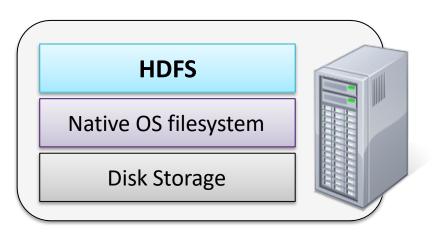
The Hadoop Distributed File System (HDFS)



HDFS: The Hadoop Distributed File System

- HDFS is a filesystem written in Java
 - Based on Google's GFS (Google File System)
- Sits on top of a native filesystem
 - Such as ext3, ext4, or xfs
- Provides redundant storage for massive amounts of data
 - Using industry-standard hardware
- At load time, data is distributed across all nodes
 - Provides for efficient processing



HDFS Features

- High performance
- Fault tolerance
- Relatively simple centralized management
 - Master-worker architecture
- Security
 - Two levels from which to choose
- Optimized for distributed processing
 - Data locality
- Scalability

HDFS Design Assumptions

Components will fail

"Modest" number of large files

- Millions of large files, not hundreds of millions of small files
- Each file is likely to be 100MB or larger
 - Multi-gigabyte files typical

Files are write-once

 Data can be appended to a file, but the file's existing contents cannot be changed

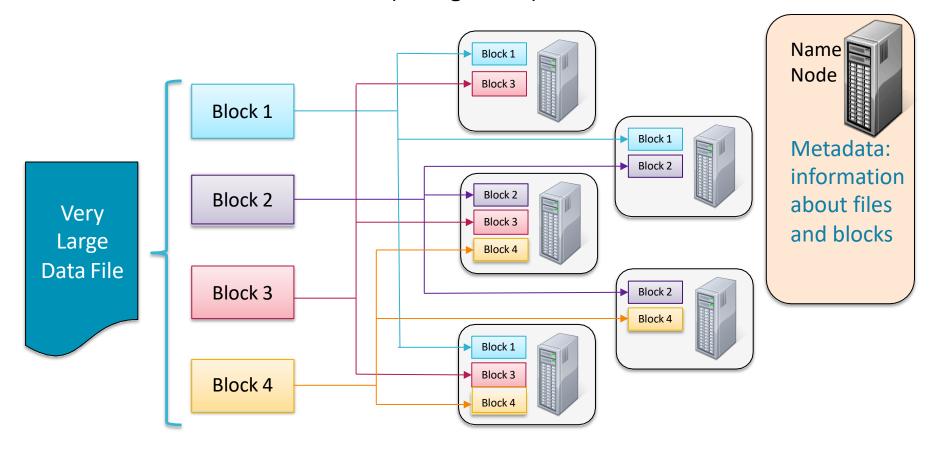
Large streaming reads

Favor high sustained throughput over low latency

HDFS Blocks

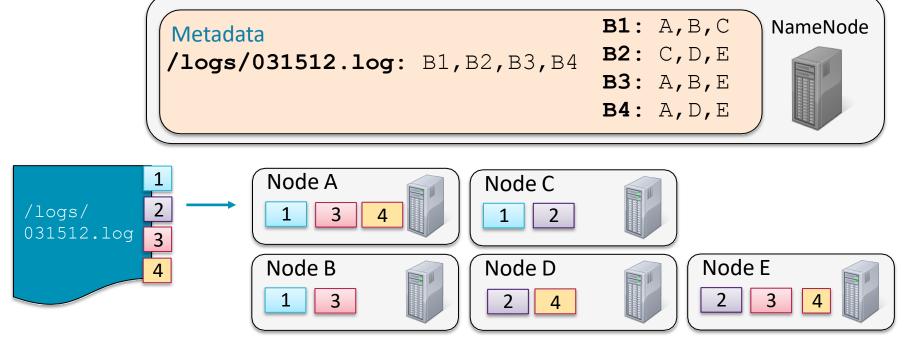
When a file is added to HDFS, it is split into blocks

