# MapReduce Program and Deployment in Cloud

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# **Outline**

- 1. Introduction to MapReduce
- 2. Writing a MapReduce Program
- 3. Hadoop Ecosystem & Running MapReduce Jobs
- 4. MapReduce Cloud Deployment
- 5. Deploying MapReduce on Cloud
- 6. Challenges, Optimization & Future Trends

# Introduction to MapReduce

**Sri Dev S (22z262)** 

# What is MapReduce?

- A programming model for processing large datasets in parallel.
- Divides tasks into smaller sub-tasks and distributes them across a cluster.
- Consists of Map, Shuffle, and Reduce phases.
- Working Principle:
  - Splits data into smaller chunks.
  - Processes them in parallel.
  - Combines results to produce final output.

# **History & Significance**

- 2004: Developed by Google (Jeffrey Dean & Sanjay Ghemawat).
- Inspired by functional programming concepts (*map* and *reduce*).
- 2006 :Apache Hadoop adopted MapReduce for large-scale data processing.

# Key Benefits:

- Handles large-scale data.
- Fault-tolerant and scalable.
- Runs on standard hardware.

# **Key Concepts – Map, Shuffle, Reduce**

## Map Phase:

- The input data is broken into small parts and sent to multiple computers.
- Each part is processed separately by a Mapper function.
- The Mapper converts the data into key-value pairs
- Example (Word Count):

Input: "Hadoop is fast. Hadoop is scalable."

Output: ("Hadoop", 1), ("is", 1), ("fast.", 1), ("Hadoop", 1), ("is", 1), ("scalable.", 1)

# **Key Concepts – Map, Shuffle, Reduce**

## **Shuffle & Sort Phase:**

- The system groups all key-value pairs that have the same key.
- It also **sorts** them so they are easy to process in the next step.

## Input (Before Shuffle & Sort):

```
("Hadoop", 1), ("Hadoop", 1), ("is", 1), ("is", 1), ("fast.", 1), ("scalable.", 1)
```

## **Output (After Shuffle & Sort):**

```
("Hadoop", [1,1]), ("is", [1,1]), ("fast.", [1]), ("scalable.", [1])
```

# **Key Concepts – Map, Shuffle, Reduce**

## **Reduce Phase:**

- The **Reducer function** adds up all the values for each key.
- It produces the **final result** with the total count for each word.

## Input (Grouped Data):

```
("Hadoop", [1,1]), ("is", [1,1]), ("fast.", [1]), ("scalable.", [1])
```

## **Final Output (Word Count):**

```
("Hadoop", 2), ("is", 2), ("fast.", 1), ("scalable.", 1)
```

# **Real-World Use Cases**

- Search Engines (Google, Bing, Yahoo) Web page indexing.
- Social Media (Facebook, Twitter) Trend analysis.
- Fraud Detection (Banks) Transaction monitoring.
- Healthcare Medical data analysis.
- Retail (Amazon, Walmart) Customer behavior insights.

# Writing a MapReduce Program

**Vishnu Barath K(22z274)** 

# Structure of a MapReduce Job

## **Reduce Phase:**

- The **Reducer function** adds up all the values for each key.
- It produces the **final result** with the total count for each word.

## Input (Grouped Data):

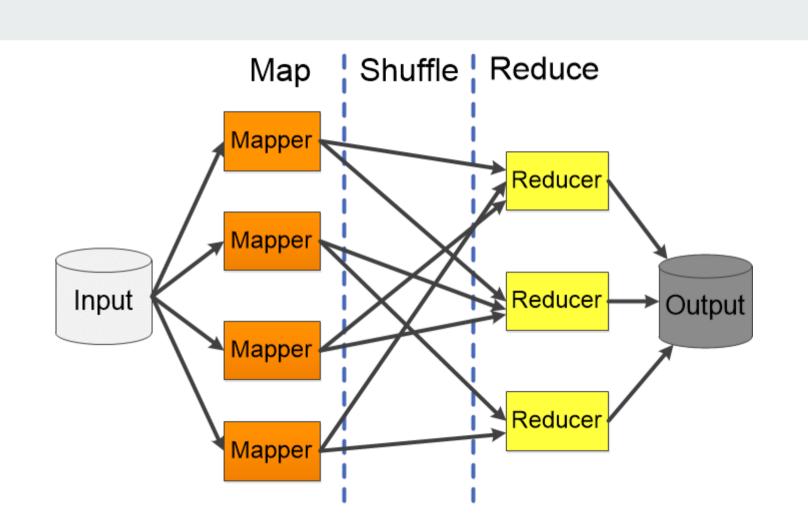
```
("Hadoop", [1,1]), ("is", [1,1]), ("fast.", [1]), ("scalable.", [1])
```

