



# Understanding & Planning Hadoop Cluster

Cloudwick  
Technologies

# Agenda

Typical Workflow

Writing Files to HDFS

Reading files from HDFS

Rack Awareness

Planning for Cluster

Choosing Right Hardware

Choosing Right Hadoop Distribution

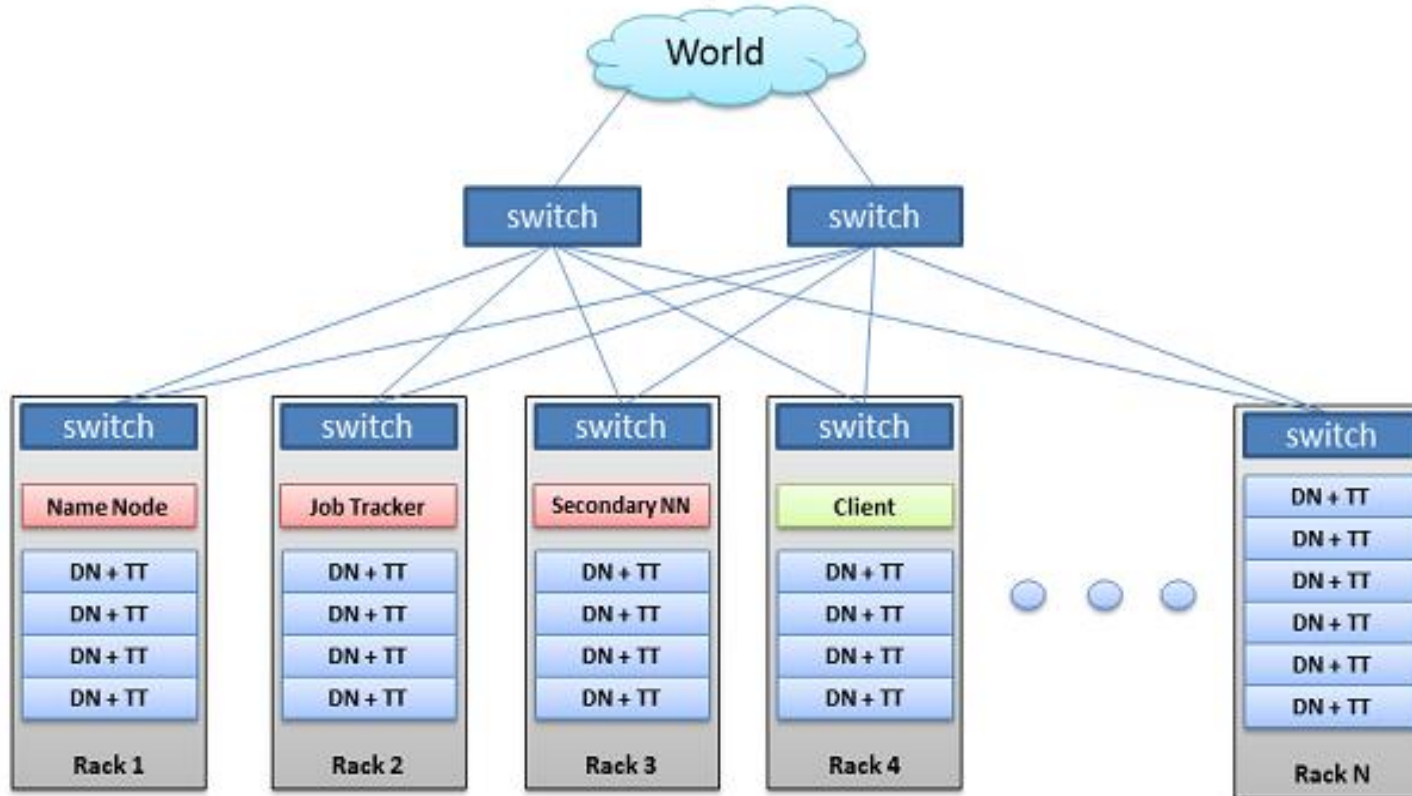


## 3 Major Categories

- # Hadoop Server Roles
- 
- ```
graph TD; Clients[Clients] --> DDP[Distributed Data Processing]; Clients --> DDS[Distributed Data Storage]; DDP --> MR[Map Reduce]; MR --> JT[Job Tracker]; JT -.-> DNTT1[Data Node & Task Tracker]; DNTT1 -.-> DNTT2[Data Node & Task Tracker]; DNTT2 -.-> DNTT3[Data Node & Task Tracker]; DDS --> HDFS[HDFS]; HDFS --> NN[Name Node]; HDFS --> SNN[Secondary Name Node]; NN -.-> DNTT1; NN -.-> DNTT2; NN -.-> DNTT3; SNN -.-> DNTT4[Data Node & Task Tracker]; SNN -.-> DNTT5[Data Node & Task Tracker]; SNTT[Slaves] --- DNTT1; SNTT --- DNTT2; SNTT --- DNTT3; SNTT --- DNTT4; SNTT --- DNTT5;
```
- The diagram illustrates the Hadoop Server Roles, organized into two main functional areas: **Distributed Data Processing** and **Distributed Data Storage**.
- Clients** (represented by a green box) interact with both processing and storage components.
- Distributed Data Processing (Map Reduce):**
- The **Job Tracker** (red box) is the central component for processing.
  - It manages **Data Node & Task Tracker** (blue boxes) which are part of the **slaves** group.
- Distributed Data Storage (HDFS):**
- The **Name Node** (red box) and **Secondary Name Node** (red box) are the central components for storage.
  - They manage **Data Node & Task Tracker** (blue boxes) which are part of the **slaves** group.
- Slaves:** The **Data Node & Task Tracker** (blue boxes) are the nodes that execute tasks and store data. They are grouped under the **slaves** label.

# Understanding Hadoop Cluster

## Hadoop Cluster



This is the typical architecture of a Hadoop cluster



# Understanding Hadoop Cluster

## Typical Workflow

- Load data into the cluster (HDFS writes)
- Analyze the data (Map Reduce)
- Store results in the cluster (HDFS writes)
- Read the results from the cluster (HDFS reads)

Sample Scenario:

