DAYANANDA SAGAR COLLEGE OF ENGINEERING (An Autonomous Institute Affiliated to VTU, Belagavi. Approved by AICTE & Camp; ISO 9001:2015 Certified)

Autonomous Institute Affiliated to VTU, Belagavi. Approved by AICTE & Description Shavige Malleshwara Hills, Kumaraswamy Layout, Bengaluru-560111)

DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING

SIXTH SEMESTER DATA ANALYTICS LABORATORY MANUAL

SUBJECT CODE: 21CSL66

ACADEMIC YEAR: 2023-24

PREPARED BY,

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DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING

VISION AND MISSION OF THE DEPARTMENT

VISION

To provide a vibrant learning environment in computer science and engineering with focus on industry needs and research, for the students to be successful global professionals contributing to the society.

MISSION

- * To adopt a contemporary teaching learning process with emphasis on hands on and collaborative learning.
- * To facilitate skill development through additional training and encourage student forums for enhanced learning.
 - * To collaborate with industry partners and professional societies and make the students industry ready.
 - * To encourage innovation through multidisciplinary research and development activities.
 - * To inculcate human values and ethics to groom the students to be responsible citizens.

CODE OF CONDUCT IN THE LAB

Do's

Students shall

- Come prepared for the program to be developed in the Laboratory.
- Report any broken plugs or exposed electrical wires to your faculty/laboratory technician immediately.
- Turn off the machine once you have finished using it.
- Maintain silence while working in the lab.
- Keep the Computer lab premises clean and tidy.
- Place backpacks under the table or computer counters.
- Treat fellow users of the laboratory, and all equipment within the laboratory, with the appropriate level of care and respect.

Don'ts

Students shall not

- Talk on cell phones in the lab.
- Eat or drink in the laboratory.
- Touch, connect or disconnect any plug or cable without the faculty/laboratory technician's permission.
- Install or download any software or modify or delete any system files on any lab computers.
- Read or modify other users' files.
- Meddle with other user's files.
- Leave their personal belongings unattended.

NOTE: We are not responsible for any theft.

COURSE OBJECTIVES

- 1. Develop proficiency in data management techniques for storing and sorting data.
- 2. Create MapReduce programs for diverse data processing tasks.
- 3. Analyze big data using linear models and spark.
- 4. Leverage HIVE for advanced data querying and analysis

COURSE OUTCOMES

At the end of the course, student will be able to:

CO1	Demonstrate file management in Hadoop.	
CO ₂	Process big data using MapReduce.	
CO3	Develop competency in leveraging relational data stores within the Hadoop.	
CO4	Demonstrate data analysis with spark and utilize MLlib for implementing	
	Regression models.	

LIST OF EXPERIMENTS

Experiment	Contents of the Experiment	CO's			
No	PART-A				
	FILE MANAGEMENT IN HADOOP				
	Implement the following file management tasks in Hadoop,				
	a) Adding Files and directories.				
	b) Retrieving files.				
1	c) Deleting files.	CO1			
	d) Create a new file in HDFS environment.				
	e) List files in HDFS.				
	f) Upload and download files in HDFS as well other properties copy file, move files and				
	remove file operations in HDFS.				
2	Implement word count / frequency program using MapReduce.	CO2			
3	Implement an MR program that processes a weather dataset.				
4	Perform data transfer operations using Sqoop in a Hadoop cluster setup.				
	Perform targeted data analysis using Apache Hive on the provided dataset,				
	a) To find the movie with the highest average rating.				
5	b) Identify the most active users based on the number of ratings submitted.				
	c) Discover movies with the highest number of positive ratings.				
	d) Find the top genres ranked by their average rating.				
	Utilize Apache Spark to achieve the following tasks for the given dataset,				
	a) Find the movie with the lowest average rating with RDD.				
6	b) Identify users who have rated the most movies.				
	c) Explore the distribution of ratings over time.				
	d)Find the highest-rated movies with a minimum number of ratings.				
	Analyze customer engagement data from an e-commerce company offering both mobile app				
7	and website platforms. Using linear regression, determine which platform the company				
	should prioritize for improvement efforts.				

_	Contents of the Experiment	
Experiment No	PART-B	CO's
	MINI PROJECT	
	a) Build a mini project utilizing Apache Spark to analyze a large dataset, focusing on a	
	specific domain.	
1	OR	CO4
	b) Design and execute a mini project focused on building a regression model using	
	PySpark.	

