→ Airlines project in PySpark

[bigdatalab456422@ip-10-1-1-204 ~]\$ ls -l airlines.csv

[bigdatalab456422@ip-10-1-1-204 ~]\$ ls -l airlines.csv -rw-rw-r-- 1 bigdatalab456422 bigdatalab456422 1821 Jun 5 09:24 airlines.csv

[bigdatalab456422@ip-10-1-1-204 \sim]\$ hadoop fs -put airlines.csv training

[bigdatalab4564 2 2@ip-10-1-1-204 $^{\sim}$]\$ hadoop fs -put airlines.csv training [bigdatalab456422@ip-10-1-1-204 $^{\sim}$]\$



[biqdatalab456422@ip-10-1-1-204 ~]\$ pyspark

→ find the year with highest revenue

```
>>> from pyspark.sql.types import StructType, StringType,
IntegerType, DoubleType, LongType
>>> from pyspark.sql.types import StructType, StringType, IntegerType, DoubleType, LongType
>>> schema2 =
StructType().add("Year", StringType(), True).add("Quarter", StringType(), True).add("Booked_seats", IntegerType(), True)
..., True)
>>> schema2 = StructType().add("Year", StringType(), True).add("ARPS", DoubleType(), True).add("Booked_seats", IntegerType(), True)
>>> schema2 = StructType().add("Year", StringType(), True).add("ARPS", DoubleType(), True).add("Booked_seats", IntegerType(), True)
```

```
>>> print(schema2)
StructType(List(StructField(Year,StringType,true),StructField(Quarter,StringType,true),StructField(ARPS,DoubleType,true),StructField(Booked_seats,IntegerType,true))))
>>> df with schema2 = spark.read.format("csv").option("header",
"True").schema(schema2).load("hdfs://nameservice1/user/bigdatalab4564
22/training/airlines.csv")
>>> df_with_schema2 = spark.read.format("csv").option("header", "True").schema(schema2).load("hdfs://nameservice1/user/bigdatalab456422/training/airlines.csv")
                                                                                                                                                       >>> df with schema2.count()
>>> df_with_schema2.count() 84 >>> ■
>>> df with schema2.show()
>>> df_with_schema2.show()
|Year|Quarter| ARPS|Booked_seats|
          1 296.9 2 296.8 3 287.51 4 287.78 1 283.97 2 275.78 3 269.49 4 278.33 4
| 1995 | 1995 | 1995 | 1996 | 1996 | 1996 | 1996 | 1996 | 1997
                         34128
30388
47808
43020
                         38952
37443
           1 283.4 2 289.44
 1997
1997
                         35067
46565
          2 | 289.44
3 | 282.27
4 | 293.51
1 | 304.74
2 | 300.97
3 | 315.25
4 | 316.18
1 | 331.74
2 | 329.34
3 | 317.22
 | 1997 |
| 1997 |
| 1998 |
| 1998 |
| 1998 |
| 1999 |
| 1999 |
                         38886
                         37454
                         37454
31315
30852
38118
35393
47453
38243
|1999|
|1999|
           3 | 317.22 |
4 | 317.93 |
only showing top 20 rows
>>>
>>> df with schema2.registerTempTable("airlines")
>>> df_with_schema2.registerTempTable("airlines")
>>>
>>> YrWiseRev= spark.sql("SELECT Year, sum(ARPS*Booked seats) AS
revenue FROM airlines GROUP BY Year ORDER BY revenue DESC")
>>> YrWiseRev= spark.sql("SELECT Year, sum(ARPS*Booked_seats) AS revenue FROM airlines GROUP BY Year ORDER BY revenue DESC")
>>> YrWiseRev.count()
>>> YrWiseRev.count()
>>> YrWiseRev.show(21)
```

