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Big Data Analytics
Lecture 2

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* Some slides are adopted from Professor Kunpeng Zhang's Big Data course @UMD

What we'll cover…

- Hadoop history and advantages

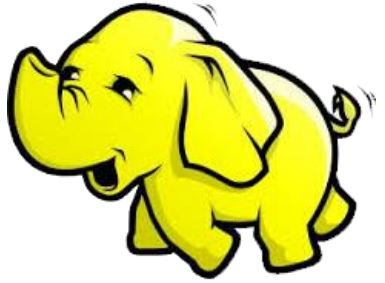
- Architecture
  - Hadoop distributed file system (HDFS)

- Hands-on **Preparation for Cloudera environment**
  - Installation of GitHub – Lab 1
  - Basic Linux commands – Lab 2

Brief History of Hadoop

Designed to answer the question:

"How to process big data with reasonable cost and time?"



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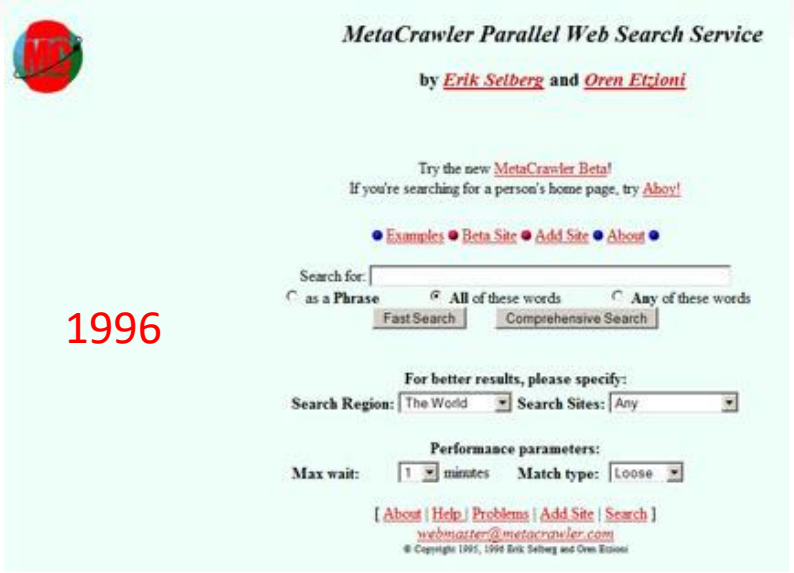
Search engines in 1990s



1996

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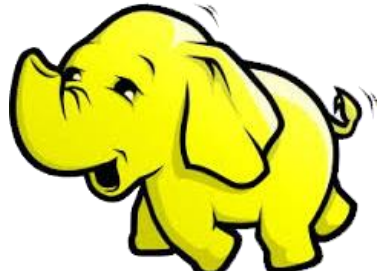

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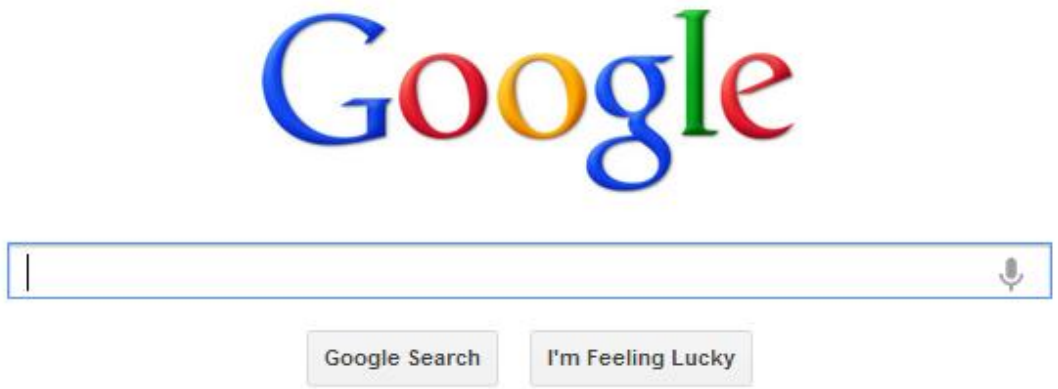
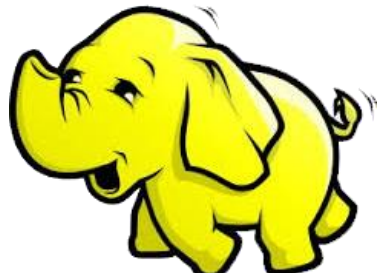
Google search engines

1998



2013



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Early Large Scale Computing

- Historically computation was processor-bound
  - Data volume has been relatively small
  - Complicated computations are performed on that data
- Advances in computer technology has historically centered around improving the power of a single machine

Distributed System: Problems

"You know you have a distributed system when the crash of a computer you've never
heard of stops you from getting any work done." –Leslie Lamport

- Distributed systems must be designed with the expectation of failure

Component Recovery

- If a component fails, it should be able to recover without restarting the entire system

- Component failure or recovery during a job must not affect the final output

Hadoop

- Based on work done by Google in the early 2000s
  - "The Google File System" in 2003
  - "MapReduce: Simplified Data Processing on Large Clusters" in 2004
- The core idea was to distribute the data as it is initially stored
  - Each node can then perform computation on the data it stores without moving the data for the initial processing

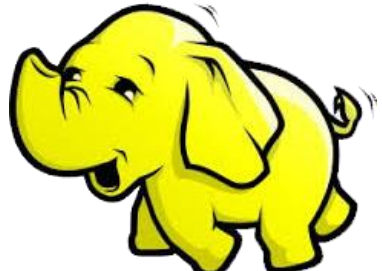
Hadoop's Developers

Doug Cutting

2005: Doug Cutting and Michael J. Cafarella developed Hadoop to support distribution for the Nutch search engine project.

The project was funded by Yahoo.

2006: Yahoo gave the project to Apache Software Foundation.



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Google Origins

2003

The Google File System

Sanjay Ghemawat, Howard Gobioff, and Shun-Tak Leung
Google*

2004

MapReduce: Simplified Data Processing on Large Clusters

Jeffrey Dean and Sanjay Ghemawat

jeff@google.com, sanjay@google.com

Google, Inc.

