7Exercise_T_SNE_Viz

February 3, 2019

```
In [1]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        %matplotlib inline
        import warnings
        warnings.filterwarnings("ignore")
        import sqlite3
        import nltk
        import string
        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
        import re
        import string
        from nltk.corpus import stopwords
        from nltk.stem import PorterStemmer
        from nltk.stem.wordnet import WordNetLemmatizer
        from gensim.models import Word2Vec
        from gensim.models import KeyedVectors
        import pickle
        from tqdm import tqdm
        import os
        from sklearn.utils import shuffle
        from sklearn.manifold import TSNE
```

```
In [2]: #defining all the required functions here
        def partition(x):
            if x < 3:
                return 'negative'
            return 'positive'
        def cleanHTML(sentence):
            cleanr = re.compile('<.*?>')
            cleantext = re.sub(cleanr, ' ', sentence)
            return cleantext
        def cleanPunctuation(sentence):
            cleaned = re.sub(r'[?|!|\'|"|#]', r'', sentence)
            cleaned = re.sub(r'[.|,|)|(||/|)', r'', sentence)
            return cleaned
        #removing stop words
        stop = set(stopwords.words('english')) #set of stopwords
        sno = nltk.stem.SnowballStemmer('english') #initializing the Snowball Stemmer
        final_dataset_path = '/home/monodeepdas112/Datasets/amazon-fine-food-reviews/final.sql
        initial_dataset_path = '/home/monodeepdas112/Datasets/amazon-fine-food-reviews/databaset
In [3]: if os.path.isfile(final_dataset_path):
            #loading the cleaned dataset if present
            con = sqlite3.connect(final_dataset_path)
            data = pd.read_sql_query('select * from Reviews', con)
            with open('positive_words.pkl', 'rb') as f:
                all_positive_words = pickle.load(f)
            with open('negitive_words.pkl', 'rb') as f:
                all_negative_words = pickle.load(f)
            #cleaning the dataset and making the finally cleaned dataset if the cleaned datase
            con = sqlite3.connect(initial_dataset_path)
            filtered_data = pd.read_sql_query('select * from Reviews where score != 3', con)
            #replacing the score column to contain only positive or negative rather than conti
            actualScore = filtered_data['Score']
            positiveNegativeScore = actualScore.map(partition)
            filtered_data['Score'] = positiveNegativeScore
            #data deduplication
            sorted_data = filtered_data.sort_values('ProductId', axis=0, ascending=True, inpla
            data = sorted_data.drop_duplicates(subset={"UserId", "ProfileName", "Time", "Text"}, ]
            #removing some wrong data points
```

```
#text pre-processing
            final_string=[]
            all_positive_words=[] # store words from +ve reviews here
            all_negative_words=[] # store words from -ve reviews here.
            for i, sent in enumerate(tqdm(data['Text'].values)):
                filtered_sentence=[]
                sent = cleanHTML(sent)#removing HTML tags
                for w in sent.split():
                    for cleaned_words in cleanPunctuation(w).split():
                        if((cleaned_words.isalpha()) & (len(cleaned_words)>2)):
                            if(cleaned_words.lower() not in stop):
                                s=(sno.stem(cleaned_words.lower()))
                                filtered_sentence.append(s)
                                if (final['Score'].values)[i] == 'positive':
                                    all_positive_words.append(s) #list of all words used to de
                                if(final['Score'].values)[i] == 'negative':
                                    all_negative_words.append(s) #list of all words used to de
                str1 = ' '.join(filtered_sentence) #final string of cleaned words
                final_string.append(str1)
            ###########---- storing the data into final.sqlite file -----###################
            data['CleanedText']=final_string #adding a column of CleanedText which displays th
            data['CleanedText'] = final['CleanedText']
            conn = sqlite3.connect(final_dataset_path)
            c=conn.cursor()
            conn.text_factory = str
            data.to_sql('Reviews', conn, schema=None, if_exists='replace', \
                         index=True, index_label=None, chunksize=None, dtype=None)
            conn.close()
            #saving the dictionaries
            with open('positive_words.pkl', 'wb') as f:
                pickle.dump(all_positive_words, f)
            with open('negitive_words.pkl', 'wb') as f:
                pickle.dump(all_negative_words, f)
In [4]: data.head()
Out[4]:
           index
                      Id ProductId
                                                                       ProfileName \
                                               UserId
        0 138706 150524 0006641040
                                      ACITT7DI6IDDL
                                                                   shari zychinski
        1 138688 150506 0006641040 A2IW4PEEKO2ROU
                                                                             Tracy
        2 138689 150507 0006641040 A1S4A3IQ2MU7V4
                                                             sally sue "sally sue"
        3 138690 150508 0006641040
                                          AZGXZ2UUK6X Catherine Hallberg "(Kate)"
        4 138691 150509 0006641040 A3CMRKGE0P909G
                                                                            Teresa
```

