Project in Spark 2017

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1 TASK

Firstly we uncompressed the data stored in ling-spam.zip folder with *Extract all* command. Secondly we open Virtual Box machine with Hortonworks, we signed in with maria_dev username and maria_dev password on Ambari available under 127.0.0.1:8080 ip address. We have selected *Files view*, than navigated to /tmp folder and created directories tmp/ling-spam/ham and ling-spam/spam. Following that we logged in with ssh credentials to Hortonworks machine

```
1 $ssh root@127.0.0.1 -p 2222
```

In the meantime upload to the virtual machine ling-spam.zip with:

```
1 $sudo scp -P 2222 ../ling-spam.zip root@127.0.0.1:/tmp/
```

We unzipped ling-spam.zip with:

```
1 $unzip ling-spam.zip -d /tmp/ling-spam
```

We putted files into /tmp/ling-spam/ folder in hdfs with:

```
1 $hdfs dfs -put ./ling-spam/ham /tmp/ling-spam/ham
```

^{2 \$}hdfs dfs -put ./ling-spam/spam /tmp/ling-spam/spam

2 TASK

Installation of sbt:

6 \$sbt package

```
$\text{$wget http://dl.bintray.com/sbt/rpm/sbt-0.13.12.rpm}$
Edit file /etc/yum.repos.d/sandbox.repo:

[sandbox]
name=Sandbox repository (tutorials)
gpgcheck=0
enabled=0
baseurl=http://dev2.hortonworks.com.s3.amazonaws.com/repo/dev/
master/utils/

$\text{yum clean all}
yum update
$\text{sudo yum localinstall sbt-0.13.12.rpm}$
$\text{sbt -update}$
$\text{sudo scp -P 2222 -r ../spamTopWords/* root@127.0.0.1:/tmp/
spamTopWords/
```

3 TASK

Firstly we created Spark Context with:

Than we called function *probaWordDir* with defined spark context as well as folder name for which we want to count words.

```
val (probaHW, nbHFiles) = probaWordDir(sc)(args(0)+"ham/*.txt")
print("number of files in "+ args(0)+"ham/*.txt" +":")
println(nbHFiles)

//process spam files
val (probaSW, nbSFiles) = probaWordDir(sc)(args(0)+"spam/*.txt")
print("number of files in "+ args(0)+"spam/*.txt" +":")
println(nbSFiles)
```

For each value and variable in the code we put its type in the comment in code in the report. Function: probaWordDir:

```
1 def probaWordDir(sc:SparkContext)(filesDir:String)
2 :(RDD[(String, Double)], Long) = {
3
```

```
4
         //sc -> class org.apache.spark.SparkContext
5
         //filesDir -> java.lang.String
6
         //read the files
7
         val rdd = sc.wholeTextFiles(filesDir)
         //rdd -> class org.apache.spark.rdd.MapPartitionsRDD
8
9
         // The number of files is counted and stored in a variable
             → nbFiles
10
         val nbFiles = rdd.count()
          //nbFiles -> long
11
          // Non informative words must be removed from the set of
12
             \hookrightarrow unique words.
         val stopWords = Set(".", ":", ",", " ", "/", "\\", "-", "'",
13
             \hookrightarrow "(", ")", "@", "Subject:")
14
          //stopWords -> class scala.collection.immutable.HashSet\
             → $HashTrieSet
15
          // get the words in an email, delete the dublicate in one
             → email, delete
                               the stop words
16
         val wordBagRdd: RDD[(String, Set[String])] = rdd.map(

    textTuple =>

17
                  (textTuple._1, textTuple._2.trim().
18
                  split("\\s+").toSet.diff(stopWords)))
19
         //wordBagRdd -> class org.apache.spark.rdd.MapPartitionsRDD
20
         // count the words in all emails
         val wordCountRdd: RDD[(String, Int)] = wordBagRdd.flatMap(x
21
             \hookrightarrow => x._2.map(y => (y, 1))).reduceByKey(_ +_)
         //wordCountRdd -> class org.apache.spark.rdd.ShuffledRDD
22
23
         //calculate the probability
         val probaWord: RDD[(String, Double)] = wordCountRdd.map(x =>
24
             \hookrightarrow (x._1, x._2.toDouble / nbFiles))
25
         //probaWord -> class org.apache.spark.rdd.MapPartitionsRDD
         return (probaWord, nbFiles)
26
27
28
```

4 TASK

We computed function: computeMutualInformationFactor with given formula:

 $P(occurs, class)log_2\left(\frac{P(occurs, class)}{P(occurs)P(class)}\right)$

```
def computeMutualInformationFactor(
probaWC: RDD[(String, Double)],//prob of just a class, some word
could not be
probaW: RDD[(String, Double)],//all words prob, all word
probaC: Double, //prb of a class : class mails / all mails
```

```
5
     probaDefault: Double // default value when a probability is
        → missing
6
  ): RDD[(String, Double)] = {
7
               //got (word,(prob for both classes, prob for class)),
                  \hookrightarrow if the prob f
                                      or class does not exist set the
                  → default
8
              val probWJoin: RDD[(String, (Double, Option[Double]))] =
                     probaW.leftOuterJoin(probaWC)// got all class
                 → probs, if not -> default
9
                                   //p(accurs) p(accurs,class)
       //probWJoin -> class org.apache.spark.rdd.MapPartitionsRDD
10
11
              val valueClassAndOcu: RDD[(String, (Double, Double))] =
                 \hookrightarrow probWJoin.map(x => (x._1, (x._2._1, x._2._2.
                 → getOrElse(probaDefault))))
12
              //calculate the formula for mutual information
13
              valueClassAndOcu.map(x => (x._1, x._2._2 * (math.log(x.
                 \leftarrow _2._2 / (x._2._1 * probaC)) / math.log(2.0)))
14
         //valueClassAndOcu -> class org.apache.spark.rdd.
             → MapPartitionsRDD
15
16
   }
```

probaWC is a RDD with the map structure: word => probability the word occurs in an email of a given class.

probaW has the map structure: word => probability the word occurs (whatever the class). probaC is the probability that an email belongs to the given class.

probaDefault is a probability when a word does not occur in both classes but only one with value given by formula:

totalNumberOfFiles

This function returns the factor of each words (so it returns a RDD) given a class value (spam or ham) and an occurrence value (true or false).

5 Task

- **a.** We computed the couples (probaWordHam, nbFilesHam) for the directory 'ham' and (probaWordSpam, nbFilesSpam) for the directory 'spam'.
- **b.** We computed the probability P(occurs, class) for each word. There are two values of class ('ham' and 'spam') and two values of occurs ('true' or 'true'). Hence, we obtained 4 RDDs, one RDD for each case: (true,ham), (true, spam), (false, ham) and (false, spam). Each RDD has the map structure: word => probability the word occurs (or not) in an email of a given class.
- **c.** We computed the mutual information of each word as a RDD with the map structure: word => MI(word). With the usage of the function computeMutualInformationFactor. If a word occurs in only one class, its joint probability with the other class takes on the

default value probaDefault defined earlier. The function computeMutualInformation-Factor is called 4 times for each possible value of P(occurs, class): (true,ham), (true, spam), (false, ham) and (false, spam).

d. The main function prints on screen the 20 top words (maximizing the mutual information value) which can be used to distinguish a spam from an ham email by using the mutual information.

We have obtained this list of top 20 words:

- (bio,23.820986127869574)
- (touch-tone,23.820986127869574)
- (woodland,23.820986127869574)
- (8080,23.820986127869574)
- (ibi,23.820986127869574)
- (wales, 23.820986127869574)
- (pearce,23.820986127869574)
- (slap,23.820986127869574)
- (commissioner,23.820986127869574)
- (n,23.820986127869574)
- (2442,23.820986127869574)
- (cake,23.820986127869574)
- (dawson,23.820986127869574)
- (detailed,23.820986127869574)
- (trilogy,23.820986127869574)
- (miranda,23.820986127869574)
- (piggy,23.820986127869574)
- (marke,23.820986127869574)
- (lightn,23.820986127869574)
- (pristine,23.820986127869574)

Once we get these top words, they can be used to classify spam emails or ham emails based on the number of occurrences of these top words in those emails. So if there is an email has a high frequency of these top words, it will be easily classified to spam emails. These top words have bigger mutual information factor, that means they appear rarely in ham emails and they are more frequently in spam emails. So we can classify ham or spam emails based on these top words.

e. These top words are also stored on HDFS in the file '/tmp/topWords.txt'