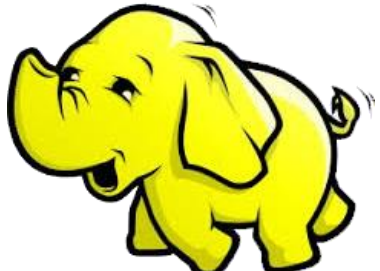
2006

Bigtable: A Distributed Storage System for Structured Data

Fay Chang, Jeffrey Dean, Sanjay Ghemawat, Wilson C. Hsieh, Deborah A. Wallach
Mike Burrows, Tushar Chandra, Andrew Fikes, Robert E. Gruber

Google, Inc.

Abstract



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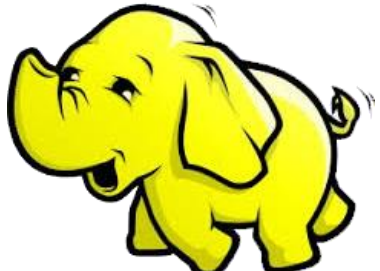
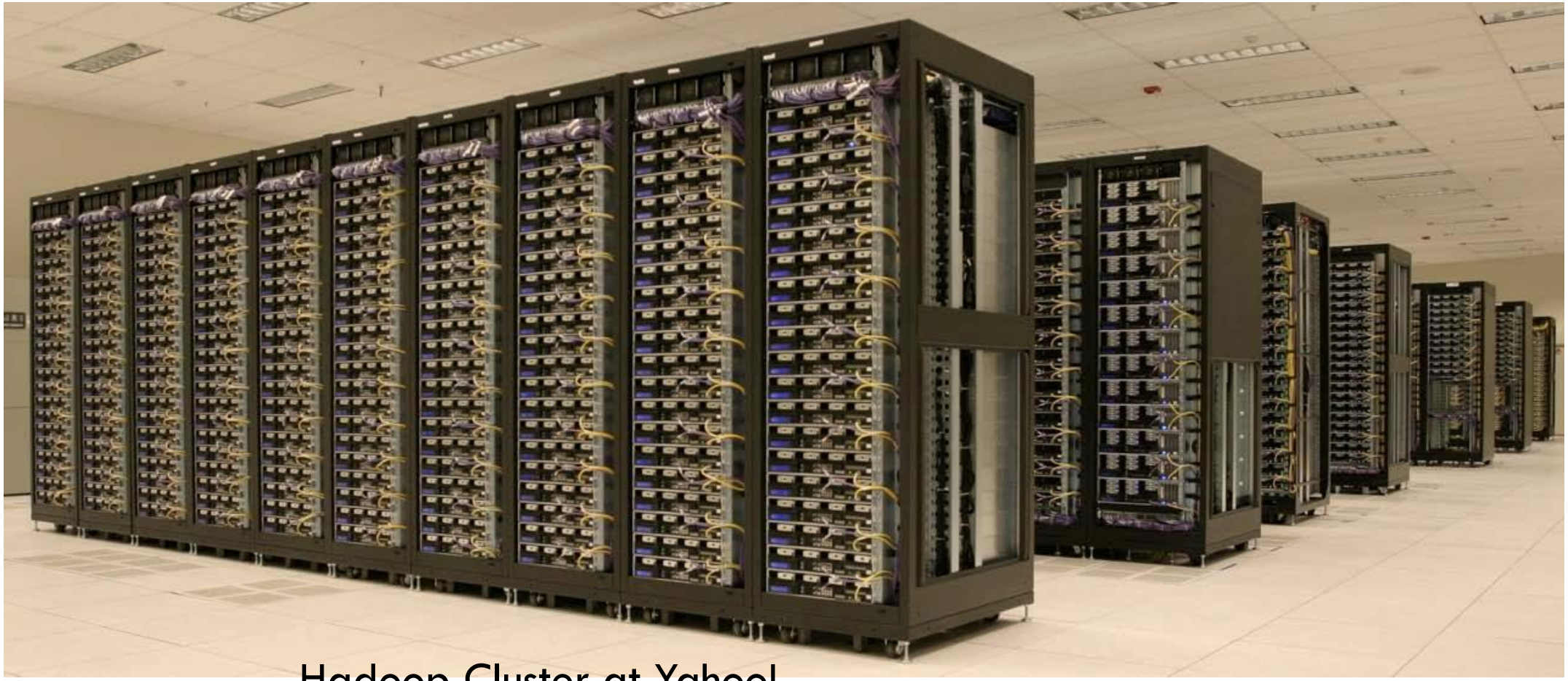
Some Hadoop Milestones

- **2008 - Hadoop Wins Terabyte Sort Benchmark (**sorted 1 terabyte of data in 209 seconds, compared to previous record of 297 seconds)
- 2009 - Avro and Chukwa became new members of Hadoop Framework family
- 2010 - Hadoop's Hbase, Hive and Pig subprojects completed, adding more computational power to Hadoop framework
- **2011 - ZooKeeper Completed**
- **2013 - Hadoop 1.1.2 and Hadoop 2.0.3 alpha.**
  - Ambari, Cassandra, Mahout have been added



One popular solution: Hadoop



Hadoop Cluster at Yahoo!

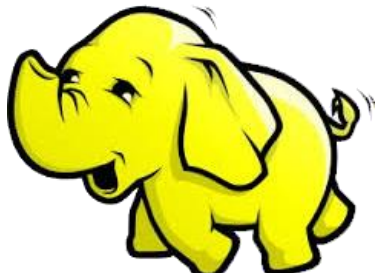
Hadoop in the Wild

- Hadoop is in use at most organizations that handle big data:
  - Yahoo!
  - Facebook
  - Amazon
  - Netflix
  - Etc…

- Some examples of scale:
  - Yahoo!'s Search Webmap runs on 10,000 core Linux cluster and powers Yahoo! Web search
  - FB's Hadoop cluster hosts 100+ PB of data (July, 2012) & growing at ½ PB/day (Nov, 2012)

Three main applications of Hadoop

- Advertisement (Mining user behavior to generate recommendations)
- Searches (group related documents)
- Security (search for uncommon patterns)



Big data storage is challenging

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- Data volumes are massive
- Reliability of storing PBs of data is challenging
- All kinds of failures: Disk/Hardware/Network Failures
- Probability of failures simply increase with the number of machines …

How much data?

