```
#Importing Libraries

import pickle
import pandas as pd
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.feature_selection import chi2
import numpy as np
```

In [2]:
```
#Accessing document uploaded

path_df = "/content/drive/My Drive/Pickles/News_dataset.pickle"

with open(path_df, 'rb') as data:
    df = pickle.load(data)
```

In [3]:
```
#checking data

df.head()
```

Out[3]:

	File_Name	Content	Category	Complete_Filename	id	News_length
0	001.txt	Ad sales boost Time Warner profit\r\n\r\nQuart...	business	001.txt-business	1	2569
1	002.txt	Dollar gains on Greenspan speech\r\n\r\nThe do...	business	002.txt-business	1	2257
2	003.txt	Yukos unit buyer faces loan claim\r\n\r\nThe o...	business	003.txt-business	1	1557
3	004.txt	High fuel prices hit BA's profits\r\n\r\nBriti...	business	004.txt-business	1	2421
4	005.txt	Pernod takeover talk lifts Domecq\r\n\r\nShare...	business	005.txt-business	1	1575

In [ ]:
```
#Chcking article

df.loc[1]['Content']
```

```
('Dollar gains on Greenspan speech\r\n\r\nThe dollar has hit its highest level against the euro in almost three months after the Federal Reserve head said the
US trade deficit is set to stabilise.\r\n\r\nAnd Alan Greenspan highlighted the US government\'s willingness to curb spending and rising household savings as
factors which may help to reduce it. In late trading in New York, the dollar reached $1.2871 against the euro, from $1.2974 on Thursday. Market concerns about
the deficit has hit the greenback in recent months. On Friday, Federal Reserve chairman Mr Greenspan\'s speech in London ahead of the meeting of G7 finance mi
nisters sent the dollar higher after it had earlier tumbled on the back of worse-than-expected US jobs data. "I think the chairman\'s taking a much more sangu
ine view on the current account deficit than he\'s taken for some time," said Robert Sinche, head of currency strategy at Bank of America in New York. "He\'s
taking a longer-term view, laying out a set of conditions under which the current account deficit can improve this year and next."\r\n\r\nWorries about the de
ficit concerns about China do, however, remain. China\'s currency remains pegged to the dollar and the US currency\'s sharp falls in recent months have theref
ore made Chinese export prices highly competitive. But calls for a shift in Beijing\'s policy have fallen on deaf ears, despite recent comments in a major Chi
nese newspaper that the "time is ripe" for a loosening of the peg. The G7 meeting is thought unlikely to produce any meaningful movement in Chinese policy. In
the meantime, the US Federal Reserve\'s decision on 2 February to boost interest rates by a quarter of a point - the sixth such move in as many months - has o
pened up a differential with European rates. The half-point window, some believe, could be enough to keep US assets looking more attractive, and could help pr
op up the dollar. The recent falls have partly been the result of big budget deficits, as well as the US\'s yawning current account gap, both of which need to
be funded by the buying of US bonds and assets by foreign firms and governments. The White House will announce its budget on Monday, and many commentators bel
ieve the deficit will remain at close to half a trillion dollars.'
```

1. Text cleaning and preparation

In [ ]:
```
#Text cleaning

df['Content_Parsed_1'] = df['Content'].str.replace("\r", " ")
df['Content_Parsed_1'] = df['Content_Parsed_1'].str.replace("\n", " ")
df['Content_Parsed_1'] = df['Content_Parsed_1'].str.replace("    ", " ")
df['Content_Parsed_1'] = df['Content_Parsed_1'].str.replace('"', '')
```

In [ ]:
```
#Text preparation

df['Content_Parsed_2'] = df['Content_Parsed_1'].str.lower()         #all to lower case

punctuation_signs = list("?:!.,;")                                 #remove punctuations
df['Content_Parsed_3'] = df['Content_Parsed_2']

for punct_sign in punctuation_signs:
    df['Content_Parsed_3'] = df['Content_Parsed_3'].str.replace(punct_sign, '')

df['Content_Parsed_4'] = df['Content_Parsed_3'].str.replace("'s", "")       #remove possessive pronouns
```

Use any 1 method for Lemmatization

In [ ]:
```
#Stemming and Lemmatization

nltk.download('punkt')
nltk.download('wordnet')

nltk.download('averaged_perceptron_tagger')
from nltk.corpus import wordnet
```

```
[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data]   Unzipping tokenizers/punkt.zip.
[nltk_data] Downloading package wordnet to /root/nltk_data...
[nltk_data]   Unzipping corpora/wordnet.zip.
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]     /root/nltk_data...
[nltk_data]   Unzipping taggers/averaged_perceptron_tagger.zip.
```

