







307401 Big Data and Data Warehouses

Introduction to Hadoop – Practical Examples









Hands On Practice using Horton Works Virtual Machine

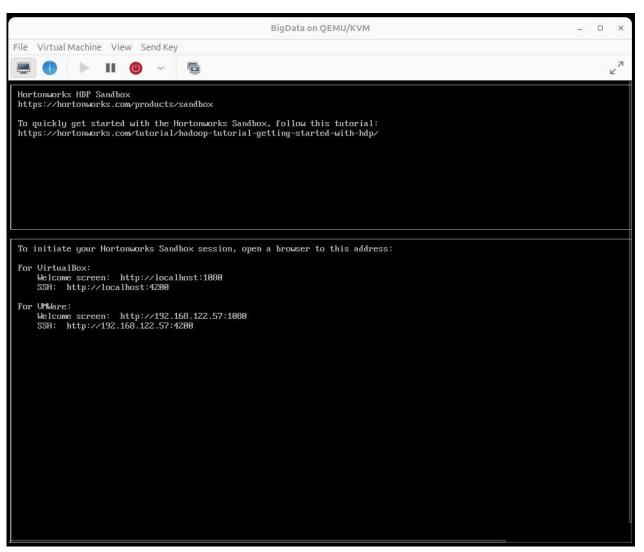






Prepare Test Environment

- Download Virtual Box: https://www.virtualbox.org/wiki/Downloads
- 2. Or Works virtual machine Version 2.6.5:
 https://archive.cloudera.com/hwx-sandbox/hdp/hdp-2.6.5/HDP-2.6.5 virtualbox 180626.ova
- 3. Import HortonWorks (.ova file) you downloaded in step 1 into Virtual Box.
- 4. Open your browser, goto: http://localhost:1080, launch HORTONWORKS main page.









Main Sand Box Web Page – localhost:1080 (port 1080)





