```
In [13]: | pip install -U gensim
         Collecting gensim
           Downloading gensim-4.2.0-cp39-cp39-win_amd64.whl (23.9 MB)
         Requirement already satisfied: scipy>=0.18.1 in c:\users\skand\anaconda3\lib\si
         te-packages (from gensim) (1.7.1)
         Requirement already satisfied: numpy>=1.17.0 in c:\users\skand\anaconda3\lib\si
         te-packages (from gensim) (1.20.3)
         Collecting smart-open>=1.8.1
           Downloading smart_open-6.0.0-py3-none-any.whl (58 kB)
         Collecting Cython==0.29.28
           Downloading Cython-0.29.28-py2.py3-none-any.whl (983 kB)
         Installing collected packages: smart-open, Cython, gensim
           Attempting uninstall: Cython
             Found existing installation: Cython 0.29.24
             Uninstalling Cython-0.29.24:
               Successfully uninstalled Cython-0.29.24
         Successfully installed Cython-0.29.28 gensim-4.2.0 smart-open-6.0.0
         Note: you may need to restart the kernel to use updated packages.
In [25]: import nltk
In [26]: | nltk.download('vader lexicon')
         [nltk_data] Downloading package vader_lexicon to
         [nltk data] C:\Users\skand\AppData\Roaming\nltk data...
Out[26]: True
In [15]: pip show gensim
         Name: gensim
         Version: 4.2.0
         Summary: Python framework for fast Vector Space Modelling
         Home-page: http://radimrehurek.com/gensim (http://radimrehurek.com/gensim)
         Author: Radim Rehurek
         Author-email: me@radimrehurek.com
         License: LGPL-2.1-only
         Location: c:\users\skand\anaconda3\lib\site-packages
         Requires: smart-open, scipy, Cython, numpy
         Required-by:
         Note: you may need to restart the kernel to use updated packages.
```

```
In [17]: | %matplotlib inline
         import re
         import sqlite3
         import pandas as pd
         import numpy as np
         import nltk
         import tqdm as tqdm
         import string
         from nltk.corpus import stopwords
         stop = stopwords.words("english")
         import matplotlib.pyplot as plt
         import numpy as np
         import datetime as dt
         from sklearn.feature extraction.text import TfidfTransformer
         from sklearn.feature extraction.text import CountVectorizer
         from sklearn.model_selection import train_test_split
         from sklearn.metrics import confusion_matrix
         from sklearn import metrics
         from sklearn.metrics import roc_curve, auc
         from nltk.stem.porter import PorterStemmer
         english_stemmer=nltk.stem.SnowballStemmer('english')
         from nltk.tokenize import word_tokenize
         from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.model selection import train test split
         from sklearn.metrics import accuracy score
         from sklearn.metrics import confusion matrix
         from math import floor,ceil
         from sklearn.svm import LinearSVC
         from keras.models import Sequential
         from keras.layers import LSTM, Dense, Embedding
         import seaborn as sns
```

```
In [2]: | df = pd.read excel(r'C:\Users\skand\Downloads\interns-tagging Skandha.xlsx')
        print (df)
             cluster id
                                                       id
                        0334a0d055104e9a931c079e338be9a1
        0
                      0
        1
                      0 796d6c25ab8849cbba427f1f3e250d80
                      0 661f5299cd8944a8a3841fd4f049dee9
        2
                      0 da831e4bc58d4505aec3c583f0248f8b
        4
                      0 0ea997675e7344419d1540d3e0bc26c3
        599
                     10 93f874167d11473f8d36d1cda0a0081c
        600
                     10 d50fe37fab064408a891aa9ef45dcd70
        601
                     10 3e1e8901d5ab4fc9b602ecfdca1220cb
        602
                     10 c84e1b1196a242d18938af6c60403afc
        603
                     10 fd4c71f399104d59ad6c1013fc414c67
                                                        phrase
                                                                   common idea
        0
                   Would use the product again if needed Joe . loyal customer
                    Have been using the product for a week now loyal customer
        1
                                                                loyal customer
        2
             Will continue to use this product when I have ...
                 Have always had good luck with this product . loyal customer
             Will continue to use This product as This prod... loyal customer
        4
        599
                                      Spray has no strong odor
                                                                   good smell
        600 Spray is nice to keep out on the porch on a su...
                                                                   good smell
                                                                good smell
        601
                   Spray does not leave any oily stinky stains
        602 Love that the scent of this spray is not chemi...
                                                                   good smell
        603
                         Bug spray does not smell nauseating .
                                                                   good smell
```

