CSC 643 – Big Data & Web Intelligence Team1.cache(Ahmad, Sean) Project 4 Submission Date: December 11, 2019

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TASK RANKING

> Member 1: Ahmad Alshawaf

Handled 50% of the project from start to complete

- Task 1: analyzing data
- Task 2: solving part 1
- Task 3: solving part 2
- Task 4: solving part 3
- Task 5: solving part 4
- Task 6: organizing report

➤ Member 2: Sean Sothey

Handled 50% of the project from start to complete

- Task 1: analyzing data
- Task 2: solving part 1
- Task 3: solving part 2
- Task 4: solving part 3
- Task 5: solving part 4
- Task 6: organizing report

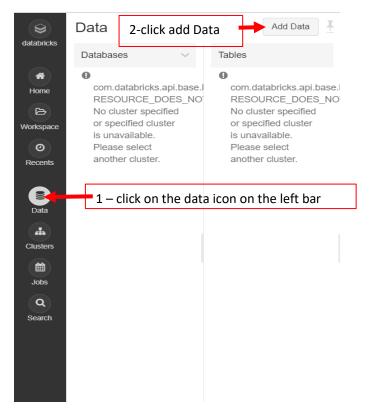
INTRODUCTION

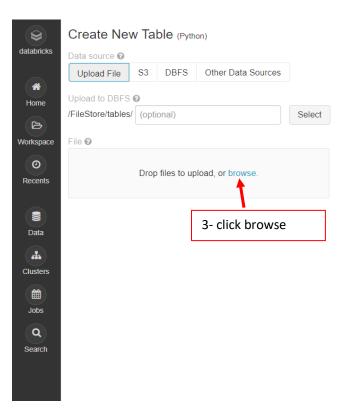
Data science has become an extremely rewarding career choice for people interested in extracting, manipulating, and generating insights out of large volumes of data. To fully leverage the power of data science, scientists often need to obtain skills in databases, statistical programming tools, and data visualizations. Companies surely need data scientists to help them empower their analytics processes, build a numbers-based strategy that will boost their bottom line, and ensure that enormous amounts of data are translated into actionable insights. In this project, we would like to introduce one of the most popular platform tools to analyze the data. Databricks is one of the most powerful cloud base computations, managed platform for running Apache Spark - that means that you do not have to learn complex cluster management concepts nor perform tedious maintenance tasks to take advantage of Spark. Databricks also provides a host of features to help its users be more productive with Spark. It's a point and click platform for those that prefer a user interface like data scientists or data analysts. However, this UI is accompanied by a sophisticated API for those that want to automate aspects of their data workloads with automated jobs. To meet the needs of enterprises, Databricks also includes features such as role-based access control and other intelligent optimizations that not only improve usability for users but also reduce costs and complexity for administrators.

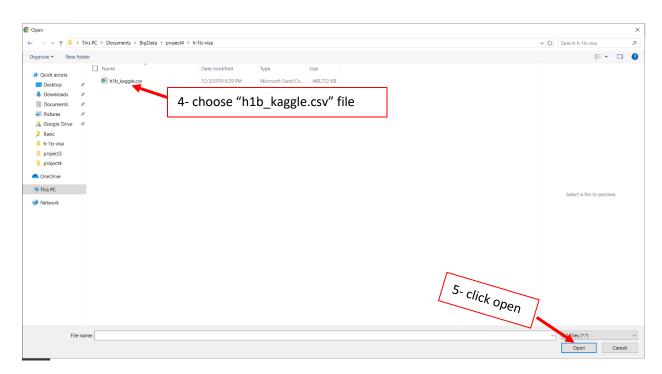
In this project, we will explore our process for discovering on numbers of petition visas base on random cases. It is worth mentioning that this notebook (in Databricks) is intended to be an academically usage, but simply a good example of the work that, a data scientist might be performing using Apache Spark and Databricks.