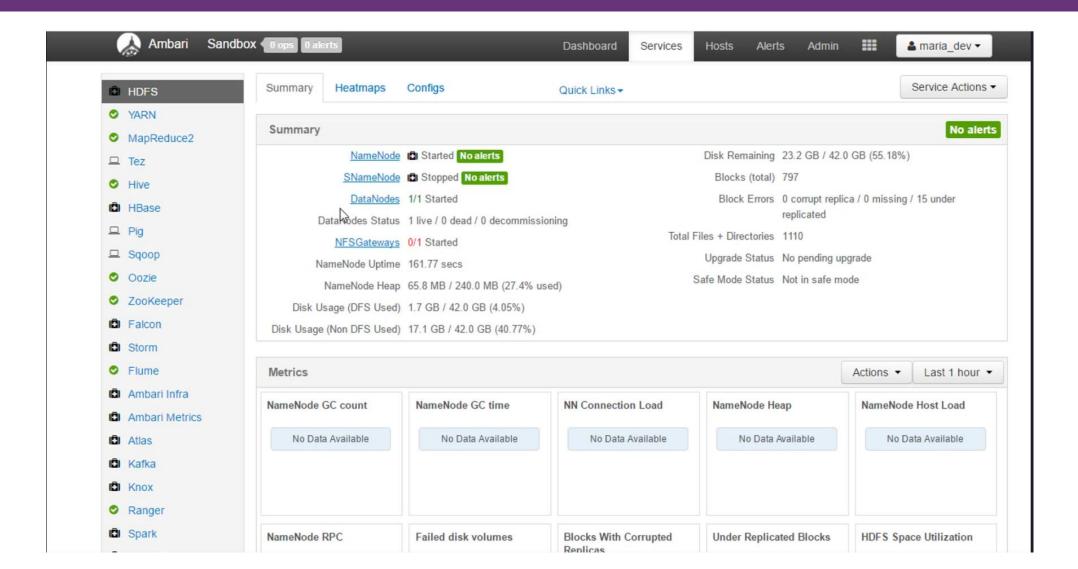








Ambari Dashboard

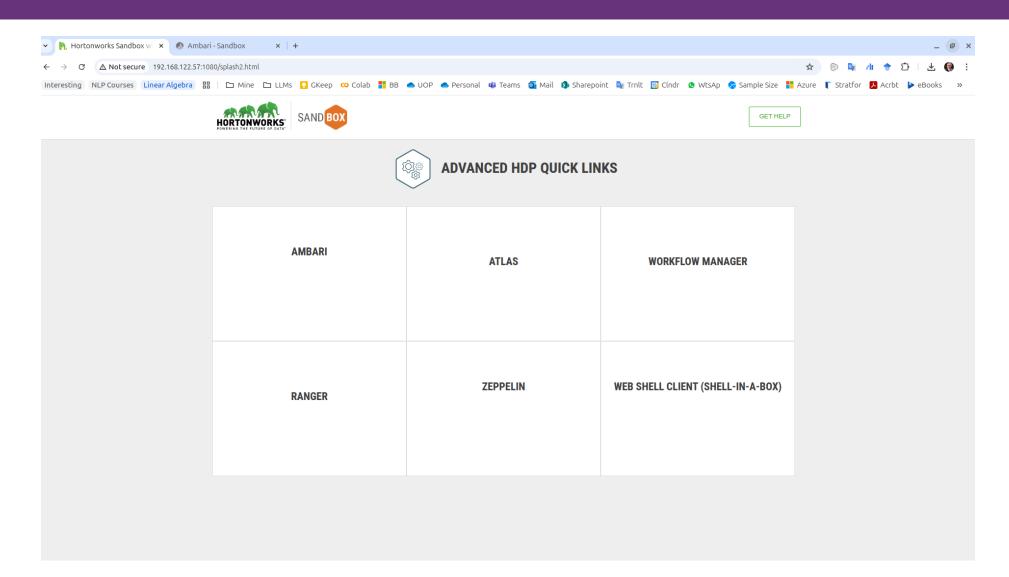








Advanced HDP Page

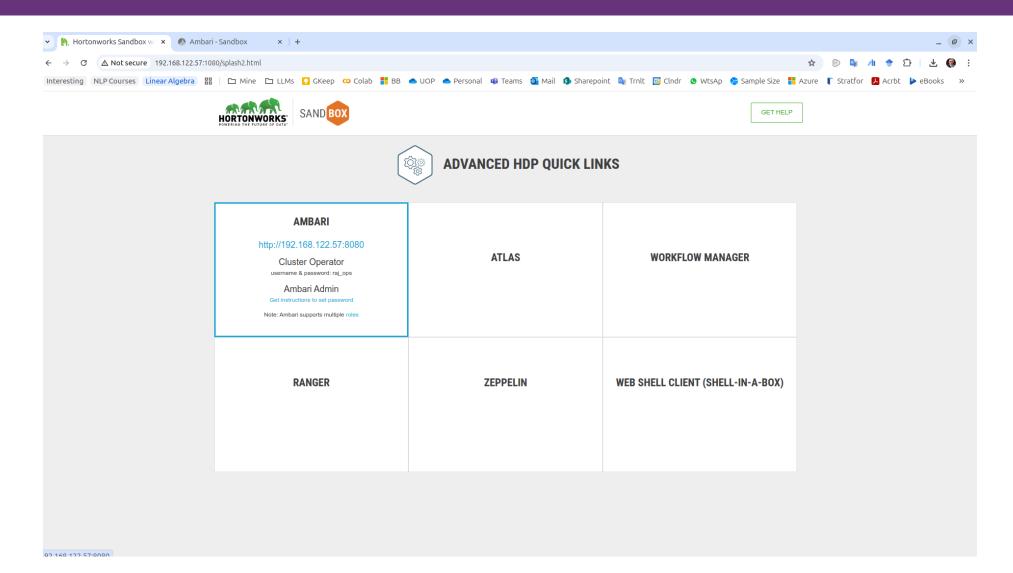








Open Ambari Dashboard









Practical (Proping)

