20CYS402 - Distributed Systems & Cloud Computing

Lab 5 - Hadoop Cluster and MapReduce

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Task 1: Hadoop Service Startup and HDFS Interaction

Objective: The primary goal of this task is to start the Hadoop cluster, verify that all its
essential background services (daemons) are running correctly, and then perform
fundamental file system operations—such as creating a directory and uploading a file—on
the Hadoop Distributed File System (HDFS).

Service Startup: Start all essential Hadoop services: NameNode, DataNode, ResourceManager, NodeManager, and SecondaryNameNode. Provide the command line instructions used to start each service.

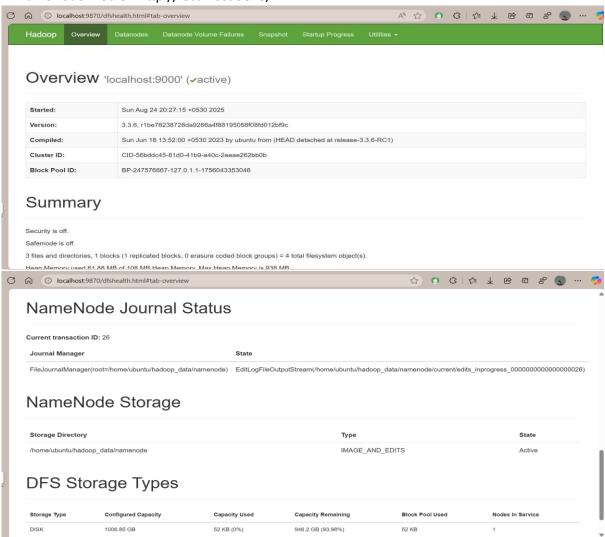
Service Verification: Use the jps command to list and verify that all core Hadoop services are running. Capture a screenshot of the terminal output showing the running services and describe the function of each service:

- NameNode: Manages the HDFS filesystem namespace.
- DataNode: Stores the actual data in HDFS.
- ResourceManager: Manages resource allocation for MapReduce jobs.
- NodeManager: Manages the execution of containers on worker nodes.
- SecondaryNameNode: Periodically checkpoints the HDFS metadata.
 - start-all.sh This script initiates all the necessary Hadoop daemons for both HDFS and YARN.
 - 2. *jps* This command lists all running Java processes, which allows you to verify that the Hadoop daemons have started.

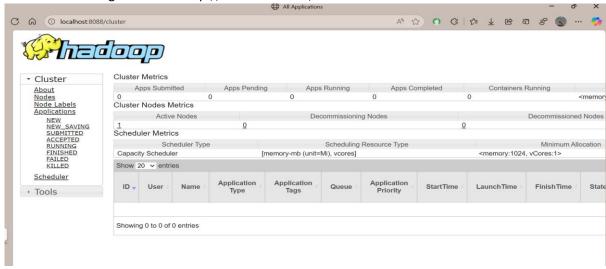
```
ini:~$ start-all.sh
WARNING: Attempting to start all Apache Hadoop daemons as ubuntu in 10 seconds.
WARNING: This is not a recommended production deployment configuration.
WARNING: Use CTRL-C to abort.
Starting namenodes on [localhost]
Starting datanodes
Starting secondary namenodes [ChitraHarini]
Starting resourcemanager
Starting nodemanagers
ubuntu@ChitraHarini:~$ jps
19090 NameNode
20213 Jps
19224 DataNode
19465 SecondaryNameNode
19675 ResourceManager
19806 NodeManager
```

Hadoop Service Testing via Browser: Use the following URLs to verify the status of your Hadoop services through a web browser:

1. NameNode Web UI: http://localhost:9870/

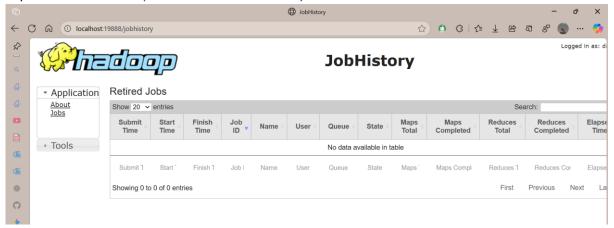


2. ResourceManager Web UI: http://localhost:8088/

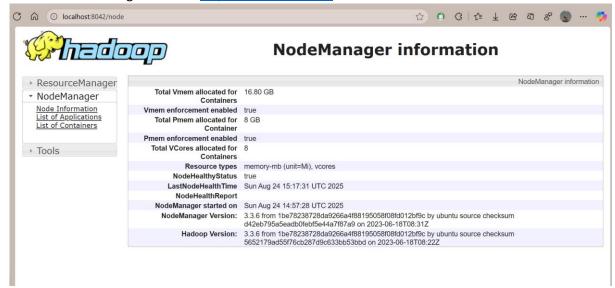


3. JobHistory Server Web UI: http://localhost:19888/

This server tracks the status of completed MapReduce jobs. It often needs to be started with a separate command *mapred* --daemon start historyserver



4. YARN NodeManager Web UI: http://localhost:8042/



HDFS Interaction:

- 3. hdfs dfs -mkdir /input Creates a new directory named input within HDFS.
- 4. *hdfs dfs -put textdata.txt /input -* Copies the local textdata.txt file into the /input directory on HDFS.
- 5. *hdfs dfs -ls /input* Lists the contents of the /input directory to confirm the file was uploaded.

```
ubuntu@ChitraHarini:~$ hdfs dfs -mkdir /input
ubuntu@ChitraHarini:~$ hdfs dfs -put ~/mapreduce/textdata.txt /input
ubuntu@ChitraHarini:~$ hdfs dfs -ls /input
Found 1 items
-rw-r--r-- 1 ubuntu supergroup 4115 2025-08-24 13:58 /input/textdata.txt
```

Explanation: This task is the foundation for all Hadoop operations. The start-all.sh script
brings the entire cluster online. HDFS provides the distributed storage layer, managed by the
NameNode (which organizes the file system) and DataNodes (which store the data). YARN
provides the processing framework, managed by the ResourceManager (the master that
allocates resources) and NodeManagers (workers on each machine). The hdfs dfs commands

- are your tools for interacting with HDFS, much like using mkdir and cp in a standard Linux environment.
- **Conclusion**: Successfully completing this task results in a fully operational single-node Hadoop cluster. It confirms that the storage and processing layers are correctly configured and ready to accept data and execute jobs. The textdata.txt file is now stored within HDFS, prepared for the MapReduce tasks.

Task 2: Implement Word Count Using MapReduce

Write a MapReduce program in Java to count the occurrences of each word in a given document. The program should map each word to a key-value pair and reduce the data by summing the counts. Run the job on your Hadoop cluster and submit the Java code, a sample input file, the output of the job, and a brief description of the steps taken.

Objective

The objective of this task is to execute a fundamental MapReduce program called **WordCount**. This program reads a text file (textdata.txt) from HDFS, processes it in a distributed manner to count the occurrences of each unique word, and writes the final counts back into a new file on HDFS. This task validates that your entire Hadoop cluster—both HDFS for storage and YARN for processing—is working correctly.

Commands

1. Run the WordCount Job

Command: hadoop jar wordcount.jar com.jordiburgos.WordCount /input/textdata.txt /output/wordcount

Explanation: This command tells the YARN ResourceManager to start a new application.

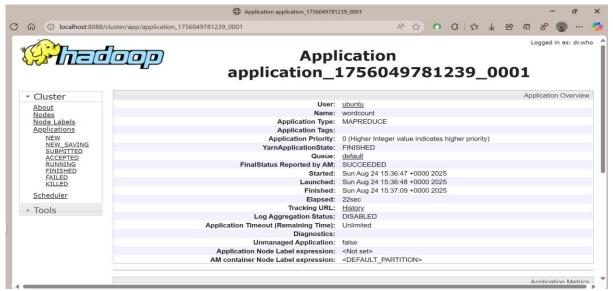
- hadoop jar wordcount.jar: Specifies the Java program to run.
- WordCount: The main class inside the JAR that contains the job configuration.
- /input/textdata.txt: The input file on HDFS that the job will read.
- /output/wordcount: The new directory on HDFS where the results will be stored. This directory must not exist before you run the command.

