PROJECT - 20 NEWSCROUP DATASET

Cleaning, Prediction and Performance analysis of text dataset.

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OBJECTIVES

- Introduction of Dataset
- Preparing Dataset
- Data Cleaning
- Vectorization: tf_idf
- Text Classifier
- Prediction on Random Test Data
- Algorithm Performance
- Model Evaluation



- The 20 newsgroups dataset comprises around 18000 newsgroups posts on 20 topics split in two subsets: one for training (or development) and the other one for testing (or for performance evaluation). The split between the train and test set is based upon a messages posted before and after a specific date.
- The 20 newsgroups collection has become a popular data set for experiments in text applications of machine learning techniques, such as text classification and text clustering.
- The data is organized into 20 different newsgroups, each corresponding to a different topic. Some of the newsgroups are very closely related to each other.



INTRODUCTION

Fetching and Displaying Dataset

The <u>sklearn.datasets.fetch_20newsgroups</u> function is a data fetching / caching functions that downloads the data archive from the original $\underline{20}$ <u>newsgroups website</u>.

- The whole dataset would be divided in 2 parts:

df_train.head()

	news	source	label
0	I was wondering if anyone out there could enli	7	rec.autos
1	A fair number of brave souls who upgraded thei	4	comp.sys.mac.hardware
2	well folks, my mac plus finally gave up the gh	4	comp.sys.mac.hardware
3	\nDo you have Weitek's address/phone number?	1	comp.graphics
4	From article <c5owcb.n3p@world.std.com>, by to</c5owcb.n3p@world.std.com>	14	sci.space

2. Test Dataset We can fetch test dataset by using command: test = fetch_20newsgroups(subset='test')

df test.head()

	news	source	label
0	I am a little confused on all of the models of	7	rec.autos
1	I'm not familiar at all with the format of the	5	comp.windows.x
2	\nln a word, yes.\n	0	alt.atheism
3	\nThey were attacking the Iraqis to drive them 17 talk		talk.politics.mideast
4	\nl've just spent two solid months arguing tha	19	talk.religion.misc

PREPARING DATASET

- In the test dataset, the data contains text with newlines, punctuation, misspellings, and other items common in text documents. To build a model, we will clean up the text by removing some of these issues.
- Data cleaning will include:
 - 1. Removing stopwords
 - 2. Filter out short words
 - 3. Lowercase and removing everything except words
 - 4. Removing of special characters
 - 5. Applying lemmatization to the text
- Using nltk library the clean data in dataframe will look like:

	news	source	label	clean_text
0	I was wondering if anyone out there could enli	7	rec.autos	wondering anyone could enlighten car saw day d
1	A fair number of brave souls who upgraded thei	4	comp.sys.mac.hardware	fair number brave soul upgraded clock oscillat
2	well folks, my mac plus finally gave up the gh	4	comp.sys.mac.hardware	well folk mac plus finally gave ghost weekend
3	\nDo you have Weitek's address/phone number?	1	comp.graphics	weitek address phone number like get informati
4	From article <c5owcb.n3p@world.std.com>, by to</c5owcb.n3p@world.std.com>	14	sci.space	article owcb world std com tombaker world std

DATA CLEANING

- In order to feed predictive or clustering models with the text data, one first need to turn the text into vectors of numerical values suitable for statistical analysis.
- This can be achieved with the utilities of the sklearn.feature_extraction.text that extract <u>TF-IDF</u> vectors of unigram tokens from a subset of 20news.
- We can import from sklearn by using command:

```
from sklearn.feature_extraction.text import
CountVectorizer, TfidfVectorizer
```

```
vectorizer = TfidfVectorizer(min_df=5,
strip_accents='ascii', analyzer='word', lowercase=True)
```

CONVERTING TEXT TO VECTORS

- The multinomial Naive Bayes classifier which is suitable for discrete classification.
- Scikit-learn has a great Class called Pipeline, which allows us to a create pipeline for a classifier, i.e. we can just add the functions that we want to use on our input data.
- Here, we are using a *TfidfVectorizer()* as vectorizer and *Multinomial* as classifier:
- It is easy for a classifier to overfit on particular things that appear in the 20 Newsgroups data, such as newsgroup headers.

```
#Initialize and fit
nb = MultinomialNB()
nb.fit(x_train, y_train)
nb.score(x_test, y_test)

# Apply to testing data
y_pred = nb.predict(x_test)
print("Accuracy is: %0 3f" % nb score(x_test, y_test))
```

```
print("Accuracy is: %0.3f" % nb.score(x_test, y_test))
print(metrics.classification_report(y_test, y_pred, target_names=test.target_names))
```

Accuracy is: 0.673

TEXT CLASSIFIER

- After training, testing and cleaning the data the model is now ready to predict the random values and in which category the data belong.
- We can provide a list of different data and test the model for its accurate prediction.
- This can be done by the following method in code:

Clean test data: [u'detroit red wing still rebuild mode find hidden gem fantasy hockey team team lot prove', 'hardwar e delivery system written', u'latest motorcycle india royal enfield including meteor himalayan classic bullet', u'pro duction image computer use medium image used graphic design printed material frequent']

```
t = vectorizer.transform(testing).toarray()
print (t.shape)
Predicted_values = nb.predict(t)
for i in Predicted_values:
    print (dataset.target_names[i])

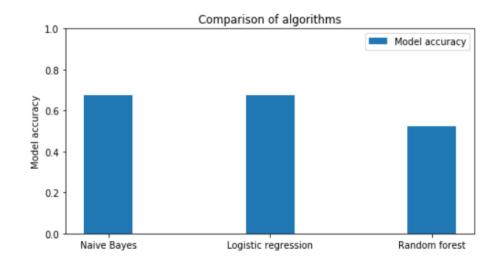
(4L, 13874L)
rec.sport.hockey
comp.sys.ibm.pc.hardware
```

rec.motorcycles comp.graphics

PREDICTION ON RANDOM TEST DATA

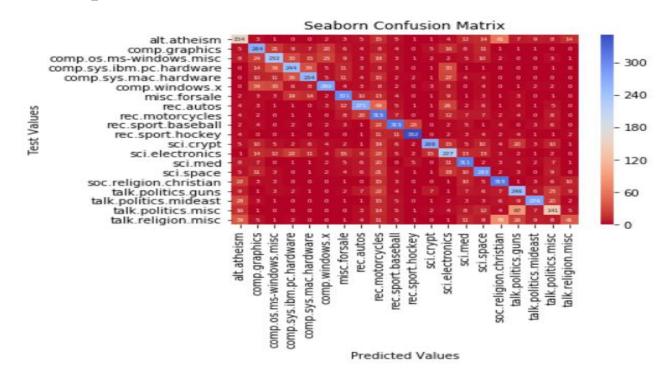


- In this analysis we can use different algorithms to get the better accuracy score.
- The algorithms used are:
 - 1. Naïve Bayes
 - 2. Logistics Regression
 - 3. Random Forest
- Each algorithm gives the different accuracy score which helps in the analysis of performance of each algorithm and the optimal performance can be found.
- Performance can be shown in the form of graph:



ALGORITHM PERFORMANCE

- We will use confusion_matrix() from sckit-learn to compare real and predicted categories.
- A confusion matrix is a technique for summarizing the performance of a classification algorithm.
- Confusion matrices are useful because they give direct comparisons of values like True Positives, False Positives, True Negatives and False Negatives.
- The output of Seaborn Confusion Matrix is:



MODEL EVALUATION



