

# **Chapter 12: Spark Standalone Cluster**

## Ex1: NLP - Tags

Requirement: Build a tags filter. Use the various NLP tools and a classifier, to predict tag for one question. In future questions could be auto-tagged by such a classifier or tags could be recommended to users prior to posting.

- Dataset: stack-overflow-data.csv. It contains Stack Overflow questions and associated tags.
- Link tham khảo: <a href="http://benalexkeen.com/multiclass-text-classification-with-pyspark/">http://benalexkeen.com/multiclass-text-classification-with-pyspark/</a>)

```
In [1]: # Link HDFS: http://172.24.40.251:50070/
 In [2]:
         import findspark
         findspark.init()
 In [3]:
         import pyspark
 In [4]:
         from pyspark.sql import SparkSession
         from pyspark import SparkContext
         from pyspark import SparkConf
 In [5]: SparkContext.setSystemProperty('spark.executor.memory', '6g')
         sc = SparkContext(master='spark://172.25.53.2:7077', appName='Stack Overflow')
 In [7]: sc
 Out[7]: SparkContext
         Spark UI (http://PM503-GV:4041)
         Version
          v2.4.4
         Master
          spark://172.25.53.2:7077
         AppName
          Stack Overflow
 In [8]:
         spark = SparkSession(sc)
In [10]:
         file name = "hdfs://172.24.40.251:19000/stack overflow data.csv"
```

```
In [11]: data = spark.read.csv(file name, inferSchema=True,header=True)
In [12]: data.show(5)
                          post
         |what is causing t...|
                                        c#|
         |have dynamic html...| asp.net|
         |how to convert a ...|objective-c|
         |.net framework 4 ...|
                                    .net
         |trying to calcula...|
                                    python|
         +----+
         only showing top 5 rows
In [13]: data.groupby('tags').count().show(30)
                   tags | count |
                 iphone| 2000|
                android | 2000 |
                     c#| 2000|
                   null | 20798 |
                asp.net | 2000 |
                   html | 2000 |
                  mysq1| 2000|
                 jquery| 2000|
             javascript | 2000 |
                    css | 2000 |
                    sq1 | 2000 |
                    c++ | 2000 |
                      c| 2000|
            objective-c| 2000|
                   java| 2000|
                    php | 2000 |
                   .net| 2000|
                    ios | 2000 |
                 python| 2000|
              angularjs | 2000 |
         |ruby-on-rails| 2000|
         +-----+
In [14]: | tags_null_data = data.filter(data.tags.isNull())
In [15]: tags_null_data.count()
Out[15]: 20798
In [16]: data = data.filter(data.tags.isNotNull())
```

```
In [17]: data.count()
Out[17]: 40000
In [18]: from pyspark.sql.functions import *
```

## **Clean and Prepare the Data**

\*\* Create a new length feature: \*\*

```
In [19]: from pyspark.sql.functions import length
In [20]: data = data.withColumn('length',length(data['post']))
In [21]: data.show()
```

what is causing t... c#| 833 |have dynamic html...| asp.net| 804 |how to convert a ...| objective-c| 755 .net framework 4 ... .net| 349 |trying to calcula...| python| 1290 how to give alias... asp.net| 309 |window.open() ret...| angularjs| 495 iphone| |identifying serve...| 424 unknown method ke...|ruby-on-rails| 2022 angularjs| from the include ... 1279 |when we need inte...| c# 995 |how to install .i...| 344 ios| |dynamic textbox t...| 389 asp.net rather than bubbl... 1338 c site deployed in ... asp.net| 349 |connection in .ne...| .net| 228 62 |how to subtract 1...| objective-c| ror console show ...|ruby-on-rails| 2594 |distance between ...| iphone| 336 l |sql query - how t...| sq1| 1037

only showing top 20 rows

In [22]: # Pretty Clear Difference
data.groupby('tags').mean().show()



```
tags|avg(length)|
               709.621
       iphone|
      android|
               1713.4345
               1145.3065
           c#|
                  999.95
      asp.net|
         html|
                891.3105
        mysq1|
                1038.561
       jquery|
                1081.507
   javascript|
                964.396
                 954.809
          css
                 870.912
          sql
          c++|
                1295.955
            c |
               1121.1115
  objective-c|
                972.8925
         java|
                1357.308
               1123.4205
          php|
                731.0075
         .net|
          ios
                970.7565
               1018.6695
       python|
    angularjs|
               1294.7545
|ruby-on-rails|
               1244.2055
  -----+
```