## **Final Output (Word Count):**

```
("Hadoop", 2), ("is", 2), ("fast.", 1), ("scalable.", 1)
```

# Structure of a MapReduce Job

- Input Data: Stored in HDFS (Hadoop Distributed File System)
- Mapper Phase: Processes input data and generates key-value pairs
- Shuffle and Sort: Groups similar keys together
- Reducer Phase: Aggregates and processes mapped data
- Output Data: Stored back in HDFS



# Writing the Mapper Class

The Mapper class processes input and emits key-value pairs

#### Implementation:

```
1 - public static class TokenizerMapper extends Mapper<Object, Text, Text,
        IntWritable> {
        private final static IntWritable one = new IntWritable(1);
        private Text word = new Text();
        public void map(Object key, Text value, Context context) throws IOException,
            InterruptedException {
            StringTokenizer itr = new StringTokenizer(value.toString());
 6
 7 +
           while (itr.hasMoreTokens()) {
                word.set(itr.nextToken());
                context.write(word, one);
10
11
12 }
```

# **Writing the Reducer Class**

The Reducer class processes the key-value pairs and aggregates results

#### Implementation

```
1 - public static class IntSumReducer extends Reducer<Text, IntWritable, Text,
       IntWritable> {
       private IntWritable result = new IntWritable();
       public void reduce(Text key, Iterable<IntWritable> values, Context context)
4 -
            throws IOException, InterruptedException {
           int sum = 0:
           for (IntWritable val : values) {
6 +
                sum += val.get();
            result.set(sum);
            context.write(key, result);
10
12 }
```

# **Input and Output Formats**

## **Input Format:**

- TextInputFormat (Default): Reads text files line by line
- **KeyValueTextInputFormat:** Splits data based on a delimiter
- SequenceFileInputFormat: Reads binary key-value pairs

## **Output Format:**

- **TextOutputFormat (Default):** Writes plain text
- SequenceFileOutputFormat: Writes binary output

# **Word Count Example:**

```
public class WordCount {
  public static class TokenizerMapper extends Mapper<Object, Text, Text, IntWritable> {
     private final static IntWritable one = new IntWritable(1);
    private Text word = new Text();
    public void map(Object key, Text value, Context context) throws IOException, InterruptedException {
       StringTokenizer itr = new StringTokenizer(value.toString());
       while (itr.hasMoreTokens()) {
         word.set(itr.nextToken());
         context.write(word, one);
  public static class IntSumReducer extends Reducer<Text, IntWritable, Text, IntWritable> {
     private IntWritable result = new IntWritable();
    public void reduce(Text key, Iterable<IntWritable> values, Context context) throws IOException, InterruptedException {
       int sum = 0;
       for (IntWritable val : values) {
         sum += val.get();
       result.set(sum);
       context.write(key, result);
```

# **Word Count Code**

```
public static void main(String[] args) throws Exception {
    Configuration conf = new Configuration();
    Job job = Job.getInstance(conf, "word count");
    FileInputFormat.addInputPath(job, new Path(args[0]));
    FileOutputFormat.setOutputPath(job, new Path(args[1]));
    System.exit(job.waitForCompletion(true) ? 0 : 1);
    }
}
```

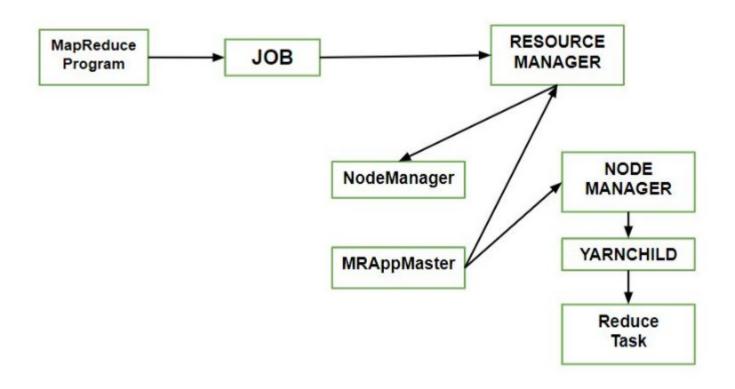
# Running & Monitoring MapReduce Jobs Hadoop (Single Vs Multi-node cluster)

Vivekanand M U (22z275)

# Overview of Hadoop and Its Components

- Hadoop is an open-source framework designed for the distributed processing of large datasets across clusters of computers using simple programming models. The core components of a hadoop ecosystem include:
  - HDFS NameNode
  - HDFS DataNode
  - YARN NodeManager
  - YARN ResourceManager
  - MapReduce

# **Running MapReduce Jobs:**



# **Submitting and Monitoring MapReduce Jobs:**

**Job** is the primary interface by which user-job interacts with the ResourceManager.

Job provides facilities to submit jobs, track their progress, access component-tasks reports and logs.

The job submission process involves:

- 1. Checking the **input and output specifications** of the job.
- 2. Computing the **InputSplit** values for the job.
- 3. Setting up the requisite accounting information for the **DistributedCache** of the job, if necessary.
- 4. Copying the job's jar and configuration to the MapReduce system directory on the FileSystem.
- 5. Submitting the job to the ResourceManager and optionally monitoring its status.