- Similar concept to native filesystems, but much larger block size
- Default block size is 512MB (configurable)



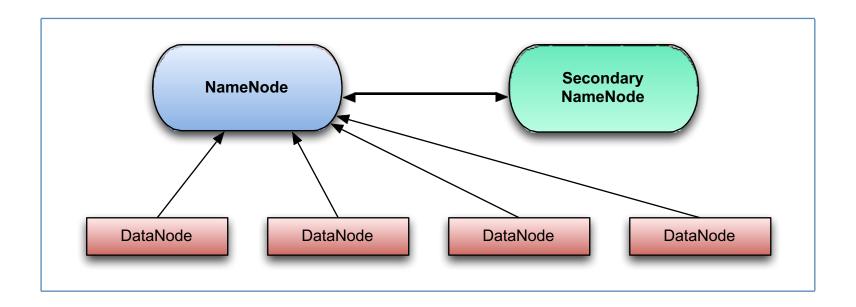
HDFS Replication

- Blocks are replicated to nodes throughout the cluster
 - Based on the *replication factor* (default is three)
- Replication increases reliability and performance
 - Reliability: data can tolerate loss of all but one replica
 - Performance: more opportunities for data locality



HDFS Without High Availability

- You can deploy HDFS with or without high availability
- Without high availability, there are three daemons
 - NameNode (master)
 - SecondaryNameNode (master)
 - DataNode (worker)



The HDFS NameNode

- The NameNode holds all metadata in RAM
 - Information about file locations in HDFS
 - Information about file ownership and permissions
 - Names of the individual blocks
 - Locations of the blocks
- Metadata is stored on disk and read when the NameNode daemon starts up
 - Filename is fsimage
- Changes to the metadata are stored in RAM
 - Changes are also written to an edits log

The Worker Nodes

- Actual contents of the files are stored as blocks on the worker nodes
- Each worker node runs a DataNode daemon
 - Controls access to the blocks
 - Communicates with the NameNode
- Blocks are simply files on the worker nodes' underlying filesystem
 - Named blk xxxxxxx
 - Nothing on the worker node provides information about what underlying file the block is a part of
 - That information is *only* stored in the NameNode's metadata
- Each block is stored on multiple different nodes for redundancy
 - Default is three replicas

The HDFS SecondaryNameNode: Caution!

The SecondaryNameNode is not a failover NameNode!

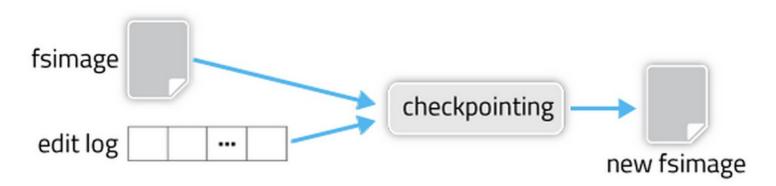
- It performs memory-intensive administrative functions for the NameNode
 - NameNode keeps information about files and blocks (the metadata) in memory
 - NameNode writes metadata changes to an edit log
 - Secondary NameNode periodically combines a prior snapshot of the file system metadata and edit log into a new snapshot
 - New snapshot is transmitted back to the NameNode
- SecondaryNameNode should run on a separate machine in a large installation
 - It requires as much RAM as the NameNode
- SecondaryNameNode only exists when high availability is not configured

File System Metadata Snapshot and Edit Log

- The fsimage file contains a file system metadata snapshot
 - It is **not** updated at every write
- HDFS write operations are recorded in the NameNode's edit log
 - The NameNode's in-memory representation of the file system metadata is also updated
- Applying all changes in the edits file(s) during a NameNode restart could take a long time
 - The files could also grow to be huge

Checkpointing the File System Metadata

- The SecondaryNameNode periodically constructs a checkpoint using this process:
 - 1. Compacts information in the edits log
 - 2. Merges it with the most recent fsimage file
 - 3. Clears the edits log
- Benefit: faster NameNode restarts
 - The NameNode can use the latest checkpoint and apply the contents of the smaller edits log