2. View the Output

Command: hdfs dfs -cat /output/wordcount/part-r-00000

 Explanation: After the job completes, this command prints the contents of the output file. The results are typically in a file named part-r-00000 (the 'r' stands for Reducer).





Code Explanation

The Java code for WordCount, as seen in your mapreduce_Help.docx, has three main parts:

- 1. The Driver (main method)
 - Purpose: This is the orchestrator. It sets up and configures the MapReduce job.
 - O What it does:
 - It takes the input path (/input/textdata.txt) and output path (/output/wordcount) as command-line arguments.
 - It tells Hadoop which class to use for the Mapper (MapForWordCount.class) and which to use for the Reducer (ReduceForWordCount.class).
 - It defines the data types for the final output key (Text for the word) and value (IntWritable for the count).
 - Finally, it submits the job to the YARN cluster and waits for it to complete.

2. The Mapper (MapForWordCount class)

- Purpose: To process the raw input data and transform it into standardized key-value pairs.
- What it does:
 - The map function receives one line of the textdata.txt file at a time.

- It converts the line to a string and splits it into an array of words using spaces as the delimiter.
- It then loops through each word. For every single word, it emits a key-value pair of (word, 1). For example, the line "the dog saw the cat" results in five outputs: (THE, 1), (DOG, 1), (SAW, 1), (THE, 1), (CAT, 1).

3. The Reducer (ReduceForWordCount class)

- Purpose: To aggregate the intermediate results from the Mapper.
- O What it does:
 - After the map phase, Hadoop's "shuffle and sort" process groups all the emitted pairs by their key. The Reducer receives one key at a time, along with a list of all its associated values.
 - For example, it will receive (THE, [1, 1]).
 - The reduce function then simply iterates through the list of values (the ones) and sums them up.
 - Finally, it writes the word and its final calculated sum to the output file in HDFS.

Conclusion

Successfully completing this task demonstrates a complete, end-to-end execution of a distributed data processing job. It proves that your Hadoop cluster can take input data from HDFS, process it in parallel across the cluster using the MapReduce framework managed by YARN, and write the final, aggregated results back to HDFS. It is the "Hello, World!" of Big Data and confirms your setup is ready for more complex analysis.

Task 3: Implement Line Count Using MapReduce

Write a MapReduce program in Java to count the number of lines in a text document. The program should map each line to a key-value pair with the line count (1 for each line) and reduce the data by summing the counts. Run the job on your Hadoop cluster and submit the Java code, a sample input file, the output of the job, and a brief description of the steps taken.

Objective

The objective of this task is to develop, compile, and execute a custom MapReduce program from scratch. This program, **LineCount**, will read the textdata.txt file from HDFS and calculate the total number of lines in the document. This demonstrates a deeper understanding of the MapReduce paradigm by applying it to a new problem.

Java Code (LineCount.java)

import java.io.IOException;

import org.apache.hadoop.conf.Configuration;

import org.apache.hadoop.fs.Path;

import org.apache.hadoop.io.IntWritable;

import org.apache.hadoop.io.LongWritable;

import org.apache.hadoop.io.Text;

import org.apache.hadoop.mapreduce.Job;

import org.apache.hadoop.mapreduce.Mapper;

 $import\ org. a pache. hadoop. mapreduce. Reducer;$

import org.apache.hadoop.mapreduce.lib.input.FileInputFormat;

