Hadoop Training Apache Spark SQL Playing With Key Features Of Spark SQL

In this series of exercise we practice SQL-related features of Spark such as SQL-like queries and DataFrame - both on files in HDFS and tables in Hive.

Getting ready

- 1.Login using ssh to edgenode cdh00.cl.ii.pw.edu.pl using your Linux account.
- 2.Go to home directory, create a data folder and copy sample Parquet file. (You can have a look at this file):

```
cp /data/local/datascience/data/users.parquet
/data/local/datascience/home/${USER}/data

cd /data/local/datascience/home/${USER}/data

head users.parquet

hdfs dfs -put users.parquet /user/${USER}

hdfs dfs -ls /user/${USER}
```

3. Start the spark-shell:

```
spark-shell --master yarn --deploy-mode client --executor-memory
2048m --num-executors 4 --conf spark.ui.port=95<YOUR-NUMBER>
```

DataFrame Operations On A Parquet File In HDFS

4. Now, load the Parquet file directly into SQL context:

```
scala> sc.setLogLevel("INFO")
scala> val user_df =
spark.sqlContext.read.parquet("/user/${USER}/users.parquet")
```

You can use the show() and printSchema() operators to preview the sample of the data and see the schema.

scala> user_df.show()

```
fname| lname|gender| birthdate|
                                               state| registration|
+---+

      1 | EDWIN | BATES |
      M | 1965-03-14 |
      Colorado |
      2013-12-25 |

      2 | ISABELLE | HEATH |
      F | 1999-09-12 |
      Oregon |
      2013-05-15 |

      3 | ANNIKA | FREEMAN |
      F | 1967-09-23 |
      Michigan |
      2013-09-22 |

      4 | VALENTINA | LARSON |
      F | 1951-07-13 |
      Wisconsin |
      2013-11-07 |

+---+
only showing top 20 rows
scala> user_df.printSchema()
root
 |-- id: integer (nullable = true)
 |-- name: string (nullable = true)
 |-- surname: string (nullable = true)
 |-- gender: string (nullable = true)
 |-- bday: string (nullable = true)
 |-- state: string (nullable = true)
```

Show only one row of the dataset.

The DataFrame API provides a set of useful operators that you can use to manipulate your dataset. Go ahead and try to run few of them!

```
scala> user_df.select("name").show()
scala> user_df.select("name", "gender").show()
scala> user_df.select("gender").distinct().show()
scala> user_df.select("name").distinct().count()
scala> user_df.select("name").take(1)
```

|-- registration: string (nullable = true)

As we reuse the same dataset over and over, let's cache it.

```
scala> val user_df_cached = user_df.cache()
```

Spark caches datasets lazily. It is cached in the executors' memory the first time we access a given dataset. To warm up the cache, run simple command:

```
scala> user_df_cached.select("name").take(1)
```

From now on, you will be processing cached data. Run the same command again to see that this time we read our data from the memory:

```
scala> user_df_cached.select("name").take(1)
```

To practice a bit, run following lines of the code and don't be afraid of modifying them to experiment:

```
scala> user_df_cached.groupBy("state").count().show()
scala> user_df_cached.groupBy("state").count().filter("count > 50").show()
scala> user_df_cached.groupBy("state").count().orderBy("count").show()
scala> user_df_cached.groupBy("state").count().orderBy(desc("count")).show()
scala> user_df_cached.groupBy("state","gender").count().orderBy(asc("state")).show()
scala> user_df_cached.groupBy("state").agg(min("bday")).show()
```

Feel free to investigate the API and try more operators.

Spark SQL Operations On Parquet File

To use Spark SQL statements on a Parquet file, we have to register it at first.

```
scala> user_df.createOrReplaceTempView("user")
```

Now you can query the user table using SQL statements:

```
scala> val count_by_gender = sql("select gender, count(*) as cnt from user
group by gender order by cnt desc")
```

```
16/02/22 15:09:38 INFO ParseDriver: Parsing command: select gender, count(*) as cnt from user group by gender order by cnt desc 16/02/22 15:09:40 INFO ParseDriver: Parse Completed
```

Print the results using the regular Spark non-SQL code:

Hurray! It works:)

Write the following query:

a) Return the state which had highest number of woman registered.

Spark SQL in Hive Context

It is high time to query some tables directly from Hive!

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```
scala> val df = spark.sqlContext.sql("SELECT * FROM
datascience.measured_data_orc")
scala> df.limit(5).show()
```

```
----+
-----+
    1|2015-05-18 19:24:00|
                  236 | 0.09920929584628235 |
kW|
       D|Warsaw-Dereniowa|
s
2|2015-05-18 19:24:00|
                  237 | 0.8467751295230832 |
                                 kW|
       D|Warsaw-Dereniowa|
s
                  238| 0.4608953190788161|
                                 kW|
    3 | 2015-05-18 19:24:00 |
       D|Warsaw-Dereniowa|
s
    4|2015-05-18 19:24:00|
                  239 | 0.029231154861803166 |
                                 kW|
s
       D|Warsaw-Dereniowa|
    5|2015-05-18 19:24:00|
                  240 | 0.7698375647136683 |
                                 kW
D|Warsaw-Dereniowa|
```

Write the following query:

a) Return the number (count) of observations and average value (MD_VALUE) for each observation timestamp (MD_TIMESTAMP)

Analysis of Iris data:

- a) Download Iris data from UCI database: https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data to CDH00.
- b) Put the file into HDFS.
- c) Create external table following the schema described in: https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.names
- d) Write SQL query which will answer the following question:
 - For each class of irises, find the average sepal length.
- e) Find the way to read csv data into spark, create spark dataframe and create an SQL query from point 4.