Main function:

```
1
   def main(args: Array[String]) {
2
3
         if(args.size > 0){
                  val conf = new SparkConf().setAppName("Spam Filter
4
                     → Application").setMaster("local")
5
              //conf -> class org.apache.spark.SparkConf
6
              //initiate spark context
7
                  val sc = new SparkContext(conf)
8
              //sc -> class org.apache.spark.SparkContext
                  println("Got the path:"+args(0))
9
10
                  // args(0) should be something like "hdfs:///project
                     → /, see readme
11
12
                  //process ham files
                  val (probaHW, nbHFiles) = probaWordDir(sc)(args(0)+"
13
                      → ham/*.txt")
                  //probaHW -> class org.apache.spark.rdd.
14
                     \hookrightarrow MapPartitionsRDD
15
              //nbHFiles -> long
16
                  //process spam files
17
                  val (probaSW, nbSFiles) = probaWordDir(sc)(args(0)+"

    spam/*.txt")

18
              //probaSW -> class org.apache.spark.rdd.MapPartitionsRDD
19
              // nbSFiles -> long
                  print("number of files in "+ args(0)+"ham/*.txt" +":
20
                     → ")
21
                  println(nbHFiles)
                  print("number of files in "+ args(0)+"spam/*.txt" +"
22
                      23
                  println(nbSFiles)
24
                  val nbFiles = nbSFiles + nbHFiles
25
26
              //nbFiles -> long
27
28
                  val probaWs = probaSW.map(x \Rightarrow (x._1,(x._2,1))).
                      \hookrightarrow union(probaHW.map(x => (x._1,(x._2,0))))
29
                  //probaWs -> class org.apache.spark.rdd.UnionRDD
30
                  val probaW = probaWs.reduceByKey((x,y) => if(y._2<1)
                     \hookrightarrow ((x._1*nbSFiles.toDouble+y._1*nbHFiles.

    toDouble)/(nbFiles.toDouble),1) else ((y._1*)
                     → nbSFiles.toDouble+x._1*nbHFiles.toDouble)/(
                     \hookrightarrow nbFiles.toDouble) ,0)) .map(x => (x._1,x._2._1
31
              //probaW -> class org.apache.spark.rdd.MapPartitionsRDD
32
33
34
35
                  //Compute the probability P(occurs, class) for each
```

```
→ word.
36
37
                  val probaH = nbHFiles.toDouble / nbFiles.toDouble //
                      \hookrightarrow the probability that an email belongs to the
                      → given class.
38
                  //probaH -> double
39
                  val probaS = nbSFiles.toDouble / nbFiles.toDouble
40
                  //probaS -> double
                  // Compute mutual information for each class and
41
                      → occurs
42
                  val MITrueHam = computeMutualInformationFactor(
                      → probaHW, probaW, probaH, 0.2 / nbFiles) // the
                      → last is a default value
43
                  //MITrueHam -> class org.apache.spark.rdd.
                      → MapPartitionsRDD
                  val MITrueSpam = computeMutualInformationFactor(
44
                      → probaSW, probaW, probaS, 0.2 / nbFiles)
45
                  //MITrueSpam -> class org.apache.spark.rdd.
                     → MapPartitionsRDD
                  val MIFalseHam = computeMutualInformationFactor(
46
                      \rightarrow probaHW.map(x => (x._1, 1.00001 - x._2)),
                      → probaW, probaH, 0.2 / nbFiles)
47
                  //MIFalseHam -> class org.apache.spark.rdd.
                     → MapPartitionsRDD
48
                  val MIFalseSpam = computeMutualInformationFactor(
                      \hookrightarrow probaSW.map(x => (x._1, 1.00001 - x._2)),
                      \hookrightarrow probaW, probaS, 0.2 / nbFiles)
49
                  //MIFalseSpam -> class org.apache.spark.rdd.
                      → MapPartitionsRDD
50
51
                  println("print top MIFalseSpam prob:")
52
53
                  MIFalseSpam.top(10)(Ordering[Double].on(x \Rightarrow x._2)).
                      → foreach{ println }
                  println("print top MIFalseHam prob:")
54
55
                  MIFalseHam.top(10)(Ordering[Double].on(x \Rightarrow x._2)).
                      → foreach{ println }
56
                  println("print top MITrueSpam prob:")
57
                  MITrueSpam.top(10)(Ordering[Double].on(x \Rightarrow x._2)).
                      → foreach{ println }
                  println("print top MITrueHam prob:")
58
59
                  MITrueHam.top(10)(Ordering[Double].on(x => x._2)).
                      → foreach{ println }
60
                  //sum the mutual information
61
62
                  val MI :RDD[(String, Double)] = MITrueHam.union(
                      → MITrueSpam).union(MIFalseHam).union(
                      \hookrightarrow MIFalseSpam).reduceByKey( (x, y) => x + y)
                  //MI -> class org.apache.spark.rdd.ShuffledRDD
63
```

```
//{\tt These} words must be also stored on HDFS in the
64
                     → file âĂIJ/tmp/topWords.txtâĂİ.
65
                  val path: String = "/tmp/topWords.txt"
                  //path -> class java.lang.String
66
67
                  val topTenWords: Array[(String, Double)] = MI.top
                      \hookrightarrow (20) (Ordering [Double] .on(x => x._2))
68
                  //topTenWords -> class [Lscala.Tuple2;
                  //save the top 20 words
69
70
                  sc.parallelize(topTenWords).keys.coalesce(1, true).
                      → saveAsTextFile(path)
71
          }
72
          else
73
                  println("Please write te directory where the ham and
74 }
```