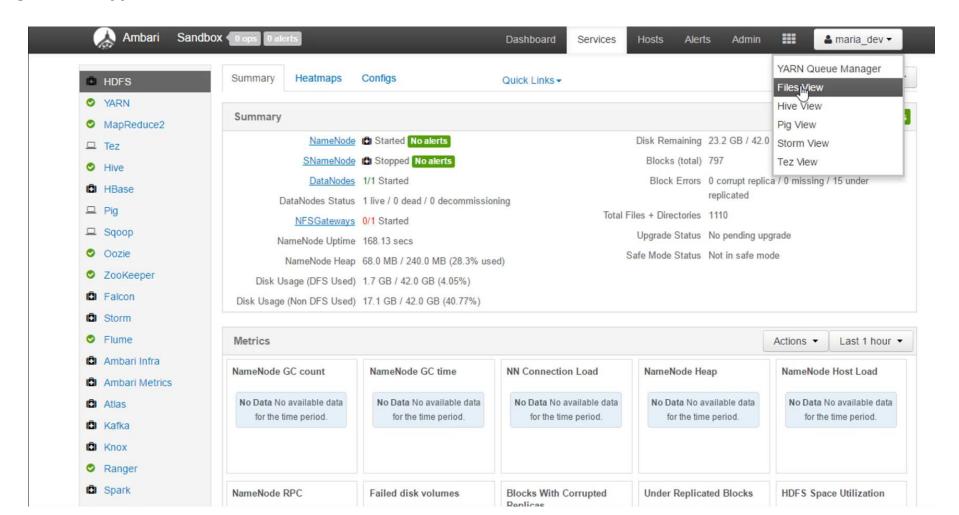






Testing HDFS

Select File View in Ambari

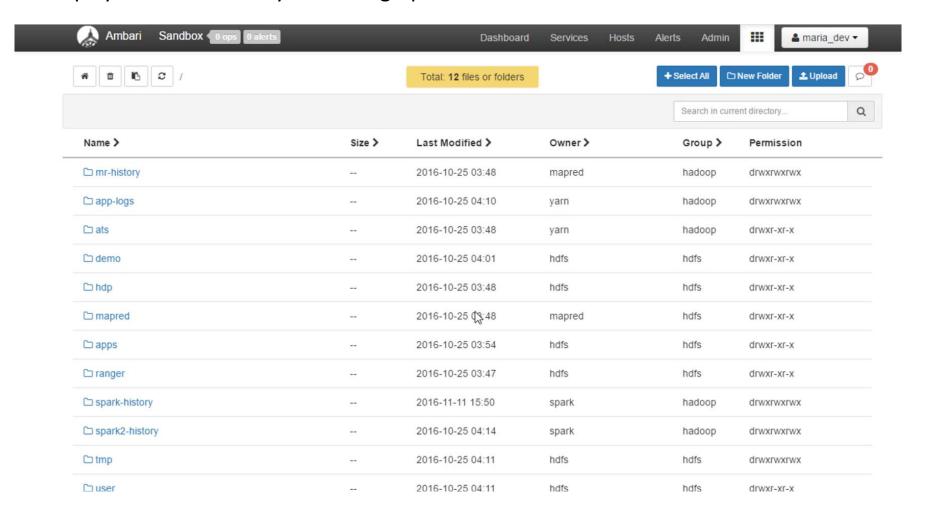








Selecting file views displays the entire file system in a graphical manner

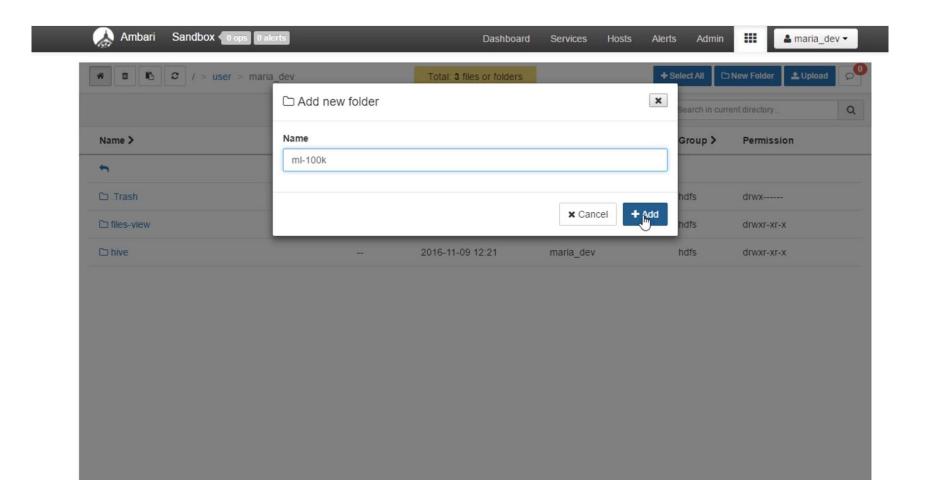








We can create a new folder to store our files.

