# 3. Post Clean TF-IDF

June 1, 2018

### 1 AAI Course: TSNE Visualization (Part III)

#### 1.1 TF-IDF & t-SNE

Data Source: The preprocessing step has produced final.sqlite file after doing the data preparation & clearning. The review text is now devoid of punctuations, HTML markups and stop words.

Objective: To plot t-SNE plot after doing TF-IDF & Truncated SVD for dimensionality reduction. The aim is to check whether there is a separation between data points.

### 1.2 Preprocessed Data Loading

```
In [1]: import sqlite3
        import pdb
        import pandas as pd
        import numpy as np
        import nltk
        import string
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.feature_extraction.text import TfidfTransformer
        from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.metrics import confusion_matrix
        from sklearn import metrics
        from sklearn.metrics import roc_curve, auc
        from nltk.stem.porter import PorterStemmer
        # using the SQLite Table to read data.
        con = sqlite3.connect('./final.sqlite')
        #filtering only positive and negative reviews i.e.
        # not taking into consideration those reviews with Score=3
        final = pd.read_sql_query("""
```

```
SELECT *
        FROM Reviews
        """, con)
        print(final.head(3))
    index
               Ιd
                    ProductId
                                       UserId
                                                          ProfileName
 138706
           150524
                   0006641040
                                ACITT7DI6IDDL
                                                      shari zychinski
  138688
           150506
                   0006641040
                               A2IW4PEEKO2ROU
                                                                Tracy
  138689
          150507
                   0006641040
                               A1S4A3IQ2MU7V4
                                               sally sue "sally sue"
                        HelpfulnessDenominator
  HelpfulnessNumerator
                                                     Score
                                                                  Time
                                                 positive
                                                             939340800
0
                      0
                      1
                                                 positive
1
                                                            1194739200
2
                      1
                                                 positive
                                                           1191456000
                                      Summary
0
                    EVERY book is educational
  Love the book, miss the hard cover version
                chicken soup with rice months
2
                                                 Text \
0 this witty little book makes my son laugh at 1...
1 I grew up reading these Sendak books, and watc...
2 This is a fun way for children to learn their ...
                                         CleanedText
0 b'witti littl book make son laugh loud recit c...
1 b'grew read sendak book watch realli rosi movi...
2 b'fun way children learn month year learn poem...
```

#### 1.3 TF-IDF & Truncated SVD

TF-IDF will produce a high dimensional vector. If the number of features is very high, it is highly recommended to use another dimensionality reduction method.

For instance, **PCA for dense data or TruncatedSVD for sparse data.** Here, as the TF-IDF data is sparse, TruncatedSVD is used.

```
from sklearn.manifold import TSNE
import pdb

num_points = 10000
# Picking the top 1000 points as TSNE takes a lot of time for 15K points
data_1000 = final_tf_idf[0:num_points,:]
print(type(final_tf_idf))
labels_1000 = final['Score'].head(num_points)

from sklearn.decomposition import TruncatedSVD
from sklearn.random_projection import sparse_random_matrix
svd = TruncatedSVD(n_components=50, n_iter=10, random_state=42)
data_1000 = svd.fit_transform(data_1000)
print(data_1000.shape)

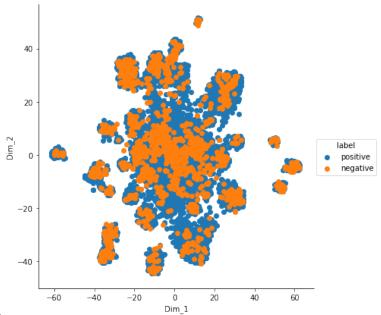
<class 'scipy.sparse.csr.csr_matrix'>
(10000, 50)
```

#### 2 t-SNE Visualization

The output of Truncated SVD is fed into t-SNE for visualization.

```
In [4]: # TSNE Plot after dimensionality reduction
        #from MulticoreTSNE import MulticoreTSNE as TSNE
        \#tsne = TSNE(n_jobs=1)\#, n_components=2,
                     random_state=0, perplexity = 100, n_iter = 1000)
        \#tsne\_data = tsne.fit\_transform(data\_1000)
        model = TSNE(n_components=2, random_state=0,
                     perplexity = 100, n_iter = 1000)
        # configuring the parameteres
        # the number of components = 2
        # default perplexity = 30
        # default learning rate = 200
        # default Maximum number of iterations for the optimization = 1000
        tsne_data = model.fit_transform(data_1000)
        # creating a new data frame which help us in ploting the result data
        tsne_data = np.vstack((tsne_data.T, labels_1000)).T
        tsne_df = pd.DataFrame(data=tsne_data, columns=("Dim_1", "Dim_2", "label"))
        # Ploting the result of tsne
```

```
sns.FacetGrid(tsne_df, hue="label", size=6).map(
    plt.scatter, 'Dim_1', 'Dim_2').add_legend()
plt.show()
```



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## 3 Observations

- 1. The **overlap is still significant using TF-IDF method. Not much improvement** is seen, in terms of separation, compared to BoW-tSNE visualization.
- 2. Hence, **we will use Word2Vec dense vector representation** for reviews and bypass Truncated SVD before feeding in to t-SNE.