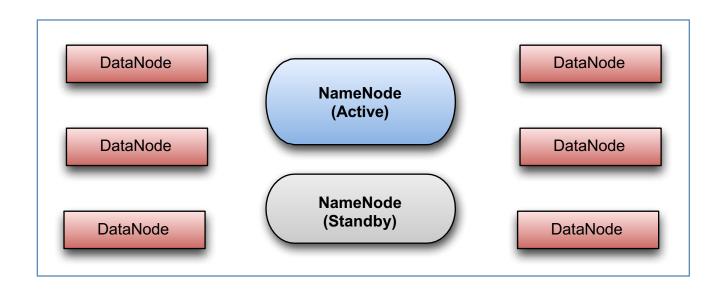


Single Point of Failure

- In this mode of operation, each Hadoop cluster has a single NameNode
 - The Secondary NameNode is *not* a failover NameNode
- The NameNode is a single point of failure (SPOF)
- In practice, this is not a major issue
 - HDFS will be unavailable until NameNode is replaced
 - There is very little risk of data loss for a properly managed system
- Recovering from a failed NameNode is relatively easy
 - We will discuss this process in detail later

HDFS With High Availability

- Deploy HDFS with high availability to eliminate the NameNode SPOF
- Two NameNodes: one active and one standby
 - Standby NameNode takes over when active NameNode fails
 - Standby NameNode also does checkpointing (SecondaryNameNode no longer needed)



HDFS Read Caching

Applications can instruct HDFS to cache blocks of a file

- Blocks are stored on the DataNode in off-heap RAM
- Can result in significant performance gains for subsequent reads
- Most useful for applications where the same file will be read multiple times

It is possible for a user to manually cache files

- Impala is caching aware
 - Tables can be cached from within the Impala shell, or set to be cached when they are created
- Caching-aware applications will attempt to read a block from the node on which it is cached

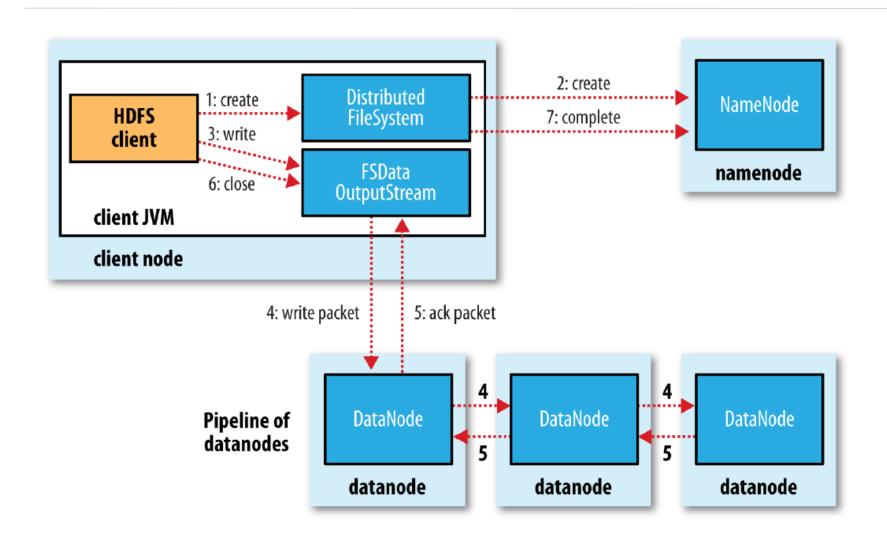
HDFS caching provides benefits over standard OS-level caching

Avoids memory-to-memory copying

Configuring HDFS Read Caching

- Cloudera Manager enables HDFS Read Caching by default
- Amount of RAM per DataNode to use for caching is controlled by dfs.datanode.max.locked.memory
 - Can be configured per role group
 - Default is 4GB
 - Set to 0 to disable caching

Anatomy of a File Write (1)



Anatomy of a File Write (2)