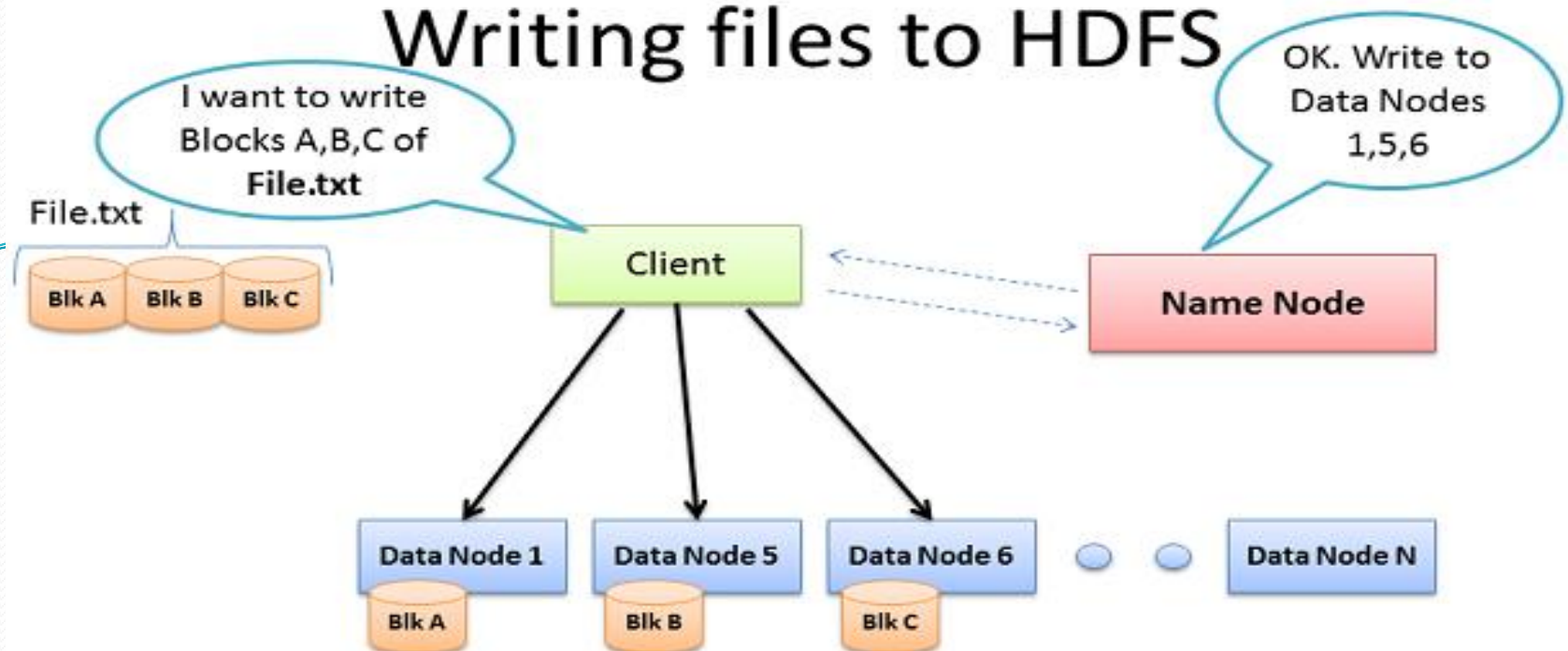
How many times did our customers type the word **“Refund”** into emails sent to customer service?

Huge file containing all emails sent to customer service



# Understanding Hadoop Cluster

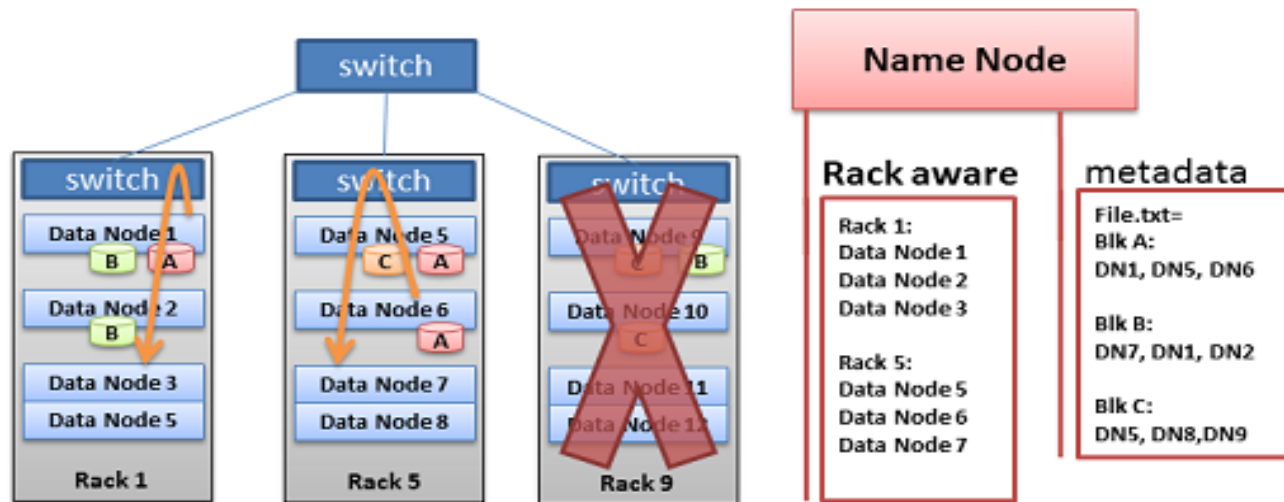
## Writing files to HDFS



- Client consults Name Node
- Client writes block directly to one Data Node
- Data Nodes replicates block
- Cycle repeats for next block

# Understanding Hadoop Cluster

## Hadoop Rack Awareness – Why?

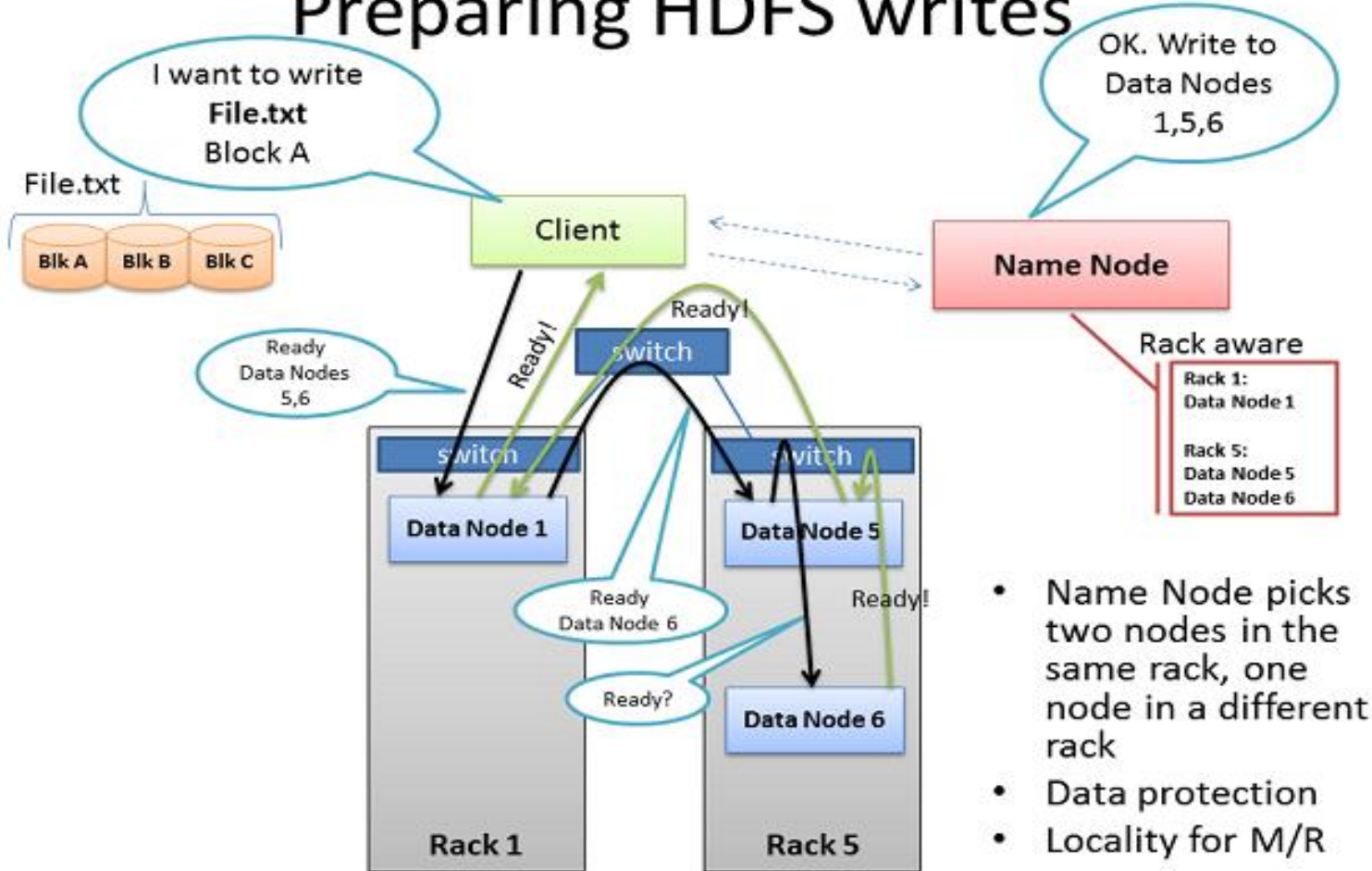


- Never loose all data if entire rack fails
- Keep bulky flows in-rack when possible
- Assumption that in-rack is higher bandwidth, lower latency



