```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")
import re
from bs4 import BeautifulSoup
from tqdm import tqdm
from nltk.stem import WordNetLemmatizer
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score,f1_score, confusion_matrix
from keras.preprocessing.text import Tokenizer
from keras.preprocessing.sequence import pad_sequences
from keras.layers import Dense , Input , LSTM , Embedding, Dropout , Activation, GRU, Flat
from keras.layers import Bidirectional, GlobalMaxPool1D
from keras.models import Model, Sequential
from keras.layers import Convolution1D
from keras import initializers, regularizers, constraints, optimizers, layers
df = pd.read_csv('food_review.csv',engine='python')
df.shape
     (40500, 3)
```

df.tail()

Text			Score	
ray	This is the second Beaba freezer tray	that we'	5	
fro	The other, 5-star, review, must be from	m a shil	1	
ned	Perfect for pills and capsules for me	ications	5	
cer	Sure, the first ingredient on this co	eal is w	2	
for A	I waited long and with bated breath for	Amazon	5	

```
df = df.drop(['Unnamed: 0'], axis = 1)
df.head()
```

```
Text Score
      0
                                                         1
           I bought these from a large chain pet store. a...
      1
             This soup is incredibly good! But honestly, I...
                                                         5
      2 Our family loves these tasty and healthy sesam...
                                                         5
import nltk
nltk.download('stopwords')
nltk.download('wordnet')
     [nltk data] Downloading package stopwords to /root/nltk data...
     [nltk data]
                   Package stopwords is already up-to-date!
     [nltk_data] Downloading package wordnet to /root/nltk_data...
     [nltk_data] Package wordnet is already up-to-date!
     True
import re
import nltk
from nltk.stem.porter import PorterStemmer
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
from nltk.tokenize import word_tokenize
corpus=[]
for i in range(0,3150):
    #replace non-letter with space
    review = re.sub('[^a-zA-Z]', ' ', df['Text'][i] )
    #convert all to lower case
    review=review.lower()
    #split
    review=review.split()
    #stemming and Lemmanization
    ps=PorterStemmer()
    lm = WordNetLemmatizer()
    #review=[ps.stem(word) for word in review if not word in set(stopwords.words('english'
    review = [lm.lemmatize(word) for word in review if word not in set(stopwords.words('en
    review=' '.join(review)
    corpus.append(review)
final_df = df[df["Score"]!=3]
final_df.shape
     (32397, 2)
final_df["Score"] = final_df["Score"].apply(lambda x : 1 if x>3 else 0)
```

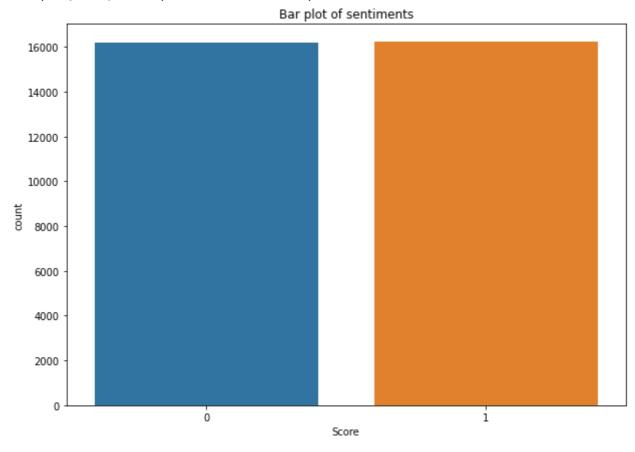
```
final_df['Score'].value_counts()
```

1 16223

0 16174 Name: Score, dtype: int64

```
plt.figure(figsize = (10,7))
sns.countplot(final_df['Score'])
plt.title("Bar plot of sentiments")
```