- Client connects to the NameNode
- 2. NameNode places an entry for the file in its metadata, returns the block name and list of DataNodes to the client
- 3. Client connects to the first DataNode and starts sending data
- 4. As data is received by the first DataNode, it connects to the second and starts sending data
- Second DataNode similarly connects to the third
- 6. ack packets from the pipeline are sent back to the client
- 7. Client reports to the NameNode when the block is written

Anatomy of a File Write (3)

If a DataNode in the pipeline fails

- The pipeline is closed
- A new pipeline is opened with the two good nodes
- The data continues to be written to the two good nodes in the pipeline
- The NameNode will realize that the block is under-replicated, and will re-replicate it to another DataNode

As blocks of data are written, the client calculates a checksum for each block

- Sent to the DataNode along with the data
- Written together with each data block
- Used to ensure the integrity of the data when it is later read

Hadoop is 'Rack-aware'

Hadoop understands the concept of 'rack awareness'

- The idea of where nodes are located, relative to one another
- Helps the ResourceManager allocate processing resources on nodes closest to the data
- Helps the NameNode determine the 'closest' block to a client during reads
- In reality, this should perhaps be described as being 'switch-aware'

HDFS replicates data blocks on nodes on different racks

- Provides extra data security in case of catastrophic hardware failure

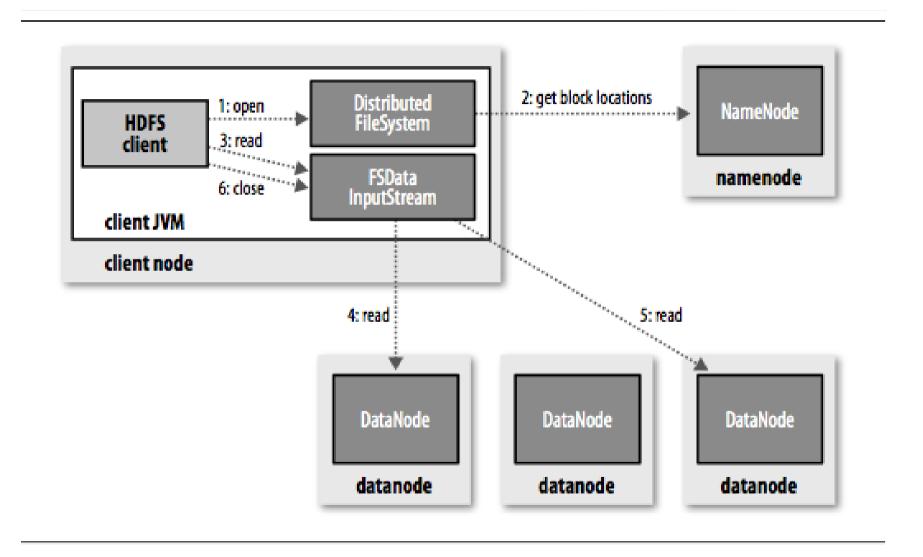
Rack-awareness is determined by a user-defined script

 Rack Topology configuration details through Cloudera Manager discussed later in this course

HDFS Block Replication Strategy

- First copy of the block is placed on the same node as the client
 - If the client is not part of the cluster, the first block is placed on a random node
 - System tries to find one which is not too busy
- Second copy of the block is placed on a node residing on a different rack
- Third copy of the block is placed on different node in the same rack as the second copy

Anatomy of a File Read (1)



Anatomy of a File Read (2)

- Client connects to the NameNode
- 2. NameNode returns the name and locations of the first few blocks of the file
 - Block locations are returned closest-first
- 3. Client connects to the first of the DataNodes, and reads the block
 - If the DataNode fails during the read, the client will seamlessly connect to the next one in the list to read the block

Dealing With Data Corruption

- As a client is reading the block, it also verifies the checksum
 - 'Live' checksum is compared to the checksum created when the block was stored
- If they differ, the client reads from the next DataNode in the list
 - The NameNode is informed that a corrupted version of the block has been found
 - The NameNode will then re-replicate that block elsewhere
- The DataNode verifies the checksums for blocks on a regular basis to avoid 'bit rot'
 - Default is every three weeks after the block was created

