Technical Report

Apache Hadoop on Data Fabric Enabled by NetApp
Hadoop Across Data Centers with NFS Connector for Hadoop and NetApp Private Storage

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1 Introduction

This document describes Apache Hadoop solutions with NetApp® FAS NFS Connector in the Data Fabric enabled by NetApp. This Data Fabric uses compute resources from Amazon Web Services (AWS) and storage from NetApp Private Storage (NPS).

This document describes the installation, underlying architecture, use cases, and integration with Hadoop in a hybrid cloud environment. Additionally, it discusses the validation of a number of important Hadoop ecosystem components, such as Impala, Hive, mrjob, Pig, and Apache Spark, with the NetApp FAS NFS Connector. In addition, this report showcases the benefits of performing cloud-based analytics with the Data Fabric enabled by NetApp.

This Data Fabric helps IT organizations to leverage the agility that the cloud promises, such as:

- **Simplified data management and control.** The NetApp universal data platform provides simplified data management and control. The platform offers customers multiple deployment and procurement options to adjust easily and efficiently to align with changing business needs.
- **Dynamic data portability across clouds.** NetApp SnapMirror® and the NetApp OnCommand® Cloud Manager portal provide the ability to move data and applications dynamically among cloud resources. Their universal data platform and dynamic data portability remove the barriers to private and public clouds to create a reliable hybrid cloud environment while maintaining your choice of resources.
- **Extensive choices for technologies, applications, and public cloud service providers.** The NetApp cloud strategy offerings comprise an evolving portfolio of solutions that span private cloud, cloud service providers, and hyperscale cloud providers. Some of the cloud offerings are:
  - NPS for Cloud: Combines the power of NetApp ONTAP® management software with multiple cloud providers such as AWS, Microsoft Azure, and IBM SoftLayer for on-demand performance computing, efficient disaster recovery, and optional multilayer backup and recovery.
  - ONTAP Cloud: Leverages the same benefits of ONTAP in the cloud with on-demand, pay-as-you-go options.
  - NetApp StorageGRID® Webscale: Provides reliable object-based storage that connects private clouds with public cloud resources, including Amazon S3.
  - NetApp AltaVault™ cloud-integrated storage: Opens new pathways to cloud-based backup and archive solutions while preserving existing storage investments.

2 Solution Overview

2.1 Data Fabric Enabled by NetApp

The Data Fabric is NetApp’s vision for the future of data management. A data fabric seamlessly connects different data management environments across disparate clouds into a cohesive, integrated whole. The Data Fabric enabled by NetApp helps organizations to maintain control and choice in how they manage, secure, protect, and access their data across the hybrid cloud, no matter where it is.

Although a data fabric evolves constantly, organizations can start leveraging it by using NetApp technologies that enable data management and seamless data movement across the hybrid cloud. For more information about the Data Fabric powered by NetApp, see WP-7218: NetApp Data Fabric Architecture Fundamentals: Building a Data Fabric Today.

2.2 NPS for AWS

NetApp offers a family of storage solutions to help customers optimize their data use through the AWS cloud. By combining the world’s number-one public cloud and the world’s number-one open networked branded storage operating system (source: IDC Worldwide Quarterly Enterprise Storage Systems Tracker 2015 Q4, March 2016 [Open Networked Enterprise Storage Systems revenue]), you can confidently
manage and maintain control of your data while gaining the flexibility and efficiency benefits of the AWS cloud. You can rent, lease, or buy data storage in or near the cloud to best match your IT resources to your business requirements. Figure 1 illustrates the NPS for AWS architecture.

Figure 1) NPS for AWS architecture.

Table 1 lists the NPS for AWS use cases.

Table 1) NPS for AWS use cases.

<table>
<thead>
<tr>
<th>Use Case</th>
</tr>
</thead>
<tbody>
<tr>
<td>Use Amazon EC2 for large, variable, and enterprise workloads.</td>
</tr>
<tr>
<td>Provide disaster recovery at a lower cost.</td>
</tr>
<tr>
<td>Leverage temporary cloud resources for data center migration and consolidation.</td>
</tr>
<tr>
<td>Adjust your capex-opex spending throughout your storage-as-a-service lifecycle.</td>
</tr>
</tbody>
</table>

NetApp provides the following flexibility and control with NPS for AWS:

- Gain public cloud benefits and proven enterprise storage with a hybrid cloud solution.
- Get secure connectivity to the AWS cloud with AWS Direct Connect.
- Meet compliance and sovereignty requirements.
- Obtain world-class physical security for your data.

NetApp storage solutions with AWS provides the following benefits:

- Obtain AWS cloud benefits with the power of ONTAP.
- Change your capex-opex mix to meet your business needs.
• Accelerate innovation and time to market, maintaining control of your data.

2.3 ONTAP Overview

ONTAP helps achieve results and get products to market faster by providing the throughput and scalability needed to meet the demanding requirements of high-performance computing and digital media content applications. ONTAP also facilitates high levels of performance, manageability, and reliability for large Linux, UNIX, and Microsoft Windows clusters.

NetApp ONTAP features include:
• Scale-up, scale-out, and scale-down are possible with numerous nodes using a global namespace.
• Storage virtualization with storage virtual machines (SVMs) eliminates the physical boundaries of a single controller (memory, CPU, ports, disks, and so on).
• Nondisruptive operations are available when you redistribute load or rebalance capacity combined with network load balancing options within the cluster for upgrading or expanding its nodes.
• NetApp storage efficiency features such as NetApp Snapshot® copies, thin provisioning, space-efficient cloning, deduplication, data compression, and NetApp RAID DP® technology are also available.

The main ONTAP components include:
• SVM: An SVM is a logical file system namespace capable of spanning beyond the boundaries of physical nodes in a cluster.
  – Clients can access virtual servers from any node in the cluster, but only through the associated LIFs.
  – Each SVM has a root volume under which additional volumes are mounted, extending the namespace.
  – Each SVM can span several physical nodes.
  – Each SVM is associated with one or more logical interfaces; clients access the data on the virtual server through the logical interfaces that can live on any node in the cluster.
• LIF: A LIF is essentially an IP address with associated characteristics, such as a home port, a list of ports for failover, a firewall policy, a routing group, and so on.
  – Client network data access is through LIFs dedicated to the SVM.
  – An SVM can have more than one LIF. You can have many clients mounting one LIF or one client mounting several LIFs. This means that IP addresses are no longer tied to a single physical interface.
• Aggregates: An aggregate is a RAID-level collection of disks; it can contain more than one RAID group.
  – Aggregates serve as resources for SVMs and are shared by all SVMs.
• Flexible volumes: A volume is a logical unit of storage. The disk space that a volume occupies is provided by an aggregate.
  – Each volume is associated with one individual aggregate, and therefore with one physical node.
  – In ONTAP, data volumes are owned by an SVM.

2.4 NetApp FAS NFS Connector for Hadoop

The NetApp FAS NFS Connector for Hadoop allows analytics software such as Apache Hadoop and Spark to use ONTAP to access data using the NFS protocol and a simple configuration file change. By using ONTAP, the connector decouples analytics from storage, leveraging the benefits of NAS to share data. ONTAP enables high availability (HA) and storage efficiency. For even higher performance, the NetApp FAS NFS Connector for Hadoop can be combined with Tachyon to build a scale-out caching tier that is backed by ONTAP.
The NetApp FAS NFS Connector for Hadoop design has two deployment options enabling NFS to be leveraged as a non-Hadoop Distributed File System (non-HDFS) or in parallel with HDFS. The deployment options are based on the use cases and the applications used in Apache Hadoop and Apache Spark. Regardless of the file system selected, the user applications don’t have to be modified.

Figure 2) Components of the NetApp FAS NFS Connector for Hadoop.

- **Connection pool**: Makes multiple connections to the storage server; allows the full use of 10GbE.
- **NFS input stream**: Issues large sequential reads (1MB by default) and uses Hadoop semantics for aggressive and intelligent prefetching.
- **File handle cache**: Reduces NFS lookup calls and uses an LRU cache.
- **NFS output stream**: Batches write operations into large inputs/outputs (I/Os) (1MB by default, 64K with ONTAP) and use Hadoop semantics to write in background and commit at task end.

For details about FAS NFS Connector for Hadoop components, see [TR-4382: NetApp FAS NFS Connector for Hadoop](#).

NetApp FAS NFS Connector for Hadoop Design

Figure 3 shows how the NetApp FAS NFS Connector for Hadoop plugs into a typical Hadoop environment.
The NetApp FAS NFS Connector for Hadoop design allows users to run one of two different deployment options:

- Run HDFS as the primary file system and use the connector to analyze data on NFS storage systems (non-HDFS) such as ONTAP. With this approach, users can immediately analyze data on existing NFS-based storage such as NetApp FAS storage arrays. Existing analytics hardware can be used to analyze NFS data, as well.
- Deploy NFS as the primary file system replacing HDFS for the entire Hadoop cluster.