 $import\ org. a pache. hadoop. mapreduce. lib. output. File Output Format;$

```
public class LineCount {
 // The Mapper Class
 public static class LineCountMapper extends Mapper<LongWritable, Text, Text, IntWritable> {
  private final static IntWritable one = new IntWritable(1);
  private Text constantKey = new Text("Total Lines");
  public void map(LongWritable key, Text value, Context context) throws IOException, InterruptedException {
   // For every line we read, we emit the same key ("Total Lines") and the value 1.
   // We don't care what the line's content is.
   context.write(constantKey, one);
  }
 // The Reducer Class
 public static class LineCountReducer extends Reducer<Text, IntWritable, Text, IntWritable> {
  public void reduce(Text key, Iterable<IntWritable> values, Context context) throws IOException, InterruptedException {
   // The reducer receives the key "Total Lines" and a list of 1s ([1, 1, 1, ...]).
   // We just need to sum them up to get the total count.
   for (IntWritable val : values) {
    sum += val.get();
   context.write(key, new IntWritable(sum));
 // The Driver
 public static void main(String[] args) throws Exception {
  Configuration conf = new Configuration();
  Job job = Job.getInstance(conf, "line count");
  job.setJarByClass(LineCount.class);
  job.setMapperClass(LineCountMapper.class);
  // Combiner is the same as the Reducer in this case, which is a common optimization.
  job.setCombinerClass(LineCountReducer.class);
  job.setReducerClass(LineCountReducer.class);
  job.setOutputKeyClass(Text.class);
  job.setOutputValueClass(IntWritable.class);
  FileInputFormat.addInputPath(job, new Path(args[0]));
  FileOutputFormat.setOutputPath(job, new Path(args[1]));
  System.exit(job.waitForCompletion(true)?0:1);
}
```

Commands:

Follow these steps in your terminal to create, compile, package, and run your program.

1. Create the Java File: Use nano to create the file and paste the code from above into it.

```
nano LineCount.java
```

2. **Compile the Code**: This command uses javac to compile your program. The -cp flag tells the compiler where to find all the necessary Hadoop library files.

```
javac -cp $(hadoop classpath) LineCount.java
```

3. **Create the JAR File**: This command packages your compiled .class files into a single linecount.jar file that you can submit to Hadoop.

```
jar -cvf linecount.jar *.class
```

4. **Run the MapReduce Job**: Execute your custom JAR on the cluster. We'll use the same input file and create a new output directory.

hadoop jar linecount.jar LineCount /input/textdata.txt /output/linecount

5. View the Output: Check the result file to see the final line count.

hdfs dfs -cat /output/linecount/part-r-00000

```
Whomtu@Chiratalarin:-/mapreduce$ nano LineCount.java
ubuntu@Chiratalarin:-/mapreduce$ nano LineCount.java
ubuntu@Chiratalarin:-/mapreduce$ javac -cp $(hadosp classpath) LineCount.java
ubuntu@Chiratalarin:-/mapreduce$ hadosp jar linecount.java LineCount.java
ubuntu@Chiratalarin:-/mapreduce$ hadosp jar linecount.java
ubuntu@Chiratalarin:-/mapreduce$ hadosp jar
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```

Code Explanation

- The Mapper (LineCountMapper): This is the simplest possible Mapper. For every single line it reads from the input file, it ignores the content of the line entirely and just outputs a static key-value pair: ("Total Lines", 1).
- The Reducer (LineCountReducer): After the map phase, the shuffle and sort process gathers all the emitted pairs. Since every key was identical, the Reducer receives just one group: the key "Total Lines" and a very long list of 1s, where the length of the list is equal to the number of lines in the original file. The Reducer's job is to simply iterate through this list and sum all the 1s to get the final total.
- The Driver (main method): Just like in WordCount, the driver configures the job. It specifies the Mapper and Reducer classes, sets the input and output paths from the command line, and defines the data types for the final output.

Conclusion By successfully writing, compiling, and running the LineCount program, you demonstrate a solid grasp of the MapReduce programming model. You have proven that you can analyze a problem (counting lines) and break it down into a parallelizable solution using the map and reduce paradigm The final output, a single file containing the total line count, is the result of your own custom-built, distributed data processing application.					
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