Use Case 2

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Introduction

The spark-shell was used to complete the following tasks.

Spark Actions & Transformations

Load the matches.csv and deliveries.csv into rdd and rdd2, respectively.

```
scala> val rddFromFile = sc.textFile("/home/gabriel/Data/matches.csv")
scala> val rdd = rddFromFile.map(f=>{ f.split(",") })
scala> val rddFromFile2 = sc.textFile("/home/gabriel/Data/deliveries.csv")
scala> val rdd2 = rddFromFile2.map(f => {f.split(",")})
```

1. Find the number of matches that took place in Hyderabad.

```
scala> rdd.
l filter(a => a(2).equals("Hyderabad")).
l count()
res30: Long = 64
```

2. Find the number matches that Sunrisers won in Hyderabad City.

```
scala> rdd.
| filter(a => a(2).equals("Hyderabad") &&
| a(10).equals("Sunrisers Hyderabad")).
| count()
res31: Long = 30
```

3. Find the number of matches Sunrisers won in each season.

```
scala> rdd.
       | filter(a => a(10).equals("Sunrisers Hyderabad")).
2
       | groupBy(a => a(1)).
       | map{ case(k,v) => (k, v.size) }.
4
5
       | collect().
6
       | sortBy(_._1).
       | foreach(println)
8 (2013,10)
9 (2014,6)
10 (2015,7)
11 (2016,11)
12 (2017,8)
13 (2018,10)
14 (2019,6)
```

4. In which season did Yuvraj Singh has been awarded maximum Player of the match title.

```
scala> rdd.

| filter(a => a(13).equals("Yuvraj Singh")).
| foreach( a => println(a(1)))

2017
2014
2009
2009
2011
```

5. Which venue hosted maximum matches.

6. Find the average win by runs in Eden Gardens in Season 2017.

```
scala> rdd.

| filter(a => a(14).equals("Eden Gardens")).

| filter(a => a(1).equals("2017")).

| map(_(11).toInt).

| mean()

res0: Double = 15.428571428571429
```

7. Find out the percentage of each dismissal over all seasons.

```
1 // Select match id and player_dismissed columns from the RDD
2 scala> val id_dismissed = rdd2.map( f => { (f(0), if(f.size < 19) "-" else f</pre>
       (18) ) } )
4 // Select match id and the season columns from the RDD
5 \text{ scala} > \text{val} \text{ id\_season} = \text{rdd.map}(f \Rightarrow \{(f(0), f(1))\})
_{7} // Join the two lists of tuples on the match id
8 scala > val joinedRdd = id_dismissed.join(id_season)
_{
m 10} // Restructure the data - put the season as the first element in the row
scala > val restruct = joinedRdd.map{ case(a,(b,c)) => (c,b,a) }.collect()
_{13} // We will use a two mutable maps: dismiss and totals
14 scala > import scala.collection.mutable.Map
_{16} // This maps season year to number of players dismissed in that season
scala> var dismiss= Map[String, Int]()
19 // This maps season year to number of deliveries
20 scala> var totals = Map[String, Int]()
22 // Count number of players dismissed per season and
23 // also count the total number of deliveries per season
24 scala> for((a,b,c) <- restruct){</pre>
      if(!dismiss.contains(a)){ dismiss(a) = 0 }
       | if(b != "-"){ dismiss(a) += 1 }
       if(!totals.contains(a)){ totals(a) = 0 }
27
          totals(a) += 1
       | }
29
31 // print the results
32 scala > for((k,v) <- totals){</pre>
33
       println("Season: "+k+"\tDismissal Rate: "+dismiss(k).toFloat /
       totals(k).toFloat)
   | }
```