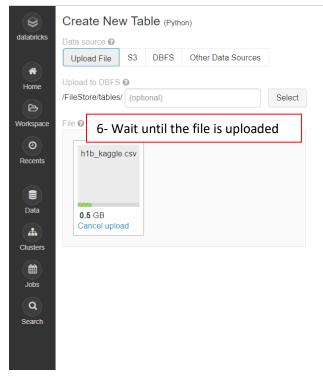
STARTER KITS

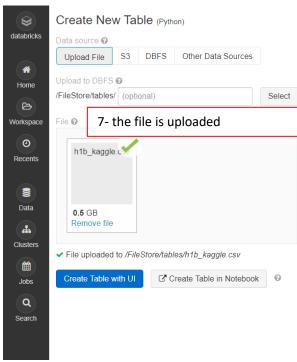
1. Loading data into Databricks



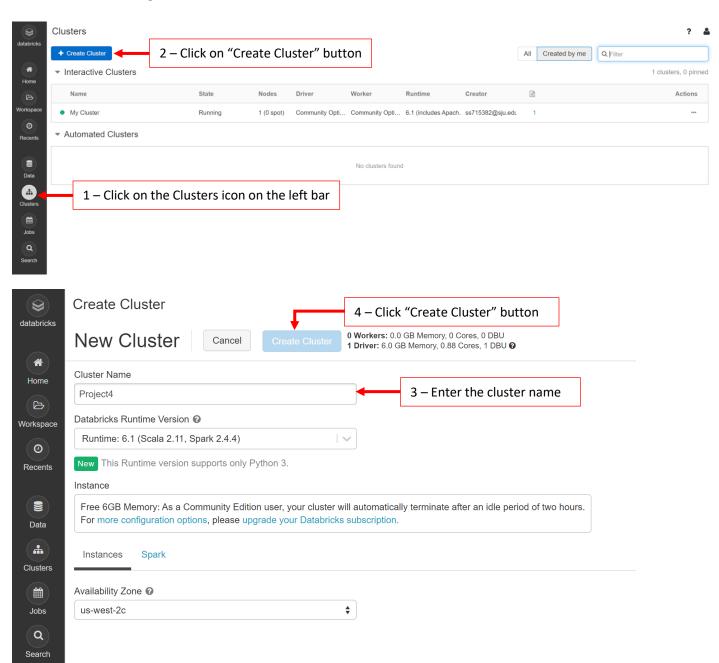




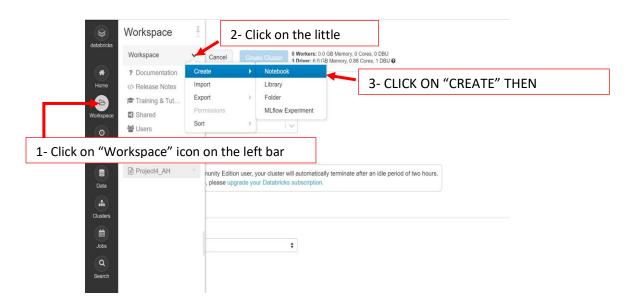


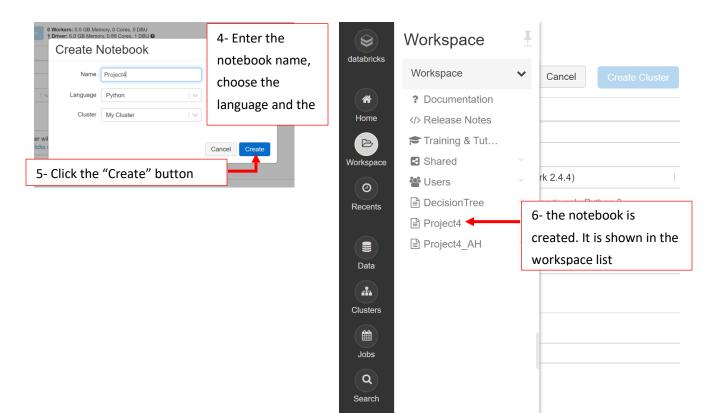


2. Creating cluster in Databricks



3. Creating notebook in Databricks





ANSWERS:

First, create RDD (Resilient Distributed Datasets) to run on Disk only, as following:

```
from pyspark import SparkContext
import pyspark

h1bDF = spark.read.csv('/FileStore/tables/h1b_kaggle.csv',
header=True,
inferSchema=True)

h1bDF.persist(pyspark.StorageLevel.DISK_ONLY)
```

Part1: Use Apache Spark Core (Transformations and Action on RDD) to answer the following queries?

