PD1-Word Count-MTI850-A24

September 24, 2025

1 PD1: Building a word count application

1.1 MTI850 - Big Data Analytics

V2.0 - Fall 2022V2.1 - Fall 2024

Equipe 9

The volume of unstructured text in existence is growing dramatically, and Spark is an excellent tool for analyzing this type of data. In this lab, we will write code that calculates the most common words in the Complete Works of William Shakespeare retrieved from Project Gutenberg. This could also be scaled to larger applications, such as finding the most common words in Wikipedia.

In this PD we will cover: * Part 1: Creating a base DataFrame and performing operations * Part 2: Counting with Spark SQL and DataFrames * Part 3: Finding unique words and a mean value * Part 4: Apply word count to a file

Note that for reference, you can look up the details of the relevant methods in Spark's Python API.

1.2 Spark Context Initialization and Imports

```
import findspark
findspark.init()

# Test module for MTI850
import testmti850

# Util module for MTI850
import utilmti850

import pyspark

from pyspark.sql import SparkSession

spark = SparkSession.builder \
.master("local") \
.appName("Web Server Log Analysis") \
.config("spark.some.config.option", "some-value") \
.getOrCreate()
```

WARNING: Using incubator modules: jdk.incubator.vector

Using Spark's default log4j profile: org/apache/spark/log4j2-defaults.properties Setting default log level to "WARN".

To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).

25/09/24 12:11:42 WARN NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicable

1.3 Part 1: Creating a base DataFrame and performing operations

In this part of the lab, we will explore creating a base DataFrame with spark.createDataFrame and using DataFrame operations to count words.

1.3.1 (1a) Create a DataFrame

We'll start by generating a base DataFrame by using a Python list of tuples and the spark.createDataFrame method. Then we'll print out the type and schema of the DataFrame. The Python API has several examples for using the createDataFrame method.

```
[2]: wordsDF = spark.createDataFrame([('cat',), ('elephant',), ('rat',), ('rat',), ('rat',), ('cat', )], ['word'])

wordsDF.show()

print (type(wordsDF))

wordsDF.printSchema()
```

[Stage 0:> (0 + 1) / 1]

```
+-----+
| word|
+-----+
| cat|
|elephant|
| rat|
| cat|
+-----+

<class 'pyspark.sql.classic.dataframe.DataFrame'>
root
|-- word: string (nullable = true)
```

1.3.2 DataFrame Visualization

For a better visualization of a DataFrame you can use 'DataFrame.limit().toPandas().head()'

1.3.3 (1b) Using DataFrame functions to add an 's'

Let's create a new DataFrame from wordsDF by performing an operation that adds an 's' to each word. To do this, we'll call the select DataFrame function and pass in a column that has the recipe for adding an 's' to our existing column. To generate this Column object you should use the concat function found in the pyspark.sql.functions module. Note that concat takes in two or more string columns and returns a single string column. In order to pass in a constant or literal value like 's', you'll need to wrap that value with the lit column function.

Please replace <FILL IN> with your solution.

After you have created pluralDF you can run the next cell which contains two tests. If you implementation is correct it will print 1 test passed for each test.

This is the general form that exercises will take. Exercises will include an explanation of what is expected, followed by code cells where one cell will have one or more <FILL IN> sections.

The cell that needs to be modified will have:

```
# TODO: Replace <FILL IN> with appropriate code
on its first line.
```

Once the <FILL IN> sections are updated and the code is run, the test cell can then be run to verify the correctness of your solution. The last code cell before the next markdown section will contain the tests.

Note: Make sure that the resulting DataFrame has one column which is named 'word'.

```
[4]: # TODO: Replace <FILL IN> with appropriate code
     from pyspark.sql.functions import lit, concat
     pluralDF = wordsDF.select(concat(wordsDF.word, lit("s")).alias("word"))
     pluralDF.show()
     pluralDF.limit(5).toPandas().head()
     ----+
          word
    +----+
          cats
    |elephants|
          rats
          ratsl
          cats
[4]:
             word
     0
             cats
     1
       elephants
     2
             rats
     3
             rats
     4
             cats
    1.3.4 Test (1b)
[5]: # TEST Using DataFrame functions to add an 's' (1b)
     testmti850.Test.assertEquals(pluralDF.first()[0], 'cats', 'incorrect result:
      ⇔you need to add an s')
     testmti850.Test.assertEquals(pluralDF.columns, ['word'], "there should be one_ |
      ⇔column named 'word'")
    1 test passed.
    1 test passed.
```