## **Feature Transformations**

```
In [23]: from bs4 import BeautifulSoup

from pyspark import keyword_only
from pyspark.ml import Transformer
from pyspark.ml.param.shared import HasInputCol, HasOutputCol
from pyspark.sql.functions import udf
from pyspark.sql.types import StringType
```

```
In [24]: class BsTextExtractor(Transformer, HasInputCol, HasOutputCol):
             @keyword only
             def init (self, inputCol=None, outputCol=None):
                  super(BsTextExtractor, self).__init__()
                  kwargs = self. input kwargs
                  self.setParams(**kwargs)
             @keyword only
              def setParams(self, inputCol=None, outputCol=None):
                  kwargs = self. input kwargs
                  return self._set(**kwargs)
             def transform(self, dataset):
                  def f(s):
                     cleaned post = BeautifulSoup(s).text
                     return cleaned post
                  t = StringType()
                  out col = self.getOutputCol()
                  in_col = dataset[self.getInputCol()]
                  return dataset.withColumn(out col, udf(f, t)(in col))
In [25]: from pyspark.ml.feature import Tokenizer, StopWordsRemover, CountVectorizer, IDF, St
```

```
text_extractor = BsTextExtractor(inputCol="post", outputCol="cleaned_post")
tokenizer = Tokenizer(inputCol="cleaned_post", outputCol="token_text")
stopremove = StopWordsRemover(inputCol='token text',outputCol='stop tokens')
count vec = CountVectorizer(inputCol='stop tokens',outputCol='c vec')
idf = IDF(inputCol="c_vec", outputCol="tf_idf")
class to num = StringIndexer(inputCol='tags',outputCol='label')
```

```
In [26]: from pyspark.ml.feature import VectorAssembler
         from pyspark.ml.linalg import Vector
```

```
In [27]:
         clean_up = VectorAssembler(inputCols=['tf_idf','length'],outputCol='features')
```

#### The Model

We'll use Naive Bayes, but feel free to play around with this choice!

```
In [28]:
         from pyspark.ml.classification import NaiveBayes
```

```
In [29]:
         # Use defaults
         nb = NaiveBayes()
```

#### **Pipeline**

```
In [30]: from pyspark.ml import Pipeline
```

```
In [31]: data_prep_pipe = Pipeline(stages=[class_to_num,text_extractor,tokenizer,stopremov
In [32]: cleaner = data_prep_pipe.fit(data)
In [33]: clean_data = cleaner.transform(data)
```

### **Training and Evaluation!**

only showing top 20 rows

```
clean_data = clean_data.select(['label', 'features'])
In [34]:
In [35]:
          clean data.show()
           |label|
                                features
            14.0 (262145, [0,1,2,3,...
              3.0 (262145, [0, 12, 31, ... ]
            16.0 (262145, [0, 1, 2, 3, ...
              8.0 (262145, [0, 18, 21, ...
             9.0 (262145, [0,1,4,8,...
              3.0 (262145, [0, 12, 21, ...
            11.0 (262145, [0,1,3,6,...
            18.0 (262145, [0,44,61,...
              6.0 (262145, [0, 1, 14, 2...
            11.0 (262145, [0, 1, 3, 4, . . .
            14.0 (262145, [0, 2, 3, 6, ...
              1.0 (262145, [0, 18, 27, ... ]
              3.0 (262145, [0, 7, 12, 1...
            13.0 (262145, [0,1,2,3,...
              3.0 (262145, [0, 11, 27, ...
              8.0 (262145, [0, 187, 23...
            16.0 (262145, [0, 10, 15, ...
              6.0 (262145, [0,1,3,12...
            18.0 (262145, [0, 30, 39, ...
              0.0 (262145, [0, 12, 15, ...
```

```
In [36]: (training,testing) = clean_data.randomSplit([0.7,0.3], seed=142)
In [41]: predictor = nb.fit(training)
In [42]: test_results = predictor.transform(testing)
```



