# Sentimental\_Analysis

## February 4, 2019

```
In [10]: import pandas as pd
         import numpy as np
         import os
         import time
         from nltk.stem.wordnet import WordNetLemmatizer
         from sklearn.feature_extraction.text import CountVectorizer
         import itertools
         from nltk.corpus import stopwords
         from nltk.tokenize import word_tokenize
         import string
         import matplotlib.pyplot as plt
         from sklearn.model_selection import train_test_split
         from sklearn import svm
         import pickle
         from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.naive_bayes import MultinomialNB
         from sklearn.decomposition import PCA
         from sklearn.decomposition import TruncatedSVD
In [2]: import nltk
        nltk.download('stopwords')
        nltk.download('wordnet')
[nltk_data] Downloading package stopwords to /root/nltk_data...
              Unzipping corpora/stopwords.zip.
[nltk_data]
[nltk_data] Downloading package wordnet to /root/nltk_data...
[nltk_data]
              Unzipping corpora/wordnet.zip.
Out[2]: True
In [3]: from google.colab import drive
        drive.mount('/content/drive')
Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client_id=947318989803-
Enter your authorization code:
ຳຳຳຳຳຳຳຳຳຳຳຳຳ
```

```
In []: data=pd.read_csv("drive/My Drive/Machine Learning-Colab/Sentimental Analysis/movie_data
  Information About Data
In [5]: print ("Number of Data Points :", data.shape[0])
        print ("Number of Features /Variables:" ,data.shape[1])
Number of Data Points: 50000
Number of Features /Variables: 2
  Data Preprocessing
In [ ]: lmtzr = WordNetLemmatizer()
        stop_words = set(stopwords.words('english'))
        def remove_punctuation(s):
            s = ''.join([i for i in s if i not in frozenset(string.punctuation)])
            return s
        def nlp_preprocessing(total_text, index, column):
            if type(total_text) is not int:
                string = ""
                for words in total_text.split():
                    # remove the special chars in review like '"#$@!%~&*()_+-~?>< etc.
                    word = ("".join(e for e in words if e.isalpha()))
                    # Conver all letters to lower-case
                    word = word.lower()
                    # stop-word removal
                    if not word in stop_words:
                        string += lmtzr.lemmatize(word) + " "
                data[column][index] = string
In [8]: start_time = time.clock()
        data['review'] = data['review'].apply(remove_punctuation)
        print("Time for Punctuation removal",(time.clock() - start_time)/60, "minutes")
        start_time = time.clock()
        for index, row in data.iterrows():
            nlp_preprocessing(row['review'], index, 'review')
        print("Time for Preprocesing",(time.clock() - start_time)/60, "minutes")
Time for Punctuation removal 0.84782265 minutes
/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:18: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html
```

```
In [ ]: pd.to_hdf('movie_data.h5','data')
  New Starting
In [2]: data=pd.read_hdf('movie_data.h5','data')
In [3]: data.shape
Out[3]: (50000, 2)
In [4]: data.columns
Out[4]: Index(['review', 'sentiment'], dtype='object')
In [5]: Y=data['sentiment']
  BOW Model
In [6]: vectorizer=CountVectorizer()
        X=vectorizer.fit_transform(data['review'])
In [7]: X.shape
Out[7]: (50000, 164728)
In [9]: X_train, X_test, y_train, y_test = train_test_split(X, Y,test_size=.2,random_state=42)
In [10]: print (X_train.shape, y_train.shape)
        print (X_test.shape, y_test.shape)
(40000, 164728) (40000,)
(10000, 164728) (10000,)
  RBF Kernel
In [11]: start_time = time.clock()
        model_svm = svm.SVC(C=1,gamma=1)
        model_svm.fit(X_train,y_train)
         y_pred= model_svm.predict(X_test)
         print((time.clock() - start_time)/60, "minutes")
102.20228001622887 minutes
In [12]: print("Accuracy", model_svm.score(X_test,y_test))
Accuracy 0.5065
```

