



Assignment3.1



VERSION AUTHOR

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LANGUAGE

Mariusz Jaskowski

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Python 2 with Spark 2.1

Assignment 3

Welcome to Assignment 3. This will be even more fun. Now we will calculate statistical measures on the test data you have created.

YOU ARE NOT ALLOWED TO USE ANY OTHER 3RD PARTY LIBRARIES LIKE PANDAS. PLEASE ONLY MODIFY CONTENT INSIDE THE FUNCTION SKELETONS Please read why:

<https://www.coursera.org/learn/exploring-visualizing-iot-data/discussions/weeks/3/threads/skjCbNgeEeapeQ5W6suLkA>
(<https://www.coursera.org/learn/exploring-visualizing-iot-data/discussions/weeks/3/threads/skjCbNgeEeapeQ5W6suLkA>). Just make sure you hit the play button on each cell from top to down. There are seven functions you have to implement. Please also make sure that on each change on a function you hit the play button again on the corresponding cell to make it available to the rest of this notebook. Please also make sure to only implement the function bodies and DON'T add any additional code outside functions since this might confuse the autograder.

So the function below is used to make it easy for you to create a data frame from a cloudant data frame using the so called "DataSource" which is some sort of a plugin which allows ApacheSpark to use different data sources.

```
In [1]: #Please don't modify this function
def readDataFrameFromCloudant(database):
    cloudantdata=spark.read.load(database, "org.apache.bahir.cloudant")

    cloudantdata.createOrReplaceTempView("washing")
    spark.sql("SELECT * from washing").show()
    return cloudantdata
```

All functions can be implemented using DataFrames, ApacheSparkSQL or RDDs. We are only interested in the result. You are given the reference to the data frame in the "df" parameter and in case you want to use SQL just use the "spark" parameter which is a reference to the global SparkSession object. Finally if you want to use RDDs just use "df.rdd" for obtaining a reference to the underlying RDD object.

Let's start with the first function. Please calculate the minimal temperature for the test data set you have created. We've provided a little skeleton for you in case you want to use SQL. You can use this skeleton for all subsequent functions. Everything can be implemented using SQL only if you like.

```
In [2]: def minTemperature(df,spark):
        return spark.sql("SELECT MIN(temperature) AS mintemp FROM washing")
        .first().mintemp
```

Please now do the same for the mean of the temperature

```
In [3]: def meanTemperature(df,spark):
        return spark.sql("SELECT AVG(temperature) AS avgtemp FROM washing")
        .first().avgtemp
```

Please now do the same for the maximum of the temperature

```
In [4]: def maxTemperature(df,spark):  
        return spark.sql("SELECT MAX(temperature) AS maxtemp FROM washing")  
        .first().maxtemp
```

Please now do the same for the standard deviation of the temperature

```
In [5]: def sdTemperature(df,spark):  
        return spark.sql("SELECT STDDEV(temperature) AS sdtemp FROM washin  
g").first().sdtemp
```

Please now do the same for the skew of the temperature. Since the SQL statement for this is a bit more complicated we've provided a skeleton for you. You have to insert custom code at four position in order to make the function work. Alternatively you can also remove everything and implement it on your own. Note that we are making use of two previously defined functions, so please make sure they are correct. Also note that we are making use of python's string formatting capabilities where the results of the two function calls to "meanTemperature" and "sdTemperature" are inserted at the "%s" symbols in the SQL string.

```
In [6]: def skewTemperature(df,spark):  
        return spark.sql("""  
        SELECT  
            (  
                1/Float(COUNT(temperature))  
            ) *  
            SUM (  
                POWER((temperature - %s), 3) / POWER(%s, 3)  
            ) AS skwtemp FROM washing""")  
            %(meanTemperature(df,spark),sdTemperature(df,spark  
        ))).first().skwtemp
```

Kurtosis is the 4th statistical moment, so if you are smart you can make use of the code for skew which is the 3rd statistical moment. Actually only two things are different.

```
In [7]: def kurtosisTemperature(df,spark):  
        return spark.sql("""  
        SELECT  
            SUM (  
                POWER((temperature - %s), 4) / POWER(%s, 4)  
            ) /  
            (  
                Float(COUNT(temperature))  
            )  
            AS krttemp FROM washing""")  
            %(meanTemperature(df,spark),sdTemperature(df,spark  
        ))).first().krttemp
```

