



MANIPAL INSTITUTE OF TECHNOLOGY
MANIPAL
(A constituent unit of MAHE, Manipal)

DSE-3264 - Big Data Analytics Laboratory Manual

Department	:	Data Science Engineering And Computer Applications			
Course Name & code	:	DSE-3264 & Big Data Analytics Laboratory			
Semester & branch	:	VI Sem & BTech Data Science & Engineering			
Name of the faculty	:	Dr. Saraswati Koppad, Dr. Shavantrevva Sangappa Bilakeri			
No of contact hours/week:		L	T	P	C
		0	0	3	1

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Course Objectives ·

1. Gain practical experience using big data tools and platforms like Hadoop, Spark, Hive, Pig, and HBase to store, process, and analyze large datasets.
2. Implement Hadoop MapReduce for processing Big Data
3. Apply data analytics techniques to solve problems and extract meaningful insights using big data frameworks.

Course Outcomes:

At the end of this course, students will have the ability to

1. Demonstrate the ability to use big data frameworks such as Hadoop, Spark
2. Apply MapReduce techniques to process large data sets.
3. Demonstrate the ability to use big data tools such as Pig, Hive, and HBase to store and process Big Data
4. Apply analytical methods and techniques to solve data-driven problems.

Evaluation plan : (Tentative)

- Internal Assessment Marks: 60%
- End semester assessment of 2-hour duration: 40 %

Evaluation pattern

Internal Marks – 60 + End Sem - 40	
Internal Marks	Internal Assessment – 40 + Mid-sem - 20
Internal Assessment	Lab observations – 2*11(22) + Viva (18)
Lab Observation	Record (1*11) + Execution (1*11)
Viva	Quiz and mini-project (18)

INSTRUCTIONS TO THE STUDENTS

Pre-Lab Session Instructions

- Be on time, adhere to the institution's rules, and maintain decorum.
- Leave your mobile phones, pen drives, and other electronic devices in your bag and keep the bag in the designated place in the lab.
- Must Sign in to the log register provided.
- Make sure to occupy the allotted system and answer the attendance.

In-Lab Session Instructions

- Follow the instructions on the allotted exercises.
- Show the program and results to the instructors on completion of experiments.
- Copy the program and results for the lab record.
- Prescribed textbooks and class notes can be kept ready for reference if required.

General Instructions for the Exercises in Lab

- Academic honesty is required in all your work. You must solve all programming assignments independently, except where group work is authorized. This means you must not take, show, give, or otherwise allow others to take your program code, problem solutions, or other work.
- The programs should meet the following criteria:
 - Programs should be interactive with appropriate prompt messages, error messages if any, and descriptive messages for outputs.
 - Programs should perform input validation (Data type, range error, etc.), give appropriate error messages, and suggest corrective actions.
 - Comments should be used to give the statement of the problem, and every function should indicate the purpose of the function, inputs, and outputs.
 - Statements within the program should be properly indented.
 - Use meaningful names for variables and functions.
 - Make use of constants and type definitions wherever needed.
 - The exercises for each week are divided into three sets:
 - Solved solutions
 - Lab exercises - to be completed during lab hours
 - Additional Exercises - to be completed outside the lab or in the lab to enhance the skill

Questions for lab tests and examinations are not necessarily limited to the questions in the manual but may involve some variations and/or combinations of the questions.

THE STUDENTS SHOULD NOT

- Possess mobile phones or any other electronic gadgets during lab hours.
- Go out of the lab without permission.
- **Change/update any configuration in your allotted system. If so, the student will lose internal assessment marks**

Week 1 - Understanding Hadoop and HDFS Basic Commands

Introduction to the Hadoop Ecosystem and HDFS

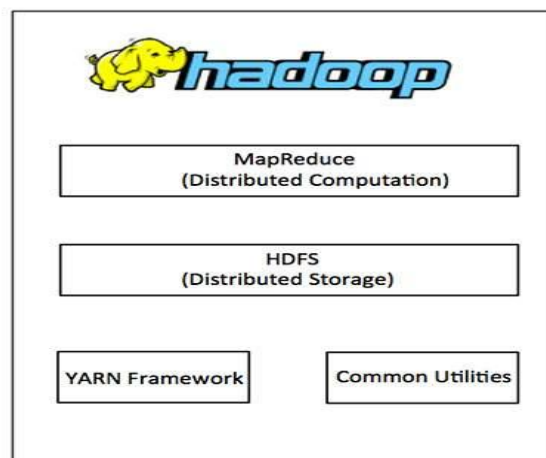
The Hadoop Ecosystem is a collection of tools and frameworks that work together to manage and process big data. HDFS (Hadoop Distributed File System) is the backbone of the Hadoop ecosystem, providing distributed storage and fault tolerance.

Hadoop Ecosystem Overview

The Hadoop Ecosystem is built around the core Hadoop components and includes tools for data storage, processing, querying, and analytics. Here's a breakdown:

Core Components:

1. **HDFS (Hadoop Distributed File System):**
 - Provides distributed storage.
 - Handles large files across multiple machines.
2. **YARN (Yet Another Resource Negotiator):**
 - Manages resources and job scheduling in the cluster.
3. **MapReduce:**
 - Programming model for data processing.



Supporting Tools:

1. **Apache Hive:**
 - Data warehouse tool for querying data in HDFS using SQL-like language (HiveQL).
2. **Apache HBase:**
 - NoSQL database for real-time read/write access to large datasets.
3. **Apache Spark:**
 - Fast, in-memory data processing engine.
4. **Apache Pig:**
 - High-level platform for creating MapReduce programs using a scripting language (Pig Latin).
5. **Apache Sqoop:**
 - Tool for transferring data between Hadoop and relational databases.
6. **Apache Flume:**
 - Tool for collecting and transferring large amounts of log data into HDFS.
7. **ZooKeeper:**
 - Centralized service for managing distributed systems.
8. **Oozie:**
 - Workflow scheduler for Hadoop jobs.

HDFS Basics

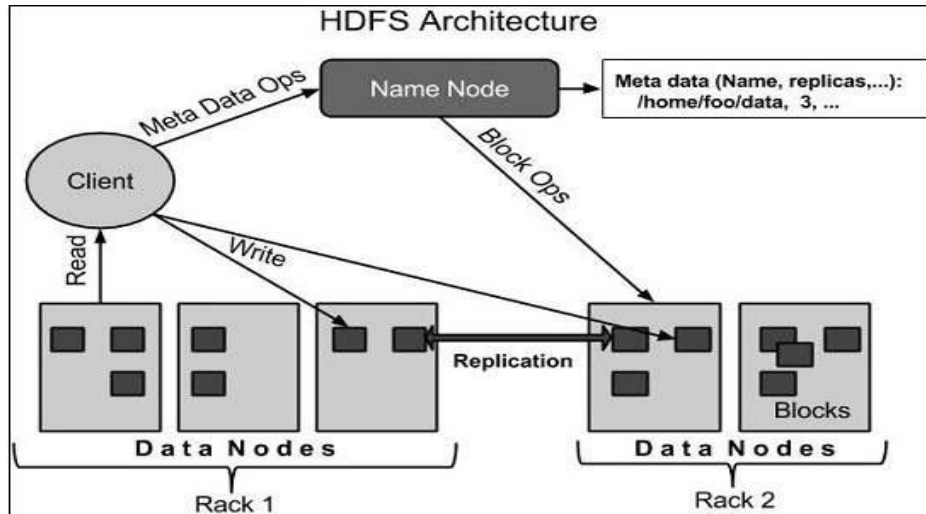
HDFS is the storage layer of Hadoop, designed for scalability and fault tolerance. It handles massive data volumes by breaking files into smaller chunks and distributing them across a cluster.

HDFS Features:

- Replication: Ensures fault tolerance by replicating data blocks (default: 3 copies).
- Write-Once, Read-Many: Optimized for batch processing.
- Scalability: Handles petabytes of data across thousands of nodes.

HDFS Architecture:

1. NameNode:
 - Maintains metadata (e.g., file structure, permissions).
 - Coordinates DataNodes but does not store data.
2. DataNodes:
 - Store actual data blocks.
 - Perform read/write operations as instructed by NameNode.
3. Secondary NameNode:
 - Periodically saves snapshots of NameNode metadata (not a backup).



Working with Hadoop:

- Open a new terminal and start the Hadoop service by following commands

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

This command initializes all the required Hadoop daemons for the cluster to become operational.

When you want to stop the services use the command:

```
stop-dfs.sh  
stop-yarn.sh
```

- Check the Status of Hadoop Services (To confirm that all necessary Hadoop services are up and running, you can check their status using the 'jps' command)

```
jps
```

The 'jps' command displays the Java Virtual Machine (JVM) processes and should show a list of Hadoop services running, such as the NameNode, DataNode, ResourceManager, and NodeManager.

```
bdalab@saraswati:~$ start-dfs.sh
Starting namenodes on [localhost]
Starting datanodes
Starting secondary namenodes [saraswati]
bdalab@saraswati:~$ start-yarn.sh
Starting resourcemanager
Starting nodemanagers
bdalab@saraswati:~$ jps
3920 NameNode
4996 Jps
4660 NodeManager
4054 DataNode
4520 ResourceManager
4252 SecondaryNameNode
bdalab@saraswati:~$
```

- **Accessing Hadoop Namenode and Resource Manager:**

To access the Hadoop Namenode, open a web browser and enter the following URL:

<http://localhost:9870>

Similarly, to access the Hadoop Resource Manager, open a web browser and enter the following URL:

<http://localhost:8088>

- **Verifying the Hadoop Cluster:**

You can check the Hadoop version by following command

```
hadoop version
```

Create Directories in HDFS:

```
hdfs dfs -mkdir /user/dir_name
```

List Directories in HDFS:

```
hdfs dfs -ls /
hdfs dfs -ls /user/
```

Transfer Files to the Hadoop File System:

```
hdfs dfs -put source-filename /destination_folder/
```


Week-1 Exercise:

1. Practice Linux commands (No need to write in the record book)
 - a. What command would you use to display the current directory?
 - b. How do you list all files, including hidden ones, in a directory?
 - c. Which command is used to create a new directory?
 - d. How can you check the currently logged-in user?
 - e. What does the pwd command do?
 - f. How do you copy a file from one location to another?
 - g. What is the command to move a file?
 - h. How can you delete a directory and all its contents?
 - i. How do you rename a file in Linux?
 - j. What is the difference between `rm` and `rmdir`?
 - k. What command is used to change the permissions of a file?
 - l. How can you view the permissions of a file?
 - m. Which command changes the owner of a file?
 - n. How do you add execute permission for the owner of a file?
 - o. What is the command to display all currently running processes?
 - p. How can you check the IP address of your system?
 - q. How do you display the contents of a file?
 - r. How can you display the current date and time in Linux?
2. Explore Basic HDFS Commands
 - a. Remove directory
 - b. View/Read File Contents
 - c. Download/copy a file from HDFS to the local system
 - d. Copy/move files within HDFS
 - e. Remove file from HDFS
 - f. View file permission

Week 2: Advanced HDFS Commands and MapReduce

Week 2 Exercises (Advanced HDFS Commands):

Create a directory in Hadoop (/user/your_reg_no/lab2). Use the same directory to keep all your files.

hdfs dfs -mkdir /user/your_reg_no/lab2

(a).

1. Upload a new file (Ex. File1.txt) into Hadoop. Demonstrate the hdfs commands for the following:
 - a. Check the permission of the file
 - b. Change the permission of the file.
 - c. Check ownership of the file
2. Upload a large file (Ex: largefile.txt) into Hadoop. Demonstrate the hdfs commands for following.
 - a. Check block details of the file
 - b. View the current replication factor for file
 - c. Modify the replication factor of the file
3. Demonstrate hdfs commands for the following tasks
 - a. Check directory size for the directory you have created (Disk usage analysis)
 - b. Find the total capacity and usage of HDFS
4. Upload a text file with sample data to the Hadoop (Ex: analytics.txt). Using HDFS commands, find
 - a. Number of lines in the file
 - b. Number of occurrences of a specific word (Ex. Hadoop) in the file.

(b). Executing simple MapReduce jobs:

What is Hadoop MapReduce?

Hadoop MapReduce is a programming model and processing framework for processing large datasets in a distributed environment. It divides the job into smaller tasks, processes them in parallel, and consolidates the results.

Key Components

1. **Map Phase:**

- Processes input data and generates intermediate key-value pairs.
- Example: Counting words in a document; the mapper emits each word as a key and the count 1 as a value.

2. **Shuffle and Sort Phase:**

- Intermediate key-value pairs are grouped and sorted by key.
- Ensures all values for a specific key are brought together.

3. **Reduce Phase:**

- Processes grouped data from the shuffle phase to generate final output.
- Example: Aggregating word counts.

MapReduce Workflow

1. **Input Data:**

- Stored in HDFS and divided into splits.

2. **Mapper:**

- Processes each split and outputs key-value pairs.

3. **Shuffle and Sort:**

- Intermediate outputs are shuffled and sorted by the framework.

4. **Reducer:**

- Processes sorted key-value pairs and generates the final result.

5. **Output:**

- Written back to HDFS.

(b). Executing sample MapReduce jobs:

What is Hadoop MapReduce?

Hadoop MapReduce is a programming model and processing framework for processing large datasets in a distributed environment. It divides the job into smaller tasks, processes them in parallel, and consolidates the results.

Key Components

4. **Map Phase:**

- Processes input data and generates intermediate key-value pairs.

- Example: Counting words in a document; the mapper emits each word as a key and the count 1 as a value.

5. Shuffle and Sort Phase:

- Intermediate key-value pairs are grouped and sorted by key.
- Ensures all values for a specific key are brought together.

6. Reduce Phase:

- Processes grouped data from the shuffle phase to generate final output.
- Example: Aggregating word counts.

MapReduce Workflow

6. Input Data:

- Stored in HDFS and divided into splits.

7. Mapper:

- Processes each split and outputs key-value pairs.

8. Shuffle and Sort:

- Intermediate outputs are shuffled and sorted by the framework.

9. Reducer:

- Processes sorted key-value pairs and generates the final result.

10. Output:

- Written back to HDFS.

Part 1: Run below commands to install python (Locally)

```
wget https://www.python.org/ftp/python/3.6.5/Python-3.6.5.tgz
tar -xvf Python-3.6.5.tgz
```

Step-1. Write a Mapper

Mapper.py: Initially the partition of content takes place based on `line.split()` function, number of partitions made = number of mapper class gets created. Mapper overrides the `—mapl` function which provides `<key, value>` pairs as the input. Even the key is repeated in same or different mapper class it doesnot matter as the default value for every key is assigned to be as “1”. A Mapper implementation may output `<key,value>` pairs using the provided Context .

Input value of the WordCount Map task will be a line of text from the input data file and the key would be the line number `<line_number, line_of_text>` . Map task outputs `<word, one>` for each word in the line of text.

Pseudo-code : mapper.py

```
#!/usr/bin/python3
"mapper.py"
import sys
for line in sys.stdin:
    # remove leading and trailing whitespace
    line = line.strip()
    # split the line into words
    words = line.split()
    # increase counters
    for word in words:
        print ('%s\t%s' % (word, 1))
```

Step-2. Write a Reducer

A Reducer collects the intermediate <key,value> output from multiple map tasks and assemble a single result. Here, the WordCount program will sum up the occurrence of each word to pairs as <word, occurrence>.

Pseudo-code: reducer.py

```
#!/usr/bin/python3
"reducer.py"
import sys
current_word = None
current_count = 0
for line in sys.stdin:
    line = line.strip()
    word, count = line.split("\t")
    count = int(count)
    if current_word == word:
        current_count += count
    else:
        if current_word:
            print ("%s\t%s" % (current_word, current_count))
        current_word = word
        current_count = 1
```

TO run word count program locally use following commands

Command: cat input.txt |python3 mapper.py

Output:

hi	1
how	1
are	1
you	1
i	1
am	1
good	1
hope	1
you	1
doing	1
good	1
too	1
how	1
about	1
you.	1
i	1
am	1
in	1
manipal	1
studying	1
Btech	1
in	1
Data	1
science.	1

Command: cat input.txt |python3 mapper.py|sort|python3 reducer.py

Output:

about	1
am	2
are	1

Btech	1
Data	1
doing	1
good	2
hi	1
hope	1
how	2
i	2
in	2
manipal	1
science.	1
studying	1
too	1
you	2
you.	1

Part 2. TO run word count program on Hadoop framework use following command:

(i) Prepare input data (inputwc.txt)

Load input.txt into Hadoop:

```
hdfs dfs -put inputwc.txt /user/studentregno/lab2
```

(ii) Run the Example MapReduce Job: (Available with Hadoop Installation)

```
hadoop jar /usr/local/hadoop/share/hadoop/mapreduce/hadoop-mapreduce-examples-3.4.1.jar  
wordcount /user/studentregno/lab2/inputwc.txt/user/studentregno/lab2/output
```

(iii) View the Results

```
hdfs dfs -cat /user/studentregno/lab2/output/part-r-00000
```


Week 2: Exercise: MapReduce with Python

1. Consider the text file (consider larger file size) of your choice and perform word count using MapReduce technique.
2. Perform Matrix operations using MapReduce by considering 3 * 3 matrix and perform following operations:
 - i. Matrix addition and subtraction
 - ii. Matrix Multiplication
 - iii. Matrix transpose

Note: Consider 3*3 matrix content as shown below

a,0,0,10

a,0,1,20

a,0,2,30

a,1,0,40

a,1,1,50

a,1,2,60

a,2,0,70

a,2,1,80

a,2,2,90

b,0,0,1

b,0,1,2

b,0,2,3

b,1,0,4

b,1,1,5

b,1,2,6

b,2,0,7

b,2,1,8

b,2,2,9

3. Create a text file containing the 20 student details such as registration number, name and marks (ex: 1001, john,45). Write a MapReduce program to sort data by student name

Week 3: MapReduce Programs

Example Program: (calculate word counts in a text file)

(i) Create a mapper script

The Mapper processes the input line by line, splitting it into words and emitting each word with a count of 1.

```
#!/usr/bin/env python3
# mapper.py
import sys

# Read lines from standard input
for line in sys.stdin:
    # Split the line into words
    words = line.strip().split()
    for word in words:
        # Emit each word with a count of 1
```

(ii) Create a Reducer script

The Reducer receives sorted input from the Mapper, groups values by key, and computes

the sum of
counts for
each word.

```
#!/usr/bin/env python3
# reducer.py
import sys

current_word = None
current_count = 0
word = None

# Read lines from standard input
for line in sys.stdin:
    # Parse the input we got from mapper
    word, count = line.strip().split("\t")
    count = int(count)

    # If this is the first word or a new word
    if current_word == word:
        current_count += count
    else:
        if current_word:
            # Emit the word and its count
            print(f"{current_word}\t{current_count}")
            current_word = word
            current_count = count

# Emit the last word
if current_word == word:
    print(f"{current_word}\t{current_count}")
```

(iii) Prepare input data (inputwc.txt)

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Manipal Academy of Higher Education
Department of Data Science and Computer Application
Manipal, Karnataka, India

Load input.txt into Hadoop:

```
hdfs dfs -put inputwc.txt /user/studentregno/lab2
```

(iv) Run the MapReduce Job:

Execute the job using Hadoop Streaming:

```
hadoop jar /usr/local/hadoop/share/hadoop/tools/lib/hadoop-streaming-3.4.1.jar -file  
./mapper.py -mapper 'python3 mapper.py' -file ./reducer.py -reducer 'python3  
reducer.py' -input /user/studentregno/lab2/input.txt -output  
/user/studentregno/lab2/output1
```

```
bdalab@saraswati:~$ hdfs dfs -put input.txt /user/studentregno/lab2/
bdalab@saraswati:~$ hadoop jar /usr/local/hadoop/share/hadoop/tools/lib/hadoop-streaming-3.4.1.jar -file ./mapper.py -mapper 'python3 map
y' -file ./reducer.py -reducer 'python3 reducer.py' -input /user/studentregno/lab2/input.txt -output /user/studentregno/lab2/output1
2025-01-10 15:43:03,579 WARN streaming.StreamJob: -file option is deprecated, please use generic option -files instead.
packageJobJar: [./mapper.py, ./reducer.py, /tmp/hadoop-unjar32849102489634973/] [] /tmp/streamjob9216975962947957627.jar tmpDir=null
2025-01-10 15:43:04,653 INFO client.DefaultNoHARMFaloverProxyProvider: Connecting to ResourceManager at /0.0.0.0:8032
2025-01-10 15:43:05,154 INFO client.DefaultNoHARMFaloverProxyProvider: Connecting to ResourceManager at /0.0.0.0:8032
2025-01-10 15:43:05,488 INFO mapreduce.JobResourceUploader: Disabling Erasure Coding for path: /tmp/hadoop-yarn/staging/bdalab/.staging/j
36491219549_0014
2025-01-10 15:43:06,738 INFO mapred.FileInputFormat: Total input files to process : 1
2025-01-10 15:43:06,951 INFO mapreduce.JobSubmitter: number of splits:2
2025-01-10 15:43:07,156 INFO mapreduce.JobSubmitter: Submitting tokens for job: job_1736491219549_0014
2025-01-10 15:43:07,156 INFO mapreduce.JobSubmitter: Executing with tokens: []
2025-01-10 15:43:07,472 INFO conf.Configuration: resource-types.xml not found
2025-01-10 15:43:07,472 INFO resource.ResourceUtils: Unable to find 'resource-types.xml'.
2025-01-10 15:43:07,583 INFO impl.YarnClientImpl: Submitted application application_1736491219549_0014
2025-01-10 15:43:07,621 INFO mapreduce.Job: The url to track the job: http://saraswati:8088/proxy/application_1736491219549_0014/
2025-01-10 15:43:07,622 INFO mapreduce.Job: Running job: job_1736491219549_0014
2025-01-10 15:43:14,757 INFO mapreduce.Job: Job job_1736491219549_0014 running in uber mode : false
2025-01-10 15:43:14,758 INFO mapreduce.Job: map 0% reduce 0%
2025-01-10 15:43:26,276 INFO mapreduce.Job: map 100% reduce 0%
2025-01-10 15:43:36,375 INFO mapreduce.Job: map 100% reduce 100%
2025-01-10 15:43:36,400 INFO mapreduce.Job: Job job_1736491219549_0014 completed successfully
2025-01-10 15:43:36,503 INFO mapreduce.Job: Counters: 54
  File System Counters
    FILE: Number of bytes read=240
    FILE: Number of bytes written=941639
    FILE: Number of read operations=0
    FILE: Number of large read operations=0
    FILE: Number of write operations=0
    HDFS: Number of bytes read=443
    HDFS: Number of bytes written=164
    HDFS: Number of read operations=11
    HDFS: Number of large read operations=0
    HDFS: Number of write operations=2
    HDFS: Number of bytes read erasure-coded=0
  Job Counters
    Launched map tasks=2
```

(v) View the Results

```
hdfs dfs -cat /user/studentregno/lab2/output1/part-00000
```

```
CPU time spent (ms)=7270
Physical memory (bytes) snapshot=1156423680
Virtual memory (bytes) snapshot=7669239808
Total committed heap usage (bytes)=1174929408
Peak Map Physical memory (bytes)=468488192
Peak Map Virtual memory (bytes)=2556018688
Peak Reduce Physical memory (bytes)=222208000
Peak Reduce Virtual memory (bytes)=2557878272
Shuffle Errors
  BAD_ID=0
  CONNECTION=0
  IO_ERROR=0
  WRONG_LENGTH=0
  WRONG_MAP=0
  WRONG_REDUCE=0
File Input Format Counters
  Bytes Read=231
File Output Format Counters
  Bytes Written=164
2025-01-10 15:43:36,503 INFO streaming.StreamJob: Output directory: /user/studentregno/lab2/output1
bdalab@saraswati:~$ hdfs dfs -cat /user/studentregno/lab2/output1/part-00000
Academy 1
Application 1
Computer 1
Data 1
Department 1
Education 1
Higher 1
India 1
Institute 1
Karnataka, 1
Manipal 3
Manipal, 1
Science 1
Technology 1
and 1
of 3
bdalab@saraswati:~$
```

Week-3 Exercise:

1. Write a Hadoop MapReduce program to count the frequency of each character in a text file.
2. Write a Hadoop MapReduce program to find the maximum temperature for each year in a weather dataset.
3. Write a Hadoop MapReduce program to find the top N records (e.g., top 3 highest scores) from a dataset.
4. Write a Hadoop MapReduce program to calculate the average value for each key in a dataset

Week-4 Advanced Programs on MapReduce

Solved Program: Implement a MapReduce program to perform secondary sorting, where records are sorted by a primary and secondary key.

Write a MapReduce program to Sort a dataset of employee records by department (primary key) and salary (secondary key) in ascending order.

Sample Input:

```
1,HR,Alice,7000
2,IT,Bob,9000
3,HR,Charlie,8000
4,IT,Dave,9500
```

Expected Output:

```
HR,Alice,7000
HR,Charlie,8000
IT,Bob,9000
IT,Dave,9500
```

Solution:

In this program, we'll sort employee records by department (primary key) and salary (secondary key) in ascending order.

Step-1 : Write mapper and reducer program as follows:

Mapper (4s_mapper.py):

- Emits the department and salary as the key.
- Emits the remaining details (employee ID and name) as the value.
- Sorting is achieved by Hadoop's built-in sorting mechanism, which sorts by key during the shuffle and sort phase.

```
#!/usr/bin/env python3
import sys

# Read input line by line
for line in sys.stdin:
    # Strip leading/trailing whitespace
    line = line.strip()

    # Split the line into fields
    fields = line.split(',')

    # Ensure the input format is valid
    if len(fields) == 4:
        employee_id, department, name, salary = fields

        # Emit department and salary as composite key, with other details as value
        print(f"{department}\t{salary}\t{employee_id},{name}")
```

Reducer (4s_reducer.py):

- Group records by department.
- Sort records within each department by salary using Python's sorted function.

```
#!/usr/bin/env python3
import sys
from collections import defaultdict

# Dictionary to store department data
department_data = defaultdict(list)

# Process each line of input
for line in sys.stdin:
    # Strip leading/trailing whitespace
    line = line.strip()

    # Split the line into department, salary, and other details
    try:
        department, salary, details = line.split('\t', 2)
        salary = int(salary) # Convert salary to integer for sorting
        department_data[department].append((salary, details))
    except ValueError:
        continue

# Iterate over each department and sort by salary
for department, records in department_data.items():
    # Sort records by salary
    sorted_records = sorted(records, key=lambda x: x[0])

    # Emit sorted records
    for salary, details in sorted_records:
        print(f"{department},{details}")
```

Step-2: Create input data (4s_input.txt) and copy it into HDFS:

```
1,HR,Eve,6000
2,HR,Alice,7000
3,HR,Charlie,8000
4,IT,Bob,9000
5,IT,Dave,9500
```

To load file on Hadoop:

```
hdfs dfs -mkdir /user/studentregno/lab4
```

```
hdfs dfs -put 4s_input.txt /user/studentregno/lab4
```

Step-3: Run the program on Hadoop

Run the MapReduce Job:

Execute the job using Hadoop Streaming:

```
hadoop jar /usr/local/hadoop/share/hadoop/tools/lib/hadoop-streaming-3.4.1.jar -file
./4s_mapper.py -mapper 'python3 4s_mapper.py' -file ./4s_reducer.py -reducer
'python3 4s_reducer.py' -input /user/studentregno/lab4/4s_input.txt -output
/user/studentregno/lab4/output1
```

```
bdaalab@saraswati:~$ hdfs dfs -put 4s_input.txt /user/studentregno/lab4
bdaalab@saraswati:~$ hadoop jar /usr/local/hadoop/share/hadoop/tools/lib/hadoop-streaming-3.4.1.jar -file ./4s_mapper.py -mapper 'python3 4s_mapper.py' -file ./4s_reducer.py -reducer 'python3 4s_reducer.py' -input /user/studentregno/lab4/4s_input.txt -output /user/studentregno/lab4/output1
2025-01-23 12:58:36,542 WARN streaming.StreamJob: -file option is deprecated, please use generic option -files instead.
packageJobJar: [./4s_mapper.py, ./4s_reducer.py, /tmp/hadoop-unjar3773630761779343200/] [] /tmp/streamjob56249820178330010.jar tmpDir=null
2025-01-23 12:58:37,453 INFO client.DefaultNoHARMFaloverProxyProvider: Connecting to ResourceManager at /0.0.0.0:8032
2025-01-23 12:58:37,714 INFO client.DefaultNoHARMFaloverProxyProvider: Connecting to ResourceManager at /0.0.0.0:8032
2025-01-23 12:58:38,683 INFO mapreduce.JobResourceUploader: Disabling Erasure Coding for path: /tmp/hadoop-yarn/staging/bdaalab/.staging/job_1737616633980_0002
2025-01-23 12:58:42,754 INFO mapred.FileInputFormat: Total input files to process : 1
2025-01-23 12:58:43,986 INFO mapreduce.JobSubmitter: number of splits:2
2025-01-23 12:58:44,813 INFO mapreduce.JobSubmitter: Submitting tokens for job: job_1737616633980_0002
2025-01-23 12:58:44,813 INFO mapreduce.JobSubmitter: Executing with tokens: []
2025-01-23 12:58:45,027 INFO conf.Configuration: resource-types.xml not found
2025-01-23 12:58:45,027 INFO resource.ResourceUtils: Unable to find 'resource-types.xml'.
2025-01-23 12:58:45,102 INFO impl.YarnClientImpl: Submitted application application_1737616633980_0002
2025-01-23 12:58:45,136 INFO mapreduce.Job: The url to track the job: http://saraswati:8088/proxy/application_1737616633980_0002/
2025-01-23 12:58:45,137 INFO mapreduce.Job: Running job: job_1737616633980_0002
2025-01-23 12:58:52,322 INFO mapreduce.Job: Job job_1737616633980_0002 running in uber mode : false
2025-01-23 12:58:52,325 INFO mapreduce.Job: map 0% reduce 0%
2025-01-23 12:58:58,524 INFO mapreduce.Job: map 100% reduce 0%
2025-01-23 12:59:03,571 INFO mapreduce.Job: map 100% reduce 100%
2025-01-23 12:59:06,642 INFO mapreduce.Job: Job job_1737616633980_0002 completed successfully
2025-01-23 12:59:06,751 INFO mapreduce.Job: Counters: 54
File System Counters
  FILE: Number of bytes read=93
  FILE: Number of bytes written=941420
  FILE: Number of read operations=0
  FILE: Number of large read operations=0
  FILE: Number of write operations=0
  HDFS: Number of bytes read=341
  HDFS: Number of bytes written=57
  HDFS: Number of read operations=11
  HDFS: Number of large read operations=0
  HDFS: Number of write operations=2
  HDFS: Number of bytes read erasure-coded=0
```



```

Combine input records=0
Combine output records=0
Reduce input groups=2
Reduce shuffle bytes=99
Reduce input records=5
Reduce output records=5
Spilled Records=10
Shuffled Maps =2
Failed Shuffles=0
Merged Map outputs=2
GC time elapsed (ms)=675
CPU time spent (ms)=3100
Physical memory (bytes) snapshot=1157881856
Virtual memory (bytes) snapshot=7671750656
Total committed heap usage (bytes)=1189609472
Peak Map Physical memory (bytes)=473272320
Peak Map Virtual memory (bytes)=2557829120
Peak Reduce Physical memory (bytes)=213995520
Peak Reduce Virtual memory (bytes)=2557796352
Shuffle Errors
BAD_ID=0
CONNECTION=0
IO_ERROR=0
WRONG_LENGTH=0
WRONG_MAP=0
WRONG_REDUCE=0
File Input Format Counters
Bytes Read=123
File Output Format Counters
Bytes Written=57
2025-01-23 12:59:06,751 INFO streaming.StreamJob: Output directory: /user/studentregno/lab4/output2
bdalab@saraswati:~$ hdfs dfs -cat /user/studentregno/lab4/output2/part-00000
HR,1,Eve
HR,2,Alice
HR,3,Charlie
IT,4,Bob
IT,5,Dave
bdalab@saraswati:~$ 

```

Week-4 Exercise:

- Write a Hadoop MapReduce program to perform a **reduce-side join** to merge two datasets based on a common key.

Hint: Join two datasets:

Students dataset:

StudentID,Name,CourseID

Courses dataset:

CourseID,CourseName,Sem

Perform an **inner join** to output:

StudentID,Name,CourseName,Sem

- Write a Hadoop MapReduce program to implement a distributed **word count** program

that filters out stop words during processing.

Hint: Count the frequency of each word in a text dataset while ignoring stop words (e.g., "is," "the," "and").

Sample Input:

Doc1 This is a sample document

Doc2 Another document with sample text

Expected Output:

Another 1

Document 2

Sample 2

Text 1

- Write a Hadoop MapReduce program to create an **inverted index** that maps each word to the list of documents in which it appears.

Hint: Generate an inverted index for a collection of documents.

Sample Input:

Doc1 Hadoop MapReduce is powerful

Doc2 MapReduce is scalable

Doc3 Hadoop and Spark are popular

Expected Output:

Hadoop Doc1,Doc3

MapReduce Doc1,Doc2

Powerful Doc1

Scalable Doc2

Spark Doc3

Popular Doc3

Week-5 Exploring Pig

What is Apache Pig?

Apache Pig is a high-level data flow scripting language used to process large datasets in Hadoop. It provides an abstraction over MapReduce, allowing users to write simpler scripts instead of complex Java code. Pig uses its own scripting language called Pig Latin, which is similar to SQL but more flexible for data transformations and ETL (Extract, Transform, Load) operations.

Key Features of Apache Pig

- Ease of Use – Uses a simple, SQL-like language (Pig Latin).
- Abstraction Over MapReduce – No need to write complex Java code.
- Scalability – Runs efficiently on large datasets.
- Extensibility – Supports User-Defined Functions (UDFs) in Python and Java.
- Flexibility – Handles both structured and semi-structured data.

Apache Pig vs. MapReduce

Feature	Apache Pig	MapReduce
Language	Pig Latin	Java
Complexity	Simple	Complex
Use Case	ETL, data transformation	Custom logic
Performance	Faster than Hive	Slower

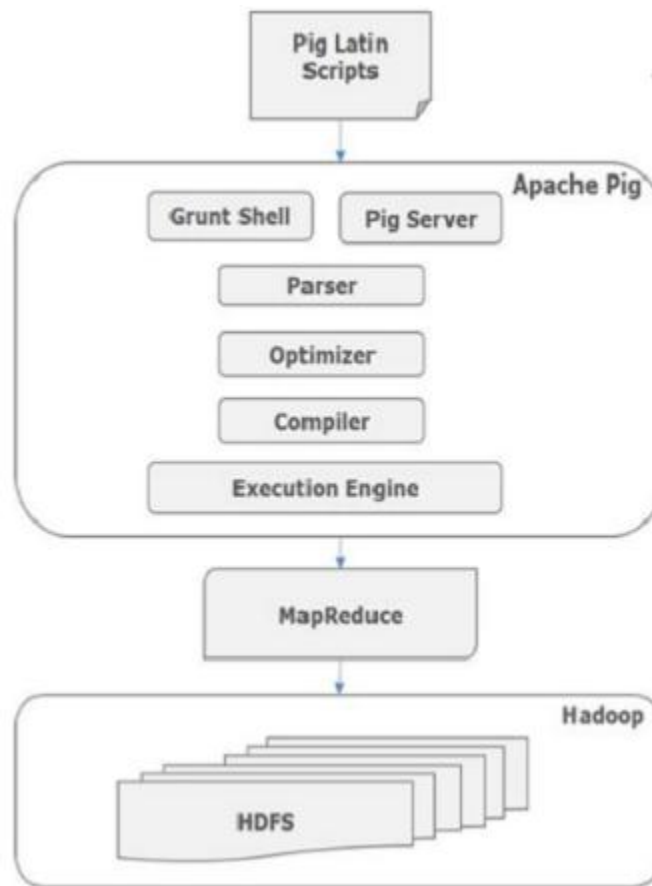
Pig Architecture

Apache Pig consists of:

- Pig Latin Interpreter – Parses and converts scripts into MapReduce jobs.
- Grunt Shell – Interactive command-line interface to run Pig scripts.
- Execution Engine – Converts scripts into optimized execution plans.
- HDFS / Local Mode – Works on Hadoop HDFS or standalone mode.

Modes of Execution:

- Local Mode – Runs without Hadoop (pig -x local).
- MapReduce Mode – Runs on Hadoop (pig -x mapreduce).



When to Use Apache Pig?

- ETL (Extract, Transform, Load) workflows
- Data preprocessing before analysis
- Handling semi-structured data like logs, JSON, CSV
- Quick prototyping compared to MapReduce

Apache Pig is widely used in big data processing pipelines at companies like Yahoo, Twitter, and LinkedIn.

Solved Exercise: First Pig Script: Word Count

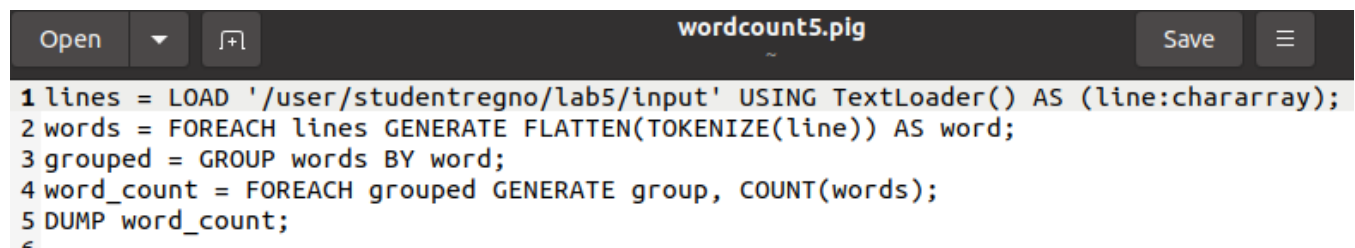
Step 1: Create a Sample Input File (input.txt)

```
Manipal Institute of Technology Manipal  
Manipal Academy of Higher Education  
Department of Data Science and Computer Application  
Manipal, Karnataka, India
```

Load it into hdfs:

```
hdfs dfs -mkdir /user/studentregno/lab5  
hdfs dfs -put input.txt /user /studentregno/lab5
```

Step 2: Write Pig Latin Script and save it as wordcount5.pig



```
Open  wordcount5.pig  Save  ≡  
1 lines = LOAD '/user/studentregno/lab5/input' USING TextLoader() AS (line:chararray);  
2 words = FOREACH lines GENERATE FLATTEN(TOKENIZE(line)) AS word;  
3 grouped = GROUP words BY word;  
4 word_count = FOREACH grouped GENERATE group, COUNT(words);  
5 DUMP word_count;
```

Step 3:

Run the Script

Switch to Pig bin directory

```
cd /usr/local/pig/bin
```

To avoid re-connecting to server, run the following command

```
rm-jobhistory-daemon.sh start historyserver
```

Run the script with following command

```
./pig -x mapreduce /home/bdalab/wordcount5.pig
```

```

bdalab@saraswati1: /usr/local/pig/bin$ ./pig -x mapreduce /home/bdalab/wordcount5.pig
2025-01-31 08:50:56,059 [main] INFO pig.ExecTypeProvider: Trying ExecType : LOCAL
2025-01-31 08:50:56,061 [main] INFO pig.ExecTypeProvider: Trying ExecType : MAPREDUCE
2025-01-31 08:50:56,061 [main] INFO pig.ExecTypeProvider: Picked MAPREDUCE as the ExecType
2025-01-31 08:50:56,116 [main] INFO org.apache.pig.Main - Apache Pig version 0.17.0 (r1797386) compiled Jun 02 2017, 15:41:58
2025-01-31 08:50:56,116 [main] INFO org.apache.pig.Main - Logging error messages to: /usr/local/pig/bin/pig_1738293656109.log
2025-01-31 08:50:56,464 [main] INFO org.apache.pig.impl.util.Util5 - Default bootup file /home/bdalab/.pigbootup not found
2025-01-31 08:50:56,536 [main] INFO org.apache.hadoop.conf.Configuration.deprecation - mapred.job.tracker is deprecated. Instead, use mapredur
ce.jobtracker.address
2025-01-31 08:50:56,536 [main] INFO org.apache.pig.backend.hadoop.executionengine.HExecutionEngine - Connecting to hadoop file system at: hdfs://localhost:9000
2025-01-31 08:50:57,142 [main] INFO org.apache.pig.PigServer - Pig Script ID for the session: PIG-wordcount5.pig-3a570a87-8288-4936-a0ad-0229
95219d85
2025-01-31 08:50:57,142 [main] WARN org.apache.pig.PigServer - ATS is disabled since yarn.timeline-service.enabled set to false
2025-01-31 08:50:57,792 [main] INFO org.apache.pig.impl.util.SpillableMemoryManager - Selected heap (PS Old Gen) of size 699400192 to monitor
, collectionUsageThreshold = 489580128, usageThreshold = 489580128
2025-01-31 08:50:57,962 [main] INFO org.apache.pig.tools.pigstats.ScriptState - Pig features used in the script: GROUP_BY
2025-01-31 08:50:57,989 [main] INFO org.apache.pig.data.SchemaTupleBackend - Key [pig.schematuple] was not set... will not generate code.
2025-01-31 08:50:58,026 [main] INFO org.apache.pig.newplan.logical.optimizer.LogicalPlanOptimizer - {RULES_ENABLED=[AddForEach, ColumnMapKeyP
rune, ConstantCalculator, GroupByConstParallelSetter, LimitOptimizer, LoadTypeCastInserter, MergeFilter, MergeForEach, NestedLimitOptimizer, P
artitionFilterOptimizer, PredicatePushdownOptimizer, PushDownForEachFlatten, PushUpFilter, SplitFilter, StreamTypeCastInserter]}
2025-01-31 08:50:58,136 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MRCompiler - File concatenation threshold: 1
00 optimizer? false
2025-01-31 08:50:58,153 [main] INFO org.apache.pig.backend.hadoop.executionengine.util.CombinerOptimizerUtil - Choosing to move algebraic for
each to combiner
2025-01-31 08:50:58,183 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MultiQueryOptimizer - MR plan size before op
timization: 1
2025-01-31 08:50:58,183 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MultiQueryOptimizer - MR plan size after opt
imization: 1
2025-01-31 08:50:58,339 [main] INFO org.apache.hadoop.yarn.client.DefaultNoHARMFalloverProxyProvider - Connecting to ResourceManager at /0.0.
0.0:8032
2025-01-31 08:50:58,792 [main] INFO org.apache.hadoop.conf.Configuration.deprecation - yarn.resourcemanager.system-metrics-publisher.enabled
is deprecated. Instead, use yarn.system-metrics-publisher.enabled
2025-01-31 08:50:58,816 [main] INFO org.apache.pig.tools.pigstats.mapreduce.MRScriptState - Pig script settings are added to the job
2025-01-31 08:50:58,828 [main] INFO org.apache.hadoop.conf.Configuration.deprecation - mapred.job.reduce.markreset.buffer.percent is deprecate
d. Instead, use mapreduce.reduce.markreset.buffer.percent
2025-01-31 08:50:58,828 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.JobControlCompiler - mapred.job.reduce.markr
eset.buffer.percent is not set, set to default 0.3
2025-01-31 08:50:59,038 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success!
2025-01-31 08:53:59,140 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success!
2025-01-31 08:53:59,152 [main] INFO org.apache.hadoop.conf.Configuration.deprecation - yarn.resourcemanager.system-metrics-publisher.enabled
is deprecated. Instead, use yarn.system-metrics-publisher.enabled
2025-01-31 08:53:59,157 [main] INFO org.apache.pig.data.SchemaTupleBackend - Key [pig.schematuple] was not set... will not generate code.
2025-01-31 08:53:59,196 [main] INFO org.apache.hadoop.mapreduce.lib.input.FileInputFormat - Total input files to process : 1
2025-01-31 08:53:59,196 [main] INFO org.apache.pig.backend.hadoop.executionengine.util.MapRedUtil - Total input paths to process : 1
(of,3)
(and,1)
(Data,1)
(India,1)
(Higher,1)
(Academy,1)
(Manipal,4)
(Science,1)
(Computer,1)
(Education,1)
(Institute,1)
(Karnataka,1)
(Department,1)
(Technology,1)
(Application,1)
2025-01-31 08:53:59,266 [main] INFO org.apache.pig.Main - Pig script completed in 3 minutes, 3 seconds and 352 milliseconds (183352 ms)
bdalab@saraswati1: /usr/local/pig/bin$

```

Working with Datasets in Pig:

Step-1: Loading Data into Pig:

Create a .csv file (employee.csv)

```

101,John,30,HR,50000
102,Alice,25,IT,60000
103,Bob,35,Finance,70000
104,Joy,36,Manager,55000

```

Load it on HDFS

hdfs dfs -put employee.csv /user/studentregno/lab5

Step 2: Write Pig Latin Script and save it as loadcsv.pig

```
1 employees = LOAD '/user/studentregno/lab5/employee.csv' USING PigStorage(',')
2 AS (emp_id:int, name:chararray, age:int, department:chararray, salary:int);
3 DUMP employees;
```

Step-3: Run the script with following command

Loading csv:

```
./pig -x mapreduce /home/bdalab/loadcsv.pig
```

```
2025-01-31 09:07:20,279 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success!
2025-01-31 09:07:20,326 [main] INFO org.apache.hadoop.conf.Configuration.deprecation - yarn.resourcemanager.system-metrics-publisher.enabled
is deprecated. Instead, use yarn.system-metrics-publisher.enabled
2025-01-31 09:07:20,329 [main] INFO org.apache.pig.data.SchemaTupleBackend - Key [pig.schematuple] was not set... will not generate code.
2025-01-31 09:07:20,359 [main] INFO org.apache.hadoop.mapreduce.lib.input.FileInputFormat - Total input files to process : 1
2025-01-31 09:07:20,359 [main] INFO org.apache.pig.backend.hadoop.executionengine.util.MapRedUtil - Total input paths to process : 1
(101,John,30,HR,50000)
(102,Alice,25,IT,60000)
(103,Bob,35,Finance,70000)
(104,Joy,36,Manager,55000)
2025-01-31 09:07:20,517 [main] INFO org.apache.pig.Main - Pig script completed in 3 minutes, 8 seconds and 480 milliseconds (188480 ms)
bdalab@saraswati:/usr/local/pig/bin$
```

Filtering Data: Find employees with a salary greater than 60,000.

Write Pig Latin Script and save it as filter.pig

```
1 employees = LOAD '/user/studentregno/lab5/employee.csv' USING PigStorage(',')
2 AS (emp_id:int, name:chararray, age:int, department:chararray, salary:int);
3 DUMP employees;
4 high_salary = FILTER employees BY salary > 60000;
5 DUMP high_salary;
```

Run the script:

```
./pig -x mapreduce /home/bdalab/filter.pig
```

```
0: Compute warning: aggregation.
2025-01-31 09:19:18,782 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success!
2025-01-31 09:19:18,788 [main] INFO org.apache.pig.data.SchemaTupleBackend - Key [pig.schematuple] was not set... will not generate code.
2025-01-31 09:19:18,815 [main] INFO org.apache.hadoop.mapreduce.lib.input.FileInputFormat - Total input files to process : 1
2025-01-31 09:19:18,815 [main] INFO org.apache.pig.backend.hadoop.executionengine.util.MapRedUtil - Total input paths to process : 1
(103,Bob,35,Finance,70000)
2025-01-31 09:19:18,856 [main] INFO org.apache.pig.Main - Pig script completed in 5 minutes, 52 seconds and 222 milliseconds (352222 ms)
bdalab@saraswati:/usr/local/pig/bin$
```

Sorting Data: Sort employees by age in descending order.

Write Pig Latin Script and save it as sortd.pig

```
1 employees = LOAD '/user/studentregno/lab5/employee.csv' USING PigStorage(',')
2 AS (emp_id:int, name:chararray, age:int, department:chararray, salary:int);
3 DUMP employees;
4 sorted_employees = ORDER employees BY age DESC;
5 DUMP sorted_employees;
```

Run the script:

```
./pig -x mapreduce /home/bdalab/sortd.pig
```

```
2025-01-31 09:33:33,256 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success!
2025-01-31 09:33:33,264 [main] INFO org.apache.pig.data.SchemaTupleBackend - Key [pig.schematuple] was not set... will not generate code.
2025-01-31 09:33:33,294 [main] INFO org.apache.hadoop.mapreduce.lib.input.FileInputFormat - Total input files to process : 1
2025-01-31 09:33:33,294 [main] INFO org.apache.pig.backend.hadoop.executionengine.util.MapRedUtil - Total input paths to process : 1
(104,Joy,36,Manager,55000)
(103,Bob,35,Finance,70000)
(101,John,30,HR,50000)
(102,Alice,25,IT,60000)
2025-01-31 09:33:33,336 [main] INFO org.apache.pig.Main - Pig script completed in 11 minutes, 35 seconds and 88 milliseconds (695088 ms)
bdalab@saraswati1: /usr/local/pig/bin$
```

Aggregations: Average Salary by Department

Write Pig Latin Script and save it as average.pig

```
1 employees = LOAD '/user/studentregno/lab5/employee.csv' USING PigStorage(',')
2 AS (emp_id:int, name:chararray, age:int, department:chararray, salary:int);
3 DUMP employees;
4 grouped_data = GROUP employees BY department;
5 avg_salary = FOREACH grouped_data GENERATE group AS department, AVG(employees.salary) AS
  avg_salary;
6 DUMP avg_salary;
```

Run the script:

```
./pig -x mapreduce /home/bdalab/average.pig
```

```
6 Compute warning aggregation
2025-01-31 09:47:39,527 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success!
2025-01-31 09:47:39,529 [main] INFO org.apache.pig.data.SchemaTupleBackend - Key [pig.schematuple] was not set... will not generate code.
2025-01-31 09:47:39,537 [main] INFO org.apache.hadoop.mapreduce.lib.input.FileInputFormat - Total input files to process : 1
2025-01-31 09:47:39,537 [main] INFO org.apache.pig.backend.hadoop.executionengine.util.MapRedUtil - Total input paths to process : 1
(HR,50000.0)
(IT,60000.0)
(Finance,70000.0)
(Manager,55000.0)
2025-01-31 09:47:39,605 [main] INFO org.apache.pig.Main - Pig script completed in 5 minutes, 50 seconds and 565 milliseconds (350565 ms)
bdalab@saraswati1: /usr/local/pig/bin$
```

Joining Two Datasets

Create a .csv file (department.csv)

```
HR,Bangalore  
IT,Bangalore  
Finance,Mumbai  
Manager,NewYork
```

Load it on HDFS

```
hdfs dfs -put department.csv /user/studentregno/lab5
```

Write Pig Latin Script and save it as join.pig

```
1 employees = LOAD '/user/studentregno/lab5/employee.csv' USING PigStorage(',')  
2 AS (emp_id:int, name:chararray, age:int, department:chararray, salary:int);  
3 departments = LOAD '/user/studentregno/lab5/department.csv' USING PigStorage(',')  
4 AS (dept_name:chararray, location:chararray);  
5 joined_data = JOIN employees BY department, departments BY dept_name;  
6 DUMP joined_data;
```

Run the script:

```
./pig -x mapreduce /home/bdalab/join.pig
```

```
2025-01-31 11:06:41,902 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success!  
2025-01-31 11:06:41,914 [main] INFO org.apache.hadoop.conf.Configuration.deprecation - yarn.resourcemanager.system-metrics-publisher.enabled  
is deprecated. Instead, use yarn.system-metrics-publisher.enabled  
2025-01-31 11:06:41,921 [main] INFO org.apache.pig.data.SchemaTupleBackend - Key [pig.schematuple] was not set... will not generate code.  
2025-01-31 11:06:41,957 [main] INFO org.apache.hadoop.mapreduce.lib.input.FileInputFormat - Total input files to process : 1  
2025-01-31 11:06:41,957 [main] INFO org.apache.pig.backend.hadoop.executionengine.util.MapRedUtil - Total input paths to process : 1  
(101,John,30,HR,50000,HR,Bangalore)  
(102,Alice,25,IT,60000,IT,Bangalore)  
(103,Bob,35,Finance,70000,Finance,Mumbai)  
(104,Joy,36,Manager,55000,Manager,NewYork)  
2025-01-31 11:06:42,104 [main] INFO org.apache.pig.Main - Pig script completed in 3 minutes, 13 seconds and 833 milliseconds (193833 ms)  
bdalab@saraswati1:usr/local/pig/bin$
```

Store Processed Data to HDFS

Write Pig Latin Script and save it as storehdfs.pig

```
1 employees = LOAD '/user/studentregno/lab5/employee.csv' USING PigStorage(',')  
2 AS (emp_id:int, name:chararray, age:int, department:chararray, salary:int);  
3 DUMP employees;  
4 STORE employees INTO '/user/studentregno/lab5/output5/' USING PigStorage(',');  
5
```


Run the script:

```
./pig -x mapreduce /home/bdalab/storehdfs.pig
```

After completion can see the data in HDFS output directory:

```
hdfs dfs -cat /user/studentregno/lab5/output5/part-m-00000
```

Week-5 Exercise (use Mapreduce mode):

1. Create a sales dataset (order_id,product,category,amount). Load it into Pig. Display the first 10 records
2. Load the sales dataset. Filter and display products where the amount is greater than 5000.
3. Load the sales dataset. Sort the products in descending order of amount. Display the sorted results.
4. Load the sales dataset. Group sales by category. Find the total sales amount for each category.
5. Create customer dataset (order_id,customer_name,city). Load sales dataset and customers datasets. Perform a JOIN operation on the order_id column. Display the results.
6. Find the top 3 most expensive products from the sales dataset.
7. Store the filtered high-value transactions (amount > 5000) in an HDFS output directory.

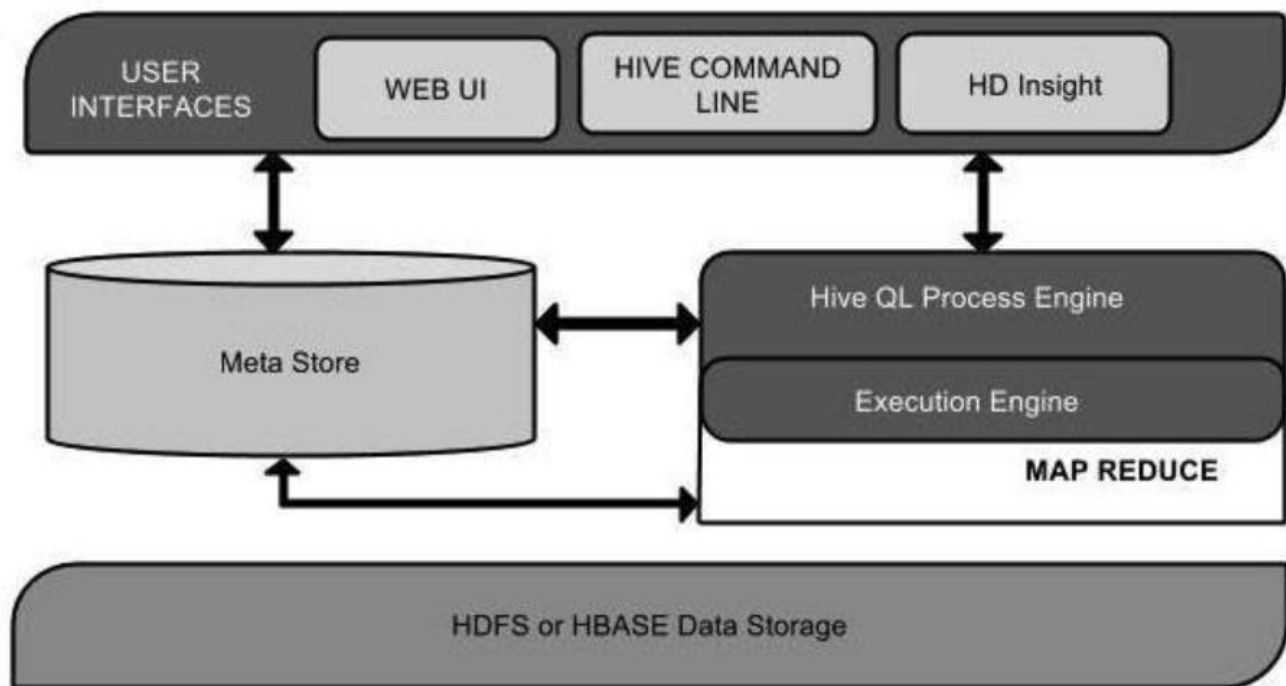
Week-6 Exploring Hive

Apache Hive is a data warehouse infrastructure built on top of Apache Hadoop that enables SQL-like querying on large datasets stored in HDFS (Hadoop Distributed File System). It is primarily used for data analysis, ETL (Extract, Transform, Load) operations, and business intelligence.

Why Use Hive?

- SQL-Like Queries – Uses HiveQL, which is similar to SQL.
- Handles Big Data – Efficiently processes petabytes of structured/unstructured data.
- Scalability – Runs on top of Hadoop, making it highly scalable.
- Schema-on-Read – Unlike traditional databases, data is interpreted at query time.
- Supports Various File Formats – Works with CSV, ORC, Parquet, JSON, and more.

Hive Architecture



Hive consists of several key components:

1. **Metastore** – Stores metadata (schema, table definitions, partitions).
2. **Driver** – Manages query execution lifecycle.
3. **Query Engine** – Converts HiveQL into execution plans.

4. **Execution Engine** – Runs queries using MapReduce, Tez, or Spark.
5. **HDFS** – Stores the actual data.

Key Features of Hive

- **Tables and Partitions** – Data is stored in tables that can be partitioned for faster queries.
- **User-Defined Functions (UDFs)** – Allows custom functions in Java/Python.
- **Joins and Aggregations** – Supports complex queries, joins, and analytics.
- **Optimization Features** – Partitioning, bucketing, indexing, and ORC file format.
- **Integration** – Works with Spark, Presto, Impala, and BI tools like Tableau.

Hive vs. Traditional Databases

Feature	Hive	Traditional RDBMS
Query Language	HiveQL (SQL-like)	SQL
Data Processing	Batch (MapReduce, Tez, Spark)	OLTP (Real-time)
Schema	Schema-on-Read	Schema-on-Write
Speed	Slower for small queries	Fast
ACID Support	Limited (Only in newer versions)	Fully Supported
Scalability	High (Handles petabytes of data)	Moderate

When to Use Hive?

- Processing large-scale data (e.g., logs, clickstream, IoT).
- Running batch analytics on Hadoop.
- ETL workloads and data preprocessing.
- Data exploration and reporting.
- Not ideal for: Real-time transactions, low-latency queries, or updates.

Working with Hive:

Start Hadoop services

```
start-dfs.sh
```

```
start-yarn.sh
```

Execute the following command to give the all permission to hive directory:

```
hdfs dfs -chmod 777 /tmp
```

Start Hive:

```
cd /usr/local/hive
```

```
bin/schematool -dbType derby -initSchema
```

if any error then only run :

```
rm -rf metastore_db
```

If no error: then

```
bin/hive
```

```
bdalab@saraswati:~$ start-dfs.sh
Starting namenodes on [localhost]
Starting datanodes
Starting secondary namenodes [saraswati]
bdalab@saraswati:~$ start-yarn.sh
Starting resourcemanager
Starting nodemanagers
bdalab@saraswati:~$ hive
SLF4J: Class path contains multiple SLF4J bindings.
SLF4J: Found binding in [jar:file:/usr/local/hive/lib/log4j-slf4j-impl-2.17.1.jar!/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: Found binding in [jar:file:/usr/local/hadoop/share/hadoop/common/lib/slf4j-reload4j-1.7.36.jar!/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: See http://www.slf4j.org/codes.html#multiple_bindings for an explanation.
SLF4J: Actual binding is of type [org.apache.logging.slf4j.Log4jLoggerFactory]
Hive Session ID = 8c45fd87-c91d-49b5-b1dc-b2933d47d1b8

Logging initialized using configuration in jar:file:/usr/local/hive/lib/hive-common-3.1.3.jar!/hive-log4j2.properties Async: true
Hive-on-MR is deprecated in Hive 2 and may not be available in the future versions. Consider using a different execution engine (i.e. spark, tez) or using Hive 1.X releases.
hive>
```

Hive Database Operations:

Create a new database

```
CREATE DATABASE sales_db;
```

List databases

```
SHOW DATABASES;
```

Use a database

```
USE sales_db;
```

Show Table list

SHOW TABLES;

Drop a database

DROP DATABASE sales_db CASCADE;

```
hive> show databases;
OK
default
Time taken: 0.05 seconds, Fetched: 1 row(s)
hive> show tables;
OK
Time taken: 0.05 seconds
hive> 
Time taken: 0.05 seconds
hive> CREATE DATABASE sales_db;
OK
Time taken: 0.546 seconds
hive> SHOW DATABASES;
OK
default
sales_db
Time taken: 0.031 seconds, Fetched: 2 row(s)
hive> USE sales_db;
OK
Time taken: 0.024 seconds
hive> DROP DATABASE IF EXISTS test_db CASCADE;
OK
Time taken: 0.018 seconds
hive>
```

Solved Exercise:

E-Commerce Sales Analytics: An e-commerce company wants to analyze sales trends and customer purchases. The data is stored in Hadoop, and they need insights into total revenue, top customers, and monthly sales trends.

Sample data: (sales_data.csv)

transaction_id	customer_id	product	category	amount	payment_type	transaction_date
1	101	Laptop	Electronics	1200	Credit Card	2023-01-05
2	102	Phone	Electronics	800	PayPal	2023-02-10
3	103	T-shirt	Clothing	50	Debit Card	2023-02-15
4	101	Headphones	Electronics	150	Credit Card	2023-03-01
5	104	Shoes	Footwear	120	Cash	2023-03-20

Create sales_data.csv (in any editor) and load it on HDFS

```
1 transaction_id,customer_id,product,category,amount,payment_type,transaction_date
2 1,101,Laptop,Electronics,1200,Credit Card,2023-01-05
3 2,102,Phone,Electronics,800,PayPal,2021-02-10
4 3,103,T-shirt,Clothing,50,Debit Card,2022-02-15
5 4,101,Headphones,Electronics,150,Credit Card,2023-03-01
6 5,104,Shoes,Footwear,120,Cash,2022-03-20
```

```
hdfs dfs -mkdir /user/studentregno/lab6
```

```
hdfs dfs -put sales_data.csv /user/studentregno/lab6
```

a. Create a Hive database and table

```
CREATE DATABASE ecommerce;
```

```
USE ecommerce;
```

```
hive> CREATE DATABASE ecommerce;
OK
Time taken: 0.122 seconds
hive> USE ecommerce;
OK
Time taken: 0.059 seconds
```

```
CREATE TABLE sales ( transaction_id INT, customer_id INT, product STRING, category
STRING, amount FLOAT, payment_type STRING, transaction_date STRING ) ROW
FORMAT DELIMITED FIELDS TERMINATED BY ',' STORED AS TEXTFILE ;
```

To skip the header information:

```
alter table sales set tblproperties ("skip.header.line.count"="1");
```

```

hive> CREATE TABLE sales ( transaction_id INT, customer_id INT, product STRING, category STRING, amount FLOAT, payment_type STRING, transaction_date STRING ) ROW FORMAT DELIMITED FIELDS TERMINATED BY ',' STORED AS TEXTFILE ;
OK
Time taken: 1.537 seconds
hive> alter table sales set tblproperties ("skip.header.line.count"="1");
OK
Time taken: 0.255 seconds
hive> DESCRIBE sales;
OK
transaction_id      int
customer_id         int
product             string
category            string
amount              float
payment_type        string
transaction_date     string
Time taken: 0.171 seconds, Fetched: 7 row(s)
hive> 

```

b. Load Data from HDFS

LOAD DATA INPATH '/user/studentregno/lab6/sales_data.csv' INTO TABLE sales;

```

hive> LOAD DATA INPATH '/user/studentregno/lab6/sales_data.csv' INTO TABLE sales;
Loading data to table ecommerce.sales
OK
Time taken: 1.246 seconds
hive> select *from sales;
OK
1      101      Laptop  Electronics      1200.0  Credit Card      2023-01-05
2      102      Phone   Electronics      800.0   PayPal          2021-02-10
3      103      T-shirt Clothing    50.0   Debit Card      2022-02-15
4      101      Headphones Electronics    150.0   Credit Card      2023-03-01
5      104      Shoes   Footwear        120.0   Cash            2022-03-20
Time taken: 3.634 seconds, Fetched: 5 row(s)
hive> select *from sales where customer_id=101;
OK
1      101      Laptop  Electronics      1200.0  Credit Card      2023-01-05
4      101      Headphones Electronics    150.0   Credit Card      2023-03-01
Time taken: 1.15 seconds, Fetched: 2 row(s)
hive> 

```

c. Analyze Total Revenue:

SELECT SUM(amount) AS total_revenue FROM sales;

```

hive> SELECT SUM(amount) AS total_revenue FROM sales;
Query ID = bdalab_20250207094827_aca6febd-4bf6-43be-b795-68be37a18e2a
Total jobs = 1
Launching Job 1 out of 1
Number of reduce tasks determined at compile time: 1
In order to change the average load for a reducer (in bytes):
  set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
  set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  set mapreduce.job.reduces=<number>
Starting Job = job_1738899093263_0001, Tracking URL = http://saraswati:8088/proxy/application_1738899093263_0001/
Kill Command = /usr/local/hadoop/bin/mapred job -kill job_1738899093263_0001
Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1
2025-02-07 09:48:47,138 Stage-1 map = 0%, reduce = 0%
2025-02-07 09:48:56,736 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 2.89 sec
2025-02-07 09:49:06,357 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 6.32 sec
MapReduce Total cumulative CPU time: 6 seconds 320 msec
Ended Job = job_1738899093263_0001
MapReduce Jobs Launched:
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 6.32 sec HDFS Read: 14392 HDFS Write: 106 SUCCESS
Total MapReduce CPU Time Spent: 6 seconds 320 msec
OK
2320.0
Time taken: 40.997 seconds, Fetched: 1 row(s)
hive> 

```

d. Identify the Most Popular Product

```

SELECT product, COUNT(*) AS sales_count
FROM sales
GROUP BY product
ORDER BY sales_count DESC
LIMIT 1;

```



```

hive> SELECT product, COUNT(*) AS sales_count FROM sales GROUP BY product ORDER BY sales_count DESC LIMIT 1;
Query ID = bdalab_20250207095238_488c25c7-4aac-4607-a278-731200aecc1d
Total jobs = 2
Launching Job 1 out of 2
Number of reduce tasks not specified. Estimated from input data size: 1
In order to change the average load for a reducer (in bytes):
  set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
  set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  set mapreduce.job.reduces=<number>
Starting Job = job_1738899093263_0002, Tracking URL = http://saraswati:8088/proxy/application_1738899093263_0002/
Kill Command = /usr/local/hadoop/bin/mapred job -kill job_1738899093263_0002
Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1
2025-02-07 09:52:50,704 Stage-1 map = 0%, reduce = 0%
2025-02-07 09:53:02,398 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 2.79 sec
2025-02-07 09:53:15,001 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 5.86 sec
MapReduce Total cumulative CPU time: 5 seconds 860 msec
Ended Job = job_1738899093263_0002
Launching Job 2 out of 2
Number of reduce tasks determined at compile time: 1
In order to change the average load for a reducer (in bytes):
  set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
  set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  set mapreduce.job.reduces=<number>
Starting Job = job_1738899093263_0003, Tracking URL = http://saraswati:8088/proxy/application_1738899093263_0003/
Kill Command = /usr/local/hadoop/bin/mapred job -kill job_1738899093263_0003
Hadoop job information for Stage-2: number of mappers: 1; number of reducers: 1
2025-02-07 09:53:31,787 Stage-2 map = 0%, reduce = 0%
2025-02-07 09:53:37,052 Stage-2 map = 100%, reduce = 0%, Cumulative CPU 2.42 sec
2025-02-07 09:53:43,387 Stage-2 map = 100%, reduce = 100%, Cumulative CPU 5.27 sec
MapReduce Total cumulative CPU time: 5 seconds 270 msec
Ended Job = job_1738899093263_0003
MapReduce Jobs Launched:
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 5.86 sec HDFS Read: 13801 HDFS Write: 224 SUCCESS
Stage-Stage-2: Map: 1 Reduce: 1 Cumulative CPU: 5.27 sec HDFS Read: 7817 HDFS Write: 112 SUCCESS
Total MapReduce CPU Time Spent: 11 seconds 130 msec
OK
Headphones 1
Time taken: 66.952 seconds, Fetched: 1 row(s)
hive> 

```

e. Analyze Sales by Payment Method

SELECT payment_type, COUNT(*) AS transaction_count

FROM sales

GROUP BY payment_type;

```

hive> SELECT payment_type, COUNT(*) AS transaction_count FROM sales GROUP BY payment_type;
Query ID = bdalab_20250207095915_de4fff6f-5589-4bf4-bcd9-d544983ba160
Total jobs = 1
Launching Job 1 out of 1
Number of reduce tasks not specified. Estimated from input data size: 1
In order to change the average load for a reducer (in bytes):
  set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
  set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  set mapreduce.job.reduces=<number>
Starting Job = job_1738899093263_0004, Tracking URL = http://saraswati:8088/proxy/application_1738899093263_0004/
Kill Command = /usr/local/hadoop/bin/mapred job -kill job_1738899093263_0004
Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1
2025-02-07 09:59:25,737 Stage-1 map = 0%, reduce = 0%
2025-02-07 09:59:37,833 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 8.88 sec
2025-02-07 09:59:46,173 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 11.72 sec
MapReduce Total cumulative CPU time: 11 seconds 720 msec
Ended Job = job_1738899093263_0004
MapReduce Jobs Launched:
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 11.72 sec HDFS Read: 14683 HDFS Write: 178 SUCCESS
Total MapReduce CPU Time Spent: 11 seconds 720 msec
OK
Cash 1
Credit Card 2
Debit Card 1
PayPal 1
Time taken: 35.547 seconds, Fetched: 4 row(s)
hive> 

```

f. Optimizing the Table with Partitioning:

For large datasets, partitioning helps speed up queries.

Enable Dynamic Partitioning:

```
SET hive.exec.dynamic.partition = true;
```

```
SET hive.exec.dynamic.partition.mode = nonstrict;
```

Create partition table (sales table by year):

```
CREATE TABLE sales_partitioned (transaction_id INT, customer_id INT, product STRING,  
category STRING, amount FLOAT, payment_type STRING) PARTITIONED BY (year INT)  
STORED AS TEXTFILE;
```

Load Data into Partitions:

```
INSERT INTO sales_partitioned PARTITION (year) SELECT transaction_id, customer_id,  
product, category, amount, payment_type, YEAR(TO_DATE(transaction_date)) AS year FROM  
sales;
```

```
hive> SET hive.exec.dynamic.partition = true;  
hive> SET hive.exec.dynamic.partition.mode = nonstrict;  
hive> INSERT INTO sales_partitioned PARTITION (year) SELECT transaction_id, customer_id, product, category, amount, payment_type, YEAR(TO_DATE  
(transaction_date)) AS year FROM sales;  
Query ID = bdalab_20250214090200_9d71c0bd-2fb1-4242-82a8-f5e1ab977db4  
Total jobs = 3  
Launching Job 1 out of 3  
Number of reduce tasks not specified. Estimated from input data size: 1  
In order to change the average load for a reducer (in bytes):  
  set hive.exec.reducers.bytes.per.reducer=<number>  
In order to limit the maximum number of reducers:  
  set hive.exec.reducers.max=<number>  
In order to set a constant number of reducers:  
  set mapreduce.job.reduces=<number>  
Starting Job = job_1739501325477_0002, Tracking URL = http://saraswati:8088/proxy/application_1739501325477_0002/  
Kill Command = /usr/local/hadoop/bin/mapred job -kill job_1739501325477_0002  
Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1  
2025-02-14 09:02:11,340 Stage-1 map = 0%, reduce = 0%  
2025-02-14 09:02:22,845 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 5.71 sec  
2025-02-14 09:02:31,172 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 8.58 sec  
MapReduce Total cumulative CPU time: 8 seconds 580 msec  
Ended Job = job_1739501325477_0002  
Stage-4 is selected by condition resolver.  
Stage-3 is filtered out by condition resolver.  
Stage-5 is filtered out by condition resolver.  
Moving data to directory hdfs://localhost:9000/user/hive/warehouse/ecommerce.db/sales_partitioned/.hive-staging_hive_2025-02-14_09-02-00_504_5  
112267488260422543-1/-ext-10000  
Loading data to table ecommerce.sales_partitioned partition (year=null)  
  
Time taken to load dynamic partitions: 0.787 seconds  
Time taken for adding to write entity : 0.003 seconds  
MapReduce Jobs Launched:  
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 8.58 sec HDFS Read: 22643 HDFS Write: 1293 SUCCESS  
Total MapReduce CPU Time Spent: 8 seconds 580 msec  
OK  
Time taken: 35.241 seconds  
hive>
```

```
SHOW PARTITIONS sales_partitioned;
```

Query the partitioned Table:

```
SELECT * FROM sales_partitioned WHERE year = 2023;
```

```
hive> SHOW PARTITIONS sales_partitioned;
OK
year=2021
year=2022
year=2023
Time taken: 0.165 seconds, Fetched: 3 row(s)
hive> SELECT * FROM sales_partitioned WHERE year = 2023;
OK
1      101      Laptop Electronics      1200.0  Credit Card      2023
4      101      Headphones Electronics      150.0   Credit Card      2023
Time taken: 0.725 seconds, Fetched: 2 row(s)
```

g. bucketed table:

Bucketing in Hive is used to **divide data into fixed-size buckets** based on a column's hash value. This improves **query performance** by reducing the number of files scanned.

Create a bucketed table:

- **CLUSTERED BY (customer_id)** → Used for bucketing
- **INTO 4 BUCKETS** → Data will be divided into 4 buckets

```
CREATE TABLE retail_sales_bucketed (
```

```
    transaction_id INT,
    customer_id INT,
    product STRING,
    category STRING,
    amount FLOAT,
    payment_type STRING
    transaction_date STRING
```

```
)
```

```
CLUSTERED BY (customer_id) INTO 4 BUCKETS STORED AS TEXTFILE;
```

```
Time taken: 0.184 seconds
hive> CREATE TABLE sales_bucketed (transaction_id INT, customer_id INT, product STRING, category STRING, amount FLOAT, payment_type STRING, transaction_date STRING) CLUSTERED BY (customer_id) INTO 4 BUCKETS STORED AS TEXTFILE;
OK
Time taken: 0.184 seconds
hive> show tables;
OK
retail_sales_partitioned
sales
sales_bucketed
sales_partitioned
Time taken: 0.05 seconds, Fetched: 4 row(s)
hive>
```

Enable Bucketing in Hive:

SET hive.enforce.bucketing = true;

Load Data into Bucketed Table:

INSERT OVERWRITE TABLE sales_bucketed SELECT * FROM sales SORT BY customer_id;

```
hive> SET hive.enforce.bucketing = true;
hive> INSERT OVERWRITE TABLE sales_bucketed SELECT * FROM sales SORT BY customer_id;
Query ID = bdalab_20250214093659_20f84efd-6aed-488a-8da3-8307ed7ba294
Total jobs = 2
Launching Job 1 out of 2
Number of reduce tasks determined at compile time: 4
In order to change the average load for a reducer (in bytes):
  set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
  set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  set mapreduce.job.reduces=<number>
Starting Job = job_1739501325477_0003, Tracking URL = http://saraswati:8088/proxy/application_1739501325477_0003/
Kill Command = /usr/local/hadoop/bin/mapred job -kill job_1739501325477_0003
Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 4
2025-02-14 09:37:13,720 Stage-1 map = 0%, reduce = 0%
2025-02-14 09:37:25,203 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 2.75 sec
2025-02-14 09:37:43,326 Stage-1 map = 100%, reduce = 25%, Cumulative CPU 7.68 sec
2025-02-14 09:37:46,524 Stage-1 map = 100%, reduce = 50%, Cumulative CPU 12.47 sec
2025-02-14 09:37:48,635 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 22.39 sec
MapReduce Total cumulative CPU time: 22 seconds 390 msec
Ended Job = job_1739501325477_0003
Loading data to table ecommerce.sales_bucketed
Launching Job 2 out of 2
Number of reduce tasks determined at compile time: 1
In order to change the average load for a reducer (in bytes):
  set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
  set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  set mapreduce.job.reduces=<number>
Starting Job = job_1739501325477_0004, Tracking URL = http://saraswati:8088/proxy/application_1739501325477_0004/
Kill Command = /usr/local/hadoop/bin/mapred job -kill job_1739501325477_0004
Hadoop job information for Stage-3: number of mappers: 1; number of reducers: 1
2025-02-14 09:38:33,280 Stage-3 map = 0%, reduce = 0%
2025-02-14 09:38:40,596 Stage-3 map = 100%, reduce = 0%, Cumulative CPU 2.68 sec
2025-02-14 09:38:48,951 Stage-3 map = 100%, reduce = 100%, Cumulative CPU 5.95 sec
MapReduce Total cumulative CPU time: 5 seconds 950 msec
Ended Job = job_1739501325477_0004
MapReduce Jobs Launched:
Stage-Stage-1: Map: 1 Reduce: 4 Cumulative CPU: 22.39 sec HDFS Read: 49645 HDFS Write: 1854 SUCCESS
Stage-Stage-3: Map: 1 Reduce: 1 Cumulative CPU: 5.95 sec HDFS Read: 18098 HDFS Write: 490 SUCCESS
Total MapReduce CPU Time Spent: 28 seconds 340 msec
OK
Time taken: 113.366 seconds
hive> █
```

Querying Data Using Buckets:

```
SELECT * FROM sales_bucketed
```

```
TABLESAMPLE(BUCKET 2 OUT OF 4 ON customer_id);
```

(**BUCKET 2 OUT OF 4** → Fetches data only from bucket 2)

```
hive> SELECT * FROM sales_bucketed TABLESAMPLE(BUCKET 2 OUT OF 4 ON customer_id);
OK
2      102      Phone  Electronics      800.0    PayPal  2021-02-10
3      103      T-shirt Clothing      50.0    Debit Card      2022-02-15
Time taken: 0.252 seconds, Fetched: 2 row(s)
hive> □
```

Week-6 Exercises:

1. A tech company wants to analyze web server logs to monitor website traffic, detect peak usage times, and track 404 errors. Do the Log analysis of server for the dataset (weblogs.csv)

log_id	ip_address	url	status_code	response_time	log_date
1	192.168.1.1	/home	200	120ms	2024-01-01
2	192.168.1.2	/product	404	60ms	2024-01-01
3	192.168.1.3	/checkout	500	200ms	2024-01-02

- a) Create a Hive table for log data. Load Data from HDFS
 - b) Find the Most Visited URLs
 - c) Detect Peak Traffic Hours
 - d) Find 404 Errors
 - e) Create a partitioned Hive table based on date. How would you insert data dynamically into partitions?
 - f) Create a Bucketed Table by status code. Load data into the bucketed table. Retrieve 404 errors (Bucketed Table)
2. Analyze movie ratings, identify top movies, and understand user behavior using **IMDb-style data**.
Dataset: movie_ratings.csv

```
movie_id,title,genre,release_year,user_id,rating,timestamp
101,The Dark Knight,Action,2008,5001,9.0,2024-01-01 12:30:00
102,Inception,Sci-Fi,2010,5002,8.8,2024-01-01 14:00:00
103,Parasite,Thriller,2019,5003,8.6,2024-01-02 16:45:00
104,Interstellar,Sci-Fi,2014,5004,8.6,2024-01-03 10:30:00
105,Avengers: Endgame,Action,2019,5005,8.4,2024-01-04 09:00:00
106,Spirited Away,Animation,2001,5006,8.6,2024-01-05 22:15:00
```

- a) Create a Hive table for movie. Load Data from HDFS
 - b) Find the Top 5 Highest Rated Movies
 - c) Find the Most Popular Genre
 - d) Extract Yearly Trends (Extract Year from Review Date)
 - e) Create a Partitioned Table by Genre. Load Data into Partitions. Retrieve All Sci-Fi Movies (from Partitioned Table)
 - f) Create a Bucketed Table by release_year. Load Data. Retrieve Ratings of Movies Released in 2019 (Bucketed Table)
3. Analyze flight delays, on-time performance, and airport efficiency using real-world airline data.
Dataset: flight_data.csv

```
flight_id, airline, flight_number, origin, destination, departure_date, departure_time, arrival_time,
delay_minutes, status
101, Delta, DL202, JFK, LAX, 2024-02-01, 08:30, 11:30, 10, On Time
102, United, UA405, ORD, SFO, 2024-02-01, 09:00, 12:00, 45, Delayed
103, American, AA100, LAX, MIA, 2024-02-02, 10:00, 14:00, 0, On Time
104, Southwest, SW305, DFW, DEN, 2024-02-02, 11:30, 13:30, 30, Delayed
105, JetBlue, JB678, BOS, ATL, 2024-02-03, 07:00, 09:30, 5, On Time
```

- a) Create a Hive table for flight data. Load Data from HDFS
- b) Find the Airline with the Most Delays
- c) Count the Number of Flights by Status (On Time, Delayed)
- d) Calculate the Average Delay per Airline
- e) Create a Partitioned Table by Departure Date. Load Data into Partitions.
- f) Create a Bucketed Table by Airline. Load data into the bucketed table. Retrieve Flights Operated by Delta (Bucketed Table)