```
In [2]: import pandas as pd
    import re
    import numpy as np
    import scipy
    import itertools
    import matplotlib
    import matplotlib.pyplot as plt
    from scipy.spatial.distance import cdist
    from collections import Counter
    from random import choice
```

```
In [3]: # Storing the training and test datasets into their respective dataframes
    trained = pd.read_csv('train_clean.csv')
    test = pd.read_csv('test_clean.csv')
```

In [4]: trained.head()

Out[4]:

	Unnamed: 0	Sentiment	Tweet
0	0	neutral	amsterdam ewr
1	1	negative	ITproblems link
2	2	positive	today staff MSP took customer service new le
3	3	negative	yet receive assistance one agents securing ne
4	4	negative	let change reservation online Im wasting time

In [5]: test.head()

Out[5]:

Tweet	Sentiment	Unnamed: 0	
jump DallasAustin market News	neutral	0	0
Chicago seen seat A AA So far great ride On	positive	1	1
need bag bouncer Get together	negative	2	2
Hey Jetblue stranded entire plane supposed go	negative	3	3
Big fail curbside baggage Pittsburgh charge	negative	4	4

```
In [8]: #Training Data
    train_unique = (list(set(trained['Tweet'].str.findall("\w+").sum()))) # Finding of
    train_unique_words = len(train_unique)

#Test Data
    test_unique = (list(set(test['Tweet'].str.findall("\w+").sum()))) # Finding all of
    test_unique_words = len(test_unique)

print("Unique words in Training Data: {}".format(train_unique_words))
print("Unique words in Test Data: {}".format(test_unique_words))
Unique words in Training Data: 12416
Unique words in Test Data: 5814
```

Feature Extraction

for sentence in test['Tweet']:
 test featurevec = []

Shape of Training Matrix: (11680 , 12416) Shape of Test Matrix: (2921 , 12416)

```
In [12]: #Calculating distances between every test instance with all the train instances.
dists = cdist(test_matrix,train_matrix,'euclidean')
```

```
In [13]: #Making an empty column in our test data for predicted labels.
    test['Predicted Label'] = ''
    dists.shape

Out[13]: (2921, 11680)

In [14]: #Function that takes a list and returns the mode of the list. If there are more to def get_mode(l):
        counting = Counter(l)
        max_count = max(counting.values())
        return choice([ks for ks in counting if counting[ks] == max_count])
```

K Nearest Neighbors & Performance Measures

```
In [16]: def cmatrix measures(k, dists, test, cmatrix):
           row_count = 0
           first max = 0
           second_max = 0
           check_tie = False
           for ls in dists:
             sorted_distances_indices = np.argsort(ls) #Getting a sorted list of indices
             knn_indices = []
             knn_indices = list(itertools.islice(sorted_distances_indices,k)) #Extracting
             knn_labels = []
             for i in knn indices:
               label = trained['Sentiment'][i] #Extracting the label of the instance by ir
               knn_labels.append(label) #Appending the label to our labels list.
             max_class = get_mode(knn_labels)
             first_max = max_class
             second_max = max(knn_labels)
             if first_max == second_max:
               check_tie = True
             predicted label = max_class
             test['Predicted Label'][row_count] = predicted_label
             row count += 1
           #Creating a frequency DataFrame that will store value counts for each tuple of
           testfreqdf = test.groupby(["Sentiment", "Predicted Label"]).size().reset index
           testfreqdf
           #Extracting values from the Frequency DataFrame and assigning to specific cells
           cmatrix['Gold Positive']['Predicted Positive'] = testfreqdf['Frequency'][8]
           cmatrix['Gold Neutral']['Predicted Positive'] = testfreqdf['Frequency'][5]
           cmatrix['Gold Negative']['Predicted Positive'] = testfreqdf['Frequency'][2]
           cmatrix['Gold Positive']['Predicted Neutral'] = testfreqdf['Frequency'][7]
           cmatrix['Gold Neutral']['Predicted Neutral'] = testfreqdf['Frequency'][4]
           cmatrix['Gold Negative']['Predicted Neutral'] = testfreqdf['Frequency'][1]
           cmatrix['Gold Positive']['Predicted Negative'] = testfreqdf['Frequency'][6]
           cmatrix['Gold Neutral']['Predicted Negative'] = testfreqdf['Frequency'][3]
           cmatrix['Gold Negative']['Predicted Negative'] = testfreqdf['Frequency'][0]
           #Extracting all three True Positives from the matrix to measure accuracy.
           TP = cmatrix['Gold Positive']['Predicted Positive']
           TNT = cmatrix['Gold Neutral']['Predicted Neutral']
           TN = cmatrix['Gold Negative']['Predicted Negative']
           total = testfreqdf['Frequency'].sum()
           accuracy = ((TP+TNT+TN)/total)*100
           accuracy = round(accuracy,2)
           accuracy_list.append(accuracy)
           #Extracting all recalls from the matrix to measure macroaveraged recall.
           recall_pos = cmatrix['Gold Positive']['Predicted Positive']/cmatrix['Gold Posit
           recall_neut = cmatrix['Gold Neutral']['Predicted Neutral']/cmatrix['Gold Neutra
           recall_neg = cmatrix['Gold Negative']['Predicted Negative']/cmatrix['Gold Negat
           macroaveraged_recall = ((recall_pos+recall_neut+recall_neg)/3)*100
```

```
macroaveraged recall = round(macroaveraged recall,2)
recall_list.append(macroaveraged_recall)
#Extracting all precisions from the matrix to measure macroaveraged precision.
precision pos = cmatrix['Gold Positive']['Predicted Positive']/(cmatrix.iloc[0]
precision_neut = cmatrix['Gold Neutral']['Predicted Neutral']/(cmatrix.iloc[1,@)
precision_neg = cmatrix['Gold Negative']['Predicted Negative']/(cmatrix.iloc[2]
macroaveraged_precision = ((precision_pos+precision_neut+precision_neg)/3)*100
macroaveraged_precision = round(macroaveraged_precision,2)
precision list.append(macroaveraged precision)
#Extracting all F1_scores from the matrix to measure macroaveraged F1_score.
F1_pos = (2*precision_pos*recall_pos)/(precision_pos+recall_pos)
F1_neut = (2*precision_neut*recall_neut)/(precision_neut+recall_neut)
F1_neg = (2*precision_neg*recall_neg)/(precision_neg+recall_neg)
F1\_score = ((F1\_pos + F1\_neut + F1\_neg)/3)*100
F1 score = round(F1 score,2)
F1_list.append(F1_score)
print("\n ("\n) Matrix with k = {}:\n".format(k))
print(cmatrix)
print("\nAccuracy with k = {0}: {1}%".format(k,accuracy))
print("Macroaveraged Precision with k = \{0\}: \{1\}%".format(k, macroaveraged preci
print("Macroaveraged Recall with k = {0}: {1}%".format(k,macroaveraged_recall))
print("Macroaveraged F1-score with k = {0}: {1}%".format(k,F1 score))
```


