Lab Assignment 2

TEAM ID : 10

Team Members :

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Part 1

Aim :

Implement MapReduce algorithm for finding Facebook common friends problem and run the MapReduce job on Apache Spark.

```
def mapper(theString):
    theString = theString.split(" ")
    user = theString[0]
    friends = theString[1]
    keyvalues = []

    for char in friends:
        keyvalues.append((''.join(sorted(user+char)), friends.replace(char, "")))

    return keyvalues


def reducer(a, b):
    newString = ''
    for char in a:
        if char in b:
            newString += char
    return newString
```

I have used two main functions :

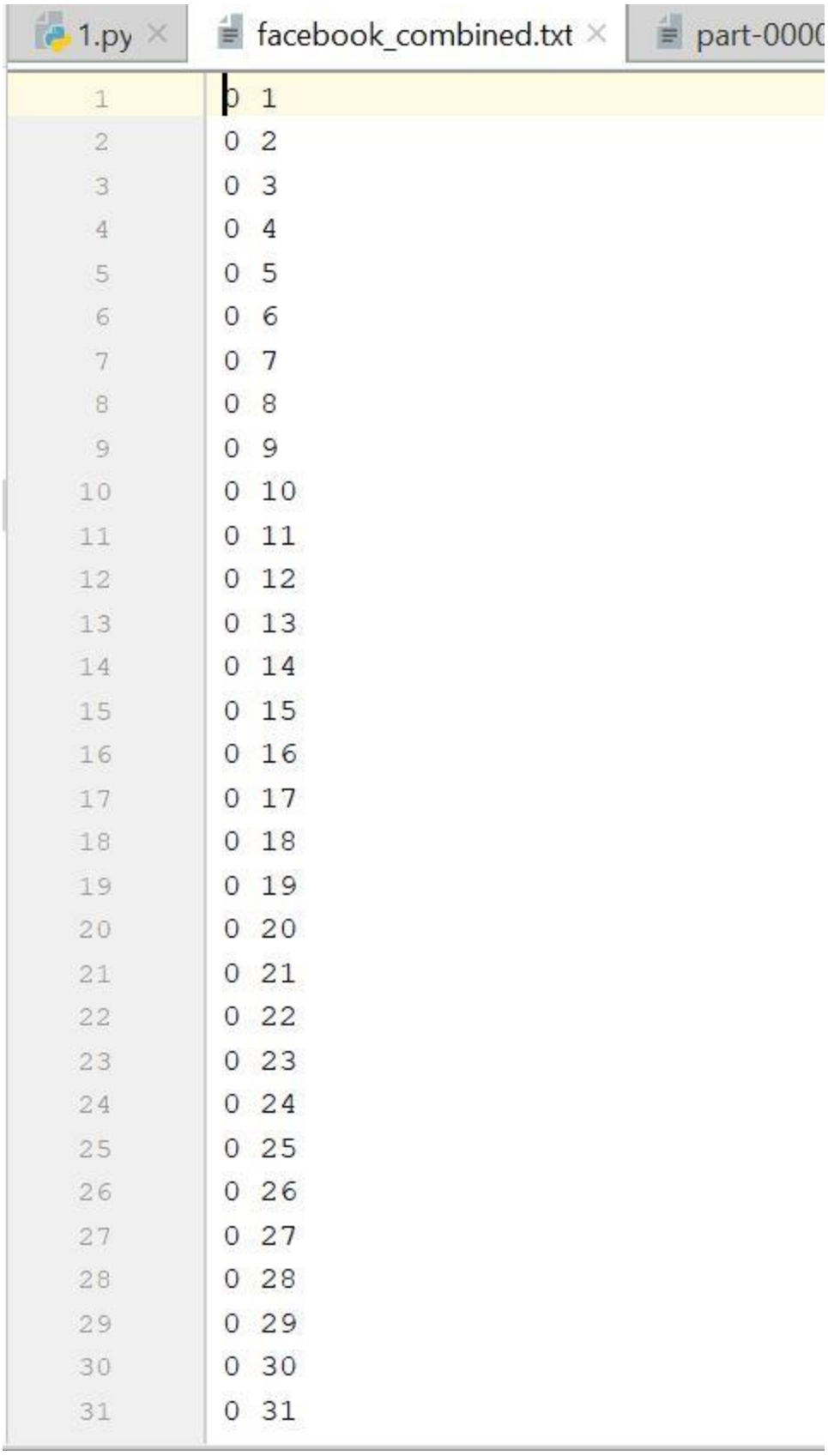
- One for the dataset given.
- Other for the simple dataset.

```
if __name__ == "__main__":
    mew = SparkContext.getOrCreate()
    lines = mew.textFile("input.txt", 1)
    newLines = lines.flatMap(mapper)
    newLines.saveAsTextFile("mapper")
    friends = newLines.reduceByKey(reducer)
    friends.coalesce(1).saveAsTextFile("reducer")
    mew.stop()

if __name__ == "__main__":
    facebook = SparkContext.getOrCreate()
    facebooklines = facebook.textFile("facebook_combined.txt", 1)
    facebookNewLines = facebooklines.flatMap(mapper)
    facebookNewLines.saveAsTextFile("facebookmapper")
    facebookfriends = facebookNewLines.reduceByKey(reducer)
    facebookfriends.coalesce(1).saveAsTextFile("facebookreducer")
    facebook.stop()
```

Input :

- Using the given dataset.


1.py
facebook_combined.txt
part-0000
0 1
0 2
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0 11
0 12
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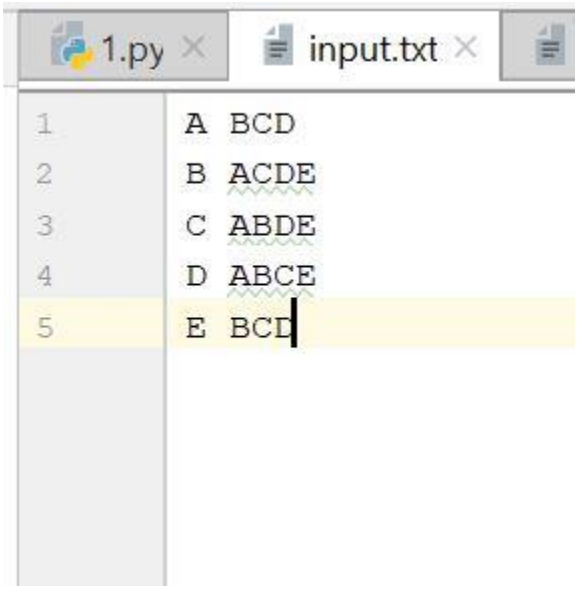
0 31

- Using the simple dataset.



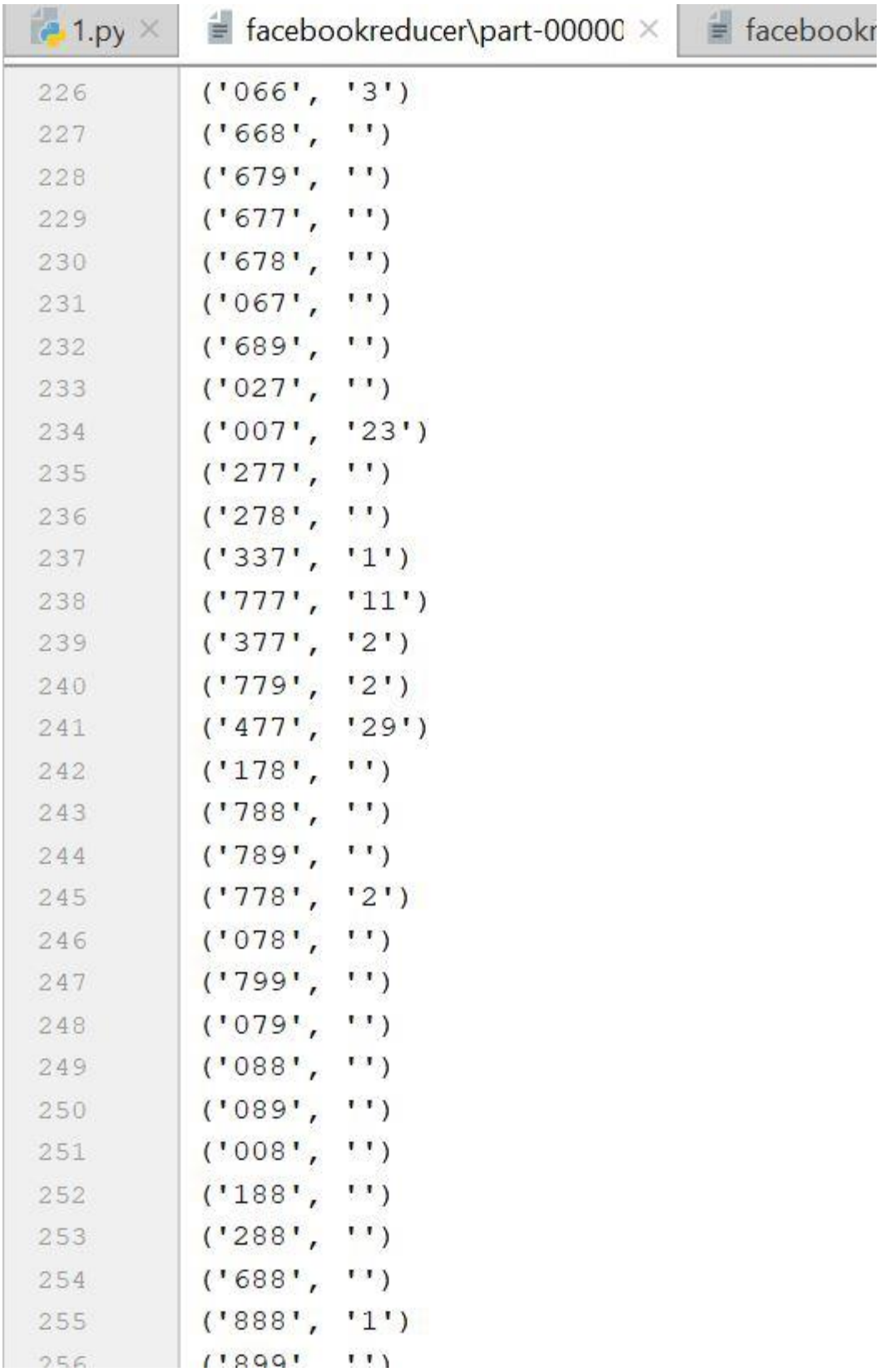
Output :

- For the given dataset.

1.py | facebookmapper\part-00000 | input.txt | faceb

The file size (5.38 MB) exceeds configured limit (2.44 MB). Code insight feature

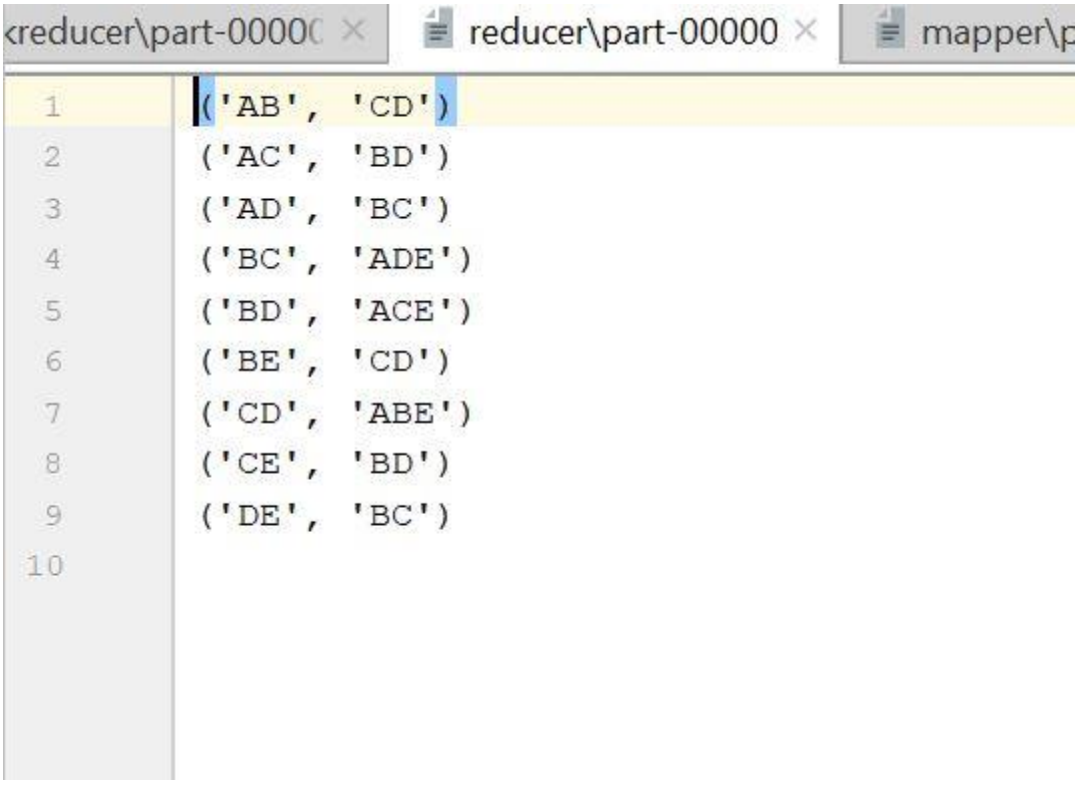
```
448  ('01', '86')
449  ('08', '16')
450  ('06', '18')
451  ('01', '87')
452  ('08', '17')
453  ('07', '18')
454  ('01', '88')
455  ('08', '1')
456  ('08', '1')
457  ('01', '89')
458  ('08', '19')
459  ('09', '18')
460  ('01', '90')
461  ('09', '10')
462  ('00', '19')
463  ('01', '9')
464  ('09', '11')
465  ('01', '9')
466  ('01', '92')
467  ('09', '12')
468  ('02', '19')
469  ('01', '93')
470  ('09', '13')
471  ('03', '19')
472  ('01', '94')
473  ('09', '14')
474  ('04', '19')
475  ('01', '95')
476  ('09', '15')
477  ('05', '19')
```



- For the simple dataset.

reducer\part-0000 × mapper\part-00000 × fa

```
('AB', 'CD')
('AC', 'BD')
('AD', 'BC')
('AB', 'CDE')
('BC', 'ADE')
('BD', 'ACE')
('BE', 'ACD')
('AC', 'BDE')
('BC', 'ADE')
('CD', 'ABE')
('CE', 'ABD')
('AD', 'BCE')
('BD', 'ACE')
('CD', 'ABE')
('DE', 'ABC')
('BE', 'CD')
('CE', 'BD')
('DE', 'BC')
```



Part 2

Aim :

1. Create a Spark DataFrame using one of datasets and try to use all different StructType.

We have used the dataset "FIFA World Cup" for this part.

```scala
def main(args: Array[String]): Unit = {

  //Setting up the Spark Session and Spark Context
  val conf = new SparkConf().setMaster("local[2]").setAppName("Task2")
  val sc = new SparkContext(conf)
  val spark = SparkSession
    .builder()
    .appName( name = "Task2")
    .config(conf =conf)
    .getOrCreate()

  Logger.getLogger( name = "org").setLevel(Level.ERROR)
  Logger.getLogger( name = "akka").setLevel(Level.ERROR)

  // We are using all 3 Fifa dataset given on Kaggle Repository
  //a.Import the dataset and create df and print Schema

  val df1 = spark.read
    .format( source = "csv")
    .option("header", "true") //reading the headers
    .option("mode", "DROPMALFORMED")
    .load( path = "WorldCups.csv")

  val df2 = spark.read
    .format( source = "csv")
    .option("header", "true") //reading the headers
    .option("mode", "DROPMALFORMED")
    .load( path = "WorldCupPlayers.csv")

