```
In [1]: import pandas as pd
   import warnings
   warnings.filterwarnings('ignore')
   import numpy as np

import matplotlib.pyplot as plt
   %matplotlib inline
   import seaborn as sns

import re
   import nltk
   from nltk.corpus import stopwords
   import string
   from sklearn.metrics import classification_report
```

```
In [2]: df = pd.read_csv("NASA.csv")
    df.drop(df.columns[[1,2,4,5,6,7,8,9,10,11]], axis=1, inplace=True)
    df
```

#### Out[2]:

	Unnamed: 0	Tweet
0	0	Here's to you, Oppy. 🥂 \n\nBefore you say #Good
1	1	Are there rivers and lakes on other worlds? Yo
2	2	We want to hear from you!\n\nJoin our series o
3	3	The @NASAExoplanets data hint that WASP-39 b,
4	4	.@NASAWebb just scored another first: a full p
14105	14105	The supermoon is here! Be sure to bundle up th
14106	14106	Ever wonder how we track supermoons 🦲 and othe
14107	14107	A supermoon is coming! Tonight, the full Moon
14108	14108	Happy New Year from space! Astronauts aboard t
14109	14109	□ Send a robot to Mars\n□ Launch @NASA_Astrona

14110 rows × 2 columns

```
In [3]: df['length'] = df['Tweet'].str.len()
df['Tweet'][0]
```

Out[3]: "Here's to you, Oppy. // \n\nBefore you say #GoodNightOppy, learn more about our Opportunity @NASAMars rover, and how a planned 90-day mission turned int o a 15-year journey of exploration and discovery: https://t.co/KTf5ECdUrO (https://t.co/KTf5ECdUrO) https://t.co/70f9CjxGrb" (https://t.co/70f9CjxGrb")

```
In [4]: df['length'].describe()
Out[4]: count
                 14110.000000
        mean
                   239.409001
                    75.212052
        std
        min
                     8.000000
        25%
                   208.000000
        50%
                   268.000000
        75%
                   297.000000
                   453.000000
        max
        Name: length, dtype: float64
```

# **Conversion of Emoji**

```
In [5]:
                            import regex
                            import emoji
                            import html.parser as html
                            emoticons = [':-)', ':)', '(:', '(-:', ':))', '((:', ':-D', ':D', 'X-D', 'XD'
                                                                                                            ';-D', ';D', '(;', '(-;', ':-(', ':(', '(:', '(-:', ':
                                                                                                            '=(', ')=', '=-0', 'O-=', ':0', 'O:', 'O:', 'O:', ':-o
                                                                                                          '^_^', '^.^', '>.>', 'T_T', 'T-T', '-.-', '*.*', '~.~'
                                                                                                              ':-<', '$_$', '8-)', ':-P', ':-p', '=P', '=p', ':*)',
                            def split_count(text):
                                          text = html.unescape(text)
                                          emoji_list = []
                                          data = regex.findall(r'\X', text)
                                          for ch in data:
                                                        if any(char in emoji.EMOJI_DATA for char in ch):
                                                                     emoji_list.append(ch)
                                          for word in text.split(' '):
                                                        if word in emoticons :
                                                                     emoji list.append(word)
                                          return emoji list
                            text = df['Tweet']
                            emoji list= []
                            for t in text:
                                          emoji list=emoji list+split count(t)
                            from collections import Counter
                            print(len(emoji list))
                            print(emoji list)
                            7026
                                                       ' 😽 ', '⊕', 'え`
                                                   '∰\u200d 🚀 ', '່ 🔆 '

    \u200d 
    \u20dd 
    \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
  \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
  \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 
   \u20dd 

                                                   ' 🐾 ',' 👨 \u200d 🎻 '
```

```
In [6]:
    import re
    from emot.emo_unicode import UNICODE_EMOJI
    def convert_emojis(text):
        for emot in UNICODE_EMOJI:
            text = text.replace(emot, " ".join(UNICODE_EMOJI[emot].replace(",",""
            return text

df['Tweet'] = df['Tweet'].apply(lambda x: convert_emojis(x))
df
```