C:\Users\aj240\AppData\Local\Temp/ipykernel_112/4185092973.py:24: SettingWithCo
pyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

test['Predicted Label'][row_count] = predicted_label

Confusion Matrix with k = 1:

		Gold	Positive	Gold	Neutral	Gold	Negative
Predicted	Positive		263		107		205
Predicted	Neutral		132		343		758
Predicted	Negative		77		165		871

Accuracy with k = 1: 50.56%

Macroaveraged Precision with k=1:50.6%Macroaveraged Recall with k=1:52.99%Macroaveraged F1-score with k=1:48.82%

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test['Predicted Label'][row_count] = predicted_label

Confusion Matrix with k = 3:

		Gold	Positive	Gold	Neutral	Gold	Negative
Predicted	Positive		273		124		235
Predicted	Neutral		135		363		783
Predicted	Negative		64		128		816

Accuracy with k = 3: 49.71%

Macroaveraged Precision with k=3:50.83%Macroaveraged Recall with k=3:53.79%Macroaveraged F1-score with k=3:48.39%

C:\Users\aj240\AppData\Local\Temp/ipykernel_112/4185092973.py:24: SettingWith
CopyWarning:

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See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

test['Predicted Label'][row_count] = predicted_label

Confusion Matrix with k = 5:

		Gold	Positive	Gold	Neutral	Gold	Negative
Predicted	Positive		272		133		250
Predicted	Neutral		138		376		814
Predicted	Negative		62		106		770

Accuracy with k = 5: 48.55%

Macroaveraged Precision with k=5: 50.64% Macroaveraged Recall with k=5: 53.58% Macroaveraged F1-score with k=5: 47.51%

C:\Users\aj240\AppData\Local\Temp/ipykernel_112/4185092973.py:24: SettingWithCo
pyWarning:

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See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

test['Predicted Label'][row_count] = predicted_label

Confusion Matrix with k = 7:

		Gold	Positive	Gold	Neutral	Gold	Negative
Predicted	Positive		275		121		247
Predicted	Neutral		153		397		871
Predicted	Negative		44		97		716

Accuracy with k = 7: 47.52%

Macroaveraged Precision with k=7:51.42% Macroaveraged Recall with k=7:53.95% Macroaveraged F1-score with k=7:47.18%

C:\Users\aj240\AppData\Local\Temp/ipykernel_112/4185092973.py:24: SettingWithCo
pyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

test['Predicted Label'][row_count] = predicted_label

Confusion Matrix with k = 10:

		Gold	Positive	Gold	Neutral	Gold	Negative
Predicted	Positive		275		112		252
Predicted	Neutral		158		424		893
Predicted	Negative		39		79		689

Accuracy with k = 10: 47.52%

Macroaveraged Precision with k = 10: 52.39% Macroaveraged Recall with k = 10: 54.92% Macroaveraged F1-score with k = 10: 47.42%

In []: