#### Dataset:

https://data.world/crowdflower/sentiment-analysis-in-text

# In [ ]:

```
import pandas as pd
# Reading the csv dataset into a pandas dataframe
df = pd.read_csv('E:/Internships/TCS-iON/Code/MyCode/Tweets/text_emotion.csv', encoding = '
# Adding a column representing 1 for positive and 0 for negative sentiments
df['senti'] = df.apply(lambda x: 1 if (x['sentiment'] == 'enthusiasm' or x['sentiment'] ==
# Deleting unnecessary columns
df = df.drop(['tweet_id', 'sentiment', 'author'], axis = 1)
# Converting the data type to string
df['content'] = df["content"].astype("str")
# Converting all text to lowercase for use
df['content'] = df['content'].str.lower()
df.head()
```

review	senti
@tiffanylue i know i was listenin to bad habi	0
layin n bed with a headache ughhhhwaitin o	0
funeral ceremonygloomy friday	0
wants to hang out with friends soon!	1
@dannycastillo we want to trade with someone w	0

i am going to start reading the harry potter series again because that is one awesome story.

```
import re
import string
from nltk import WordNetLemmatizer
from nltk.stem.snowball import SnowballStemmer
from nltk.corpus import stopwords
# Initialising the nltk stop_words, stemmer and lemmatizer functions
stop_words = set(stopwords.words("english"))
lemmatizer = WordNetLemmatizer()
stemmer = SnowballStemmer("english")
# Creating a function for text cleaning
def textCleanser(myText):
    # Converting each tweet to string
   myText = str(myText)
    # Removing the name titles and the period symbols after it
   myText = re.sub(r'[mdsr]r(s)?\.', '', myText)
    # Removing the '@username' mentions
   myText = re.sub(r'@\w+\s', '', myText)
    # Removing punctuation
   myPunct = string.punctuation
   punctToSpace = str.maketrans(myPunct, len(myPunct)*' ')
   myText = myText.translate(punctToSpace)
    # Removing urls
   myText = re.sub(r'((http(s?)?)://?)(www\.?).+\.com', '', myText)
   myText = re.sub(r'http(s?))', '', myText)
    # Removing numbers
   myText = re.sub(r'\d+', '', myText)
    # Removing stopwords
   myText = [word for word in myText.split(' ') if not word in stop words]
   myText = [word for word in myText if word != '']
    # Lemmatizing the text
   myText = [lemmatizer.lemmatize(token) for token in myText]
    # Stemming the text
    # myText = [stemmer.stem(token) for token in myText]
    return myText
for i in range(len(df['content'])):
    df['content'][i] = textCleanser(df['content'][i])
df.head()
```

review	senti
[know, listenin, bad, habit, earlier, started,	0
[layin, n, bed, headache, ughhhh, waitin, call]	0
[funeral, ceremony, gloomy, friday]	0
[want, hang, friend, soon]	1
[want, trade, someone, houston, ticket, one]	0

['going', 'start', 'reading', 'harry', 'potter', 'series', 'one', 'awesome', 'story']

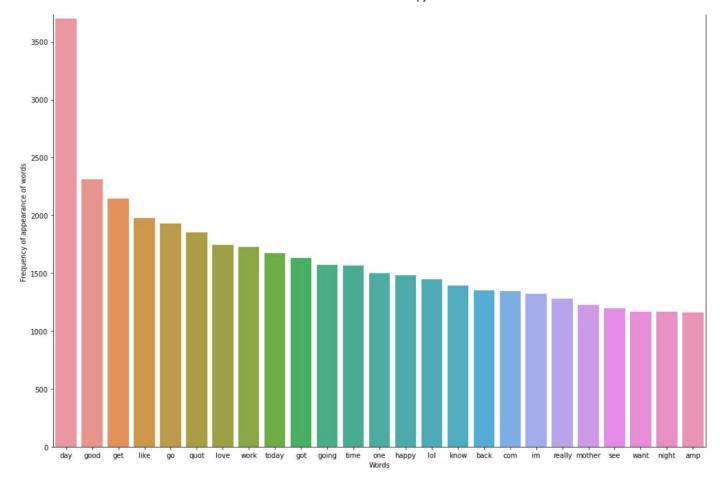
```
myReviews = []
for i in range(len(df['content'])):
    for j in df['content'][i]:
        if j != 'br' and j != 'http':
            myReviews.append(j)
```

## In [ ]:

```
from collections import Counter
import collections
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.base import BaseEstimator, TransformerMixin
from sklearn.feature extraction.text import CountVectorizer, TfidfVectorizer
from sklearn.model selection import GridSearchCV, train test split
from sklearn.pipeline import Pipeline, FeatureUnion
from sklearn.metrics import classification report
from sklearn.naive_bayes import MultinomialNB
from sklearn.linear model import LogisticRegression
import numpy as np
np.random.seed(1234)
# Initialising the Count Vectorizer
cv = CountVectorizer()
myBow = cv.fit_transform(myReviews)
wordFrequency = dict(zip(cv.get_feature_names(), np.asarray(myBow.sum(axis = 0)).ravel()))
wordCounter = collections.Counter(wordFrequency)
# Storing the frequency of appearance of words
dfWordCounter = pd.DataFrame(wordCounter.most common(25), columns = ['word', 'frequency'])
```

#### In [ ]:

```
# Plotting the top 25 most frequently occurring words
plt.close('all')
fig, ax = plt.subplots(figsize = (17, 12))
sns.barplot(x = 'word', y = 'frequency', data = dfWordCounter, ax = ax)
sns.set_palette('pastel')
plt.xlabel('Words')
plt.ylabel('Frequency of appearance of words')
plt.show()
```



```
# Defining a dummy function for the tokenizer and preprocessor inputs of the vectorizers so
def dummy_fun(doc):
    return doc
tfidf = TfidfVectorizer(sublinear_tf=True, max_df = 2000, min_df = 50, norm='12', encoding=
features = []
features = tfidf.fit_transform(df.content).toarray()
labels = df.senti
print(features.shape)
```

### Features Shape:

(40000, 823)

### In [ ]:

```
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.pipeline import Pipeline
# Splitting the dataframe to training and testing datasets
X_train, X_test, y_train, y_test = train_test_split(df["review"], df['senti'],test_size=0.2
vectorizerCount = CountVectorizer(tokenizer = dummy_fun, preprocessor = dummy_fun)
X_train_counts = vectorizerCount.fit_transform(X_train)
tfidf_transformer = TfidfTransformer()
X_train_tfidf = tfidf_transformer.fit_transform(X_train_counts)
nBayes = MultinomialNB().fit(X_train_tfidf, y_train)
```

```
In [ ]:
```

```
y_= nBayes.predict(vectorizerCount.transform(X_test))
```

```
# Testing the data
from sklearn.metrics import accuracy_score
print(accuracy_score(y_test, y_,normalize=True))
```

Accuracy:

71.31%