data = data[data.HelpfulnessNumerator<=data.HelpfulnessDenominator]

```
0
                              0
                                                         positive
                                                                    939340800
        1
                              1
                                                                   1194739200
                                                      1 positive
        2
                              1
                                                      1 positive
                                                                   1191456000
        3
                              1
                                                      1 positive
                                                                   1076025600
                              3
        4
                                                        positive
                                                                   1018396800
                                              Summary \
                            EVERY book is educational
        0
          Love the book, miss the hard cover version
        1
        2
                        chicken soup with rice months
        3
               a good swingy rhythm for reading aloud
        4
                      A great way to learn the months
                                                        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 ...
        4 This is a book of poetry about the months of t...
                                                 CleanedText
        0 witti littl book make son laugh loud recit car...
        1 grew read sendak book watch realli rosi movi i...
        2 fun way children learn month learn poem throug...
        3 great littl book read nice rhythm well good re...
        4 book poetri month year goe month cute littl po...
In [5]: min_data_points = 1000
In [6]: pos_data = data[data.Score=='positive'].loc[:, ['CleanedText', 'Score']]
       pos_data = pos_data.sample(min_data_points)
       neg_data = data[data.Score=='negative'].loc[:, ['CleanedText', 'Score']]
       neg_data = neg_data.sample(min_data_points)
       part_data = pos_data.append(neg_data)
        part_data = shuffle(part_data)
        cleaned_text = part_data['CleanedText'].tolist()
In [7]: #Function to plot t-SNE
        def plot t SNE(features, labels):
           plt.figure(figsize=(25, 25))
            ind=1
            for p in range(1, 61, 20):
                for i in range(1000, 5001, 1000):
                    #defining the t-SNE model
                    model = TSNE(n_components=2, random_state=0, perplexity=p, n_iter=i)
                    tsne_data = model.fit_transform(features)
```

HelpfulnessDenominator

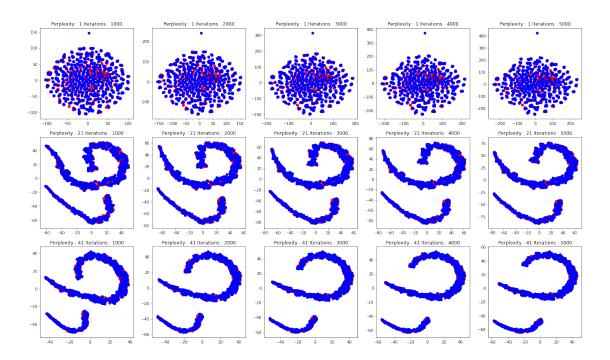
HelpfulnessNumerator

Score

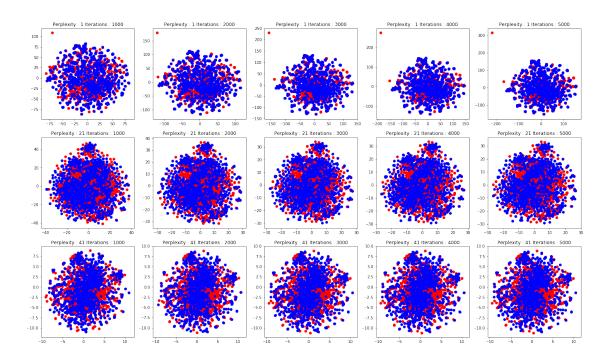
Time

```
#creating a new dataframe which help in plotting the result data
                    tsne_data = np.vstack((tsne_data.T, labels)).T
                    tsne_df = pd.DataFrame(data=tsne_data, columns=('Dim_1', 'Dim_2', 'label')
                    pos = tsne_df[tsne_df.label=='positive']
                    neg = tsne_df[tsne_df.label=='negative']
                    #Plotting the t-sne plot
                    plt.subplot(5, 5, ind)
                    plt.scatter(pos.iloc[:,0], pos.iloc[:,1], c='r')
                    plt.scatter(neg.iloc[:,0], neg.iloc[:,1], c='b')
                    plt.title('Perplexity : {0} Iterations : {1}'.format(p, i))
                    ind=ind+1
            plt.show()
        list_of_sentences=[]
        for sentence in cleaned_text:
            list_of_sentences.append(sentence.split())
        #Training my own Word2Vec model
        w2v_model=Word2Vec(list_of_sentences, size=50, workers=4)
0.1 t-SNE with Word2Vec
In [8]: #Making the sentence vectors
        sent_vectors = []
```

