In [1]: !pip install scikit-learn

!pip install imbalanced-learn

!pip install imblearn

Requirement already satisfied: scikit-learn in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/ lib/python3.9/site-packages (1.2.0)

Requirement already satisfied: scipy>=1.3.2 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/ lib/python3.9/site-packages (from scikit-learn) (1.10.0)

Requirement already satisfied: threadpoolctl>=2.0.0 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/lib/python3.9/site-packages (from scikit-learn) (3.1.0)

Requirement already satisfied: numpy>=1.17.3 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silico n/lib/python3.9/site-packages (from scikit-learn) (1.23.2)

Requirement already satisfied: joblib>=1.1.1 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silico n/lib/python3.9/site-packages (from scikit-learn) (1.2.0)

Requirement already satisfied: imbalanced-learn in /Users/pravinanandpawar/miniforge3/envs/tensorflow sili con/lib/python3.9/site-packages (0.10.1)

Requirement already satisfied: joblib>=1.1.1 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silico n/lib/python3.9/site-packages (from imbalanced-learn) (1.2.0)

Requirement already satisfied: scipy>=1.3.2 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/ lib/python3.9/site-packages (from imbalanced-learn) (1.10.0)

Requirement already satisfied: threadpoolctl>=2.0.0 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/lib/python3.9/site-packages (from imbalanced-learn) (3.1.0)

Requirement already satisfied: numpy>=1.17.3 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silico n/lib/python3.9/site-packages (from imbalanced-learn) (1.23.2)

Requirement already satisfied: scikit-learn>=1.0.2 in /Users/pravinanandpawar/miniforge3/envs/tensorflow s ilicon/lib/python3.9/site-packages (from imbalanced-learn) (1.2.0)

Requirement already satisfied: imblearn in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/lib/ python3.9/site-packages (0.0)

Requirement already satisfied: imbalanced-learn in /Users/pravinanandpawar/miniforge3/envs/tensorflow sili con/lib/python3.9/site-packages (from imblearn) (0.10.1)

Requirement already satisfied: numpy>=1.17.3 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silico n/lib/python3.9/site-packages (from imbalanced-learn->imblearn) (1.23.2)

Requirement already satisfied: joblib>=1.1.1 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silico n/lib/python3.9/site-packages (from imbalanced-learn->imblearn) (1.2.0)

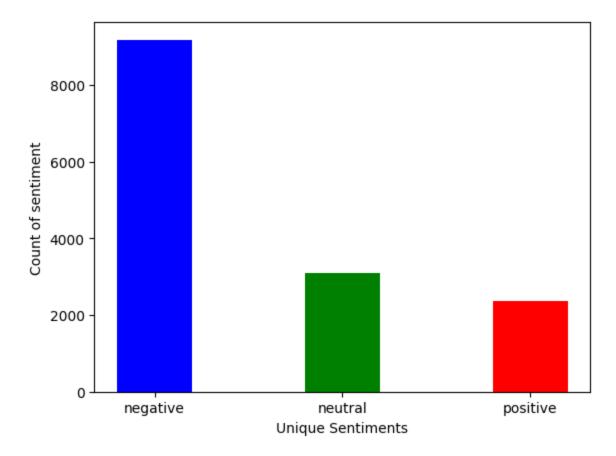
Requirement already satisfied: scipy>=1.3.2 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/ lib/python3.9/site-packages (from imbalanced-learn->imblearn) (1.10.0)

Requirement already satisfied: scikit-learn>=1.0.2 in /Users/pravinanandpawar/miniforge3/envs/tensorflow s ilicon/lib/python3.9/site-packages (from imbalanced-learn->imblearn) (1.2.0)

