```
In [1]:
#pip install contractions
In [2]:
import pandas as pd
import numpy as np
import nltk
nltk.download('wordnet')
nltk.download('stopwords')
nltk.download('omw-1.4')
import re
from bs4 import BeautifulSoup
import contractions
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.model selection import train test split
from sklearn.metrics import classification_report
from sklearn.metrics import confusion matrix
from \ sklearn.metrics \ import \ precision\_recall\_fscore\_support \ as \ score
[nltk_data] Downloading package wordnet to
                C:\Users\USER\AppData\Roaming\nltk data...
[nltk_data]
[nltk data]
              Package wordnet is already up-to-date!
[nltk_data] Downloading package stopwords to
                C:\Users\USER\AppData\Roaming\nltk_data...
[nltk data]
[nltk_data]
              Package stopwords is already up-to-date!
[nltk\_data] \ Downloading \ package \ omw-1.4 \ to
                C:\Users\USER\AppData\Roaming\nltk_data...
[nltk_data]
              Package omw-1.4 is already up-to-date!
[nltk_data]
In [ ]:
```

Read Data

```
In [3]:

df= pd.read_table('amazon_reviews_us_Beauty_v1_00.tsv',on_bad_lines='skip',low_memory=False)
```

Keep Reviews and Ratings

```
In [4]:
df_fnl= df[['review_body','star_rating']]
In [5]:
df_fnl
Out[5]:
                                              review body star rating
        0
                              Love this, excellent sun block!!
                                                                       5
        1
                The great thing about this cream is that it do...
                                                                       5
        2
                Great Product, I'm 65 years old and this is al...
                                                                       5
            I use them as shower caps & conditioning caps....
                                                                       5
        4
                This is my go-to daily sunblock. It leaves no ...
                                                                       5
 5094302
              After watching my Dad struggle with his scisso...
                                                                       5
                                                                       3
 5094303 Like most sound machines, the sounds choices a...
 5094304
               I bought this product because it indicated 30 ...
 5094305
             We have used Oral-B products for 15 years; thi...
```

We form three classes and select 20000 reviews randomly from each class.

```
In [6]:

df_final=df_fnl.replace({'star_rating':{2:1,3:2,4:3,5:3,'5':3,'2':1,'3':2,'4':3,'1':1}})
```

```
In [7]:
df_final.isnull().sum()
Out[7]:
review_body
                  400
star_rating
                   10
dtype: int64
In [8]:
df_final.dropna(inplace=True)
In [9]:
s0 = df_final[df_final['star_rating'].eq(1)].sample(20000).index
s1 = df_final[df_final['star_rating'].eq(2)].sample(20000).index
s2 = df_final[df_final['star_rating'].eq(3)].sample(20000).index
In [10]:
df_fi = df_final.loc[s0.union(s1).union(s2)]
In [11]:
df_fi
Out[11]:
                                      review_body star_rating
      78 Like all of Dove's Men+ line these are good bu...
                                                            2
     233
                                                            3
            I love this product! It always leaves my hair ...
     367
                                                            3
     527
          I ordered this almost a year ago to replace my...
                                                            1
     539
                               I dont like this product.
 5094018
                                                            2
           The bubble spa is very very loud, the air comi...
 5094035
            the Quiet mode was very loud!....and I still h...
 5094066
            This hair dryer is maybe only slightly quieter...
 5094081
           It works for less than a minute and then the b...
In [12]:
df_fi.isnull().sum()
                  0
review_body
star_rating
dtype: int64
```

Data Cleaning

Pre-processing

```
In [13]:

dt_before_clean=df_fi['review_body'].apply(len).mean()

In [15]:

def remove_alphanumeric(s):
    s=s.lower()
    s=s.strip()
    s=contractions.fix(s)
    s=s.replace(r'<[^<>\}^*\>', '')
    s=s.replace(r'http\s+', '').replace(r'www\S+', '')
    s=re.sub(r'[^a-zA-Z]', '',s)
    return re.sub(' +', '',s)
In [16]:

df_fi['review_body']=df_fi['review_body'].apply(remove_alphanumeric)
```

```
In [17]:

dt_after_clean=df_fi['review_body'].apply(len).mean()

In [18]:

print("Average length of reviews before and after data cleaning:",dt_before_clean,',',dt_after_clean)

Average length of reviews before and after data cleaning: 269.4664 , 261.4241666666667

In [ ]:
```

remove the stop words

```
In [19]:

dt_before_preprocess=df_fi['review_body'].apply(len).mean()

In [20]:

# from nltk.corpus import stopwords
# stop = stopwords.words('english')
# df_fi['review_body'] = df_fi['review_body'].apply(lambda x: ' '.join([word for word in x.split() if word not in (stop)]))
```

```
In [21]:
df_fi
Out[21]:
                                             review_body star_rating
                                                                      2
            like all of dove s men line these are good but...
      233
             i love this product it always leaves my hair s...
                                                                      3
      367
                                        wonderful product
      527 i ordered this almost a year ago to replace my...
      539
                                  i do not like this product
 5094018 the bubble spa is very very loud the air comin...
 5094035 the guiet mode was very loud and i still had a...
              this hair dryer is maybe only slightly quieter...
```

perform lemmatization

5094081 it works for less than a minute and then the b...

```
In [22]:
```

```
from nltk.stem import WordNetLemmatizer
lemmatizer=WordNetLemmatizer()
def lemmatize_words(text):
    words = text.split()
    words = [lemmatizer.lemmatize(word,pos='v') for word in words]
    return ' '.join(words)

df_fi['review_body']=df_fi['review_body'].apply(lemmatize_words)
```

```
In [23]:
df_fi
                                   review_body star_rating
                                                        2
     78
           like all of dive s men line these be good but ...
    233
                                                        3
           i love this product it always leave my hair so...
    367
                                wonderful product
                                                        3
    527
          i order this almost a year ago to replace my f...
    539
                            i do not like this product
 5094018 the bubble spa be very very loud the air come ...
 5094035
          the quiet mode be very loud and i still have a...
5094066
           this hair dryer be maybe only slightly quieter...
5094081
          it work for less than a minute and then the bl...
 5094261 the three speed be fine and the unit generate ...
In [24]:
dt_after_preprocess=df_fi['review_body'].apply(len).mean()
In [25]:
print("Average length of reviews before and after data preprocessing:",dt_before_preprocess,',',dt_after_preprocess)
Average length of reviews before and after data preprocessing: 261.4241666666667 , 251.40098333333333
TF-IDF Feature Extraction
In [26]:
imp_features = TfidfVectorizer(ngram_range=(1,3))
x = imp_features.fit_transform(df_fi['review_body'])
In [27]:
X_train, X_test, Y_train, Y_test = train_test_split(x, df_fi['star_rating'], test_size=0.2, random_state=42)
Y_train=Y_train.astype('int')
Y_test=Y_test.astype('int')
Perceptron
In [28]:
from sklearn.linear_model import Perceptron
clf_percep=Perceptron(tol=1e-3, random_state=0)
clf_percep.fit(X_train,Y_train)
Out[28]:
 ▼ Perceptron
Perceptron()
In [29]:
predicted\_perceptron = clf\_percep.predict(X\_test)
In [30]:
target_names = ['class 1', 'class 2', 'class 3']
\verb|print(classification_report(Y_test, predicted_perceptron, target_names=target_names)||
                             recall f1-score
               precision
                                                  support
     class 1
                     0.72
                                0.73
                                           0.72
                                                      4021
                                                      3909
     class 2
                     0.63
                                0.61
                                           0.62
                     0.80
                                                      4070
     class 3
                                0.81
                                           0.81
```

0.72

0.72

0.72

0.72

0.72

0.71

0.72

accuracy

macro avg

weighted avg

12000

12000

12000

```
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                                                               HW1-CSCI544 - Jupyter Notebook
 In [31]:
 precision, recall, fscore, support = score(Y_test, predicted_perceptron)
 # print('precision: {}'.format(precision))
 # print('recall: {}'.format(recall))
# print('fscore: {}'.format(fscore))
# print('support: {}'.format(support))
 In [32]:
 print('------')
 for i in range(3):
     print('class '+ str(i+1),'Precision:',precision[i],',','Recall:', recall[i],',', 'F1 score:',fscore[i],',','average:',((precision[i]+
  -----Perceptron Classification------
 class 1 Precision: 0.7201382033563672 , Recall: 0.7256901268341208 , F1 score: 0.7229035055122012 , average: 0.722910611900
 8963
 class 2 Precision: 0.6285034373347436 , Recall: 0.6080839089281146 , F1 score: 0.6181250812638149 , average: 0.618237475842
 class 3 Precision: 0.7959673547767643 , Recall: 0.8147420147420148 , F1 score: 0.805245264691598 , average: 0.8053182114034
 SVM
 In [33]:
 from sklearn.svm import LinearSVC
 clf_SVM= LinearSVC(random_state=42, tol=1e-5)
 {\tt clf\_SVM.fit}(X\_{\tt train,Y\_{\tt train}})
 Out[331:
                LinearSVC
  LinearSVC(random state=42, tol=1e-05)
 In [34]:
 predicted_SVM=clf_SVM.predict(X_test)
 In [35]:
 target names = ['class 1', 'class 2', 'class 3']
 print(classification_report(Y_test, predicted_SVM, target_names=target_names))
               precision
                           recall f1-score
                                             support
      class 1
                   0.75
                             0.75
                                      0.75
                                                4021
      class 2
                   0.65
                             0.64
                                      0.65
                                                3909
      class 3
                   0.82
                             0.84
                                      0.83
                                                4070
                                      0.74
     accuracy
                                               12000
                   0.74
                             0.74
    macro avg
                                      0.74
                                               12000
```

```
weighted avg
                   0.74
                              0.74
                                        0.74
                                                 12000
```

In [36]:

```
precision, recall, fscore, support = score(Y_test, predicted_SVM)
print('precision: {}'.format(precision))
print('recall: {}'.format(recall))
print('fscore: {}'.format(fscore))
print('support: {}'.format(support))
```

precision: [0.75099305 0.65316257 0.82103232] recall: [0.75230042 0.63929394 0.83636364] fscore: [0.75164617 0.64615385 0.82862707] support: [4021 3909 4070]

```
In [37]:
for i in range(3):
      print('class '+ str(i+1),'Precision:',precision[i],',','Recall:', recall[i],',', 'F1 score:',fscore[i],',','average:',((precision[i]+
------Support Vector Machine Classification------
class 1 Precision: 0.7509930486593843 , Recall: 0.7523004227804029 , F1 score: 0.7516461672257424 , average: 0.751646546221
8432
class 2 Precision: 0.6531625718766335 , Recall: 0.639293937068304 , F1 score: 0.6461538461538462 , average: 0.6462034516995
945
class 3 Precision: 0.8210323203087313 , Recall: 0.83636363636363 , F1 score: 0.8286270691333981 , average: 0.828674341935
2552
Logistic Regression
In [38]:
from sklearn.linear model import LogisticRegression
clf_Logistic= LogisticRegression(random_state=42,max_iter=1000000)
clf_Logistic.fit(X_train,Y_train)
Out[38]:
                              LogisticRegression
LogisticRegression(max_iter=1000000, random_state=42)
In [39]:
predicted_Log=clf_Logistic.predict(X_test)
In [40]:
target_names = ['class 1', 'class 2', 'class 3']
print(classification_report(Y_test, predicted_Log, target_names=target_names))
                      precision
                                            recall f1-score
                                                                           support
        class 1
                               0.75
                                               0.74
                                                                0.75
                                                                                4021
        class 2
                               0.64
                                               0.66
                                                                0.65
                                                                                3909
                               0.83
                                               0.82
                                                                                4070
        class 3
                                                                0.82
      accuracy
                                                                0.74
                                                                               12000
     macro avg
                               0.74
                                               0.74
                                                                0.74
                                                                               12000
weighted avg
                               0.74
                                               0.74
                                                                0.74
                                                                              12000
In [41]:
precision, recall, fscore, support = score(Y_test, predicted_Log)
print('precision: {}'.format(precision))
print('recall: {}'.format(recall))
print('fscore: {}'.format(fscore))
print('support: {}'.format(support))
precision: [0.74743686 0.64492571 0.82679901]
recall: [0.74334743 0.65515477 0.81867322]
fscore: [0.74538653 0.65
                                                  0.82271605]
support: [4021 3909 4070]
In [42]:
print('-------)
for i in range(3):
      print('class '+ str(i+1),'Precision:',precision[i],',','Recall:', recall[i],',', 'F1 score:',fscore[i],',','average:',((precision[i]+
-----Logistic Regression Classification------
class 1 Precision: 0.7474368592148037 , Recall: 0.7433474260134295 , F1 score: 0.7453865336658354 , average: 0.745390272964
class \ 2 \ Precision: \ 0.6449257114077058 \ , \ Recall: \ 0.655154771041187 \ , \ F1 \ score: \ 0.65 \ , \ average: \ 0.6500268274829643 \ , \ average: \ 0.65002682748299643 \ , \ average: \ 0.65002682748299643 \ , \ average: \ 0.650026827482
class 3 Precision: 0.8267990074441688 , Recall: 0.8186732186732186 , F1 score: 0.8227160493827161 , average: 0.822729425166
```

Naive Bayes

```
In [43]:
from sklearn.naive bayes import MultinomialNB
clf naive = MultinomialNB()
clf_naive.fit(X_train, Y_train)
Out[43]:
▼ MultinomialNB
MultinomialNB()
In [44]:
predicted_naive = clf_naive.predict(X_test)
In [45]:
conf=confusion_matrix(Y_test,predicted_naive)
In [46]:
target_names = ['class 1', 'class 2', 'class 3']
print(classification_report(Y_test, predicted_naive, target_names=target_names))
             precision
                         recall f1-score
                                          support
    class 1
                 0.77
                           0.71
                                    0.74
                                             4021
    class 2
                 0.58
                           0.77
                                    0.66
                                             3909
    class 3
                 0.90
                           0.69
                                    0.78
                                             4070
   accuracy
                                    0.72
                                             12000
                 0.75
                           0.72
                                    0.73
   macro avg
                                            12000
                                            12000
weighted avg
                 0.75
                           0.72
                                    0.73
In [ ]:
In [47]:
precision, recall, fscore, support = score(Y_test, predicted_naive)
print('precision: {}'.format(precision))
print('recall: {}'.format(recall))
print('fscore: {}'.format(fscore))
print('support: {}'.format(support))
precision: [0.77223427 0.57868314 0.89626556]
recall: [0.70828152 0.76669225 0.68992629]
fscore: [0.73887664 0.65955106 0.77967514]
support: [4021 3909 4070]
In [ ]:
In [48]:
print('-----
                            -----')
for i in range(3):
   print('class '+ str(i+1),'Precision:',precision[i],',','Recall:', recall[i],',', 'F1 score:',fscore[i],',','average:',((precision[i]+
-----Naive Bayes Classification------
class 1 Precision: 0.7722342733188721 , Recall: 0.7082815220094504 , F1 score: 0.7388766376961993 , average: 0.739797477674
8405
class 2 Precision: 0.5786831434639892 , Recall: 0.7666922486569455 , F1 score: 0.6595510563380282 , average: 0.668308816152
9875
class 3 Precision: 0.8962655601659751 , Recall: 0.6899262899262899 , F1 score: 0.7796751353602667 , average: 0.788622328484
1772
In [ ]:
```