

IDENTIFICATION OF TOP -10
PLATFORM AND GENRE FOR
GAMES USING HADOOP

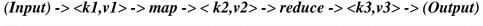
ABSTRACT

This report presents a project that utilizes Hadoop MapReduce chaining to identify the top 10 video game platform and genre combinations based on sales data. By leveraging Hadoop's MapReduce framework, the project aims to showcase its potential in handling large-scale datasets and emphasize the significance of data analysis in the gaming industry. The project involves mapping video game sales records to identify platform and genre attributes, followed by aggregating the sales data for each combination. A second MapReduce chain is employed to sort the aggregated data and determine the top 10 combinations, providing valuable insights for game developers, marketers, and investors. The project highlights Hadoop's scalability, flexibility, and its role in facilitating datadriven decision-making in the gaming industry.

Tanvi Mathur
Data Science & Analytics Student

Introduction

Map Reduce is a Hadoop data analysis technique that allows data to expand and be managed across several computer nodes. It filters and turns input into a value that the reducer may aggregate overusing basic units called as mappers and reducers. Lists and key/value pairs are the basics of map reduction, and they serve as constraints for the map reduce method. The programme takes a list of key-value pairs as input (). Following the scanning of the input file, mapper's map function divides each line into a key value pair. The output is added to a long list of pairs, which are then joined to form a new pair. With this design, the reducer evaluates each component individually. The MapReduce framework's primary features are its simplicity, scalability, and extensibility, which allow it to process huge amounts of information (Lam, 2011).



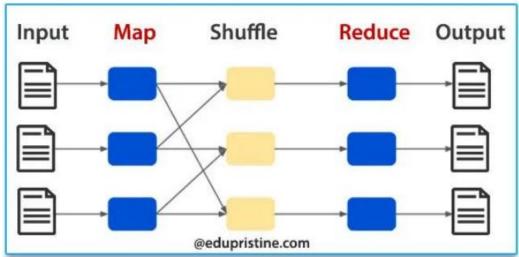


Figure 2: MapReduce Steps

Problem Statement and Data:

Firstly, the problem statement for this report is "What are the top 10 genre console combination which are popular among the buyers based on total sales?", the objective of this problem statement is to analyse the global sales for each genre of video games across different gaming consoles. The analysis will help us understand which genre-console combinations are most popular among buyers. This information can help game developers to focus on popular combinations, improve game quality and drive sales revenue. The dataset used for the analysis was obtained from the Kaggle website which comprises of 13764 rows and 14 columns. The collection of data is widely ranged from the year 2000 to 2020. In the below table, each parameter is extensively described in the data lexicon.

Parameters	Synopsis	Data Type
Title	Name of the game	Categorical
Console	Platform of the game	Categorical
Genre	Genre of the game	Categorical
Publisher	Publisher of the game	Categorical
Vg_score	The vgchartz critical score (out of 10)	Float
Critic_score	The metacritic critical score (out of 10)	Float
User_score	The user critical score (out of 10)	Float
Total_shipped	The total number shipped in millions	Float
Asian_sales	Asian region sales in millions	Float
North_American_sales	North American region sales in millions	Float
Japan_sales	Japan region sales in millions	Float
European_sales	European region sales in millions	Float
Global_sales	Worldwide sales in millions	Float
Release_year	Year game was released	Numeric

Table1: Description of video game sales data

Design & Implementation:

For this project, we will create a script in Eclipse using the programming language 'Java,' then export it to see what output is created when such code is executed on the dataset. These outputs will then serve as the foundation for the report's findings section. This problem statement can be answered using the **MapReduce chaining**, to obtain the top 10 genre console combinations with highest sales which will help developers identify the popular gaming combination which they should consider while creating new games. The flow of the solution is such that we are going to run two MapReduce jobs in sequence, thus the output of first job will be an input for the second job.

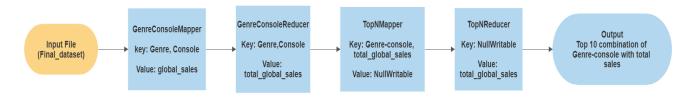


Figure 3: Flow diagram

Let go ahead and have a look at the different files we created in Eclipse to run the MapReduce flow and achieve the desired outcome.