#### In [43]: test\_results.show()

```
----+-----
|label|
                 features|
                               rawPrediction|
                                                    probability|predictio
  0.0|(262145, [0,1,4,11...|[-21169.668465935...|[1.21437232270286...|
                                                                    12.
01
  0.0|(262145,[0,1,4,14...)[-1167.4513441585...][1.0,7.6094842671...]
                                                                     0.
0
  0.0|(262145, [0,1,5,14...][-11564.690744964...][1.0,0.0,0.0,0.0,...]
                                                                     0.
01
  0.0|(262145, [0,1,5,14...][-10439.407817798...][1.93390066856389...]
                                                                    12.
  0.0|(262145,[0,1,7,9,...)[-7336.5344958838...][1.0,1.5922509763...]
                                                                     0.
0
  0.0|(262145, [0,1,7,11...][-8486.4019708681...][1.0,2.0825090588...]
                                                                     0.
  0.0|(262145,[0,1,9,10...|[-3738.5573599120...|[7.30286711565549...|
                                                                    12.
01
  0.0|(262145,[0,1,9,12...][-6617.1798543878...][0.99995740820888...]
                                                                     0.
01
  0.0|(262145, [0,1,9,12...|[-1693.4933675940...|[2.32251400846856...|
                                                                    12.
  0.0|(262145,[0,1,10,1...|[-3989.2579384985...|[1.0,3.3928388326...]
                                                                     0.
01
  0.0|(262145,[0,1,10,1...)[-2373.9652840133...][1.0,2.7652251546...]
                                                                     0.
  0.0|(262145, [0,1,11,1...][-8818.6290325138...][1.0,1.4580653626...]
                                                                     0.
01
  0.0|(262145,[0,1,11,1...|[-4285.8899650084...|[1.0,9.2886878940...]
                                                                     0.
01
  0.0|(262145, [0,1,11,1...][-2804.2071232226...][1.0,3.5487447155...]
                                                                     0.
0 l
  0.0|(262145, [0,1,11,1...][-2115.4985343601...][6.48717059350829...]
                                                                     4.
01
  0.0|(262145, [0,1,12,1...][-2828.9699626115...][0.13219295903883...]
                                                                    12.
01
  0.0|(262145,[0,1,12,1...)[-5997.8626004770...][1.0,1.4086114099...]
                                                                     0.
01
  0.0|(262145, [0,1,12,1...|[-3875.2807055205...|[1.0,2.0818433425...|
                                                                     0.
  0.0|(262145, [0,1,12,1...|[-5370.3720000707...|[1.0,2.0080477642...|
                                                                     0.
0
  0.0|(262145,[0,1,12,1...][-7260.3672598605...][1.0,1.6809702518...]
                                                                     0.
only showing top 20 rows
```

localhost:8888/notebooks/Chapter12/Ex1 NLP TagFilters.ipynb



```
In [44]: # Create a confusion matrix
    test_results.groupBy('label', 'prediction').count().show()
```

```
|label|prediction|count|
   8.0
              3.0
 16.0
              8.0
                     20
 19.0
              5.0
                      1
 10.0
              1.0
                      3 |
 12.0
              5.0
                      2
   0.0|
             12.0
                    169
   0.01
              8.0
                     17
   1.0
             19.0
                      9|
   1.0
             12.0
                      1
   7.0
              3.0
                     11
 15.0
             11.0
                     12
 19.0
             12.0
                     1
 17.0
             7.0
                     12
 17.0
             19.0
                     15
             17.0
 11.0
                      7
   8.0
              6.0
                      1|
 17.0
              9.0
                      6
   4.0
              6.0
                      1|
   3.0
              9.01
                      3|
   3.0
              5.0
                      1|
only showing top 20 rows
```

```
In [45]: from pyspark.ml.evaluation import MulticlassClassificationEvaluator
```

```
In [46]: acc_eval = MulticlassClassificationEvaluator()
    acc = acc_eval.evaluate(test_results)
    print("Accuracy of model at predicting: {}".format(acc))
```

Accuracy of model at predicting: 0.7148108175153408

```
In [48]: # save hdfs
nb.save("hdfs://172.24.40.251:19000/NB_TagFilters_model")
```

- Not very good result! (~72%)
- Solution: Try switching out the classification models! Or even try to come up with other engineered features!...