Regardless of the approach taken, the NetApp FAS NFS Connector for Hadoop allows analytics to use existing NetApp technologies such as Snapshot, RAID DP, SnapMirror data replication, and NetApp FlexClone® volumes to provide a rich set of data management features for use with Hadoop.

### 2.5 Solution Benefits and Use Cases

#### Cross-Data-Center Deployments

The decoupled design of NetApp FAS NFS Connector for Hadoop allows independent scaling of compute and storage layers. This capability provides the flexibility to place your analytics compute tier on cloud architectures such as Amazon EC2 while keeping the data safely on storage you control on the premises. In these scenarios, up-front hardware purchases are replaced by the pay-as-you-go cloud model specifically on compute. Cross-data-center deployments benefit from products, such as AWS Direct Connect and NPS, that enable high-bandwidth connections between private data centers and public clouds. Figure 4 shows an example of a cross-data-center deployment with Hadoop enabled by technologies such as ONTAP Cloud, NPS, and SnapMirror.
**Figure 4** Cross-data-center deployments.

**Note:** NPS enables enterprise customers to leverage the performance, availability, security, and compliance of NetApp storage with the economics, elasticity, and time-to-market benefits of the public cloud.

**On-Demand Big Data Analytics**

You can create analytics applications in public clouds, such as in AWS, for your enterprise by building out compute resources ranging from a few to thousands of servers in minutes. Then you can direct them to operate on the specific data in different locations, and then power them down when finished. This process enables you to perform analytics in less time and at a lower cost compared to that of maintaining physical server resources for analytics. NetApp products such as SnapMirror and NPS for AWS enable this use case by providing users with the flexibility to put data where it can be easily accessed by flexible analytics compute resources in the cloud.

**Smart Applications**

You can capture and store terabytes of data per hour from various sources such as website clickstreams, financial transactions, social media feeds, IT logs, and location tracking events. This data can be ingested to centralized NetApp storage controllers or transferred to a colocation facility using SnapMirror. The data can then be processed using cloud compute power such as AWS and machine learning to create real time predictions. Figure 5 shows how the typical Internet of Things sources are delivered to a cloud-based analytics environment using NetApp SnapMirror. There the data is analyzed using cloud-based compute resources with results provided in smart applications such as smartphone devices.
Data Warehousing

You can optimize query performance and reduce costs by deploying a data warehousing architecture in a cloud such as AWS. The Apache Hadoop framework can perform an extract, transform, and load (ETL) process and load the processed data into NPS for big data analytics and business intelligence applications.

2.6 Key Hadoop Components

This section describes the key Hadoop components that were tested and validated with NetApp FAS NFS Connector for Hadoop in NPS/AWS to validate a Hadoop environment using the following use cases:

- **MapReduce functionality and performance** were tested using the following applications:
  - TeraGen: Generates the input data for TeraSort
  - TeraSort: Sorts the data generated
  - TeraValidate: Validates the sorted data output
  - DFSIO: Distributed I/O benchmark for read and write tests on the HDFS and non-HDFS file systems. Tests were conducted in the following areas:
    - Write performance: Benchmark for measuring the write throughput capabilities of the Hadoop cluster
    - Read performance: Benchmark for measuring the read throughput capabilities of the Hadoop cluster

  **Note:** For more validation information, see the section “MapReduce Functionality and Performance Validation.”

- **Data analysis:**
  - Apache Impala: A SQL query engine on Hadoop through which you can query data in HDFS, non-HDFS, or Apache Hbase.
  - Apache Hive: A data warehouse infrastructure built on top of Hadoop for providing data summarization, query, and analysis through SQL.
  - Apache Pig: A high-level platform for creating MapReduce programs used with Hadoop. The platform uses Pig Latin, which abstracts the programming from the Java MapReduce idiom into a MapReduce programming high level such as SQL for RDBMS.
Note: For more validation information, see the section “Data Analysis Validation.”

- Streaming:
  - mrjobs: A Python 2.6+/3.3+ package that helps you write and run Hadoop streaming jobs.

Note: For more validation information, see the section “Mrjob.”

3 Solution Architecture

3.1 Network Architecture

The NetApp test setup consisted of a Hadoop cluster with five slaves and one master, which are m4.10xlarge AWS instances. Based on the workload and performance requirements for your specific environment, you can adjust the number and type of AWS instances. The Hadoop nodes running as AWS instances access the NetApp FAS storage located in an NPS colocation facility through Direct Connect routers using two 10GbE redundant connections. One network connection is active and the other is passive. This enables users to achieve a network throughput of approximately 1GBps between the AWS compute resources and the NetApp storage hosted in the NPS colocation facility. It is possible to increase the available bandwidth between the AWS Hadoop nodes and the storage by configuring multiple 10GbE connections.

Two FAS8060 HA pairs of NetApp storage controllers are connected to layer 3 switches in the Equinix data center. The storage controller provides four volumes for the AWS Hadoop instances through the NetApp FAS NFS Connector for Hadoop using 10GbE (TCP/IP) connection.

Figure 6 is an illustration of the network architecture.
A total of eight LIFs were created for data access, two LIFs for each FAS storage controller. The disk shelves are connected to the FAS controller using 6Gb SAS cables. In this setup, the AWS instances can access the Internet through an AWS public gateway. In Figure 6, each storage controller has 2 x 10Gb Ethernet connections to Layer 3 network switches, which are connected to AWS Direct Connect routers.

### 3.2 Storage Architecture

Table 1 provides details about the hardware and software components and the configurations used in the NetApp test setup.

**Note:** An AWS infrastructure with two aggregates was used in the test setup. Each aggregate contained seven volumes. Two volumes from each aggregate were used for Hadoop testing.

Figure 7 is an illustration of the storage architecture.
Figure 7) Storage architecture.

Table 1) Key components of the validated configuration.

<table>
<thead>
<tr>
<th>Component</th>
<th>Products</th>
<th>Details</th>
</tr>
</thead>
</table>
| Storage      | NetApp FAS8060 with clustered Data ONTAP 8.3 or later | - Four controllers  
- Two aggregates of size 14 x 3.64TB each  
- Each aggregate has seven volumes  
- SATA disk type  
- 7200 RPM  
- RAID DP |
| AWS instances| m4.10xlarge                                    | 40 vCPU  
160GB RAM                                                                 |
| Networking   | AWS Direct Connect routers                     | 10Gb                                                                     |
| OS           | Red Hat Enterprise Linux (RHEL)                | 7.2 or later works in most of the Linux                                |
To validate the performance and stability of this solution, NetApp leveraged a number of standard Hadoop workload tools including TeraGen, TeraSort, TeraValidate, and DFSIO. The results of these tests are described in the following sections.

4.1 TeraGen, TeraSort, and TeraValidate

TeraSort is one of the mostly widely used Hadoop benchmarks. Most of the Hadoop distributions bundle it with their distributions. TeraGen generates random data as the input to the TeraSort tool. TeraGen runs on multiple nodes with multiple tasks by specifying the number of maps, which is based on the number of rows generation and the number of cluster nodes. TeraSort sorts the data generated by TeraGen. TeraValidate makes sure that the output produced by TeraSort is globally sorted.

The TeraGen, TeraSort, and TeraValidate operations were performed on the AWS instances whose data was accessed from NPS using the NetApp FAS NFS Connector for Hadoop and the NFS protocol. For these tests, NetApp created and processed a series of ever larger data sets ranging in size from 93GB–to 466GB. NetApp also measured the elapsed time for each of the tools to complete their processing. Generally, the time required to complete the test depends on the size of the dataset, as shown in Figure 8.

Figure 8 graphs the TeraGen, TeraSort, and TeraValidate results.
In the tested configuration, there was one 10GbE direct connection between the AWS instance and storage; NetApp discovered that this was the primary factor limiting the performance during these tests. As mentioned above, it is possible to increase the number of active connections between the Hadoop nodes and the NPS-based storage if additional performance capacity is required. In any case, NetApp believes that these tests validate the overall stability of the configuration.

Because of the limited network usage as well as the NFS Connector caching feature, the write (TeraGen) and read (TeraValidate) operations were faster than the read/write (TeraSort) operation. For better performance with NFS Connector, make sure to provide multiple direct (active) connections between the AWS instance and NPS.

4.2 DFSIO

During testing, NetApp used the HiBench/DFSIO-E tool to measure the read and write performance of MapReduce jobs behavior between the AWS compute resources and NPS storage through the NetApp FAS NFS Connector for Hadoop. NetApp configured 136 maps per NodeManager, as recommended by the Cloudera tuning YARN. Figure 9 and Figure 10 illustrate the write and read throughput per AWS Hadoop instance.

Note: AWS does not support jumbo frames for Direct Connect. The MTU to be used is 1,500.
Based on the read and write performance results, the AWS Hadoop instance provides better throughput with 4GB file size through NetApp FAS NFS Connector for Hadoop. Additionally, the maximum throughput of 1GBps between AWS instances and NPS was achieved when the throughput of all the AWS instances was combined and was the limiting factor in these tests. Both the FAS8060 storage and AWS-based Hadoop nodes were capable of delivering better performance if additional 10GbE links were applied between AWS and the NPS-based storage systems.