1st method for lemmatization

In [ ]:
```
#Stemming and Lemmatization

wordnet_lemmatizer = WordNetLemmatizer()
nrows = len(df)
lemmatized_text_list = []

for row in range(0, nrows):

    # Create an empty list containing lemmatized words
    lemmatized_list = []

    # Save the text and its words into an object
    text = df.loc[row]['Content_Parsed_4']
    text_words = text.split(" ")

    # Iterate through every word to lemmatize
    for word in text_words:
        lemmatized_list.append(wordnet_lemmatizer.lemmatize(word, pos="v"))

    # Join the list
    lemmatized_text = " ".join(lemmatized_list)

    # Append to the list containing the texts
    lemmatized_text_list.append(lemmatized_text)

df['Content_Parsed_5'] = lemmatized_text_list
```

In [ ]:
```
df['Content_Parsed_5']
```

Out[ ]:
```
0       ad sales boost time warner profit quarterly pr...
1       dollar gain on greenspan speech the dollar hav...
2       yukos unit buyer face loan claim the owners of...
3       high fuel price hit ba profit british airways ...
4       pernod takeover talk lift domecq share in uk d...
                              ...
2220    bt program to beat dialler scam bt be introduc...
2221    spam e-mail tempt net shoppers computer users ...
2222    be careful how you code a new european directi...
2223    us cyber security chief resign the man make su...
```

```
2224    lose yourself in online game online role play ...
Name: Content_Parsed_5, Length: 2225, dtype: object
```

2nd method for lemmatization

In [ ]:
```
lemmatizer = WordNetLemmatizer()

# function to convert nltk tag to wordnet tag
def nltk_tag_to_wordnet_tag(nltk_tag):
    if nltk_tag.startswith('J'):
        return wordnet.ADJ
    elif nltk_tag.startswith('V'):
        return wordnet.VERB
    elif nltk_tag.startswith('N'):
        return wordnet.NOUN
    elif nltk_tag.startswith('R'):
        return wordnet.ADV
    else:
        return None

def lemmatize_sentence(sentence):
    #tokenize the sentence and find the POS tag for each token
    nltk_tagged = nltk.pos_tag(nltk.word_tokenize(sentence))
    #tuple of (token, wordnet_tag)
    wordnet_tagged = map(lambda x: (x[0], nltk_tag_to_wordnet_tag(x[1])), nltk_tagged)
    lemmatized_sentence = []
    for word, tag in wordnet_tagged:
        if tag is None:
            #if there is no available tag, append the token as is
            lemmatized_sentence.append(word)
        else:
            #else use the tag to lemmatize the token
            lemmatized_sentence.append(lemmatizer.lemmatize(word, tag))
    return " ".join(lemmatized_sentence)
```

```
nrows = len(df)
lemmatized_text_list = []

for row in range(0, nrows):
    lemmatized_text = lemmatize_sentence(df.loc[row]['Content_Parsed_4'])
    lemmatized_text_list.append(lemmatized_text)

df['Content_Parsed_5'] = lemmatized_text_list
```

In [ ]:
```
df['Content_Parsed_5']
```

Use any 1 method for stop word

In [ ]:
```
#Downloading

nltk.download('stopwords')
```

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data]   Unzipping corpora/stopwords.zip.
```

Out[ ]:True

In [ ]:
```
#Removing stop words

stop_words = list(stopwords.words('english'))
```

1st Method

In [ ]:
```
df['Content_Parsed_6'] = df['Content_Parsed_5']

for stop_word in stop_words:

    regex_stopword = r"\b" + stop_word + r"\b"
    df['Content_Parsed_6'] = df['Content_Parsed_6'].str.replace(regex_stopword, '')
```

In [ ]:
```
df.loc[5]['Content_Parsed_6']
```

```
C'japan narrowly escape recession japan economy teeter   brink   technical recession   three months  september figure show revise figure indicate growth   01%
 -   similar-sized contraction   previous quarter   annual basis  data suggest annual growth   02% suggest  much  hesitant recovery   previously  think  commo
 n technical definition   recession  two successive quarter  negative growth  government  keen  play   worry implications   data  maintain  view  japan econom
 y remain   minor adjustment phase   upward climb    monitor developments carefully say economy minister heizo takenaka    face   strengthen yen make export l
 ess competitive  indications  weaken economic condition ahead observers  less sanguine  paint  picture   recovery much patchier  previously think say paul sh
 eard economist  lehman brothers  tokyo improvements   job market apparently  yet  fee   domestic demand  private consumption   02%   third quarter'
```

2nd Method

In [ ]:
```
stop_list_final=[]
nrows = len(df)
stopwords_english = stopwords.words('english')

for row in range(0, nrows):