There are a total of 5 java classes in our MapReduce project namely:

- 1. GenresConsoleMapper The first Mapper class
- 2. GenresConsoleReducer The first Reducer class
- 3. TopNMapper– The second Mapper class
- 4. TopNReducer–The second Reducer class
- 5. TopGenresConsoleSales The main driver class
- 6. Final_dataset The input text file

Mapper 1: GenreConsoleMapper

```
☐ GenresConsoleMapper.java 
☐

  package org.myorg;
      //Importing the libraries that contain the classes and methods needed in the Mapper class
  40 import java.io.IOException;
5 import org.apache.hadoop.io.*;
6 import org.apache.hadoop.mapreduce.*;
  8 /*Creating a subclass (child class) called GenresConsoleMapper from a parent class called Mapper using the keyword extends */
 11 public class GenresConsoleMapper extends Mapper<LongWritable, Text, Text, DoubleWritable>
                                                                    //<Input Key & Value datatype, Output Key & Value datatype>
 14 //Melissa K.and the Heart of Gold Collectors Edition,PC,Adventure,SDP Games,9,10,2,0.18,0.95,4.45,1.13,6.66,13.19,1971
          @Override /*@override is used to change the original behavior of the method "map" in the parent Class "Mapper" */
           // the map method takes three parameters that are a key, a value, and a context as input parameters
           public void map(LongWritable key, Text value, Context context) throws IOException, InterruptedException {
          /*create a variable line of data type String, convert value into string and store it in line and
   Split the input text line using a comma as separator and store the result in the 'line' array.*/
   String[] line = value.toString().split(","); /*Melissa K.and the Heart of Gold Collectors Edition,PC,Adventure,SDP Games,
   9,10,2,0.18,0.95,4.45,1.13,6.66,13.19,1971*/
           /*Concatenate the 2nd(genre) and 3rd element(console) of the 'line' array using a hyphen (-) as a separator by
and store the result in the 'genreConsole' variable.*/
String genreConsole = line[2] + "-" + line[1]; //Adventure-PC
      // Parse the 12th element(global sales) of the 'line' array as a double and store the result in the 'sales' variable.
                 double sales = Double.parseDouble(line[12]); //13.19
      /*Use context that was defined as a parameter in the map method above to emit new key-value pair
as output that is 'genreConsole' variable as key and 'sales' as value. */
    context.write(new Text(genreConsole), new DoubleWritable(sales)); //Adventure-PC 13.19
```

Figure 5: GenreConsoleMapper

The First mapper deal with the input from the text file it will read the input as:

```
// Melissa K.and the Heart of Gold Collectors Edition, PC, Adventure, SDP Games, 9, 10, 2, 0.18, 0.95, 4.45, 1.13, 6.66, 13.19, 1971
```

In the map method, we must take a close look at our key and value. Since the analysis, we must calculate for the genre and console combination, Genre and Console will be the key. Since we are interested in finding total global sales for combination the global_sales will be value.

Line 22 splits the data and stores it in a array. Next the Line 26 here from the above data genre and console are picked and concatenated with a hyphen as shown below.

```
//Adventure-PC
```

While line 29 fetch the data which is present at 12th index in the input file that is shown below.

Finally in line 33 the data is passed as key value pair to the reducer where here the key is Adventure-PC and 13.19 is value as shown below.

```
//Adventure-PC 13.19
```

Reducer 1: GenreConsoleReducer

```
☑ GenresConsoleReducer.java 

1 package org.myorg;
       //Importing the libraries that contain the classes and methods needed in the Reducer class
   40 import java.io.10Exception;
5 import org.apache.hadoop.io.*;
6 import org.apache.hadoop.mapreduce.*;
          *Creating a subclass (child class) called GenresConsoleReducer from a parent
  own/vicating a subclass (thirt class) rated defines on softeneducer from a parent of class called Reducer using the keyword extends */
10 public class GenresConsoleReducer extends Reducer<Text, DoubleWritable, Text, DoubleWritable> {
                                                                          //<Input Key & Value datatype, Output Key & Value datatype>
            /*@override is used to change the original \underline{\text{behavior}} of the method "reduce" in the parent Class "Reducer" */
 13<sup>©</sup>
14
15<sup>©</sup>
16
17

18

19

20

21

22

23

24

25

26

27

28

29

30

31

}

32

33
                 the reduce method takes three parameters that are a key, a value, and a context as input parameters
            public void reduce(Text key, Iterable<DoubleWritable> values, Context context) throws IOException, InterruptedException {
                   //initialize the variable totalSales with 0 value.
                  double totalSales = 0;
/* Using for loop on input values, such that for each value in values, value is added to totalSales variable.*/
for (DoubleWritable value : values) {
                        totalSales += value.get();
                  /*Use context that was defined as a parameter in the map method above to emit new key-value pair to output using input key and 'totalSales' as the value.*/
                  context.write(key, new DoubleWritable(totalSales));
```