**History logs:** mapred job -history output.jhist

**Job.submit()**: Submit the job to the cluster and return immediately. **Job.waitForCompletion(boolean)**: Submit the job to the cluster and wait for it to finish.

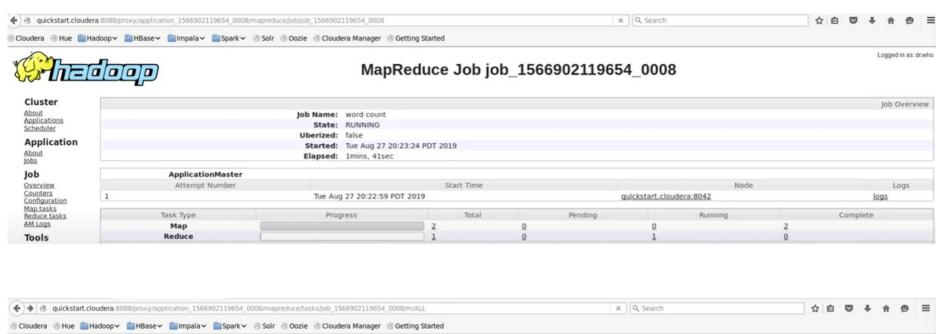
## **Monitor MapReduce Jobs using Cloudera:**

- Cloudera is a company that provides Hadoop-based big data solutions.
- It offers Cloudera Distribution for Hadoop (CDH), a Hadoop ecosystem with tools like HDFS, YARN, Hive, and Spark.
- It simplifies Hadoop cluster management and enhances enterprise security and performance.
- It can be local if installed on a single-node cluster using Cloudera QuickStart VM.

[cloudera@quickstart hadoop\_codes\_simple\_wc]\$ yarn jar wc.jar WordCount /user/cloudera/data/wc\_3/input /user/cloudera/hadoop/data/wc\_3/output\_rec\_part\_3 19/08/27 20:22:45 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0:8032

```
19/08/27 20:22:49 INFO mapreduce.JobSubmitter: number of splits:2
19/08/27 20:22:50 INFO mapreduce.JobSubmitter: Submitting tokens for job: job_1566902119654_0008
19/08/27 20:22:51 INFO impl.YarnClientImpl: Submitted application application 1566902119654_0008
19/08/27 20:22:51 INFO mapreduce.Job: The url to track the job: http://quickstart.cloudera:8088/proxy/application_1566902119654_0008/
19/08/27 20:22:51 INFO mapreduce.Job: Running job: job_1566902119654_0008
^C[cloudera@quickstart hadoop_codes_simple_wc]$ http://quickstart.cloudera:8088/proxy/application_1566902119654_0008/
```

# **Monitor MapReduce Jobs using Cloudera:**



#### Logged in as: dr.who Map Tasks for job\_1566902119654\_0008 ▶ Cluster Show 20 - entries Search: · Application Task Status State Start Time Finish Time Elapsed Time Progress - Job task 1566902119654 0008 m 000000 map SUCCEEDED Tue Aug 27 20:23:30 -0700 2019 Tue Aug 27 20:24:42 -0700 2019 1mins, 12sec Overview task 1566902119654 0008 m 000001 map SUCCEEDED Tue Aug 27 20:23:30 -0700 2019 Tue Aug 27 20:24:42 -0700 2019 1mins, 12sec Counters Showing 1 to 2 of 2 entries Configuration

# **Configuration (Core XML Files)**

## **Core-site.xml: (Hadoop Configuration)**

Specifies the HDFS NameNode location (fs.defaultFS property), which is the entry point for the entire Hadoop filesystem.

## Hdfs-site.xml: (HDFS Configuration)

Defines the replication factor for data blocks, controlling how many copies of your data are stored across the cluster for fault tolerance.

# **Configuration (Core XML Files)**

### mapred-site.xml (MapReduce Configuration):

Configures which framework will be used for processing data (typically YARN), determining how MapReduce jobs are executed.

### Yarn-site.xml: (YARN Configuration)

Specifies the auxiliary services needed for applications, particularly the shuffle operation that's critical for MapReduce jobs to function.

```
<name>yarn.resourcemanager.hostname
```

# **Word Count Example: (Single-node)**

- The input file (input.txt) is stored locally on disk (not in HDFS).
- The whole file is processed as one logical block (since there is no HDFS splitting).
- The single node runs the entire Mapper logic.

```
(hadoop, 1)
(mapreduce, 1)
(hadoop, [1, 1, 1])
(hadoop, 1)
(spark, 1)
(hdfs, [1])
(hdfs, 1)
(hdfs, 1)
(spark, 1)
(spark, 1)
```

The final result is stored in file:///output/part-r-00000.