```

2. Perform 10 intuitive questions in Dataset.

```
// Find the winner by years using WorldCup view
val Q = spark.sql( sqlText = "select Winner, Country, Year from WC Order By Country ")
Q.show()

//Find the goals by years using WorldCup view
val Q1 = spark.sql( sqlText = "select QualifiedTeams, MatchesPlayed, Year from WC WHERE Country = 'Brazil' Order
Q1.show()

//Cities that hosted highest world cup matches on view wcMatches
val Q2 = spark.sql( sqlText = "select Count(City),City from Matches Group By City")
Q2.show()

//Teams with the most world cup final victories on WorldCup view
val Q3 = spark.sql( sqlText = "select Count(Winner),Winner,Attendance from WC Group By Winner, Attendance")
Q3.show()

// Display all Stage Finalers in the year 1934
val Q4 = spark.sql( sqlText = "select * from Matches where Stage='Final' AND Year  = 1934 ")
Q4.show()

//matches held by coach CAUDRON Raoul (FRA)
val Q5 =spark.sql( sqlText = "select * from Players where `Coach Name` = 'CAUDRON Raoul (FRA)'")
Q5.show()

//No of matches in year 1934 and in san siro stadium
val Q6 = spark.sql( sqlText = "select count(*) from Matches where year=1934 AND Stadium = 'San Siro' ")
Q6.show()
```

```
//No of matches in year 1934 and in san siro stadium
val Q6 = spark.sql( sqlText = "select count(*) from Matches where year=1934 AND Stadium = 'San Siro' ")
Q6.show()

//number of matches that held in Estadio Centenario stadium
val Q7 = spark.sql( sqlText = "select count(*) from Matches where Stadium = 'Estadio Centenario'")
Q7.show()

//Country which hoster World Cup highest number of times
val Q8 = spark.sql( sqlText = "select Count(Country),Country,Year from WC Group by Country,Year")
Q8.show()

//Stadium with highest number of matches
val Q9 = spark.sql( sqlText = "select Count(Stadium),Stadium from Matches Group By Stadium")
Q9.show()

val Q10 = spark.sql( sqlText = "select `Player Name`, Position from Players where Position = 'GK' ")
Q10.show()



//HomeTeam Goals Count and their stage by Years
val Q11 = spark.sql( sqlText = "select `Home Team Name`,Stage,Year FROM Matches Group By Year,`Home Team Name`,Stage")
Q11.show()

// Away Team Goals and their stage
val Q12 = spark.sql( sqlText = "select `Away Team Name`,Stage,Year from Matches Group By Year,`Away Team Name`,Stage")
Q12.show()
```

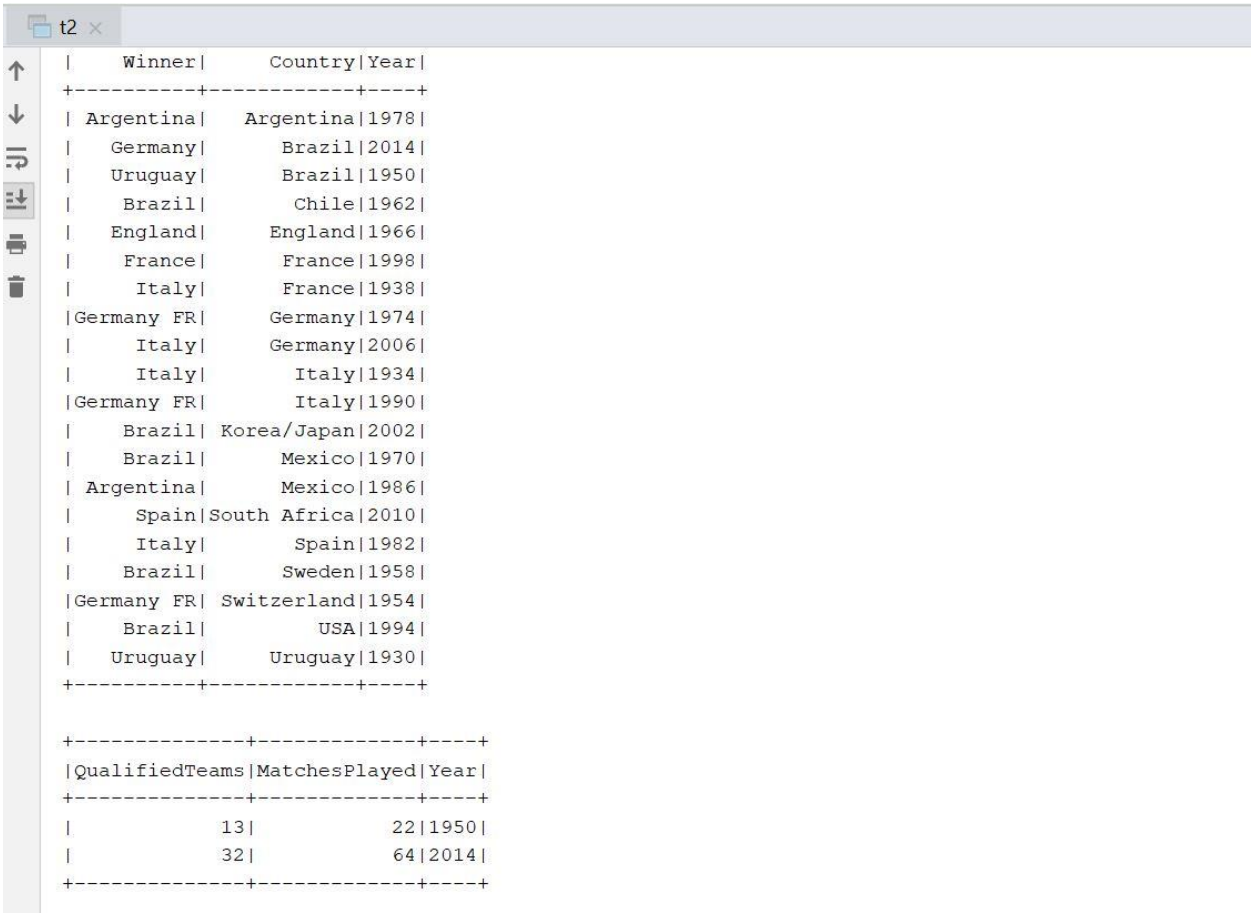
3. Perform any 5 queries in Spark RDD's and Spark Data Frames.

```scala
val csv = sc.textFile( path = "WorldCups.csv")

val h1 = csv.first()

val data = csv.filter(line => line != h1)

data.foreach(println)

val rdd = data.map(line=>line.split( regex = ",")).collect()

//rdd.foreach(println)
//RDD Highest Numbers of goals
val rdd1 = data.filter(line => line.split( regex = ",")(0) == "2006").map(line => (line.split( regex = ",")(0),
  (line.split( regex = ",")(1)), (line.split( regex = ",")(2)), (line.split( regex = ",")(3)) ) )
rdd1.foreach(println)

// Dataframe
df1.select( col = "Year", cols = "Country", "Winner").filter( conditionExpr = "Year =2006").show( numRows = 10)

// Dataframe SQL
val dfQ1 = spark.sql( sqlText = "select Year, Country, Winner FROM WC WHERE Year = 2006 order by Year Desc Limit 10").show()
```

```scala
val rdd2 = data.filter(line => (line.split( regex = ",")(2)=="Italy" ))
  .map(line=> (line.split( regex = ",")(0),line.split( regex = ",")(2),line.split( regex = ",")(3),line.split( regex = ",")(4),lin
    ,line.split( regex = ",")(5))).collect()
rdd2.foreach(println)

// Using Dataframe
df1.select( col = "Year", cols = "Winner","Runners-Up","Third", "Fourth").filter( conditionExpr = "Winner == 'Italy'").show( numRo

// usig Spark SQL
val DFQ2 = spark.sql( sqlText = "select * from WC where Winner = 'Italy' order by Year").show( numRows = 10)


// Details of years ending in ZERO
// RDD
val rdd3 = data.filter(line => (line.split( regex = ",")(7)>"16" ))
  .map(line=> (line.split( regex = ",")(0),line.split( regex = ",")(2),line.split( regex = ",")(6), line.split( regex = ",")(7))).
rdd3.foreach(println)

//DataFrame
df1.select( col = "Year", cols = "Winner","QualifiedTeams").filter( conditionExpr = "QualifiedTeams > 16").show( numRows = 10)

//DF - SQL
val DFQ3 = spark.sql( sqlText = "SELECT Year, Winner, QualifiedTeams from WC where QualifiedTeams > 16  ").show( numRows = 10)
```

```scala
// Using Dataframe
df1.select( col = "Year", cols = "Country","Fourth").filter( conditionExpr = "Country==Fourth").show( numRows = 10)

// usig Spark SQL
val DFQ4 = spark.sql( sqlText = "select Year,Country,Fourth from WC where Country = Fourth order by Year").show()

//Max matches played
//RDD
val rdd5 = data.filter(line=>line.split( regex = ",")(8) > "55")
  .map(line=> (line.split( regex = ",")(0),line.split( regex = ",")(8),line.split( regex = ",")(3))).collect()
rdd5.foreach(println)

// DataFrame
df1.filter( conditionExpr = "MatchesPlayed > 55").show()

// Spark SQL
val DFQ5 = spark.sql( sqlText = " Select * from WC where MatchesPlayed in " +
  "(Select Max(MatchesPlayed) from WC )" ).show()

}
```

Output :



t2

```
|    Winner|     Country|Year|
+----------+------------+----+
| Argentina|   Argentina|1978|
|   Germany|      Brazil|2014|
|   Uruguay|      Brazil|1950|
|    Brazil|       Chile|1962|
|   England|     England|1966|
|    France|      France|1998|
|     Italy|      France|1938|
|Germany FR|     Germany|1974|
|     Italy|     Germany|2006|
|     Italy|       Italy|1934|
|Germany FR|       Italy|1990|
|    Brazil| Korea/Japan|2002|
|    Brazil|      Mexico|1970|
| Argentina|      Mexico|1986|
|     Spain|South Africa|2010|
|     Italy|       Spain|1982|
|    Brazil|      Sweden|1958|
|Germany FR| Switzerland|1954|
|    Brazil|         USA|1994|
|   Uruguay|     Uruguay|1930|
+----------+------------+----+

+--------------+-------------+----+
|QualifiedTeams|MatchesPlayed|Year|
+--------------+-------------+----+
|            13|           22|1950|
|            32|           64|2014|
+--------------+-------------+----+
```

t2

count(City)	City
4	Daegu
9	Paris
4	Natal
6	San Francisco
10	Santiago De Chile
1	Eskilstuna
3	La Coru?A
3	Bilbao
4	Geneva
1	Le Havre
4	Verona
3	Kobe
8	Solna
5	Liverpool
3	Gwangju
4	Cuiaba
3	Niigata
17	Guadalajara
6	Boston
7	Madrid

only showing top 20 rows

t2

count(Winner)	Winner	Attendance
1	Germany	3.386.810
1	Italy	375.700
1	Germany FR	768.607
1	France	2.785.100
1	Brazil	819.810
1	Argentina	1.545.791
1	Uruguay	1.045.246
1	Spain	3.178.856
1	Germany FR	2.516.215
1	Germany FR	1.865.753
1	Brazil	893.172
1	Italy	3.359.439
1	Argentina	2.394.031
1	Brazil	1.603.975
1	Italy	363.000
1	Brazil	2.705.197
1	Uruguay	590.549
1	Italy	2.109.723
1	Brazil	3.587.538
1	England	1.563.135

Year	Datetime	Stage	Stadium	City	Home Team Name	Home Team Goals	Away Team Goals	Away Team Name	Win conditions	Attendance	Half-time
1934	10 Jun 1934 - 17:30	Final	Nazionale PNF	Rome	Italy	2	1	Czechoslovakia	Italy win after e...	55000	

RoundID	MatchID	Team Initials	Coach Name	Line-up	Shirt Number	Player Name	Position	Event
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Alex THEPOT	GK	null
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Marcel LANGILLER	null	G40'
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Ernest LIBERATI	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Andre MASCHINOT	null	G43' G87'
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Etienne MATTLER	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Marcel PINEL	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Alex VILLAPLANE	C	null
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Lucien LAURENT	null	G19'
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Marcel CAPELLE	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Augustin CHANTREL	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	S	0	Edmond DELFOUR	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	N	0	Celestin DELMER	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	N	0	Andre TASSIN	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	N	0	Nouma ANDOIRE	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	N	0	Jean LAURENT	null	null
201	1096	FRA	CAUDRON Raoul (FRA)	N	0	Emile VEINANTE	null	null
201	1085	FRA	CAUDRON Raoul (FRA)	S	0	Alex THEPOT	GK	null
201	1085	FRA	CAUDRON Raoul (FRA)	S	0	Alex VILLAPLANE	C	null
201	1085	FRA	CAUDRON Raoul (FRA)	S	0	Lucien LAURENT	null	null
201	1085	FRA	CAUDRON Raoul (FRA)	S	0	Marcel CAPELLE	null	null

only showing top 20 rows

count(1)
3

```
|count(1)|
+--------+
|      10|
+--------+
```

count(Country)	Country	Year
1	Italy	1934
1	Spain	1982
1	France	1938
1	South Africa	2010
1	France	1998
1	Chile	1962
1	Sweden	1958
1	USA	1994
1	England	1966
1	Switzerland	1954
1	Mexico	1986
1	Korea/Japan	2002
1	Uruguay	1930
1	Argentina	1978
1	Mexico	1970
1	Brazil	1950
1	Germany	1974
1	Italy	1990
1	Brazil	2014
1	Germany	2006

```
+--------------+--------------------+
|count(Stadium)|             Stadium|
+--------------+--------------------+
|             8|           Cuauhtemoc|
|             6|       Parque Central|
|             3|        Idrottsparken|
|             5|          Waldstadion|
|             1|                Friuli|
|             3|        Jose Zorrilla|
|             3|Old Trafford Stadium|
|             3|            San Mames|
|             3|       Miyagi Stadium|
|             6|FIFA World Cup St...|
|             6|Royal Bafokeng Sp...|
|             3|        Nuevo Estadio|
|             4|       Arena Amazonia|
|            11|Nou Camp - Estadi...|
|             4|    Santiago Bernabeu|
|             3| Osaka Nagai Stadium|
|             6|Estadio Jos Mar...|
|             2|Ramon Sanchez Piz...|
|             4|      Renato Dall Ara|
|             4|   Pontiac Silverdome|
+--------------+--------------------+
only showing top 20 rows
```

```
+-------------------+--------+
|        Player Name|Position|
+-------------------+--------+
|        Alex THEPOT|      GK|
|    Oscar BONFIGLIO|      GK|
|      Jimmy DOUGLAS|      GK|
|      Arnold BADJOU|      GK|
|     Milovan JAKSIC|      GK|
|               JOEL|      GK|
|     Ion LAPUSNEANU|      GK|
|    Juan VALDIVIESO|      GK|
|       Angel BOSSIO|      GK|
|        Alex THEPOT|      GK|
|     Roberto CORTES|      GK|
|       Isidoro SOTA|      GK|
|     Milovan JAKSIC|      GK|
|     Jesus BERMUDEZ|      GK|
|      Jimmy DOUGLAS|      GK|
|      Modesto DENIS|      GK|
|Enrique BALLESTRERO|      GK|
|       Jorge PARDON|      GK|
|     Roberto CORTES|      GK|
|        Alex THEPOT|      GK|
+-------------------+--------+
only showing top 20 rows
```

```
+--------------------+--------------------+----+
|      Home Team Name|               Stage|Year|
+--------------------+--------------------+----+
|             Belgium|             Group 1|1982|
|            Portugal|             Group F|1986|
|              Brazil|             Group 3|1962|
|          Germany FR|      Quarter-finals|1966|
|          Germany FR|Match for third p...|1970|
|             Denmark|             Group A|2002|
|              France|             Group A|2010|
|             Hungary|      Quarter-finals|1954|
|          Germany FR|             Group 1|1974|
|             England|             Group F|1990|
|          Germany FR|             Group B|1974|
|           Argentina|         Round of 16|2006|
|            Portugal|             Group G|2010|
|             Germany|         Round of 16|2014|
|          Yugoslavia|             Group 2|1958|
|           Argentina|             Group 4|1974|
|            Paraguay|             Group 4|1930|
|          Germany FR|         Semi-finals|1966|
|"rn"">Republic of...|             Group E|1994|
|            Honduras|             Group H|2010|
+--------------------+--------------------+----+
only showing top 20 rows
```

```
+--------------------+--------------------+----+
|      Away Team Name|               Stage|Year|
+--------------------+--------------------+----+
|         Switzerland|             Group 2|1962|
|             Belgium|             Group 1|1982|
|            Portugal|             Group F|1986|
|             Morocco|             Group 4|1970|
|             Uruguay|Match for third p...|1970|
|                 USA|         Round of 16|1994|
|             Denmark|             Group A|2002|
|              France|             Group A|2010|
|          Germany FR|             Group 1|1974|
|             Denmark|         Round of 16|1998|
|          Germany FR|             Group B|1974|
|               Spain|      Quarter-finals|1994|
|            Portugal|             Group G|2010|
|          Yugoslavia|             Group 2|1958|
|           Argentina|             Group 4|1974|
|                 USA|         Round of 16|2014|
|            Paraguay|             Group 4|1930|
|         Netherlands|         Round of 16|1990|
|"rn"">Republic of...|             Group E|1994|
|            Honduras|             Group H|2010|
+--------------------+--------------------+----+
only showing top 20 rows
```

```
1930,Uruguay,Uruguay,Argentina,USA,Yugoslavia,70,13,18,590.549
1934,Italy,Italy,Czechoslovakia,Germany,Austria,70,16,17,363.000
1978,Argentina,Argentina,Netherlands,Brazil,Italy,102,16,38,1.545.791
1938,France,Italy,Hungary,Brazil,Sweden,84,15,18,375.700
1982,Spain,Italy,Germany FR,Poland,France,146,24,52,2.109.723
1950,Brazil,Uruguay,Brazil,Sweden,Spain,88,13,22,1.045.246
1986,Mexico,Argentina,Germany FR,France,Belgium,132,24,52,2.394.031
1954,Switzerland,Germany FR,Hungary,Austria,Uruguay,140,16,26,768.607
1990,Italy,Germany FR,Argentina,Italy,England,115,24,52,2.516.215
1958,Sweden,Brazil,Sweden,France,Germany FR,126,16,35,819.810
1994,USA,Brazil,Italy,Sweden,Bulgaria,141,24,52,3.587.538
1962,Chile,Brazil,Czechoslovakia,Chile,Yugoslavia,89,16,32,893.172
1998,France,France,Brazil,Croatia,Netherlands,171,32,64,2.785.100
1966,England,England,Germany FR,Portugal,Soviet Union,89,16,32,1.563.135
2002,Korea/Japan,Brazil,Germany,Turkey,Korea Republic,161,32,64,2.705.197
1970,Mexico,Brazil,Italy,Germany FR,Uruguay,95,16,32,1.603.975
2006,Germany,Italy,France,Germany,Portugal,147,32,64,3.359.439
2010,South Africa,Spain,Netherlands,Germany,Uruguay,145,32,64,3.178.856
1974,Germany,Germany FR,Netherlands,Poland,Brazil,97,16,38,1.865.753
2014,Brazil,Germany,Argentina,Netherlands,Brazil,171,32,64,3.386.810
(2006,Germany,Italy,France)
+----+--------+------+
|Year|Country|Winner|
+----+--------+------+
|2006|Germany| Italy|
+----+--------+------+
```

```
(1934,Italy,Czechoslovakia,Germany,Austria,Austria)
(1938,Italy,Hungary,Brazil,Sweden,Sweden)
(1982,Italy,Germany FR,Poland,France,France)
(2006,Italy,France,Germany,Portugal,Portugal)
+----+------+--------------+-------+--------+
|Year|Winner|    Runners-Up|  Third|  Fourth|
+----+------+--------------+-------+--------+
|1934| Italy|Czechoslovakia|Germany| Austria|
|1938| Italy|       Hungary| Brazil|  Sweden|
|1982| Italy|    Germany FR| Poland|  France|
|2006| Italy|        France|Germany|Portugal|
+----+------+--------------+-------+--------+

+----+-------+------+--------------+-------+--------+-----------+--------------+-------------+----------+
|Year|Country|Winner|    Runners-Up|  Third|  Fourth|GoalsScored|QualifiedTeams|MatchesPlayed|Attendance|
+----+-------+------+--------------+-------+--------+-----------+--------------+-------------+----------+
|1934|  Italy| Italy|Czechoslovakia|Germany| Austria|         70|            16|           17|   363.000|
|1938| France| Italy|       Hungary| Brazil|  Sweden|         84|            15|           18|   375.700|
|1982|  Spain| Italy|    Germany FR| Poland|  France|        146|            24|           52| 2.109.723|
|2006|Germany| Italy|        France|Germany|Portugal|        147|            32|           64| 3.359.439|
+----+-------+------+--------------+-------+--------+-----------+--------------+-------------+----------+

(1982,Italy,146,24)
(1986,Argentina,132,24)
(1990,Germany FR,115,24)
(1994,Brazil,141,24)
(1998,France,171,32)
(2002,Brazil,161,32)
(2006,Italy,147,32)
(2010,Spain,145,32)
(2014,Germany,171,32)
```

```
|1990|Germany FR|            24|
|1994|    Brazil|            24|
|1998|    France|            32|
|2002|    Brazil|            32|
|2006|     Italy|            32|
|2010|     Spain|            32|
|2014|   Germany|            32|
+----+----------+--------------+

(2014,Brazil,Brazil)
+----+-------+------+
|Year|Country|Fourth|
+----+-------+------+
|2014| Brazil|Brazil|
+----+-------+------+

+----+-------+------+
|Year|Country|Fourth|
+----+-------+------+
|2014| Brazil|Brazil|
+----+-------+------+

(1998,64,Brazil)
(2002,64,Germany)
(2006,64,France)
(2010,64,Netherlands)
(2014,64,Argentina)
+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+
|Year|     Country| Winner| Runners-Up|      Third|        Fourth|GoalsScored|QualifiedTeams|MatchesPlayed|Attendance|
+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+
|1998|      France| France|     Brazil|    Croatia|   Netherlands|        171|            32|           64| 2.785.100|
```

```
+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+
|Year|     Country| Winner| Runners-Up|      Third|        Fourth|GoalsScored|QualifiedTeams|MatchesPlayed|Attendance|
+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+
|1998|      France| France|     Brazil|    Croatia|   Netherlands|        171|            32|           64| 2.785.100|
|2002| Korea/Japan| Brazil|    Germany|     Turkey|Korea Republic|        161|            32|           64| 2.705.197|
|2006|     Germany|  Italy|     France|    Germany|      Portugal|        147|            32|           64| 3.359.439|
|2010|South Africa|  Spain|Netherlands|    Germany|       Uruguay|        145|            32|           64| 3.178.856|
|2014|      Brazil|Germany|  Argentina|Netherlands|        Brazil|        171|            32|           64| 3.386.810|
+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+

+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+
|Year|     Country| Winner| Runners-Up|      Third|        Fourth|GoalsScored|QualifiedTeams|MatchesPlayed|Attendance|
+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+
|1998|      France| France|     Brazil|    Croatia|   Netherlands|        171|            32|           64| 2.785.100|
|2002| Korea/Japan| Brazil|    Germany|     Turkey|Korea Republic|        161|            32|           64| 2.705.197|
|2006|     Germany|  Italy|     France|    Germany|      Portugal|        147|            32|           64| 3.359.439|
|2010|South Africa|  Spain|Netherlands|    Germany|       Uruguay|        145|            32|           64| 3.178.856|
|2014|      Brazil|Germany|  Argentina|Netherlands|        Brazil|        171|            32|           64| 3.386.810|
+----+------------+-------+-----------+-----------+--------------+-----------+--------------+-------------+----------+


Process finished with exit code 0
```