Figure 6: GenreConsoleReducer

As we go to the reducer in Line 10 key is passed as [{Adventure-PC,13.19}, {Adventure-PC,6.89},}, {Adventure-PC,11.36},},{Puzzel-DS,9.56}]. The line 22 is a for loop which based on the genre console combination value help generate the total sales. From line 28 we will get total sales for different key, for instance [{Adventure-PC 31.44}]. So overall, the first MapReduce generates the total sales for each combination of genre and console. The output file for first MapReduce is shown below.

```
41 Racing-MS
                                                               354.960000000000004
 1 Adventure-DS
                3503.5900000000033
                                          42 Racing-PC
                                                               7794.7700000000003
 2 Adventure-Linux
                         630.06000000000002
                                           43 Racing-PS
                                                               2071.42000000000014
                 229.62
 3 Adventure-MS
                                          44 Racing-PS2
                                                               2765.269999999999
                 24680.950000000002
 4 Adventure-PC
                                           45 Racing-PS3
                                                               738.1899999999999
 5 Adventure-PS
                 2900.3000000000015
                                           46 Racing-PS4
                                                               899.199999999995
 6 Adventure-PS2
                 5592.769999999999
                                                               53.83
                                           47 Racing-PS5
 7 Adventure-PS3
                 807.74000000000001
                                           48 Racing-PSN
                                                               2243.11
 8 Adventure-PS4
                 2450.2999999999984
                                                               408.8499999999999
                                           49 Shooter-DS
 9 Adventure-PS5
                 147.59
                                           50 Shooter-Linux
                                                               91.96
                 2149.17000000000005
10 Adventure-PSN
                                           51 Shooter-MS
                                                               786.5300000000001
11 Adventure-iOS
                 146.350000000000002
                                           52 Shooter-PC
                                                               15370.119999999986
12 Board Game-DS
                 45.44
                                                               1471.5399999999997
                                           53 Shooter-PS
13 Board Game-Linux
                         29.23
                                           54 Shooter-PS2
                                                               2062.40000000000005
14 Board Game-PC
                 1371.58000000000004
                                           55 Shooter-PS3
                                                               1406.4199999999987
15 Board Game-PS3 67.42999999999999
                                           56 Shooter-PS4
                                                               2149.51
16 Board Game-PS4 678.3599999999998
                                           57 Shooter-PS5
                                                               141.57
17 Board Game-PS5 166.17
                                           58 Shooter-PSN
                                                               4119.4200000000002
18 Board Game-iOS 74.32000000000001
                                                               67.83
                                           59 Shooter-iOS
19 Fighting-DS
                 349.55000000000001
                                          60 Simulation-DS
                                                               3483,119999999998
20 Fighting-Linux 22.65
                                                                        288.550000000000007
                                          61 Simulation-Linux
21 Fighting-MS
                 191.55
                                          62 Simulation-MS
                                                              112.67999999999999
22 Fighting-PC
                 1916.1699999999996
                                           63 Simulation-PC
                                                               13109.74999999999
23 Fighting-PS
                 1620.71000000000003
                                           64 Simulation-PS
                                                               1894.99
24 Fighting-PS2
                 1843.06000000000002
                                           65 Simulation-PS2 1572.1200000000006
25 Fighting-PS3
                 554.2499999999999
                                           66 Simulation-PS3
                                                               324.07999999999987
26 Fighting-PS4
                 952.039999999998
                                          67 Simulation-PS4
                                                               868.25000000000003
27 Fighting-PS5
                 8.81
                                           68 Simulation-PS5
                                                               53.78
28 Fighting-PSN
                 1661.50000000000007
                                           69 Simulation-PSN
                                                               1055.98000000000002
29 Puzzle-DS
                 3365.659999999997
                                           70 Simulation-iOS
                                                               125.15
30 Puzzle-Linux
                 176.65
                                           71 Sports-DS
                                                               1774.34
                 205.12999999999997
31 Puzzle-MS
                                           72 Sports-Linux
                                                               62.2
32 Puzzle-PC
                 8207.870000000008
                                           73 Sports-MS
                                                               744.13
33 Puzzle-PS
                 1727.52
                                           74 Sports-PC
                                                               8130.909999999999
34 Puzzle-PS2
                 626.93000000000001
                                           75 Sports-PS
                                                               3730.2199999999975
35 Puzzle-PS3
                 193.9
                                          76 Sports-PS2
                                                               4446.689999999997
36 Puzzle-PS4
                 928.10000000000004
                                           77 Sports-PS3
                                                               1579.83000000000006
37 Puzzle-PSN
                 3317.5600000000018
                                           78 Sports-PS4
                                                               1396.7599999999993
38 Puzzle-iOS
                 121.52
                                           79 Sports-PS5
                                                               44.84
                 735.10000000000001
                                                               1905.4199999999996
39 Racing-DS
                                           80 Sports-PSN
40 Racing-Linux
                                          81 Sports-iOS
                 30.22
```