>>> print(schema2)

```
>>> YrWiseRev.show(21)
                          6.636320871E7
6.262417585000001E7
6.237899057E7
6.219912728E7
     2014
     2012
     2008 5.7653170760000005E7
     2007 5.730921607E7
2001 5.553377999999999E7
    |2010| 5.486152129E7
|2000| 5.2342926550000004E7
   | 2000 | 5.2342926550000004E7 | 2011 | 5.188228622E7 | 2004 | 5.0631364949999996E7 | 2006 | 5.0437898419999994E7 | 2003 | 4.9273210891 | 2005 | 4.79791465E7 | 2002 | 4.7991465E7 | 2009 | 4.674644659E7 | 2005 | 4.637678624E7 | 2005 | 4.637678624E7 | 2006 | 4.63767862E7 | 2006 | 4.63767862E7 | 2006 | 4.63767862E7 | 2006 | 4.63767862E7 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 | 2006 
     1996
1997
                                                4.635877803E7
                                                4.538523616E7
     1995
    1998
                                                4.203571778F7
   >>>
>>> YrWiseRev= spark.sql("SELECT Year, sum(ARPS*Booked seats)/1000000
AS revenue_in_mill FROM airlines GROUP BY Year ORDER BY
 revenue_in mill DESC")
 >>> YrWiseRev= spark.sql("SELECT Year, sum(ARPS*Booked_seats)/1000000 AS revenue_in_mill FROM airlines GROUP BY Year ORDER BY revenue_in_mill DESC")
>>>
>>> YrWiseRev.count()
   >>> YrWiseRev.count()
>>> YrWiseRev.show(21)
   >>> YrWiseRev.show(21)
                                                             revenue
                        66.36320871
62.62417585000001
62.37899057
62.19912728
57.65317076
   2013
    |2007
|2007
|2001
|2010
|2000
                          57.30921607
55.5337799999999
                                                54.86152129
52.34292655
     2011
                                                 51.88828622
50.63136495
   | 2004 | 50.63156495 | 2006 | 50.437898419999996 | 2003 | 49.27321083 | 1999 | 48.75771448 | 2002 | 47.4991465 | 2009 | 46.746446590000005 | 2005 | 46.35877803 | 1997 | 45.38523616 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.4940332 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.494032 | 47.49
     1995
    1998
                                                42.03571778
>>> YrWiseRev= spark.sql("SELECT Year,
round(sum(ARPS*Booked seats)/1000000, 2) AS revenue in mill FROM
airlines GROUP BY Year ORDER BY revenue in mill DESC")
```

>>> YrWiseRev= spark.sql("SELECT Year, round(sum(ARPS*Booked_seats)/1000000, 2) AS revenue_in_mill FROM airlines GROUP BY Year ORDER BY revenue_in_mill DESC")
>>>

>>> YrWiseRev.show(21)

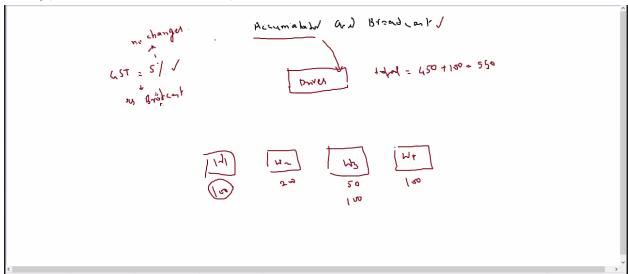
>>>

→ find the year with highest number of PAX flown

```
>>> YrWisePax = spark.sql("SELECT Year, sum(Booked_seats) AS
total_pax FROM airlines GROUP BY Year ORDER BY total_pax DESC")
>>> YrWisePax = spark.sql("SELECT Year, sum(Booked_seats) AS total_pax FROM airlines GROUP BY Year ORDER BY total_pax DESC")
>>> YrWisePax.count()
>>> YrWisePax.count()
```

>>>

→ Types of variables in Spark



a. Accumulators

- i. Are at driver/spark app
- ii. Value can change as it sums up the values at each of the worker nodes
- iii. Active throughout entire session
- iv. Need to re-initialize it to zero to reset accumulator

b. Broadcast variables

- i. Are like constant/immutable variable
- ii. Can be used in any program, in any stage
- iii. Declared at driver node, Can be used from any of the worker nodes

\rightarrow Note for port number

- a. check core-site.xml
- b. hdfs://localhost:54310

→ Retail Project for accumulator & broadcast variables

ightarrow write a program using shared variables like broadcast and accumulators to calculate total tax and total profit using Retail data