1. File Management in Hadoop

Implement the following file management tasks in Hadoop,

- a) Adding Files and directories.
- b) Retrieving files.
- c) Deleting files.
- d) Create a new file in HDFS environment.
- e) List files in HDFS.
- f) Upload and download files in HDFS as well other properties copy file, move files and remove file operations in HDFS.

File management in Hadoop is primarily done through the Hadoop Distributed File System (HDFS), which is designed to store large data sets reliably, and to stream those data sets at high bandwidth to user applications. To manage files in HDFS, you use the Hadoop File System shell (hadoop fs) or the HDFS shell (hdfs dfs), which provides a variety of commands for managing files and directories.

Below are the steps and commands for performing the listed file management tasks in Hadoop:

- a) Adding Files and directories.
 - Creating a Directory: To create a directory in HDFS, use the mkdir command.
 hdfs dfs -mkdir /user/hadoop/mydirectory
 - Adding Files to HDFS: To add a file from your local file system to HDFS, use the
 put command.

hdfs dfs -put localfile.txt /user/hadoop/mydirectory

• Upload a directory to HDFS:

hdfs dfs -put localdirectory /path/in/hdfs

b) Retrieving Files.

• To retrieve files from HDFS to your local file system, use the **get** command:

hdfs dfs -get /path/in/hdfs/localfile.txt /path/in/local

- c) Deleting files.
 - **Deleting a File**: To delete a file from HDFS, use the **rm** command.

hdfs dfs -rm /user/hadoop/mydirectory/localfile.txt

• **Deleting a Directory**: To delete a directory and its contents, use the **rm** -**r** command.

hdfs dfs -rm -r /user/hadoop/mydirectory

- d) Create a new file in HDFS environment.
 - Creating a new empty file in HDFS is not directly supported by a command. However, you can create an empty file locally and then upload it to HDFS.

touch local_empty_file.txt

hdfs dfs -put local_empty_file.txt /path/in/hdfs/emptyfile.txt

- e) List files in HDFS.
 - To list files and directories in HDFS, use the **ls** command:

hdfs dfs -ls /path/in/hdfs

Upload and download files in HDFS as well other properties copy file, move files and remove file operations in HDFS.					
	Upload Files: hdfs dfs -put localfile.txt /path/in/hdfs				
 Download 	Files : hdfs d	lfs -get /pa	th/in/hdfs/	/localfile.tz	xt /path/in/local
 Copy Files: 	Copy Files: hdfs dfs -cp /path/in/hdfs/sourcefile.txt /path/in/hdfs/destination/				
 Move Files 	: hdfs dfs -m	nv/path/in/	/hdfs/sour	cefile.txt /	path/in/hdfs/destination/
• Remove	Files:	hdfs	dfs	-rm	/path/in/hdfs/localfile.ta

2. Implement Word Count / Frequency Program Using MapReduce.

Steps to be followed:

- Step-1: Open Eclipse → then select File → New → Java Project → Name it WordCount → then Finish.
- Step-2: Create Three Java Classes into the project.

File \rightarrow New \rightarrow Class

Name them WCDriver (having the main function), WCMapper and WCReducer.

- Step-3: You have to include two Reference Libraries,
 Right Click on Project → then select Build Path → Click on Configure Build
 Path → Add External JARs (Share → Hadoop). In this add JARs of Client,
 Common, HDFS, MapReduce and YARN → Click on Apply and Close.
- Step-4: **Mapper Code** which should be copied and pasted into the **WCMapper** Java Class file.

// Importing libraries

import java.io.IOException;

import org.apache.hadoop.io.IntWritable;

import org.apache.hadoop.io.LongWritable;

import org.apache.hadoop.io.Text;

import org.apache.hadoop.mapred.MapReduceBase;

import org.apache.hadoop.mapred.Mapper;

import org.apache.hadoop.mapred.OutputCollector;