#### Out[6]:

	Unnamed: 0	Tweet	length
0	0	Here's to you, Oppy. clinking_glasses\n\nBefor	245
1	1	Are there rivers and lakes on other worlds? Yo	293
2	2	We want to hear from you!\n\nJoin our series o	302
3	3	The @NASAExoplanets data hint that WASP-39 b, $\dots$	175
4	4	.@NASAWebb just scored another first: a full p	189
14105	14105	The supermoon is here! Be sure to bundle up th	174
14106	14106	Ever wonder how we track supermoons full_moon	140
14107	14107	A supermoon is coming! Tonight, the full Moon	255
14108	14108	Happy New Year from space! Astronauts aboard t	254
14109	14109	white_medium_square Send a robot to Mars\nwhit	272

14110 rows × 3 columns

### **Data Preprocessing**

```
In [7]: df.isnull().any()

Out[7]: Unnamed: 0   False
         Tweet          False
          length          False
          dtype: bool
```

```
In [8]: df['dup'] = df.duplicated(subset=None, keep='first')
    del df['dup']
    df
```

#### Out[8]:

	Unnamed: 0	Tweet	length
0	0	Here's to you, Oppy. clinking_glasses\n\nBefor	245
1	1	Are there rivers and lakes on other worlds? Yo	293
2	2	We want to hear from you!\n\nJoin our series o	302
3	3	The @NASAExoplanets data hint that WASP-39 b, $\dots$	175
4	4	.@NASAWebb just scored another first: a full p	189
14105	14105	The supermoon is here! Be sure to bundle up th	174
14106	14106	Ever wonder how we track supermoons full_moon	140
14107	14107	A supermoon is coming! Tonight, the full Moon	255
14108	14108	Happy New Year from space! Astronauts aboard t	254
14109	14109	white_medium_square Send a robot to Mars\nwhit	272

14110 rows × 3 columns

```
In [9]: def text_lowering(text):
    text = text.lower()
    return text
df['Tweet'] = df['Tweet'].apply(lambda x: text_lowering(x))
df
```

#### Out[9]:

Unnamed: 0		Tweet	length
0	0	here's to you, oppy. clinking_glasses\n\nbefor	245
1	1	are there rivers and lakes on other worlds? yo	293
2	2	we want to hear from you!\n\njoin our series o	302
3	3	the @nasaexoplanets data hint that wasp-39 b, $\dots$	175
4	4	.@nasawebb just scored another first: a full p	189
14105	14105	the supermoon is here! be sure to bundle up th	174
14106	14106	ever wonder how we track supermoons full_moon $\dots$	140
14107	14107	a supermoon is coming! tonight, the full moon $\dots$	255
14108	14108	happy new year from space! astronauts aboard t	254
14109	14109	white_medium_square send a robot to mars\nwhit	272

```
In [10]: def remove_html_tags(text):
    html=re.compile(r'<.*?>')
    text = html.sub(r'',text)
    return text
df['Tweet'] = df['Tweet'].apply(lambda x: remove_html_tags(x))
df
```

#### Out[10]:

	Unnamed: 0	Tweet	length
0	0	here's to you, oppy. clinking_glasses\n\nbefor	245
1	1	are there rivers and lakes on other worlds? yo	293
2	2	we want to hear from you!\n\njoin our series o	302
3	3	the @nasaexoplanets data hint that wasp-39 b,	175
4	4	.@nasawebb just scored another first: a full p	189
14105	14105	the supermoon is here! be sure to bundle up th	174
14106	14106	ever wonder how we track supermoons full_moon	140
14107	14107	a supermoon is coming! tonight, the full moon	255
14108	14108	happy new year from space! astronauts aboard t	254
14109	14109	white_medium_square send a robot to mars\nwhit	272

```
In [11]: def replace_uderScores(tweet):
    return tweet.replace("_"," ")
def remove_url_tags(text):
    text = re.sub(r"http\S+", "",text)
    return text
df['Tweet'] = df['Tweet'].apply(lambda x: remove_url_tags(x))
df['Tweet'] = df['Tweet'].apply(lambda x: replace_uderScores(x))
df
```