1. What is the total number of rows in this dataset?

```
h1bDF.count()
Out[3]: 3002458
```

2. How many different employers filed for H-1B visa?

```
employers = h1bDF.select('EMPLOYER_NAME').distinct()
employers.count()
Out[4]: 236015
```

3. Show 10 different employers ordered by employer name in descending order.

```
employers.orderBy(['EMPLOYER NAME'], descending=True).show(10)
 ----+
       EMPLOYER_NAME
+----+
|"""EXCELLENT COMP...|
|"""I HAVE A DREAM...|
  """K"" LINE AMERICA
 """K"" LINE AMERICA"
 """K"" LINE LOGIS...|
 """K"" LINE LOGIS...
|"""K"" LINE LOGIS...|
 "CREPE ""N"" TEARIA
|"CRISP MEDIA, INC...|
|"E-2I ""EVOLUTION...|
+----+
only showing top 10 rows
```

Part2: Use Apache Spark Core (Transformations and Action on RDD) to answer the following queries?

1. How many visa petitions are there for each 'case status' (e.g., 1000 withdrawn, 100 denied)? Order results by the number of visa petitions (i.e., count) in ascending order. Note, ascending order is the default, thus, there is no need to specify that.

2. Repeat the above query to perform in-memory processing. Databricks will display execution time for each 'cell', take a screenshot of the cell including the running times for both in-memory and on-disk.

```
h1bDF.cache()
```

#What is the total number of rows in this dataset?

```
h1bDF.count()
Out[8]: 3002458
```

```
#Part1 Use Apache Spark Core (Transformations and Action on RDD) to answer the following queries?

#1 What is the total number of rows in this dataset?

hlbDF.count()
```

▶ (1) Spark Jobs Out[18]: 3002458

Command took 0.28 seconds -- by ss715382@sju.edu at 12/10/2019, 1:28:22 PM on My Cluster

Figure 1: Part 1-1 without cache

```
#What is the total number of rows in this dataset?
hlbDF.count()
```

▶ (1) Spark Jobs

Out[19]: 3002458

Command took 0.12 seconds -- by ss715382@sju.edu at 12/10/2019, 1:28:50 PM on My Cluster

Figure 2: Part 1-1 with cache

#How many different employers filed for H-1B visa?

```
employers = h1bDF.select('EMPLOYER_NAME').distinct()
employers.count()
Out[9]: 236015
```

```
#2 How many different employers filed for H-1B visa?
employers = h1bDF.select('EMPLOYER_NAME').distinct()
employers.count()

(1) Spark Jobs
```

▶ ■ employers: pyspark.sql.dataframe.DataFrame = [EMPLOYER NAME: string]

Out[3]: 236015

Command took 7.63 seconds -- by ss715382@sju.edu at 12/10/2019, 1:19:22 PM on My Cluster

Figure 4: Part 1-2 without cache

```
#How many different employers filed for H-1B visa?
employers = h1bDF.select('EMPLOYER_NAME').distinct()
employers.count()
```

▶ (1) Spark Jobs

employers: pyspark.sql.dataframe.DataFrame = [EMPLOYER_NAME: string]

Out[8]: 236015

Command took 2.76 seconds -- by ss715382@sju.edu at 12/10/2019, 1:19:23 PM on My Cluster