import org.apache.hadoop.mapred.Reporter;

```
implements
public
           class
                    WCMapper
                                   extends
                                               MapReduceBase
Mapper<LongWritable, Text, Text, IntWritable>
  // Map function
 public void map(LongWritable key, Text value, OutputCollector<Text,
 IntWritable> output, Reporter rep) throws IOException
   {
     String line = value.toString();
     // Splitting the line on spaces
     for (String word : line.split(" "))
     {
       if (word.length() > 0)
        {
          output.collect(new Text(word), new IntWritable(1));
Step-5: Reducer Code which should be copied and pasted into the WCReducer
Java Class file.
// Importing libraries
import java.io.IOException;
import java.util.Iterator;
import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.io.Text;
```

```
import org.apache.hadoop.mapred.MapReduceBase;
  import org.apache.hadoop.mapred.OutputCollector;
  import org.apache.hadoop.mapred.Reducer;
  import org.apache.hadoop.mapred.Reporter;
  public class WCReducer extends MapReduceBase implements Reducer<Text,
  IntWritable, Text, IntWritable>
  // Reduce function
  public void reduce(Text key, Iterator<IntWritable> value, OutputCollector<Text,
  IntWritable> output, Reporter rep) throws IOException
        int count = 0;
        // Counting the frequency of each words
        while (value.hasNext())
        {
              IntWritable i = value.next();
              count += i.get();
        output.collect(key, new IntWritable(count));
  }
• Step-6: Driver Code which should be copied and pasted into the WCDriver Java
  Class file.
  // Importing libraries
  import java.io.IOException;
  import org.apache.hadoop.conf.Configured;
  import org.apache.hadoop.fs.Path;
  import org.apache.hadoop.io.IntWritable;
  import org.apache.hadoop.io.Text;
```

```
import org.apache.hadoop.mapred.FileInputFormat;
import org.apache.hadoop.mapred.FileOutputFormat;
import org.apache.hadoop.mapred.JobClient;
import org.apache.hadoop.mapred.JobConf;
import org.apache.hadoop.util.Tool;
import org.apache.hadoop.util.ToolRunner;
public class WCDriver extends Configured implements Tool
public int run(String args[]) throws IOException
      if (args.length < 2)
            System.out.println("Please give valid inputs");
            return -1;
      JobConf conf = new JobConf(WCDriver.class);
      FileInputFormat.setInputPaths(conf, new Path(args[0]));
      FileOutputFormat.setOutputPath(conf, new Path(args[1]));
      conf.setMapperClass(WCMapper.class);
      conf.setReducerClass(WCReducer.class);
      conf.setMapOutputKeyClass(Text.class);
      conf.setMapOutputValueClass(IntWritable.class);
      conf.setOutputKeyClass(Text.class);
      conf.setOutputValueClass(IntWritable.class);
      JobClient.runJob(conf);
      return 0;
// Main Method
```

```
public static void main(String args[]) throws Exception
{
    int exitCode = ToolRunner.run(new WCDriver(), args);
    System.out.println(exitCode);
}
```

• Step-7: Now you have to make a jar file.

Right Click on Project → Click on Export → Select export destination as Jar File → Name the jar File (WordCount.jar) → Click on next → at last Click on Finish.

• Step-8: Open the terminal and change the directory to the workspace.

You can do this by using "cd workspace/" command.

Now, Create a text file (WCFile.txt) and move it to HDFS.

For that open terminal and write the below code (remember you should be in the same directory as jar file you have created just now),

cat WCFile.text

• Step-9: Now, run the below command to copy the file input file into the HDFS,

hadoop fs -put WCFile.txt WCFile.txt

• Step-10: Now to run the jar file, execute the below code,

hadoop jar wordcount.jar WCDriver WCFile.txt WCOutput

• Step-11: After Executing the code, you can see the result in WCOutput file or by writing following command on terminal,

hadoop fs -cat WCOutput/part-00000

3. Implement An Mr Program That Processes A Weather Dataset.

Steps to be followed:

 Step-1: We can download the dataset from this <u>Link</u>, For various cities in different years. choose the year of your choice and select any one of the data textfile for analysing.

We can get information about data from README.txt file available on the NCEI website.

- Step-2: Make a project in Eclipse with below steps:
 - First Open Eclipse → then select File → New → Java Project → Name it MyProject → then select use an execution environment → choose JavaSE-1.8 then next → Finish.
 - In this Project Create Java class with name MyMaxMin → then click Finish.
 - Copy the below source code to this **MyMaxMin** java class.

```
import org.apache.hadoop.conf.Configuration;
import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.fs.Path;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.mapreduce.Job;
import org.apache.hadoop.mapreduce.lib.input.FileInputFormat;
import org.apache.hadoop.mapreduce.lib.output.FileOutputFormat;
public class maxtemperature {
    public static void main(String[] args) throws Exception {
```

Configuration conf = new Configuration();

```
Job job = Job.getInstance(conf, "maxtemperature");
             job.setJarByClass(maxtemperature.class);
             // TODO: specify a mapper
             job.setMapperClass(MaxTempMapper.class);
             // TODO: specify a reducer
             job.setReducerClass(MaxTempReducer.class);
             // TODO: specify output types
             job.setOutputKeyClass(Text.class);
             job.setOutputValueClass(IntWritable.class);
             // TODO: specify input and output DIRECTORIES (not files)
             FileInputFormat.setInputPaths(job, new Path(args[0]));
             FileOutputFormat.setOutputPath(job, new Path(args[1]));
             if (!job.waitForCompletion(true))
                   return;
       }
 }
 import java.io.IOException;
 import org.apache.hadoop.io.LongWritable;
 import org.apache.hadoop.io.Text;
 import org.apache.hadoop.io.IntWritable;
 import org.apache.hadoop.mapreduce.Mapper;
 public class MaxTempMapper extends Mapper<LongWritable, Text, Text,
IntWritable > {
 public void map(LongWritable key, Text value, Context context)
```

```
throws IOException, InterruptedException {
String line=value.toString();
String year=line.substring(15,19);
int airtemp;
if(line.charAt(87)== '+')
airtemp=Integer.parseInt(line.substring(88,92));
else
       airtemp=Integer.parseInt(line.substring(87,92));
       String q=line.substring(92,93);
       if(airtemp!=9999&&q.matches("[01459]"))
       {
             context.write(new Text(year),new IntWritable(airtemp));
import java.io.IOException;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.mapreduce.Reducer;
public class MaxTempReducer extends Reducer<Text, IntWritable, Text,
IntWritable>
public void reduce(Text key, Iterable<IntWritable> values, Context
context)
      throws IOException, InterruptedException
{
      int maxvalue=Integer.MIN VALUE;
      for (IntWritable value : values)
```

```
{
maxvalue=Math.max(maxvalue, value.get());
}
context.write(key, new IntWritable(maxvalue));
}
```

- Now we need to add external jar for the packages that we have import. Download the jar package Hadoop Common and Hadoop MapReduce Core according to the Hadoop version.
- Now we add these external jars to our **MyProject**.

Right Click on MyProject \rightarrow then select Build Path \rightarrow Click on Configure Build Path and select Add External jars.... and add jars from its download location then click \rightarrow Apply and Close.

• Now export the project as jar file.

Right-click on MyProject choose Export.. and go to Java \rightarrow JAR file click \rightarrow Next and choose your export destination then click \rightarrow Next. choose Main Class as MyMaxMin by clicking \rightarrow Browse and then click \rightarrow Finish \rightarrow Ok.

• Step-4: Start the Hadoop daemons.

```
start-dfs.sh
start-yarn.sh
```

hdfs dfs -ls /

• Step-5: Move the dataset to Hadoop HDFS.

```
hdfs dfs -put /file_path /destination
In below command / shows the root directory of our HDFS,
hdfs dfs -put /home/..../datasetname.txt /
```

• Step-6: Now Run your Jar File with below command and produce the output				
in MyOutput File.				
hadoop jar /jar_file_location /dataset_location_in_HDFS /output-file_name				
hadoop jar //Project.jar /datasetname.txt /MyOutput				
• Step-7: Now Move to localhost:50070/, under utilities select Browse the file system and download part-r-00000 in / MyOutput directory to see result.				
• Step-8: See the result in downloaded file.				

4. Perform data transfer operations using Sqoop in a Hadoop cluster setup.

Prerequisites:

- a) Hadoop Cluster: Ensure your Hadoop cluster is up and running.
- b) **Sqoop Installation**: Sqoop should be installed and configured in your Hadoop cluster.
- c) **Database Access**: Access to the source database (e.g., MySQL, PostgreSQL) with the required JDBC drivers installed on the Sqoop nodes.
- d) **JDBC Driver**: Ensure the JDBC driver for your database is available in the Sqoop lib directory.
- Step-1: Connect to the MySQL Database
 cp mysql-connector-java-*.jar \$SQOOP_HOME/lib/
- Step-2: Create the Sqoop Import and Export Command sqoop import \

```
--connect
```