Data Reliability and Recovery

DataNodes send heartbeats to the NameNode

- Every three seconds

After a period without any heartbeats, a DataNode is assumed to be lost

- NameNode determines which blocks were on the lost node
- NameNode finds other DataNodes with copies of these blocks
 - These DataNodes are instructed to copy the blocks to other nodes
- Three-fold replication is actively maintained

A DataNode can rejoin a cluster after being down for a period

- The NameNode will ensure that blocks are not over-replicated by instructing DataNodes to remove excess copies
 - Note that this does not mean all blocks will be removed from the DataNode which was temporarily lost!

The NameNode Is Not a Bottleneck

- Note: the data never travels via a NameNode
 - For writes
 - For reads
 - During re-replication

NameNode: Memory Allocation (1)

- When a NameNode is running, all metadata is held in RAM for fast response
- Default Java Heap Size of the NameNode is 1GB
 - At least 1GB recommended for every million HDFS blocks
- Items stored by the NameNode:
 - Filename, permissions, etc.
 - Block information for each block

NameNode: Memory Allocation (2)

Why HDFS prefers fewer, larger files:

- Consider 1GB of data, HDFS block size 128MB
- Stored as 1 x 1GB file
 - Name: 1 item
 - Blocks: 8 items
 - Total items in memory: 9
- Stored as 1000 x 1MB files
 - Names: 1000 items
 - Blocks: 1000 items
 - Total items in memory: 2000

HDFS File Permissions

- Files in HDFS have an owner, a group, and permissions
 - Very similar to Unix file permissions
- File permissions are read (x), write (w), and execute (x) for each of owner, group, and other
 - x is ignored for files
 - For directories, x means that its children can be accessed
- HDFS permissions are designed to stop good people doing foolish things
 - Not to stop bad people doing bad things!
 - HDFS believes you are who you tell it you are

Hadoop Security Overview

Authentication

- Proving that a user or system is who he or she claims to be
- Hadoop can provide strong authentication control via Kerberos
 - Cloudera Manager simplifies Kerberos deployment
- Authentication via LDAP is available with Cloudera Enterprise

Authorization (access control)

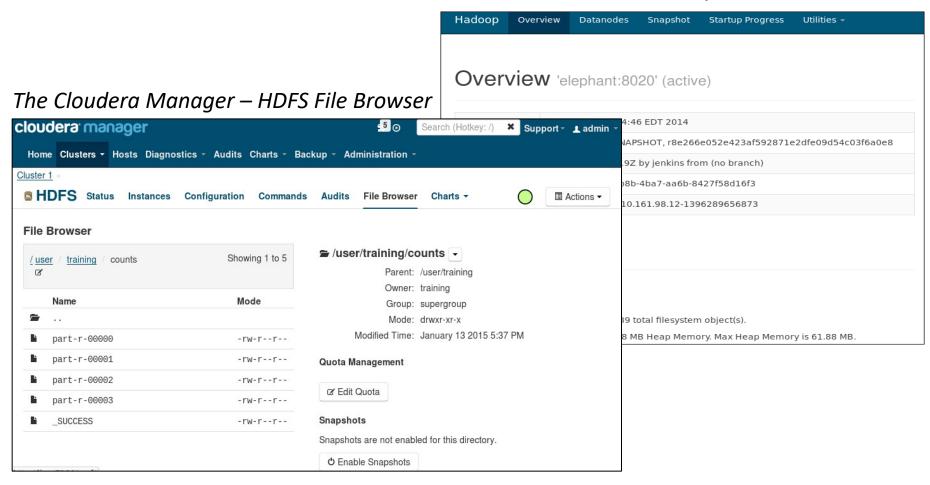
- Allowing people or systems to do some things but not other things
- CDH has traditional POSIX-style permissions for files and directories
- Access Control Lists (ACLs) for HDFS
- Role-based access control provided with Apache Sentry