Figure 7: Output File for GenresConsole MapReduce

The output from first MapReduce is all possible combinations of genre and console with their total sales. So, we can make out that there are a total of 81 combinations, and each have a different sale which basically indicates the popularity of them. Going further in the next MapReduce the outcome is further processed to receive the top 10 combinations based on sales.

Mapper 2: TopNMapper

Figure 8: TopNMapper

The input for the second mapper is the output from **GenresConsoleReducer**. It is read as shown below.

```
/*[{Adventure-DS 3503.5900000000033} {Adventure-Linux 630.0600000000002} {Adventure-MS 229.62} {Adventure-PC 24680.950000000002}]*/
```

LongWritable is a byte offset from the current line in the input file, Text is the actual data on the line, and Context is used to emit intermediate key-value pairs. In line 18 the value input is written as the key output with a NullWritable value. This means that for each input line, the mapper will output a key-value pair where the key is the input line, and the value is NullWritable which is then passed to the next reducer.

Reducer 2: TopNReducer

```
☑ TopNReducer.java 🛭
      //Importing the libraries that contain the classes and methods needed in the Reducer class
  40 import java.io.10Exception;
5 import org.apache.hadoop.io.*;
6 import org.apache.hadoop.mapreduce.*;
  8⊖/*Creating a subclass (child class) called TopNReducer from a parent
  public class TopNReducer extends Reducer<Text, NullWritable, Text, DoubleWritable> {
         /*Initialize integer variable TOP with value 10 which is used to keep track of the number of top genre-console combinations to be output.*/
private static final int TOP= 10;
12
13
14
15
16
17
18
19<sup>©</sup>
20
21
•22
23
24
25
         /*Initialize private instance variable top10 which is a TreeMap object that maps sales to their corresponding genre-console combinations.*/
private java.util.TreeMap<Double, String> top10;
          /*@override is used to setup method from the parent class.*/
         /*initialize the topN instance variable with a new TreeMap object. This method is called once at the start of the reduce task.*/
public void setup(Context context) throws IOException, InterruptedException {
    top10 = new java.util.TreeMap⇔();
 26⊖
           /*@override is used to change the original behavior of the method "reduce" in the
27
28⊜
           parent Class "Reducer" */
           @Override
 29
30
^31
           // the reduce method takes three parameters that are a key, a value, and a context as input parameters
           public void reduce(Text key, Iterable<NullWritable> values, Context context) throws IOException, InterruptedException {
32
33
34
                /*create a variable line of data type String, convert value into string and store it in line and Split the input text line using a tab("\t") as separator and store the result in the 'line' array.*/
 35
               String[] line = key.toString().split("\t");
 36
37
                 //Store the key that is genre-console combination into genreConsole variable
 38
                String genreConsole = line[0];
 39
40
             //Store the value that is total sales variable totalSales.
                double totalSales = Double.parseDouble(line[1]);
 42
 43
                // Add the current genre-console combination to the TreeMap
                top10.put(totalSales, genreConsole);
 45
 46
                 // Keep only the top 10 entries in the TreeMap
                if (top10.size() > TOP) {
 48
                      top10.remove(top10.firstKey());
 49
```

```
51
52
53⊜
         //@Override is used for the cleanup method from the parent class.
         @Override
         // The cleanup method takes context object as parameter and return void.
.55
         public void cleanup(Context context) throws IOException, InterruptedException {
56
      /* The java.util.Set object stores Map.Entry objects with key of type Double and value of type String.

The top 10 entries are retrieved from the TreeMap in descending order using the descendingMap() method. */
57
58
59
              java.util.Set<java.util.Map.Entry<Double, String>> entries = top10.descendingMap().entrySet();
60
61
      /*For loop iterate on each entry, storing the key (total sales) to sales and value (genre-console) to variable gc*/
              for (java.util.Map.Entry<Double, String> entry : entries) {
                  double sales = entry.getKey();
String gc = entry.getValue();
63
64
65
                         returns the key-value pair, here key is genre-console combination(Text) and value is total sales(double).
                  context.write(new Text(gc), new DoubleWritable(sales));
66
67
68
69 }
        }
```