 $jdbc:mysql://<MYSQL_HOST>:<MYSQL_PORT>/<DATABASE_NAME>\setminus (ACCORNING OF ACCORNING O$

- --username <USERNAME> \
- --password <PASSWORD> \
- --table <TABLE NAME> \
- --target-dir < HDFS_TARGET_DIR > \
- --num-mappers 1

```
--connect
jdbc:mysql://<MYSQL_HOST>:<MYSQL_PORT>/<DATABASE_NAME> \
--username <USERNAME> \
--password <PASSWORD> \
--table <TABLE_NAME> \
--export-dir <HDFS_EXPORT_DIR> \
--input-fields-terminated-by ',' \
--num-mappers 1
```

• Step-3: Execute the Sqoop Command

Run the Sqoop import command in your Hadoop cluster terminal.

• Step-4: Verify the Imported and Exported Data

hadoop fs -ls <HDFS_TARGET_DIR>
SELECT * FROM <TABLE NAME>;

- 5. Perform targeted data analysis using Apache Hive on the provided dataset,
 - a) To find the movie with the highest average rating.
 - b) Identify the most active users based on the number of ratings submitted.
 - c) Discover movies with the highest number of positive ratings.
 - d) Find the top genres ranked by their average rating.

Steps to be followed:

Prerequisites:

- 1. **Apache Hive Setup**: Ensure you have Apache Hive installed and configured.
- 2. **Dataset**: Load the provided dataset into Hive tables. Assume we have two tables: **movies** and **ratings**.
 - movies table has columns: movie_id, title, genres.
 - ratings table has columns: user_id, movie_id, rating.

How to transfer dataset from windows to VM?

Step-1: Go to VM → Network → Adapter 1 → Attached to: Select Bridge adapter → Name: Intel® Ethernet Connection → Advanced → Promiscuous mode: Allow all → OK

Step-2: Run Hadoop

Step-3: At the top right an option is available with double headed arrow marks → Enable connection → Connect the information (IP address will be found here).

Step-4: Go to terminal in Hadoop \rightarrow mkdir windows \rightarrow cd windows \rightarrow ls

Step-5: Go to windows and check if the dataset is available in either **documents** or **pictures**, if not put it in the said folders.

Step-6: Go to CMD → cd pictures/documents → scp movies.csv hadoop@172.25.4.67:/home/hadoop/windows

It will ask for password: hadoop

Step-7: Go to home/hadoop/windows/movies.csv and check for the dataset which will be moved from windows to Hadoop.

```
Step-8: Go to terminal in VM and execute the following commands,
```

```
> startCDH
> 1s
>cp /home/hadoop/windows/movies.csv /home/hadoop/
>1s
>hadoop fs -put movies.csv movies1.csv (Give any name of your choice)
>hadoop fs -ls
>hadoop fs -ls /user
>hadoop fs -ls /user/hive (Check for warehouse)
>hive
hive> create table if not exists movies(
id int,
name string,
genre string)
 row format delimited
fields terminated by ',';
hive> select * from movies;
hive> load data local inpath 'movies.csv' into table movies;
Copying data from file:/home/hadoop/movie.csv
Copying file: file:/home/hadoop/movie.csv
Loading data to table default.movies
```

```
hive> select * from movies;
hive> select id from movies;
hive> select name from movies;
hive> show tables;
hive> drop table movies;
```

a. To find the movie with the highest average rating.

```
select movie_id, avg(rating) as avg_rating from ratings group by movie_id order by avg_rating desc limit 1;
```

b. Identify the most active users based on the number of ratings submitted.

```
select user_id, count(*) as num_ratings
from ratings
group by user_id
order by num_ratings desc
limit 10; -- you can adjust the limit as needed
```

c. Discover movies with the highest number of positive ratings.

```
assuming a positive rating is 4 or 5:
select movie_id, count(*) as positive_ratings
from ratings
where rating >= 4
group by movie_id
order by positive_ratings desc
limit 10; -- you can adjust the limit as needed
```

d. Find the top genres ranked by their average rating.

```
this step requires joining the movies and ratings tables,
-- explode genres and calculate average ratings for each genre
with exploded_genres as (
    select movie_id, explode(split(genres, '[|]')) as genre
    from movies
),
genre_ratings as (
    select g.genre, avg(r.rating) as avg_rating
    from exploded_genres g
    join ratings r on g.movie_id = r.movie_id
    group by g.genre
)
select genre, avg_rating
from genre_ratings
order by avg_rating desc;
```

- 6. Utilize Apache Spark to achieve the following tasks for the given dataset,
- a) Find the movie with the lowest average rating with RDD.
- b) Identify users who have rated the most movies.
- c) Explore the distribution of ratings over time.
- d)Find the highest-rated movies with a minimum number of ratings.

