UE Large Scale Processing @ ESIGELEC 2019/2020

14 - Spark SQL - Data Format/Sources and UDF

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Data Format / Sources

- Supported <u>data formats</u>:
 - CSV: comma separated values
 - JSON: JavaScript Object Notation
 - Avro: row-based self-describing binary storage data format with nested data
 - Parquet: columnar-based binary storage format with nested data
 - ORC (Optimized Row Columnar): columnar-based groups of rows in stripes
 - Table: access external RDBMS tables and view
 - and possibly more ...

Data Format / Sources – Semi-Structured (JSON)

- SparkSQL can automatically infers a schema from a set of JSON records
- Schema inference algorithm works in one pass over the data
- Can run a sample of the data if desired

CREATE TABLE tweets USING json OPTIONS (path "my.json")

SELECT loc.lat , loc.long FROM tweets

WHERE text LIKE '%Spark%' AND tags IS NOT NULL

```
"text": "This is a tweet about #Spark",
    "tags ": ["# Spark "],
    "loc ": {" lat ": 45.1 , "long ": 90}
}{
    "text": "This is another tweet",
    "tags ": [],
    "loc ": {" lat ": 39, "long ": 88.5}
}{
    "text": "A #tweet without #location",
    "tags ": ["# tweet", "# location "]
```

Data Format / Sources

- Supported data sources:
 - Local & external files
 - HDFS
 - JDBC
 - RDD, DataSet or DataFrame
 - And many other connectors...

Check: Databricks Data Sources documentation

for more informations

- Most are common connectors provided as Spark packages:
 - spark-avro
 - spark-redshift
 - spark-csv
 - deep-spark
 - spark-mongdb
 - spark-cassandra-connector
 - couchbase-spark-connector
 - pyspark-csv
 - pyspark-elastic
 - elasticsearch-hadoop
 - and more...

Data Format / Sources - spark-packages.org

- Community-Managed
- References to:
 - third-party libraries
 - add-ons
 - Applications...

Easy to use via Maven coordinates:

```
spark-submit \
 --packages <package id>:<version> \
 --conf key=value

    Or via JARs:

spark-submit \
 --jars "spark-csv_2.10-1.0.0.jar, commons-
csv-1.1.jar" \
 --conf key=value
```

DataFrameReader – one class to read all!

- Package:
 - org.apache.spark.sql
- Class
 - DataFrameReader
- Use the "read" attribute/method from:
 - SQLContext
 - HiveContext
 - SparkSession

Note: You can also use the SparkContext to read data but only into a RDD

- Then you can use directly the corresponding input format method :
 - json()
 - csv()
 - parquet()
 - orc()
 - text()
- Or the format / load methods for more flexibility

Note: You can use <u>DataStreamReader</u> for streaming data

<u>DataFrameReader</u> – PySpark Examples

Read a CSV file:

```
from os.path import expanduser
home = expanduser("~") + "/esigelec-ue-lsp-hdp/spark-3.0.0"
path = "file://" + home + "/examples/src/main/resources"
# read with the csv method
df = spark.read \
 .option("sep", ";") \
 .option("inferSchema", "true") \
 .option("header", "true") \
 .csv(path + "/people.csv")
# read with the format/load method
df = spark.read \
 .option("sep", ";") \
 .option("inferSchema", "true") \
 .option("header", "true") \
 .format("csv") \
 .load(path + "/people.csv")
```

• Read a PARQUET file:

```
from os.path import expanduser
home = expanduser("~") + "/esigelec-ue-lsp-hdp/spark-3.0.0"
path = "file://" + home + "/examples/src/main/resources"

# read with the parquet method
df = spark.read.parquet(path + "/users.parquet")

# read with the format/load method
df = spark.read.format("parquet").load(path + "/users.parquet")

# read with the load method
df = spark.read.load(path + "/users.parquet")
```

DataFrameWriter – one class to write all!

- Package:
 - org.apache.spark.sql
- Class
 - DataFrameWriter
- Use the "write" attribute/method from:
 - DataFrame
 - Dataset

- Then you can use directly the corresponding output format method :
 - p json()
 - csv()
 - parquet()
 - orc()
 - text()
- Or the save / saveAsTable methods for more flexibility

<u>DataFrameWriter</u> – PySpark Examples

Write a CSV file:

```
from os.path import expanduser
home = expanduser("~") + "/esigelec-ue-lsp-hdp/spark-3.0.0"
path = "file://" + home + "/examples/src/main/resources"
# write with the csv method
df.write \
 .option("sep", ";") \
 .option("inferSchema", "true") \
 .option("header", "true") \
 .csv(path + "/df-csv-folder")
# write with the format/save method
df. write \
 .option("sep", ";") \
 .option("inferSchema", "true") \
 .option("header", "true") \
 .format("csv") \
 .save(path + "/df-csv-save-folder")
```

• Write a PARQUET file:

```
from os.path import expanduser
home = expanduser("~") + "/esigelec-ue-lsp-hdp/spark-3.0.0"
path = "file://" + home + "/examples/src/main/resources"

# write with the parquet method
df.write.parquet(path + "/df-parquet-folder")

# read with the format/load method
df.write.format("parquet").save(path + "/df-parquet-save-folder")

# write with the load method
df.write.save(path + "/df-parquet-save-direct-folder")
```

SparkSQL Caching

- SparkSQL can materialize (cache) data in memory (columnar storage)
- Columnar cache reduce the memory
- In interactive mode, iterative algorithms will benefit from caching
- Caching can be invoked using:
 - cache() API on DataFrame/Dataset
 - cacheTable() API on tables
 - CACHE TABLE sql statement on tables

SparkSQL UDF (User Defined Function)

- Supports existing Apache Hive UDFs
- SparkSQL provides a series of native UDFs (pyspark.sql.functions)
- Can be developed in :
 - Java
 - Scala
 - Python
- Note: SparkSQL will not try to optimize the code

```
from pyspark import SparkContext, SparkConf
from pyspark.sql import SparkSession
from os.path import expanduser
home = expanduser("~") + "/esigelec-ue-lsp-hdp/spark-
3.0.0"
path = "file://" + home + "/examples/src/main/resources"
sparkSession = SparkSession.builder \
  .master("spark://localhost:7077") \
  .appName("SparkSQL App") \
  .getOrCreate()
def squaredPy(s): return s * s
sparkSession.udf.register("squaredPy", squaredPy, LongType())
sparkSession.range(1, 20).createOrReplaceTempView("test")
sparkSession.sql("select id, squaredPy(id) from test").show()
```