Figure 9: TopNReducer

The TopNReducer class is responsible for the reduce phase. The input key is the genre-console combination, and the input values are NullWritable objects. There is a variable TOP_N set to 10 and another variable topN as a TreeMap object.

A **TreeMap** in java is a data structure that holds key-value pairs in sorted order either ascending or descending and includes efficient methods for locating, inserting, and deleting entries. It is implemented as a Red-Black Tree, which allows it to maintain element order even after element insertion and deletion. The most significant advantages of utilising a TreeMap is that it guarantees logarithmic time complexity for most operations such as adding, removing, and searching items.

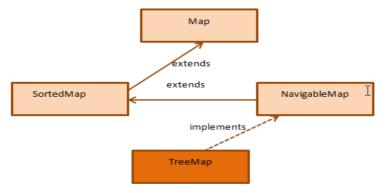


Figure 10: TreeMap Flow

In reducer the TreeMap is utilised to store genre-console combinations as keys and their associated sales as values. The line 47 of code is a if statement where the size of TreeMap is checked with each new entry and if the size exceeds 10, the lowest entry is removed.

In the cleanup method entries are retrieved from the TreeMap in descending order using the descendingMap() method as done in line 60, finally a for loop is used to iterate through the top 10 entries, storing the genre-console combination as a Text object and the total sales as a DoubleWritable object. The context object is then used to emit the key-value pairs as output records.

Main Driver Class: TopGenresConsoleSales

```
    TopGenresConsoleSales.java 
    package org.myorg;

    //Importing the libraries that contain the classes and methods needed in the driver class
    // import org.apache.hadoop.conf.Configuration;
    import org.apache.hadoop.fs.Path;
    import org.apache.hadoop.io.*;
    import org.apache.hadoop.mapreduce.*;
    import org.apache.hadoop.mapreduce.lib.input.FileInputFormat;
    import org.apache.hadoop.mapreduce.lib.output.FileOutputFormat;
```

Figure 11: Libraries of Driver Class

Despite the idea that the mapper and reducer representations are all that is needed to conduct a MapReduce job, MapReduce requires one additional piece of code: the driver class, which interfaces with the Hadoop framework and specifies the configuration parameters needed to run a MapReduce job. This includes instructing Hadoop on which Mapper and Reducer classes to use, where to get input data and in what format to find it, where to store output data, and how to organise and arrange it.

```
public class TopGenresConsoleSales {
13∈
         public static void main(String[] args) throws Exception {
              if (args.length != 4)
                    System.err.println("Usage: TopGenresConsoleSales <input path> <output path>");
                   System.exit(-1);
              ,/create the "<u>conf"</u> object from the "Configuration" class, which provides access to configuration parameters necessary for <u>Hadoop</u> job.

Configuration conf = new Configuration();
19
20
              // Create the first MapReduce job to get the total sales for each genre-console combination
Job job1 = Job.getInstance(conf, "GenresConsoleSales");
22
23
24
25
26
27
28
29
30
            //specifying the driver class in the JAR file
              job1.setJarByClass(TopGenresConsoleSales.class);
            //Setting the mapper and reducer for the job
              job1.setMapperClass(GenresConsoleMapper.class);
job1.setReducerClass(GenresConsoleReducer.class);
31
32
           //Setting the data type for the output key class to be Hadoop's data type Text
//Setting the data type for the output Value class to be Hadoop's data type DoubleWritable
33
34
35
              job1.setOutputKeyClass(Text.class);
job1.setOutputValueClass(DoubleWritable.class);
               FileInputFormat.addInputPath(job1, new Path(args[1]))
              FileOutputFormat.setOutputPath(job1, new Path(args[2]));
            //Submit the job to the cluster and wait for it to complete
               job1.waitForCompletion(true);
```

Figure 12: JOB 1 of Driver Class

In the above code line 19 is creating an object for the Configuration class which provides access to configuration parameters necessary for Hadoop job. The Job1 is created with the name "GenresConsoleSales". The jar is set to the driver class (TopGenresConsoleSales) and the mapper and reducers are set to run the first MapReduce jobs namely GenresConsoleMapper and GenresConsoleReducer respectively. The setOutputKeyClass and setOutputValueClass are used to set the output key class to Text and the output value class to DoubleWritable.