5 Data Analysis Validation

This section discusses additional validation testing that NetApp performed using Hadoop Ecosystem Applications with the NetApp FAS NFS Connector for Hadoop. For this test, NetApp used Impala, Hive, and Pig. The goal of these tests was not to measure performance of the applications but to simply
validate that their basic functionality worked correctly when used with NFS Connector with the NFS protocol.

5.1 Impala

For validation of the NFS Connector with Impala, NetApp used Cloudera Impala 2.6.0, an open-source massively parallel processing SQL query engine for data stored in a computer cluster running Apache Hadoop.

For the NFS Connector to work with Impala, refer to the following procedure:

Note

The IMPALA-1850 fix allows the non-HDFS file system to be set as fs.defaultFS. Refer to the following link for the procedure and code review: http://gerrit.cloudera.org:8080/#/c/1121/.

Basic Validation of Impala with NFS Connector

For the basic validation, NetApp used the following methodology:

1. Create an external table from the CSV file containing all of the records currently available to the Impala application. This file resides in an NFS file system and can be accessed using the NFS Connector.

2. Run a select query to validate the structure of the external table records.

```
[ip-172-30-1-57.ec2.internal:21000] > create external table tab3_nfs_new
()
(id INT,
policynumber INT,
addedrundatecyclenumber INT,
col_4 BOOLEAN,
col_5 DOUBLE,
policydate TIMESTAMP,
state string
)
row format delimited fields terminated by ',',
location 'nfs://172.30.254.29:2049/Hadoop_Vol_1/tab3_nfs';
Query: create external table tab3_nfs_new
()
(id INT,
policynumber INT,
addedrundatecyclenumber INT,
col_4 BOOLEAN,
col_5 DOUBLE,
policydate TIMESTAMP,
state string
)
row format delimited fields terminated by ',',
location 'nfs://172.30.254.29:2049/Hadoop_Vol_1/tab3_nfs'
Fetched 0 row(s) in 0.51s
[ip-172-30-1-57.ec2.internal:21000] > select * from tab3_nfs_new;
Query: select * from tab3_nfs_new
+----+---------+----------------+--------+-------+---------+---------+
| id | policynumber | addedrundatecyclenumber | col_4 | col_5 | policydate | state |
|----|-------------+------------------------+--------+-------+---------+---------+
| 1  | 58337846    | 77964269               | false | 1.1   | 2016-04-06 13:23:01 | AL    |
| 2  | 14570261    | 36802977               | false | 2.2   | 2016-04-06 13:23:01 | AL    |
```

Query: select * from tab3_nfs_new
Fetched 0 row(s) in 0.51s
For more complicated Impala queries that join multiple tables such as select and joins, perform the following procedure to configure the NFS data through the NFS Connector for Hadoop to access the data from Cloudera Impala:

1. Retain the default file system (HDFS) and restart the Hadoop cluster.
2. Verify that the default file system is HDFS by running the `hadoop fs -df -h` command.
3. Create and verify (select query) an external table based on the CSV file, which is located in the non-HDFS file system through the NFS Connector.
4. Verify the `desc` and `select` queries for the NFS tables. A message is displayed indicating that it failed for the NFS tables.
5. Restart only the Impala instances from the Impala page in Cloudera Manager.
6. Run the `select` and `desc` query works for table `tab1` and `tab2`; their location is `nfs://172.30.254.29:2049/user/impala/sample_data/tab1` and `nfs://172.30.254.29:2049/user/impala/sample_data/tab2`.

**Note:** For testing purposes, NetApp updated the `/usr/lib64/cmf/service/impala/impala.sh` file in one of the Impala servers as well as the metadata and catalog server.

```bash
export HADOOP_CONF_CLASSPATH=$(/bin/hadoop classpath)

JDBC_JARS="$CLOUDERA_MYSQL_CONNECTOR_JAR:$CLOUDERA_POSTGRESQL_JDBC_JAR:$CLOUDERA_ORACLE_CONNECTOR_JAR:$HADOOP_CONF_CLASSPATH"
if [ "impalad" = "$1" ]; then
  exec "$IMPALA_HOME/../../bin/impalad" --flagfile=${FLAG_FILE} --abort_on_config_error=false
  ${IMPALA_SERVER_ARGS}
elif [ "statestored" = "$1" ]; then
  exec "$IMPALA_HOME/../../bin/statestored" --flagfile=${FLAG_FILE} --abort_on_config_error=false
  ${IMPALA_SERVER_ARGS}
elif [ "catalogd" = "$1" ]; then
  exec "$IMPALA_HOME/../../bin/catalogd" --flagfile=${FLAG_FILE} --abort_on_config_error=false
  ${IMPALA_SERVER_ARGS}
fi
```

7. Update the `fs.defaultfs` in `core-site.xml` using Snippet (Safety Value) to change the default file system from HDFS to NFS.
8. After the update, do not restart the HDFS instances. Wait until the cluster perspective of the primary file systems is HDFS before restarting the service.
9. Restart the stale config from the HDFS panel and then redeploy the configuration.

**Note:** During testing, NetApp changed the configuration of only one Impala server. You can try testing the configuration for multiple servers.
10. The default file system changes from HDFS to NFS.
11. Verify the join query.

**Note:** For more information, see the section “Impala Qualification with NetApp FAS NFS Connector for Hadoop for Join Query” in the appendix.

### 5.2 Hive

To validate the NetApp FAS NFS Connector with Hive, NetApp used Cloudera Hive to run SQL queries on custom-built warehouse data using the CSV file that contains the records accessible to the database for select and join operations in the AWS Hadoop instance and NPS storage through NetApp FAS NFS Connector. In this validation, NetApp kept the CSV file data in an NPS location that was accessed using the NFS protocol through the NetApp FAS NFS Connector for Hadoop. Figure 11 shows the HiveQL query execution with Hadoop instances running in AWS.

**Figure 11) Hive with NetApp NFS FAS Connector for Hadoop.**

During testing, NetApp had to validate the Hive functionality through SQL queries through NetApp FAS NFS Connector for Hadoop. NetApp created two tables with several million records and successfully performed the select, create, desc, and join queries. The join queries used MapReduce programs to run on the CSV files that contain the records accessible to the database. These records reside in the NFS volume through NetApp FAS NFS Connector for Hadoop.

NetApp tested the following queries in Hive through NFS Connector; you can run other queries based on workload.

1. **Based on the testing, set** `hive.auto.convert.join` **to false before running the queries.**
   ```sql
   set hive.auto.convert.join=false;
   ```

2. **Change** `mapred.reduce.tasks` **from 1 to 20 or the desired value in the Hive configuration.**

3. **Verify the database by running the** `show` **command.**
   ```sql
   show databases
   ```
4. Change into the NFS-specific database by running the `use` command.

```sql
use nfs_db;
```

5. Verify the tables by running the `show` command.

```sql
show tables
```

6. Verify the number of records in the first table by running the `select` query.

```sql
select count(*) from tab3_nfs_new;
```

7. Verify the number of records in the second table by running the `select` query.

```sql
select count(*) from tab4_nfs_new;
```

8. Run the join query by running the `select` query on both tables.

```sql
select count(*) from tab3_nfs_new join tab4_nfs_new where tab3_nfs_new.id = tab4.nfs_new.id;
```

9. You can run all of the SQL queries; for example, to limit the number of records, run the `limit` query.

```sql
select count(*) from tab3_nfs_new join tab4_nfs_new where tab3_nfs_new.id = tab4.nfs_new.id limit 100;
```

**Note:** For detailed results, see the section titled “Hive” in the appendix.

Based on the testing validation, NetApp was able to run the join query against two tables as well as basic SQL queries.

### 6 Apache Pig

Apache Pig is a client-side application for data analysis and processing on Hadoop. The application executes as MapReduce jobs and uses a textual language known as Pig Latin for processing data.
Figure 12 shows the Pig deployment in NPS. For the basic testing validation, NetApp used the following methodology with Pig Latin:

1. Read the data from the CSV file that contains the records accessible to the database by using a LOAD statement.
2. Used transformation statements to process the data.
3. Generated the output of the results by using the DUMP statement.

   **Note:** For the DUMP query to generate data from the NFS data tables, see the section “Pig” in the appendix.

Based on the Pig validation through NFS Connector, NetApp was able to execute and process the Pig Latin queries in the Hadoop AWS instance on data that is in NPS.

   **Note:** NetApp will provide the Pig performance information in the performance validation update of this paper.

7 Mrjob

The mrjob can be used to execute the Hadoop streaming jobs using non-Java programs such as Python. The following example is an mrjob query on NFS data. The data is kept in the NFS file shares and accessed by using NFS Connector.
For our streaming job validation of the NFS Connector, NetApp used the open-source WordCount python program to count the number of characters, lines, and words from the CSV file that is accessed from the NFS volume from NPS storage through the NFS protocol. Figure 13 shows the workflow of streaming jobs with mapper and reducers jobs running in stream for map and reduce operations.

Figure 13 ) Hadoop streaming job with NFS FAS Connector for Hadoop.