# **Word Count Example: (Multi-node)**

- The input file (input.txt) is uploaded to HDFS.
- HDFS splits the file into blocks (default 128MB or 256MB).
- Blocks are replicated across 3 worker nodes for fault tolerance.
- Each block is processed independently on different nodes. Each worker node runs its own Mapper tasks.

## 1. Block Splitting & Map Phase:

Node 1 Mapper:	Node 2 Mapper:	Node 3 Mapper:
(hadoop, 1) (mapreduce, 1)	(hadoop, 1) (spark, 1) (hadoop, 1)	(hdfs, 1) (spark, 1)

## 2. Partition Phase:

partition = key.hashCode() % numReducers

```
"hadoop" hash \rightarrow Partition 0 , "mapreduce" hash \rightarrow Partition 1 "spark" hash \rightarrow Partition 2 , "hdfs" hash \rightarrow Partition 0
```

# **Word Count Example: (Multi-node)**

## 3. Shuffle and Sort:

Reducer 0 receives:	Reducer 1 receives:	Reducer 2 receives:
After sorting: (hadoop, 1), (hadoop, 1), (hadoop, 1), (hdfs, 1)	(mapreduce, 1)	After sorting: (spark, 1), (spark, 1)

# 4. Reducer Execution

### (Distributed):

- Each Reducer runs on a separate node.
- The final result is written back to HDFS.
- The output is stored in multiple partitioned files (part-r-00000, part-r-00001, etc.).

# Hadoop: Single-Node vs Multi-Node Configuration

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Configuration	Single-Node	Multi-Node	Why the Difference?	

fs.defaultFS = hdfs://master:9000

mapreduce.framework.name = yarn

yarn.resourcemanager.hostname =

master

dfs.replication = 3

File

core-site.xml

hdfs-site.xml

mapred-

site.xml

yarn-site.xml

fs.defaultFS =

hdfs://localhost:9000

dfs.replication = 1

Not required

mapreduce.framework.name = local

filesystem.

across different nodes for fault tolerance.

using YARN resource management.

resource allocation across nodes.

Multi-node uses a remote **master** node instead of **localhost**, allowing distributed access to the

Single-node doesn't need data replication, while multi-node maintains 3 copies of data blocks

Single-node processes jobs locally, while multi-node distributes computation across the cluster

YARN resource manager is essential for multi-node clusters to coordinate job scheduling and

# MapReduce Cloud Deployment

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# What is Cloud Deployment?

- Cloud deployment means running applications and services on cloud infrastructure instead of local servers.
- It provides on-demand access to computing resources like storage, processing power, and networking.
- Used for scalability, cost-efficiency, and flexibility.
- Cloud providers handle hardware, updates, and security, reducing operational overhead.
- Applications can be **accessed from anywhere**, ensuring collaboration and remote work.

# Why deploy MapReduce in the Cloud? MapReduce processes large datasets, and cloud platforms help by providing:

- Scalability Handle increasing data volume dynamically.
- Cost Efficiency Pay only for what you use (No upfront investment).
- Managed Services Cloud providers handle infrastructure setup.
- **Faster Processing** Distributed resources improve performance.

# Running Hadoop on the Cloud

#### What's wrong with on-premise Hadoop?

#### **How does running Hadoop on the Cloud help?**



#### Rigid and unscalable infrastructure

- Buy too much: waste money & have idle resources lying around
- Buy too little: slow performance



### Flexible to changes

- Quickly obtain what you need, resize & tear down what you no longer need
- Pay as you go



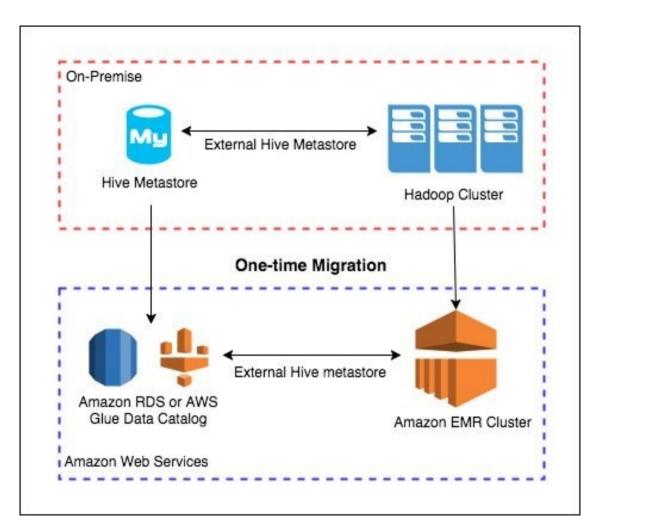
#### **Time-consuming setup & maintenance**

- Purchasing and configuring new computers for Hadoop clusters take too long
- Fixing or replacing malfunctioned hardware could impact business operations



#### Managed setup & maintenance

- Launch and allocate new resources for Hadoop needs on the fly
- Hardware failure is managed by cloud providers



# **Cloud Platforms Supporting**

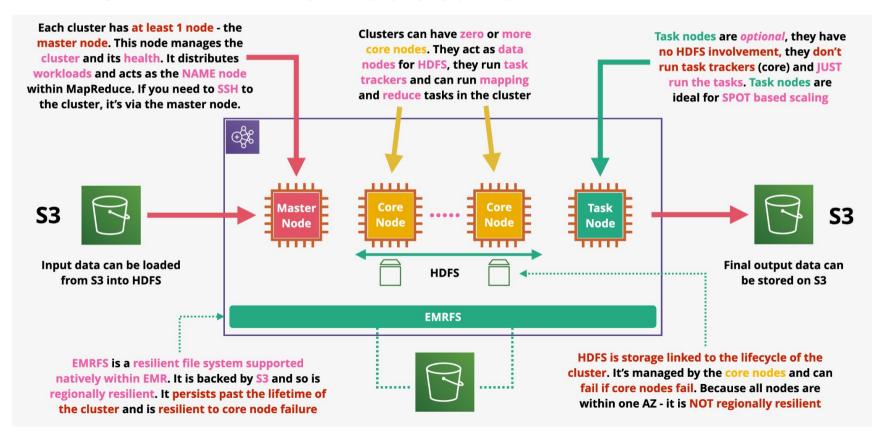
Amazon Web	Microsoft Azure	Google Cloud Platform
Services		
EMR	HDInsight	Cloud Dataflow
Kinesis	Stream Analytics, Data Lake Analytics, Data Lake Store	Cloud Dataprep (Beta)
Glue	Data Factory, Data Category	Cloud Dataproc
Athena	Data Catalog	Cloud Datalab
QuickSight	Power BI, Power BI Embedded	Genomics
ElasticSearch Service	Marketplace - ElasticSearch	Google Data Studio BETA
Redshift	SQL Data Warehouse	Big Query
AWS Data Pipeline	Azure Bot Service	

#### **AWS EMR**

Amazon EMR (Elastic MapReduce) is a cloud-based big data processing service.

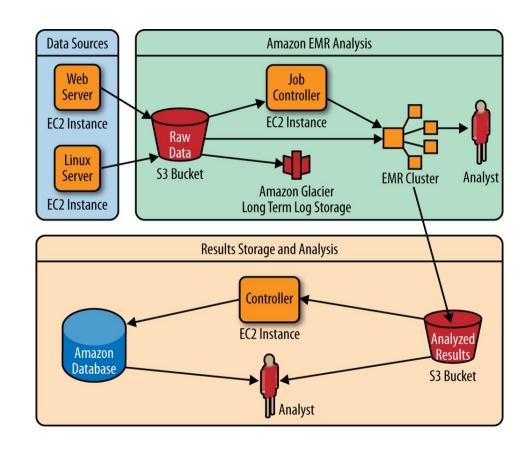
- It helps run **Apache Hadoop, Spark, Presto, and other big data frameworks** efficiently.
- Uses **EC2 instances as nodes** to process large-scale datasets in a distributed manner.
- Stores and retrieves data from Amazon S3 instead of HDFS for better scalability.
- Supports Auto Scaling, reducing costs by adjusting resources dynamically.
- Ideal for data analytics, machine learning, and log processing.

#### **AWS EMR Architecture**



# Deployment of MapReduce

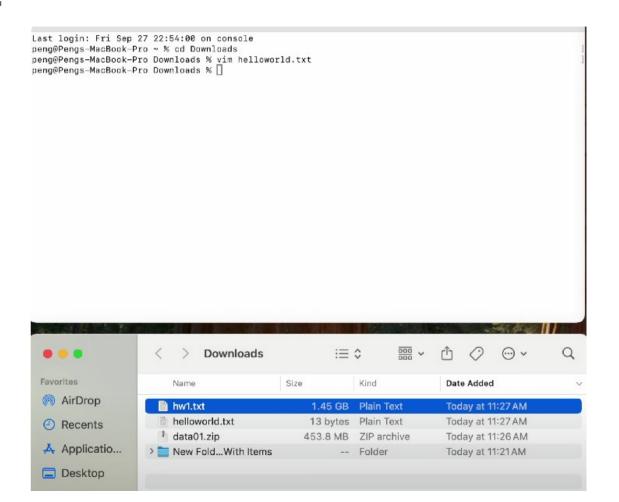
- Amazon EC2 (Elastic Compute Cloud)
- Amazon S3 (Simple Storage Service)
- AWS EMR (Elastic MapReduce)
- Amazon Glacier (Long-Term Storage)
- AWS IAM (Identity & Access Management)
- Amazon Database (RDS/DynamoDB)