1.3.5 (1c) Length of each word

Now use the SQL length function to find the number of characters in each word. The length function is found in the pyspark.sql.functions module.

```
[6]: # TODO: Replace <FILL IN> with appropriate code from pyspark.sql.functions import length
```

```
pluralLengthsDF = pluralDF.select(length(pluralDF.word))
pluralLengthsDF.show()
pluralLengthsDF.limit(5).toPandas().head()
```

```
+-----+
|length(word)|
+-----+
| 4|
| 9|
| 4|
| 4|
+------+
```

```
[6]: length(word)
0 4
1 9
2 4
3 4
4 4
```

1.3.6 Test (1c)

```
[7]: # TEST Length of each word (1c)
from collections.abc import Iterable
asSelf = lambda v: map(lambda r: r[0] if isinstance(r, Iterable) and len(r) ==_u
-1 else r, v)

testmti850.Test.assertEquals(set(asSelf(pluralLengthsDF.collect())), {4, 9, 4,_u
-4, 4},
'incorrect values for pluralLengths')
```

1 test passed.

1.4 Part 2: Counting with Spark SQL and DataFrames

Now, let's count the number of times a particular word appears in the 'word' column. There are multiple ways to perform the counting, but some are much less efficient than others.

A naive approach would be to call **collect** on all of the elements and count them in the driver program. While this approach could work for small datasets, we want an approach that will work for any size dataset including terabyte- or petabyte-sized datasets. In addition, performing all of the work in the driver program is slower than performing it in parallel in the workers. For these reasons, we will use data parallel operations.

```
[14]: # TODO: Replace <FILL IN> with appropriate code
wordCountsDF = wordsDF.groupBy(wordsDF.word).count()
wordCountsDF.show()
```

```
| word|count|
+-----+
| rat| 2|
| cat| 2|
|elephant| 1|
```

1.4.1 Test (2a)

1 test passed.

1.5 Part 3: Finding unique words and a mean value

1.5.1 (3a) Unique words

Calculate the number of unique words in wordsDF. You can use other DataFrames that you have already created to make this easier.

```
wordsDF: ['word']
pluralDF: ['word']
pluralLengthsDF: ['length(word)']
wordCountsDF: ['word', 'count']
```

```
[11]: # TODO: Replace <FILL IN> with appropriate code
uniqueWordsCount = wordCountsDF.count()
print(uniqueWordsCount)
```

3

1.5.2 Test (3a)

```
[12]: # TEST Unique words (3a)
testmti850.Test.assertEquals(uniqueWordsCount, 3, 'incorrect count of unique
→words')
```

1 test passed.

1.5.3 (3b) Means of groups using DataFrames

Find the mean number of occurrences of words in wordCountsDF.

You should use the mean GroupedData method to accomplish this. Note that when you use groupBy you don't need to pass in any columns. A call without columns just prepares the DataFrame so that aggregation functions like mean can be applied.

1.6 Part 4: Apply word count to a file

In this section we will finish developing our word count application. We'll have to build the wordCount function, deal with real world problems like capitalization and punctuation, load in our data source, and compute the word count on the new data.

1.6.1 (4a) The wordCount function

First, define a function for word counting. You should reuse the techniques that have been covered in earlier parts of this lab. This function should take in a DataFrame that is a list of words like wordsDF and return a DataFrame that has all of the words and their associated counts.

```
[19]: # TODO: Replace <FILL IN> with appropriate code

def wordCount(wordListDF):
    """Creates a DataFrame with word counts.

Args:
```

```
wordListDF (DataFrame of str): A DataFrame consisting of one string

column called 'word'.

Returns:

DataFrame of (str, int): A DataFrame containing 'word' and 'count'

columns.

"""

return wordListDF.groupBy(wordsDF.word).count()

wordCount(wordsDF).show()
```

```
+----+
| word|count|
+----+
| rat| 2|
| cat| 2|
|elephant| 1|
```

1.6.2 Test (4a)

1 test passed.

1.6.3 (4b) Capitalization and punctuation

Real world files are more complicated than the data we have been using in this lab. Some of the issues we have to address are: + Words should be counted independent of their capitalization (e.g., Spark and spark should be counted as the same word). + All punctuation should be removed. + Any leading or trailing spaces on a line should be removed.

Define the function removePunctuation that converts all text to lower case, removes any punctuation, and removes leading and trailing spaces. Use the Python regexp_replace module to remove any text that is not a letter, number, or space. If you are unfamiliar with regular expressions, you may want to review this tutorial from Google. Also, this website is a great resource for debugging your regular expression.

You should also use the trim and lower functions found in pyspark.sql.functions.

