Week 5: Brandon Mather DSC550-T301

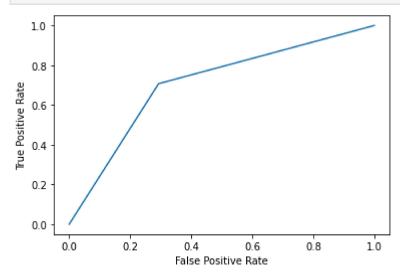
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In [2]:
         import pandas as pd
          import numpy as np
          import unicodedata
          import sys
          from nltk.corpus import stopwords
          from nltk.stem.porter import PorterStemmer
          from sklearn.feature_extraction.text import CountVectorizer
          from sklearn.feature extraction import DictVectorizer
          from sklearn.feature extraction.text import TfidfVectorizer
          from sklearn.model_selection import train_test_split
          from sklearn.linear model import LogisticRegression
          from sklearn.metrics import confusion_matrix
          from sklearn.metrics import accuracy score
          from sklearn.metrics import precision score
          from sklearn.metrics import recall_score
          from sklearn.metrics import f1 score
          from sklearn import metrics
          import matplotlib.pyplot as plt
          from sklearn import tree
         #1 Get the stemmed data using the same process you did in Week 3.
 In [5]:
          data = pd.read_csv("labeledTrainData.tsv", header=0, \
                              delimiter="\t", quoting=3)
         review = data["review"]
 In [6]:
         def decapitalizer(string: str) -> str:
 In [7]:
              return string.lower()
         noncapital = [decapitalizer(string) for string in review]
 In [8]:
         punctuation = dict.fromkeys(i for i in range(sys.maxunicode)
 In [9]:
              if unicodedata.category(chr(i)).startswith('P'))
In [10]:
         noperiods = [string.translate(punctuation) for string in noncapital]
In [11]:
         stop words = stopwords.words('english')
In [12]:
         nostopwords = [word for word in noperiods if word not in stop words]
In [13]:
         porter = PorterStemmer()
In [14]:
         appliedstemmer = [porter.stem(word) for word in nostopwords]
In [15]: #2 Split this into a training and test set.
         x = appliedstemmer
         y = data.sentiment
In [16]: x_train, x_test, y_train, y_test = train_test_split(x,y, test_size=0.2)
```

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#3 Fit and apply the tf-idf vectorization to the training set.
In [17]:
          tfidf vectorizer = TfidfVectorizer()
         tfidf train = tfidf vectorizer.fit transform(x train)
In [18]:
In [19]:
         #4 Apply but DO NOT FIT the tf-idf vectorization to the test set (Why?).
         tfidf_test = tfidf_vectorizer.transform(x_test)
         #5 Train a logistic regression using the training data.
In [17]:
         logit = LogisticRegression()
         model = logit.fit(tfidf_train, y_train)
In [18]:
In [21]: #6 Find the model accuracy on test set.
          score = accuracy score(y test,x test)
          score
         C:\Users\brand\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:217: Fu
         tureWarning: elementwise comparison failed; returning scalar instead, but in the futu
         re will perform elementwise comparison
           score = y_true == y_pred
         0.0
Out[21]:
         #7 Create a confusion matrix for the test set predictions.
In [26]:
          pred = model.predict(tfidf_test)
In [27]:
         confusion_matrix(y_test, pred)
         array([[2178, 290],
Out[27]:
                [ 267, 2265]], dtype=int64)
         #8 Get the precision, recall, and F1-score for the test set predictions.
In [30]:
          precision_score(y_test, pred)
         0.8864970645792564
Out[30]:
In [31]:
         recall_score(y_test, pred)
         0.8945497630331753
Out[31]:
In [32]:
         f1_score(y_test, pred)
         0.890505209357185
Out[32]:
         #9 Create a ROC curve for the test set.
In [36]:
         y_pred_proba = logit.predict_proba(tfidf_test)[::,1]
In [37]:
         fpr, tpr, _ = metrics.roc_curve(y_test, y_pred_proba)
In [38]:
         plt.plot(fpr,tpr)
          plt.ylabel('True Positive Rate')
          plt.xlabel('False Positive Rate')
          plt.show()
```

```
1.0 - 0.8 - 0.6 0.6 0.4 - 0.2 0.4 0.6 0.8 1.0 False Positive Rate
```

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#10 Pick another classification model you learned about this week and repeat steps (5)
 In [ ]:
          #Decision Tree
 In [3]:
          clf = tree.DecisionTreeClassifier()
          clf = clf.fit(tfidf_train, y_train)
In [21]:
         Y_pred = clf.predict(tfidf_test)
In [24]:
          score_2 = accuracy_score(y_test, Y_pred)
In [25]:
          score 2
         0.707
Out[25]:
          confusion_matrix(y_test, Y_pred)
In [26]:
         array([[1809, 750],
Out[26]:
                 [ 715, 1726]], dtype=int64)
In [27]:
          precision_score(y_test, Y_pred)
         0.697092084006462
Out[27]:
In [28]:
          recall_score(y_test, Y_pred)
         0.7070872593199509
Out[28]:
          f1_score(y_test, Y_pred)
In [29]:
         0.7020540980272524
Out[29]:
         y_pred_proba_2 = clf.predict_proba(tfidf_test)[::,1]
In [30]:
In [31]:
         fpr_2, tpr_2, _ = metrics.roc_curve(y_test, y_pred_proba_2)
In [32]:
          plt.plot(fpr_2,tpr_2)
          plt.ylabel('True Positive Rate')
```

plt.xlabel('False Positive Rate')
plt.show()



In []: