

# Exercise\_8

June 30, 2019

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### 0.1.1 Exercise 1 - Part (a)

```
In [0]: #import libraries
        from pyspark import SparkContext
        from pyspark.sql import SQLContext
        import pandas as pd
        from pyspark.sql import Row
        import numpy as np

In [0]: sc = SparkContext()
        sqlContext = SQLContext(sc)

In [0]: #List of words
        a = ["spark", "rdd", "python", "context", "create", "class"]
        b = ["operation", "apache", "scala", "lambda", "parallel", "partition"]

        #Making RDDs
        rdd_A = sc.parallelize(a)
        rdd_B= sc.parallelize(b)

        #Mapping the names of RDD
        A = rdd_A.map(lambda x: Row(name = x))
        B = rdd_B.map(lambda x: Row(name = x))

        #Creating dataframe from RDD
        dfA = sqlContext.createDataFrame(A)
        dfB = sqlContext.createDataFrame(B)

        #Creating Alias
        df1 = dfA.alias("df1")
        df2 = dfB.alias("df2")
```

### 0.1.2 Right outer join:

A RIGHT OUTER JOIN is one of the JOIN operations that allow you to specify a JOIN clause. It preserves the unmatched rows from the second (right) table, joining them with a NULL in the shape of the first (left) table.

```
In [35]: #Right Outer Join
         right_outer = df1.join(other=df2,on="name",how='right_outer')
         print("Right Outer Join:")
         right_outer.show()
```

Right Outer Join:

```
+-----+
|    name|
+-----+
|operation|
|  lambda|
|partition|
| parallel|
|    scala|
|   apache|
+-----+
```

### 0.1.3 Full outer join:

A FULL OUTER JOIN combines the results of both left and right outer joins and returns all (matched or unmatched) rows from the tables on both sides of the join clause

```
In [36]: #Full Outer Join
         full_outer = df1.join(other=df2,on="name",how='full_outer')
         print("Full Outer Join:")
         full_outer.show()
```

Full Outer Join:

```
+-----+
|    name|
+-----+
|operation|
|  lambda|
| context|
|partition|
|  create|
|    rdd|
| parallel|
|    scala|
|   apache|
|    spark|
|    class|
|   python|
+-----+
```

```
In [37]: #Mapping the RDD
map_rdd = full_outer.rdd.map(lambda x: sum([word.count('s') for word in x]))
map_df = map_rdd.map(lambda x: Row(name = x))
map_df = sqlContext.createDataFrame(map_df)
print("Mapped Dataframe with count of 's'")
map_df.show()

#Reducing the RDD
reduce_rdd=map_rdd.reduce(lambda x,y: x+y)
print("Using Map-Reduce, the character \"s\" appears",reduce_rdd,"times in all
a and b.\n" )
```

Mapped Dataframe with count of 's'

```
+----+
|name|
+----+
|  0 |
|  0 |
|  0 |
|  0 |
|  0 |
|  0 |
|  0 |
|  0 |
|  1 |
|  0 |
|  1 |
|  2 |
|  0 |
+----+
```

Using Map-Reduce, the character "s" appears 4 times in all a and b.

```
In [38]: #Aggregate function
count = full_outer.rdd.aggregate(0, lambda i, x: i + x[0].count('s'),
                                  lambda i, j: i+j)
print("Using aggregate function, the character \"s\" appears",count,
      "times in all a and b." )
```

Using aggregate function, the character "s" appears 4 times in all a and b.

#### 0.1.4 Exercise 1 - Part (b)

```
In [31]: json_file = "gdrive/My Drive/DDA/Spark1/students.json"
df = sqlContext.read.json(json_file)
print("Students records: \n")
df.show()
```

Students records:

course	dob	first_name	last_name	points	s_id
Humanities and Art	October 14, 1983	Alan	Joe	10	1
Computer Science	September 26, 1980	Martin	Genberg	17	2
Graphic Design	June 12, 1982	Athur	Watson	16	3
Graphic Design	April 5, 1987	Anabelle	Sanberg	12	4
Psychology	November 1, 1978	Kira	Schommer	11	5
Business	17 February 1981	Christian	Kiriam	10	6
Machine Learning	1 January 1984	Barbara	Ballard	14	7
Deep Learning	January 13, 1978	John	null	10	8
Machine Learning	26 December 1989	Marcus	Carson	15	9
Physics	30 December 1987	Marta	Brooks	11	10
Data Analytics	June 12, 1975	Holly	Schwartz	12	11
Computer Science	July 2, 1985	April	Black	null	12
Computer Science	July 22, 1980	Irene	Bradley	13	13
Psychology	7 February 1986	Mark	Weber	12	14
Informatics	May 18, 1987	Rosie	Norman	9	15
Business	August 10, 1984	Martin	Steele	7	16
Machine Learning	16 December 1990	Colin	Martinez	9	17
Data Analytics	null	Bridget	Twain	6	18
Business	7 March 1980	Darlene	Mills	19	19
Data Analytics	June 2, 1985	Zachary	null	10	20

```
In [32]: from pyspark.sql.functions import mean,col
        avg = df.select(mean(col('points')).alias('mean')).collect()
        df = df.na.fill(avg[0]['mean'])
        print("Replacing the null values in column points by mean of all points: \n")
        df.show()
```

Replacing the null values in column points by mean of all points:

course	dob	first_name	last_name	points	s_id
Humanities and Art	October 14, 1983	Alan	Joe	10	1
Computer Science	September 26, 1980	Martin	Genberg	17	2
Graphic Design	June 12, 1982	Athur	Watson	16	3
Graphic Design	April 5, 1987	Anabelle	Sanberg	12	4
Psychology	November 1, 1978	Kira	Schommer	11	5
Business	17 February 1981	Christian	Kiriam	10	6
Machine Learning	1 January 1984	Barbara	Ballard	14	7
Deep Learning	January 13, 1978	John	null	10	8

Machine Learning	26 December 1989	Marcus	Carson	15	9
Physics	30 December 1987	Marta	Brooks	11	10
Data Analytics	June 12, 1975	Holly	Schwartz	12	11
Computer Science	July 2, 1985	April	Black	11	12
Computer Science	July 22, 1980	Irene	Bradley	13	13
Psychology	7 February 1986	Mark	Weber	12	14
Informatics	May 18, 1987	Rosie	Norman	9	15
Business	August 10, 1984	Martin	Steele	7	16
Machine Learning	16 December 1990	Colin	Martinez	9	17
Data Analytics	null	Bridget	Twain	6	18
Business	7 March 1980	Darlene	Mills	19	19
Data Analytics	June 2, 1985	Zachary	null	10	20

```
In [33]: df = df.na.fill({'dob':'unknown','last_name':'--'})
print("Replacing values in column dob and last_name by 'unknown' and '--': \n")
df.show()
```

Replacing values in column dob and last\_name by 'unknown' and '--':

course	dob	first_name	last_name	points	s_id
Humanities and Art	October 14, 1983	Alan	Joe	10	1
Computer Science	September 26, 1980	Martin	Genberg	17	2
Graphic Design	June 12, 1982	Athur	Watson	16	3
Graphic Design	April 5, 1987	Anabelle	Sanberg	12	4
Psychology	November 1, 1978	Kira	Schommer	11	5
Business	17 February 1981	Christian	Kiriam	10	6
Machine Learning	1 January 1984	Barbara	Ballard	14	7
Deep Learning	January 13, 1978	John	--	10	8
Machine Learning	26 December 1989	Marcus	Carson	15	9
Physics	30 December 1987	Marta	Brooks	11	10
Data Analytics	June 12, 1975	Holly	Schwartz	12	11
Computer Science	July 2, 1985	April	Black	11	12
Computer Science	July 22, 1980	Irene	Bradley	13	13
Psychology	7 February 1986	Mark	Weber	12	14
Informatics	May 18, 1987	Rosie	Norman	9	15
Business	August 10, 1984	Martin	Steele	7	16
Machine Learning	16 December 1990	Colin	Martinez	9	17
Data Analytics	unknown	Bridget	Twain	6	18
Business	7 March 1980	Darlene	Mills	19	19
Data Analytics	June 2, 1985	Zachary	--	10	20

```

In [0]: from pyspark.sql.functions import mean,col
        avg = df.select(mean(col('points')).alias('mean')).collect()
        df = df.na.fill({'dob':'January 20, 1995','last_name':'--'})

In [21]: from dateutil import parser
        import datetime
        from pyspark.sql.types import TimestampType,DateType
        from pyspark.sql.functions import UserDefinedFunction,col,date_format
        udf = UserDefinedFunction(lambda x:parser.parse(x), TimestampType())
        ts_df = df.withColumn("dob_timestamp",udf(df.dob))
        func = UserDefinedFunction(lambda x: datetime.datetime
                                   .strptime(str(x), '%Y-%m-%d %H:%M:%S'),
                                   TimestampType())

        df_upd = ts_df.withColumn('dob', date_format(func(col('dob_timestamp')),
                                                    'dd-MM-yyyy'))

        dd = df_upd.drop('dob_timestamp')
        print("Dates changed to 'DD-MM-YYYY' format: \n")
        dd.show()