Part 3

Aim :

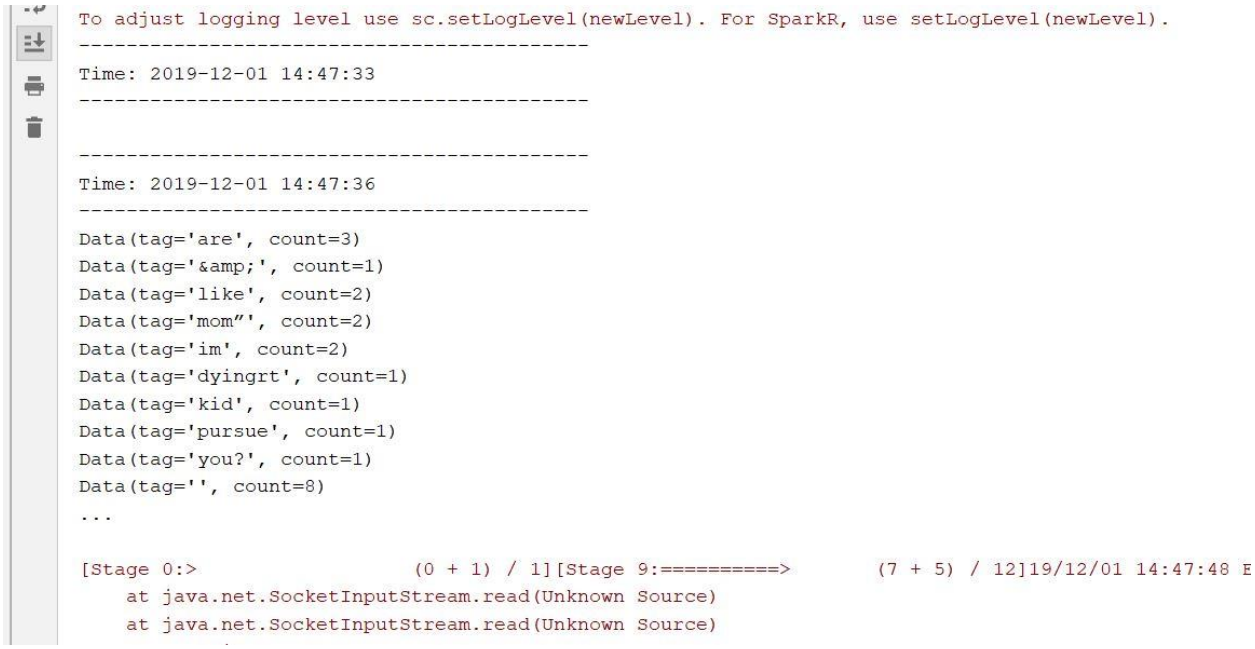
Spark Streaming : Perform Word-Count on Twitter Streaming Data using Spark.

```
CONSUMER_KEY = 'y37L6Vykcr0AvxvDf10axhKsc'
CONSUMER_SECRET = 'K8Sk4VSDTp0ijSgqBQ5tk8eAXfa1gcQbNoGkm8a3KKzDTdz2a9'
ACCESS_TOKEN = '2886203293-rx1AypFuSuAmNrjLeFI0ShrwpUbz8R2SZuRDU0H'
ACCESS_TOKEN_SECRET ='qKpWIQ7ujh8aA1eHlVzyrGcptWqjXZh9rUaTxn0T5yN7x'

def validTweet(str_tweet):
    json_tweet = json.loads(str_tweet)
    return False if list(json_tweet.keys())[0] == 'delete' or list(json_tweet.keys())[0] == 'limit' else True

class TwitterStreamListener(tweepy.StreamListener):
    def __init__(self, csocket):
        self.client_socket = csocket

    def on_data(self, data):
        if validTweet(data):
            tweet = json.loads(data)
            self.client_socket.send(tweet["text"].encode('utf-8'))

    def on_error(self, status):
        print(status)
```

1.py 2.py

```
from pyspark import SparkContext
from pyspark.streaming import StreamingContext
from collections import namedtuple

import os
os.environ["SPARK_HOME"] = "C:\spark-2.4.4-bin-hadoop2.7"
os.environ["HADOOP_HOME"] = "C:\\winutils"

sc = SparkContext(appName="Lab 4")

# Change log level to error
logger = sc._jvm.org.apache.log4j
logger.LogManager.getRootLogger().setLevel(logger.Level.ERROR)

ssc = StreamingContext(sc, 3)

Tweet = namedtuple("Data", ("tag", "count"))

# Split each line into words and use map reduce to count occurance of token then print word count
ssc.socketTextStream("localhost", 9000).flatMap(lambda line: line.split(" ")).map(lambda word: (word.lower(), 1)).reduceByKey

# Start spark streaming
ssc.start()
ssc.awaitTermination()
```

Output :



```
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
-------------------------------------------
Time: 2019-12-01 14:47:33
-------------------------------------------

-------------------------------------------
Time: 2019-12-01 14:47:36
-------------------------------------------
Data(tag='are', count=3)
Data(tag='&amp;', count=1)
Data(tag='like', count=2)
Data(tag='mom”', count=2)
Data(tag='im', count=2)
Data(tag='dyingrt', count=1)
Data(tag='kid', count=1)
Data(tag='pursue', count=1)
Data(tag='you?', count=1)
Data(tag='', count=8)
...

[Stage 0:>                  (0 + 1) / 1][Stage 9:==========>       (7 + 5) / 12]19/12/01 14:47:48 E
    at java.net.SocketInputStream.read(Unknown Source)
    at java.net.SocketInputStream.read(Unknown Source)
```

Part 4

Aim :

Spark Graphx Task

We used the dataset Nashville for this task.

