# Reviews

December 16, 2023

## 0.0.1 Data Gathering

Downloading the data

```
[1]: | !kaggle datasets download -d d4rklucif3r/restaurant-reviews
```

restaurant-reviews.zip: Skipping, found more recently modified local copy (use --force to force download)

Extracting the compressed file

The dataset is extracted

### 0.0.2 Importing the Dependencies

```
[3]: import pandas as pd
import numpy as np
from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
import re
```

Creating a dataframe

```
[4]: tsv_file_path = ('C:\\Users\\tejus\\Desktop\\6th Sem Project\\Restaurant

→Reviews (NLP)\\Restaurant_Reviews.tsv')

r_reviews = pd.read_csv(tsv_file_path, sep='\t')
```

```
[5]: r_reviews.size
```

```
[5]: 2000
 [6]: r_reviews.shape
 [6]: (1000, 2)
 [7]: r_reviews.head()
 [7]:
                                                      Review Liked
      0
                                   Wow... Loved this place.
      1
                                         Crust is not good.
                                                                  0
      2
                 Not tasty and the texture was just nasty.
                                                                  0
      3 Stopped by during the late May bank holiday of ...
      4 The selection on the menu was great and so wer...
 [8]: r_reviews.isnull().sum()
 [8]: Review
                0
      Liked
      dtype: int64
 [9]: r_reviews['Liked'].value_counts()
 [9]: Liked
      1
           500
      0
           500
      Name: count, dtype: int64
[10]: r_reviews.head()
[10]:
                                                      Review Liked
                                   Wow... Loved this place.
      0
                                         Crust is not good.
                                                                  0
      1
                 Not tasty and the texture was just nasty.
                                                                  0
      3 Stopped by during the late May bank holiday of ...
                                                                1
      4 The selection on the menu was great and so wer...
     0.0.3 Data Preprocessing
     Stemming converts a word to its root word
[11]: def stemming(content):
          stemmed_content = re.sub('[^a-zA-Z]', ' ', content)
          stemmed_content = stemmed_content.lower()
          stemmed_content = stemmed_content.split(sep=' ')
          stemmed_content = [PorterStemmer().stem(word) for word in stemmed_content_
```

→if word not in stopwords.words('english')]

stemmed\_content = ' '.join(stemmed\_content)

```
return stemmed_content
     creating a new column for the stemmed reviews
[12]: r_reviews['Stemmed_Reviews'] = r_reviews['Review'].apply(stemming)
[13]: r_reviews.head()
[13]:
                                                      Review Liked \
                                   Wow... Loved this place.
      0
                                                                1
      1
                                         Crust is not good.
                                                                  0
      2
                 Not tasty and the texture was just nasty.
                                                                  0
      3 Stopped by during the late May bank holiday of ...
                                                                1
      4 The selection on the menu was great and so wer...
                                             Stemmed_Reviews
      0
                                                 love place
                                         WOW
      1
                                                 crust good
      2
                                        tasti textur nasti
      3 stop late may bank holiday rick steve recommen...
      4
                                   select menu great price
     0.0.4 Splitting the data into Training set and Testing set
[14]: X = r_reviews['Stemmed_Reviews'].values
      Y = r_reviews['Liked'].values
[15]: X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.
       →2,stratify=Y,random_state=2)
     Converting words into vectors
[16]: vectorizer = TfidfVectorizer()
     Saving the vectorizer to be used in the app
[17]: X_train = vectorizer.fit_transform(X_train)
      X_test = vectorizer.transform(X_test)
      import pickle
      filename = 'vectorizer.pkl'
      pickle.dump(vectorizer, open(filename, 'wb'))
[18]: print(X_train)
       (0, 892)
                      0.3292152335601234
       (0, 365)
                      0.3905782593861427
       (0, 30)
                      0.3905782593861427
```

```
(0, 1372)
              0.28602868670508325
(0, 1369)
              0.35186250062333335
(0, 300)
              0.30656796649691337
(0, 391)
              0.290499474797314
(0, 392)
              0.30656796649691337
(0, 76)
              0.2780357843366841
(0, 478)
              0.176729367184008
(1, 311)
              0.6913616616002672
(1, 834)
              0.651273787246651
(1, 478)
              0.31282823867854653
(2, 1168)
              0.508602366570897
(2, 566)
              0.4174849217841828
(2, 680)
              0.42869679219480356
(2, 615)
              0.508602366570897
(2, 239)
              0.35294853149107464
(3, 1131)
              0.5525769612282082
(3, 1250)
              0.5205363708646262
(3, 1204)
              0.5525769612282082
(3, 107)
              0.34403385072792303
(4, 72)
              0.3256194422164578
(4, 142)
              0.3256194422164578
(4, 1108)
              0.29135594162429207
    :
(795, 1190)
              0.3053019226759833
(795, 1229)
              0.3402973136508531
(795, 985)
              0.255638003441639
(795, 590)
              0.3313973884561668
(795, 513)
              0.26528290970241614
(795, 914)
              0.18205428020544953
(795, 525)
              0.19845710492961235
(796, 569)
              0.792390353732586
(796, 745)
              0.6100143664796324
(797, 1320)
              0.6668466950220887
(797, 977)
              0.5794663922337555
(797, 1057)
              0.468544753048962
(798, 1001)
              0.3768528122914655
(798, 417)
              0.35500140075386427
(798, 897)
              0.33949757753596876
(798, 1277)
              0.35500140075386427
(798, 54)
              0.2901166432215648
(798, 145)
              0.33949757753596876
(798, 874)
              0.33949757753596876
(798, 693)
              0.20905858539086558
(798, 1322)
              0.25276140846606804
(798, 239)
              0.26151991305768607
(799, 1116)
              0.6556077434275575
(799, 1073)
              0.5526064269822653
(799, 146)
              0.514591705739337
```