[bigdatalab456422@ip-10-1-1-204 ~]\$ hadoop fs -ls

```
[bigdatalab456422@ip-10-1-1-204 ~]$ hadoop fs -ls Found 10 items drwx----- - bigdatalab456422 bigdatalab456422
                                        0 2023-05-31 11:00 .Trash
0 2023-06-05 09:28 .sparkStaging
0 2023-05-30 11:35 .staging
40990862 2023-06-02 10:33 NYSE.csv
0 2023-05-19 11:52 data
0 2023-05-19 07:12 hive
50 2023-05-16 12:43 newfile.txt
0 2023-05-30 10:43 sales
0 2023-05-20 10:02 student
0 2023-05-20 10:02 student
0 2023-05-05 09:25 training
[bigdatalab456422@ip-10-1-1-204 ~]$
[bigdatalab456422@ip-10-1-1-204 ~]$ hadoop fs -mkdir retail
[bigdatalab456422@ip-10-1-1-204 \sim]$ hadoop fs -mkdir retail [bigdatalab456422@ip-10-1-1-204 \sim]$
[bigdatalab456422@ip-10-1-1-204 ~]$ hadoop fs -ls retail
[bigdatalab456422@ip-10-1-1-204 ~]$ hadoop fs -ls retail [bigdatalab456422@ip-10-1-1-204 ~]$
[bigdatalab456422@ip-10-1-1-204 ~]$ hadoop fs -put D11 D12 D01 D02
[bigdatalab456422@ip-10-1-1-204 ~]$ hadoop fs -put D11 D12 D01 D02 retail [bigdatalab456422@ip-10-1-1-204 ~]$
>>> retailRDD =
sc.textFile("hdfs://nameservice1/user/bigdatalab456422/retail")
>>> retailRDD = sc.textFile("hdfs://nameservice1/user/bigdatalab456422/retail")
>>> retailRDD.count()
>>> retailRDD.count()
>>>
>>> retailRDD.getNumPartitions()
>>> retailRDD.getNumPartitions()
>>> gst = sc.broadcast(5.00)
>>> gst = sc.broadcast(5.00)
>>>
>>> totalTax = sc.accumulator(0.00)
>>> totalTax = sc.accumulator(0.00)
>>>
>>> totalProfit = sc.accumulator(0.00)
>>> totalProfit = sc.accumulator(0.00)
>>> arrayRDD = retailRDD.map(lambda a : a.split(";"))
>>> arrayRDD = retailRDD.map(lambda a : a.split(";"))
>>> taxAndProfit = arrayRDD.map(lambda a : (
float(a[8])*qst.value/100, (float(a[8]) - float(a[7]))))
>>> taxAndProfit = arrayRDD.map(lambda a : ( float(a[8])*gst.value/100 , (float(a[8]) - float(a[7]))))
>>>
```

```
>>> for a in taxAndProfit.take(5):
            print(a)
>>> for a in taxAndProfit.take(5):
... print(a)
(2.6, 8.0)
(6.45, -21.0)
(1.95, 4.0)
(5.95, 25.0)
(7.95, 59.0)
>>>
>>> for line in taxAndProfit.collect():
              totalTax += line[0]
>>> for line in taxAndProfit.collect():
... totalTax += line[0]
>>> print(totalTax)
>>> print(totalTax) 5392053.600018393
>>> for line in taxAndProfit.collect():
              totalProfit += line[1]
. . .
>>> for line in taxAndProfit.collect():
... totalProfit += line[1]
>>> print(totalProfit)
>>> print(totalProfit)
16163257.0
# does not work as totalProfit is of type Accumulator, so it needs to accessed using value
>>> print(totalProfit/1000000)
>>> print(totalProfit/1000000)
Traceback (most recent call last):
File "cstdin", line 1, in <module>
TypeError: unsupported operand type(s) for /: 'Accumulator' and 'int'
>>> print(totalProfit.value/1000000)
>>> print(totalProfit.value/1000000)
# find net profit after deducting tax
>>> print(totalProfit.value - totalTax.value)
>>> print(totalProfit.value - totalTax.value) 10771203.399981607
>>> print(totalProfit)
>>> print(totalProfit)
16163257.0
>>> totalProfit = sc.accumulator(0.0)
>>> totalProfit = sc.accumulator(0.0)
```

```
>>> print(totalProfit)
>>> print(totalProfit)
0.0
```

```
The Cast Format Weve Help

File Cast Format Veve Help

write a program using shared variables like broadcast and accumulators to calculate total tax and total profit

using Retail data

retailRDD = sc.textFile("hdfs://nameservice1/user/bigdatalab45644/retail")

retailRDD.count()

retailRDD.getNumPartitions()

gst = sc.broadcast(5.00)

totalTax = sc.accumulator(0.00)

totalProfit = sc.accumulator(0.00)

arrayRDD = retailRDD.map(lambda a : a.split(";"))

taxAndProfit = arrayRDD.map(lambda a : (float(a[8])*gst.value/100 , (float(a[8]) - float(a[7]))))

for line in taxAndProfit.collect():

... totalTax += line[0]

...

print(totalTax)

for line in taxAndProfit.collect():

... totalProfit += line[1]

...
```

Previous exam pattern

1. MapReduce Problem 1 no

- 2. Hive Queries 5 no
- 3. Spark Queries 5 no