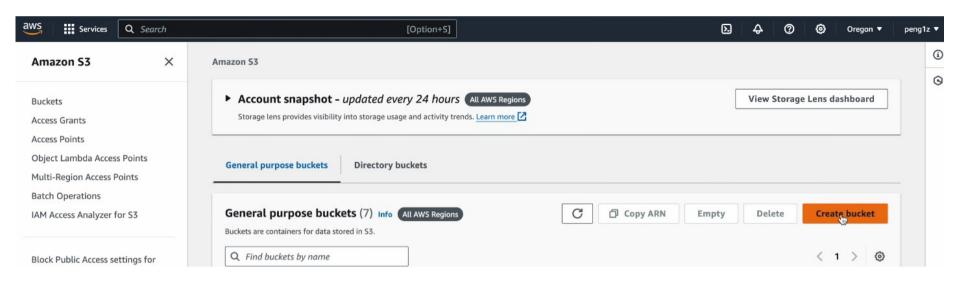


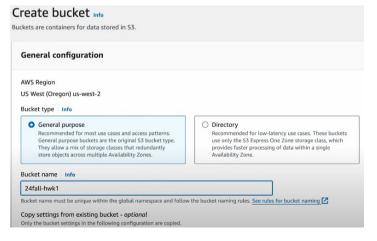
### Deploying MapReduce on AWS EMR

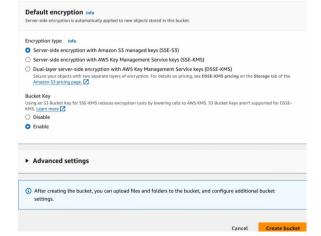
**Mohamed Muzammil J (22Z239)** 

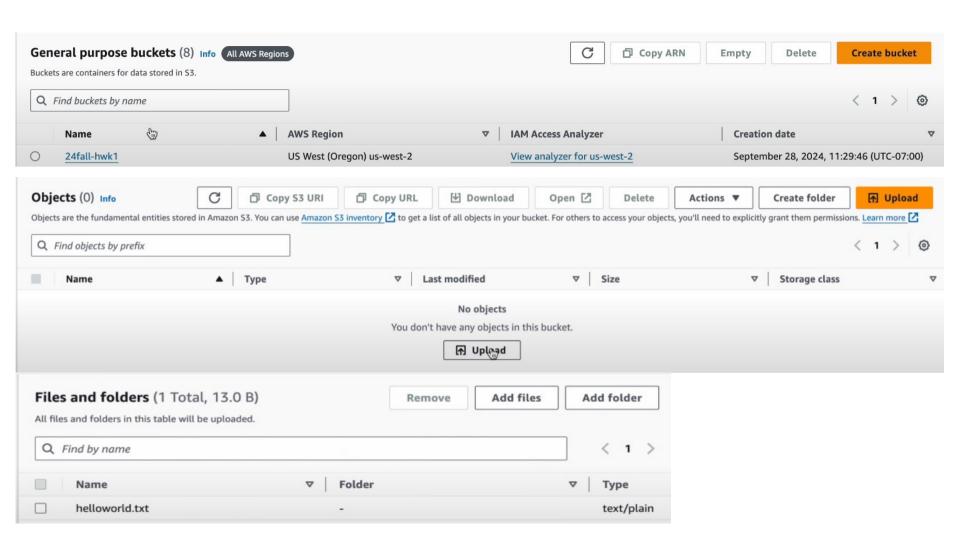
#### **EXAMPLE**

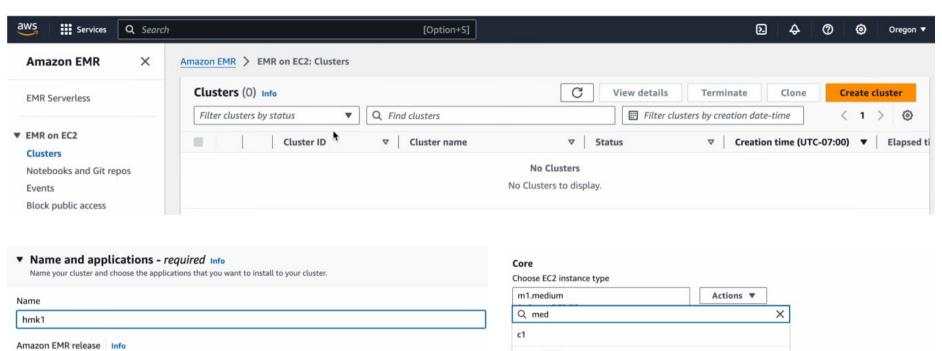


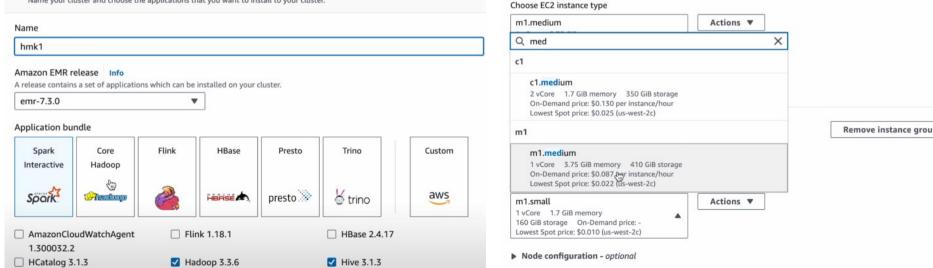


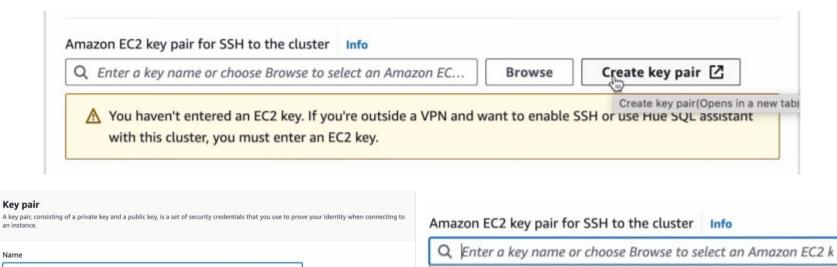






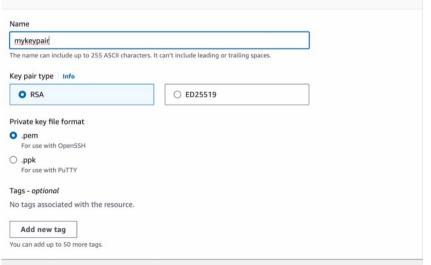




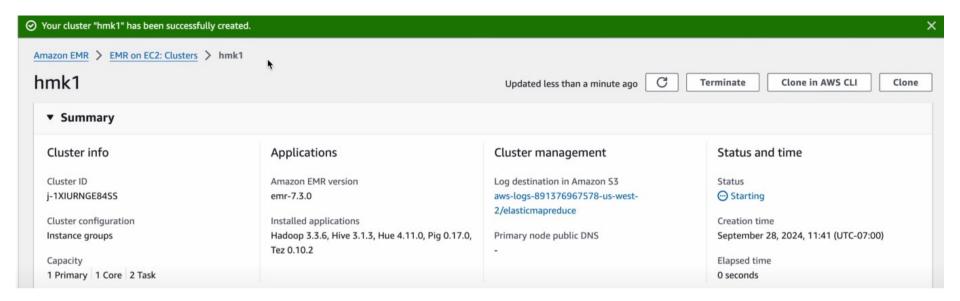


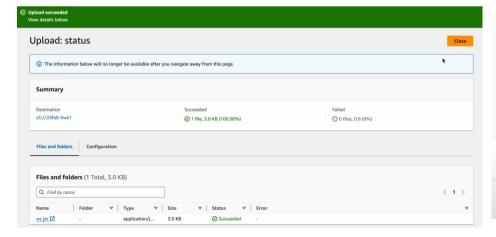
Create key pair

Cancel

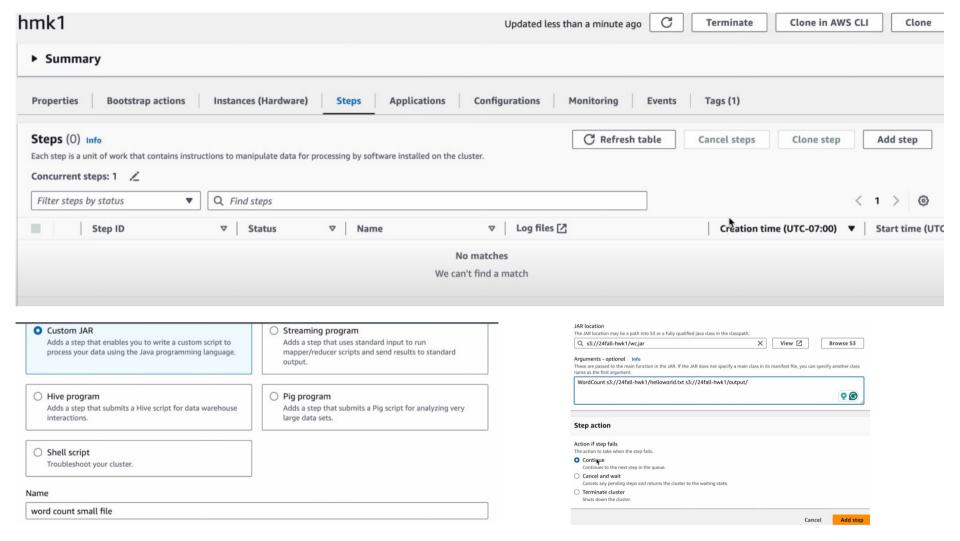


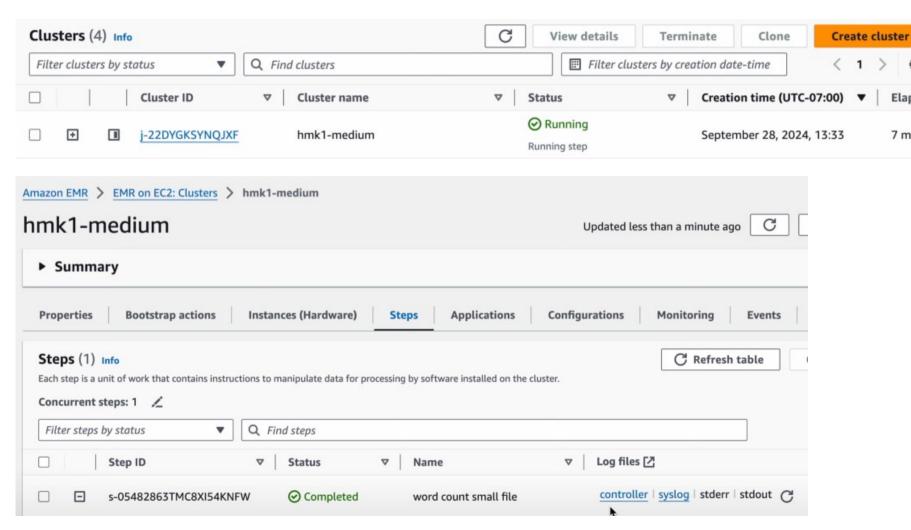






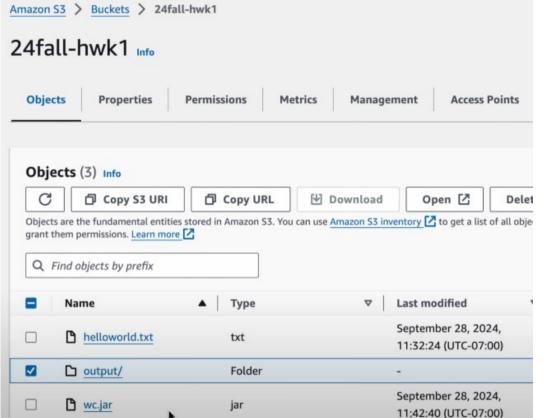






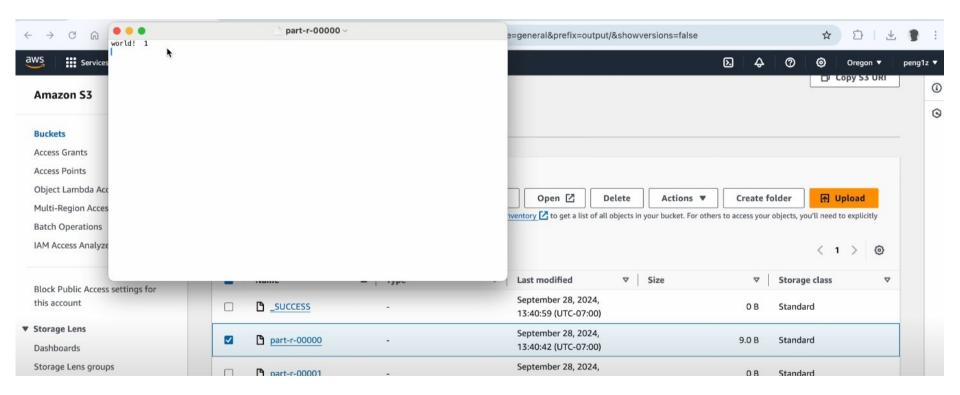
Elapsed ti

7 minutes



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	<b>p</b> art-r-00002	-	

#### RESULT



## Common Challenges in MapReduce & Cloud Deployment

#### **Challenges in MapReduce:**

- Scalability Issues Handling very large datasets efficiently
- I/O Bottlenecks Disk-based processing slows down execution
- Fault Tolerance Overhead Checkpointing and recovery take time
- Complex Debugging Difficult to troubleshoot errors

#### **Challenges in Cloud Deployment:**

- Security & Data Privacy Sensitive data risks in cloud storage
- Resource Management Cost optimization and underutilization
- Network Latency Delays due to distributed infrastructure
- Vendor Lock-in Dependence on a single cloud provider

# Optimization Techniques for Better Performance

#### **MapReduce Optimization:**

- Combiner Functions Reduce the amount of data transferred between nodes
- Data Locality Process data closer to where it is stored
- Efficient Partitioning & Scheduling Balance workload distribution
- Compression & In-Memory Processing Reduce storage & improve speed

#### **Cloud Optimization:**

- Auto-scaling Dynamically allocate resources based on demand
- Containerization (Docker, Kubernetes) Efficient resource utilization
- Edge Computing Reduce latency by processing data closer to the source
- Caching & Load Balancing Improve response times and availability

## **Future of Big Data Processing**

Spark vs. MapReduce

Feature	MapReduce	Spark
Speed	Slower (Disk-based)	Faster (In-memory)
Ease of Use	Complex (Java, MR jobs)	Simple (Scala, Python)
Performance	Batch Processing	Real-time & Batch