Introduction to Hadoop

CSE 569

Content

- Distributed system, DFS
- Hadoop
- Architecture of Hadoo
- Hadoop usage at Facebook

Operating systems

- Operating system Software that supervises and controls tasks on a computer. Individual OS:
 - Batch processing ◊ jobs are collected, placed in a queue, no interaction with job during processing
 - Time shared \(\) computing resources are provided to different users, interaction with program during execution
 - RT systems ◊ fast response, can be interrupted

Distributed Systems

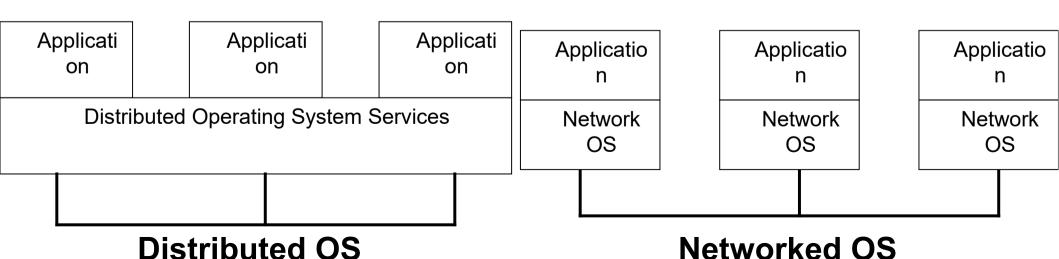
- Consists of a number of computers that are connected and managed so that they automatically share the job processing load among the constituent computers.
- A distributed operating system is one that appears to its users as a traditional uniprocessor system, even though it is actually composed of multiple processors.
- It gives a single system view to its users and provides a single service.
- Users are transparent to location of files. It provides a virtual computing environment.

Eg The Internet, ATM banking networks, mobile computing networks, Global Positioning Systems and Air Traffic Control

DISTRIBUTED SYSTEM IS A COLLECTION OF INDEPENDENT COMPUTERS THAT APPEARS TO IS USERS AS A SINGLE COHERENT SYSTEM

Network Operating System

- In a network operating system the users are aware of the existence of multiple computers.
- The operating system of individual computers must have facilities to have communication and functionality.
- Each machine runs its own OS and has its own user.
- Remote login and file access
- Less transparent but more independency



DFS

- Resource sharing is the motivation behind distributed Systems. To share files ◊ file system
 - File System is responsible for the organization, storage, retrieval, naming, sharing, and protection of files.
 - The file system is responsible for controlling access to the data and for performing low-level operations such as buffering frequently used data and issuing disk I/O requests
 - The goal is to allow users of physically distributed computers to share data and storage resources by using a common file system.

What is Hadoop

- Open source software platform for scalable, distributed computing
- Hadoop provides fast and reliable analysis of both structured data and unstructured data
- Apache Hadoop software library is essentially a framework that allows for the distributed processing of large datasets across clusters of computers using a simple programming model.
- Hadoop can scale up from single servers to thousands of machines, each offering local computation and storage.

What is Hadoop?..Contd

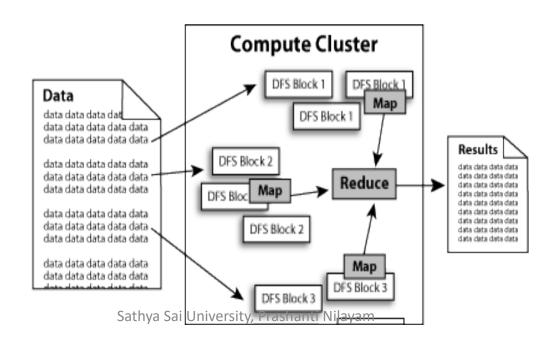
- Software platform that lets one easily write and run applications that process vast amounts of data. It includes:
 - MapReduce offline computing engine
 - HDFS Hadoop distributed file system
- Here's what makes it especially useful:
 - Scalable: It can reliably store and process petabytes.
 - **Economical:** It distributes the data and processing across clusters of commonly available computers (in thousands).
 - Efficient: By distributing the data, it can process it in parallel on the nodes where the data is located.
 - Reliable: It automatically maintains multiple copies of data and automatically redeploys computing tasks based on failures.

Who uses Hadoop?