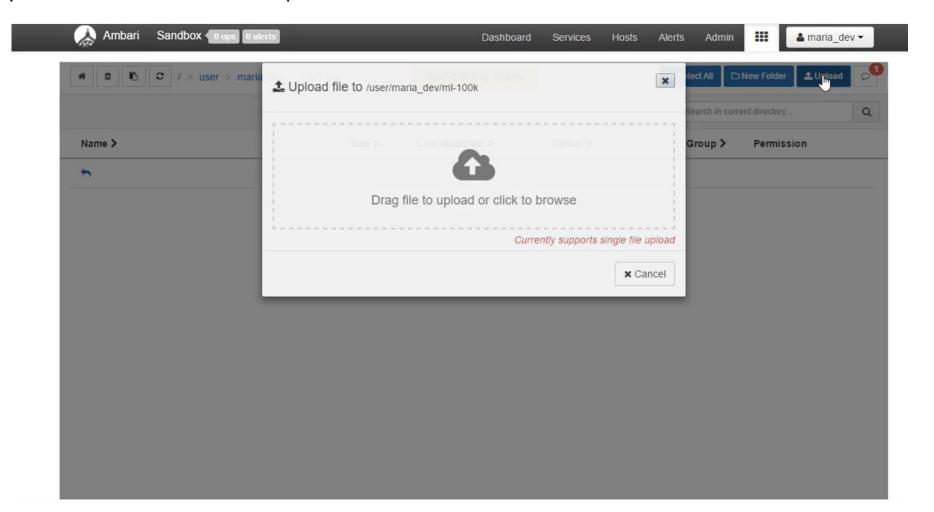








Now we can upload our data to the Hadoop cluster

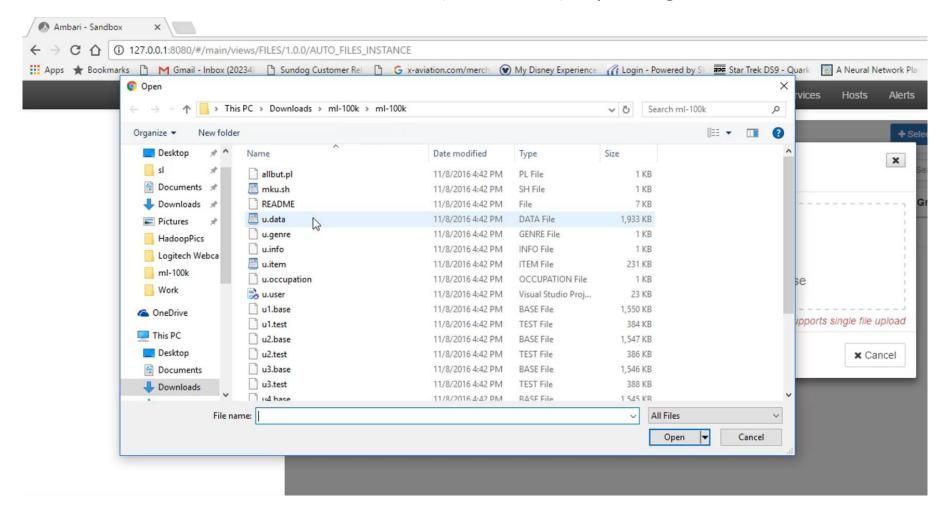








One we upload a file, HDFS distribute it across our cluster (Data Nodes) depending on file size and cluster size

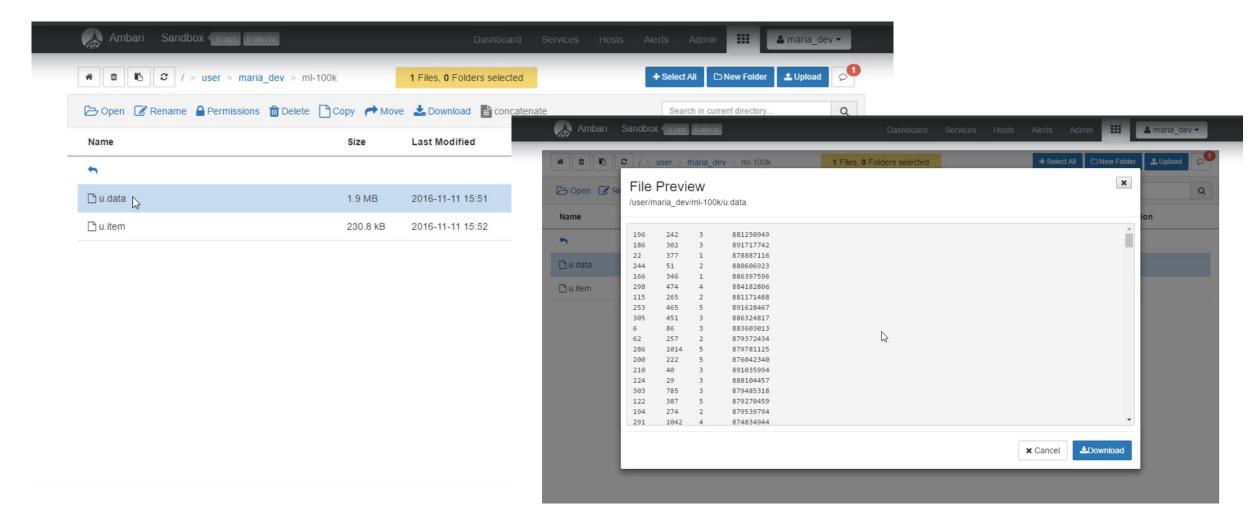








We can do file operations using Ambari interface (Open, Rename, Change Permissions, Delete, Copy, Download...etc.)