Figure 3: Part 1-2 with cache

#Show 10 different employers ordered by employer name in descending order.

```
employers.orderBy(['EMPLOYER_NAME'], descending=True).show(10)
```

```
1 #3 Show 10 different employers ordered by employer name in descending order.
employers.orderBy(['EMPLOYER_NAME'], descending=True).show(10)
▶ (1) Spark Jobs
     EMPLOYER_NAME
+----+
|"""EXCELLENT COMP...|
|"""I HAVE A DREAM...|
| """K"" LINE AMERICA|
|"""K"" LINE AMERICA"|
|"""K"" LINE LOGIS...|
|"""K"" LINE LOGIS...|
|"""K"" LINE LOGIS...|
| "CREPE ""N"" TEARIA|
|"CRISP MEDIA, INC...|
|"E-2I ""EVOLUTION...|
+----
only showing top 10 rows
Command took 6.16 seconds -- by ss715382@sju.edu at 12/10/2019, 1:19:23 PM on My Cluster
```

Figure 5: Part 1-3 without cache

```
#Show 10 different employers ordered by employer name in descending order.
employers.orderBy(['EMPLOYER_NAME'], descending=True).show(10)
```

▶ (1) Spark Jobs

```
Command took 2.44 seconds -- by ss715382@sju.edu at 12/10/2019, 1:19:23 PM on My Cluster
```

Figure 6: Part 1-3 with cache

#How many visa petitions are there for each 'case status' (e.g., 1000 withdrawn, 100 denied)? Order results by the number of visa petitions (i.e., count) in ascending order. Note, ascending order is the default, thus, there is no need to specify that.

```
h1bDF.groupBy('CASE_STATUS').count().orderBy('count').show()
```

```
#Part2 Use Apache Spark Core (Transformations and Action on RDD) to answer the following queries?

#1 How many visa petitions are there for each 'case status' (e.g., 1000 withdrawn, 100 denied)? Order order. Note, ascending order is the default, thus, there is no need to specify that.

hlbDF.groupBy('CASE_STATUS').count().orderBy('count').show()
```

```
▶ (1) Spark Jobs
```

Command took 5.31 seconds -- by ss715382@sju.edu at 12/10/2019, 1:19:23 PM on My Cluster

Figure 7: Part 2-1 without cache

```
#Part2 Use Apache Spark Core (Transformations and Action on RDD) to answer the following queries?
#How many visa petitions are there for each 'case status' (e.g., 1000 withdrawn, 100 denied)? Order order. Note, ascending order is the default, thus, there is no need to specify that.

hlbDF.groupBy('CASE_STATUS').count().orderBy('count').show()
```

▶ (1) Spark Jobs

Command took 1.96 seconds -- by ss715382@sju.edu at 12/10/2019, 1:19:23 PM on My Cluster

Figure 8:Part 2-1 with cache

	Command	With cache	Without cache
Part 1 - 1	h1bDF.count()	0.12	0.28
Part 1 - 2	employers = h1bDF.select('EMPLOYER_NAME').distinct() employers.count()	2.76	7.63
part 1 - 3	employers.orderBy(['EMPLOYER_NAME'], descending=True).show(10)	2.44	6.16
part 2 - 1	h1bDF.groupBy('CASE_STATUS').count().orderBy('count').show()	1.96	5.31