NetApp tested the following queries in streaming through NFS Connector; however, you can run other queries based on your workload:

1. Log in as a Hadoop user such as yarn user.
2. Copy the CSV file from the OS to the Hadoop non-HDFS file system. In this example, the NFS file system through NPS storage through the NFS Connector.

3. Verify the CSV file by using the Hadoop CLI command.

4. Run the Python `word_count` program against the CSV file and verify the results. Verify the following command results:

```
(root@ip-172-30-1-117 ~) # su = yarn
Last login: Tue Apr 12 11:15:35 EDT 2016 on pts/0
[yarn@ip-172-30-1-117 ~] $ hadoop fs -ls -h /
Found 2 items
drwxrwxrwt - hdfs supergroup 0 2016-02-25 16:52 /tmp
drwx-xr-x - hdfs supergroup 0 2016-04-04 14:04 /user
[yarn@ip-172-30-1-117 ~] $ hadoop dfs -put /tmp/tab_allstate10_n1.csv /tmp
DEPRECATED: Use of this script to execute hdfs command is deprecated.
Instead use the hdfs command for it.
[yarn@ip-172-30-1-117 ~] $ hadoop fs -ls -h /tmp
Found 4 items
drwxrwxrwx - hdfs supergroup 0 2016-04-12 11:37
/tmp/.cloudera_health_monitoring_canary_files
drwx-wx-x - hive supergroup 0 2016-04-04 20:18 /tmp/hive
drwxrwxrwt - mapred hadoop 0 2016-04-04 14:07 /tmp/logs
-rw-r--r-- 1 yarn supergroup 220.3 M 2016-04-12 11:38 /tmp/tab_allstate10_n1.csv
[yarn@ip-172-30-1-117 ~] $

[yarn@ip-172-30-1-117 ~] $ hadoop fs -ls -h /tmp
Found 4 items
drwxrwxrwx - hdfs supergroup 0 2016-04-12 11:37
/tmp/.cloudera_health_monitoring_canary_files
drwx-wx-x - hive supergroup 0 2016-04-04 20:18 /tmp/hive
drwxrwxrwt - mapred hadoop 0 2016-04-04 14:07 /tmp/logs
-rw-r--r-- 1 yarn supergroup 220.3 M 2016-04-12 11:38 /tmp/tab_allstate10_n1.csv
[yarn@ip-172-30-1-117 ~] $

(yarn@ip-172-30-1-117 ~) $ python mr_word_count.py /tmp/tab_allstate10_n1.csv
No configs found; falling back on auto-configuration
Creating temp directory /tmp/mr_word_count.yarn.20140412.153925.700419
Running step 1 of 1...
Streaming final output from /tmp/mr_word_count.yarn.20140412.153925.700419/output...
"chars" 227246458
"lines" 3716616
"words" 7433232
Removing temp directory /tmp/mr_word_count.yarn.20140412.153925.700419...
[yarn@ip-172-30-1-117 ~] $
```

8 Apache Spark

Apache Spark is a fast, general engine for large-scale data processing on a cluster (including Map and Reduce) as well as for nonbatch processing models. The engine works near-real-time processing and allows configurable in-memory data caching for efficient iteration. It processes applications in a distributed manner across worker nodes to provide distributed storage that is horizontally scalable and fault tolerant. Spark applications are written in Python, Scala, or Java, and consist of one or more tasks. The Spark applications have a Spark driver that runs in YARN client mode (the driver runs on the client) mode or YARN cluster mode (the driver runs on the cluster on the application master, which continue to run even if the client disconnects).

Spark applications include one or more Spark jobs. Jobs run tasks in the executors and the executors run in YARN containers. Each executor runs in a single container. Executors exist throughout the life of an application. The executor is fixed once the application starts and YARN does not resize the already allocated container. An executor can run tasks concurrently on in-memory data.

Apache service has two versions: Spark on YARN, which requires HDFS and YARN and is supported in the latest version of the popular Hadoop distributions; and Spark standalone. Both versions come with the Spark History server.

Figure 14 shows the Spark deployment modes. In the testing validation, NetApp wanted to make sure that Spark applications would run with NetApp NFS Connector in NPS and provide some guidelines based on basic performance results with the WordCount program in Spark through the client deployment mode.
To validate whether the Apache Spark functionality worked with the NFS Connector, NetApp completed the following procedure:

1. In the Hadoop cluster, we modified the primary file system (HDFS) to non-HDFS (in this case, NFS) by changing the `fs.defaultFS` parameter and NFS Connector–specific dependency parameters in `core-site.xml`.

2. We updated the `users.json` and `groups.json` with the Spark user and group details, which are part of NFS Connector configurations.

3. We copied the NFS Connector to the Spark library locations.

4. We updated the connector with execute permissions by running the `chmod` command (if you do not have execute permission).

5. We verified the basic WordCount functionality by running the cluster deploy mode.

The following command validates JavaWordCount:

```
spark-submit --class org.apache.spark.examples.JavaWordCount --master yarn --deploy-mode client
--num-executors 300 --executor-memory 2G --conf "spark.kryoserializer.buffer.max=512m" --conf "spark.rdd.compress=true" --conf "spark.storage.memoryFraction=1" --conf="spark.core.connection.ack.wait.timeout=600" --conf="spark.akka.frameSize=320" --conf "spark.default.parallelism=320" --conf="spark.executor.cores=2" /opt/cloudera/parcels/CDH/jars/spark-examples-1.5.0-cdh5.5.2-hadoop2.6.0-cdh5.5.2.jar
/Hadoop_Vol_1/tab_allstate600g.csv /Hadoop_Vol_1/resultJavaWordCount_600g
```

```
spark-submit --class org.apache.spark.examples.JavaWordCount --master yarn --deploy-mode cluster
--num-executors 300 --executor-memory 2G --conf "spark.kryoserializer.buffer.max=512m" --conf "spark.rdd.compress=true" --conf "spark.storage.memoryFraction=1" --conf="spark.core.connection.ack.wait.timeout=600" --conf="spark.akka.frameSize=320" --conf "spark.default.parallelism=320" --conf="spark.executor.cores=2" /opt/cloudera/parcels/CDH/jars/spark-examples-1.5.0-cdh5.5.2-hadoop2.6.0-cdh5.5.2.jar
/Hadoop_Vol_1/tab_allstate500g.csv /Hadoop_Vol_1/resultJavaWordCount_500g
```
Figure 15 shows the test results with an m4.10xlarge instance. The results indicate that for smaller file sizes of up to 100GB, the WordCount operation completes in approximately the same time as the file fits in AWS Instance memory. In addition, the operation doesn’t require access to the disk. A file size greater than 200GB requires storage access with completion time increasing with the word file size.

Summary

This report provides detailed information about NetApp FAS NFS Connector deployment in the Data Fabric enabled by NetApp Data Fabric, which uses compute resources from AWS and storage from NPS through the underlying architecture and use cases. This document demonstrates NFS Connector’s basic Hadoop MapReduce functionality validation. It also demonstrates Hadoop Ecosystem validation for Hive, Pig, and mrjobs in AWS with NPS and basic performance results for some of them. In addition, it shows that the NFS Connector works with Apache Spark, which is demonstrated through a sample program. As described in this document, NFS Connector supports key Hadoop ecosystem projects such as Apache Hadoop, Apache Spark, Hive, Pig, Impala, and mrjobs in AWS with NPS.

Appendixes

Impala (Create Table and Select Query)

For testing purposes, NetApp created a table and executed the select query through Impala.

1. Create a table with a comma-separated list of rows.
2. Copy the rows into Hadoop using the hadoop fs -put command through NFS Connector.

```
root@ip-172-30-1-117 nfsconnectorjars_from_1211# hadoop fs -put tab_allstate10.csv
nfs://172.30.254.29:2049/Hadoop_Vol_1/
Picked up _JAVA_OPTIONS: -Xmx140000M
```
3. Verify the files in the Hadoop NFS file system.

```
[root@ip-172-30-1-117 nfsc0nnectorjars_from_121]# hadoop fs -ls /
Picked up _JAVA_OPTIONS: -Xmx140000M
Store with ep Endpoint: host-nfs://172.30.254.29:2049/ export=/ has fsId 2147484980
Found 10 items
drwxrwxrwx - root root  4096 2016-04-25 13:05 /.snapshot
drwxrwxrwx - root root  4096 2016-04-25 13:41 /Hadoop_Vol_1
drwxrwxrwx - root root  4096 2016-04-25 13:10 /Hadoop_Vol_2
drwxrwxrwx - root root  4096 2016-04-25 13:10 /Hadoop_Vol_3
drwxrwxrwx - root root  4096 2016-04-25 13:10 /Hadoop_Vol_4
drwxrwxrwx - root root  4096 2016-03-14 16:53 /benchmarks
-rw-r--r--  1 root root  61326 2016-03-08 14:21 /terasort-1.0-SNAPSHOT-2.jar
drwxrwxrwx - root root  4096 2016-04-18 19:28 /tmp
drwxrwxrwx - root root  4096 2016-03-14 10:04 /ttt
drwxrwxrwx - root root  4096 2016-04-04 17:01 /user
```

```
```
16/04/25 13:42:17 INFO nfs.NFSv3FileSystem: Group config file:
/etc/NetAppNFSConnector/conf/groups.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=null path=/Hadoop_Vol_1/ has fsId 2147484980
Found 4 items
-rw-r--r-- 1 root root 4296 2016-04-18 16:58 /Hadoop_Vol_1/README.rst
-rw-r--r-- 1 root root 904546018 2016-04-25 13:41 /Hadoop_Vol_1/tab_allstate10.csv
-rw-r--r-- 1 root 0 2016-03-07 12:25 /Hadoop_Vol_1/test
-rw-r-xr-x 1 root root 4096 2016-03-14 10:08 /Hadoop_Vol_1/tttttttt
[root@ip-172-30-1-117 nfsconnectorjars_from_121]

16/04/25 11:59:01 INFO nfs.NFSv3FileSystem: User config file:
/etc/NetAppNFSConnector/conf/users.json
Picked up _JAVA_OPTIONS: -Xmx140000M
16/04/25 11:59:01 INFO nfs.NFSv3FileSystem: Group config file:
/etc/NetAppNFSConnector/conf/groups.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=null path=/Hadoop_Vol_1/ has fsId 2147484980
Found 4 items
-rw-r--r-- 1 root root 4.2 K 2016-04-18 16:58 /Hadoop_Vol_1/README.rst
drwxrwxrwx - impala impala 4 K 2016-04-25 17:36 /Hadoop_Vol_1/tab3 nfs
-rw-r--r-- 1 root 1 0 2016-03-07 12:25 /Hadoop_Vol_1/test
drwxr-xr-x - root root 4 K 2016-03-14 10:08 /Hadoop_Vol_1/tttttttt
[root@ip-172-30-1-117 ~]# hadoop fs -ls -h /Hadoop_Vol_1
Picked up _JAVA_OPTIONS: -Xmx140000M
16/04/25 11:59:15 INFO nfs.NFSv3FileSystem: User config file:
/etc/NetAppNFSConnector/conf/users.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=null path=/Hadoop_Vol_1/ has fsId 2147484980
Found 4 items
-rw-r--r-- 1 root root 4.2 K 2016-04-18 16:58 /Hadoop_Vol_1/README.rst
drwxrwxrwx - impala impala 4 K 2016-04-25 17:36 /Hadoop_Vol_1/tab3 nfs
-rw-r--r-- 1 root 1 0 2016-03-07 12:25 /Hadoop_Vol_1/test
drwxr-xr-x - root root 4 K 2016-03-14 10:08 /Hadoop_Vol_1/tttttttt
[root@ip-172-30-1-117 ~]# hadoop fs -ls -h /Hadoop_Vol_1/tab3 nfs
Picked up _JAVA_OPTIONS: -Xmx140000M
16/04/25 11:59:15 INFO nfs.NFSv3FileSystem: User config file:
/etc/NetAppNFSConnector/conf/users.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=null path=/Hadoop_Vol_1/ has fsId 2147484980
Found 4 items
-rw-r--r-- 1 root root 862.6 M 2016-04-25 13:41 /Hadoop_Vol_1/tab3 nfs/tab_allstate10.csv

4. Create a folder and move the CSV file into the new folder.

[impala@ip-172-30-1-117 ~]$ hadoop fs -mkdir nfs://172.30.254.29:2049/Hadoop_Vol_1/tab3 nfs
DEPRECATED: Use of this script to execute hdfs command is deprecated.
Instead use the hdfs command for it.
Picked up _JAVA_OPTIONS: -Xmx140000M
16/04/25 17:36:54 INFO nfs.NFSv3FileSystem: User config file:
/etc/NetAppNFSConnector/conf/users.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=null path=/Hadoop_Vol_1/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=./ path=./ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=./ path=./ has fsId 2147484980
[impala@ip-172-30-1-117 ~]$ hadoop fs -mv
nfs://172.30.254.29:2049/Hadoop_Vol_1/tab_allstate10.csv
nfs://172.30.254.29:2049/Hadoop_Vol_1/tab3 nfs
DEPRECATED: Use of this script to execute hdfs command is deprecated.
Instead use the hdfs command for it.
Picked up _JAVA_OPTIONS: -Xmx140000M
16/04/25 17:37:23 INFO nfs.NFSv3FileSystem: User config file:
/etc/NetAppNFSConnector/conf/users.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=null path=/Hadoop_Vol_1/ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=./ path=./ has fsId 2147484980
Store with ep Endpoint: host=nfs://172.30.254.37:2049/ export=./ path=./ has fsId 2147484980
[impala@ip-172-30-1-117 ~]$
5. Create an external table based on the CSV file.

```sql
(ip-172-30-1-57.ec2.internal:21000] > create external table tab3_nfs_new
(
  id INT,
policynumber INT,
addedrundatecyclenumber INT,
col_4 BOOLEAN,
col_5 DOUBLE,
policydate TIMESTAMP,
state string
)
row format delimited fields terminated by ','
location 'nfs://172.30.254.29:2049/Hadoop_Vol_1/tab3_nfs';
Query: create external table tab3_nfs_new
(
  id INT,
policynumber INT,
addedrundatecyclenumber INT,
col_4 BOOLEAN,
col_5 DOUBLE,
policydate TIMESTAMP,
state string
)
row format delimited fields terminated by ','
location 'nfs://172.30.254.29:2049/Hadoop_Vol_1/tab3_nfs'
Fetched 0 row(s) in 0.51s
[ip-172-30-1-57.ec2.internal:21000] > select * from tab3_nfs_new;
Query: select * from tab3_nfs_new
+-----+----------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------
| id  | policynumber  | addedrundatecyclenumber | col_4 | col_5      | policydate              | state       | id  | policynumber  | addedrundatecyclenumber | col_4 | col_5      | policydate              | state       | id  | policynumber  | addedrundatecyclenumber | col_4 | col_5      | policydate              | state       |
+-----+----------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------------------------+-------------------------+-----+-----+-------
```
Impala Qualification with NetApp FAS NFS Connector for Hadoop for Join Query

This section details the Impala qualification for the join query between two external tables that are created from the CSV files.

1. HDFS is the default file system. We don’t have NFS in core-site.xml Snippet (Safety Value). Update the core-site.xml file with the new NFS Connector configuration to use NFS instead of HDFS.
2. Restart the cluster.

3. Verify the status of the cluster.

4. The details of the cluster host are displayed.
5. Select the Status tab and verify that the health status of the Impala server is good (green).

6. Select the Instance tab to display the details of the Impala instance.
7. From the CLI, verify that the default file system is HDFS.

```
[impala@ip-172-30-1-117 ~]$ hostname
ip-172-30-1-117.ec2.internal
[impala@ip-172-30-1-117 ~]$ id
uid=979(impala) gid=975(impala) groups=975(impala),980(hive)
[impala@ip-172-30-1-117 ~]$ hadoop dfs -df -h
DEPRECATED: Use of this script to execute hdfs command is deprecated.
Instead use the hdfs command for it.
```

```
Filesystem Size Used Available Use%
hdfs://ip-172-30-1-117.ec2.internal:8020 112.4 G 1.4 G 85.2 G 1%
```

```
[impala@ip-172-30-1-117 ~]$ hadoop dfs -ls nfs://172.30.254.29:2049/user/impala/sample_data/tab*
DEPRECATED: Use of this script to execute hdfs command is deprecated.
Instead use the hdfs command for it.
```

```
16/04/07 14:38:09 INFO nfs.NFSv3FileSystem: User config file: /etc/NetAppNFSConnector/conf/users.json
16/04/07 14:38:09 INFO nfs.NFSv3FileSystem: Group config file: /etc/NetAppNFSConnector/conf/groups.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
Found 1 items
Found 1 items
```

8. Check the table's CSV files in the NFS location, which can be accessed by the NFS Connector using the /etc/NetAppNFSConnector/conf/nfs-mapping.json configuration through the secondary file system in Hadoop. The current primary file system until now has been HDFS.

```
[impala@ip-172-30-1-117 ~]$ impala-shell -i ip-172-30-1-57.ec2.internal -c
Starting Impala Shell without Kerberos authentication
Connected to ip-172-30-1-57.ec2.internal:21000
Server version: impalad version 2.3.0-cdh5.5.2 RELEASE (build cc1125f10419a7269366f17f950f5f7b24b07ac64)
******************************************************************************
Welcome to the Impala shell. Copyright (c) 2015 Cloudera, Inc. All rights reserved.
(Impala Shell v2.3.0-cdh5.5.2 (cc1125f) built on Mon Jan 25 16:03:27 PST 2016)
```

When pretty-printing is disabled, you can use the 'output_delimiter' flag to set
10. The desc and select queries fail for NFS tables.

```sql
[ip-172-30-1-57.ec2.internal:21000] > desc formatted tab1;
Query: describe formatted tab1
ERROR: AnalysisException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: ExecutionException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: TableLoadingException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: TableLoadingException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil

[ip-172-30-1-57.ec2.internal:21000] > desc formatted tab2;
Query: describe formatted tab2
ERROR: AnalysisException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: ExecutionException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: TableLoadingException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: TableLoadingException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil

[ip-172-30-1-57.ec2.internal:21000] > select * from tab1;
Query: select * from tab1
ERROR: AnalysisException: Failed to load metadata for table: 'tab1'
CAUSED BY: TableLoadingException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: ExecutionException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil

[ip-172-30-1-57.ec2.internal:21000] > select * from tab2;
```
Query: `select * from tab2`
ERROR: AnalysisException: Failed to load metadata for table: 'tab2'
CAUSED BY: TableLoadingException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: ExecutionException: java.lang.NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil
CAUSED BY: NoClassDefFoundError: Could not initialize class com.cloudera.impala.util.HdfsCachingUtil

11. Restart only the Impala instances from the Impala page.
12. The `select` and `desc` query should now work for tables `tab1` and `tab2`; their locations are
nfs:///172.30.254.29:2049/user/impala/sample_data/tab1 and

```
[impala@ip-172-30-1-117 ~]$ impala-shell -i ip-172-30-1-57.ec2.internal -c
Starting Impala Shell without Kerberos authentication
Connected to ip-172-30-1-57.ec2.internal:21000
Server version: impalad version 2.3.0-cdh5.5.2 RELEASE (build cc1125f10419a7269366f7f950f57b24b07acd64)
***********************************************************************************
Welcome to the Impala shell. Copyright (c) 2015 Cloudera, Inc. All rights reserved.
(Impala Shell v2.3.0-cdh5.5.2 (cc1125f) built on Mon Jan 25 16:03:27 PST 2016)
To see a summary of a query's progress that updates in real-time, run 'set
LIVE_PROGRESS=1;'.
***********************************************************************************
[ip-172-30-1-57.ec2.internal:21000] > use nfs_db;
Query: use nfs_db
[ip-172-30-1-57.ec2.internal:21000] > select * from tab2;
Query: select * from tab2
+-----+-------+---------------------+
| id  | col_1 | col_2              |
+-----+-------+---------------------+
| 1   | true  | 12789.123          |
| 2   | false | 1243.5             |
| 3   | false | 24453.325          |
| 4   | false | 2423.325           |
| 5   | true  | 243.325            |
| 60  | false | 243565423.325      |
| 70  | true  | 243.325            |
| 80  | false | 243423.325         |
| 90  | true  | 243.325            |
+-----+-------+---------------------+
Fetched 9 row(s) in 0.99s
[ip-172-30-1-57.ec2.internal:21000] > select * from tab1;
Query: select * from tab1
+-----+-------+-------+-------+
| id  | col_1 | col_2 | col_3 |
+-----+-------+-------+-------+
| 1   | true  | 123.123| 2012-10-24 08:55:00 |
| 2   | false | 1243.5 | 2012-10-25 13:40:00 |
| 3   | false | 24453.325| 2008-08-22 09:33:21.123000000 |
| 4   | false | 243423.325| 2007-05-12 22:32:32.133454000 |
| 5   | true  | 243.325| 1953-04-22 09:11:33 |
+-----+-------+-------+-------+
Fetched 5 row(s) in 0.29s
[ip-172-30-1-57.ec2.internal:21000] > desc formatted tab1;
Query: describe formatted tab1
+-----------------------------+-----------------------------+-----------------------------+
| name | type                        |
| comment |
+-----------------------------+-----------------------------+-----------------------------+
| # col_name                  | data_type                  |
| comment | | | |
| NULL | | | |
| id | int | | |
| NULL | | | |
| col_1 | boolean | | |
| NULL | | | |
| col_2 | double | | |
| NULL | | | |
| col_3 | timestamp | | |
| NULL | | | |
| # Detailed Table Information | NULL | | |
| NULL | | | |
```
| Database: | nfs_db |
| Owner: | impala |
| CreateTime: | Mon Apr 04 19:14:54 EDT 2016 |
| LastAccessTime: | UNKNOWN |
| Protect Mode: | None |
| Retention: | 0 |
| Location: | nfs://172.30.254.29:2049/user/impala/sample_data/tab1 |
| Table Type: | MANAGED_TABLE |
| Table Parameters: | NULL |
| false | COLUMN_STATS_ACCURATE |
| numFiles | 0 |
| numRows | -1 |
| rawDataSize | -1 |
| totalSize | 0 |
| transient_lastDdlTime | 1459811694 |
| # Storage Information | NULL |
| SerDe Library: | org.apache.hadoop.hive.serde2.lazy.LazySimpleSerDe |
| InputFormat: | org.apache.hadoop.mapred.TextInputFormat |
| OutputFormat: | org.apache.hadoop.hive.ql.io.HiveIgnoreKeyTextOutputFormat |
| Compressed: | No |
| Num Buckets: | 0 |
| Bucket Columns: | [] |
| Sort Columns: | [] |
| Storage Desc Params: | NULL |
| field.delim | , |
| serialization.format | , |
| # Storage Information | NULL |

Fetched 35 row(s) in 0.05s
| col_1 | boolean | NULL |
| col_2 | double | NULL |
|       | boolean | NULL |
| # Detailed Table Information | NULL |
| Database: | nfs_db | NULL |
| Owner: | impala | NULL |
| CreateTime: | Mon Apr 04 19:21:10 EDT 2016 | NULL |
| LastAccessTime: | UNKNOWN | NULL |
| Protect Mode: | None | NULL |
| Retention: | 0 | NULL |
| Location: | nfs://172.30.254.29:2049/user/impala/sample_data/tab2 | NULL |
| Table Type: | MANAGED_TABLE | NULL |
| Table Parameters: | NULL | NULL |
|COLUMN_STATS_ACCURATE | false |
|numFiles | 0 |
|numRows | -1 |
|rawDataSize | -1 |
totalSize | 0 |
transient_lastDdlTime | NULL |
1459812070 | NULL |
# Storage Information | NULL |
|SerDe Library: | org.apache.hadoop.hive.serde2.lazy.LazySimpleSerDe | NULL |
|InputFormat: | org.apache.hadoop.mapred.TextInputFormat | NULL |
|OutputFormat: | org.apache.hadoop.hive.ql.io.HiveIgnoreKeyTextOutputFormat | NULL |
|Compressed: | No | NULL |
|Num Buckets: | 0 | NULL |
|Bucket Columns: | [] | NULL |
|Sort Columns: | [] | NULL |
|Storage Desc Params: | NULL | NULL |
|field.delim | , |
|serialization.format | , |

Fetched 34 row(s) in 0.01s
[ip-172-30-1-57.ec2.internal:21000] > select * from tab1 join tab2;
Query: select * from tab1 join tab2
ERROR: Failed to connect to FS: nfs://172.30.254.29:2049/
Error(255): Unknown error 255
13. For testing purposes, update the `/usr/lib64/cmfservice/impala/impala.sh` file in one of the Impalad servers (172.30.1.57) as well as the metadata and catalogd server (172.30.1.117).

```bash
export HADOOP_CONF_CLASSPATH=$(/bin/hadoop classpath)

JDBC_JARS="$CLOUDERA_MYSQL_CONNECTOR_JAR:$CLOUDERA_POSTGRESQL_JDBC_JAR:$CLOUDERA_ORACLE_CONNECTOR_JAR:$HADOOP_CONF_CLASSPATH"

if [ "impalad" = "$1" ]; then
    exec "$IMPALA_HOME/../../bin/impalad" --flagfile=${FLAG_FILE} --abort_on_config_error=false ${IMPALA_SERVER_ARGS}
elif [ "statestore" = "$1" ]; then
    exec "$IMPALA_HOME/../../bin/statestored" --flagfile=${FLAG_FILE} --abort_on_config_error=false ${IMPALA_SERVER_ARGS}
elif [ "catalogd" = "$1" ]; then
    exec "$IMPALA_HOME/../../bin/catalogd" --flagfile=${FLAG_FILE} --abort_on_config_error=false ${IMPALA_SERVER_ARGS}
fi
```

14. Update `fs.defaultfs` in `core-site.xml` by using Snippet (Safety Value) to change the default file system from HDFS to NFS.

```xml
<property>
  <name>fs.defaultFS</name>
  <value>nfs://172.30.254.29:2049</value>
</property>

<property>
  <name>fs.AbstractFileSystem.nfs.impl</name>
  <value>org.apache.hadoop.fs.nfs.NFSv3AbstractFilesystem</value>
</property>

<property>
  <name>fs.nfs.impl</name>
  <value>org.apache.hadoop.fs.nfs.NFSv3FileSystem</value>
</property>

<property>
  <name>fs.nfs.configuration</name>
  <value>/etc/NetAppNFSConnector/conf/nfs-mapping.json</value>
</property>
```
15. After the update, do not restart the HDFS instances.

```
[impala@ip-172-30-1-117 ~]$ hadoop dfs df -h
DEPRECATED: Use of this script to execute hdfs command is deprecated.
Instead use the hdfs command for it.
```

```
Filesystem                     Size   Used  Available  Use%
hdfs://ip-172-30-1-117.ec2.internal:8020  112.4 G  1.4 G  85.2 G  1%
```
17. Restart stale config from the HDFS panel and redeploy the configuration.
18. After a successful restart, verify that the configuration was changed for only one Impala server.
19. Verify that the default file system changed from HDFS to NFS.

```bash
[impala@ip-172-30-1-117 ~]$ hadoop dfs -df -h
DEPRECATED: Use of this script to execute hdfs command is deprecated. Instead use the hdfs command for it.

16/04/07 16:15:31 INFO nfs.NFSv3FileSystem: Group config file: /etc/NetAppNFSConnector/conf/groups.json
Store with ep Endpoint: host=nfs://172.30.254.29:2049/export=/ has fsId 2147484980

Filesystem Size Used Available Use%
--- ----------- ------- ------- ---
nfs://172.30.254:29:2049 8.0 E 0 8.0 E 0%

[impala@ip-172-30-1-117 ~]$ 
```

20. Verify the join query.

```bash
[impala@ip-172-30-1-117 ~]$ impala-shell -i ip-172-30-1-57.ec2.internal -c
Starting Impala Shell without Kerberos authentication
Connected to ip-172-30-1-57.ec2.internal:21000
Server version: impalad version 2.3.0-cdh5.5.2 RELEASE (build cc1125f10419a726936f7f950f57b24b07ad64)
***********************************************************************************
Welcome to the Impala shell. Copyright (c) 2015 Cloudera, Inc. All rights reserved.
Impala Shell v2.3.0-cdh5.5.2 (cc1125f) built on Mon Jan 25 16:09:27 PST 2016
The HISTORY command lists all shell commands in chronological order.
***********************************************************************************
```
<table>
<thead>
<tr>
<th>Query: show databases</th>
</tr>
</thead>
<tbody>
<tr>
<td>name</td>
</tr>
<tr>
<td>_impala_builtins</td>
</tr>
<tr>
<td>default</td>
</tr>
<tr>
<td>nfs_db</td>
</tr>
<tr>
<td>------------------</td>
</tr>
<tr>
<td>Fetched 3 row(s) in 0.09s</td>
</tr>
<tr>
<td>Query: use nfs_db</td>
</tr>
<tr>
<td>Query: show tables</td>
</tr>
<tr>
<td>name</td>
</tr>
<tr>
<td>tab1</td>
</tr>
<tr>
<td>tab2</td>
</tr>
<tr>
<td>test1_nfs</td>
</tr>
<tr>
<td>test2_nfs</td>
</tr>
<tr>
<td>test_nfs</td>
</tr>
<tr>
<td>------------------</td>
</tr>
<tr>
<td>Fetched 5 row(s) in 0.01s</td>
</tr>
<tr>
<td>Query: describe formatted tab1</td>
</tr>
<tr>
<td>name</td>
</tr>
<tr>
<td>comment</td>
</tr>
<tr>
<td># col_name</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>NULL</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>NULL</td>
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<tr>
<td></td>
</tr>
<tr>
<td>NULL</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>NULL</td>
</tr>
<tr>
<td># Detailed Table Information</td>
</tr>
<tr>
<td>Database:</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Owner:</td>
</tr>
<tr>
<td>NULL</td>
</tr>
<tr>
<td>NULL</td>
</tr>
<tr>
<td>NULL</td>
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<tr>
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<td>NULL</td>
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<tr>
<td>NULL</td>
</tr>
<tr>
<td>NULL</td>
</tr>
<tr>
<td>name</td>
</tr>
<tr>
<td>--------</td>
</tr>
<tr>
<td>id</td>
</tr>
<tr>
<td>col_1</td>
</tr>
<tr>
<td>col_2</td>
</tr>
</tbody>
</table>

**Storage Information**

- SerDe Library: org.apache.hadoop.hive.serde2.lazy.LazySimpleSerDe
- OutputFormat: org.apache.hadoop.hive.ql.io.HiveIgnoreKeyTextOutputFormat
- Compressed: No
- Num Buckets: 0
- Bucket Columns: []
- Sort Columns: []
- Storage Desc Params: field.delim, serialization.format

```
Fetched 35 row(s) in 0.06s
[ip-172-30-1-57.ec2.internal:21000] > desc formatted tab2;
Query: describe formatted tab2
```
null
<table>
<thead>
<tr>
<th>id</th>
<th>col_1</th>
<th>col_2</th>
<th>col_3</th>
<th>id</th>
<th>col_1</th>
<th>col_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>1</td>
<td>true</td>
<td>12789.123</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>2</td>
<td>false</td>
<td>1243.5</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>3</td>
<td>false</td>
<td>24453.325</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>4</td>
<td>false</td>
<td>2423.3254</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>5</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>60</td>
<td>false</td>
<td>243565423.325</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>70</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>80</td>
<td>false</td>
<td>243423.325</td>
</tr>
<tr>
<td>1</td>
<td>true</td>
<td>123.123</td>
<td>2012-10-24 08:55:00</td>
<td>90</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>1</td>
<td>true</td>
<td>12789.123</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>2</td>
<td>false</td>
<td>1243.5</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>3</td>
<td>false</td>
<td>24453.325</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>4</td>
<td>false</td>
<td>2423.3254</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>5</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>60</td>
<td>false</td>
<td>243565423.325</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>70</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>80</td>
<td>false</td>
<td>243423.325</td>
</tr>
<tr>
<td>1</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>90</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>1</td>
<td>true</td>
<td>12789.123</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>2</td>
<td>false</td>
<td>1243.5</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>3</td>
<td>false</td>
<td>24453.325</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>4</td>
<td>false</td>
<td>2423.3254</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>5</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>60</td>
<td>false</td>
<td>243565423.325</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>70</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>80</td>
<td>false</td>
<td>243423.325</td>
</tr>
<tr>
<td>2</td>
<td>false</td>
<td>1243.5</td>
<td>2012-10-25 13:40:00</td>
<td>90</td>
<td>true</td>
<td>243.325</td>
</tr>
<tr>
<td>3</td>
<td>false</td>
<td>24453.325</td>
<td>2008-08-22 09:33:21.123000000</td>
<td>1</td>
<td>true</td>
<td>12789.123</td>
</tr>
<tr>
<td>3</td>
<td>false</td>
<td>24453.325</td>
<td>2008-08-22 09:33:21.123000000</td>
<td>2</td>
<td>false</td>
<td>1243.5</td>
</tr>
<tr>
<td>3</td>
<td>false</td>
<td>24453.325</td>
<td>2008-08-22 09:33:21.123000000</td>
<td>3</td>
<td>false</td>
<td>24453.325</td>
</tr>
<tr>
<td>3</td>
<td>false</td>
<td>24453.325</td>
<td>2008-08-22 09:33:21.123000000</td>
<td>4</td>
<td>false</td>
<td>2423.3254</td>
</tr>
<tr>
<td>3</td>
<td>false</td>
<td>24453.325</td>
<td>2008-08-22 09:33:21.123000000</td>
<td>5</td>
<td>true</td>
<td>243.325</td>
</tr>
</tbody>
</table>
Picked up _JAVA_OPTIONS:

Picked up _JAVA_OPTIONS:

Hive

1. Find the available databases.

[root@ip-172-30-1-117 ~]# hive
Picked up _JAVA_OPTIONS: -Xmx140000M
Picked up _JAVA_OPTIONS: -Xmx140000M
Picked up _JAVA_OPTIONS: -Xmx140000M

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2. Find the tables in the NFS database.

hive> use nfs_db;
OK
Time taken: 0.036 seconds
hive> show tables;
OK
allstatetest
tab1
tab2
tab3 nfs
tab3 nfs_new
tab4 nfs
tab4 nfs_new
test1 nfs
test2 nfs
test nfs
Time taken: 0.029 seconds, Fetched: 10 row(s)
hive>

3. Find the number of records in the table; for example, tab3 nfs_new.

hive> select count(*) from tab3 nfs_new;
Query ID = root_20160509120909_3f8673b8-3fdd-4615-b265-82808fb2bed3
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.reducers=<number>
Starting Job = job_1462586435822_0026, Tracking URL = http://ip-172-30-1-117.ec2.internal:8088/proxy/application_1462586435822_0026/
Kill Command = /opt/cloudera/parcels/CDH-5.5.2-1.cdh5.5.2.p0.4/lib/hadoop/bin/hadoop job -kill job_1462586435822_0026
Hadoop job information for Stage-1: number of mappers: 3; number of reducers: 1
2016-05-09 12:09:42,712 Stage-1 map = 0%, reduce = 0%
2016-05-09 12:09:54,002 Stage-1 map = 67%, reduce = 0%, Cumulative CPU 22.26 sec
2016-05-09 12:10:01,222 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 24.14 sec
MapReduce Total cumulative CPU time: 24 seconds 140 msec
Ended Job = job_1462586435822_0026
MapReduce Jobs Launched:
Stage-Stage-1: Map: 3 Reduce: 1 Cumulative CPU: 24.14 sec HDFS Read: 0 HDFS Write: 0 SUCCESS
Total MapReduce CPU Time Spent: 24 seconds 140 msec
OK
14216336
Time taken: 26.245 seconds, Fetched: 1 row(s)

4. Find the number of records in the table; for example, tab4 nfs_new.

hive> select count(*) from tab4 nfs_new;
Query ID = root_20160509121010_56761d96-35a9-4f4b-a4c7-89bf47db4223
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>
```


In order to change the average load for a reducer (in bytes):

```
set hive.exec.reducers.bytes.per.reducer=<number>
```

MapReduce job information for Stage-1: number of mappers: 6; number of reducers: 28

2016-05-09 12:13:04,082 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 25.09 sec
2016-05-09 12:13:07,180 Stage-1 map = 0%, reduce = 0%
2016-05-09 12:10:46,654 Stage-1 map = 33%, reduce = 0%, Cumulative CPU 7.3 sec
2016-05-09 12:10:47,687 Stage-1 map = 78%, reduce = 0%, Cumulative CPU 22.03 sec
2016-05-09 12:10:49,746 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 23.63 sec
2016-05-09 12:10:56,938 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 25.09 sec

MapReduce Total cumulative CPU time: 25 seconds 90 msec

Ended Job = job_1462586435822_0028

Total MapReduce CPU Time Spent: 25 seconds 90 msec

OK
14987634
Time taken: 27.456 seconds, Fetched: 1 row(s)

5. Run the join query to find the total records in the two tables. This example uses tables `tab3_nfs_new` and `tab4_nfs_new`.

```
hive> select count(*) from tab3_nfs_new join tab4_nfs_new where tab3_nfs_new.id = tab4_nfs_new.id;
Query ID = root_20160509121212_eb6b05ed-51da-4140-af87-ffcc33af8aa51
Total jobs = 2
Launching Job 1 out of 2
Number of reduce tasks not specified. Estimated from input data size: 28
In order to change the average load for a reducer (in bytes):

```
set hive.exec.reducers.bytes.per.reducer=<number>
```

In order to set the constant number of reducers:

```
set mapreduce.job.reduces=<number>
```


MapReduce job information for Stage-1: number of mappers: 6; number of reducers: 28

2016-05-09 12:13:04,082 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 25.09 sec
2016-05-09 12:13:17,422 Stage-1 map = 11%, reduce = 0%, Cumulative CPU 56.4 sec
2016-05-09 12:13:18,453 Stage-1 map = 33%, reduce = 0%, Cumulative CPU 78.76 sec
2016-05-09 12:13:20,510 Stage-1 map = 44%, reduce = 0%, Cumulative CPU 82.06 sec
2016-05-09 12:13:21,537 Stage-1 map = 61%, reduce = 0%, Cumulative CPU 102.72 sec
2016-05-09 12:13:23,587 Stage-1 map = 78%, reduce = 0%, Cumulative CPU 111.01 sec
2016-05-09 12:13:23,691 Stage-1 map = 89%, reduce = 0%, Cumulative CPU 123.09 sec
2016-05-09 12:13:28,718 Stage-1 map = 94%, reduce = 0%, Cumulative CPU 125.01 sec
2016-05-09 12:13:30,773 Stage-1 map = 96%, reduce = 0%, Cumulative CPU 128.32 sec
2016-05-09 12:13:31,801 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 129.94 sec
2016-05-09 12:13:40,030 Stage-1 map = 100%, reduce = 86%, Cumulative CPU 356.9 sec
2016-05-09 12:13:42,090 Stage-1 map = 100%, reduce = 90%, Cumulative CPU 380.01 sec
2016-05-09 12:13:43,117 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 466.15 sec

MapReduce Total cumulative CPU time: 7 minutes 46 seconds 150 msec

Ended Job = job_1462586435822_0028

Total MapReduce CPU Time Spent: 7 minutes 46 seconds 150 msec

OK
14987634
Time taken: 27.456 seconds, Fetched: 1 row(s)

6. Run the join query between two tables. This example uses tables `tab3_nfs_new` and `tab4_nfs_new`.

```hive
hive> select * from tab3_nfs_new join tab4_nfs_new where tab3_nfs_new.id = tab4_nfs_new.id limit 100;
```

In order to set a constant number of reducers:
```bash
set hive.exec.reducers.max=<number>
```

In order to set the maximum number of reducers:
```bash
set hive.exec.reducers.max=<number>
```


Hadoop job information for Stage-1: number of mappers: 6; number of reducers: 28

```
2016-05-09 12:14:56,684 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 332.55 sec
2016-05-09 12:14:55,938 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 227.44 sec
2016-05-09 12:14:54,438 Stage-1 map = 87%, reduce = 0%, Cumulative CPU 219.39 sec
2016-05-09 12:14:53,871 Stage-1 map = 78%, reduce = 0%, Cumulative CPU 200.48 sec
2016-05-09 12:14:53,303 Stage-1 map = 6%, reduce = 0%, Cumulative CPU 124.21 sec
```

```
2016-05-09 12:13:56,977 Stage-2 map = 0%, reduce = 0%
2016-05-09 12:13:57,134 Stage-2 map = 100%, reduce = 0%, Cumulative CPU 2.12 sec
2016-05-09 12:14:03,291 Stage-2 map = 100%, reduce = 100%, Cumulative CPU 3.58 sec
```

MapReduce Total cumulative CPU time: 3 seconds 580 msec
Ended Job = job_1462586435822_0029

MapReduce Jobs Launched:
```
Stage-Stage-1: Map: 6 Reduce: 28 Cumulative CPU: 466.15 sec HDFS Read: 0 HDFS Write: 0 SUCCESS
Stage-Stage-2: Map: 1 Reduce: 1 Cumulative CPU: 3.58 sec HDFS Read: 0 HDFS Write: 0 SUCCESS
```

Total MapReduce CPU Time Spent: 7 minutes 49 seconds 730 msec
OK
799657050
Time taken: 66.984 seconds, Fetched: 1 row(s)

hive>

Note: The records shown in this example are truncated for brevity.

Parameters:

- `fs.default.name` is deprecated. Instead, use `fs.defaultFS`.

```
28 64369998 97268880  true  117.8  2016-04-06
13:45:12 VT 28  9215379  59750993  true  69.8  2016
-04-06 13:38:06 MA
28 64369998 97268880  true  117.8  2016-04-06
13:45:12 VT 28  9215379  59750993  true  69.8  2016
-04-06 13:38:06 MA
28 64369998 97268880  true  117.8  2016-04-06
13:45:12 VT 28  9215379  59750993  true  69.8  2016
-04-06 13:38:06 MA
```

---

**Pig**

1. Start the Pig shell.

```
[impala@ip-172-30-1-117 ~]$ pig
log4j:WARN No appenders could be found for logger (org.apache.hadoop.util.Shell).
log4j:WARN Please initialize the log4j system properly.
log4j:WARN See http://logging.apache.org/log4j/1.2/faq.html#noconfig for more info.
2016-04-07 12:00:05,019 [main] INFO org.apache.hadoop.conf.Configuration.deprecation -.mapred.job.tracker is deprecated. Instead, use mapreduce.jobtracker.address
2016-04-07 12:00:05,038 [main] INFO org.apache.pig.impl.util.Utils
2016-04-07 12:00:05,054 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,065 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,069 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,073 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,077 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,079 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,084 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,088 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,092 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,094 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,098 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
2016-04-07 12:00:05,107 [main] INFO org.apache.hadoop.conf.Configuration.deprecation
```

```
2. Read the data from the NFS location by using the LOAD statement.

```
grunt> A = LOAD 'nfs://172.30.254.29:2049/user/impala/sample_data/tab2/tab2.csv' USING PigStorage(',');
Store with ep Endpoint: host=nfs://172.30.254.29:2049/ export=/ path=/ has fsId 2147484980
```

```
grunt> describe A;
A: {id: int, col_1: boolean, col_2: float}
```

3. Generate the output of the NFS table.

```
grunt> dump A;
```

---

**Note:** The output from the previous command was truncated for brevity.
Input(s):
Successfully read 9 records from:
"nfs://172.30.254.29:2049/user/impala/sample_data/tab2/tab2.csv"

Output(s):
Successfully stored 9 records in: "nfs://172.30.254.29:2049/tmp/temp1489487504/tmp942386654"

Counters:
Total records written : 9
Total bytes written : 0
Spillable Memory Manager spill count : 0
Total bags proactively spilled: 0
Total records proactively spilled: 0

Job DAG:
job_1460041670554_0004

References
The following references were used in this technical report:

- Cloudera Impala
  https://en.wikipedia.org/wiki/Cloudera_Impala

- Apache Hive
  https://en.wikipedia.org/wiki/Apache_Hive

- Mrjobs
  https://github.com/Yelp/mrjob

- TR-4133: NetApp Private Storage for AWS

- TR-4067: Clustered Data ONTAP NFS Best Practice and Implementation Guide
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