- Facebook
  - 500 TB per day
- Yahoo
  - Over 170 PB
- eBay
  - Over 6 PB

- Getting the data to the processors becomes the bottleneck

Hadoop offers

- Redundant, Fault-tolerant data storage
- Parallel computation framework
- Job coordination

Hadoop offers

- Redundant, Fault-tolerant data storage
- Parallel computation framework
- Job coordination

Programmers

No longer need to worry about

Q: Where file is located?

Q: How to handle failures & data lost?

Q: How to divide computation?

Q: How to program for scaling?

Who uses Hadoop?

Social

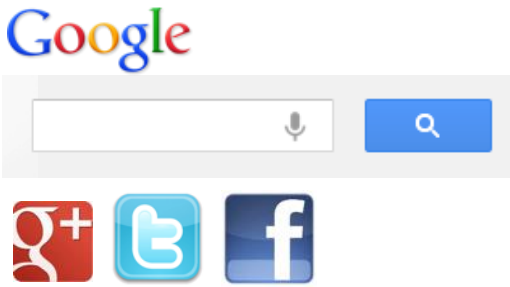
User Tracking & Engagement

Homeland Security

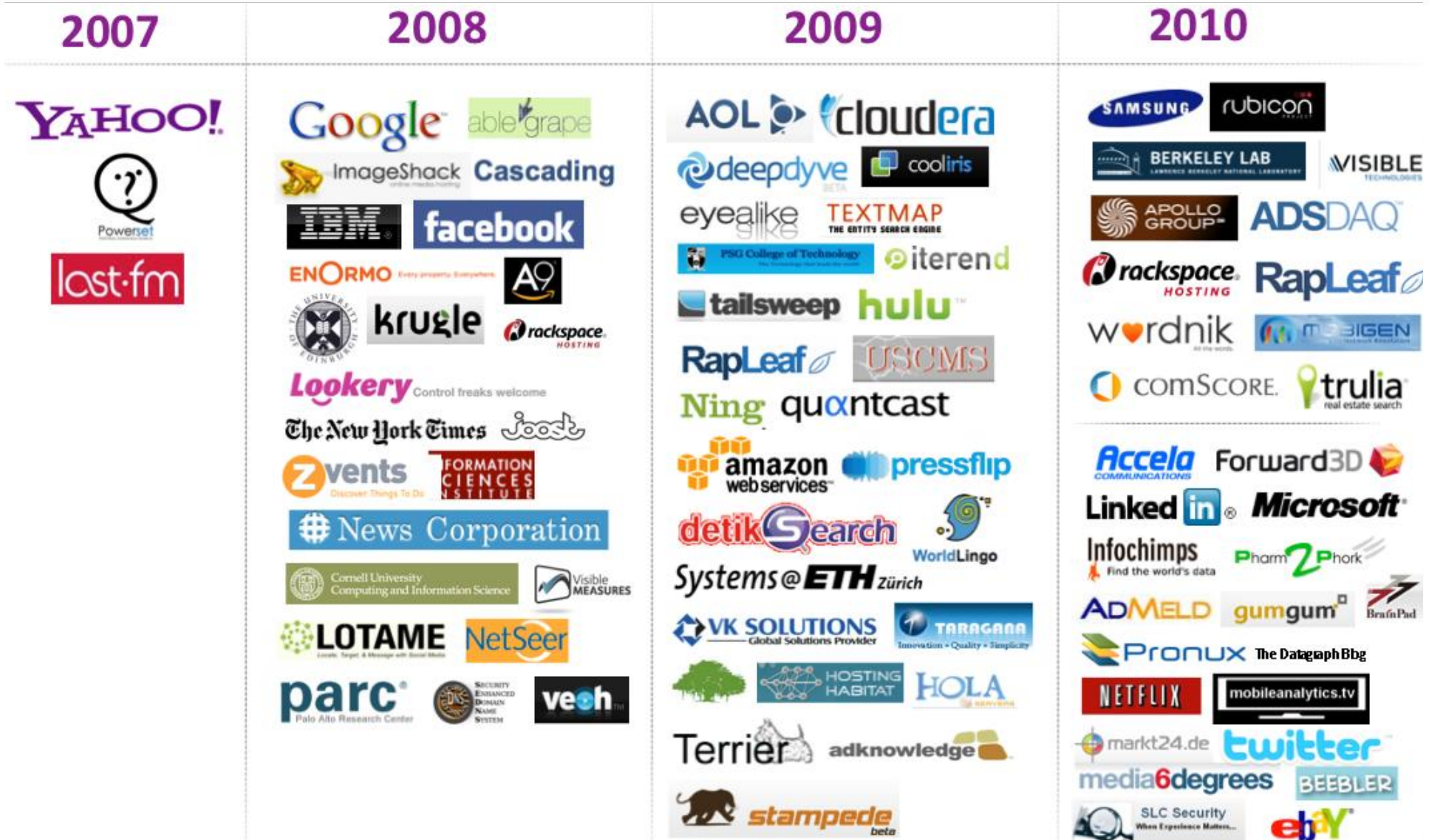
eCommerce

Financial Services

Real Time Search

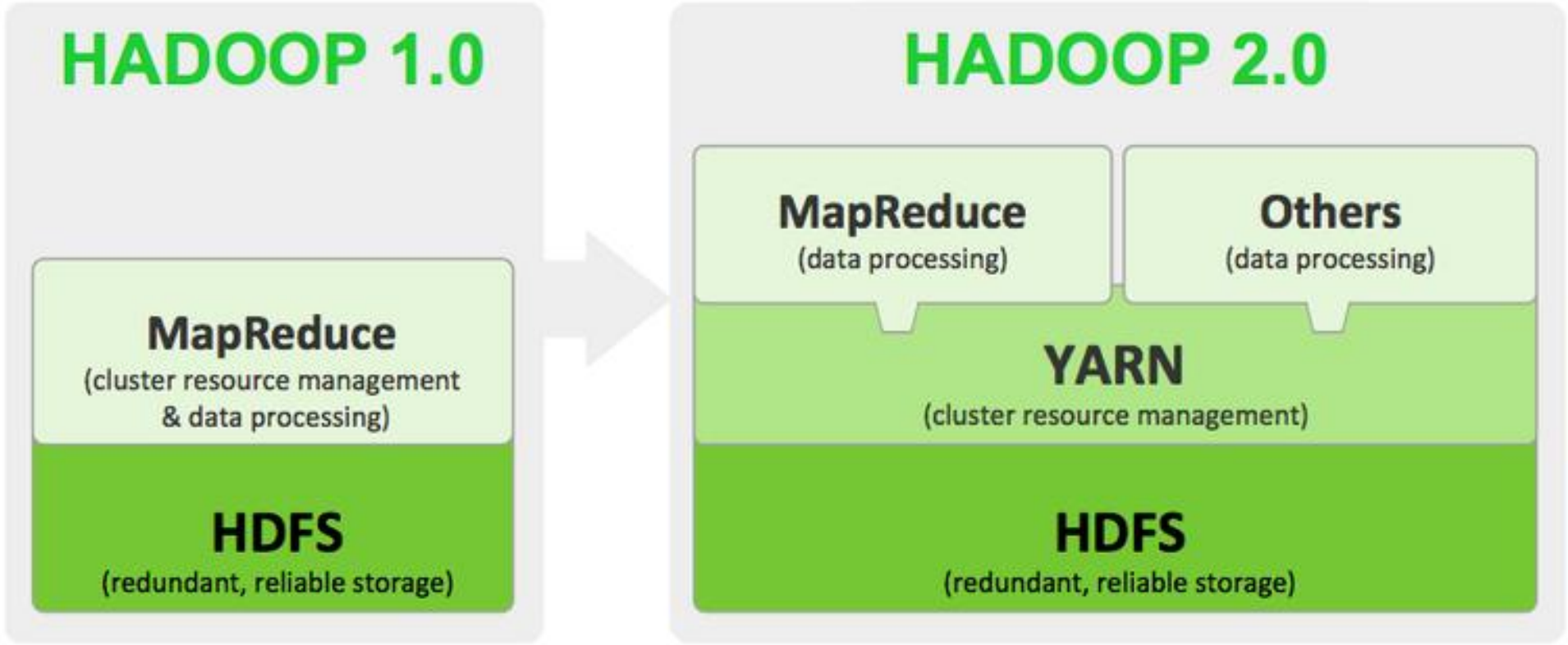


Who uses Hadoop?



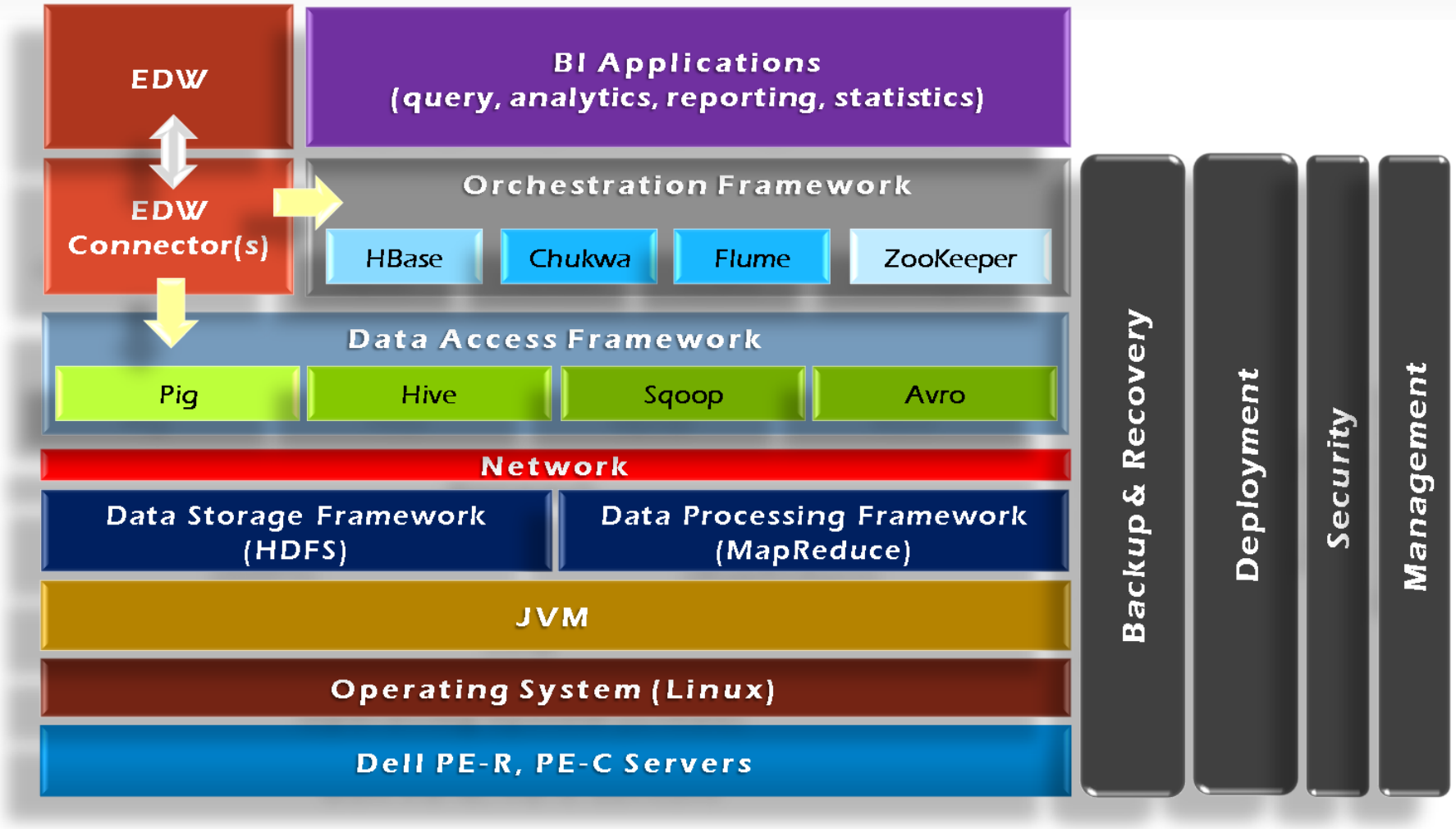
Hadoop big picture

MapReduce: Programming
YARN: Resource Scheduling
Hadoop: Data Management



HDFS + MapReduce is core to get work done

Hadoop Framework Tools



Hadoop goals

- **Scalable**: Petabytes (10^{15} Bytes) of data on thousands on nodes
- **Economical**: Commodity components only
- **Reliable**: fault tolerance