Note that you shouldn't use any RDD operations or need to create custom user defined functions (udfs) to accomplish this task

```
[24]: # TODO: Replace <FILL IN> with appropriate code
     from pyspark.sql.functions import regexp_replace, trim, col, lower
     def removePunctuation(column):
          """Removes punctuation, changes to lower case, and strips leading and \sqcup
       \hookrightarrow trailing spaces.
         Note:
             Only spaces, letters, and numbers should be retained. Other characters \Box
       ⇒should should be
             eliminated (e.g. it's becomes its). Leading and trailing spaces should \sqcup
       \hookrightarrow be\ removed\ after
             punctuation is removed.
         Args:
             column (Column): A Column containing a sentence.
         Returns:
             Column: A Column named 'sentence' with clean-up operations applied.
         return trim(lower(regexp_replace(column, "[^a-zA-Z0-9]", "")))
     sentenceDF = spark.createDataFrame([('Hi, you!',),
                                              (' No under score!',),
                                              (' * Remove punctuation then,
      ⇔spaces * ',)], ['sentence'])
     sentenceDF.show(truncate=False)
      (sentenceDF
      .select(removePunctuation(col('sentence')))
       .show(truncate=False))
     sentence
     +----+
     |Hi, you!
     | No under_score!
             Remove punctuation then spaces * |
     +----+
                                                                       (0 + 1) / 1
     [Stage 57:>
```

|trim(lower(regexp_replace(sentence, [^a-zA-Z0-9], , 1)))|

1.6.4 Test (4b)

```
[25]: # TEST Capitalization and punctuation (4b)

testPunctDF = spark.createDataFrame([(" The Elephant's 4 cats. ",)])

testmti850.Test.assertEquals(testPunctDF.select(removePunctuation(col('_1'))).

→first()[0],

'the elephants 4 cats',

'incorrect definition for removePunctuation function')
```

1 test passed.

1.7 (4c) Load a text file

For the next part of this lab, we will use the Complete Works of William Shakespeare from Project Gutenberg. To convert a text file into a DataFrame, we use the sqlContext.read.text() method. We also apply the recently defined removePunctuation() function using a select() transformation to strip out the punctuation and change all text to lower case. Since the file is large we use show(15), so that we only print 15 lines.

```
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     | have to check the laws of the country where you are located before using
     Ithis ebook
     |title the complete works of william shakespeare
     |author william shakespeare
     only showing top 15 rows
[39]: shakespeareDF.limit(10).toPandas().head()
[39]:
                                                       text
      O project gutenbergs the complete works of willi...
      2 this ebook is for the use of anyone anywhere i...
      3 most other parts of the world at no cost and w...
      4 whatsoever you may copy it give it away or re...
```

1.7.1 (4d) Words from lines

Before we can use the wordcount() function, we have to address two issues with the format of the DataFrame: + The first issue is that that we need to split each line by its spaces. + The second issue is we need to filter out empty lines or words.

Apply a transformation that will split each 'sentence' in the DataFrame by its spaces, and then transform from a DataFrame that contains lists of words into a DataFrame with each word in its own row. To accomplish these two tasks you can use the split and explode functions found in pyspark.sql.functions.

Once you have a DataFrame with one word per row you can apply the DataFrame operation where to remove the rows that contain '.'

Note that shakeWordsDF should be a DataFrame with one column named word.

```
+----+
       word
+----+
    project
| gutenbergs|
        the
   complete
      worksl
         ofl
    william|
|shakespeare|
         by |
    william|
|shakespeare|
       this
      ebookl
         isl
        for
        the
        usel
         ofl
     anyone
   anywhere
only showing top 20 rows
961306
```

1.7.2 Test (4d)

```
[43]: # TEST Remove empty elements (4d)

testmti850.Test.assertEquals(shakeWordsDF.count(), 961306, 'incorrect value for

⇔shakeWordCount')

testmti850.Test.assertEquals(shakeWordsDF.columns, ['word'], "shakeWordsDF

⇔should only contain the Column 'word'")
```

- 1 test passed.
- 1 test passed.

1.7.3 (4e) Count the words

We now have a DataFrame that is only words. Next, let's apply the wordCount() function to produce a list of word counts. We can view the first 20 words by using the show() action; however, we'd like to see the words in descending order of count, so we'll need to apply the orderBy DataFrame method to first sort the DataFrame that is returned from wordCount().

You'll notice that many of the words are common English words. These are called stopwords. In a later lab, we will see how to eliminate them from the results.

```
[46]: # TODO: Replace <FILL IN> with appropriate code
from pyspark.sql.functions import desc

topWordsAndCountsDF = shakeWordsDF.groupBy("word").count().

→orderBy(desc("count"))

topWordsAndCountsDF.show(20)
```

```
(0 + 1) / 1]
[Stage 78:>
+---+
|word|count|
+---+
| the|30205|
| and | 28386 |
   i|21949|
| to|20923|
of | 18822 |
   a|16182|
| you|14437|
| my|13180|
| in|12232|
|that|11776|
  is| 9713|
| not| 9066|
|with| 8528|
| me| 8263|
```

1.7.4 Test (4e)

[Stage 81:> (0 + 1) / 1]

1 test passed.

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