FileInputFormat is used to specify the input and output folders. Finally, the task is sent to the Hadoop cluster using waitForCompletion, and the software waits for it to finish.

```
// Create the second MapReduce job to get the top 10 genre-console combinations by total sales

Job job2 = Job.getInstance(conf, "TopGenresConsoleSales");
45
          //specifying the driver class in the JAR file
            job2.setJarByClass(TopGenresConsoleSales.class);
48
49
          //Setting the mapper and reducer for the job
            job2.setMapperClass(TopNMapper.class);
50
51
            job2.setReducerClass(TopNReducer.class);
52
          //Set the data type for the output key class to be Text and
53
          //set the data type for the output Value class to be NullWritable
54
            job2.setMapOutputKeyClass(Text.class);
55
            job2.setMapOutputValueClass(NullWritable.class);
56
57
58
            //Setting the data type for the output key class to be Hadoop's data type Text and
59
            //Setting the data type for the output Value class to be Hadoop's data type DoubleWritable
60
            job2.setOutputKeyClass(Text.class);
61
            job2.setOutputValueClass(DoubleWritable.class);
62
          //Set the path of the input and output directories
63
            FileInputFormat.addInputPath(job2, new Path(args[2]));
64
65
            FileOutputFormat.setOutputPath(job2, new Path(args[3]));
66
          //Submit the job to the cluster and wait for it to complete
67
            job2.waitForCompletion(true);
68
        }
69
70 }
71
```

Figure 13: JOB 2 of Driver Class

The Job2 is created with the name "GenresConsoleSales". The jar is set to the driver class (TopGenresConsoleSales) and the mapper and reducers are set to run the first MapReduce jobs namely TopNMapper and TopNReducer respectively. The setOutputKeyClass and setOutputValueClass are used to set the output key class to Text and the output value class to DoubleWritable. FileInputFormat is used to specify the input and output folders. Finally, the task is sent to the Hadoop cluster using waitForCompletion, and the software waits for it to finish.

STAGE	INPUT	INPUT TYPE	OUTPUT	OUTPUT TYPE
Mapper 1	Key: Input	Key:LongWritable	Key: Genre,	Key: Text
[GenresConsoleMapper]	data file Location. Value: Line by Line Data	Value: Text	Console. Value:Global_sales	Value:DoubleWritable
Reducer 1	Key: Genre,	Key: Text	Key:Combination	Key: Text
[GenresConsoleReducer]	Console. Value: Global_sales	Value:DoubleWritable	of Genre Console Value:Total Global_sales	Value:DoubleWritable
Mapper 2	Key: Output	Key:LongWritable	Key:Line by Line	Key: Text
[TopNMapper]	file of	Value: Text	data	Value:NullWritable

	Reducer 1 Location. Value: Line by Line Data		Value:NULL	
Reducer 2	Key: Genre,	Key: Text	Key:Genre,Console	Key: Text
[TopNReducer]	Console,	Value:NullWritable	Value:Total	Value:DoubleWritable
	Total		Global_sales	
	Global_sales.			
	Value: NULL			

Table2: Description of Key-Value pair for each MapReduce

Result & Evaluation:

When the job is finished, a text file with the result is saved in the HDFS output directory. As a result, it must be transferred from the HDFS to the local system before it can be accessed.

```
Bytes Written=297
(base) hadoop@hadoop-VirtualBox:~$ hdfs dfs -get output1 output1
(base) hadoop@hadoop-VirtualBox:~$
```

Figure 14: Setup the environment & run the program in Hadoop

The Figure 13 clearly mention every command that needs to be executed so that we receive the output. When we execute the last command "hdfs dfs-get output1 output1" then the output file which existed in Hadoop is copied on the local system for the user to view. The below image is the final output for my execution.