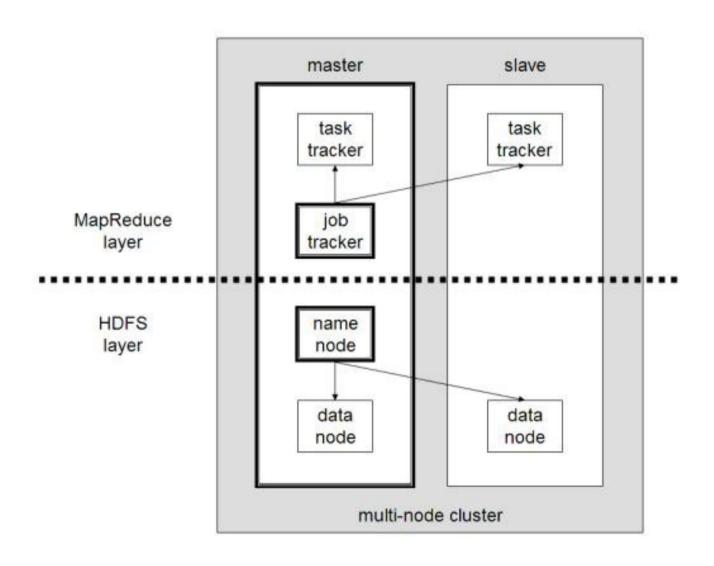
- Amazon/A9
- Facebook
- Google
- IBM
- Joost
- Last.fm
- New York Times
- PowerSet
- Veoh
- Yahoo!

What does it do?

- Hadoop implements Google's MapReduce, using HDFS
- MapReduce divides applications into many small blocks of work.
- HDFS creates multiple replicas of data blocks for reliability, placing them on compute nodes around the cluster.
- MapReduce can then process the data where it is located.
- Hadoop 's target is to run on clusters of the order of 10,000-nodes.



Hadoop Cluster Architecture:



HDFS

- HDFS uses a master/slave architecture where master consists of a single NameNode that manages the file system metadata and one or more slave DataNodes that store the actual data.
- A file in an HDFS namespace is split into several blocks and those blocks are stored in a set of DataNodes. The NameNode determines the mapping of blocks to the DataNodes.
- The DataNodes takes care of read and write operation with the file system. They also take care of block creation, deletion and replication based on instruction given by NameNode.

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 DataNode failure

NameNode

- DFS Master "Namenode"
 - Manages the file system namespace
 - Controls read/write access to files
 - Manages block replication
 - Checkpoints namespace and journals namespace changes for reliability

Metadata of Name node in Memory

- The entire metadata is in main memory
- No demand paging of FS metadata

Types of Metadata:

List of files, file and chunk namespaces; list of blocks, location of replicas; file attributes etc.

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

DATA NODES

- Serve read/write requests from clients
- Perform replication tasks upon instruction by namenode

Data nodes act as:

- 1) A Block Server
 - Stores data in the local file system
 - Stores metadata of a block (e.g. CRC)
 - Serves data and metadata to Clients
- 2) Block Report: Periodically sends a report of all existing blocks to the NameNode
- 3) Periodically sends heartbeat to NameNode (detect node failures)
- 4) Facilitates Pipelining of Data (to other specified DataNodes)

Data Model

- Data is organized into files and directories
- Files are divided into uniform sized blocks and distributed across cluster nodes
- Replicate blocks to handle hardware failure
- Checksums of data for corruption detection and recovery

- Assumes commodity hardware that fails
 - Files are replicated to handle hardware failure
 - Checksums for corruption detection and recovery
 - Continues operation as nodes / racks added / removed
- Optimized for fast batch processing
 - Data location exposed to allow computes to move to data
 - Stores data in chunks/blocks on every node in the cluster
 - Provides VERY high aggregate bandwidth

- Files are broken in to large blocks.
 - Typically 128 MB block size
 - Blocks are **replicated** for reliability
 - One replica on local node,
 another replica on a remote rack,
 Third replica on local rack,
 Additional replicas are randomly placed
- Understands rack locality
 - Data placement exposed so that computation can be migrated to data
- Client talks to both NameNode and DataNodes
 - Data is not sent through the namenode, clients access data directly from DataNode
 - Throughput of file system scales nearly linearly with the number of nodes.

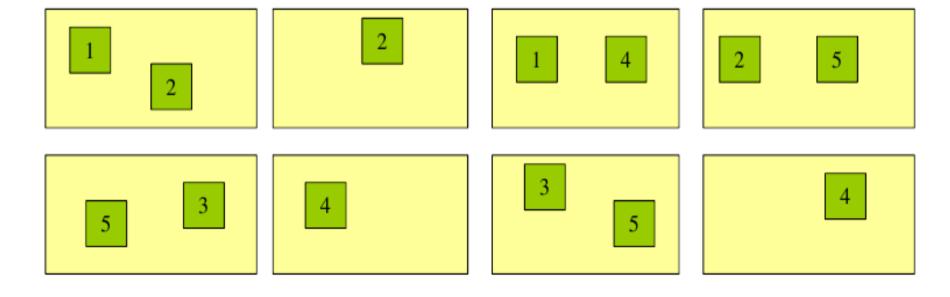
Block Placement

Namenode

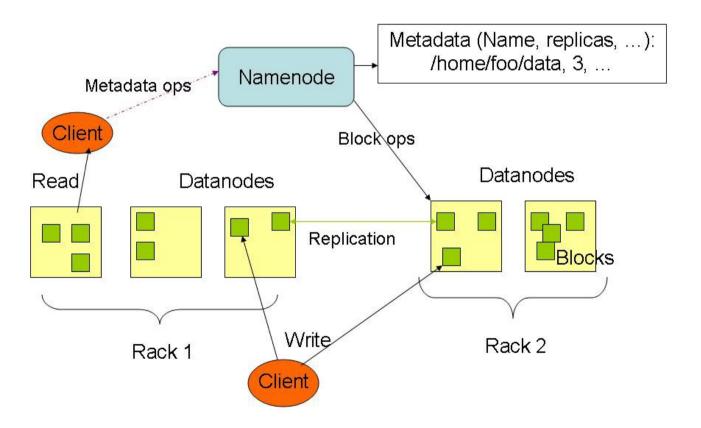
name:/users/foo/myFile - copies:2, blocks:{1,3}

name:/users/bar/someData.gz, copies:3, blocks:{2,4,5}

Datanodes



HDFS Architecture



MapReduce

The term MapReduce actually refers to the following two different tasks that Hadoop programs perform:

The Map Task: This is the first task, which takes input data and converts it into a set of data, where individual elements are broken down into tuples key/value pairs.

The Reduce Task: This task takes the output from a map task as input and combines those data tuples into a smaller set of tuples. The reduce task is always performed after the map task.

How does MapReduce work?

- The run time partitions the input and provides it to different Map instances;
- Map (key, value) ◊ (key', value')
- The run time collects the (key', value') pairs and distributes them to several Reduce functions so that each Reduce function gets the pairs with the same key'.
- Each Reduce produces a single (or zero) file output.
- Map and Reduce are user written functions

Example MapReduce: To count the occurrences of words in the given set of documents

```
map(String key, String value):
// key: document name; value: document contents; map (k1,v1) ◊
   list(k2,v2)
for each word w in value: EmitIntermediate(w, "1");
(Example: If input string is ("Saibaba is God. I am I"), Map produces
   {<"Saibaba",1">, <"is", 1>, <"God", 1>, <"I",1>, <"am",1>,<"I",1>}
reduce(String key, Iterator values):
// key: a word; values: a list of counts; reduce (k2,list(v2)) ◊ list(v2)
int result = 0;
for each v in values:
result += ParseInt(v);
Emit(AsString(result));
(Example: reduce("I", <1,1>) ◊ 2)
```