Data Pre Processing

[604 rows x 4 columns]

Cleaning

```
In [18]: clean_Text = []
for review in df['phrase']:
        clean_Text.append( " ".join(cleaning(review)))

clean_summary = []
for review in df['common idea']:
        clean_summary.append( " ".join(cleaning(review)))
```

Top Word Count In Text(Review)

```
In [19]: Top_Words_Review =pd.Series(' '.join(clean_Text).lower().split()).value_counts()|
         print ("Top Count Words Used In Review", Top_Words_Review)
         Top Count Words Used In Review product
                                                       102
         safe
         use
                        79
         smell
                        72
         around
                        57
         pest
                        47
         recommend
                        46
         pet
                        38
         good
                        38
         great
                        36
         dtype: int64
In [20]: Top_Words_Summary = pd.Series(' '.join(clean_summary).lower().split()).value_cour
         print ("Top Count Words Used In Summary", Top_Words_Summary)
         Top Count Words Used In Summary effici
                                                      142
         product
                     142
         good
                     141
         smell
                     141
         safe
                     117
         kid
                     117
         pet
                     117
         loyal
                      62
                      62
         custom
         valu
                      61
         dtype: int64
```

Tf-idf

```
In [23]: from sklearn.feature_extraction.text import TfidfVectorizer
    vectorizer = TfidfVectorizer(min_df=4, max_features = 10000)
    vz = vectorizer.fit_transform(clean_Text)
    tfidf = dict(zip(vectorizer.get_feature_names(), vectorizer.idf_))
```

Sentiment Analysis

```
In [27]: | from nltk.sentiment.vader import SentimentIntensityAnalyzer
         Senti = SentimentIntensityAnalyzer()
         sample_review = clean_Text[:5]
         for sentence in sample_review:
             sentence
             ss = Senti.polarity_scores(sentence)
             for k in sorted(ss):
                  print('{0}: {1}, '.format(k, ss[k]))
             print(sentence)
         compound: 0.0,
         neg: 0.0,
         neu: 1.0,
         pos: 0.0,
         would use product need joe
         compound: 0.0,
         neg: 0.0,
         neu: 1.0,
         pos: 0.0,
         use product week
         compound: 0.0,
         neg: 0.0,
         neu: 1.0,
         pos: 0.0,
         continu use product issu
         compound: 0.7096,
         neg: 0.0,
         neu: 0.253,
         pos: 0.747,
         alway good luck product
         compound: 0.0,
         neg: 0.0,
         neu: 1.0,
         pos: 0.0,
         continu use product product get job done
```

K means

```
In [28]: | from sklearn.cluster import MiniBatchKMeans
         num_clusters = 10
         kmeans_model = MiniBatchKMeans(n_clusters=num_clusters, init='k-means++', n_init=
                                   init_size=1000, batch_size=1000, verbose=False, max_iter
         kmeans = kmeans_model.fit(vz)
         kmeans_clusters = kmeans.predict(vz)
         kmeans_distances = kmeans.transform(vz)
         sorted_centroids = kmeans.cluster_centers_.argsort()[:, ::-1]
         terms = vectorizer.get_feature_names()
         for i in range(num_clusters):
             print("Cluster %d:" % i)
             for j in sorted_centroids[i, :5]:
                  print(' %s' % terms[j])
             print()
         Cluster 0:
          recommend
          would
          friend
          product
          anyon
         Cluster 1:
          safe
          around
          kid
          children
          use
         Cluster 2:
           smell
          chemic
          work
          bad
          spray
         Cluster 3:
          odor
          product
          strong
          bad
          effect
         Cluster 4:
          price
          great
          good
          worth
          afford
         Cluster 5:
          pet
          friend
          safe
          stuff
```

	good
	Cluster 6: scent strong lemon fresh nice
	Cluster 7: use product year sever continu
	Cluster 8: bug ant deliveri valu fast
	Cluster 9: pest control awesom effect keep
In []:	
In []:	
In []:	