## [19]: print(X\_test)

```
(0, 1012)
              1.0
(1, 1322)
              1.0
(2, 1236)
              0.423284623964311
(2, 1076)
              0.3760578251624208
(2, 291)
              0.6227159389161827
(2, 76)
              0.5400328681378932
(3, 74)
              1.0
(4, 1225)
              0.5199124394067061
(4, 492)
              0.40808338783594494
(4, 233)
              0.41684062678974443
(4, 207)
              0.48976586677556594
(4, 163)
              0.38669405415860425
(5, 239)
              0.7810980019325923
(5, 74)
              0.6244084491556086
(6, 1292)
              0.4633955888426175
(6, 913)
              0.37961404738057924
(6, 525)
              0.2621377899243931
(6, 467)
              0.5332730613009541
(6, 163)
              0.39663125253043513
(6, 23)
              0.3615867447417573
(7, 1175)
              0.7366661152982865
(7, 538)
              0.1998466746585458
(7, 525)
              0.18105923718083705
(7, 426)
              0.36833305764914326
(7, 414)
              0.3104649342041163
(193, 406)
              0.5445116292761637
(193, 397)
              0.5129386986511408
(194, 1177)
              0.8619831211522782
(194, 1076)
              0.5069369771959595
(195, 1346)
              0.36547429159010664
(195, 1236)
              0.2696614628437105
(195, 1078)
              0.43539042746189216
(195, 1075)
              0.33957759871549603
(195, 786)
              0.35392845770798087
(195, 582)
              0.41996800674984397
(195, 402)
              0.33957759871549603
(195, 74)
              0.2681082660987881
(196, 914)
              0.6391742977803287
(196, 538)
              0.7690619071680924
(197, 1107)
              0.6185806414059921
(197, 1076)
              0.398311190032479
(197, 61)
              0.6772785143297118
(198, 1150)
              0.39111119944149897
(198, 1061)
              0.4872094265555146
(198, 918)
              0.517198642547495
```

```
(198, 404)0.33984429608443073(198, 107)0.32200734571230993(198, 23)0.35068745660501993(199, 1073)0.6962561726198929(199, 604)0.7177933839822555
```

#### 0.0.5 Model creation and Training

```
[20]: model = LogisticRegression(max_iter=20)
```

Training the Model

```
[21]: model.fit(X_train,Y_train)
```

[21]: LogisticRegression(max\_iter=20)

## 0.0.6 Evaluating the performance of the Model

```
[22]: X_train_prediction = model.predict(X_train)
Accuracy = accuracy_score(Y_train, X_train_prediction)
print(f"The accuracy on training data is {Accuracy}")
```

The accuracy on training data is 0.95125

```
[23]: X_test_prediction = model.predict(X_test)
Accuracy = accuracy_score(Y_test, X_test_prediction)
print(f"The accuracy on test data is {Accuracy}")
```

The accuracy on test data is 0.775

## 0.0.7 Saving the model for further use

```
[24]: import pickle

filename = 'trained_model.sav'
pickle.dump(model, open(filename, 'wb'))
```