Time for Preprocesing 120.33746326666666 minutes

```
In [20]: svm_pkl_model=open("svm_model_rbf_bow.pkl","wb")
In [21]: pickle.dump(model_svm,svm_pkl_model)
In [22]: svm_pkl_model.close()
       Linear Kernel
In [23]: start_time = time.clock()
                       model_svm = svm.SVC(kernel='linear',C=1,gamma=1)
                       model_svm.fit(X_train,y_train)
                       print((time.clock() - start_time)/60, "minutes")
144.38460584875347 minutes
In [24]: print("Accuracy", model_svm.score(X_test, y_test))
Accuracy 0.8625
In [25]: svm_pkl_model=open("svm_model_linear_bow.pkl","wb")
                       pickle.dump(model_svm,svm_pkl_model)
                       svm_pkl_model.close()
       Tf-Idf Model
In [8]: tfidf_review_vectorizer_train1 = TfidfVectorizer()
                    tfidf_review_features_train1= tfidf_review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_transform(data['review_vectorizer_train1.fit_train1.fit_train1.fit_train1.fit_train1.fit_tr
In [ ]: X_train1, X_test1, y_train1, y_test1 = train_test_split(tfidf_review_features_train1, '
                    print (X_train1.shape, y_train1.shape)
                    print (X_test1.shape, y_test1.shape)
       Linear kernel
In [27]: start_time = time.clock()
                       model_svm = svm.SVC(kernel='linear',C=1,gamma=1)
                       model_svm.fit(X_train1,y_train1)
                       print((time.clock() - start_time)/60, "minutes")
                       print("Accuracy:",model_svm.score(X_test1,y_test1))
26.57927562000556 minutes
Accuracy: 0.896
In [28]: svm_pkl_model=open("svm_model_linear_TfIDF.pkl","wb")
                       pickle.dump(model_svm,svm_pkl_model)
                       svm_pkl_model.close()
       Rbf Kernel
```

```
In [29]: start_time = time.clock()
         model_svm = svm.SVC(C=1,gamma=1)
         model_svm.fit(X_train1,y_train1)
         print((time.clock() - start_time)/60, "minutes")
         print("Accuracy:",model_svm.score(X_test1,y_test1))
52.326281867188605 minutes
Accuracy: 0.89736
In [30]: svm_pkl_model=open("svm_model_rbf_TfIDF.pkl","wb")
         pickle.dump(model_svm,svm_pkl_model)
         svm_pkl_model.close()
1 Naive bayes Model
Tf-Idf Model
In [35]: start_time = time.clock()
         clf = MultinomialNB()
         clf.fit(X_train1, y_train1)
         print("Accuracy:",clf.score(X_test1,y_test1))
         print((time.clock() - start_time)/60, "minutes")
Accuracy: 0.86528
0.0027276288708283874 minutes
In [36]: nb_pkl_model=open("nb_model_multinb.pkl","wb")
         pickle.dump(clf,nb_pkl_model)
         nb_pkl_model.close()
   Bow Model
In [37]: start_time = time.clock()
         clf = MultinomialNB()
         clf.fit(X_train, y_train)
         print("Accuracy:",clf.score(X_test,y_test))
         print((time.clock() - start_time)/60, "minutes")
Accuracy: 0.8598
0.01507356905543323 minutes
In [38]: nb_pkl_model=open("nb_model_multinb_Bow.pkl","wb")
         pickle.dump(clf,nb_pkl_model)
         nb_pkl_model.close()
```

## 2 Dimensionality Reduction

Latent Semantic analysis on BOW Model

```
In [11]: start_time = time.clock()
         svd = TruncatedSVD(n_components=1000, random_state=42)
         bow_X_svd=svd.fit_transform(X)
         print((time.clock() - start_time)/60, "minutes")
4.982865505514079 minutes
In [12]: svd.singular_values_.size
Out[12]: 1000
In [13]: bow_X_svd.size
Out[13]: 50000000
In [14]: arr=svd.explained_variance_ratio_
In [15]: plt.plot(np.cumsum(arr))
         plt.xlabel('number of components')
         plt.ylabel('cumulative explained variance');
         plt.show()
           0.7
           0.6
        cumulative explained variance
           0.5
           0.4
           0.3
           0.2
           0.1
                            200
                  0
                                        400
                                                    600
                                                               800
                                                                          1000
                                    number of components
```