# Understanding Hadoop Cluster

## Preparing HDFS writes

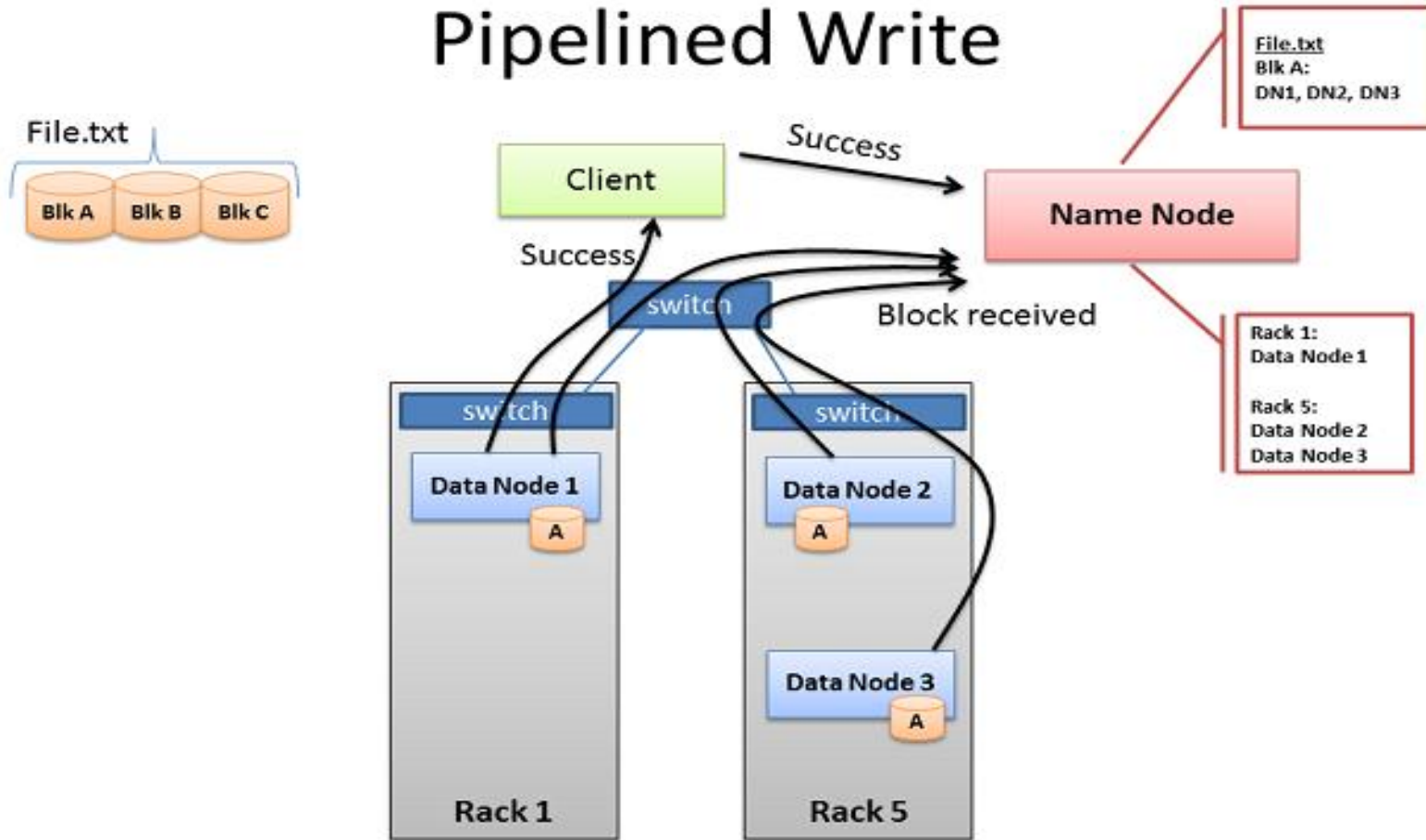


- Name Node picks two nodes in the same rack, one node in a different rack
- Data protection
- Locality for M/R