Steps to be followed:

Prerequisites:

- Apache Spark Setup: Ensure you have Apache Spark installed and configured. (If not go to terminal and install pyspark using pip install pyspark)
- Dataset: Load the dataset.
 - o movies dataset has columns: movie id, title, genres.
 - o ratings dataset has columns: user_id, movie_id, rating.
- Loading Data: First, load the data into Spark RDDs or DataFrames.

```
from pyspark.sql import SparkSession
# Create RDDs
#movies_rdd = movies_df.rdd
#ratings rdd = ratings df.rdd
# Initialize Spark session
spark = SparkSession.builder.appName("MovieRatingsAnalysis").getOrCreate()
# Load datasets
movies df
                      spark.read.csv("/Users/amithpradhaan/Desktop/ml-latest-small/movies.csv",
header=True, inferSchema=True)
ratings df
                      spark.read.csv("/Users/amithpradhaan/Desktop/ml-latest-small/ratings.csv",
header=True, inferSchema=True)
# Create RDDs
movies rdd = movies df.rdd
ratings rdd = ratings df.rdd
```

```
a) Find the Movie with the Lowest Average Rating Using RDD.
# Compute average ratings
avg ratings rdd = ratings rdd.map(lambda x: (x['movieId'], (x['rating'], 1))) \setminus
  .reduceByKey(lambda x, y: (x[0] + y[0], x[1] + y[1])) \setminus
  .mapValues(lambda x: x[0] / x[1])
# Find the movie with the lowest average rating
lowest avg rating = avg ratings rdd.sortBy(lambda x: x[1]).first()
print(f"Movie with the lowest average rating: {lowest avg rating}")
b) Identify Users Who Have Rated the Most Movies.
# Compute number of ratings per user
user ratings count = ratings rdd.map(lambda x: (x['userId'], 1))
  .reduceByKey(lambda x, y: x + y) \setminus
  .sortBy(lambda x: x[1], ascending=False)
# Get top users
top_users = user_ratings_count.take(10)
print(f"Top users by number of ratings: {top users}")
c) Explore the Distribution of Ratings Over Time.
from pyspark.sql.functions import from unixtime, year, month
# Convert timestamp to date and extract year and month
ratings_df = ratings_df.withColumn("year", year(from_unixtime(ratings_df['timestamp']))) \
              .withColumn("month", month(from unixtime(ratings df['timestamp'])))
# Group by year and month to get rating counts
ratings_over_time = ratings_df.groupBy("year", "month").count().orderBy("year", "month")
```

```
# Show distribution

ratings_over_time.show()

d) Find the Highest-Rated Movies with a Minimum Number of Ratings.

Set a minimum number of ratings, ex: 100.

# Compute average ratings and count ratings per movie

movie_ratings_stats = ratings_rdd.map(lambda x: (x['movieId'], (x['rating'], 1))) \

.reduceByKey(lambda x, y: (x[0] + y[0], x[1] + y[1])) \

.mapValues(lambda x: (x[0] / x[1], x[1])) # (avg_rating, count)

# Filter movies with a minimum number of ratings
```

Find the highest-rated movies

 $min_ratings = 100$

 $highest_rated_movies = qualified_movies.sortBy(lambda~x:~x[1][0],~ascending=False).take(10)$

print(f"Highest-rated movies with at least {min_ratings} ratings: {highest_rated_movies}")

qualified movies = movie ratings stats.filter(lambda x: $x[1][1] \ge min ratings$)

EXPECTED OUTPUTS:

- a) Lowest Average Rating Movie:
 - Movie with the lowest average rating: (movieId, averageRating)
- b) Top Users by Number of Ratings:
 - Top users by number of ratings: [(userId1, count1), (userId2, count2), ...]
- c) Distribution of Ratings Over Time:

+		+
year +		count
1996		58
1996	4	165
1996	5	832
1996	6	882
1996	7	490
1996	8	1010
1996	9	384
1996	10	935
1996	11	978
1996	12	306
1997	1	250
1997	2	323
1997	3	398
1997	4	219
1997	5	303
1997	6	84
1997	7	70
1997	9	236
1997	10	1
1997		j 2 j
+		+

d) Highest-Rated Movies with Minimum Number of Ratings:

• Highest-rated movies with at least 100 ratings: [(movieId1, (averageRating1, count1)), (movieId2, (averageRating2, count2)), ...]