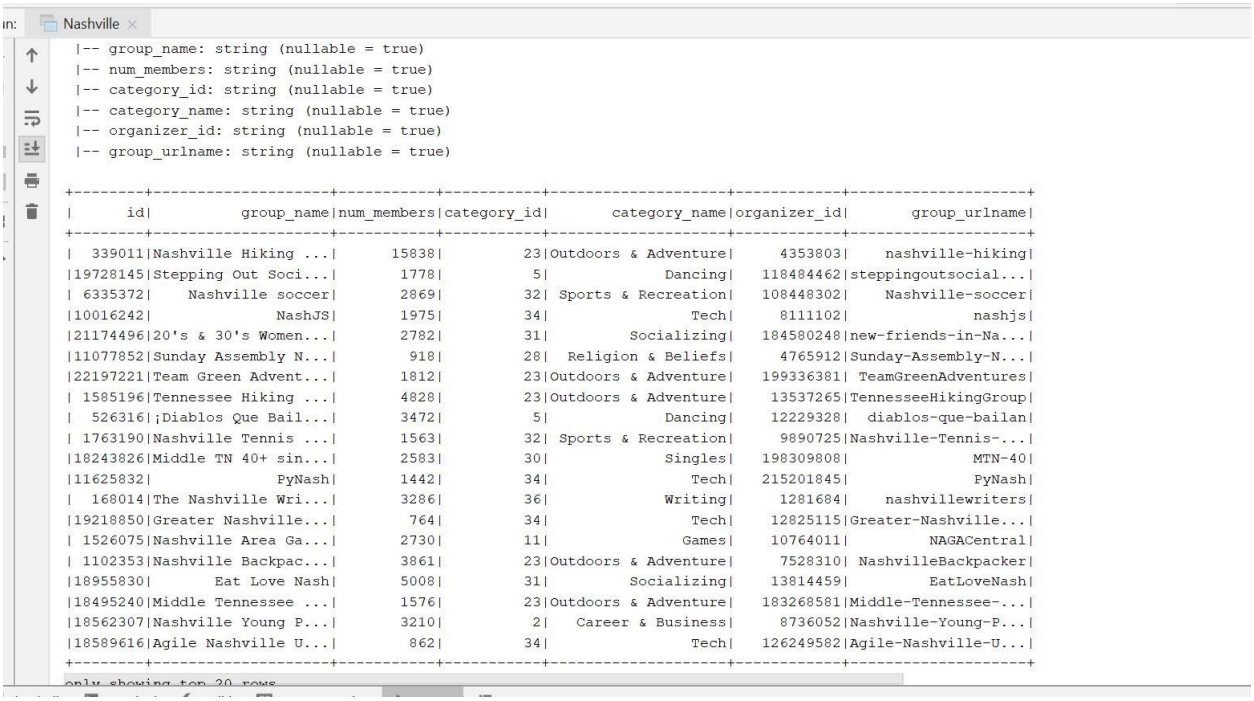
```scala
object Nashville {
  def main(args: Array[String]): Unit = {
    System.setProperty("hadoop.home.dir", "C:\\winutils");
    val conf = new SparkConf().setMaster("local[2]").setAppName("PAGE_RANK")
    val sc = new SparkContext(conf)
    val spark = SparkSession
      .builder()
      .appName( name = "PAGE_RANK")
      .config(conf =conf)
      .getOrCreate()


    Logger.getLogger( name = "org").setLevel(Level.ERROR)
    Logger.getLogger( name = "akka").setLevel(Level.ERROR)

    val groups_df = spark.read
      .format( source = "csv")
      .option("header", "true") //reading the headers
      .option("mode", "DROPMALFORMED")
      .load( path = "meta-groups.csv")

    val edges_df = spark.read
      .format( source = "csv")
      .option("header", "true") //reading the headers
      .option("mode", "DROPMALFORMED")
      .load( path = "group-edges.csv")

```

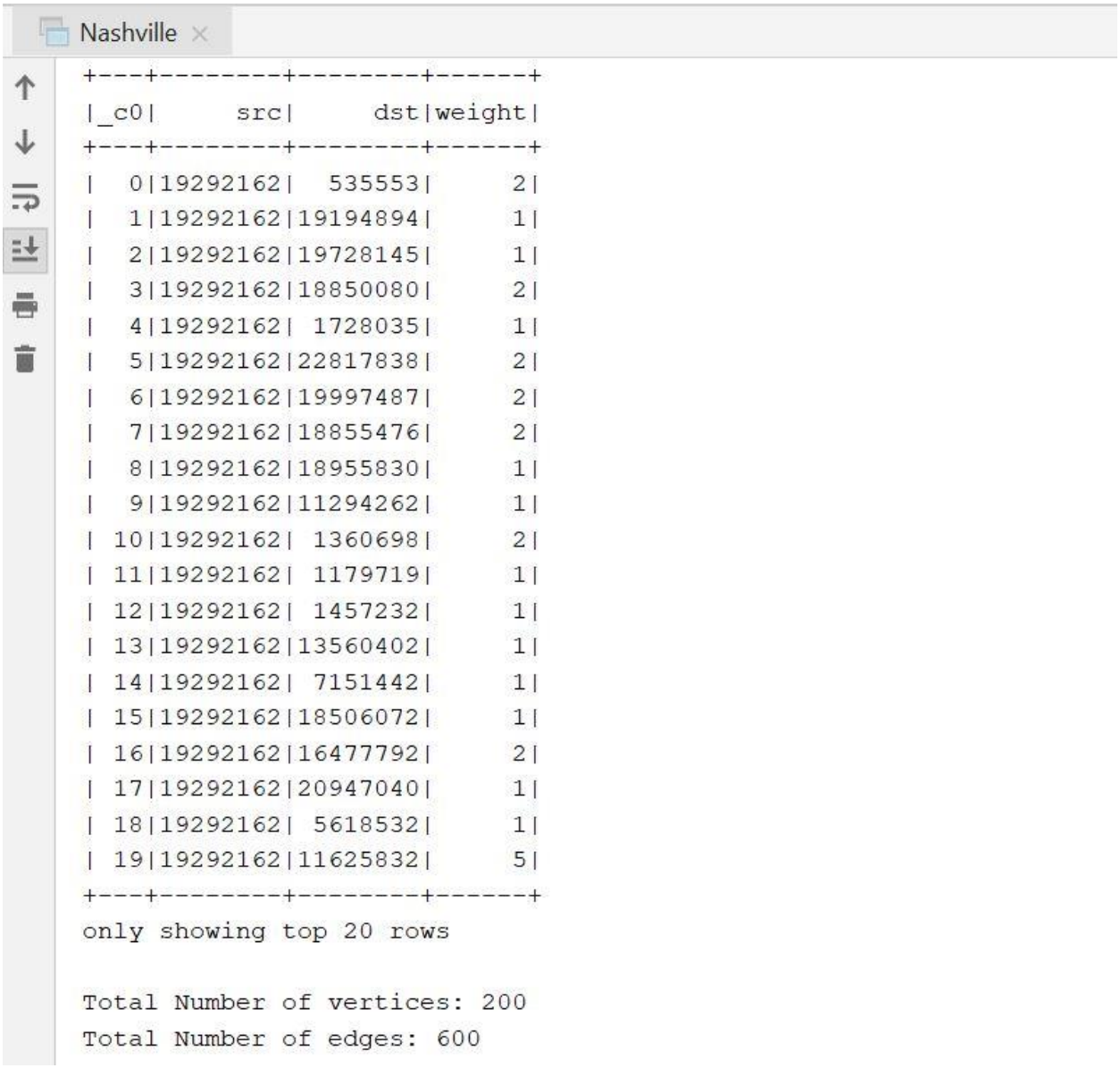
1. Perform Page Rank

build.sbt × 1.scala × group-edges.csv ×

```scala
    val vertices = g2
      .withColumnRenamed( existingName = "group_id",  newName = "id").limit(200)
      .distinct()

    val edges = e2
      .withColumnRenamed( existingName = "group1",  newName = "src").limit(600).distinct()
      .withColumnRenamed( existingName = "group2",  newName = "dst").limit(600).distinct()


    val graph = GraphFrame(vertices, edges)

    edges.cache()
    vertices.cache()
    graph.vertices.show()
    graph.edges.show()


    println("Total Number of vertices: " + graph.vertices.count)
    println("Total Number of edges: " + graph.edges.count)


    val stationPageRank = graph.pageRank.resetProbability( value = 0.15).tol( value = 0.01).run()
    stationPageRank.vertices.show()
    stationPageRank.edges.show()
  }

}
```

Output :