Hadoop Distributed File System HDFS

Overview

- Responsible for storing data on the cluster
- Data files are split into blocks and distributed across the nodes in the cluster
- Each block is replicated multiple times

HDFS Basic Concepts

- HDFS is a file system written in Java based on the Google's GFS
- Provides redundant storage for massive amounts of data

HDFS

- Master-Slave architecture
- Single NameNode
  - Sometimes a backup: secondary NameNode
- Many (Thousands) DataNodes
- Files are split into fixed sized blocks and stored on data nodes (default: 64MB)
- Data blocks are replicated for fault tolerance and fast access (default: 3)

HDFS Basic Concepts

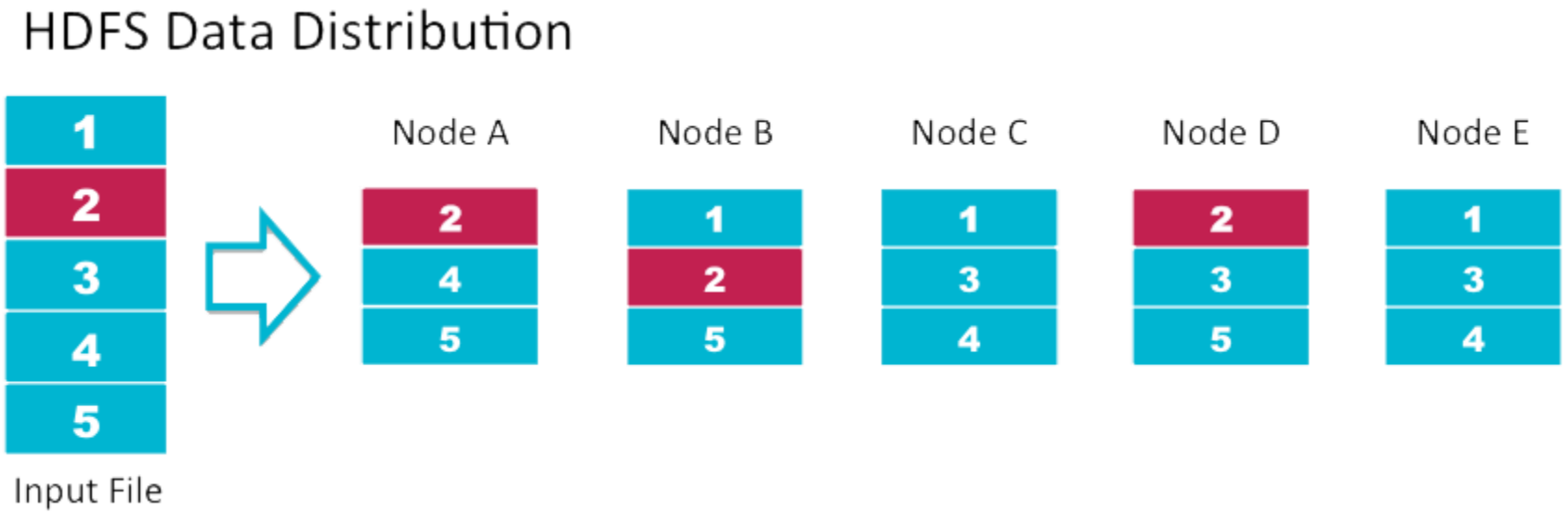
- HDFS works best with a smaller number of large files
  - Millions as opposed to billions of files
  - Typically 100MB or more per file
- Files in HDFS are write once
- Optimized for streaming reads of large files and not random reads

How are Files Stored

- Files are split into blocks
- Blocks are split across many machines at load time
  - Different blocks from the same file will be stored on different machines
- Blocks are replicated across multiple machines
- The NameNode keeps track of which blocks make up a file and where they are stored

Data Replication

- Default replication is 3-fold



Hadoop's Architecture

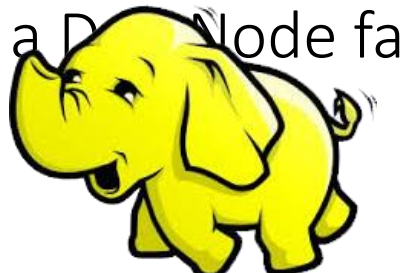
- NameNode:

- Stores metadata for the files, like the directory structure of a typical FS.

- The server holding the NameNode instance is quite crucial, as there is only one.

- Transaction log for file deletes/adds, etc. Does not use transactions for whole blocks or file-streams, only metadata.

- Handles creation of more replica blocks when necessary after a D[illegible]Node failure