7. Analyze customer engagement data from an e-commerce company offering both mobile app and website platforms. Using linear regression, determine which platform the company should prioritize for improvement efforts.

Prerequisites:

- Apache Spark Setup: Ensure you have Apache Spark installed and configured.

 (If not go to terminal and install pyspark using pip install pyspark)
- **Dataset**: The dataset that we make use of is cruise_ship_info
 The dataset contains 159 instances with 9 features.

The Description of dataset is as below:

- Ship Name
- Cruise Line
- Age (as of 2013)
- Tonnage (1000s of tons)
- passengers (100s)
- Length (100s of feet)
- Cabins (100s)
- Passenger Density
- Crew (100s)

Program:

import pyspark

from pyspark.sql import SparkSession

spark=SparkSession.builder.appName('housing price model').getOrCreate()

#create spark dataframe of input csv file

df=spark.read.csv('/content/drive/MyDrive/cruise_ship_info.csv',inferSchema=True,header=True)

df.show(10)

#prints structure of dataframe along with datatype

df.printSchema()

#In our predictive model, below are the columns

df.columns

from pyspark.ml.feature import StringIndexer

indexer=StringIndexer(inputCol='Cruise line',outputCol='cruise cat')

indexed=indexer.fit(df).transform(df)

#above code will convert string to numeric feature and create a new dataframe

#new dataframe contains a new feature 'cruise cat' and can be used further

#feature cruise cat is now vectorized and can be used to fed to model

for item in indexed.head(5):

```
print(item)
print('\n')
from pyspark.ml.linalg import Vectors
from pyspark.ml.feature import VectorAssembler
#creating vectors from features
#Apache MLlib takes input if vector form
assembler=VectorAssembler(inputCols=['Age',
 'Tonnage',
 'passengers',
 'length',
 'cabins',
 'passenger density',
 'cruise cat'],outputCol='features')
output=assembler.transform(indexed)
output.select('features','crew').show(5)
#final data consist of features and label which is crew.
final data=output.select('features','crew')
#splitting data into train and test
train data,test data=final data.randomSplit([0.7,0.3])
train data.describe().show()
test data.describe().show()
#import LinearRegression library
from pyspark.ml.regression import LinearRegression
#creating an object of class LinearRegression
#object takes features and label as input arguments
ship lr=LinearRegression(featuresCol='features',labelCol='crew')
#pass train data to train model
trained_ship_model=ship_lr.fit(train_data)
#evaluating model trained for Rsquared error
ship results=trained ship model.evaluate(train data)
print('Rsquared Error :',ship results.r2)
#testing Model on unlabeled data
#create unlabeled data from test data
unlabeled_data=test_data.select('features')
```

```
unlabeled_data.show(5)
predictions=trained_ship_model.transform(unlabeled_data)
predictions.show()
```

EXPECTED OUTPUT:

(4.0,220.0,54.0,1 21.271739496668904 (5.0,86.0,21.04,9 9.114033898203381 (5.0,160.0,36.34, 15.293748410828057 (6.0,112.0,38.0,9 11.112622870326526 (7.0,89.6,25.5,9 10.824133054440468 (8.0,77.499,19.5, 8.430096442438705 (8.0,91.0,22.44,9 9.958161654581318 (9.0,59.058,17.0, 7.340683021224405 (9.0,110.0,29.74, 11.970586563083168 (9.0,113.0,26.74, 11.371217092465935 (10.0,58.825,15.6 7.080310095115826 (10.0,68.0,10.8,7 6.754771632667488 (10.0,90.09,25.01 8.749834607555409 (10.0,105.0,27.2, 11.21389654880134 (10.0,138.0,31.14 13.168398774919039 (11.0,90.09,25.01 8.747297495806473 (11.0,91.0,20.32, 9.227370515613016 (11.0,91.62700000 9.19180119229311 (11.0,110.0,29.74 11.976769354346743	+	++
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only showing top 20 rows