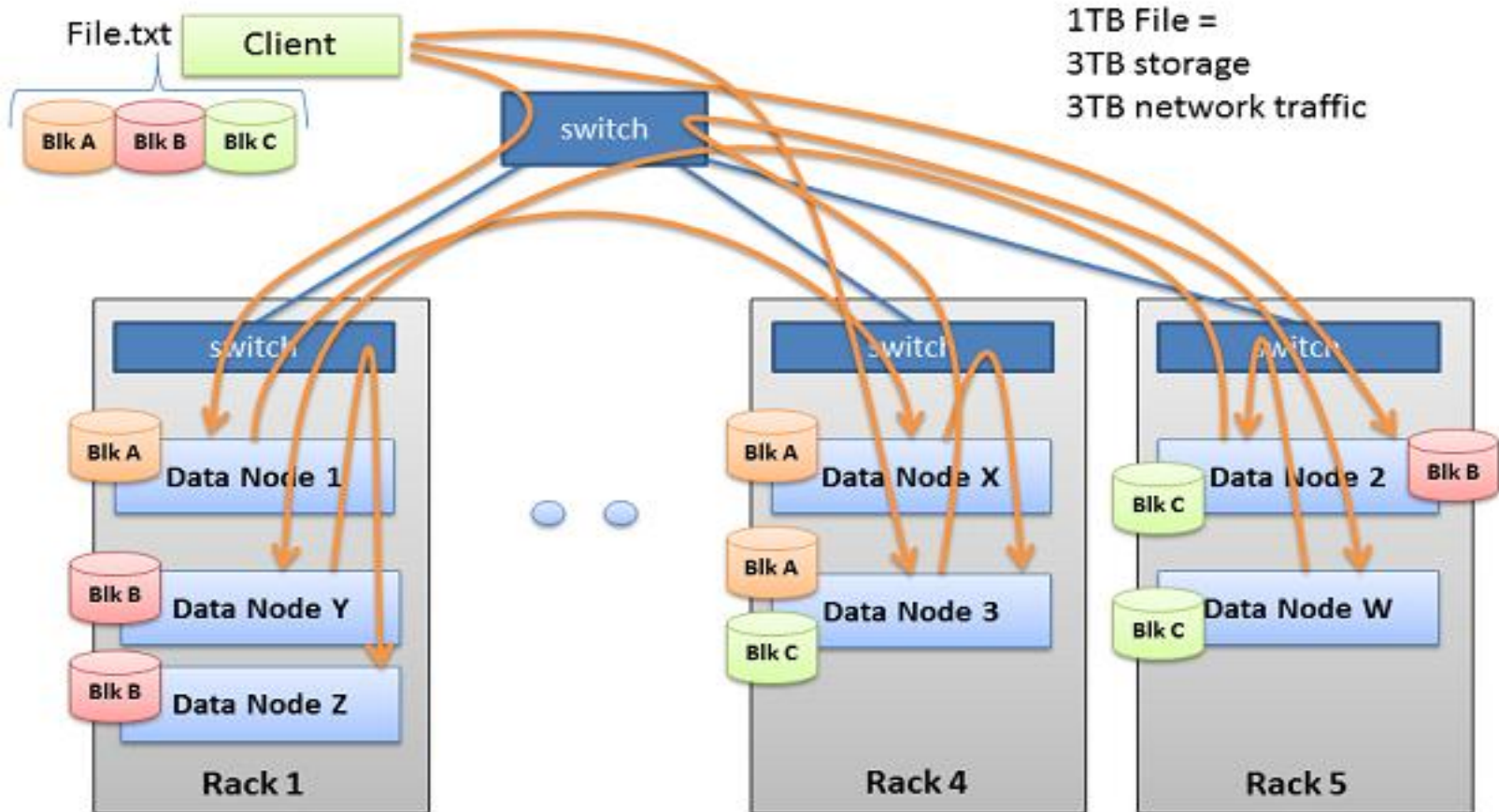
# Understanding Hadoop Cluster

## Pipelined Write



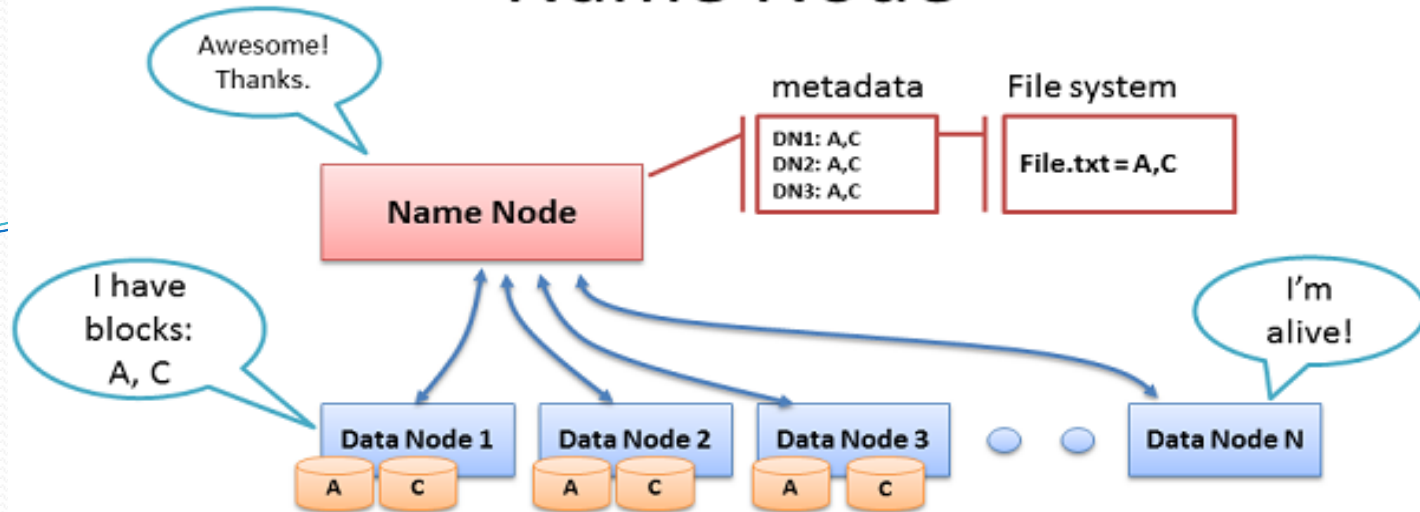
# Understanding Hadoop Cluster

## Multi-block Replication Pipeline



# Understanding Hadoop Cluster

## Name Node

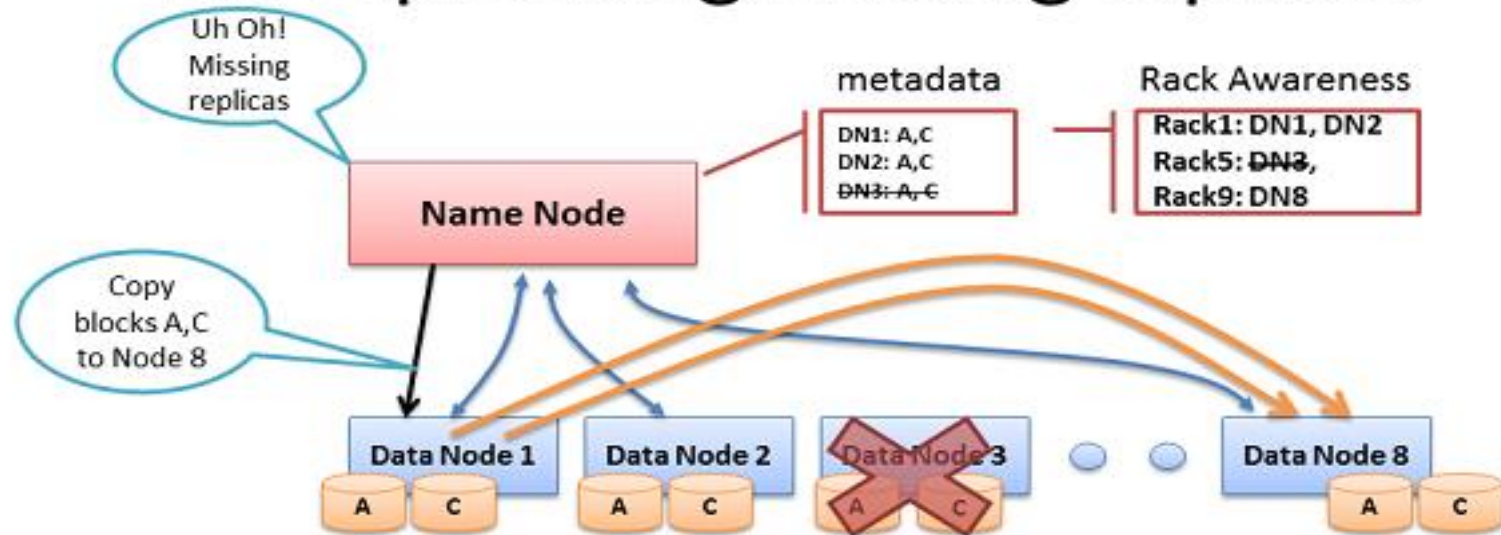


- Data Node sends Heartbeats
- Every 10<sup>th</sup> heartbeat is a Block report
- Name Node builds metadata from Block reports
- TCP – every 3 seconds
- If Name Node is down, HDFS is down



# Understanding Hadoop Cluster

## Re-replicating missing replicas

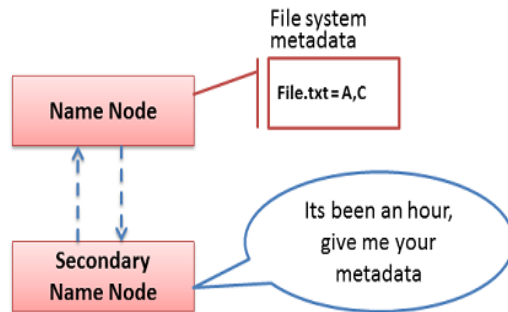


- Missing Heartbeats signify lost Nodes
- Name Node consults metadata, finds affected data
- Name Node consults Rack Awareness script
- Name Node tells a Data Node to re-replicate