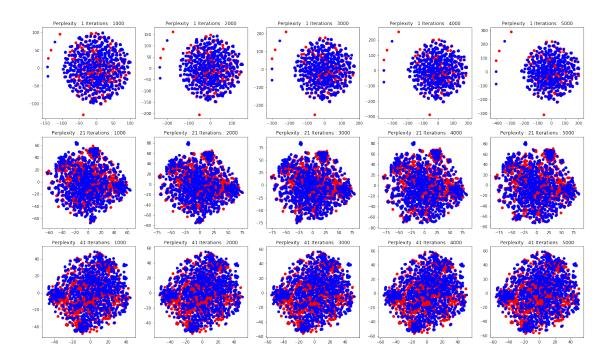
```
for sent in tqdm(list_of_sentences):
            sent_vec = np.zeros(50)
            cnt_words = 0
            for word in sent:
                if word in w2v_model:
                    word_vec = w2v_model.wv[word]
                    sent_vec += word_vec
                    cnt_words += 1
            if cnt_words != 0:
                sent_vec /= cnt_words
            sent_vectors.append(sent_vec)
        print(len(sent_vectors))
        print(len(sent_vectors[0]))
        #setting up t-SNE using Average Word2Vec
        features = sent_vectors
        labels = part_data['Score'].values
        #plotting t-SNE
        plot_t_SNE(features, labels)
100%|| 2000/2000 [00:01<00:00, 1962.76it/s]
```



0.2 t-SNE with Bag of Words



0.3 t-SNE with TF-IDF



0.4 t-SNE with TF-IDF weighted Word2Vec

```
In [11]: #initializing tf-idf model with ngram_range = (1,3)
         tfidf_vec = TfidfVectorizer(ngram_range=(1,3), max_features=1000)
         tfidf_vec.fit_transform(cleaned_text)
         #creating a dictionary of idf values
         idf_dictionary = dict(zip(tfidf_vec.get_feature_names(), list(tfidf_vec.idf_)))
         #getting the names of the features
         tfidf_features = tfidf_vec.get_feature_names()
         #creating the features
         tfidf_sent_vectors = [];
         row=0;
         for sent in tqdm(list_of_sentences):
             sent_vec = np.zeros(50)
             weight_sum =0;
             for word in sent:
                 if word in w2v_model and word in tfidf_features:
                     vec = w2v_model.wv[word]
                     tf_idf = idf_dictionary[word]*(sent.count(word)/len(sent))
                     sent_vec += (vec * tf_idf)
                     weight_sum += tf_idf
             if weight_sum != 0:
                 sent_vec /= weight_sum
             tfidf_sent_vectors.append(sent_vec)
```

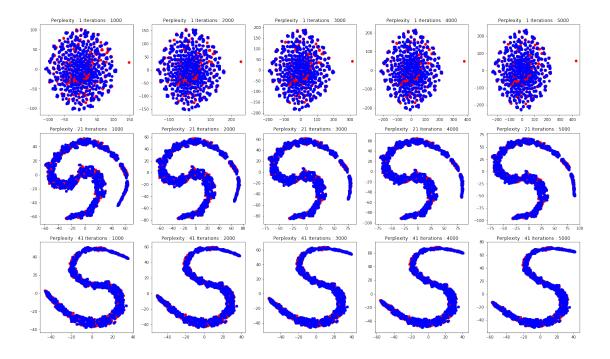
```
row += 1
print(len(tfidf_sent_vectors))
print(len(tfidf_sent_vectors[0]))

features = tfidf_sent_vectors
labels = part_data['Score'].values

plot_t_SNE(features, labels)

100%|| 2000/2000 [00:02<00:00, 962.75it/s]

2000
50</pre>
```



0.5 Conclusion

I had to limit the max_features to 1000 and use 1000 data pts and had to refer from assignment 3 to optimize the performance of Tf-Idf vectorizer to use the dictionary approach for idf features. But when I was trying to multiply TF-IDF Vectors with W2V then I was getting the error as Dimentionality Mismatch which is logical though as both are having different lengths. So Please tell me what should be done about it ??