Output:

```
1 Season: 2016 Dismissal Rate: 0.047247447
2 Season: 2010 Dismissal Rate: 0.050006896
3 Season: 2019 Dismissal Rate: 0.04725014
4 Season: 2013 Dismissal Rate: 0.050173298
5 Season: 2009 Dismissal Rate: 0.0513009
6 Season: 2015 Dismissal Rate: 0.050615296
7 Season: 2018 Dismissal Rate: 0.050314907
8 Season: 2012 Dismissal Rate: 0.048291776
9 Season: 2011 Dismissal Rate: 0.047786985
10 Season: 2014 Dismissal Rate: 0.04713287
11 Season: 2008 Dismissal Rate: 0.05115279
12 Season: 2017 Dismissal Rate: 0.051291298
```

8. Which stadium is best suited to bat first.

```
scala> rdd.
| map(x => (x(14), if(x(11) != "win_by_runs") x(11).toInt else 0)).
| reduceByKey(_ + _).
| sortBy(-1 * _._2).
| take(1).
| foreach(println)
| (M Chinnaswamy Stadium, 1310)
```

Checking our answers with the DataFrame API

This section completes the tasks from the previous section again, except this time using the DataFrame API. The purpose of this is to double-check the answers from the previous section.

Load the matches.csv and deliveries.csv into dataframes df and df2, respectively.

```
scala> val df = spark.read.option("header",true).csv("/home/gabriel/Downloads
    /matches.csv")

scala> val df2 = spark.read.option("header",true).csv("/home/gabriel/
    Downloads/deliveries.csv")
```

1. Find the number of matches that took place in Hyderabad.

2. Find the number matches that Sunrisers won in Hyderabad City.

```
scala> df.
| where(col("winner") === "Sunrisers Hyderabad").
| where(col("city") === "Hyderabad").
| groupBy("winner").
| count().
| show()
| winner|count|
| Usunrisers Hyderabad| 30|
| the count of the count
```

3. Find the number of matches Sunrisers won in each season.

```
scala> df.
       | where(col("winner") === "Sunrisers Hyderabad").
       | groupBy("season").
      | count().
4
5
      | show()
6 +----+
7 |season|count|
8 +----+
9 | 2016| 11|
           6 |
8 |
10 | 2019|
11 | 2017|
12 | 2014|
13 | 2013 | 10 |
14 | 2018 | 10 |
15 | 2015 | 7 |
16 +----+
```

4. In which season did Yuvraj Singh has been awarded maximum Player of the match title.

```
scala> df.
| select("season").
| where(col("player_of_match") === "Yuvraj Singh").
| show()
|-----+
| season|
| +----+
| 2017|
| 2009|
| 2009|
| 2009|
| 2011|
| 2011|
| 2014|
| 3 +----+
```

5. Which venue hosted maximum matches.

```
scala> df.
groupBy("venue").
locunt().
lorderBy(desc("count")).
lshow(1)

+-----+
lvenue|count|
+-----+
leden Gardens| 77|
only showing top 1 row
```

6. Find the average win by runs in Eden Gardens in Season 2017.

7. Find out the percentage of each dismissal over all seasons.

```
scala> df2.
       | join(df, df2("match_id") === df("id"), "left").
       | groupBy("season").
       | agg((sum(when($"player_dismissed".isNull,0).otherwise(1))/count("*")).
      | as("player_dismissed : fraction null")).
      | show()
6
7 +----
8 |season|player_dismissed : fraction null|
9 +----+
10 | 2016|
                     0.04724744608399546|
                     0.0482917768897394|
11 | 2012|
12 | 2019|
                       0.047250139586823|
                   0.051291299956716205|
0.047132867132867136|
13 | 2017 |
14 | 2014|
                  0.0501/02
0.051300896663236804.
0.050314905528341496|
0.04778698642214777|
15 | 2013|
16 | 2009 |
17 | 2018 |
18 | 2011|
19 | 2008|
              0.05061529446234984
20 | 2015|
21 | 2010|
                      0.05000689750310388|
22 +-----
```

8. Which stadium is best suited to bat first.

```
scala> df.
| groupBy("venue").
| agg(sum("win_by_runs").as("win_by_runs_by_venue")).
| orderBy(desc("win_by_runs_by_venue")).
| show(1)
| venue|win_by_runs_by_venue|
| venue|win_by_runs_by_venue|
| the control of the control o
```