Data encryption

- OS filesystem-level, HDFS-level, and Network-level options
- We will cover Hadoop security in more depth later

Web UIs for HDFS

The NameNode Web UI on port 50070



Options for Accessing HDFS

From the command line

- FsShell: hdfs dfs

From Cloudera Manager

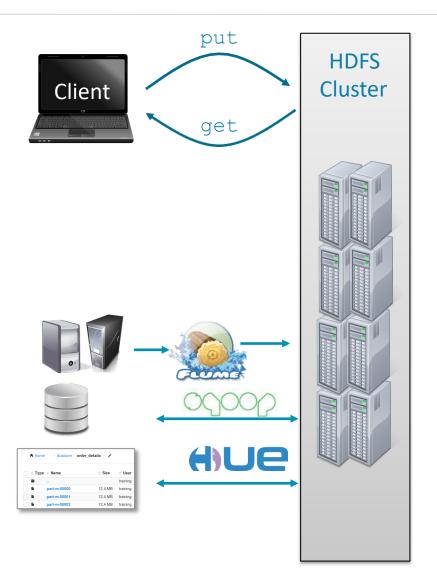
- HDFS page, File Browser tab
hdfs://host:port/file...

From the NameNode Web UI

- Utilities > Browse the file system

Other programs

- Java API
 - Used by MapReduce,
 Spark, Impala, Hue,
 Sqoop, Flume, etc.
- RESTful interface



Accessing HDFS via the Command Line

- HDFS is not a general purpose filesystem
 - Not built into the OS, so only specialized tools can access it
- End users typically access HDFS via the hdfs dfs command
 - Actions are specified with subcommands (prefixed with a minus sign)
 - Most subcommands are similar to corresponding UNIX commands
- Display the contents of the /user/fred/sales.txt file

```
$ hdfs dfs -cat /user/fred/sales.txt
```

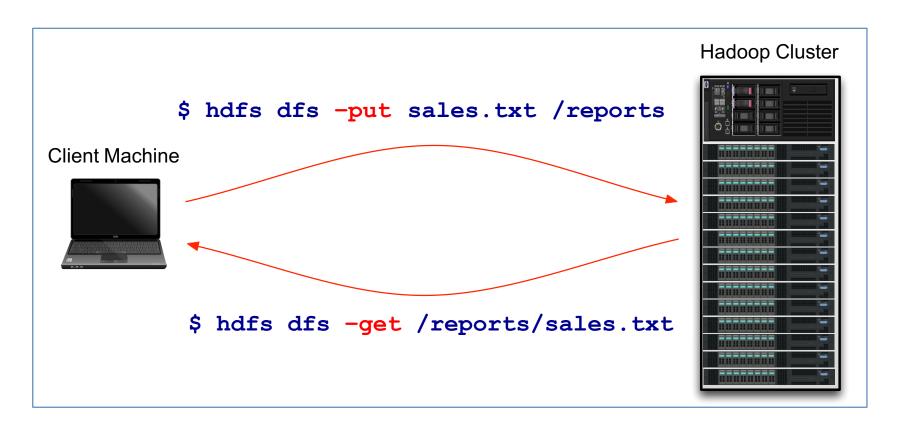
Create a directory (below the root) called reports

```
$ hdfs dfs -mkdir /reports
```

Copying Local Data To and From HDFS Using the Command Line

Remember that HDFS is distinct from your local filesystem

- The hdfs dfs -put command copies local files to HDFS
- The hdfs dfs -get fetches a local copy of a file from HDFS



More hdfs dfs Command Examples

Copy file input.txt from local disk to the user's directory in HDFS

```
$ hdfs dfs -put input.txt input.txt
```

- This will copy the file to /user/username/input.txt
- Get a directory listing of the HDFS root directory

```
$ hdfs dfs -ls /
```

Delete the file /reports/sales.txt

```
$ hdfs dfs -rm /reports/sales.txt
```

Hands-On Exercise: Exploring HDFS