# **Sentiment Analysis using textblob**

```
In [1]:
        1 #!pip install textblob
        2 import textblob
        1 from textblob import TextBlob
In [2]:
        1 txt1 = TextBlob('This was a good movie')
In [3]:
        1 txt1.sentiment
In [4]:
Out[4]: Sentiment(polarity=0.7, subjectivity=0.6000000000000001)
In [5]:
        1 txt2 = TextBlob('This was a bad movie')
        1 txt2.sentiment
In [6]:
1 txt3 = TextBlob('This was an okay movie')
In [7]:
        1 txt3.sentiment
In [8]:
Out[8]: Sentiment(polarity=0.5, subjectivity=0.5)
```

# **Sentiment Analysis using Vader**

```
In [9]: 1 import vaderSentiment
```

```
In [10]:
             !pip install vaderSentiment
         Requirement already satisfied: vaderSentiment in c:\users\yashm\anaconda3\lib\site-packages (3.3.2)
         Requirement already satisfied: requests in c:\users\yashm\anaconda3\lib\site-packages (from vaderSentiment) (2.25.1)
         Requirement already satisfied: chardet<5,>=3.0.2 in c:\users\yashm\anaconda3\lib\site-packages (from requests->vaderSen
         timent) (3.0.4)
         Requirement already satisfied: idna<3,>=2.5 in c:\users\yashm\anaconda3\lib\site-packages (from requests->vaderSentimen
         t) (2.8)
         Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\yashm\anaconda3\lib\site-packages (from requests->vade
         rSentiment) (1.25.8)
         Requirement already satisfied: certifi>=2017.4.17 in c:\users\yashm\anaconda3\lib\site-packages (from requests->vaderSe
         ntiment) (2019.11.28)
         WARNING: You are using pip version 20.3.3; however, version 22.2.2 is available.
         You should consider upgrading via the 'C:\Users\yashm\anaconda3\python.exe -m pip install --upgrade pip' command.
           1 from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
In [11]:
In [12]:
             sentiment analyser = SentimentIntensityAnalyzer()
In [13]:
           1 sent = "This was a great movie"
In [14]:
              sent analysis = sentiment analyser.polarity scores(sent)
           1 sent analysis
In [15]:
Out[15]: {'neg': 0.0, 'neu': 0.494, 'pos': 0.506, 'compound': 0.6249}
In [16]:
           1 sent analysis
Out[16]: {'neg': 0.0, 'neu': 0.494, 'pos': 0.506, 'compound': 0.6249}
```

### **Excercise**

```
In [17]:
             import pandas as pd
           3 data = pd.read_csv('Tweets.csv')
           1 data[['airline_sentiment','airline_sentiment_confidence','text']]
In [18]:
           2
           3
Out[18]:
```

text	airline_sentiment_confidence	airline_sentiment	
@VirginAmerica What @dhepburn said.	1.0000	neutral	0
@VirginAmerica plus you've added commercials t	0.3486	positive	1
@VirginAmerica I didn't today Must mean I n	0.6837	neutral	2
@VirginAmerica it's really aggressive to blast	1.0000	negative	3
@VirginAmerica and it's a really big bad thing	1.0000	negative	4
@AmericanAir thank you we got on a different f	0.3487	positive	14635
@AmericanAir leaving over 20 minutes Late Flig	1.0000	negative	14636
@AmericanAir Please bring American Airlines to	1.0000	neutral	14637
@AmericanAir you have my money, you change my	1.0000	negative	14638
@AmericanAir we have 8 ppl so we need 2 know h	0.6771	neutral	14639

14640 rows × 3 columns

# **Excercise:**

- 1. Preprocess the text with the techniques dicussed yesterday
- 2. Creat a user defined function that returns the compound sentiment from the vader function
- 3. If the compound sentiment is greater than 0.5 it is positive and if it is less than -0.5 it is negative and if it is between -0.5 to 0.5 it is neutral
- 4. Compare the above output with the column of airline sentiment and evaluate your accuracy

```
1 df = data[['airline sentiment', 'airline sentiment confidence', 'text']].iloc[:500]
In [19]:
             from nltk.stem import SnowballStemmer
              from nltk.tokenize import word tokenize
           5
           6
           7
              def remove punc(string):
                  punc = '''!()-[]{};:'"\,<>./?@#$%^&* ~'''
           9
                  for char in string:
          10
          11
                      if char in punc:
                           string = string.replace(char,"")
          12
          13
                  return string
          14
              def stem text(string):
          15
                  ps = SnowballStemmer(language = 'english')
          16
                  words = word tokenize(string)
          17
                  sentence = []
          18
                  for word in words:
          19
          20
                      sentence.append(ps.stem(word))
                  return " ".join(sentence)
          21
          22
          23
              def lower(string):
                  return string.lower()
          24
          25
          26
          27
          28
              def clean text(string):
                  string = remove punc(string)
          29
                  string = stem text(string)
          30
                  return string.lower()
          31
          32
          33
          34 df['clean text'] = df['text'].apply(clean text)
```

```
In [24]:
           1 def vader sentiment analyser(sent):
                  sentiment analyser = SentimentIntensityAnalyzer()
           2
                  compound = sentiment analyser.polarity scores(sent)['compound']
           3
                  if compound>0.5:
           4
           5
                      return 'positive'
           6
                  elif compound<-0.5:</pre>
                      return 'negative'
           7
           8
                  else:
           9
                      return 'neutral'
          10
           1 df['vader score'] = df['clean text'].apply(vader sentiment analyser)
In [25]:
           1 from sklearn.metrics import classification report
In [26]:
           1 print(classification report(df['vader score'],df['airline sentiment']))
In [27]:
                                     recall f1-score
                        precision
                                                         support
             negative
                             0.21
                                       0.95
                                                  0.34
                                                              40
               neutral
                             0.86
                                       0.40
                                                  0.55
                                                             363
              positive
                                                  0.54
                             0.44
                                       0.69
                                                              97
                                                  0.50
                                                             500
              accuracy
            macro avg
                                                  0.48
                                                             500
                             0.51
                                       0.68
         weighted avg
                             0.73
                                                  0.53
                                       0.50
                                                             500
 In [ ]:
```