Memento

"A retentive memory may be a good thing, but the ability to forget is the true token of greatness." – Elbert Hubbard

Spark: Custom UDF Example

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UDF (User defined functions) and UDAF (User defined aggregate functions) are key components of big data languages such as Pig and Hive. They allow to extend the language constructs to do adhoc processing on distributed dataset. Previously I have blogged about how to write custom UDF/UDAF in Pig (here) and Hive(Part I & II). In this post I will focus on writing custom UDF in spark. UDF and UDAF is fairly new feature in spark and was just released in Spark 1.5.1. So its still in evolution stage and quite limited on things you can do, especially when trying to write generic UDAFs. I will talk about its current limitations later on.

As a motivating example assume we are given some student data containing student's name, subject and score and we want to convert numerical score into ordinal categories based on the following logic:

- A -> if score >= 80
- B -> if score >= 60
- C -> if score >= 35
- D -> otherwise

Below is the relevant python code if you are using pyspark.

```
1
     # Generate Random Data
     import itertools
     import random
     students = ['John', 'Mike', 'Matt']
     subjects = ['Math', 'Sci', 'Geography', 'History']
     random.seed(1)
 6
 7
     data = []
 8
 9
     for (student, subject) in itertools.product(students, subjects):
10
         data.append((student, subject, random.randint(0, 100)))
11
12
     # Create Schema Object
13
     from pyspark.sql.types import StructType, StructField, IntegerType, StringType
14
     schema = StructType([
15
                 StructField("student", StringType(), nullable=False),
                 StructField("subject", StringType(), nullable=False),
16
                 StructField("score", IntegerType(), nullable=False)
17
18
         ])
19
20
     # Create DataFrame
21
     from pyspark.sql import HiveContext
22
     sqlContext = HiveContext(sc)
23
     rdd = sc.parallelize(data)
24
     df = sqlContext.createDataFrame(rdd, schema)
25
26
     # Define udf
     from pyspark.sql.functions import udf
27
     def scoreToCategory(score):
28
29
         if score >= 80: return 'A'
         elif score >= 60: return 'B'
30
31
         elif score >= 35: return 'C'
32
         else: return 'D'
33
```

```
udfScoreToCategory=udf(scoreToCategory, StringType())
df.withColumn("category", udfScoreToCategory("score")).show(10)
```

Line 2-10 is the basic python stuff. We are generating a random dataset that looks something like this:

student	subject	score
John	Math	13
	•••	
Mike	Sci	45
Mike	Geography	65
	•••	

Next line 12-24 are dealing with constructing the dataframe. The main part of the code is in line 27-34. We first define our function in a normal python way.

Below is scala example of the same:

```
// Construct Dummy Data
import util.Random
import org.apache.spark.sql.Row
implicit class Crossable[X](xs: Traversable[X]) {
   def cross[Y](ys: Traversable[Y]) = for { x <- xs; y <- ys } yield (x, y)
}

val students = Seq("John", "Mike", "Matt")
val subjects = Seq("Math", "Sci", "Geography", "History")
val random = new Random(1)</pre>
```

```
10
     val data =(students cross subjects).map{x => Row(x. 1, x. 2, random.nextInt(100))}.t
11
12
     // Create Schema Object
     import org.apache.spark.sql.types.{StructType, StructField, IntegerType, StringType}
13
14
     val schema = StructType(Array(
                  StructField("student", StringType, nullable=false),
StructField("subject", StringType, nullable=false),
15
16
                  StructField("score", IntegerType, nullable=false)
17
18
         ))
19
20
     // Create DataFrame
21
     import org.apache.spark.sql.hive.HiveContext
22
     val rdd = sc.parallelize(data)
     val df = sqlContext.createDataFrame(rdd, schema)
23
24
25
     // Define udf
26
     import org.apache.spark.sql.functions.udf
     def udfScoreToCategory=udf((score: Int) => {
27
28
              score match {
29
              case t if t >= 80 => "A"
30
              case t if t >= 60 => "B"
31
              case t if t >= 35 => "C"
32
              case => "D"
33
         }})
     df.withColumn("category", udfScoreToCategory(df("score"))).show(10)
34
```

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