Table 1: time compression between in-disk and in-memory processing

2. Use Python, R, or Scala to create a new data-frame (RDD) that contains employer names only. Then, iterate (i.e., loop) through the newly created data-frame to print all rows of data/employer names, and count/print number of rows. Hint, a transformation returns a new data-frame. Be patient, this might take a while.

```
employersDF = h1bDF.select('EMPLOYER_NAME')
employersDF.count()
Out[91]: 3002458
```

```
for row in employersDF.collect():
print(row)
 Row(EMPLOYER_NAME='UNIVERSITY OF MICHIGAN')
 Row(EMPLOYER_NAME='GOODMAN NETWORKS, INC.')
 Row(EMPLOYER_NAME='PORTS AMERICA GROUP, INC.')
 Row(EMPLOYER_NAME='GATES CORPORATION, A WHOLLY-OWNED SUBSIDIARY OF TOMKINS
 PLC')
 Row(EMPLOYER_NAME='PEABODY INVESTMENTS CORP.')
 Row(EMPLOYER_NAME='BURGER KING CORPORATION')
 Row(EMPLOYER NAME='BT AND MK ENERGY AND COMMODITIES')
 Row(EMPLOYER_NAME='GLOBO MOBILE TECHNOLOGIES, INC.')
 Row(EMPLOYER_NAME='ESI COMPANIES INC.')
 Row(EMPLOYER NAME='LESSARD INTERNATIONAL LLC')
 Row(EMPLOYER_NAME='H.J. HEINZ COMPANY')
 Row(EMPLOYER_NAME='DOW CORNING CORPORATION')
 Row(EMPLOYER_NAME='ACUSHNET COMPANY')
 Row(EMPLOYER_NAME='BIOCAIR, INC.')
 Row(EMPLOYER_NAME='NEWMONT MINING CORPORATION')
 Row(EMPLOYER_NAME='VRICON, INC.')
 Row(EMPLOYER_NAME='CARDIAC SCIENCE CORPORATION')
 Row(EMPLOYER_NAME='WESTFIELD CORPORATION')
 Row(EMPLOYER_NAME='QUICKLOGIX LLC')
 Row(EMPLOYER_NAME='MCCHRYSTAL GROUP, LLC')
```

Part3: Use Apache Spark Core and Spark SQL to answer the following queries. Hint, you might need to create a temporary view/table in order to perform SQL queries. Cache both RDDs and Views to perform in-memory processing.

```
# Create a VIEW to perform SQL queries
#done the RDD cahche in the above part
h1bDF.createOrReplaceTempView("h1bView")
# cache table
spark.catalog.cacheTable("h1bView")
```

1. How many visa petitions were denied in 2016?

```
%sql
select count(CASE_STATUS)
from h1bView
where CASE_STATUS == 'DENIED' and YEAR == '2016';
count(CASE_STATUS)
9175
```

2. Show different employers in California and Pennsylvania who have certified visa petitions in 2013. Order results by employer names in ascending order.

```
%sql
select distinct EMPLOYER_NAME
from h1bView
where WORKSITE regexp 'CALIFORNIA | PENNSYLVANIA' and CASE_STATUS ==
'CERTIFIED' and YEAR == '2013'
order by EMPLOYER_NAME;
EMPLOYER_NAME
1 WAY SOLUTIONS, INC.
1SEO.COM
24/7 MEDIA, INC.
3A SOFT INC.
3CUBE SOLUTIONS INC
31 INFOTECH, INC
3I INFOTECH, INC.
3I SOLUTIONS, INC.
3K TECHNOLOGIES LLC
```

3. Is it possible to perform SQL queries directly on RDD?

It is possible in a way that you create your own version of RDD called schemaRDD. To standard RDD functions, SchemaRDDs can be used in relational queries. At the core of this component, a new type of RDD, <u>SchemaRDD</u>, are composed <u>Row</u> objects along with a schema that describes the data types of each column in the row. A SchemaRDD is similar to a table in a traditional relational database. A SchemaRDD can be created from an existing RDD, <u>Parquet</u> file, a JSON dataset, etc.

Importing a SQLContext brings an implicit into scope that automatically converts a standard RDD whose elements are scala case classes into a SchemaRDD. A SchemaRDD can be registered as a table in the <u>SQLContext</u> that was used to create it. Once an RDD has been registered as a table, it can be used in the FROM clause of SQL statements.

Here is the example:

```
// One method for defining the schema of an RDD is to make a case class with the desired column
// names and types.
case class Record(key: Int, value: String)

val sc: SparkContext // An existing spark context.
val sqlContext = new SQLContext(sc)

// Importing the SQL context gives access to all the SQL functions and implicit conversions.
import sqlContext._

val rdd = sc.parallelize((1 to 100).map(i => Record(i, s"val_$i")))
// Any RDD containing case classes can be registered as a table. The schema of the table is
// automatically inferred using scala reflection.
rdd.registerAsTable("records")

val results: SchemaRDD = sql("SELECT * FROM records")
```