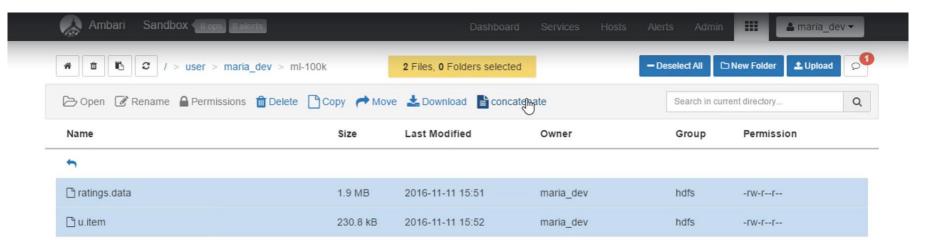








We can even concatenate multiple files to download them together



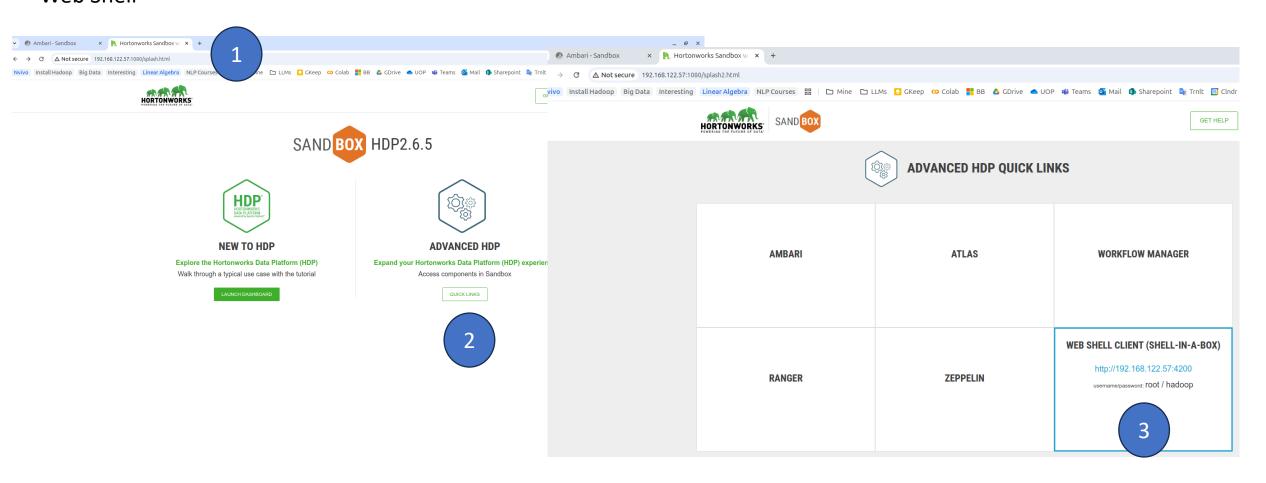






HDFS Command Line Access

Open HortonWorks Sandbox (usually http://localhost:1080, Open Advanced HDP Quick Links Page, Open Web Shell





Basic HDFS Commands

1. Check HDFS directory structure

2. Create a directory in HDFS

bash

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hdfs dfs -mkdir /user/pig/logs

3. Upload a local file to HDFS

bash

hdfs dfs -put access_log.txt /user/pig/logs/

4. List contents of a directory

5. View contents of a file

bash

hdfs dfs -cat /user/pig/logs/access_log.txt





6. Copy a file from HDFS to your local machine

7. Remove a file or directory

bash

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hdfs dfs -rm /user/pig/logs/access_log.txt