Text(0.5, 1.0, 'Bar plot of sentiments')



```
def decontract(text):
    text = re.sub(r"won\'t", "will not", text)
    text = re.sub(r"can\'t", "can not", text)
    text = re.sub(r"n\'t", " not", text)
    text = re.sub(r"\'re", " are", text)
    text = re.sub(r"\'s", " is", text)
    text = re.sub(r"\'d", " would", text)
    text = re.sub(r"\'ll", " will", text)
    text = re.sub(r"\'t", " not", text)
    text = re.sub(r"\'ve", " have", text)
    text = re.sub(r"\'we", " am", text)
    return text
```

'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over'

```
'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any',
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', '
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now',
            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'd
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn'
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn
            'won', "won't", 'wouldn', "wouldn't"])
lemmatizer = WordNetLemmatizer()
def preprocess text(review):
                                                      # removing website links
    review = re.sub(r"http\S+", "", review)
    review = BeautifulSoup(review, 'lxml').get_text() # removing html tags
    review = decontract(review)
                                                       # decontracting
    review = re.sub("\S*\d\S*", "", review).strip()
                                                      # removing the words with numeric
    review = re.sub('[^A-Za-z]+', ' ', review)
                                                      # removing non-word characters
    review = review.lower()
                                                        # converting to lower case
    review = [word for word in review.split(" ") if not word in stop_words] # removing sto
    review = [lemmatizer.lemmatize(token, "v") for token in review] #Lemmatization
    review = " ".join(review)
    review.strip()
    return review
final_df['Text'] = final_df['Text'].apply(lambda x: preprocess_text(x))
final_df['Text'].head()
          buy large chain pet store read review check ba...
          soup incredibly good honestly look better deal...
          family love tasty healthy sesame honey almonds...
          local auto shop offer free customers try twice...
          bring bottle one carry pocket home fell love v...
     Name: Text, dtype: object
train df, test df = train test split(final df, test size = 0.2, random state = 42)
print("Training data size : ", train_df.shape)
print("Test data size : ", test_df.shape)
 T→ Training data size : (25917, 2)
     Test data size: (6480, 2)
top words = 6000
tokenizer = Tokenizer(num words=top words)
tokenizer.fit on texts(train df['Text'])
list_tokenized_train = tokenizer.texts_to_sequences(train_df['Text'])
max review length = 1691
X_train = pad_sequences(list_tokenized_train, maxlen=max_review_length)
```

y_train = train_df['Score']

```
embedding_vecor_length = 231
model = Sequential()
model.add(Embedding(top words+1, embedding vecor length, input length=max review length))
model.add(LSTM(100))
model.add(Dense(1, activation='sigmoid'))
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
model.summary()
    Model: "sequential"
    Layer (type)
                              Output Shape
                                                     Param #
    ______
    embedding (Embedding)
                              (None, 1691, 231)
                                                     1386231
    1stm (LSTM)
                              (None, 100)
                                                     132800
                              (None, 1)
    dense (Dense)
                                                     101
    ______
    Total params: 1,519,132
    Trainable params: 1,519,132
    Non-trainable params: 0
model.fit(X_train,y_train, epochs=1, batch_size=64, validation_split=0.2)
    <keras.callbacks.History at 0x7fadc020cc50>
list_tokenized_test = tokenizer.texts_to_sequences(test_df['Text'])
X_test = pad_sequences(list_tokenized_test, maxlen=max_review_length)
y test = test df['Score']
prediction = model.predict(X test)
y_pred = (prediction > 0.5)
print("Accuracy of the model : ", accuracy score(y pred, y test))
print('F1-score: ', f1_score(y_pred, y_test))
print('Confusion matrix:')
confusion_matrix(y_test,y_pred)
    Accuracy of the model : 0.8487654320987654
    F1-score: 0.8429487179487181
    Confusion matrix:
    array([[2870, 370],
          [ 610, 2630]])
from sklearn.metrics import classification report
print(classification_report(y_test,y_pred))
                precision
                            recall f1-score
                                             support
```

3240

3240

0.89

0.81

0.85

0.84

0.82

0.88

0

1

accuracy			0.85	6480
macro avg	0.85	0.85	0.85	6480
weighted avg	0.85	0.85	0.85	6480