Part4: Pre-define the Schema manually (inferring the schema) in order to use Data Aggregation functions and compare numeric values (e.g., <, >, =) using Apache Spark Core. You will need to import the following libraries to pre-define/infer schema and use aggregation functions:

```
from pyspark.sql.functions import min, max, avg
from pyspark.sql.types import StructType, StructField, IntegerType,
StringType
from pyspark.sql.types import *
```

1. Since SOC_NAME column does not specify what sort of data it contains, change the column name to OCCUPATION_CODE for a coherent column name. Take a screenshot of the Schema, to verify that the column name has changed. Hint, use DF.printSchema() to view column names and data types.

```
h1bDF = h1bDF.withColumnRenamed('SOC NAME', 'OCCUPATION CODE')
#convert PREVAILING WAGE from datatype to decimal
h1bDF = h1bDF.withColumn('PREVAILING WAGE',
h1bDF.PREVAILING WAGE.cast(DecimalType(20)))
h1bDF.printSchema()
   |-- _c0: integer (nullable = true)
   |-- CASE_STATUS: string (nullable = true)
   |-- EMPLOYER_NAME: string (nullable = true)
   |-- OCCUPATION_CODE: string (nullable = true)
   |-- JOB_TITLE: string (nullable = true)
   |-- FULL_TIME_POSITION: string (nullable = true)
   |-- PREVAILING_WAGE: decimal(20,0) (nullable = true)
   |-- YEAR: string (nullable = true)
   |-- WORKSITE: string (nullable = true)
   |-- lon: string (nullable = true)
   |-- lat: string (nullable = true)
```

2. Show the lowest, highest, and average wage/salary of all filed visa petitions?

```
#filter to get only the legit value
h1bDF filter =
h1bDF.select('PREVAILING WAGE').filter(h1bDF.PREVAILING WAGE>0)
h1bDF filter.agg(max("PREVAILING WAGE")).show()
h1bDF filter.agg(min("PREVAILING WAGE")).show()
h1bDF filter.agg(avg("PREVAILING WAGE")).show()
+----+
|max(PREVAILING_WAGE)|
         6997606720
|min(PREVAILING_WAGE)|
                15|
|avg(PREVAILING_WAGE)|
        147000.3827
  _____+
```

3. Print an appropriate message if there are wages that are greater than 80,000. If yes, print how many, else, print "none".

```
from pyspark.sql.functions import col, when
h1bDF_filter.withColumn('wage_grt80000', when(col('PREVAILING_WAGE') >
80000, "Yes").otherwise("No")).show()
h1bDF_filter.where(h1bDF_filter['PREVAILING_WAGE']>80000).count()
|PREVAILING_WAGE|wage_grt80000|
        187200|
                       Yes
       241842
                      Yes
        99986
                     Yes
        99986
        187200
                     Yes
                      Yes
        215862
        192088
                       Yes
         95296
        149594
                       Yes
        226699
                       Yes
        187200
                       Yes
        159370
                       Yes
         98550
                       Yes
        130853
                       Yes
         52416
                       No
        130853
                       Yes
         89107
                       Yes
        130853
                       Yes
        102190|
                       Yes
        197683
                       Yes
only showing top 20 rows
```

alternative solution:

```
count = 0
for x in hlbDF_filter.collect():
    if x.PREVAILING_WAGE > 80000:
        count+=1
    if count>0:
        print("Yes, there are wages greater than $80,000 : " + str(count))
    else:
        print("None")
        * (1) Spark Jobs
        Yes, there are wages greater than $80,000 : 789875
```

REFERENCES:

• Data provider:

https://www.kaggle.com/nsharan/h-1b-visa

• Databricks Documentation:

https://docs.databricks.com/

• SchemaRDD:

https://spark.apache.org/docs/1.0.2/api/scala/index.html#org.apache.spark.sql.SchemaRDD

• PySpark - StorageLevel:

https://www.tutorialspoint.com/pyspark/pyspark_storagelevel.htm

• CSC643 class's material:

 $\underline{http://people.sju.edu/^bforoura/tutorials/databricks/introduction.htm}$

• Notebook in HTML:

file:///C:/Users/seans/Downloads/Project4.html