hdfs dfs -rm -r /user/pig/logs

Check Disk Usage

hdfs dfs -du -h /user/







Practical Practical







Movie Rating Example

0 50 5 881250949

0 172 5 881250949

0 133 1 881250949

196 242 3 881250949

186 302 3 891717742

22 377 1 878887116

244 51 2 880606923

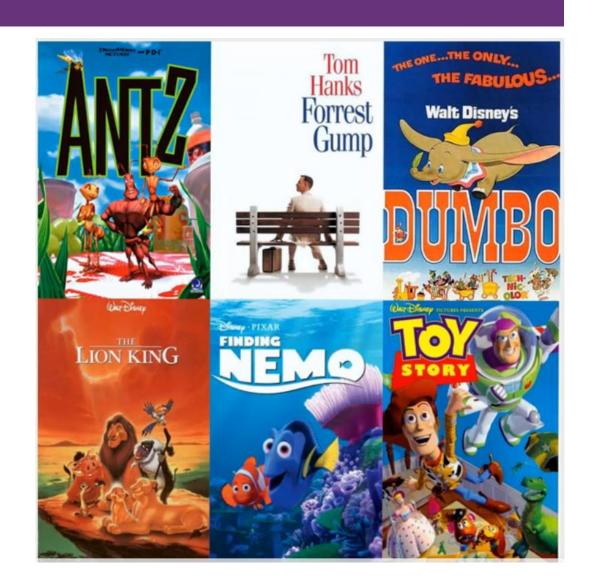
166 346 1 886397596

298 474 4 884182806

115 265 2 881171488

253 465 5 891628467

305 451 3 886324817

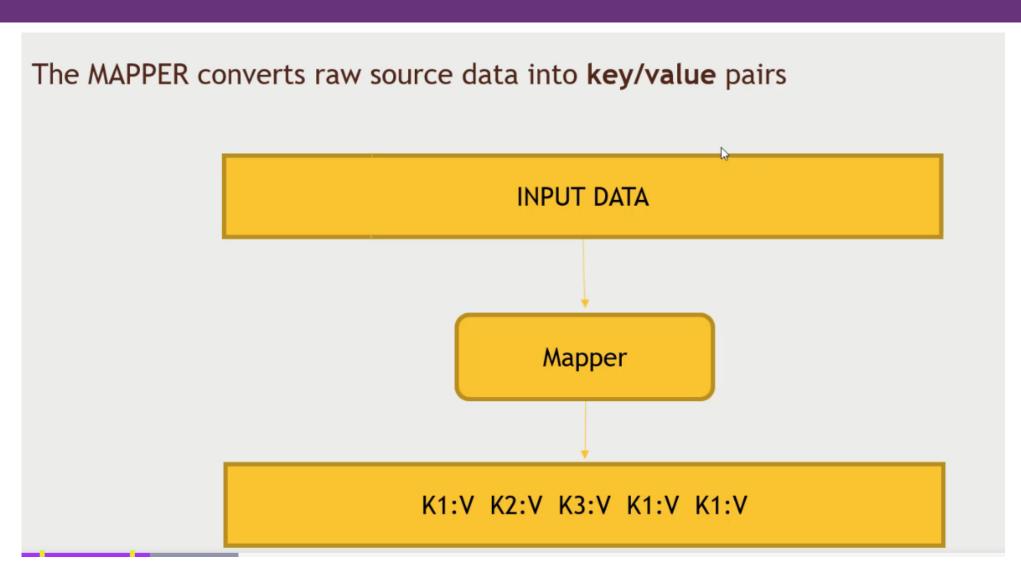








Mapping Phase









Map Reduce Movie Count Example

Writing the Mapper

```
USER ID | MOVIE ID | RATING | TIMESTAMP
      196 242 3
                   881250949
                                                  Shuffle
      186 302 3
                   891717742
                                                          1 -> 1, 1
                                                  & Sort
      196 377 1
                   878887116
                                          1,1
                                                          2 -> 1, 1
      244 51 2
                   880606923
                                          2,1
                                                          3 -> 1, 1
      166 346 1
                   886397596
                                          1,1
                                                          4 -> 1
      186 474 4
                   884182806
      186 265 2
                                          2,1
                  881171488
```

```
from mrjob.job import MRJob
from mrjob.step import MRStep
class RatingsBreakdown(MRJob):
   def steps(self):
       return [
           MRStep(mapper=self.mapper get ratings,
                   reducer=self.reducer count ratings)
   def mapper_get_ratings(self, _, line):
        (userID, movieID, rating, timestamp) = line.split('\t')
       yield rating, 1
   def reducer_count_ratings(self, key, values):
       yield key, sum(values)
if name == ' main ':
   RatingsBreakdown.run()
```







Workflow Scenario from the Client to NameNode/DataNode

1. Client Submission:

• The client submits a MapReduce job to the Hadoop cluster. The job includes the Mapper and Reducer code, and configuration parameters. The data is already stored in the Hadoop Distributed File System (HDFS)

2. ResourceManager and Job Scheduling:

- The ResourceManager receives the job submission and delegates the job to the ApplicationsManager.
- The ApplicationsManager allocates a new ApplicationMaster (AppMaster) instance for the job.

3. ApplicationMaster Initialization:

• The ApplicationMaster negotiates resources from the ResourceManager and requests containers (execution environments) on various NodeManagers (Data Nodes).

4. Data Splitting:

- The ApplicationMaster communicates with the NameNode to locate the input data blocks stored across the DataNodes.
- The input data is split into chunks, and the splits are assigned to the containers on the DataNodes where the data is located.

5. Mapper Execution:

- The NodeManagers launch Mapper tasks in the allocated containers.
- Each Mapper processes its assigned data chunk, producing intermediate key-value pairs.

6. Shuffling and Sorting:

The intermediate key-value pairs are shuffled and sorted by key. This involves transferring data across the network to group all values for the same key together.

7. Reducer Execution:

The NodeManagers launch Reducer tasks in containers.

Each Reducer processes its assigned key group, aggregating the values to produce the final output.

8. Writing the Output:

The Reducers write the final output to HDFS, which is managed by the NameNode.

The ApplicationMaster monitors the job's progress and updates the ResourceManager.

9. Job Completion:

Once all Mapper and Reducer tasks are complete, the ApplicationMaster informs the ResourceManager.

The client is notified of the job's completion, and the final results are available in HDFS















Hive Practical Example

Access Hive Console in Ambari

- In Ambari (left sidebar), click Hive.
- Look for a link under Quick Links called Hive View 2.0.
- Click it this will open a web-based
 Hive editor (SQL interface).

Key Insights

- Hive works well for **batch queries** on huge datasets.
- SQL-like syntax (HiveQL) makes it easy for analysts to use.
- Not suited for real-time access that's where HBase fits.

```
Step-by-Step Example in Hive
1. Create an External Table
  sql

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  CREATE EXTERNAL TABLE sales (
    transaction_id STRING,
    customer_id STRING,
    product_id STRING,
    amount DOUBLE,
    transaction date STRING
  ROW FORMAT DELIMITED
  FIELDS TERMINATED BY ','
  STORED AS TEXTFILE
  LOCATION '/user/hive/sales_data/';
2. Sample Queries
a. Total sales per product
  sql

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                                                                                   * Edit
  SELECT product_id, SUM(amount) AS total_sales
  FROM sales
  GROUP BY product_id;
b. Top 5 customers by total spend
  SELECT customer_id, SUM(amount) AS total_spent
  FROM sales
  GROUP BY customer id
  ORDER BY total_spent DESC
  LIMIT 5;
```







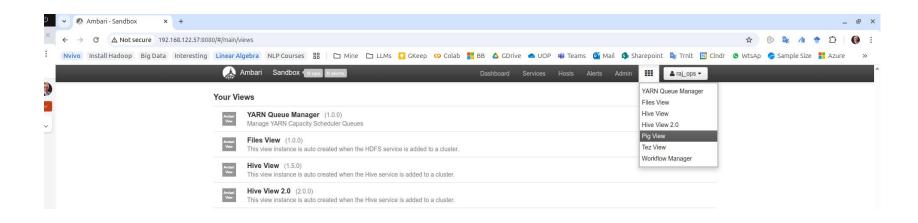
Practical Apache Pig

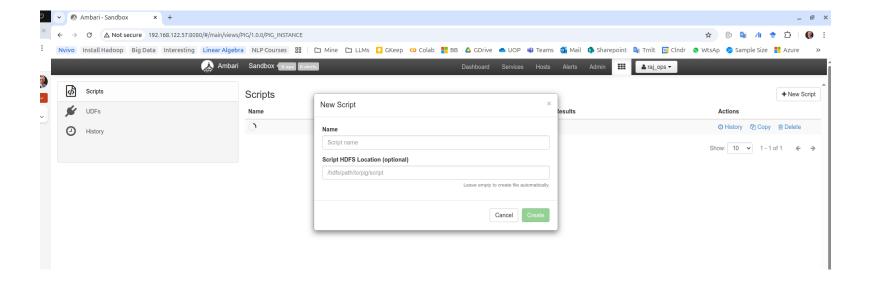






Open Pig View and Create a New Script











Scenario: Analyze Website Logs with Apache Pig

Objective:

Extract insights from a web server log file (e.g., count hits per IP address, find most visited pages).