    # Create an empty list containing no stop words
    stop_list = []

    # Save the text and its words into an object
    text = df.loc[row]['Content_Parsed_5']
    text_words = text.split(" ")

    # Iterate through every word to remove stopwords
    for word in text_words:
        if (word not in stopwords_english):
          stop_list.append(word)

    # Join the list
    stop_text = " ".join(stop_list)

    # Append to the list containing the texts
    stop_list_final.append(stop_text)

df['Content_Parsed_6'] = stop_list_final
```

In [ ]:
```
df.loc[5]['Content_Parsed_6']
```

```
C'japan narrowly escape recession japan economy teeter brink technical recession three months september figure show revise figure indicate growth 01% - simila
 r-sized contraction previous quarter annual basis data suggest annual growth 02% suggest much hesitant recovery previously think common technical definition
 recession two successive quarter negative growth government keen play worry implications data maintain view japan economy remain minor adjustment phase upwar
 d climb monitor developments carefully say economy minister heizo takenaka face strengthen yen make export less competitive indications weaken economic condi
 tion ahead observers less sanguine paint picture recovery much patchier previously think say paul sheard economist lehman brothers tokyo improvements job mar
 ket apparently yet fee domestic demand private consumption 02% third quarter'
```

In [ ]:
```
#Checking data

df.head(1)
```

O...

	File_Name	Content	Category	Complete_Filename	id	News_length	Content_Parsed_1	Content_Parsed_2	Content_Parsed_3	Content_Parsed_4	Content_Parsed_5	Content_Parsed_6
0	001.txt	Ad sales boost Time Warner profit\r\n\r\nQuart...	business	001.txt-business	1	2569	Ad sales boost Time Warner profit Quarterly pr...	ad sales boost time warner profit quarterly pr...	ad sales boost time warner profit quarterly pr...	ad sales boost time warner profit quarterly pr...	ad sales boost time warner profit quarterly pr...	ad sales boost time warner profit quarterly pr...

In [ ]:
```
#Removing the old content_parsed columns

list_columns = ["File_Name", "Category", "Complete_Filename", "Content", "Content_Parsed_6"]
df = df[list_columns]

df = df.rename(columns={'Content_Parsed_6': 'Content_Parsed'})
```

In [ ]:
```
df.head()
```

Out[ ]:

	File_Name	Category	Complete_Filename	Content	Content_Parsed
0	001.txt	business	001.txt-business	Ad sales boost Time Warner profit\r\n\r\nQuart...	ad sales boost time warner profit quarterly pr...
1	002.txt	business	002.txt-business	Dollar gains on Greenspan speech\r\n\r\nThe do...	dollar gain greenspan speech dollar hit hi...

	File_Name	Category	Complete_Filename	Content	Content_Parsed
2	003.txt	business	003.txt-business	Yukos unit buyer faces loan claim\r\n\r\nThe o...	yukos unit buyer face loan claim owners emba...
3	004.txt	business	004.txt-business	High fuel prices hit BA's profits\r\n\r\nBriti...	high fuel price hit ba profit british airways ...
4	005.txt	business	005.txt-business	Pernod takeover talk lifts Domecq\r\n\r\nShare...	pernod takeover talk lift domecq share uk dri...

2. Label coding

In [ ]:
```
#Generating new column for Category codes

category_codes = {
    'business': 0,
    'entertainment': 1,
    'politics': 2,
    'sport': 3,
    'tech': 4
}

