

Overview of HDFS Architecture

- Able to explain HDFS design goals
- Factors impacting design
- Design approach
- Explain basic HDFS architecture

HDFS Design Concept

- Scalable distributed filesystem
- Distribute data on local disks on several nodes
- Low cost commodity hardware



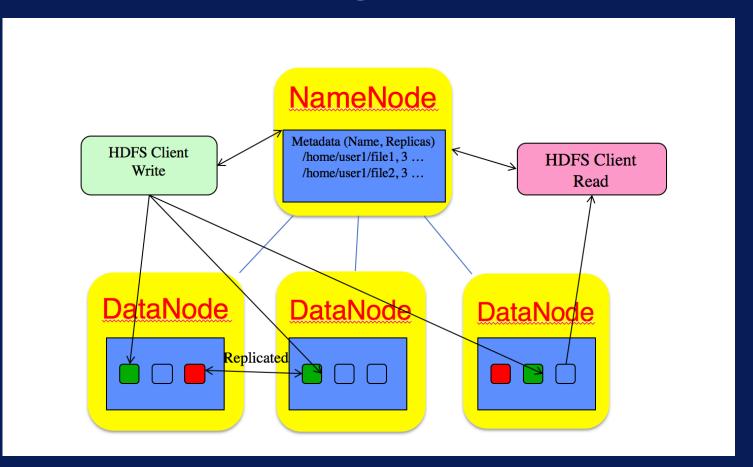
HDFS Design Factors

- Hundreds/Thousands of nodes =>
 - Need to handle node/disk failures
- Portability across heterogeneous hardware/software
- Handle large data sets
- High throughput

Approach to meet HDFS design goals

- Simplified coherency model write once read many.
- Data Replication helps handle hardware failures
- Move computation close to data
- Relax POSIX requirements increase throughput

HDFS Architecture



Summary of HDFS Architecture

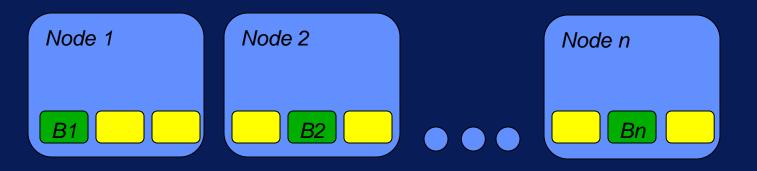
- Single NameNode a master server that manages the file system namespace and regulates access to files by clients.
- Multiple DataNodes typically one per node in the cluster. Functions:
 - Manage storage
 - Serving read/write requests from clients
 - Block creation, deletion, replication based on instructions from NameNode

Performance Envelope of HDFS

- Able to determine number of blocks for a given file size
- Key HDFS and system components impacted by block size
- Impact of small files on HDFS and system

Recall: HDFS Architecture

 Distribute data on local disks on several nodes.



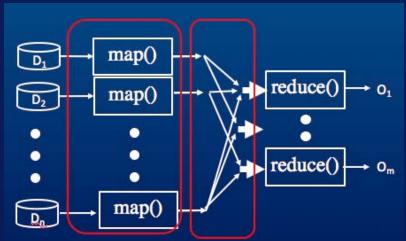
HDFS block size

- Default block size is 64MB
- Good for large files!
- So a 10GB file will be broken into: 10x1024/64 = 160 blocks.



Importance of #blocks in a file

- NameNode memory usage: Every block represented as object (default replication this will be further increased 3X)
- Number of map tasks: data typically processed block at a time



Large #small files: Impact on NameNode

Memory usage ~ 150 bytes per object
1 billion objects => 300GB memory!

• Network load – Number of checks with datanodes proportional to number of blocks.

Large #small files: Performance Impact

- Map tasks depends on #blocks 10GB of data, 32k file size => 327680 map tasks
 - ⇒lots of queued tasks
 - ⇒large overhead of spin up/tear down for each task
 - ⇒Inefficient disk I/O with small sizes

HDFS optimized for large files

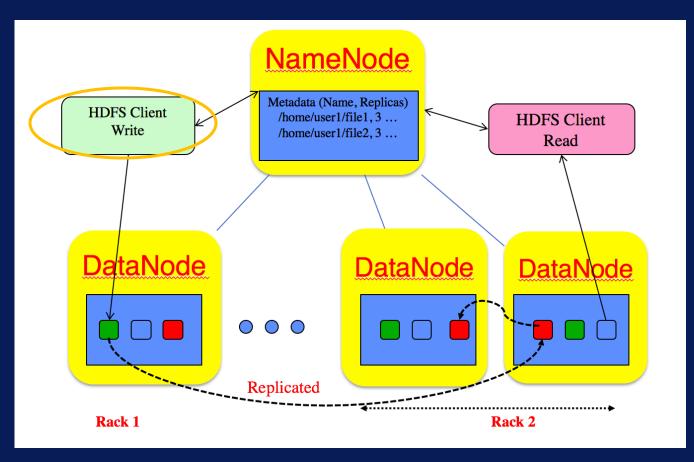
 Key takeaway – lots of small files is bad!