Key Highlights

- Demonstrates Pig's ability to process **semi-structured text**.
- Shows how to write a data pipeline using LOAD, GROUP, and FOREACH.
- Ideal for log analysis, ETL pipelines, or preprocessing before Hive or HBase.

Step 1: Sample Log Data (upload to HDFS)

Save the following as access_log.txt and upload to HDFS (e.g., /user/pig/logs/):

```
192.168.1.1 /index.html
192.168.1.2 /contact.html
192.168.1.1 /index.html
192.168.1.3 /about.html
192.168.1.2 /index.html
```

(/about.html,1)
(/contact.html,1)
(/index.html,3)

Step 2: Pig Script to Count Hits per Page

```
logs = LOAD '/user/pig/logs/access_log.txt' USING PigStorage('
') AS (ip:chararray, page:chararray);
grouped_pages = GROUP logs BY page;
page_counts = FOREACH grouped_pages GENERATE group AS page,
COUNT(logs) AS hits;
DUMP page_counts;
Expected Output
```





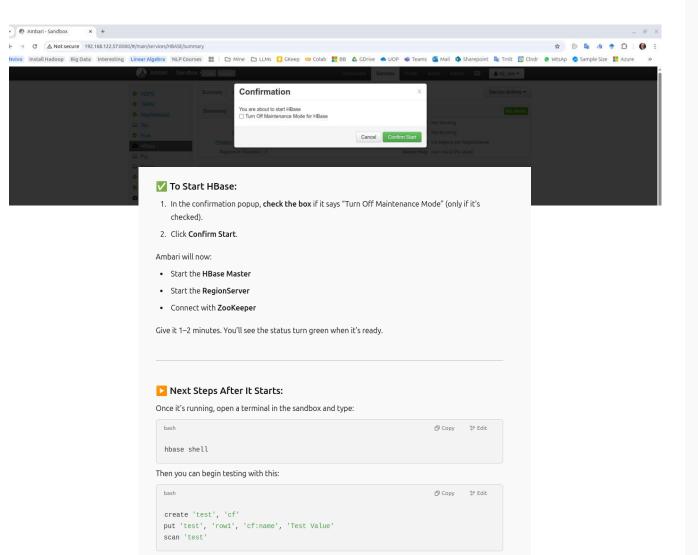


Practical





Practical HBase



✓ Step 1: Start HBase from Ambari

- Go to the Ambari Dashboard.
- 2. Click on **HBase** in the left panel.
- 3. Click the Service Actions dropdown on the top-right.
- Select Start → This will start the HBase Master and RegionServer(s).
- 5. Wait for all green checkmarks and no alerts.

✓ Step 2: Access the HBase Shell

Once HBase is running:

- 1. Open your Sandbox terminal (you can SSH into it or use the web terminal).
- 2. Run:

bash

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hbase shell

You'll see the hbase(main):001:0> prompt, where you can start testing commands.

✓ Step 3: Run Sample Commands

Try these basic ones to verify:

```
create 'test', 'cf'
put 'test', 'row1', 'cf:name', 'Hello HBase'
get 'test', 'row1'
scan 'test'
```







Practical & kafka







Practical Scenario: Real-Time Sensor Data Streaming with Apache Kafka

• **Scenario:** A company is monitoring temperature sensors in a factory. Each sensor sends data every few seconds.

• The goal is to:

- Stream the sensor readings in real time.
- Store them in HDFS for long-term analysis.
- Optionally process them live using Spark or Storm.

Components Involved:

- Kafka → for streaming data
- \circ **HDFS** \rightarrow for storing data
- Optional: Flume or Spark Streaming for processing







Practical Scenario: Real-Time Sensor Data Streaming with Apache Kafka

Step-by-Step Workflow

Step 1: Start Kafka Services, ensure ZooKeeper and Kafka are running in Ambari.

Step 2: Create a Kafka Topic

• kafka-topics --create --zookeeper localhost:2181 --replication-factor 1 --partitions 1 --topic sensor_topic

Step 3: Simulate a Sensor Producer (Terminal)

- kafka-console-producer --broker-list localhost:6667 --topic sensor topic
- Type some simulated sensor data:
- sensor1,2025-03-23T09:00,22.5
- sensor2,2025-03-23T09:01,21.8
- sensor1,2025-03-23T09:02,22.7

Step 4: Start a Kafka Consumer (Another Terminal)

• kafka-console-consumer --bootstrap-server localhost:6667 --topic sensor_topic --from-beginning

Expected Output: You'll see the same messages printed in real-time — demonstrating **decoupled data streaming**.