Just a hint. This can be solved easily using SQL as well, but as shown in the lecture also using RDDs.

```
In [19]: def meanHardness(df,spark):  
        return spark.sql("SELECT AVG(hardness) as avghard from washing").fi  
rst().avghard
```

```
def sdHardness(df,spark):
    return spark.sql("SELECT STDDEV(hardness) as sdhard from washing").
first().sdhard
def correlationTemperatureHardness(df,spark):
    return spark.sql("""
SELECT
    (
        SUM((temperature-%s) * (hardness-%s)) / Float(COUNT(temperatur
e))
    ) /
    (
        %s * %s
    )
AS coritemphrd FROM washing
""").first().coritemphrd
```

PLEASE DON'T REMOVE THIS BLOCK - THE FOLLOWING CODE IS NOT GRADED

axx

PLEASE DON'T REMOVE THIS BLOCK - THE FOLLOWING CODE IS NOT GRADED

Now it is time to connect to the cloudant database. Please have a look at the Video "Overview of end-to-end scenario" of Week 2 starting from 6:40 in order to learn how to obtain the credentials for the database. Please paste this credentials as strings into the below code

TODO Please provide your Cloudant credentials here

[illegible]

```
In [10]: spark = SparkSession\
        .builder\
        .appName("Cloudata Spark SQL Example in Python using temp tables")\
        .config("cloudant.host",hostname)\
        .config("cloudant.username", user)\
        .config("cloudant.password",pw)\
        .getOrCreate()\
        cloudantdata=readDataFrameFromCloudant(database)
```

```

+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
|           _id|           _rev|count|flowrate|fluidlevel|fr
equency|hardness|speed|temperature|           ts|voltage|
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
|28e0a6047fbb61bc1...|1-0516b5694f9cde5...|    1|    11|acceptable|
null|    72| null|    96|1530406897515|    null|
|28e0a6047fbb61bc1...|1-0516b5694f9cde5...|    1|    11|acceptable|
null|    72| null|    96|1530406897515|    null|

```

28e0a6047fbb61bc1... 1-e0a4aer254349et... 1	null	null
null null 1056 null 1530406901519	null	
28e0a6047fbb61bc1... 1-4b63d38095a17e4... 9	11	acceptable
null 76 null 99 1530406905525	null	
28e0a6047fbb61bc1... 1-6306e41a8743c5f... 2	null	null
null null 1033 null 1530406906520	null	
28e0a6047fbb61bc1... 1-fdfed10e51e00a5... 13	11	acceptable
null 76 null 86 1530406909534	null	
28e0a6047fbb61bc1... 1-6daa45c1e6e8c47... 3	null	null
null null 1031 null 1530406911522	null	
28e0a6047fbb61bc1... 1-0abe6486f53bb03... 17	11	acceptable
null 74 null 96 1530406913540	null	
28e0a6047fbb61bc1... 1-d71c991a062ff5e... 24	11	acceptable
null 79 null 89 1530406920547	null	
28e0a6047fbb61bc1... 1-fd2035da60487ea... 35	11	acceptable
null 70 null 97 1530406931572	null	
28e0a6047fbb61bc1... 1-4d616ab1c995cf9... 54	11	acceptable
null 70 null 90 1530406950615	null	
28e0a6047fbb61bc1... 1-988eb68f432da53... 64	11	acceptable
null 71 null 92 1530406960633	null	
28e0a6047fbb61bc1... 1-d7bd3b8754dd584... 70	11	acceptable
null 74 null 95 1530406966643	null	
28e0a6047fbb61bc1... 1-a0044025cea1578... 75	11	acceptable
null 80 null 87 1530406971649	null	
28e0a6047fbb61bc1... 1-09e963537e45973... 76	11	acceptable
null 73 null 92 1530406972652	null	
28e0a6047fbb61bc1... 1-c72148411906d3f... 84	11	acceptable
null 77 null 84 1530406980664	null	
28e0a6047fbb61bc1... 1-79943ce50a55c94... 95	11	acceptable
null 72 null 100 1530406991684	null	
28e0a6047fbb61bc1... 1-76437369291ad0e... 123	11	acceptable
null 110 null 83 1530407019738	null	
28e0a6047fbb61bc1... 1-c60795fbbbd2cbd... 149	11	acceptable
null 79 null 85 1530407045789	null	
28e0a6047fbb61bc1... 1-f126164c620852a... 156	11	acceptable
null 78 null 85 1530407052804	null	
28e0a6047fbb61bc1... 1-69b968fbc8c9945... 158	11	acceptable
null 72 null 94 1530407054806	null	

only showing top 20 rows

```
In [11]: minTemperature(cloudantdata,spark)
```

Out[11]: 80

```
In [12]: meanTemperature(cloudantdata,spark)
```

```
Out[12]: 90.20837209302326
```