#### Out[11]:

	Unnamed: 0	Tweet	length
0	0	here's to you, oppy. clinking glasses\n\nbefor	245
1	1	are there rivers and lakes on other worlds? yo	293
2	2	we want to hear from you!\n\njoin our series o	302
3	3	the @nasaexoplanets data hint that wasp-39 b, $\dots$	175
4	4	.@nasawebb just scored another first: a full p	189
14105	14105	the supermoon is here! be sure to bundle up th	174
14106	14106	ever wonder how we track supermoons full moon $\dots$	140
14107	14107	a supermoon is coming! tonight, the full moon	255
14108	14108	happy new year from space! astronauts aboard t	254
14109	14109	white medium square send a robot to mars\nwhit	272

```
In [12]: PUNCT_TO_REMOVE = string.punctuation
    print(PUNCT_TO_REMOVE)
    def remove_punctuation(text):
        return text.translate(str.maketrans('', '', PUNCT_TO_REMOVE))

df["Tweet"] = df["Tweet"].apply(lambda text: remove_punctuation(text))
    df
```

!"#\$%&'()\*+,-./:;<=>?@[\]^\_`{|}~

#### Out[12]:

	Unnamed: 0	Tweet	length
0	0	heres to you oppy clinking glasses\n\nbefore y	245
1	1	are there rivers and lakes on other worlds you	293
2	2	we want to hear from you\n\njoin our series of	302
3	3	the nasaexoplanets data hint that wasp39 b aka	175
4	4	nasawebb just scored another first a full prof	189
14105	14105	the supermoon is here be sure to bundle up the	174
14106	14106	ever wonder how we track supermoons full moon	140
14107	14107	a supermoon is coming tonight the full moon wi	255
14108	14108	happy new year from space astronauts aboard th	254
14109	14109	white medium square send a robot to mars\nwhit	272

```
In [13]: import nltk
    nltk.download('stopwords')
    import nltk
    #nltk.download('omw-1.4')
    STOPWORDS = set(stopwords.words('english'))
    def remove_stopwords(text):
        return " ".join([word for word in str(text).split() if word not in STOPWOID
    df["Tweet"] = df["Tweet"].apply(lambda text: remove_stopwords(text))
    df
```

[nltk\_data] Downloading package stopwords to
[nltk\_data] C:\Users\peram\AppData\Roaming\nltk\_data...
[nltk\_data] Package stopwords is already up-to-date!

#### Out[13]:

	Unnamed: 0	Tweet	length
0	0	heres oppy clinking glasses say goodnightoppy	245
1	1	rivers lakes worlds bet like earth saturn's mo	293
2	2	want hear join series virtual inperson meeting	302
3	3	nasaexoplanets data hint wasp39 b aka bocaprin	175
4	4	nasawebb scored another first full profile ato	189
14105	14105	supermoon sure bundle lead "pack" outside view	174
14106	14106	ever wonder track supermoons full moon lunar e	140
14107	14107	supermoon coming tonight full moon near closes	255
14108	14108	happy new year space astronauts aboard space s	254
14109	14109	white medium square send robot mars white medi	272

```
In [14]: from nltk.corpus import wordnet
                                       from nltk.stem import WordNetLemmatizer
                                       import nltk
                                       nltk.download('wordnet')
                                       # nltk.download('omw-1.4')
                                       nltk.download('averaged_perceptron_tagger')
                                       lemmatizer = WordNetLemmatizer()
                                       wordnet_map = {"N":wordnet.NOUN, "V":wordnet.VERB, "J":wordnet.ADJ, "R":wordnet
                                       def lemmatize_words(text):
                                                        pos_tagged_text = nltk.pos_tag(nltk.word_tokenize(text))
                                                        return " ".join([lemmatizer.lemmatize(word, wordnet_map.get(pos[0], wordnet_ma
                                       df["Tweet"] = df["Tweet"].apply(lambda text: lemmatize words(text))
                                       df
                                        [nltk_data] Downloading package wordnet to
                                        [nltk data]
                                                                                                          C:\Users\peram\AppData\Roaming\nltk data...
```