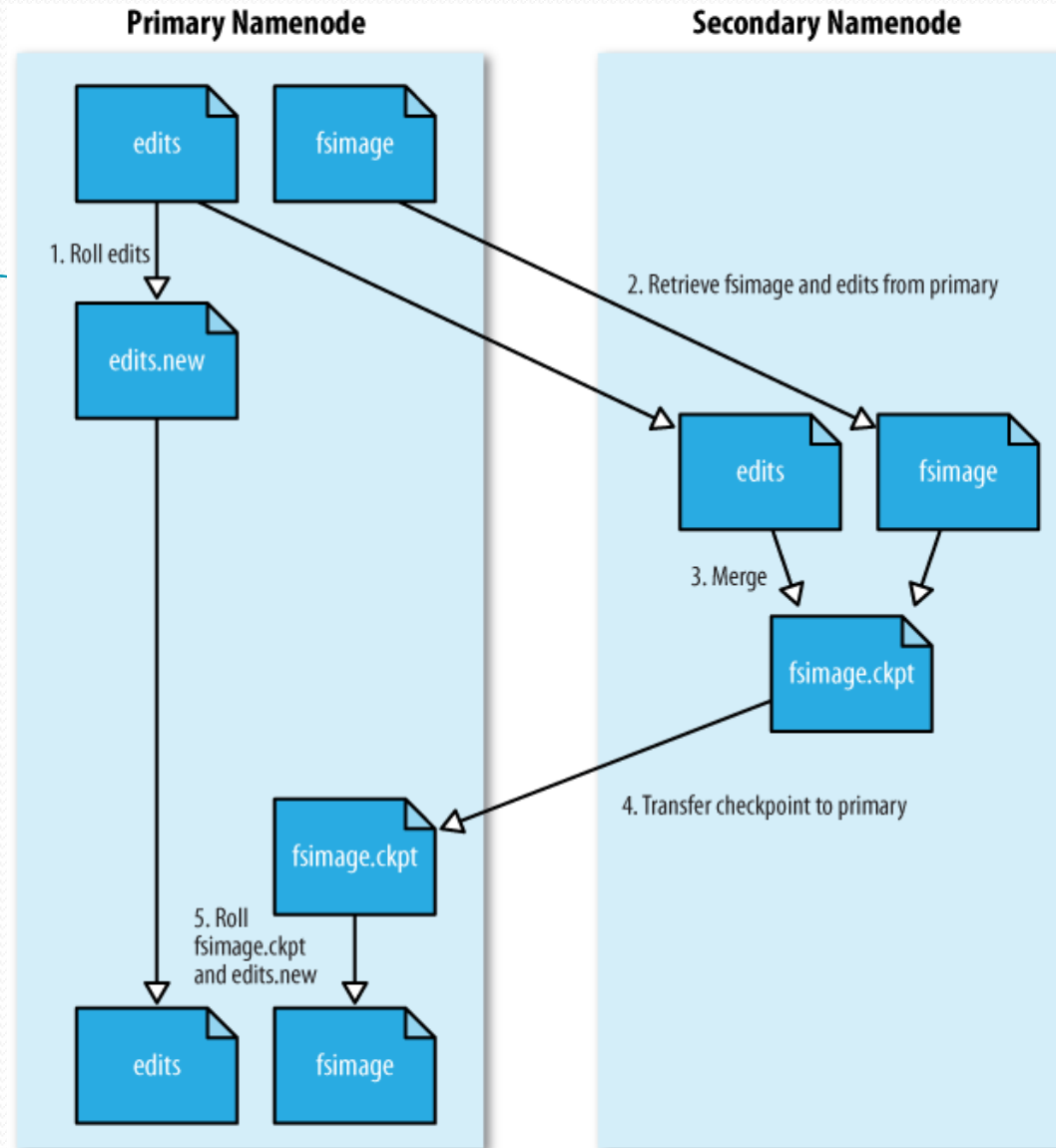


# Understanding Hadoop Cluster

## Secondary Name Node

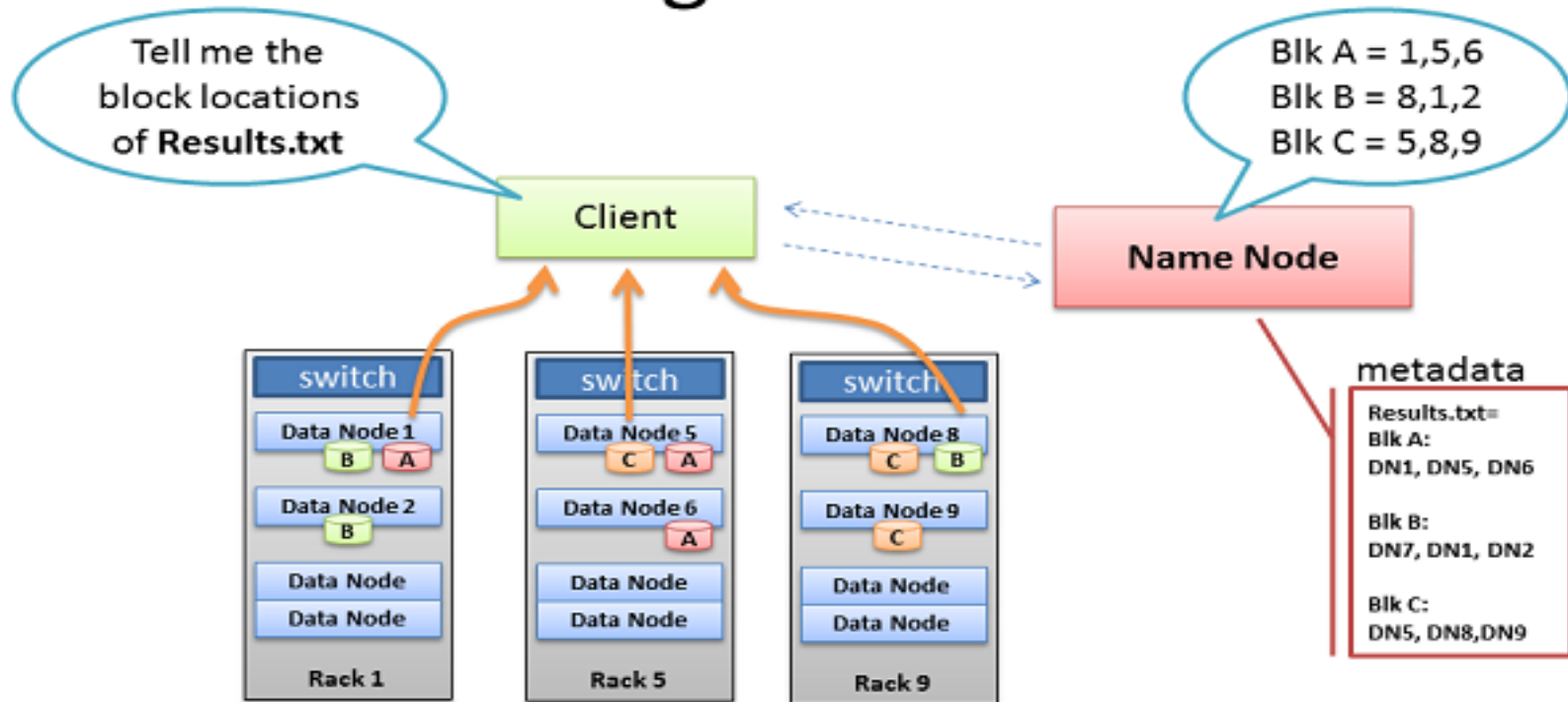


- Not a hot standby for the Name Node
- Connects to Name Node every hour\*
- Housekeeping, backup of Name Node metadata
- Saved metadata can rebuild a failed Name Node