```

Dates changed to 'DD-MM-YYYY' format:

```

+-----+-----+-----+-----+-----+-----+
|      course|      dob|first_name|last_name|points|s_id|
+-----+-----+-----+-----+-----+-----+
|Humanities and Art|14-10-1983|      Alan|      Joe|      10|      1|
|  Computer Science|26-09-1980|    Martin|   Genberg|      17|      2|
|   Graphic Design|12-06-1982|    Athur|   Watson|      16|      3|
|   Graphic Design|05-04-1987|  Anabelle|   Sanberg|      12|      4|
|      Psychology|01-11-1978|      Kira|  Schommer|      11|      5|
|      Business|17-02-1981|  Christian|   Kiriam|      10|      6|
|  Machine Learning|01-01-1984|  Barbara|   Ballard|      14|      7|
|    Deep Learning|13-01-1978|     John|      --|      10|      8|
|  Machine Learning|26-12-1989|   Marcus|   Carson|      15|      9|
|      Physics|30-12-1987|     Marta|   Brooks|      11|     10|
|   Data Analytics|12-06-1975|     Holly|  Schwartz|      12|     11|
|  Computer Science|02-07-1985|     April|    Black|      11|     12|
|  Computer Science|22-07-1980|     Irene|  Bradley|      13|     13|
|      Psychology|07-02-1986|      Mark|    Weber|      12|     14|
|    Informatics|18-05-1987|     Rosie|   Norman|       9|     15|
|      Business|10-08-1984|   Martin|   Steele|       7|     16|
|  Machine Learning|16-12-1990|     Colin|  Martinez|       9|     17|
|   Data Analytics|20-01-1995|  Bridget|    Twain|       6|     18|
|      Business|07-03-1980|  Darlene|    Mills|      19|     19|
|   Data Analytics|02-06-1985|  Zachary|      --|      10|     20|
+-----+-----+-----+-----+-----+-----+

```

```
In [22]: from pyspark.sql.functions import lit,year
df_upd = df_upd.withColumn('age',2019 - year(col('dob_timestamp')))
df_upd = df_upd.drop('dob_timestamp')
print("Updated records with the current age of students: \n")
df_upd.show()
```

Updated records with the current age of students:

course	dob	first_name	last_name	points	s_id	age
Humanities and Art	14-10-1983	Alan	Joe	10	1	36
Computer Science	26-09-1980	Martin	Genberg	17	2	39
Graphic Design	12-06-1982	Athur	Watson	16	3	37
Graphic Design	05-04-1987	Anabelle	Sanberg	12	4	32
Psychology	01-11-1978	Kira	Schommer	11	5	41
Business	17-02-1981	Christian	Kiriam	10	6	38
Machine Learning	01-01-1984	Barbara	Ballard	14	7	35
Deep Learning	13-01-1978	John	--	10	8	41
Machine Learning	26-12-1989	Marcus	Carson	15	9	30
Physics	30-12-1987	Marta	Brooks	11	10	32
Data Analytics	12-06-1975	Holly	Schwartz	12	11	44
Computer Science	02-07-1985	April	Black	11	12	34
Computer Science	22-07-1980	Irene	Bradley	13	13	39
Psychology	07-02-1986	Mark	Weber	12	14	33
Informatics	18-05-1987	Rosie	Norman	9	15	32
Business	10-08-1984	Martin	Steele	7	16	35
Machine Learning	16-12-1990	Colin	Martinez	9	17	29
Data Analytics	20-01-1995	Bridget	Twain	6	18	24
Business	07-03-1980	Darlene	Mills	19	19	39
Data Analytics	02-06-1985	Zachary	--	10	20	34

```
sd = df.select(std(col('points')).alias('std')).collect()
sd = sd[0]['std']
df_pnt = df_upd.withColumn('points',when(df_upd.points >= sd+avg[0]['mean'],20)
                           .otherwise(df_upd.points))
print("Updated points using one standard deviation: \n")
```

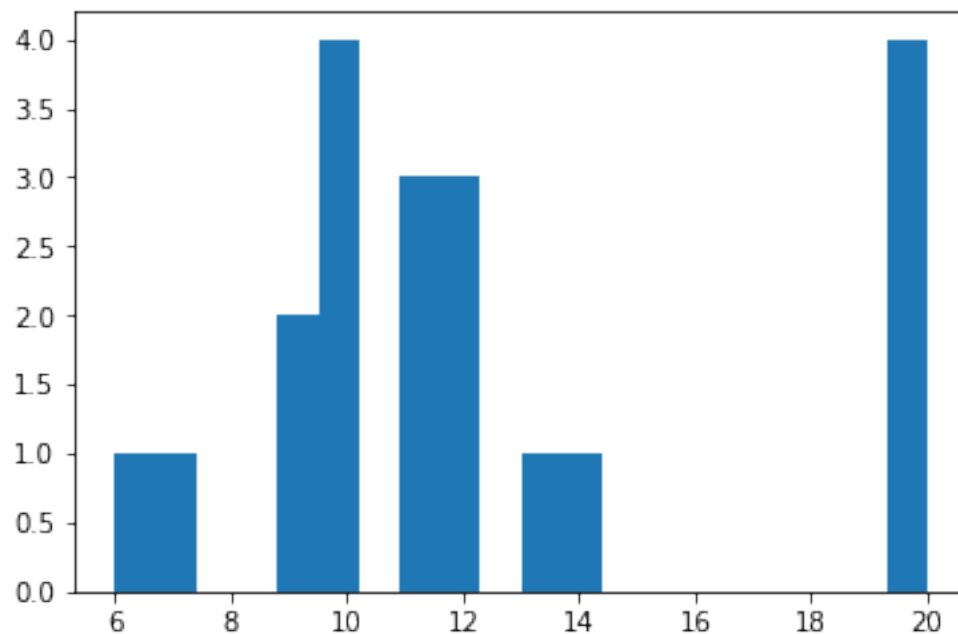
Updated points using one standard deviation:

course	dob	first_name	last_name	points	s_id	age
Humanities and Art	14-10-1983	Alan	Joe	10	1	36
Computer Science	26-09-1980	Martin	Genberg	20	2	39

	Graphic Design 12-06-1982	Athur	Watson	20	3	37
	Graphic Design 05-04-1987	Anabelle	Sanberg	12	4	32
	Psychology 01-11-1978	Kira	Schommer	11	5	41
	Business 17-02-1981	Christian	Kiriam	10	6	38
	Machine Learning 01-01-1984	Barbara	Ballard	14	7	35
	Deep Learning 13-01-1978	John	--	10	8	41
	Machine Learning 26-12-1989	Marcus	Carson	20	9	30
	Physics 30-12-1987	Marta	Brooks	11	10	32
	Data Analytics 12-06-1975	Holly	Schwartz	12	11	44
	Computer Science 02-07-1985	April	Black	11	12	34
	Computer Science 22-07-1980	Irene	Bradley	13	13	39
	Psychology 07-02-1986	Mark	Weber	12	14	33
	Informatics 18-05-1987	Rosie	Norman	9	15	32
	Business 10-08-1984	Martin	Steele	7	16	35
	Machine Learning 16-12-1990	Colin	Martinez	9	17	29
	Data Analytics 20-01-1995	Bridget	Twain	6	18	24
	Business 07-03-1980	Darlene	Mills	20	19	39
	Data Analytics 02-06-1985	Zachary	--	10	20	34
+-----+-----+-----+-----+-----+-----+						

```
In [27]: import matplotlib.pyplot as plt
pt = df_pnt.toPandas()['points']
plt.hist(pt,bins=20)
plt.show()
```

Histogram of new points:





## 0.1.5 Exercise 2

```
In [0]: #import libraries
        from pyspark import SparkContext
        from pyspark.sql import SQLContext
        import pandas as pd
        from pyspark.sql import Row
        import numpy as np
        from pyspark.sql.functions import *
        from pyspark.sql.types import IntegerType
        from pyspark.sql import Window
        from pyspark.sql.functions import mean as avg, stddev as stdd
```

In order to get the sessions of the user, the lags are found from which the difference in active time of the user is calculated. This is then checked if the user has exceeded 30 minutes are not. If yes, it's considered as timed-out session and if no, it's an active session

```
In [42]: dat_file = "gdrive/My Drive/DDA/Spark1/tags.dat"

        df = sqlContext.read.option("delimiter",":").csv(dat_file)
        df = df.selectExpr("_c0 as UserID", "_c2 as MovieID", "_c4 as Tags",
                           "_c6 as Timestamp")
        df = df.withColumn("UserID", df["UserID"].cast(IntegerType()))
        df = df.withColumn("MovieID", df["MovieID"].cast(IntegerType()))
        ts_w = Window.partitionBy("UserID").orderBy(asc("Timestamp"))
        df = df.withColumn('lag', lag(df.Timestamp).over(ts_w))
        df = df.withColumn('difference', when((df.Timestamp - df.lag)/60 < 30, 1)
                                         .otherwise(0))
        df = df.withColumn('session', sum('difference').over(ts_w))
        df = df.drop('lag', 'difference')
        print("Tagging session for each user: \n")
        df.show()

        print("*****\n\n")

        df = df.withColumn('lag', lag(df.Timestamp).over(ts_w))
        df = df.withColumn('difference', when((df.Timestamp - df.lag) > 30*60, 1)
                                         .otherwise(0))
        df = df.withColumn('session', sum('difference').over(ts_w))
        df = df.drop('lag', 'difference')

        max_freq = df.groupBy("UserID").max("session")
        max_freq = max_freq.orderBy('max(session)', ascending=False)
        print("Frequency of tagging: \n")
        max_freq.show()
```

```

print("*****\n\n")

m = df.groupBy("UserID").mean("session").orderBy('avg(session)',
                                                    ascending=False)

print("Mean and Standard deviation of the tagging frequency of\n each user: \n")
sd = df.groupBy("UserID").agg(stddev("session"))

msd = m.join(sd,"UserID",how='right_outer').orderBy('avg(session)',
                                                    ascending=False)

msd.show()

print("*****\n\n")

mean = df.agg(avg("session")).collect()[0]['avg(session)']
std = df.agg(stddev("session")).collect()[0]['stddev_samp(session)']
print("Mean and Standard deviation of the tagging frequency\n across users: \n")
print("Mean: ",mean)
print("Standard Deviation: ",std)

print("*****\n\n")
print("List of users with a mean tagging frequency within two\n standard \
deviation from the mean frequency for across users: \n")
m = m.withColumn("Flag",when(m['avg(session)'] < 2*std, 1).otherwise(0))
users_list = m.filter(m.Flag == 1).select('UserID').distinct()
users_list.show()

```

Tagging session for each user:

UserID	MovieID	Tags	Timestamp	session
6658	2712	unwatchable	1140486822	0
6658	288	annoying	1140486947	1
10817	158	Christina Ricci	1218451667	0
10817	3826	Kevin Bacon	1218452067	1
10817	3826	elizabeth shue	1218452092	2
10817	7451	Lindsay Lohan	1218452569	3
10817	1367	glen close	1218466235	3
12046	1610	cold war	1222049475	0
12046	1222	Vietnam War	1222049571	1
12046	750	dark comedy	1226230439	1
12046	750	Stanley Kubrick	1226230442	2
12046	750	cold war	1226230454	3
12046	750	satire	1226230466	4
12046	778	imdb top 250	1226230555	5
12046	778	black comedy	1226230582	6
12046	48774	end of the world	1226230975	7

12046	48774	war 1226230978	8
12046	48774	imdb top 250 1226230981	9
12046	48774	dystopia 1226230983	10
12046	48774	apocalypse 1226230992	11

+-----+-----+-----+-----+-----+

only showing top 20 rows

\*\*\*\*\*

Frequency of tagging:

+-----+-----+

UserID	max(session)
--------	--------------

+-----+-----+

10555	884
23172	476
146	332
33384	243
47448	198
34745	143
11898	126
30167	114
64633	107
8041	103
41838	99
6362	94
23388	84
18015	77
23032	72
49882	72
59092	71
50970	70
2643	68
32828	64

+-----+-----+

only showing top 20 rows

\*\*\*\*\*

Mean and Standard deviation of the tagging frequency of each user:

+-----+-----+-----+

UserID	avg(session)	stddev_samp(session)
--------	--------------	----------------------

+-----+-----+-----+

10555	520.4491017964071	226.60815651786262
-------	-------------------	--------------------

23172	233.75731284085276	143.7536630775717
146	127.94781553398059	88.3217936310053
47448	90.14512195121951	66.04907830853023
11898	61.71298174442191	39.88577311890879
33384	53.68281938325991	71.10460757488028
34745	52.15585443037975	42.93201365252578
64633	50.52549575070822	34.70998468723264
41838	42.367198838896954	31.872750688848402
6362	42.28125	25.116150339042868
23388	37.94854586129754	26.888318827426215
50970	33.81666666666667	20.285826861639716
8041	33.44179104477612	30.78448433585449
32828	28.953929539295395	12.565504107840338
48621	28.08076923076923	16.96206349067229
19460	27.875	15.358259215357908
49882	27.476462196861625	16.51963600542648
24221	26.5472972972973	13.190049901325256
39689	26.10236220472441	14.808746153191747
69388	25.571428571428573	16.550370366667796

+-----+-----+-----+-----+

only showing top 20 rows

\*\*\*\*\*

Mean and Standard deviation of the tagging frequency  
across users:

Mean: 56.551276417660596

Standard Deviation: 146.6106950491872

\*\*\*\*\*

List of users with a mean tagging frequency within two  
standard deviation from the mean frequency for across users:

+-----+

|UserID|

+-----+

| 43527|

| 18979|

| 24171|

| 12046|

| 36538|

| 53565|

| 65867|

| 57380|

| 10817|

```
| 6658|
| 14570|
| 15846|
| 16574|
| 25462|
| 26583|
| 32445|
| 41946|
| 47711|
| 49308|
| 51123|
+-----+
only showing top 20 rows
```

## 0.1.6 Bonus

```
In [0]: movies_dat_file = "gdrive/My Drive/DDA/Spark1/movies.dat"

df_movie = sqlContext.read.option("delimiter",":").csv(movies_dat_file)
df_movie= df_movie.selectExpr("_c0 as MovieID","_c2 as Title","_c4 as Genre")
df_movie.show()

ratings_dat_file = "gdrive/My Drive/DDA/Spark1/ratings.dat"

df_ratings = sqlContext.read.option("delimiter",":").csv(ratings_dat_file)
df_ratings = df_ratings.selectExpr("_c0 as UserID","_c2 as MovieID",
                                   "_c4 as Rating","_c6 as Timestamp")

df_ratings.show()
```

```
+-----+-----+-----+
|MovieID|          Title|          Genre|
+-----+-----+-----+
|      1| Toy Story (1995)|Animation|Childre...| |
|      2| Jumanji (1995)|Adventure|Childre...|
|      3|Grumpier Old Men ...|      Comedy|Romance|
|      4|Waiting to Exhale...|      Comedy|Drama|
|      5|Father of the Bri...|      Comedy|
|      6| Heat (1995)|Action|Crime|Thri...|
|      7| Sabrina (1995)|      Comedy|Romance|
|      8| Tom and Huck (1995)|Adventure|Children's|
|      9| Sudden Death (1995)|      Action|
|     10| GoldenEye (1995)|Action|Adventure|...|
|     11|American Presiden...|Comedy|Drama|Romance|
|     12| Dracula|      null|
|     13| Balto (1995)|Animation|Children's|
|     14| Nixon (1995)|      Drama|
```

15	Cutthroat Island ...	Action Adventure ...
16	Casino (1995)	Drama Thriller
17	Sense and Sensibi...	Drama Romance
18	Four Rooms (1995)	Thriller
19	Ace Ventura	null
20	Money Train (1995)	Action

```
+-----+-----+-----+-----+
```

only showing top 20 rows

```
+-----+-----+-----+-----+
```

UserID	MovieID	Rating	Timestamp
--------	---------	--------	-----------

```
+-----+-----+-----+-----+
```

1	1193	5	978300760
1	661	3	978302109
1	914	3	978301968
1	3408	4	978300275
1	2355	5	978824291
1	1197	3	978302268
1	1287	5	978302039
1	2804	5	978300719
1	594	4	978302268
1	919	4	978301368
1	595	5	978824268
1	938	4	978301752
1	2398	4	978302281
1	2918	4	978302124
1	1035	5	978301753
1	2791	4	978302188
1	2687	3	978824268
1	2018	4	978301777
1	3105	5	978301713
1	2797	4	978302039

```
+-----+-----+-----+-----+
```

only showing top 20 rows

In [0]: *#Merging both the dataframes*

```
merged = df_movie.join(df_ratings, 'MovieID', 'inner')
```

```
merged.show()
```

```
+-----+-----+-----+-----+-----+-----+
```

MovieID	Title	Genre	UserID	Rating	Timestamp
---------	-------	-------	--------	--------	-----------

```
+-----+-----+-----+-----+-----+-----+
```

1193	One Flew Over the...	Drama	1	5	978300760
661	James and the Gia...	Animation Childre...	1	3	978302109
914	My Fair Lady (1964)	Musical Romance	1	3	978301968
3408	Erin Brockovich (...)	Drama	1	4	978300275

2355	Bug's Life, A (1998)	Animation Childre...	1	5	978824291
1197	Princess Bride, T...	Action Adventure ...	1	3	978302268
1287	Ben-Hur (1959)	Action Adventure ...	1	5	978302039
2804	Christmas Story, ...	Comedy Drama	1	5	978300719
594	Snow White and th...	Animation Childre...	1	4	978302268
919	Wizard of Oz, The...	Adventure Childre...	1	4	978301368
595	Beauty and the Be...	Animation Childre...	1	5	978824268
938	Gigi (1958)	Musical	1	4	978301752
2398	Miracle on 34th S...	Drama	1	4	978302281
2918	Ferris Bueller's ...	Comedy	1	4	978302124
1035	Sound of Music, T...	Musical	1	5	978301753
2791	Airplane! (1980)	Comedy	1	4	978302188
2687	Tarzan (1999)	Animation Children's	1	3	978824268
2018	Bambi (1942)	Animation Children's	1	4	978301777
3105	Awakenings (1990)	Drama	1	5	978301713
2797	Big (1988)	Comedy Fantasy	1	4	978302039

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only showing top 20 rows

```
In [0]: columns_to_drop = ['MovieID', 'Genre', 'UserID', 'Timestamp']
        test = merged.drop(*columns_to_drop)
        test.show()
```

Title	Rating
One Flew Over the...	5
James and the Gia...	3
My Fair Lady (1964)	3
Erin Brockovich (...)	4
Bug's Life, A (1998)	5
Princess Bride, T...	3
Ben-Hur (1959)	5
Christmas Story, ...	5
Snow White and th...	4
Wizard of Oz, The...	4
Beauty and the Be...	5
Gigi (1958)	4
Miracle on 34th S...	4
Ferris Bueller's ...	4
Sound of Music, T...	5
Airplane! (1980)	4
Tarzan (1999)	3
Bambi (1942)	4
Awakenings (1990)	5
Big (1988)	4

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only showing top 20 rows

```
In [0]: test = test.withColumn("Rating", test["Rating"].cast(IntegerType()))
a = test.rdd.groupByKey().mapValues(lambda x: sum(x) / len(x))
c = sqlContext.createDataFrame(a).orderBy('_2',ascending = False)
c = c.filter(c._2 == 5.0)
c = c.selectExpr("_1 as MovieID", "_2 as Avg_Rating")
print("Movies with maximum average rating: \n")
c.show()
```

Movies with maximum average rating:

```
+-----+-----+
|          MovieID|Avg_Rating|
+-----+-----+
|Gate of Heavenly ...|      5.0|
|Smashing Time (1967)|      5.0|
|      Lured (1947)|      5.0|
|One Little Indian...|      5.0|
|      Baby, The (1973)|      5.0|
|Schlafes Bruder (...|      5.0|
|Follow the Bitch ...|      5.0|
|Bittersweet Motel...|      5.0|
|Ulysses (Ulisse) ...|      5.0|
|Song of Freedom (...|      5.0|
+-----+-----+
```

```
In [0]: columns_to_drop = ['MovieID', 'Title', 'UserID', 'Timestamp']
test = merged.drop(*columns_to_drop)
test.show()

test = test.withColumn("Rating", test["Rating"].cast(IntegerType()))
```

```
+-----+-----+
|          Genre|Rating|
+-----+-----+
|          Drama|      5| | |
|Animation|Childre...|      3|
|      Musical|Romance|      3|
|          Drama|      4|
|Animation|Childre...|      5|
|Action|Adventure|...|      3|
|Action|Adventure|...|      5|
|      Comedy|Drama|      5|
|Animation|Childre...|      4|
|Adventure|Childre...|      4|
|Animation|Childre...|      5|
|          Musical|      4|
```



	Drama	4
	Comedy	4
	Musical	5
	Comedy	4
Animation Children's		3
Animation Children's		4
	Drama	5
	Comedy Fantasy	4

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only showing top 20 rows

```
In [0]: a = test.rdd.groupByKey().mapValues(lambda x: sum(x) / len(x))
        c = sqlContext.createDataFrame(a).orderBy('_2',ascending = False)
        c = c.selectExpr("_1 as Genre", "_2 as Avg_Rating")
        c.show()
        maxx = c.rdd.max(key=lambda x:x[1])
        print("Genre with maximum average rating: \n",maxx)
```

Genre	Avg_Rating
-------	------------

+-----+

Animation Comedy ...	4.473837209302325
Film-Noir Mystery	4.367424242424242
Adventure War	4.34610705596107
Film-Noir Romance...	4.29438202247191
Film-Noir Sci-Fi	4.273333333333333
Crime Film-Noir	4.264129181084199
Film-Noir	4.258104738154613
Action Adventure ...	4.251655629139073
Adventure Childre...	4.247962747380675
Drama Film-Noir	4.218152866242038
Film-Noir Thriller	4.206757438224912
Crime Film-Noir M...	4.2020547945205475
Comedy Mystery Ro...	4.184158415841584
Comedy Drama Musical	4.179785330948121
Comedy Mystery Th...	4.168154761904762
Action Crime Drama	4.151277918489523
Action Adventure ...	4.147826086956521
Comedy Drama Western	4.141263940520446
Crime Film-Noir M...	4.126734158230221
Action Sci-Fi Thr...	4.125824175824176

+-----+

only showing top 20 rows

Genre with maximum average rating:

Row(Genre='Animation|Comedy|Thriller', Avg\_Rating=4.473837209302325)

```
In [0]: u_rat = merged.groupby('UserID').agg(countDistinct("MovieID"))
u_rat = u_rat.filter(u_rat['count(DISTINCT MovieID)'] > 40)
merged = merged.join(u_rat,"UserID","inner")
columns_to_drop = ['MovieID', 'Title', 'Genre', 'count(DISTINCT MovieID)',
                    'Timestamp']
test = merged.drop(*columns_to_drop)
test.show()
```

```
+-----+-----+
|UserID|Rating|
+-----+-----+
| 1090|    3|
| 1090|    3|
| 1090|    4|
| 1090|    3|
| 1090|    4|
| 1090|    4|
| 1090|    3|
| 1090|    3|
| 1090|    4|
| 1090|    3|
| 1090|    3|
| 1090|    3|
| 1090|    4|
| 1090|    2|
| 1090|    4|
| 1090|    2|
| 1090|    3|
| 1090|    4|
| 1090|    3|
| 1090|    3|
```

```
+-----+-----+
only showing top 20 rows
```

```
In [0]: test = test.withColumn("Rating", test["Rating"].cast(IntegerType()))
a = test.rdd.groupByKey().mapValues(lambda x: sum(x) / len(x))
c = sqlContext.createDataFrame(a).orderBy('_2',ascending = True)
c = c.selectExpr("_1 as UserID", "_2 as Avg_Rating")
c.show()
minn = c.rdd.min(key=lambda x:x[1])
print("User with minimum average rating: \n",minn)
```

```
+-----+-----+
|UserID|    Avg_Rating|
+-----+-----+
| 3598|1.0153846153846153|
```

	4486	1.0588235294117647
	2744	1.3043478260869565
	4539	1.815126050420168
	5850	1.8448275862068966
	5334	1.9272727272727272
	5686	2.0452830188679245
	3209	2.0608695652173914
	1608	2.0833333333333335
	4575	2.088
	4916	2.088235294117647
	1747	2.138888888888889
	1761	2.15929203539823
	1340	2.1627329192546583
	163	2.1828793774319064
	1100	2.1988188976377954
	5039	2.202777777777778
	2106	2.245555555555557
	1630	2.264957264957265
	203	2.2913385826771653

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only showing top 20 rows

User with minimum average rating:

Row(UserID='3598', Avg\_Rating=1.0153846153846153)