#### Out[14]:

	Unnamed: 0	Tweet	length
0	0	here oppy clink glass say goodnightoppy learn	245
1	1	river lakes world bet like earth saturn 's mo	293
2	2	want hear join series virtual inperson meeting	302
3	3	nasaexoplanets data hint wasp39 b aka bocaprin	175
4	4	nasawebb score another first full profile atom	189
14105	14105	supermoon sure bundle lead " pack " outside vi	174
14106	14106	ever wonder track supermoons full moon lunar e	140
14107	14107	supermoon come tonight full moon near close po	255
14108	14108	happy new year space astronauts aboard space s	254
14109	14109	white medium square send robot mar white mediu	272

14110 rows × 3 columns

## **Adding Labels**

```
In [15]: | nltk.download('vader lexicon')
         from nltk.sentiment.vader import SentimentIntensityAnalyzer
         sentiments = SentimentIntensityAnalyzer()
         df["Positive"] = [sentiments.polarity_scores(i)["pos"] for i in df["Tweet"]]
         df["Negative"] = [sentiments.polarity_scores(i)["neg"] for i in df["Tweet"]]
         df["Neutral"] = [sentiments.polarity_scores(i)["neu"] for i in df["Tweet"]]
         df['Compound'] = [sentiments.polarity_scores(i)["compound"] for i in df["Twee")
         score = df["Compound"].values
         sentiment = []
         for i in score:
             if i > 0 :
                 sentiment.append(1)
             elif i < 0 :
                 sentiment.append(-1)
             else:
                 sentiment.append(0)
         df['Sentiment'] = sentiment
         df.head()
```

#### Out[15]:

	Unnamed: 0	Tweet	length	Positive	Negative	Neutral	Compound	Sentiment
0	0	here oppy clink glass say goodnightoppy learn	245	0.227	0.000	0.773	0.5719	1
1	1	river lakes world bet like earth saturn 's mo	293	0.091	0.000	0.909	0.3612	1
2	2	want hear join series virtual inperson meeting	302	0.137	0.000	0.863	0.3612	1
3	3	nasaexoplanets data hint wasp39 b aka bocaprin	175	0.000	0.154	0.846	-0.4767	-1
4	4	nasawebb score another first full profile atom	189	0.162	0.000	0.838	0.4019	1

In [16]: df.drop(df.columns[[3,4,5,6]], axis=1, inplace=True)
df

Out[16]:

Unnamed: 0		Tweet	length	Sentiment
0	0	here oppy clink glass say goodnightoppy learn	245	1
1	1	river lakes world bet like earth saturn 's mo	293	1
2	2	want hear join series virtual inperson meeting	302	1
3	3	nasaexoplanets data hint wasp39 b aka bocaprin	175	-1
4	4	nasawebb score another first full profile atom	189	1
14105	14105	supermoon sure bundle lead " pack " outside vi	174	1
14106	14106	ever wonder track supermoons full moon lunar e	140	0
14107	14107	supermoon come tonight full moon near close po	255	0
14108	14108	happy new year space astronauts aboard space s	254	1
14109	14109	white medium square send robot mar white mediu	272	0

14110 rows × 4 columns

In [17]: df['Sentiment'].value\_counts()

Out[17]:

8636
 4234

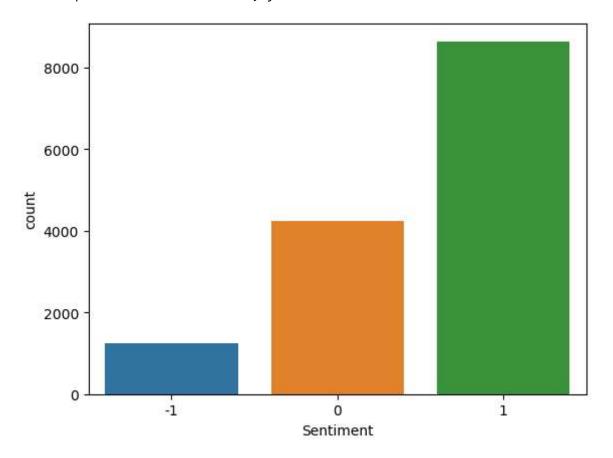
-1 1240

Name: Sentiment, dtype: int64

### **Data Visualization**

```
In [18]: sns.countplot(x ='Sentiment', data = df)
```

Out[18]: <AxesSubplot:xlabel='Sentiment', ylabel='count'>



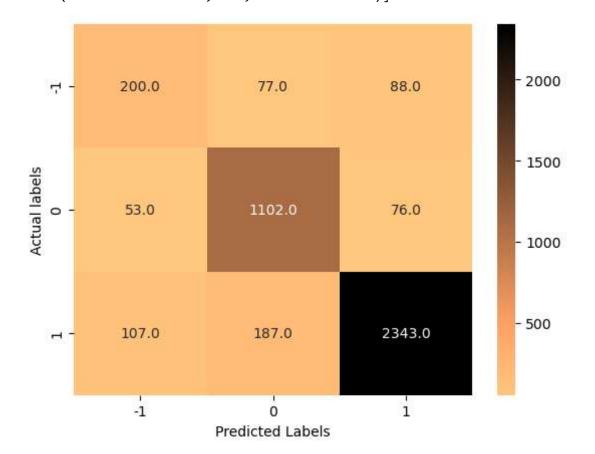
### **Model Building**

```
In [19]: from sklearn.feature_extraction.text import CountVectorizer
    cv = CountVectorizer(max_df=0.90,min_df=1,max_features = 14110,stop_words='en
    x = cv.fit_transform(df['Tweet'])
```

In [20]: from sklearn.model\_selection import train\_test\_split
X\_train,X\_test,y\_train,y\_test = train\_test\_split(x,df['Sentiment'],test\_size=

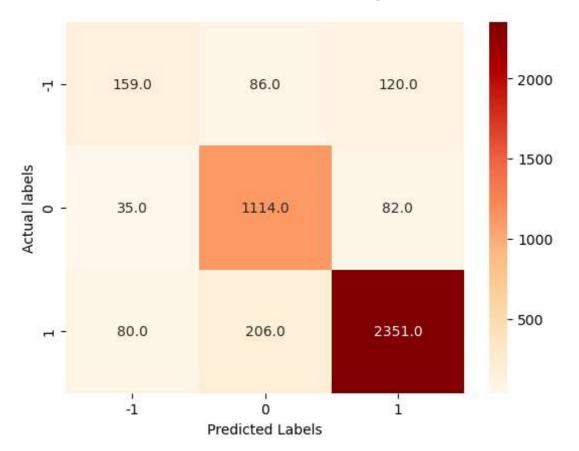
```
In [21]: from sklearn import svm
    from sklearn.metrics import accuracy_score,confusion_matrix
    model1 = svm.SVC(kernel='linear', C=1)
    model1.fit(X_train,y_train)
    y_pred = model1.predict(X_test)
    print("accuracy_score",accuracy_score(y_test,y_pred))
    print(classification_report(y_test, y_pred))
    cm = confusion_matrix(y_test, y_pred)
    sns.heatmap(cm,annot=True,fmt=".1f",cmap="copper_r",xticklabels=[-1,0,1],ytic
```

```
accuracy_score 0.8610914245216159
              precision
                            recall f1-score
                                                support
                    0.56
                              0.55
          -1
                                         0.55
                                                    365
           0
                    0.81
                              0.90
                                         0.85
                                                   1231
           1
                    0.93
                              0.89
                                         0.91
                                                   2637
                                         0.86
                                                   4233
    accuracy
   macro avg
                    0.77
                              0.78
                                         0.77
                                                   4233
weighted avg
                    0.86
                              0.86
                                         0.86
                                                   4233
```

