

1. Text Classification of News Articles using NLP.

Article Id – Article id unique given to the record Article – Text of the header and article Category – Category of the article (tech, business, sport, entertainment, politics) Consider BBC News as corpus for implementing question 1

```
In [2]: import numpy as np  
import pandas as pd
```

Load the Dataset

```
In [3]: #Upload the Data:  
df_BBCNews = pd.read_csv(r'C:\Users\raju\StockMktPrediction/BBC News.csv')
```

```
In [4]: print(df_BBCNews.head())
```

	ArticleId	Text	Category	\
0	1833	worldcom ex-boss launches defence lawyers defe...	business	
1	154	german business confidence slides german busin...	business	
2	1101	bbc poll indicates economic gloom citizens in ...	business	
3	1976	lifestyle governs mobile choice faster bett...	tech	
4	917	enron bosses in \$168m payout eighteen former e...	business	

text

```
0 worldcom exboss launches defence lawyers defen...
1 german business confidence slides german busin...
2 bbc poll indicates economic gloom citizens in ...
3 lifestyle governs mobile choice faster better ...
4 enron bosses in 168m payout eighteen former en...
```

```
In [5]: import csv  
  
filename = r'C:\Users\raju\StockMktPrediction/BBC News.csv'  
unique_ids = set()  
  
with open(filename, 'r') as csvfile:  
    reader = csv.DictReader(csvfile)
```

```
for row in reader:  
    id = row['ArticleId']  
    if id not in unique_ids:  
        unique_ids.add(id)  
  
print(unique_ids)
```

```
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Tokenization and Cleaning with NLTK

```
In [6]: import nltk  
from nltk.corpus import stopwords  
nltk.download('stopwords')
```

```
[nltk_data] Downloading package stopwords to  
[nltk_data]      C:\Users\raju\AppData\Roaming\nltk_data...  
[nltk_data] Package stopwords is already up-to-date!
```

```
Out[6]: True
```

Filter out STOP Words

```
In [7]: # Define the stop words to be removed  
stop_words = set(stopwords.words('english'))
```

```
In [8]: # Define a function to remove stop words from a string  
def remove_stop_words(text):  
    Text = ' '.join([word for word in text.split() if word.lower() not in stop_words])  
    return text
```

```
In [9]: # Apply the function to the text column  
df_BBCNews['Text'] = df_BBCNews['Text'].apply(remove_stop_words)
```

```
In [10]: df_BBCNews.to_csv(r'C:\Users\raju\StockMktPrediction/BBC News.csv', index=False)
```

```
In [11]: import string  
import re  
  
# Load the CSV file into a Pandas DataFrame  
df = pd.read_csv(r'C:\Users\raju\StockMktPrediction/BBC News.csv')
```

```
In [12]: def clean_text(text):  
    # Remove punctuation  
    text = text.translate(str.maketrans('', '', string.punctuation))  
    # Remove extra white spaces
```

```
text = re.sub('\s+', ' ', text)
# Convert to lowercase
text = text.lower()
return text
```

```
In [13]: df['text'] = df['Text'].apply(clean_text)
```

```
In [14]: df.to_csv(r'C:\Users\raju\StockMktPrediction/BBC News.csv', index=False)
```

```
In [15]: from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score
import pandas as pd
```

```
In [16]: # Load the labeled news articles
df = pd.read_csv(r'C:\Users\raju\StockMktPrediction/BBC News.csv')
```

```
In [17]: df = pd.read_csv(r'C:\Users\raju\StockMktPrediction/BBC News.csv')
```

```
In [18]: X_train, X_test, y_train, y_test = train_test_split(df['text'], df['Category'], test_size=0.2)
```

```
In [19]: # Create a TF-IDF vectorizer
vectorizer = TfidfVectorizer()
```

```
In [20]: # Fit the vectorizer on the training data and transform the data into vectors
X_train_vec = vectorizer.fit_transform(X_train)
X_test_vec = vectorizer.transform(X_test)
```

```
In [21]: # Train a SVM classifier on the training data
clf = SVC(kernel='linear')
clf.fit(X_train_vec, y_train)
```

```
Out[21]: SVC(kernel='linear')
```

```
In [22]: # Predict the categories of the test data using the trained classifier
y_pred = clf.predict(X_test_vec)
```

```
In [23]: # Evaluate the performance of the classifier  
accuracy = accuracy_score(y_test, y_pred)  
print('Accuracy:', accuracy)
```

Accuracy: 0.9899328859060402

```
In [ ]:
```