Hadoop's Architecture

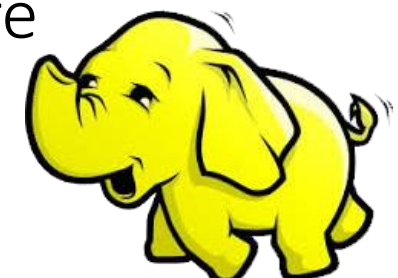
- **DataNode:**

- Stores the actual data in HDFS

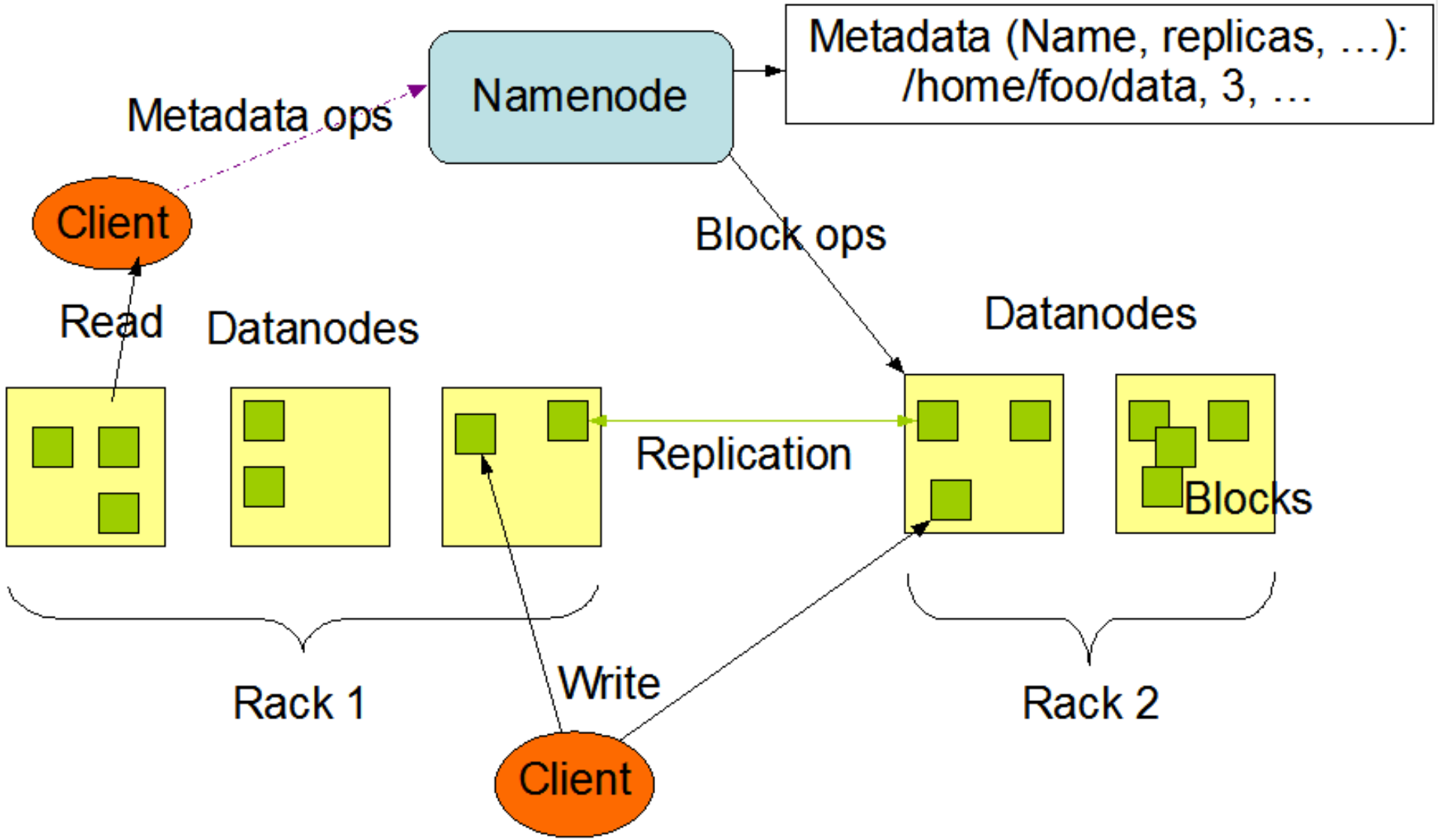
- Can run on any underlying filesystem (ext3/4, NTFS, etc)

- Notifies NameNode of what blocks it has

- NameNode replicates blocks 2x in local rack, 1x elsewhere



HDFS Architecture



Data Retrieval

- When a client wants to retrieve data
  - Communicates with the NameNode to determine which blocks make up a file and on which data nodes those blocks are stored
  - Then communicated directly with the data nodes to read the data

MapReduce

Distributing computation across nodes

MapReduce Overview

- A method for distributing computation across multiple nodes
- Each node processes the data that is stored at that node
- Consists of two main phases
  - Map
  - Reduce

MapReduce Features

- Automatic parallelization and distribution
- Fault-Tolerance
- Provides a clean abstraction for programmers to use

The Mapper

- Reads data as key/value pairs
  - The key is often discarded
- Outputs zero or more key/value pairs