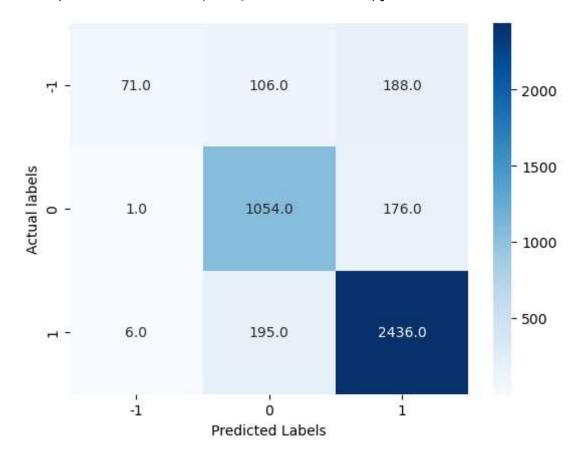


```
In [22]: from sklearn.tree import DecisionTreeClassifier
    from sklearn.metrics import accuracy_score,confusion_matrix,classification_re
    model2 = DecisionTreeClassifier()
    model2.fit(X_train, y_train)
    y_pred = model2.predict(X_test)
    print("accuracy_score",accuracy_score(y_test,y_pred))
    cm = confusion_matrix(y_test, y_pred)
    print(classification_report(y_test, y_pred))
    sns.heatmap(cm,annot=True,fmt=".1f",cmap="OrRd",xticklabels=[-1,0,1],yticklabels=[-1,0,1],yticklabels=[-1,0,1]
```

```
accuracy_score 0.8561304039688165
              precision
                            recall f1-score
                                                support
                   0.58
                              0.44
          -1
                                         0.50
                                                    365
           0
                   0.79
                              0.90
                                         0.84
                                                   1231
           1
                   0.92
                              0.89
                                         0.91
                                                   2637
                                        0.86
                                                   4233
    accuracy
                   0.76
                              0.74
                                         0.75
                                                   4233
   macro avg
weighted avg
                   0.85
                              0.86
                                         0.85
                                                   4233
```

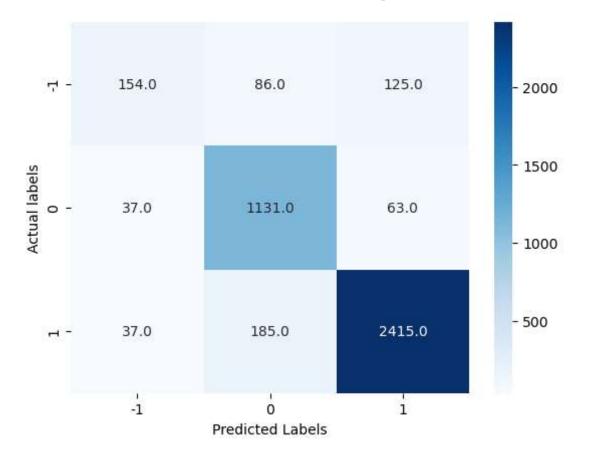


```
accuracy_score 0.8412473423104181
              precision
                            recall f1-score
                                                support
                   0.91
                              0.19
          -1
                                         0.32
                                                    365
           0
                   0.78
                              0.86
                                         0.82
                                                   1231
           1
                   0.87
                              0.92
                                         0.90
                                                   2637
                                        0.84
                                                   4233
    accuracy
                   0.85
                              0.66
                                         0.68
                                                   4233
   macro avg
weighted avg
                   0.85
                              0.84
                                         0.82
                                                   4233
```



```
In [28]: from sklearn.ensemble import VotingClassifier
    from sklearn.metrics import accuracy_score
    final_model = VotingClassifier(estimators=[('svm', model1), ('dt', model2), (
        final_model.fit(X_train,y_train)
        y_pred = final_model.predict(X_test)
        print("accuracy_score",accuracy_score(y_test,y_pred))
        print(classification_report(y_test, y_pred))
        cm = confusion_matrix(y_test, y_pred)
        sns.heatmap(cm,annot=True,fmt=".1f",cmap="Blues",xticklabels=[-1,0,1],ytickla
```

```
accuracy_score 0.8740845735884716
              precision
                            recall f1-score
                                                support
                   0.68
                              0.42
          -1
                                        0.52
                                                    365
           0
                   0.81
                              0.92
                                        0.86
                                                   1231
           1
                   0.93
                              0.92
                                        0.92
                                                   2637
                                        0.87
                                                   4233
    accuracy
                   0.80
                              0.75
                                        0.77
                                                   4233
   macro avg
weighted avg
                   0.87
                              0.87
                                        0.87
                                                   4233
```



In [ ]:	
In [ ]:	