Lab 4.5: Splitting Text Data

Overview

In this lab, we'll work with text data. The data is regularly structured, but since it's in text format, Spark can't deduce the structure on its own.

We'll use some of the DataFrame tools to create a DataFrame with a schema that's easy to use for querying the data.

The data we'll use contains page view data from Wikimedia. For more info on the data itself, see the notes at the end of the lab.

Builds on

None

Run time

30-40 minutes

Wikimedia PageView Data File

We provide a data file (**spark-labs/data/wiki-pageviews.txt**) that contains a dump of pageview data for many projects under the Wikimedia umbrella.

The data file has lines containing four fields.

- 1. Domain.project (e.g. "en.b")
- 2. Page name (e.g. "AP_Biology/Evolution")
- 3. Page view count (e.g. 1)
- 4. Total response size in bytes (e.g. 10662 but for this particular dump, value is always o).

The data is in simple text format, so when we read it in, we get a dataframe with a single column - a string containing all the data in each row. This is cumbersome to work with. In this lab, we'll apply a better schema to this data.

Tasks

- Create a DataFrame by reading in the page view data in spark-labs/data/wikipageviews.txt.
- Once you've created it, view a few lines to see the format of the data.

• You'll see that you have one line of input per dataframe row.

```
// Scala
> val viewsDF=spark.read.text("spark-labs/data/wiki-pageviews.txt")
```

```
# Python
> viewsDF=spark.read.text("spark-labs/data/wiki-pageviews.txt")
```

Split the Lines

Our first step in creating an easier to use schema is splitting each row into separate columns. We'll use the ${\tt split}()$ function defined in the Spark SQL funtions. We've used this in our word count examples in the main manual.

Tasks

- Create a dataframe by splitting each line up.
 - o Use split() to split on a whitespace (pattern of "\\s+")
 - Python: You'll need to import from the functions module as shown in the Python example below
 - Call the resulting column splitLineDF
 - To rename a column, use Scala:
 - as(newName) Python: alias(newName)

```
// Scala
> val splitViewsDF = // ...
```

```
# Python
# import everyting in the functions module
> from pyspark.sql.functions import *
> splitViewsDF = // ...
```

- View the schema of this DataFrame.
- Display a few rows in the DataFrame (there are many).
 - $_{\text{o}}$ If the display of a row is truncated, you can pass an argument to $\mathtt{show}\,()$ to prevent truncation.
 - o Scal:a show(false)

- o Python: show(truncate=False)
- This is slightly better, but still unwieldy. We'll
- apply a finer grained schema next.

Create a Better Schema

We'll create a dataframe with an easier-to-use schema containing the following columns which align with the data in the views file.

- 1. domain: String The domain.project data.
- 2. pageName: String The page name.
- 3. viewCount: Integer The view count.
- 4. size: Long The response size (always o in this file, but we'll keep it in our dataframe).

Tasks

- Use a select to create a dataframe with the schema above.
 - You can select the nth element of a row containing an array via syntax like this:
 - Scala: splitLine(n)
 - **Python:** splitLine[n]
 - $_{\text{o}}$ You can cast an element using <code>cast(dataType)</code> .
 - Providing data type to convert to, e.g. "integer", "long", "boolean", etc.
 - You can name a selected column using:
 - Scala: as (newName)
 - Python: alias (newName)
- Once you've created the dataframe, then: View
 - the schema.
 - View a few rows of the data.
 - Your data should look something like the below.

```
+----+
|domain| pageName|viewCount|size|
+----+
   aa|Main Page|
                      0 1
   aa|Main page|
                  1 |
                      0 |
   aa|User:Savh|
                  11
                      0 1
   aa|Wikipedia|
                  11
                      0 1
 aa.b|User:Savh|
                  1 |
                      0 1
 ----+
```

Try Some Queries

Tasks

Try some queries on the dataframe you've created, e.g.

- Find rows where the viewCount > 500, and display 5 of them.
 - Do the same query as above, but also filter for where the domain is "en".
- Find the 5 rows with the largest viewCount and a domain of "en".
 - o Python: When ordering with orderBy(), you can use either of the following

```
orderBy(column.asc()) /orderBy(colum.desc())
orderBy("columName", ascending=True) /
orderBy("columName", ascending=False)
```

Optional Tasks

Try some other queries, e.g.

- Find the 5 rows with the largest viewCount and a domain of "en", and where the pageName doesn't contain a colon (":").
 - To check if a column contains a string use:

```
Scala: yourColumn.contains("string_to_match")
Python: yourColumn.like("%string_to_match%")
```

- o To negate in a DataFrame boolean expression, you use
 - Scala: ! (exclamation)Python: ~ (tilde)
- [Optional] Think up your own and try them out.

Summary

We can see that text data can require a little more work than data like JSON with a pre-existing structure. Once you've restructured it with a clear schema, which is not usually di"cult, then the full power of DataFrames can be easily applied.

Notes on Data File

The data file we use was downloaded from: Wikimedia March data. For

more information on Wikimedia data, you can see these links:

- Pageview API
- <u>Field Description</u> this is for the older (deprecated) Pagecount API, but describes the dump files from the current API also.

