TSNE_TF-IDF

July 7, 2018

1 TSNE TF-IDF Visualization For Amazon Fine Food Reviews

This Dataset conists of reviews of fine foods from amazon. which includes: - Reviews from Oct 1999 - Oct 2012 - Total of 568,454 reviews - Given by 256,059 users - For 74,258 products

2 Data Cleaning and Loading

The same data is cleaned, by removing the duplicates and the reviews for which HelpfulnessNumerator is greater than HelpfulnessDenominator. So it is reduced to 364171 reviews with same 10 columns. This data with 364171 reviews is stored in a SQLite Database named 'final_sqlite' and the table for these reviews is 'Reviews'.

We load the data using SQLite in to pandas dataframe

```
In [1]: import sqlite3
        import pandas as pd
        conn = sqlite3.connect('final_sqlite')
        data = pd.read_sql_query('''select * from Reviews ''', conn)
        print(data.shape)
(364171, 12)
In [2]: def convert(x):
             '''To convert the reviews to positive(1) or negative(0)'''
            if x<3:
                return 0
            else:
                return 1
        score = data['Score'].map(convert)
        print(data.shape)
        print(score.shape)
        print(data.head())
(364171, 12)
(364171,)
                                                                ProfileName \
    index
              Ιd
                  ProductId
                                       UserId
```

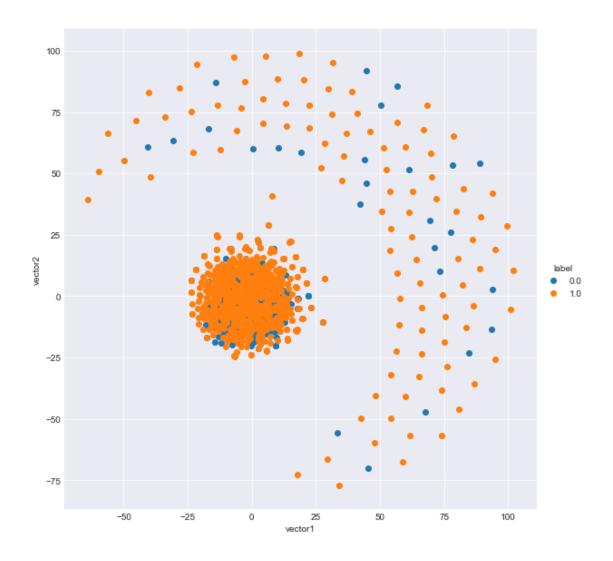
```
138706 150524
                  0006641040
                                ACITT7DI6IDDL
                                                           shari zychinski
1
  138688 150506
                  0006641040 A2IW4PEEK02R0U
                                                                     Tracy
2
  138689
          150507
                   0006641040 A1S4A3IQ2MU7V4
                                                     sally sue "sally sue"
                                               Catherine Hallberg "(Kate)"
3
  138690
          150508
                   0006641040
                                  AZGXZ2UUK6X
  138691 150509
                  0006641040 A3CMRKGE0P909G
                                                                     Teresa
  HelpfulnessNumerator
                         HelpfulnessDenominator
                                                              Time
0
                                              0
                                                         939340800
1
                      1
                                              1
                                                        1194739200
2
                      1
                                              1
                                                     4
                                                        1191456000
3
                                              1
                                                     5
                                                        1076025600
                      1
4
                                              4
                                                        1018396800
                      3
                                      Summary
0
                    EVERY book is educational
  Love the book, miss the hard cover version
1
2
                chicken soup with rice months
3
       a good swingy rhythm for reading aloud
              A great way to learn the months
4
                                                Text \
  this witty little book makes my son laugh at 1...
  I grew up reading these Sendak books, and watc...
2 This is a fun way for children to learn their ...
3 This is a great little book to read aloud- it ...
 This is a book of poetry about the months of t...
                                         ClearedText
 b'witti littl book make son laugh recit car dr...
 b'grew read sendak watch realli rosi movi inco...
2 b'fun way children learn month year learn poem...
3 b'great littl book read nice rhythm well good ...
4 b'book poetri month goe month cute littl poem ...
```

Here we determine a review as Positive or Negative by using the score. If score is more than 3 then it is considered as a positive and negative if it is less than 3 and will ignore if score is 3, as we can't decide whether it will fall into positive or negative category. The data which is in the Reviews table is queried/saved without the reviews with score 3.

3 TF-IDF

```
(364171, 115281)
(364171,)
In [6]: from sklearn.preprocessing import StandardScaler
        features = StandardScaler(with_mean=False).fit_transform(features)
        print(features.shape)
(364171, 115281)
In [7]: f_1k = features[0:1000].todense()
        s_1k = score[0:1000]
        print(f_1k.shape)
(1000, 115281)
   TSNE Visualization of TF-IDF
In [9]: from sklearn.manifold import TSNE
        model = TSNE(n_components =2, random_state = 0)
        tsne_data = model.fit_transform(f_1k)
        print(tsne_data.shape)
(1000, 2)
  We choose 1000 records to visualize data using TSNE
In [10]: import numpy as np
         c_data = np.vstack((tsne_data.T, s_1k)).T
         df = pd.DataFrame(c_data, columns =( 'vector1', 'vector2', 'label'))
        print(df.shape)
(1000, 3)
In [12]: import seaborn as sns
         import matplotlib.pyplot as plt
         sns.set_style('darkgrid')
         sns.FacetGrid(df, hue = 'label', size =8).map(plt.scatter,'vector1', 'vector2').add_le
         plt.show()
```

print(score.shape)



We choose 2000 records to visualize data using TSNE

```
In [15]: c_data = np.vstack((tsne_data.T, s_2k)).T
         df = pd.DataFrame(c_data, columns =( 'vector1', 'vector2', 'label'))
         print(df.shape)
(2000, 3)
In [16]: sns.set_style('darkgrid')
         sns.FacetGrid(df, hue = 'label', size =8).map(plt.scatter,'vector1', 'vector2').add_left
         plt.show()
       100
        75
       50
                                                                                 0.0
        0
       -25
       -50
             -100
                             -50
```

5 Observations:

Both the plots doesnot give much variations and we cannot bifercate or differentiate whether a review is positive or negative. Both the reviews are spread accross the graph. So we will proceed with other approaches and check whether they can be bifercated.