Solutions:

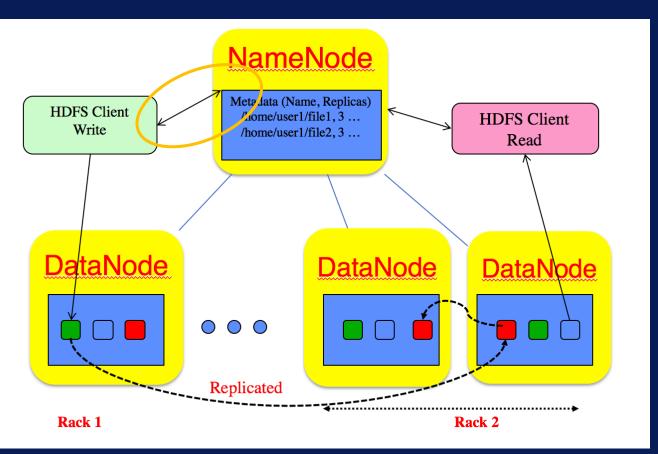
- Merge/Concatenate files
- Sequence files
- HBase, HIVE configuration
- CombineFileInputFormat

Write/Read processes on HDFS

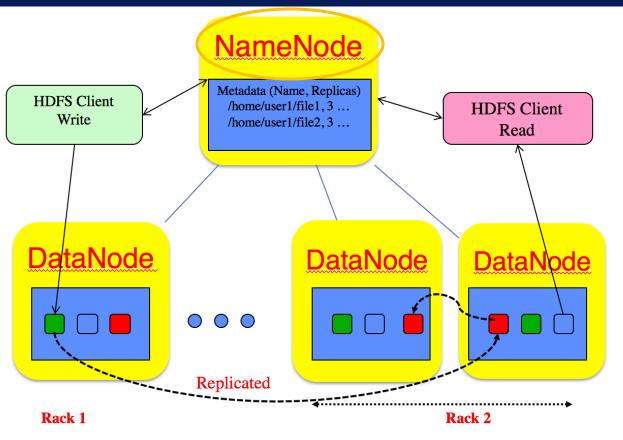
- Able to explain write process in HDFS
- Detail the replication pipeline process
- Explain read process in HDFS



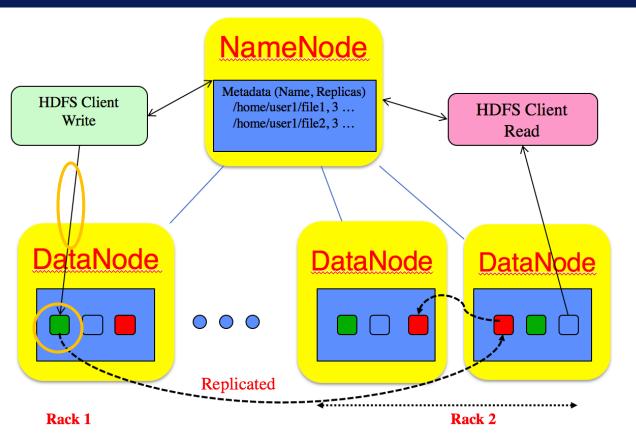
Client request to create file



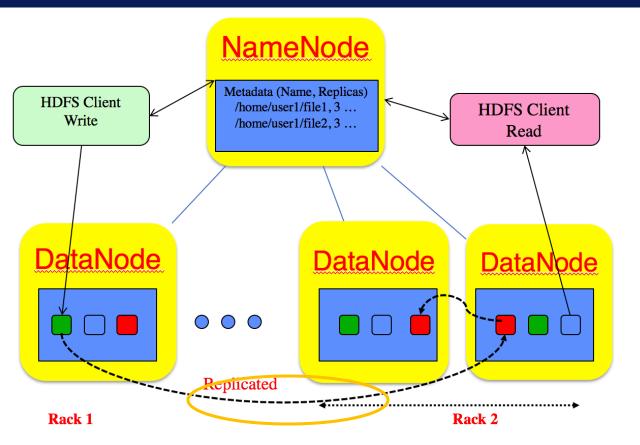
NameNode contacted once a block of data is accumulated



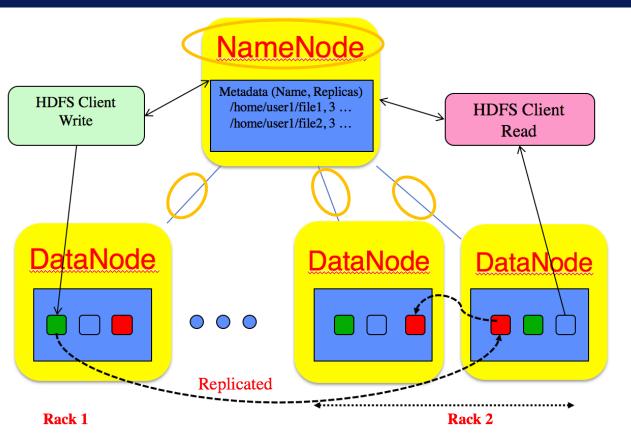
- NameNode responds with list of DataNodes
- Rack aware



First DataNode receives data, writes to local and forwards to second DataNode

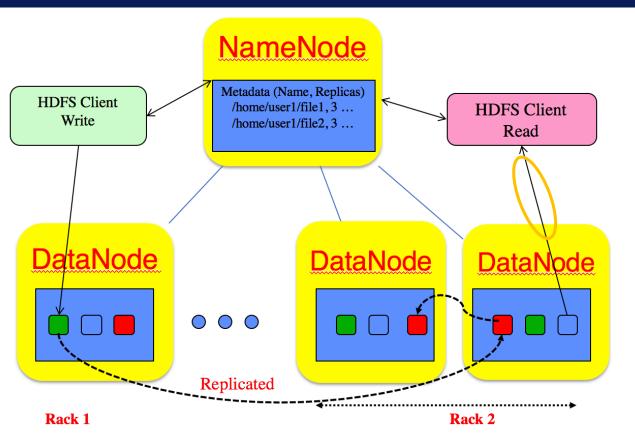


First DataNode
 receives data,
 writes to local
 and forwards
 to second
 DataNode



- NameNode commits file creation into persistent store.
- Receives
 heartbeat and
 block reports

Read Process in HDFS



- Client getsDataNode listfromNameNode
- Read from replica closest to reader.