### **Use LogisticRegression/Random Forest**

## **Logistic Regression**

In [49]: from pyspark.ml.classification import RandomForestClassifier, LogisticRegression

```
In [50]: | lg = LogisticRegression(maxIter=20, regParam=0.3, elasticNetParam=0)
         predictor_1 = lg.fit(training)
In [51]:
In [52]:
         test results 1 = predictor 1.transform(testing)
In [53]: # Create a confusion matrix
         test_results_1.groupBy('label', 'prediction').count().show()
         +----+
         |label|prediction|count|
            8.0
                       3.0
                              47
           16.0
                       8.0
                              10|
           19.0
                       5.0
                               1
           10.0
                       1.0
                               1|
                               2 |
            2.0
                       0.0
           15.0
                      16.0
                               1
           12.0
                       5.0
                               3 |
            0.0
                      12.0
                              94
            0.01
                       8.0
                               8
            1.0
                      19.0
                              12
            1.0
                      12.0
                               3|
            7.0
                       3.0
                               61
                               5 |
           15.0
                      11.0
           19.0
                      12.0
                               5 |
           11.0
                      17.0
                              15
           17.0
                       7.0
                              13
                      19.0
           17.0
                               2|
           17.0
                       9.0
                              11
                               5 |
            8.0
                       6.0
            6.0
                       1.0
                               2
         only showing top 20 rows
         acc_eval = MulticlassClassificationEvaluator()
In [54]:
         acc_1 = acc_eval.evaluate(test_results_1)
         print("Accuracy of model at predicting: {}".format(acc 1))
         Accuracy of model at predicting: 0.7182544021412621
In [55]:
         ## It's not better result!!!
In [57]: # save hdfs
         lg.save("hdfs://172.24.40.251:19000/LG TagFilters model")
```

#### **Random forest**

```
In [58]: rf = RandomForestClassifier(labelCol="label", \
                                    featuresCol="features", \
                                    numTrees = 500, \
                                    maxDepth = 5, \
                                    maxBins = 64)
In [59]: predictor 2 = rf.fit(training)
In [60]: test results 2 = predictor 2.transform(testing)
In [61]: # Create a confusion matrix
         test_results_2.groupBy('label', 'prediction').count().show()
         +----+
         |label|prediction|count|
            8.0
                       3.0
                             33|
           16.0
                      8.0
                             31
           19.0
                       5.0
                              1
            0.0
                      12.0
                             268
            1.0
                      19.0
                              5 |
            1.0
                      12.0
                             19
            0.0
                      8.0
                              3 |
                             28
           19.0
                      12.0
           7.0
                      3.0
                              1
           15.0
                      11.0
                              1
           11.0
                      17.0
                              4
           17.0
                      7.0
                             16
            8.0
                      6.0
                              4|
          17.0
                      9.0
                              5 |
            6.0
                       1.0
                              1
                      9.0
            3.0
                              1
            4.0
                      6.01
                              1
                              9|
            6.0
                      8.0
           15.0
                      19.0
                              1|
                              3|
          14.0
                      7.0
```

only showing top 20 rows

test\_results\_2.groupBy('prediction').count().show() In [62]:



```
-----
|prediction|count|
       8.0 | 1046 |
             375 l
       0.01
       7.0
             582
      18.0
             426
       1.0
             524
       4.0
             762
       11.0
             580
       14.0
             251
       3.0
             491
      19.0
             396
       2.0
             647
      17.0
             402
      10.0
             624
      13.0
             720
       6.0
             578
      15.0
             605
       5.0
             520
       9.0
             592
      16.0
             656
      12.0 | 1113 |
+----+
acc eval = MulticlassClassificationEvaluator()
acc_2 = acc_eval.evaluate(test_results_2)
print("Accuracy of model at predicting: {}".format(acc_2))
Accuracy of model at predicting: 0.7231428382621157
```

```
In [63]:
```

```
In [64]: ## It has higher accuracy but is not a better result!!!
```

```
In [66]:
         # save hdfs
         rf.save("hdfs://172.24.40.251:19000/RF_TagFilters_model")
```

```
In [67]: | sc.stop()
```