Project Report 24

Text Classification

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Question:

Using vector semantics, we can easily convert a given text into its corresponding vector form. Given any text, first pre process the text and convert it into a vector using BoW methods. Given this vector, implement your own classifier to classify the vector is pre-defined categories. You may use of these datasets for training and for defining the categories: 14 Best Text classification Datasets for Machine Learning

Prerequisites

What things you need to install the software and how to install them:

Python 3.6 This setup requires that your machine has latest version of python. The following url https://www.python.org/downloads/ can be referred to download python. Once you have python downloaded and installed, you will need to setup PATH variables (if you want to run python program directly, detail instructions are below in how to run software section). To do that check this: https://www.pythoncentral.io/add-python-to-path-python-is-not-recognized-asan-internal-or-externalcommand/. Setting up PATH variable is optional as you can also run program without it and more instruction are given below on this topic.

Second and easier option is to download anaconda and use its anaconda prompt to run the commands. To install anaconda check this url https://www.anaconda.com/download/ You will also need to download and install below 3 packages after you install either python or anaconda from the steps above Sklearn (scikit-learn) numpy scipy if you have chosen to install python 3.6

Implementation

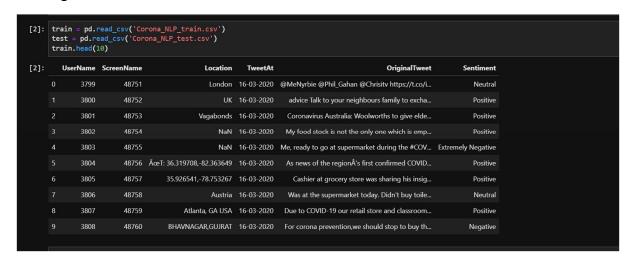
Importing the libraries and dataset

```
Launcher

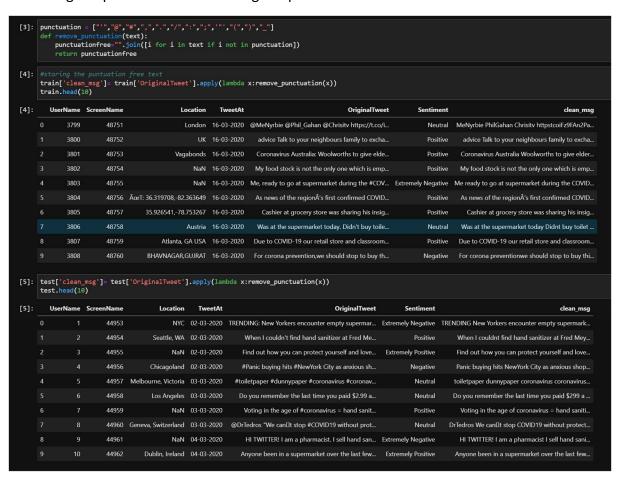
Last Modified
a month ago
8 days ago

Reference from nitk.acom import PorterStemmer
from nitk.acom import traingles, extraction import traingles, extraction import traingles, extraction, text import TfidfVectorizer
from sklearn.model_selection import traingles plact
from sklearn.model_selection import traingles plact
from sklearn.model_selection import traingles
from sklearn.maker_selection.text import acomp
from sklearn.maker_selection.text import acomp
from sklearn.maker_selection.text import traingles
from sklearn.maker_selection.text import traingles
from sklearn.maker_selection.text import traingles
from tensorflow.keras.utils import to_categorical
import tensorflow.keras.utils import to_categorical
```

Loading the dataset:



Removing the punctuation and storing the punctuation free text



Removing the unwanted Columns

```
[7]: unwanted_cols = ['UserName', 'ScreenName', 'Location', 'TweetAt', 'OriginalTweet', 'clean_msg', 'Sentiment']
[8]: train_X = train.drop(unwanted_cols, axis=1)
     train_X = train['Sentiment']
test_X = test.drop(unwanted_cols, axis=1)
test_Y = test['Sentiment']
stop_words = nltk.corpus.stopwords.words('english')
     stop_words[:5]
[8]: ['i', 'me', 'my', 'myself', 'we']
[9]: train_X['length'] = train_X['msg_lower'].apply(len)
     train_X.head(10)
[9]:
                                     msg_lower length
     0 menyrbie philgahan chrisitv httpstcoifz9fan2pa...
     1 advice talk to your neighbours family to excha...
                                                 237
     2 coronavirus australia woolworths to give elder...
                                                 124
     3 my food stock is not the only one which is emp...
                                                 284
     4 me ready to go at supermarket during the covid...
                                                 287
     5 as news of the regionâ's first confirmed covid...
                                                 238
     6 cashier at grocery store was sharing his insig...
                                                 168
     7 was at the supermarket today didnt buy toilet ...
                                                 107
     8 due to covid-19 our retail store and classroom...
     9 for corona preventionwe should stop to buy thi...
                                                 260
[12]: def pre_process(text):
              text = [word for word in text.split() if word.lower() not in stop_words]
              words =
              for i in text:
                        words += (ps.stem(i))+" "
              return words
[13]: textFeatures_train = train_X['msg_lower'].copy()
         textFeatures_train = textFeatures_train.apply(pre_process)
         textFeatures_train[:5]
               menyrbi philgahan chrisitv httpstcoifz9fan2pa ...
[13]: 0
               advic talk neighbour famili exchang phone numb...
               coronaviru australia woolworth give elderli di...
        2
               food stock one empti pleas dont panic enough f...
        3
               readi go supermarket covid19 outbreak im paran...
        Name: msg_lower, dtype: object
```

```
[13]: textFeatures_train = train_X['msg_lower'].copy()
       textFeatures_train = textFeatures_train.apply(pre_process)
       textFeatures_train[:5]
 [13]: 0 menyrbi philgahan chrisitv httpstcoifz9fan2pa ...
           advic talk neighbour famili exchang phone numb...
coronaviru australia woolworth give elderli di...
           food stock one empti pleas dont panic enough f\dots
            readi go supermarket covid19 outbreak im paran...
       Name: msg_lower, dtype: object
 [14]: textFeatures_test = test_X['msg_lower'].copy()
textFeatures_test = textFeatures_test.apply(pre_process)
       textFeatures_test[:5]
 [14]: 0
           trend new yorker encount empti supermarket she...
            couldnt find hand sanit fred meyer turn amazon...
                         find protect love one coronaviru ?
            panic buy hit newyork citi anxiou shopper stoc...
           toiletpap dunnypap coronaviru coronavirusaustr...
       Name: msg_lower, dtype: object
 [15]: textFeatures_combined = pd.concat([textFeatures_train, textFeatures_test], axis = 0)
       len(textFeatures_combined)
 [15]: 44955
[16]: (44955, 546515)
[17]: train_X = features_combined[:len(textFeatures_train)]
        test_X = features_combined[len(textFeatures_train):]
        train_X.shape, test_X.shape
[17]: ((41157, 546515), (3798, 546515))
[18]: len(train_Y), len(test_Y)
[18]: (41157, 3798)
[19]: target_names = train_Y.unique()
        target_names
[19]: array(['Neutral', 'Positive', 'Extremely Negative', 'Negative',
                'Extremely Positive'], dtype=object)
[20]:
         # Prediction using Support Vector Machine
         svc = SVC(kernel='sigmoid', gamma=1.0)
[21]: svc.fit(train_X, train_Y)
[21]: SVC(gamma=1.0, kernel='sigmoid')
```

```
[23]: # accuracy_score(labels_test, prediction)
      print(classification_report(test_Y, prediction, target_names = target_names))
                          precision
                                      recall f1-score support
                                        0.39
                Neutral
                              0.72
                                                  0.50
                                                             592
                Positive
                                                  0.58
                              0.77
                                        0.47
                                                             599
      Extremely Negative
                              0.50
                                        0.56
                                                  0.53
                                                            1041
               Negative
                              0.61
                                        0.63
                                                  0.62
                                                             619
      Extremely Positive
                              0.48
                                        0.67
                                                  0.56
                                                             947
                                                  0.56
                                                            3798
                accuracy
                              0.62
                                        0.54
                                                  0.56
              macro avg
                                                            3798
            weighted avg
                              0.59
                                        0.56
                                                  0.56
                                                            3798
[24]: # Prediction using Multinomial Naive Bayes Model
      mnb = MultinomialNB(alpha=0.2)
      mnb.fit(train_X, train_Y)
      prediction = mnb.predict(test_X)
      print(classification_report(test_Y, prediction, target_names = target_names))
                                      recall f1-score support
                          precision
                Neutral
                              0.81
                                        0.07
                                                  0.13
                                                             592
                Positive
                              0.77
                                        0.14
                                                  0.24
                                                             599
      Extremely Negative
                              0.39
                                        0.50
                                                  0.43
                                                            1041
               Negative
                              0.64
                                                             619
                                        0.18
                                                  0.28
      Extremely Positive
                              0.33
                                        0.75
                                                  0.46
                                                             947
                                                  0.38
                                                            3798
                accuracy
              macro avg
                              0.59
                                        0.33
                                                  0.31
                                                            3798
            weighted avg
                               0.54
                                         0.38
                                                  0.34
                                                            3798
```