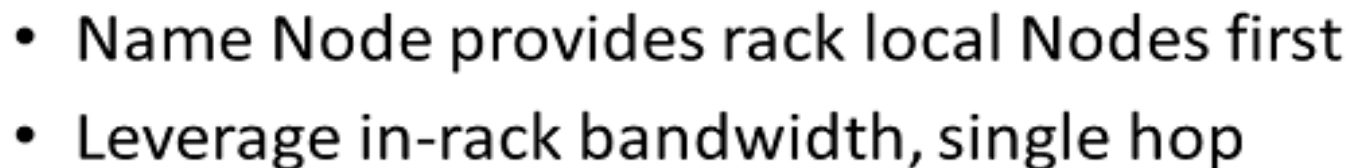
# Understanding Hadoop Cluster

## Client reading files from HDFS



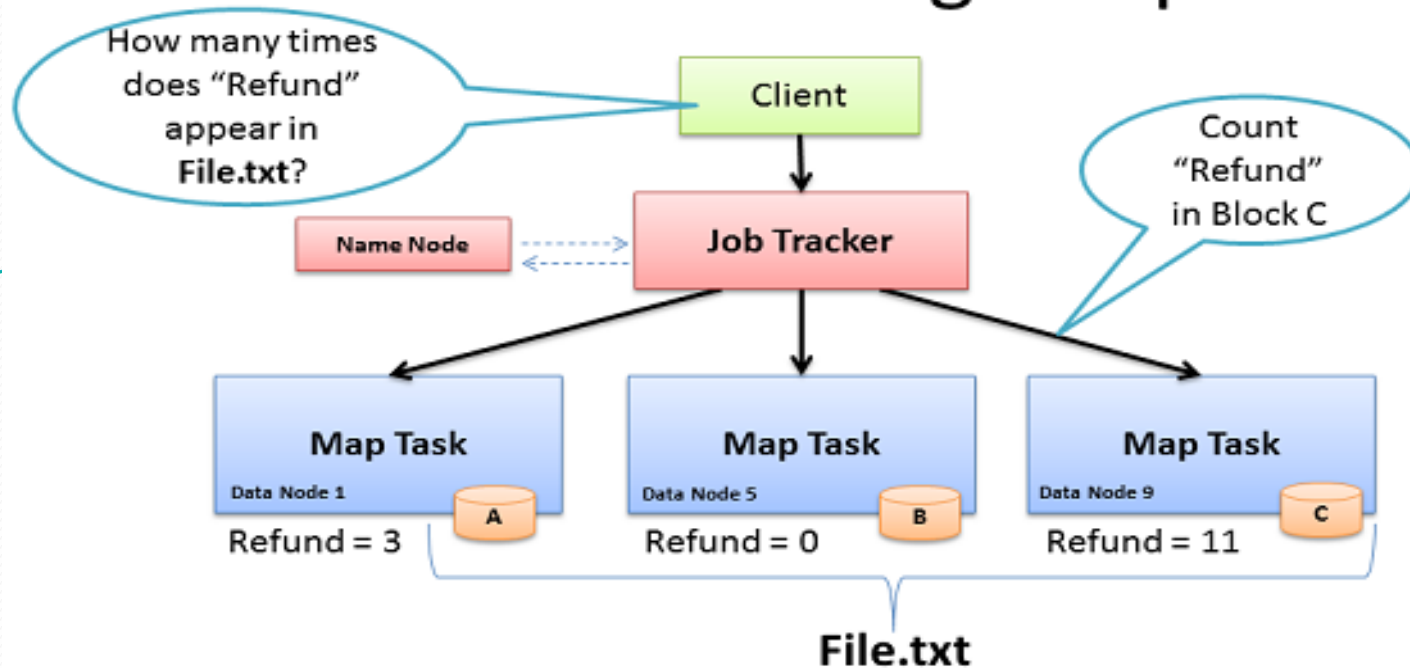
- Client receives Data Node list for each block
- Client picks first Data Node for each block
- Client reads blocks sequentially

## Data Node reading files from HDFS



# Understanding Hadoop Cluster

## Data Processing: Map

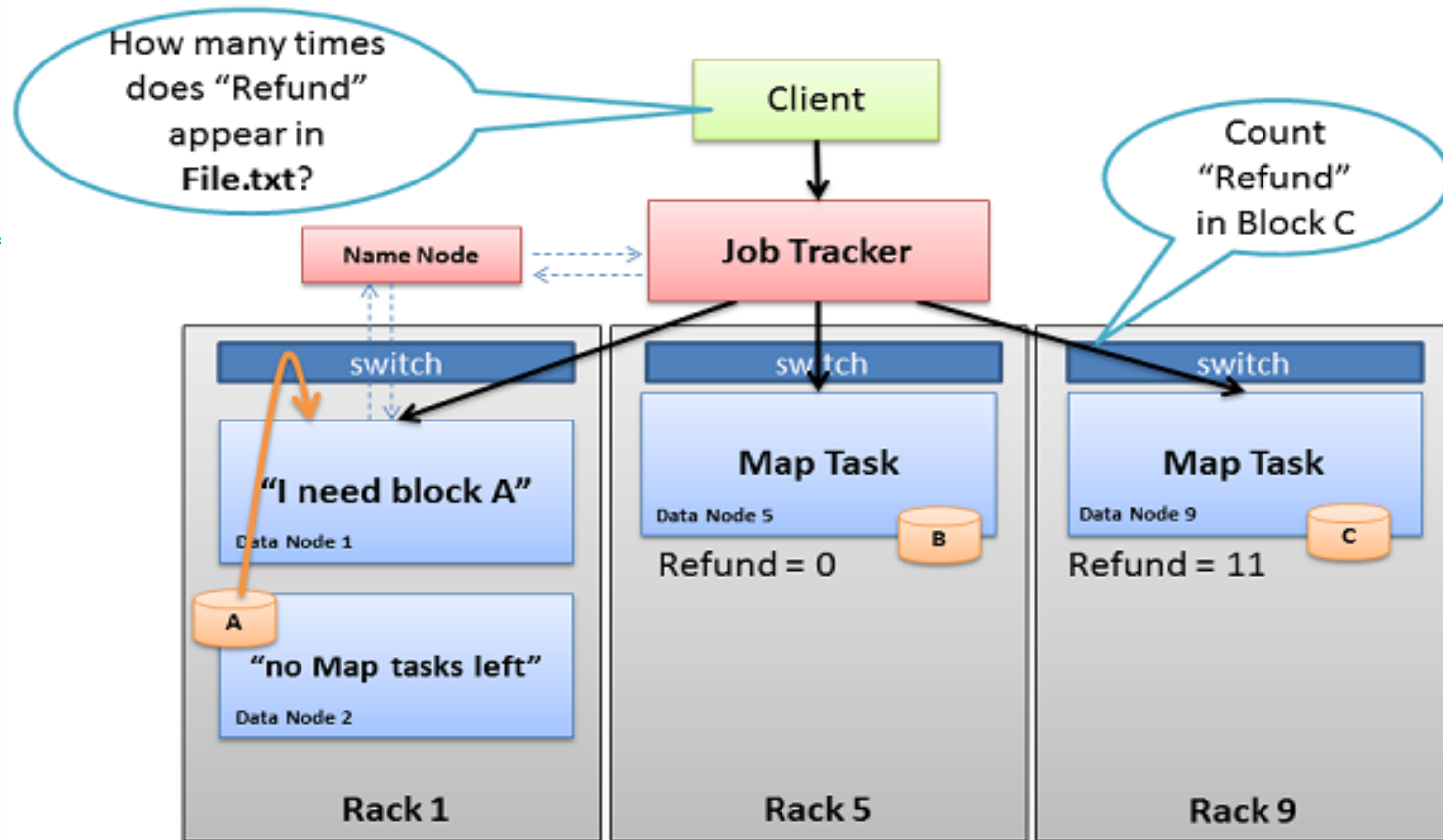


- **Map:** "Run this computation on your local data"
- Job Tracker delivers Java code to Nodes with local data