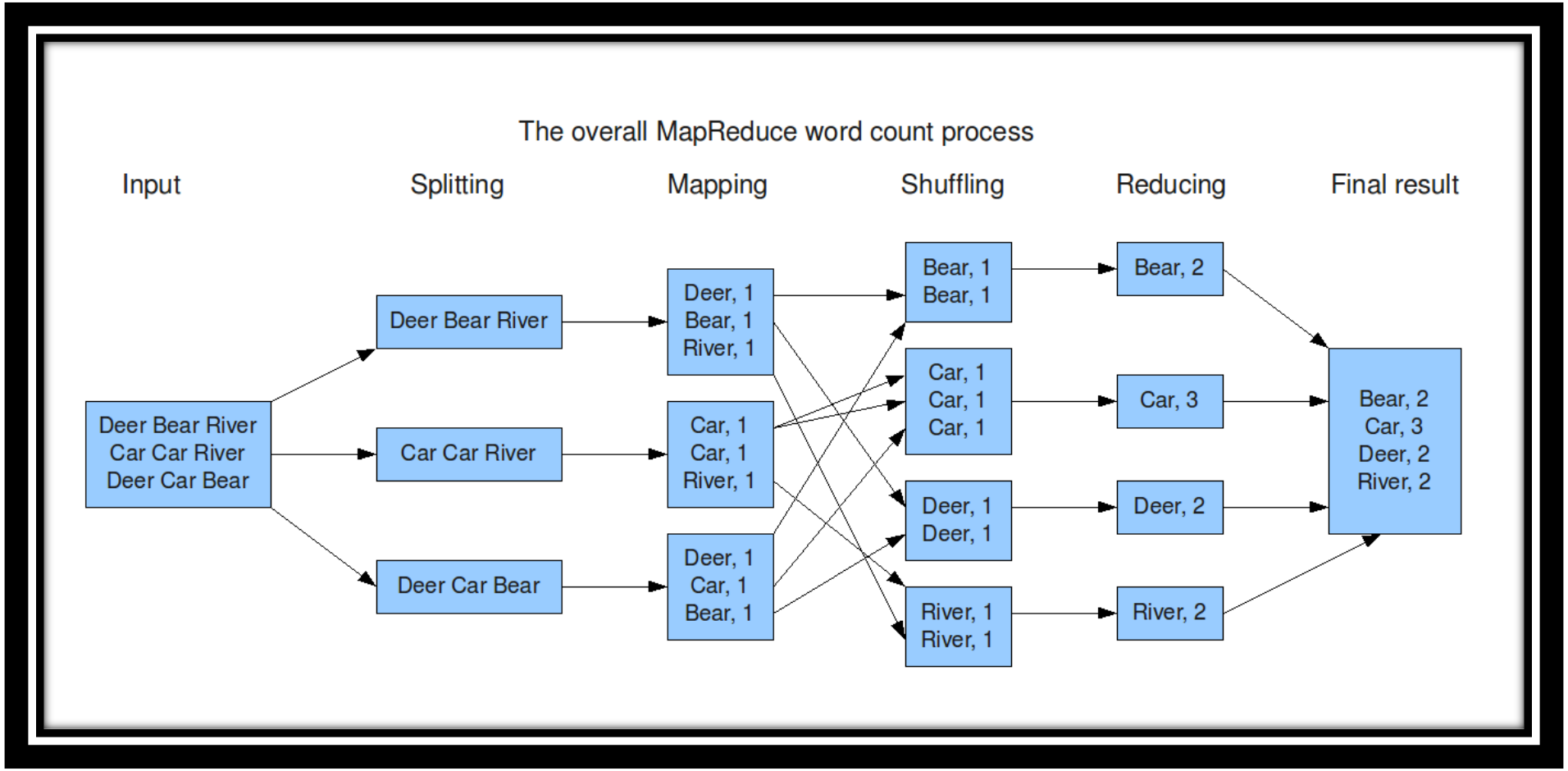
Shuffle and Sort

- Output from the mapper is sorted by key
- All values with the same key are guaranteed to go to the same machine

The Reducer

- Called once for each unique key
- Gets a list of all values associated with a key as input
- The reducer outputs zero or more final key/value pairs
  - Usually just one output per input key

MapReduce: Word Count



Hadoop Architecture

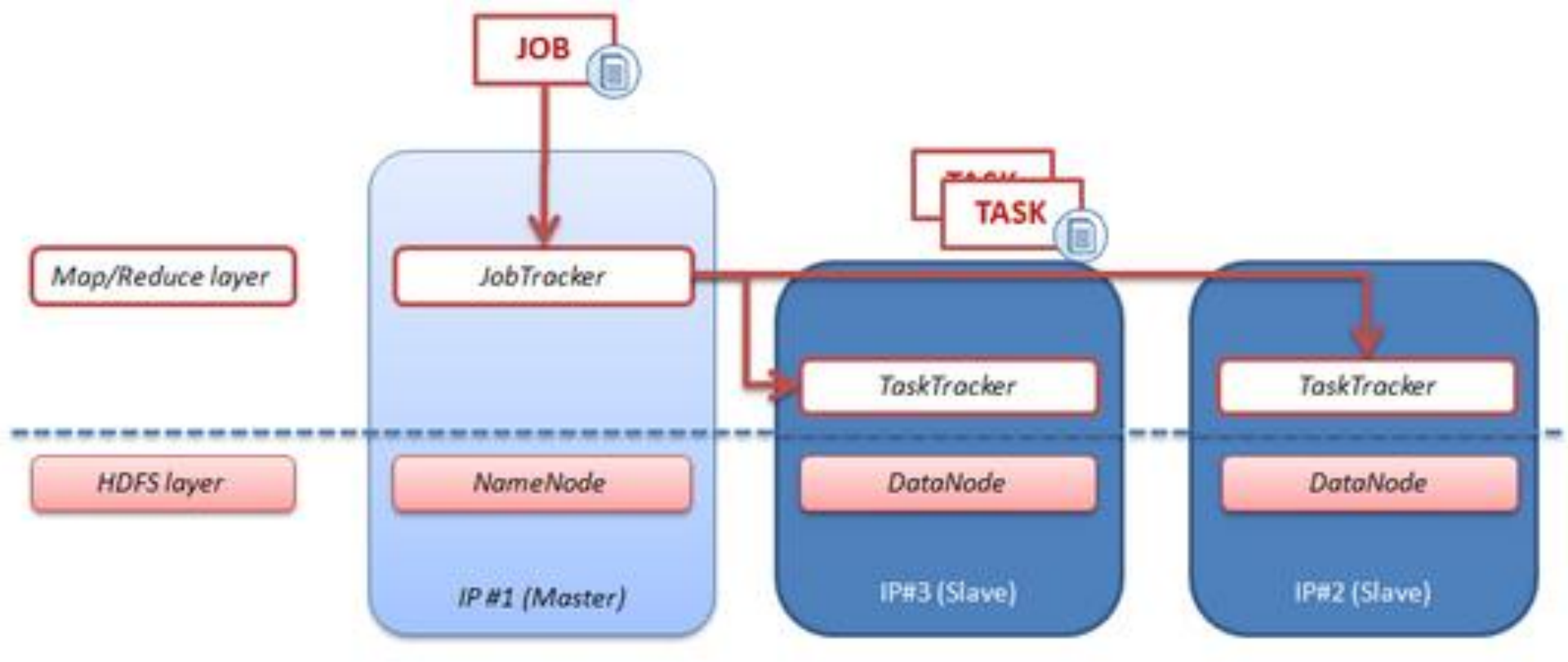
What parts actually make up a Hadoop cluster

- NameNode
  - Holds the metadata for the HDFS
- Secondary NameNode
  - Performs housekeeping functions for the NameNode
- DataNode
  - Stores the actual HDFS data blocks
- JobTracker
  - Manages MapReduce jobs
- TaskTracker
  - Monitors individual Map and Reduce tasks

JobTracker and TaskTracker

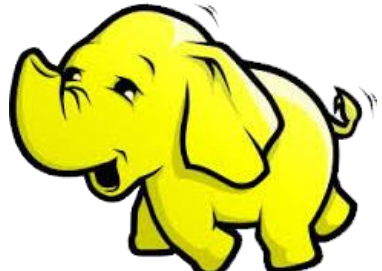
- JobTracker
  - Determines the execution plan for the job
  - Assigns individual tasks
- TaskTracker
  - Keeps track of the performance of an individual mapper or reducer