Requirement already satisfied: threadpoolctl>=2.0.0 in /Users/pravinanandpawar/miniforge3/envs/tensorflow silicon/lib/python3.9/site-packages (from imbalanced-learn->imblearn) (3.1.0)

```
In [2]: from sklearn.model_selection import train_test_split
         from sklearn.metrics import accuracy_score,f1_score
         from wordcloud import WordCloud, STOPWORDS
         from sklearn.feature_extraction.text import TfidfVectorizer
         from imblearn.over_sampling import SMOTE
         from sklearn.svm import SVC
         import matplotlib.pyplot as plt
         import pandas as pd
         import numpy as np
         df = pd.read_csv('Tweets.csv')
In [3]:
         df.head()
Out[3]:
                      tweet_id airline_sentiment airline_sentiment_confidence negativereason negativereason_confidence
                                                                                                                    airline airl
                                                                                                                     Virgin
         0 570306133677760513
                                         neutral
                                                                    1.0000
                                                                                     NaN
                                                                                                                   America
                                                                                                                     Virgin
                                                                                                            0.0000
         1 570301130888122368
                                        positive
                                                                    0.3486
                                                                                     NaN
                                                                                                                   America
                                                                                                                     Virgin
         2 570301083672813571
                                                                    0.6837
                                         neutral
                                                                                     NaN
                                                                                                                   America
                                                                                                                     Virgin
         3 570301031407624196
                                        negative
                                                                    1.0000
                                                                                 Bad Flight
                                                                                                            0.7033
                                                                                                                   America
                                                                                                                     Virgin
         4 570300817074462722
                                        negative
                                                                    1.0000
                                                                                 Can't Tell
                                                                                                            1.0000
                                                                                                                   America
In [4]: listAirline = df['airline'].unique()
         listAirline
```

```
Out[4]: array(['Virgin America', 'United', 'Southwest', 'Delta', 'US Airways',
                'American'], dtype=object)
In [5]: listSentiment = df['airline_sentiment'].unique()
         listSentiment
Out[5]: array(['neutral', 'positive', 'negative'], dtype=object)
In [6]: sentiment details = df['airline sentiment'].value counts()
In [7]: idx = sentiment_details.index
         idx[0]
Out[7]: 'negative'
In [8]: sentiment details.values
Out[8]: array([9178, 3099, 2363])
In [9]: def plot(x, y, x_lable, y_label):
             plt.bar( x, y, color=['blue','Green','red', 'brown', 'olive'],
                     width = 0.4)
             plt.xlabel(x lable)
             plt.ylabel(y label)
             plt.show()
         plot(sentiment details.index, sentiment details.values, "Unique Sentiments", "Count of sentiment")
In [10]:
```

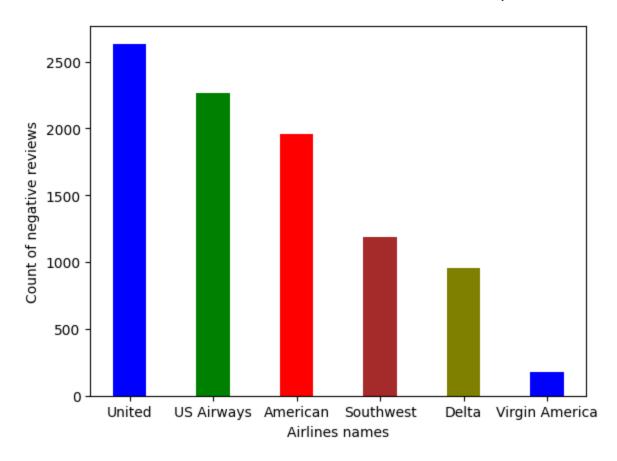


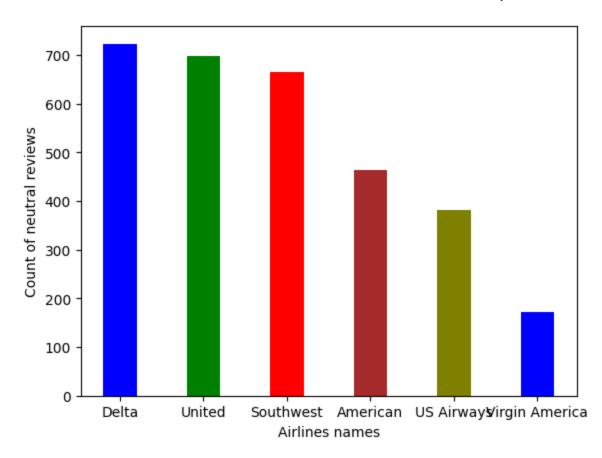
```
In [11]: df_companies = df[['airline_sentiment', 'airline']]
    df_companies
```