# Understanding Hadoop Cluster

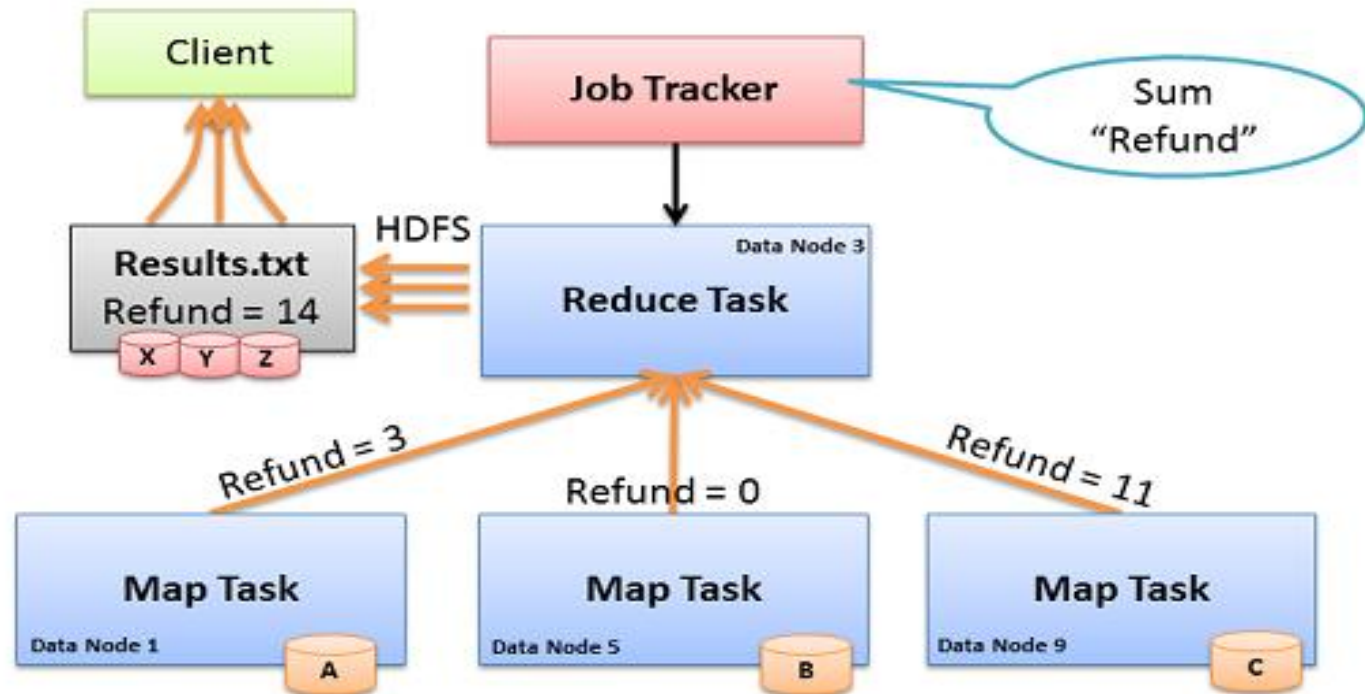
## What if data isn't local?



- Job Tracker tries to select Node in same rack as data
- Name Node rack awareness

# Understanding Hadoop Cluster

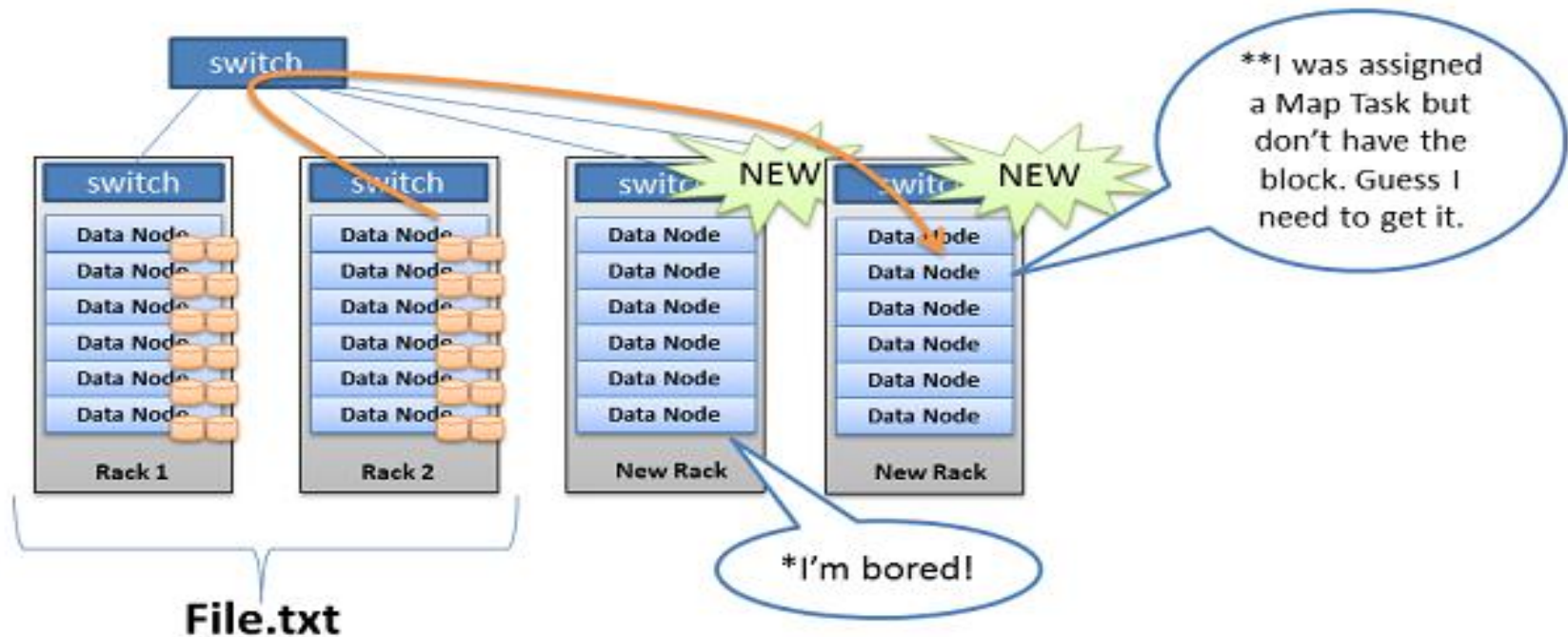
## Data Processing: Reduce



- **Reduce:** “Run this computation across Map results”
- Map Tasks send output data to Reducer over the network
- Reduce Task data output written to and read from HDFS