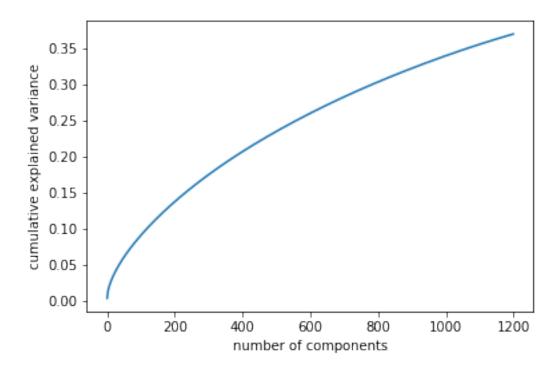
```
In [16]: X_train, X_test, y_train, y_test = train_test_split(bow_X_svd, Y,test_size=.2,random_s
         print (X_train.shape, y_train.shape)
        print (X_test.shape, y_test.shape)
(40000, 1000) (40000,)
(10000, 1000) (10000,)
  SVM-Linear model
In [17]: start_time = time.clock()
         model_svm = svm.SVC(kernel='linear',C=1,gamma=1)
         model_svm.fit(X_train,y_train)
         print((time.clock() - start_time)/60, "minutes")
         print("Accuracy:",model_svm.score(X_test,y_test))
339.4825693216163 minutes
Accuracy: 0.8677
In [19]: svm_pkl_model=open("SVM_Bow_svd_linear.pkl","wb")
         pickle.dump(model_svm,svm_pkl_model)
         svm_pkl_model.close()
  SVM-RBf model
In [20]: start_time = time.clock()
         model_svm = svm.SVC(C=1,gamma=1)
         model_svm.fit(X_train,y_train)
         print((time.clock() - start_time)/60, "minutes")
         print("Accuracy:",model_svm.score(X_test,y_test))
102.48039741238603 minutes
Accuracy: 0.507
In [22]: svm_pkl_model=open("SVM_Bow_svd_rbf.pkl","wb")
         pickle.dump(model_svm,svm_pkl_model)
         svm_pkl_model.close()
  Naive Bayes-Multinomial
In [23]: start_time = time.clock()
         clf = MultinomialNB()
         clf.fit(X_train, y_train)
         print("Accuracy:",clf.score(X_test,y_test))
         print((time.clock() - start_time)/60, "minutes")
```

```
<ipython-input-23-e3872ec50907> in <module>()
          1 start time = time.clock()
          2 clf = MultinomialNB()
    ----> 3 clf.fit(X_train, y_train)
          4 print("Accuracy:",clf.score(X_test,y_test))
          5 print((time.clock() - start_time)/60, "minutes")
        C:\ProgramData\Anaconda3\lib\site-packages\sklearn\naive_bayes.py in fit(self, X, y, s
                    self.feature_count_ = np.zeros((n_effective_classes, n_features),
        602
        603
                                                    dtype=np.float64)
    --> 604
                    self._count(X, Y)
        605
                    alpha = self._check_alpha()
        606
                    self._update_feature_log_prob(alpha)
        C:\ProgramData\Anaconda3\lib\site-packages\sklearn\naive_bayes.py in _count(self, X, Y
                    """Count and smooth feature occurrences."""
        706
                    if np.any((X.data if issparse(X) else X) < 0):</pre>
        707
    --> 708
                        raise ValueError("Input X must be non-negative")
        709
                    self.feature_count_ += safe_sparse_dot(Y.T, X)
                    self.class_count_ += Y.sum(axis=0)
        710
        ValueError: Input X must be non-negative
In [ ]: nb_pkl_model=open("nb_multinb_Bow_svd.pkl","wb")
        pickle.dump(clf,nb_pkl_model)
        nb_pkl_model.close()
  LSA on TF-Idf Model
In [22]: start_time = time.clock()
         svd = TruncatedSVD(n_components=1200, random_state=42)
         tfidf_X_svd=svd.fit_transform(tfidf_review_features_train1)
         print((time.clock() - start_time)/60, "minutes")
5.798072893612114 minutes
In [23]: arr=svd.explained_variance_ratio_
In [24]: plt.plot(np.cumsum(arr))
         plt.xlabel('number of components')
```

Traceback (most recent call last)

ValueError

```
plt.ylabel('cumulative explained variance');
plt.show()
```



## 3 PCA

#### PCA on Bow

```
pca_X_bow=pca.fit_transform(tfidf_review_features_train1)
         print((time.clock() - start_time)/60, "minutes")
        TypeError
                                                   Traceback (most recent call last)
        <ipython-input-26-adecf8d266b1> in <module>()
          1 start_time = time.clock()
          2 pca = PCA(n_components=150)
    ----> 3 pca_X_bow=pca.fit_transform(tfidf_review_features_train1)
          4 print((time.clock() - start_time)/60, "minutes")
        C:\ProgramData\Anaconda3\lib\site-packages\sklearn\decomposition\pca.py in fit_transfo
        346
        347
                    11 11 11
                    U, S, V = self._fit(X)
    --> 348
                    U = U[:, :self.n_components_]
        349
        350
        C:\ProgramData\Anaconda3\lib\site-packages\sklearn\decomposition\pca.py in _fit(self, it)
                    # This is more informative than the generic one raised by check_array.
        364
                    if issparse(X):
        365
                        raise TypeError('PCA does not support sparse input. See '
    --> 366
        367
                                         'TruncatedSVD for a possible alternative.')
        368
        TypeError: PCA does not support sparse input. See TruncatedSVD for a possible alternat
In [ ]: arr=pca.explained_variance_ratio_
        plt.plot(np.cumsum(arr))
        plt.xlabel('number of components')
        plt.ylabel('cumulative explained variance');
        plt.show()
```