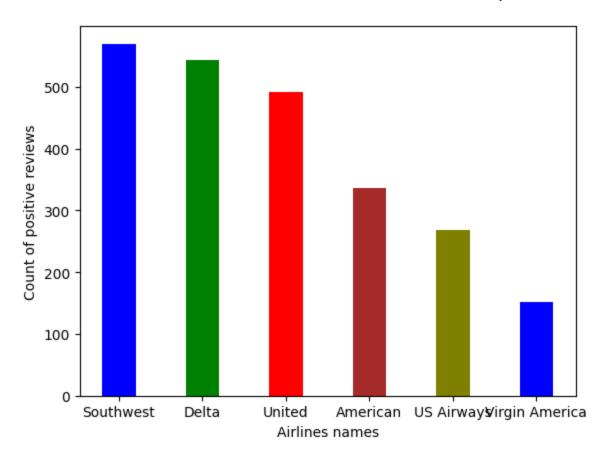
Out[11]:		airline_sentiment	airline
	0	neutral	Virgin America
	1	positive	Virgin America
	2	neutral	Virgin America
	3	negative	Virgin America
	4	negative	Virgin America
	•••		
	14635	positive	American
	14636	negative	American
	14637	neutral	American
	14638	negative	American
	14639	neutral	American

14640 rows × 2 columns

```
In [12]: for x in sentiment_details.index:
    airline_review = df_companies[df_companies.airline_sentiment == x]['airline'].value_counts()
    plot(airline_review.index, airline_review.values, "Airlines names", "Count of "+x+" reviews")
```





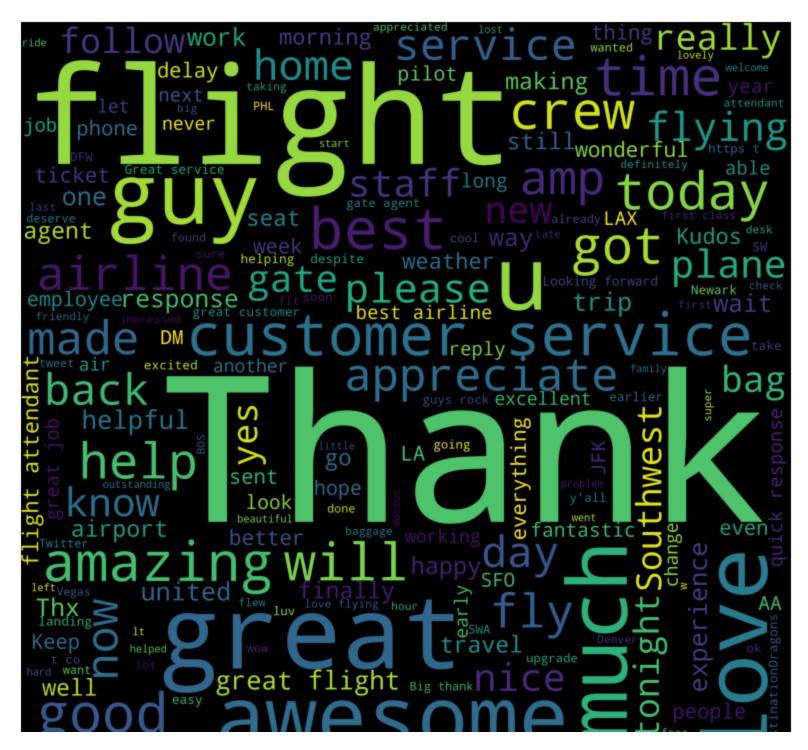


```
In [13]: # Preprocessing
In [14]: df = df[df.airline_sentiment_confidence > 0.7]
    airline_df = df[['airline_sentiment', 'text']]
    airline_df
```

Out[14]:	airline_sentiment		text
	0	neutral	@VirginAmerica What @dhepburn said.
	3	negative	@VirginAmerica it's really aggressive to blast
	4	negative	@VirginAmerica and it's a really big bad thing
	5	negative	@VirginAmerica seriously would pay \$30 a fligh
	9	positive	@VirginAmerica it was amazing, and arrived an
	•••		
	14631	negative	@AmericanAir thx for nothing on getting us out
	14633	negative	@AmericanAir my flight was Cancelled Flightled
	14636	negative	@AmericanAir leaving over 20 minutes Late Flig
	14637	neutral	@AmericanAir Please bring American Airlines to
	14638	negative	@AmericanAir you have my money, you change my

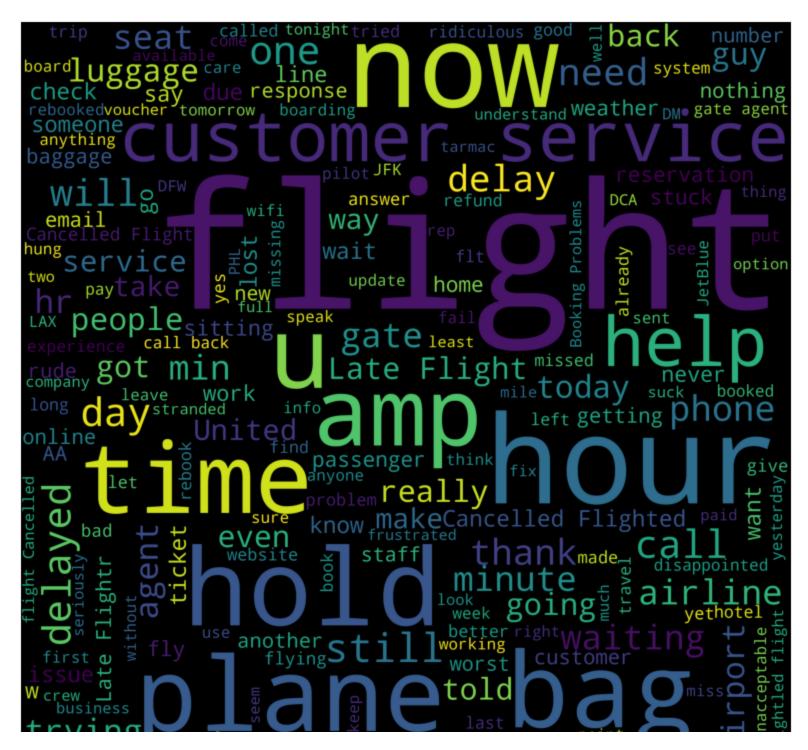
10760 rows × 2 columns

```
In [17]: # Positive
plotWordCloud('positive')
```





In [18]: # Negative
plotWordCloud('negative')





```
In [19]: # Encoding
         y = pd.factorize(airline_df['airline_sentiment'])
Out[19]: (array([0, 1, 1, ..., 1, 0, 1]),
          Index(['neutral', 'negative', 'positive'], dtype='object'))
In [20]: # load data
         X = airline df['text']
         tf idf features = TfidfVectorizer()
         X = tf idf features.fit transform(X)
         print(X[0])
           (0, 9665)
                         0.4326189761894123
           (0, 3954)
                         0.7682167910584247
           (0, 12094) 0.31014240570373325
           (0, 11872)
                         0.3556620187214576
In [21]: smote = SMOTE()
         X, y = smote.fit resample(X, y[0])
In [22]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=31)
In [23]: # SVC
In [24]: model = SVC(class weight='balanced', C=1,qamma='scale', kernel='rbf')
         model.fit(X train, y train)
         y pred = model.predict(X test)
In [25]: accuracy_score = accuracy_score(y_test, y_pred)
         f1 = f1 score(y test,y pred,average='weighted')
In [26]: accuracy_score
Out[26]: 0.9670936395759717
```

```
In [27]: f1
Out[27]: 0.9671904410391737
In [28]: # Encoding test_case = tf_idf_features.transform(['flying with @united is always a great experience'])
In [29]: model.predict(test_case)
Out[29]: array([2])
In []:
In []:
```