# Category mapping
df['Category_Code'] = df['Category']
df = df.replace({'Category_Code':category_codes})
```

In [ ]:
```
df.head()
```

Out[ ]:

	File_Name	Category	Complete_Filename	Content	Content_Parsed	Category_Code
0	001.txt	business	001.txt-business	Ad sales boost Time Warner profit\r\n\r\nQuart...	ad sales boost time warner profit quarterly pr...	0
1	002.txt	business	002.txt-business	Dollar gains on Greenspan speech\r\n\r\nThe do...	dollar gain greenspan speech dollar hit hi...	0
2	003.txt	business	003.txt-business	Yukos unit buyer faces loan claim\r\n\r\nThe o...	yukos unit buyer face loan claim owners emba...	0
3	004.txt	business	004.txt-business	High fuel prices hit BA's profits\r\n\r\nBriti...	high fuel price hit ba profit british airways ...	0
4	005.txt	business	005.txt-business	Pernod takeover talk lifts Domecq\r\n\r\nShare...	pernod takeover talk lift domecq share uk dri...	0

3. Train - test split

In [ ]:
```
X_train, X_test, y_train, y_test = train_test_split(df['Content_Parsed'],
                                                    df['Category_Code'],
                                                    test_size=0.15,
                                                    random_state=8)
```

4. Text representation

TF-IDF Vectors

unigrams & bigrams corresponding to a particular category

In [ ]:
```
# Parameter election
ngram_range = (1,2)
min_df = 10
max_df = 1.
max_features = 300
```

In [ ]:
```
tfidf = TfidfVectorizer(encoding='utf-8',
                        ngram_range=ngram_range,
                        stop_words=None,
                        lowercase=False,
                        max_df=max_df,
                        min_df=min_df,
                        max_features=max_features,
                        norm='l2',
                        sublinear_tf=True)

features_train = tfidf.fit_transform(X_train).toarray()
labels_train = y_train
print(features_train.shape)

features_test = tfidf.transform(X_test).toarray()
labels_test = y_test
print(features_test.shape)
```

```
(1891, 300)
(334, 300)
```

In [ ]:
```
from sklearn.feature_selection import chi2
import numpy as np

for Product, category_id in sorted(category_codes.items()):
    features_chi2 = chi2(features_train, labels_train == category_id)
    indices = np.argsort(features_chi2[0])
    feature_names = np.array(tfidf.get_feature_names())[indices]
    unigrams = [v for v in feature_names if len(v.split(' ')) == 1]
    bigrams = [v for v in feature_names if len(v.split(' ')) == 2]
    print("# '{}' category:".format(Product))
    print("  . Most correlated unigrams:\n. {}".format('\n. '.join(unigrams[-5:])))
    print("  . Most correlated bigrams:\n. {}".format('\n. '.join(bigrams[-2:])))
    print("")
```

```
# 'business' category:
  . Most correlated unigrams:
. market
. price
. economy
. growth
. bank
  . Most correlated bigrams:
. last year
. year old

# 'entertainment' category:
  . Most correlated unigrams:
. tv
. music
. star
. award
. film
  . Most correlated bigrams:
. mr blair
. prime minister

# 'politics' category:
  . Most correlated unigrams:
. minister
. blair
. party
. election
. labour
  . Most correlated bigrams:
. prime minister
. mr blair

# 'sport' category:
  . Most correlated unigrams:
. win
. side
. game
. team
. match
  . Most correlated bigrams:
. say mr
. year old

# 'tech' category:
  . Most correlated unigrams:
```

```
. digital
. technology
. computer
. software
. users
  . Most correlated bigrams:
. year old
. say mr
```

In [ ]:
```
bigrams
```

Out[ ]:
```
['tell bbc', 'last year', 'prime minister', 'mr blair', 'year old', 'say mr']
```

Unigrams are more relevnat to the category as compared with bigrams

In [ ]:
```
# X_train
with open('/content/drive/My Drive/Pickles/X_train.pickle', 'wb') as output:
    pickle.dump(X_train, output)

# X_test
with open('/content/drive/My Drive/Pickles/X_test.pickle', 'wb') as output:
    pickle.dump(X_test, output)

# y_train
with open('/content/drive/My Drive/Pickles/y_train.pickle', 'wb') as output:
    pickle.dump(y_train, output)

# y_test
with open('/content/drive/My Drive/Pickles/y_test.pickle', 'wb') as output:
    pickle.dump(y_test, output)

# df
with open('/content/drive/My Drive/Pickles/df.pickle', 'wb') as output:
    pickle.dump(df, output)

# features_train
with open('/content/drive/My Drive/Pickles/features_train.pickle', 'wb') as output:
    pickle.dump(features_train, output)

# labels_train
with open('/content/drive/My Drive/Pickles/labels_train.pickle', 'wb') as output:
    pickle.dump(labels_train, output)

# features_test
with open('/content/drive/My Drive/Pickles/features_test.pickle', 'wb') as output:
    pickle.dump(features_test, output)

# labels_test
with open('/content/drive/My Drive/Pickles/labels_test.pickle', 'wb') as output:
    pickle.dump(labels_test, output)

# TF-IDF object
with open('/content/drive/My Drive/Pickles/tfidf.pickle', 'wb') as output:
    pickle.dump(tfidf, output)
```