
         x.shape
         y.shape
Out[20]: (20800,)
In [21]: #split the data into train and test sets
         x train , x test , y train, y test = train test split(x, y, test size=0.20)
In [22]: #initialize TF-IDF Vectorizer
         vect=TfidfVectorizer()
In [23]: x train=vect.fit transform(x train) #fit and transform the training data
         x test=vect.transform(x test) #transform the test data
In [24]: #check the shape of test data
         x test.shape
Out[24]: (4160, 155510)
```

```
In [25]: #initialize Decision Tree Classifier
         model=DecisionTreeClassifier()
In [26]: #fit the model on the training data
         model.fit(x train, y train)
Out[26]: DecisionTreeClassifier()
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
         #make predictions on the test data
In [27]:
         prediction=model.predict(x test)
In [28]: #display the predictions
         prediction
Out[28]: array([0, 1, 0, ..., 0, 0, 0], dtype=int64)
In [29]: #calculate the accuracy of the model on the test data
         model.score(x_test, y_test)
Out[29]: 0.8831730769230769
```

```
In [33]: import seaborn as sns
    from sklearn.metrics import confusion_matrix

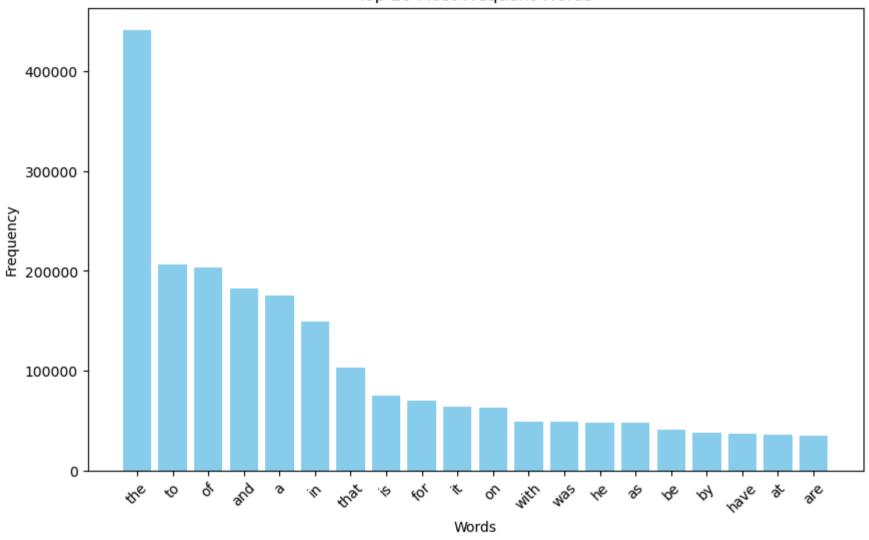
# Compute confusion matrix
cm = confusion_matrix(y_test, prediction)

# Plot confusion matrix using Seaborn's heatmap
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, cmap='Blues', fmt='g') # 'g' ensures integer format for annotations
plt.title("Confusion Matrix")
plt.xticks(range(2), ['Reliable', 'Unreliable'])
plt.yticks(range(2), ['Reliable', 'Unreliable'])
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.show()
```



```
In [34]: from collections import Counter
         # Process a subset of data
         subset df = df.sample(frac=0.5) # Adjust fraction as needed
         # Tokenize all texts and count word frequencies
         words = ' '.join(subset df['text']).split()
         word counts = Counter(words)
         # Get top N most frequent words
         top n = 20
         top words = dict(word counts.most common(top n))
         # Plot bar plot for the top N most frequent words
         plt.figure(figsize=(10, 6))
         plt.bar(top words.keys(), top words.values(), color='skyblue')
         plt.xlabel('Words')
         plt.ylabel('Frequency')
         plt.title('Top {} Most Frequent Words'.format(top n))
         plt.xticks(rotation=45)
         plt.show()
```

Top 20 Most Frequent Words



```
In [35]: #save the vectorizer
pickle.dump(vect, open('vector.pkl', 'wb'))
```

```
In [36]: #save the model
         pickle.dump(model, open('model.pkl', 'wb'))
In [37]: #load the vectorizer
         vector form=pickle.load(open('vector.pkl', 'rb'))
In [38]: # Load the model
         load model=pickle.load(open('model.pkl', 'rb'))
In [39]: #define a function to predict fake news
         def fake news(news):
             news=stemming(news) #perform stemming on the news
             input data=[news]
             vector form1=vector form.transform(input_data) #transform the input data using the loaded vectorizer
             prediction = load model.predict(vector form1) #predict using the Loaded model
             return prediction
In [40]: #test the function with a sample news article
         val=fake news("""In these trying times, Jackie Mason is the Voice of Reason. [In this week's exclusive clip for Breitk
In [41]: #print the prediction
         if val==[0]:
             print('reliable')
         else:
             print('unreliable')
         reliable
```

Accuracy: 0.8831730769230769 Precision: 0.8806327900287632 Recall: 0.8857280617164899 F1-score: 0.8831730769230769

```
In [43]: #Trying other Machine Learning Algorithms for better Results

#Using Random Forest Classifier
from sklearn.ensemble import RandomForestClassifier

# Initialize Random Forest Classifier
rf_model = RandomForestClassifier()

# Fit the model on the training data
rf_model.fit(x_train, y_train)

# Make predictions on the test data
rf_prediction = rf_model.predict(x_test)

# Calculate the accuracy of the model
rf_accuracy = rf_model.score(x_test, y_test)

# Print the accuracy
print("Random Forest Classifier Accuracy:", rf_accuracy)
```

Random Forest Classifier Accuracy: 0.9221153846153847

```
In [44]: #Ensemble Methods (Using AdaBoost Classifier)
         from sklearn.ensemble import AdaBoostClassifier
         # Initialize AdaBoost Classifier
         adaboost model = AdaBoostClassifier()
         # Fit the model on the training data
         adaboost model.fit(x train, y train)
         # Make predictions on the test data
         adaboost prediction = adaboost model.predict(x test)
         # Calculate the accuracy of the model
         adaboost accuracy = adaboost model.score(x test, y test)
         # Print the accuracy
         print("AdaBoost Classifier Accuracy:", adaboost accuracy)
         C:\Users\DISHA\anaconda3\lib\site-packages\sklearn\ensemble\ weight boosting.py:519: FutureWarning: The SAMME.R algo
         rithm (the default) is deprecated and will be removed in 1.6. Use the SAMME algorithm to circumvent this warning.
           warnings.warn(
         AdaBoost Classifier Accuracy: 0.9394230769230769
In [ ]: # Using a Support Vector Machine (SVM)
         from sklearn.svm import SVC
         svc model = SVC()
         svc model.fit(x train, y train)
         svc prediction = svc model.predict(x test)
         svc accuracy = svc_model.score(x_test, y_test)
         print("SVC Accuracy:", svc accuracy)
In [ ]:
```