- Optional: Save Kafka Stream to HDFS
- Use Flume or Kafka Connect HDFS sink to write streamed data into HDFS.

Key Takeaways:

- Kafka is ideal for real-time ingestion.
- Decouples data producers and consumers.
- Topics enable scalable, fault-tolerant messaging across multiple services.
- Would you like a follow-up scenario showing how Hive or Spark can consume data from Kafka?















What is Apache Oozie?

Oozie is a **workflow scheduler system** for managing Hadoop jobs (MapReduce, Hive, Pig, Sqoop, etc.) as **directed acyclic graphs (DAGs)**.

What You Can Do with Workflow Manager:

- Create and visualize **Oozie workflows** using a graphical interface.
- Schedule workflows to run at a specific time or in response to events.
- Chain multiple jobs (e.g., run Pig \rightarrow then Hive \rightarrow then Email Notification).
- Define decision points, forks, and joins in the workflow logic.

Example Scenario: A student wants to automate the following daily:

- Run a Pig job to process logs.
- Run a Hive job to summarize the results.
- Archive the data using HDFS commands.
- With Workflow Manager, they can design and monitor this process without writing XML manually.







Practical Scenario: Daily Data Pipeline Automation with Oozie

Scenario: A company wants to automate a daily pipeline that:

- Runs a Pig job to clean raw log data.
- Runs a **Hive query** to generate a summary report.
- Moves the output to an archive folder in HDFS.

Workflow Steps in Oozie

- **Start node** begins the workflow.
- Pig action cleans the data (e.g., filters malformed records).
- **Hive action** aggregates metrics (e.g., page views per user).
- FS action moves the final result to /data/archive/YYYY-MM-DD/.
- End node completes the workflow.

How to Implement in Hortonworks

- Open Workflow Manager in Ambari (as shown in your screenshot).
- Create a new workflow.
- Drag and drop actions:
 - Pig: run clean_logs.pig
 - Hive: run summary_query.hql
 - FS: define move operation
- Add a coordinator to schedule it daily at midnight.
- Deploy and run the workflow.
- Example Hive Query (Step 2)
- SELECT user id, COUNT(*) AS pageviews
- FROM cleaned logs
- GROUP BY user_id;

Key Highlights

- Automates multi-step workflows with scheduling.
- Centralized control, retry on failure, alerting.
- Supports Pig, Hive, Spark, Sqoop, Shell, Java, and more.







Oozie Workflow Manager (GUI) – Visual Design in Ambari

Steps to Create the Workflow:

- Open Ambari → Workflow Manager → Create Workflow.
- Name it: daily-log-workflow.
- Drag and drop the following components onto the canvas:
 - Start
 - Pig Action
 - Script: /user/oozie/scripts/clean_logs.pig
 - Input path: /data/logs/raw/
 - Output path: /data/logs/cleaned/
 - Hive Action
 - Script: /user/oozie/scripts/summary_query.hql
 - FS Action
 - Move /data/logs/summary/to /data/logs/archive/\${YEAR}-\${MONTH}-\${DAY}/
 - End
- Link them in sequence: Start → Pig → Hive → FS → End
- Click Save, then go to the Coordinator tab to schedule it daily at midnight.







Daily Log Workflow – XML Version

```
<workflow-app name="daily-log-workflow" xmlns="uri:oozie:workflow:0.5">
    <start to="clean-logs"/>
    <action name="clean-logs">
        <pig>
            <script>clean_logs.pig</script>
            <param>input=/data/logs/raw/</param>
            <param>output=/data/logs/cleaned/</param>
        </pig>
        <ok to="summarize"/>
        <error to="fail"/>
    </action>
    <action name="summarize">
        <hive xmlns="uri:oozie:hive-action:0.5">
            <script>summary_query.hql</script>
        </hive>
        <ok to="archive"/>
        <error to="fail"/>
    </action>
    <action name="archive">
        <fs>
            <move source="/data/logs/summary/" target="/data/logs/archive/${YEAR}-${MONTH}-${DAY}/"/>
        </fs>
        <ok to="end"/>
        <error to="fail"/>
    </action>
    <kill name="fail">
        <message>Workflow failed</message>
    </kill>
    <end name="end"/>
</workflow-app>
```







Optional: Coordinator XML (to schedule it daily)

```
<coordinator-app name="daily-log-coordinator"</pre>
   xmlns="uri:oozie:coordinator:0.4" frequency="1"
   start="2025-03-24T00:00Z" end="2025-12-31T00:00Z" timezone="UTC">
    <action>
       <workflow>
            <app-path>/user/oozie/workflows/daily-log-workflow</app-path>
            <configuration>
                cproperty>
                    <name>YEAR</name>
                    <value>${coord:formatTime(coord:actualTime(), 'yyyy')}</value>
                </property>
                cproperty>
                    <name>MONTH</name>
                    <value>${coord:formatTime(coord:actualTime(), 'MM')}</value>
                </property>
                cproperty>
                    <name>DAY</name>
                    <value>${coord:formatTime(coord:actualTime(), 'dd')}</value>
                </property>
            </configuration>
        </workflow>
   </action>
</coordinator-app>
```