Adventure-PC	24680.95000000002
Shooter-PC	15370.119999999986
Simulation-PC	13109.74999999999
Puzzle-PC	8207.870000000008
Sports-PC	8130.90999999999
Racing-PC	7794.770000000003
Adventure-PS2	5592.76999999999
Sports-PS2	4446.689999999997
Shooter-PSN	4119.420000000002
Sports-PS	3730.2199999999975

Figure 15: Final Output – Top 10 Genre Console combination with respect to total sales

Recall the research question: What are the top 10 genre console combinations that are popular among buyers based on total sales?

To answer this question, we started with tones of data and using the jobs first figured out the different combinations of Genre and console and then their total sales. Furthermore, we analysed the long list of genre console combinations and then using another job came down to top 10 combinations. This analysis is helpful to answer the question, it's clear that the most popular combination among the buyers is Adventure and PC with 24680.95 as total sales. The top 10 combinations are displayed as output which clearly indicates the interest of the buyers.

To better evaluate these results, some simple informative graphics can be produced with Tableau:

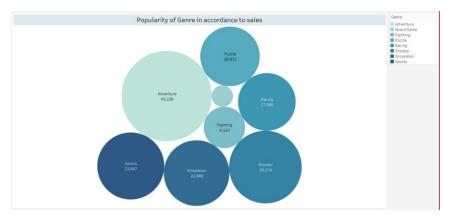


Figure 16: Popularity of genre with respect to sales

The above visualizations depict the relation of different genre and total global sales.

The bubble chart highlights the popularity of the 8-genre based on the total sales, bigger the bubble greater the contribution in the sales. From this we can easily identify that Adventure is the most popular among buyers.

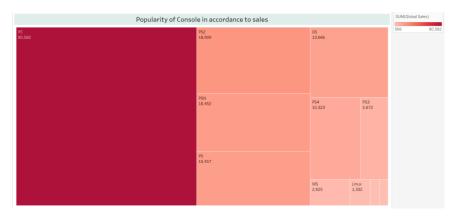


Figure 17: Popularity of console with respect to sales

The above visualizations depict the relation of different console and total global sales.

The TreeMap chart highlights the popularity of the 11-console based on the total sales, bigger the map area greater the contribution in the sales. From this we can easily identify that PC has been major contributor and is the most popular among buyers.

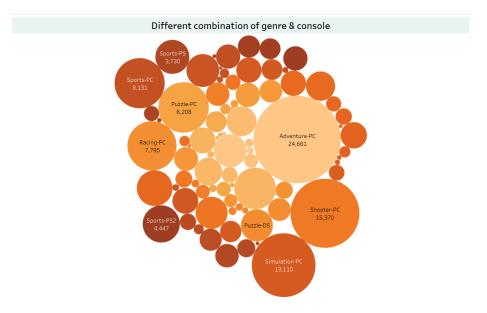


Figure 18: Visual Representation of the different combination of genre & console

From the first MapReduce job we identified that there are 81 combinations of genre and combinations, and their total global sales are calculated. The above bubble chart represents these combinations visually. The chart also highlights some of the top selling combinations like Adventure and PC (24,681), Shooter and PC (15,370) among others.

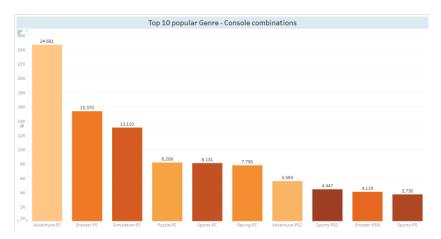


Figure 19: Visual Representation of MapReduce Job 2 (Top 10 combination based on sales)

This visualization depicts our outcome of MapReduce job that are the top 10 combinations which are most popular among buyers as their sales are maximum. Adventure and PC are the most popular combination among buyers with 24,681 million sales.

Critical Reflection:

During this project the knowledge on Hadoop MapReduce has been applied to answer a research question. Apart from the basic understanding of MapReduce new techniques were learnt to solve the problem in efficient way, like chained **MapReduce** and **TreeMap**.

To begin with two different MapReduce algorithms were used, one for finding total sales for different combinations of genre and console, and another one to perform sorting and providing top 10 combination with highest total sales. These two functionalities were executed sequentially one after another through chain MapReduce functionality.

The possible improvements or enhancements that can be made are as follows:

- 1. Run parallel MapReduce to calculate the average user score and combine it with current MapReduce to further narrow down the finding of the popularity of genre console based on user score.
- 2. In the current sequential MapReduce, we can add total number of shipment and use it along with the worldwide sales to find the revenue generated through the sales of different genre console combinations.