## **COMP 543 Assignment #4**

Due March 30th at 11:55PM.

### 1 Description

In this assignment, you will be implementing a kNN classifier to classify text documents. "Classification" is the task of labeling documents based upon their contents. The implementation will be in Python, on top of Spark. You will be asked to perform three subtasks, covering data preparation and classification.

#### 2 Data

You will be dealing with the widely-used "20 newsgroups" data set. A newsgroup post is like an old-school blog post, and this data set has 19,997 such posts from 20 different categories, according to where the blog post was made. The 20 categories are listed in the file categories.txt. The category name can be extracted from the name of the document. For example, the document with identifier 20\_newsgroups/comp.graphics/37261 is from the comp.graphics category. The document with the identifier 20\_newsgroups/sci.med/59082 is from the sci.med category. The data file has one line per document "39 MB of text. It can be accessed at:

https://s3.amazonaws.com/risamyersbucket/comp543\_Lab5/20\_news\_same\_line.txt or as direct S3 address, so you can use it in a Spark job:

s3://risamyersbucket/comp543\_Lab5/20\_news\_same\_line.txt

#### 3 The Tasks

There are three separate tasks that you need to complete to finish the assignment.

#### 3.1 Task 1

First, you need to write Spark code that builds a dictionary that includes the 20,000 most frequent words in the training corpus - this was part of Lab 5. When you do this, please start with my code, so that we know that everyone has the same dictionary. Then, you need to use this dictionary to create an RDD where each document is represented as one entry in the RDD. Specifically, the key of the document is the document identifier (like 20\_newsgroups/comp.graphics/37261) and the value is a NumPy array with 20,000 entries, where the *i*th entry in the array is the number of times that the *i*th word in the dictionary appears in the document.

Once you do this, print out the arrays that you have created for documents

```
20_newsgroups/comp.graphics/37261,
20_newsgroups/talk.politics.mideast/75944, and
20_newsgroups/sci.med/58763
```

Since each array is going to be huge, with a lot of zeros, the thing that you want to print out is just the non-zero entries in the array (that is, for an array a, print out a [a.nonzero ()].

#### 3.2 Task 2

It is often difficult to classify documents accurately using raw count vectors. Thus, the next task is to write some more Spark code that converts each of those 19,997 count vetors to TF-IDF vectors "term frequency-inverse document frequency vectors"). The ith entry in a TF-IDF vector for document d is computed as:

$$TF(i,d) \times IDF(i)$$

Where TF(i, d) is:

#### Number of occurrences of word i in dTotal number of words in d

Note that the "Total number of words" is not the number of distinct words. The "total number of words" in "Today is a great day today" is six. And the IDF(i) is:

# $log \frac{Size \ of \ corpus \ (number \ of \ docs)}{Number \ of \ documents \ having \ word \ i}$

Again, once you do this, print out the arrays that you have created for documents:

```
20_newsgroups/comp.graphics/37261,
20_newsgroups/talk.politics.mideast/75944 and
20_newsgroups/sci.med/58763
```

Again, print out just the non-zero entries.

#### 3.3 Task 3

Next, your task is to build a kNN classifier, embodied by the Python function predictLabel. This function will take as input a text string and a number k, and then output the name of one of the 20 newsgroups. This name is the new group that the classifier thinks that the text string is "closest" to. It is computed using the classical kNN algorithm. This algorithm first converts the input string into a TF-IDF vector (using the dictionary and count information computed over the original corpus). It then finds the k documents in the corpus that are "closest" to the query vector (where distance is computed using the  $L_2$  norm), and returns the newsgroup label that is most frequent in those top k. Ties go to the label with the closest corpus document. Once you have written your function, run it on the following (each is an except from a Wikipedia article, chosen to match one of the 20 newsgroups).

#### 4 Turnin

Create a single document that has results for all three tasks. For the soft copy, turn in this document as well as all of your code. For your soft copy, please zip up all of your code and your document (use .gz or .zip only, please!), or else attach each piece of code as well as your document to your submission individually. No PDFs of code, please!

## 5 Grading

Each task is worth 33% of the overall grade.