High-level architecture of Hadoop



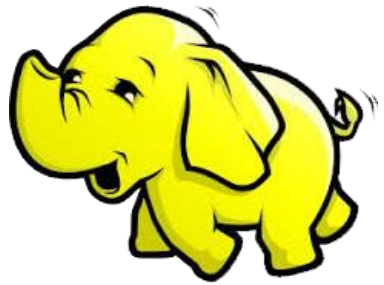
Hadoop's Architecture

- Hadoop Distributed Filesystem
- Tailored to needs of MapReduce
- Targeted towards many reads of filestreams
- Writes are more costly
- High degree of data replication (3x by default)
- No need for RAID on normal nodes
- Large blocksize (64MB)
- Location awareness of DataNodes in network



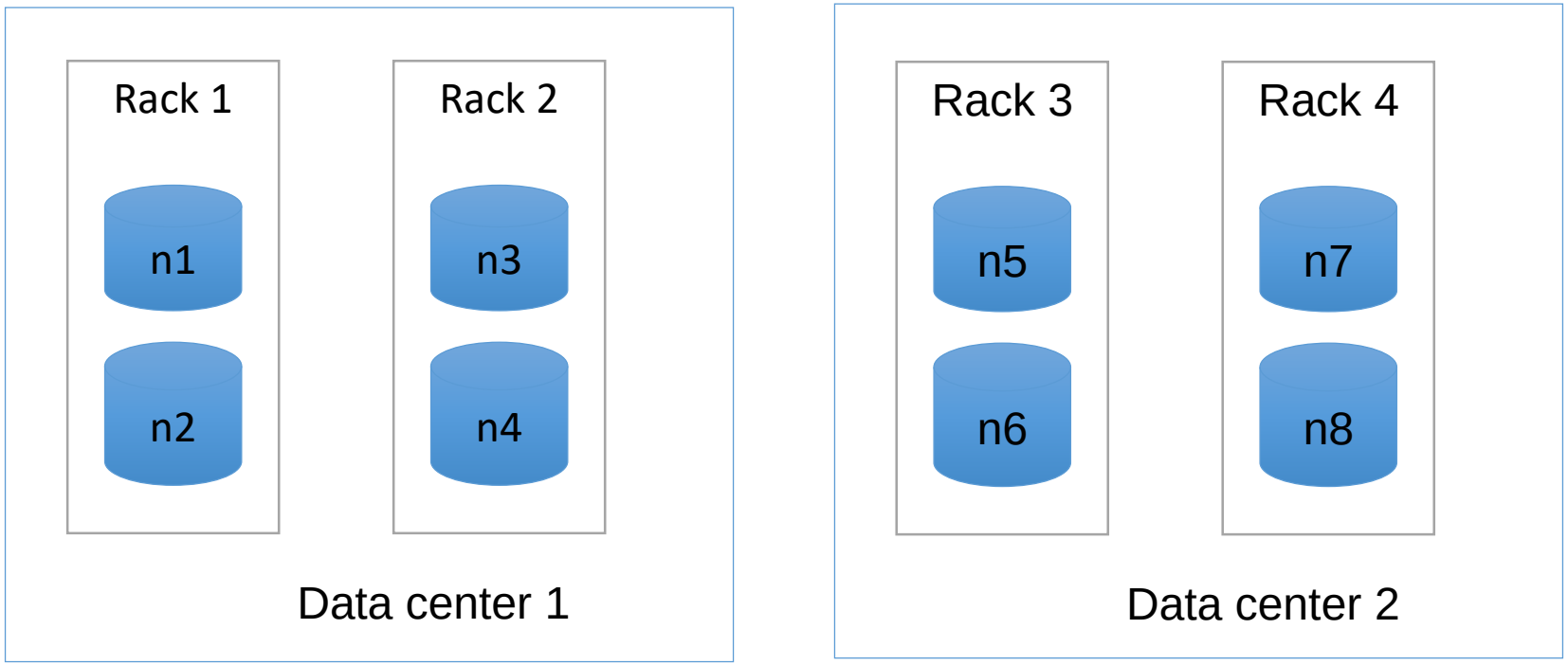
Hadoop's Architecture

- Distributed, with some centralization
- Main nodes of cluster are where most of the computational power and storage of the system lies
- Main nodes run **TaskTracke**r to accept and reply to MapReduce tasks, and also **DataNode** to store needed blocks closely as possible
- Central control node runs **NameNode** to keep track of HDFS directories & files, and **JobTracker** to dispatch compute tasks to TaskTracker
- Written in Java, also supports Python and Ruby



HDFS network topology

- HDFS takes a simple approach:
  - ❑See the network as a tree
  - ❑Distance between two nodes is the sum of their distances to their closest common ancestor



HDFS filesystem commands

1. List the contents of a directory
   - `hadoop fs -ls`
2. Create a directory in HDFS at given path(s)
   - `hadoop fs -mkdir <directory name>`

3. Upload and download a file in HDFS
   - Upload: `hadoop fs -put <local file> <remote path>`
   - Download: `hadoop fs -get <file in HDFS> <local path>`

4. See contents of a file
   - `hadoop fs -cat <filename>`
5. Delete a file/directory in HDFS
   - `hadoop fs -rm/rmr <file or directory>`

HDFS filesystem commands

6. Move file from source to destination
   - `hadoop fs -mv <src> <dst>`
7. Report the amount of space used and availability
   - `hadoop fs -df hdfs:/`
8. How much space a directory occupies
   - `hadoop fs -du -s -h <dir name>`
9. Change permission of files
   - `sudo hadoop fs -chmod 600 <file>`
10. Change owner and group of files
    - `sudo hadoop fs -chown root:root <file>`

*HDFS admin commands

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- DFSAdmin command
  - -report: reports basic statistics of HDFS
  - -safemode: though usually not required, an administrator can manually enter or leave safemode
    - enter, leave, get, wait
  - -refreshNodes: updates the set of hosts allowed to connect to namenode

Usage: hadoop dfsadmin [-report] [-safemode enter | leave | get | wait] [-refreshNodes] [-finalizeUpgrade] [-upgradeProgress status | details | force] [-metasave filename] [-setQuota <quota> <dirname>...<dirname>] [-clrQuota <dirname>...<dirname>] [-help [cmd]]

Installation and configuration

Next week: Lab 3: Cloudera Hadoop Installation

Questions?

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