# Understanding Hadoop Cluster

## Unbalanced Cluster

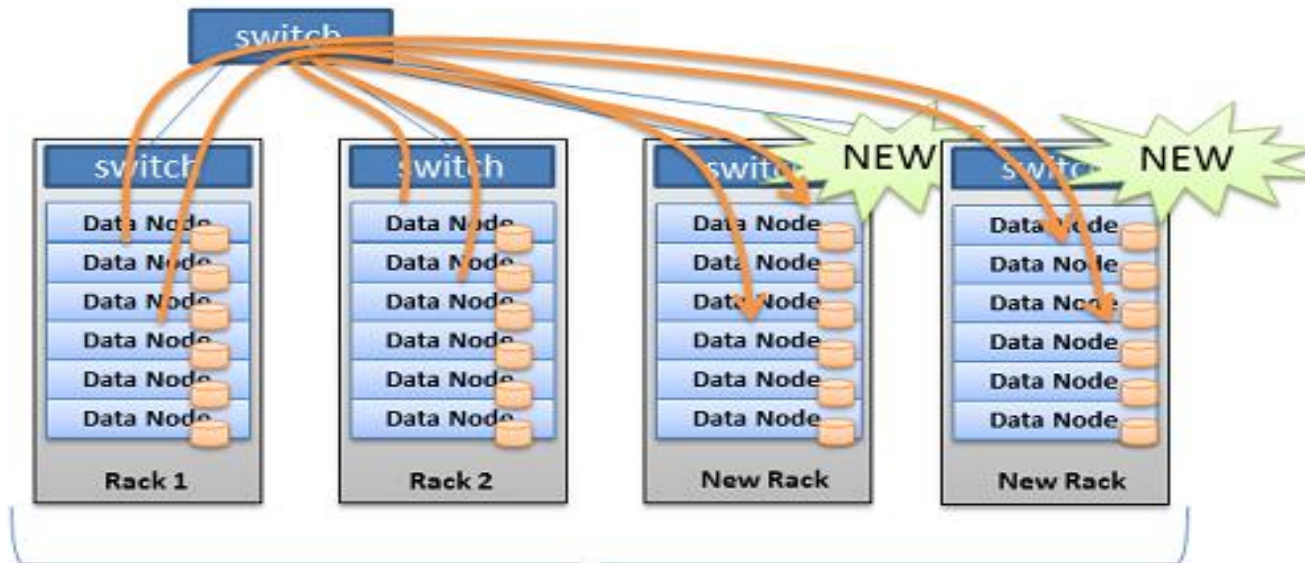


- Hadoop prefers local processing if possible
- New servers underutilized for Map Reduce, HDFS\*
- More network bandwidth, slower job times\*\*



# Understanding Hadoop Cluster

## Cluster Balancing



**File.txt**

`brad@cloudera-1:~$hadoop balancer`

- Balancer utility (if used) runs in the background
- Does not interfere with Map Reduce or HDFS
- Default rate limit 1 MB/s



# General Planning Considerations



Handle  
More  
Data



At  
Lower  
Cost



In  
Less  
Time



With  
Less  
Power

# Best Practices

- Start with small cluster ( 4 to 10 nodes) and grow as and when required.

Cluster can be grown whenever there is a

- ✓ Increase in computation power needed
- ✓ Increase in data to be stored
- ✓ Increase in amount of memory to process tasks
- ✓ Increase in data transfer between data nodes

Cluster Growth based on Storage Capacity:

| Data Growth<br>TB/Week | Replication<br>Factor | Intermediate<br>& Log Files | Overall Space needed<br>per week |
|------------------------|-----------------------|-----------------------------|----------------------------------|
| 2                      | 3                     | 30%                         | 7.8                              |

Two Machines with 1X4TB are needed.

# Where to Optimize?

## Hardware



## Software



# Choosing Right Hardware

## Master Node:

- Single Point of Failure
- 32 GB RAM
- Dual Xeon E5600 or better (Quad core)
- Dual Power supply for Redundancy
- 4 x 500 GB 7200 rpm SATA drives
- Dual 1 Gb Ethernet cards

## Data Nodes:

- 4 1TB hard disks in a JBOD (Just a Bunch Of Disks) configuration. No RAID.
- 2 quad core CPUs, running at least 2-2.5GHz
- 16-24GBs of RAM (24-32GBs if you're considering HBase)
- Gigabit Ethernet

## Master Node:

- No Commodity Hardware
- RAIDed hard drives
- Backup Metadata to an NFS Mount
- RAM Thumb rule: 1 GB per 1 million blocks of data. 32GB for 100 nodes.
- If Metadata is lost, whole cluster is lost. Use expensive Name Node.

## # of Tasks per Core:

2 Cores - Datanode and Tasktracker  
Thumb Rule – 1 Core can run 1.5 Mappers or Reducers

## Amount of RAM:

Thumb Rule: 1G per Map or Red task  
RAM for Hbase Region Server:  $0.01 \times \text{<dataset size> / <number of slaves>}$



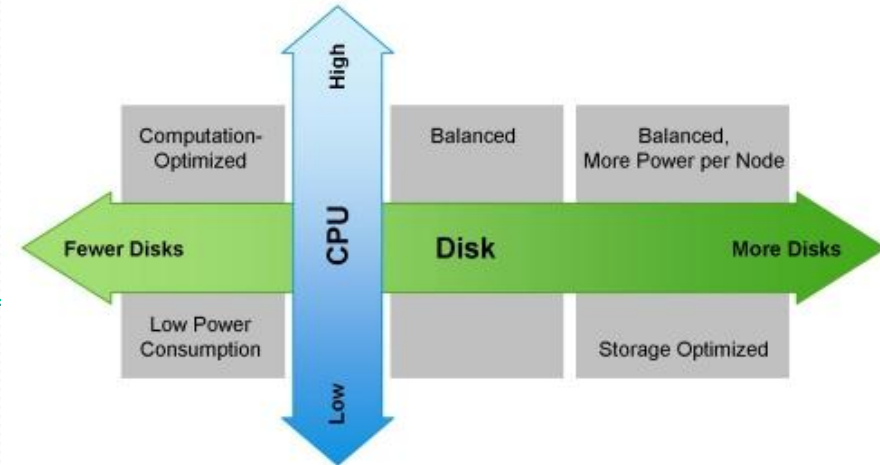
# Choosing Right Hardware based on different workloads

## Light Processing Configuration

**(1U/machine):** Two quad core CPUs, 8GB memory, and 4 disk drives (1TB or 2TB). Note that CPU-intensive work such as natural language processing involves loading large models into RAM before processing data and should be configured with 2GB RAM/core instead of 1GB RAM/core.

## Balanced Compute Configuration

**(1U/machine):** Two quad core CPUs, 16 to 24GB memory, and 4 disk drives (1TB or 2TB) directly attached using the motherboard controller. These are often available as twins with two motherboards and 8 drives in a single 2U cabinet.



## Storage Heavy Configuration

**(2U/machine):** Two quad core CPUs, 16 to 24GB memory, and 12 disk drives (1TB or 2TB). The power consumption for this type of machine starts around ~200W in idle state and can go as high as ~350W when active.

## Compute Intensive Configuration

**(2U/machine):** Two quad core CPUs, 48-72GB memory, and 8 disk drives (1TB or 2TB). These are often used when a combination of large in-memory models and heavy reference data caching is required.

# Choosing Right Software

- Using Linux distribution based on Kernel version 2.6.30 or later recommended
- Java 6u14 is or later is recommended
- Default Linux open file descriptor limit is set to 1024 which is too low for Hadoop daemons. This should be increased to 64000 using `/etc/security/limits.conf`
- If Linux Kernel 2.6.28 is used, default open ePoll file descriptor limit is 128 which is too low for Hadoop. This should be increased to 4096 in `/etc/sysctl.conf`
- Adopt a Packaged Hadoop Distribution to Reduce Technical Risk and Increase the Speed of Implementation
- Be Selective About Which Hadoop Projects to Implement
- Use Hadoop in the Cloud for Proof of Concept



# Questions?