Nashville
|-- group_name: string (nullable = true)
|-- num_members: string (nullable = true)
|-- category_id: string (nullable = true)
|-- category_name: string (nullable = true)
|-- organizer_id: string (nullable = true)
|-- group_urlname: string (nullable = true)
+--------+--------------------+-----------+-----------+-------------------+------------+--------------------+
| id| group_name|num_members|category_id| category_name|organizer_id| group_urlname|
+--------+--------------------+-----------+-----------+-------------------+------------+--------------------+
339011	Nashville Hiking ...	15838	23	Outdoors & Adventure	4353803	nashville-hiking
19728145	Stepping Out Soci...	1778	5	Dancing	118484462	steppingoutsocial...
6335372	Nashville soccer	2869	32	Sports & Recreation	108448302	Nashville-soccer
10016242	NashJS	1975	34	Tech	8111102	nashjs
21174496	20's & 30's Women...	2782	31	Socializing	184580248	new-friends-in-Na...
11077852	Sunday Assembly N...	918	28	Religion & Beliefs	4765912	Sunday-Assembly-N...
22197221	Team Green Advent...	1812	23	Outdoors & Adventure	199336381	TeamGreenAdventures
1585196	Tennessee Hiking ...	4828	23	Outdoors & Adventure	13537265	TennesseeHikingGroup
526316	¡Diablos Que Bail...	3472	5	Dancing	12229328	diablos-que-bailan
1763190	Nashville Tennis ...	1563	32	Sports & Recreation	9890725	Nashville-Tennis-...
18243826	Middle TN 40+ sin...	2583	30	Singles	198309808	MTN-40
11625832	PyNash	1442	34	Tech	215201845	PyNash
168014	The Nashville Wri...	3286	36	Writing	1281684	nashvillewriters
19218850	Greater Nashville...	764	34	Tech	12825115	Greater-Nashville...
1526075	Nashville Area Ga...	2730	11	Games	10764011	NAGACentral
1102353	Nashville Backpac...	3861	23	Outdoors & Adventure	7528310	NashvilleBackpacker
18955830	Eat Love Nash	5008	31	Socializing	13814459	EatLoveNash
18495240	Middle Tennessee ...	1576	23	Outdoors & Adventure	183268581	Middle-Tennessee-...
18562307	Nashville Young P...	3210	2	Career & Business	8736052	Nashville-Young-P...
18589616	Agile Nashville U...	862	34	Tech	126249582	Agile-Nashville-U...
+--------+--------------------+-----------+-----------+-------------------+------------+--------------------+
only showing top 20 rows


Nashville
+---+--------+--------+------+
|_c0| src| dst|weight|
+---+--------+--------+------+
0	19292162	535553	2
1	19292162	19194894	1
2	19292162	19728145	1
3	19292162	18850080	2
4	19292162	1728035	1
5	19292162	22817838	2
6	19292162	19997487	2
7	19292162	18855476	2
8	19292162	18955830	1
9	19292162	11294262	1
10	19292162	1360698	2
11	19292162	1179719	1
12	19292162	1457232	1
13	19292162	13560402	1
14	19292162	7151442	1
15	19292162	18506072	1
16	19292162	16477792	2
17	19292162	20947040	1
18	19292162	5618532	1
19	19292162	11625832	5
+---+--------+--------+------+
only showing top 20 rows
Total Number of vertices: 200
Total Number of edges: 600

Nashville

```
Total Number of edges: 600
+--------+--------------------+-----------+-----------+--------------------+------------+--------------------+--------------------+
|      id|          group_name|num_members|category_id|       category_name|organizer_id|       group_urlname|           pagerank|
+--------+--------------------+-----------+-----------+--------------------+------------+--------------------+--------------------+
|  405938|MTRAS ~ MidTn Rob...|        525|         34|                Tech|     3246917|         robotics-71|0.9711094925952904|
|18616278|Franklin Develope...|        629|         34|                Tech|   170855672|franklin-develope...|1.00498540512768433|
|19416348|Bellevue Business...|        298|          2|   Career & Business|    83272622|Bellevue-Business...| 1.021979821248102|
|24125934|Murfreesboro Web ...|         43|         34|                Tech|   178742432|Murfreesboro-Web-...|0.9711094925952904|
|22736876|Business Connecti...|        126|          2|   Career & Business|   191532521|Brentwood-Rowdy-R...| 0.980707667812802|
|18494105|The Iron Yard - N...|       1491|         34|                Tech|   104388972|The-Iron-Yard-Nas...| 1.001343744530452|
|18529135|Franklin AM - Net...|        360|          2|   Career & Business|    34583172|Franklin-AM-Netwo...| 1.001343744530452|
|11625832|              PyNash|       1442|         34|                Tech|   215201845|              PyNash|1.0211705633376058|
|20135961|20s/30s Nashville...|       1124|         31|         Socializing|   198403977|Nashville-Online-...|0.9711094925952904|
|  535553|             Nash.rb|        881|         34|                Tech|    14344641|              nashrb|1.0115723881200942|
|19528743|Nashville Real Es...|        441|          2|   Career & Business|   144256692|Nashville-Real-Es...|0.9917455693129404|
|15335602|Brentwood TN Conv...|        315|         16|Language & Ethnic...|   153513242|Brentwood-TN-Conv...|1.0175289027403736|
|18314164|              NashBI|        784|         34|                Tech|   183427754|              NashBI| 0.980707667812802|
| 6707902|Data Science Nash...|       1046|         34|                Tech|    14589429|Data-Science-Nash...|1.00498540512768433|
|  541319|The Nashville Son...|       2644|         21|               Music|     2984170|       vocalists-164| 1.021979821248102|
|20493986|  R-Ladies Nashville|        210|          2|   Career & Business|   213434886|   rladies-nashville| 0.980707667812802|
|  339011|Nashville Hiking ...|      15838|         23|Outdoors & Adventure|     4353803|    nashville-hiking|1.0725584406541067|
|20583464|Mediumship and In...|        234|         22|New Age & Spiritu...|     5212354|Mediumship-and-In...| 0.980707667812802|
| 4126912|Nashville Online ...|       1532|          2|   Career & Business|    44942272|    nashville-online| 1.021979821248102|
|22197221|Team Green Advent...|       1812|         23|Outdoors & Adventure|   199336381| TeamGreenAdventures|1.0563732824441852|
+--------+--------------------+-----------+-----------+--------------------+------------+--------------------+--------------------+
only showing top 20 rows
```

Nashville

```
only showing top 20 rows

+---+--------+--------+------+--------------------+
|_c0|     src|     dst|weight|              weight|
+---+--------+--------+------+--------------------+
|268| 1585196|18506072|     6|0.011627906976744186|
|441| 1417288| 1498076|     1|               0.025|
|324| 1179719|19030621|     1|  0.0196078431372549|
|388| 1417288|18955830|     1|               0.025|
|364| 1179719|19934054|     1|  0.0196078431372549|
|555| 3047512|18955830|     1| 0.06666666666666667|
|371| 1179719|23674770|     3|  0.0196078431372549|
|215| 1585196|20166757|     1|0.011627906976744186|
| 23|19292162|16487812|     5|0.029411764705882353|
|495|  168014| 4126912|     1|               0.025|
|391| 1417288| 1772099|     1|               0.025|
|394| 1417288|  168014|     1|               0.025|
|417| 1417288| 1358081|     2|               0.025|
|241| 1585196| 1307837|     1|0.011627906976744186|
| 36|19292162|19654655|     1|0.029411764705882353|
|162|20135961|18506072|     3|                0.25|
|239| 1585196|15335602|     1|0.011627906976744186|
|205| 1585196|22023226|     1|0.011627906976744186|
|317| 1179719|18855476|     1|  0.0196078431372549|
|234| 1585196|11131552|     2|0.011627906976744186|
+---+--------+--------+------+--------------------+
only showing top 20 rows
```

2. State importance of using graphx on the chosen dataset.

Graphx are mainly used for distributed processing of graphs. For example, where a graph is very large with huge no:of vertices and edges then it is difficult to process on a single state machine. Then we need to use parallel computation. Here, we used group-id to produce vertices and group-1,group-2 taken from the dataset which is used to produce edges for the graphs.