MapReduce Engine

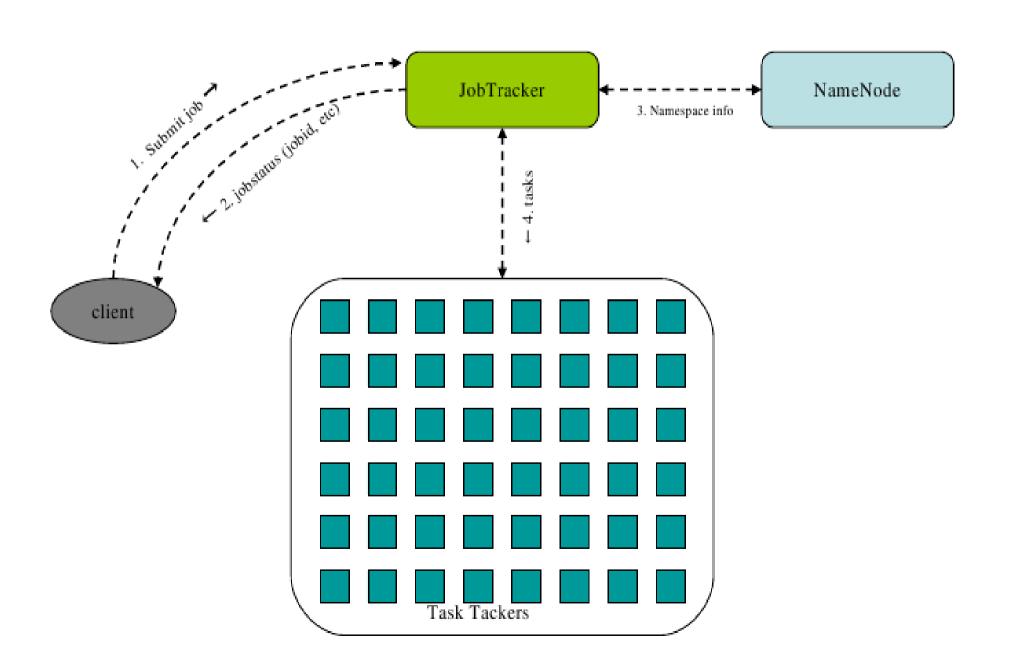
Map/Reduce Master "Jobtracker"

- Accepts MR jobs submitted by users
- Assigns Map and Reduce tasks to Tasktrackers
- Monitors task and tasktracker status, re-executes tasks upon failure

Map/Reduce Slaves "Tasktrackers"

- Run Map and Reduce tasks upon instruction from the Jobtracker
- Manage storage and transmission of intermediate output.

JOBTRACKER, TASKTACKER AND JOBCLIENT



* Spackage WordFress * Saubpackage Default_Theme

Hadoop's Architecture: MapReduce Engine

<!DOCTYPE html PUBLIC "-//WSG chtml xmlns="http://www.wd. chead profile="http://

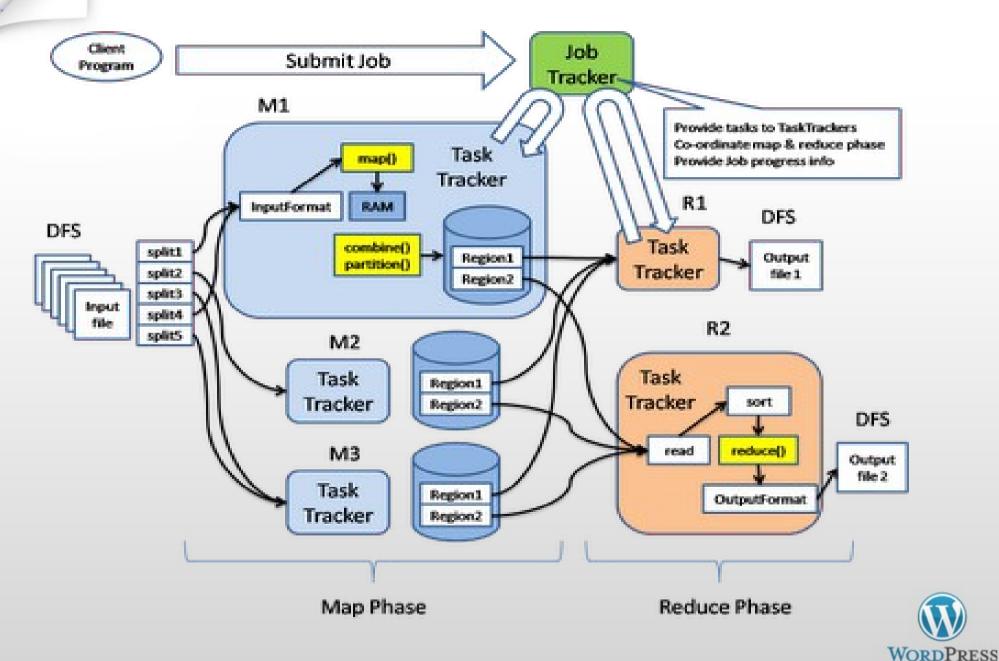
cmeta http-equiv="Cor

ctitle><?php wp

clink relets

clink rel-





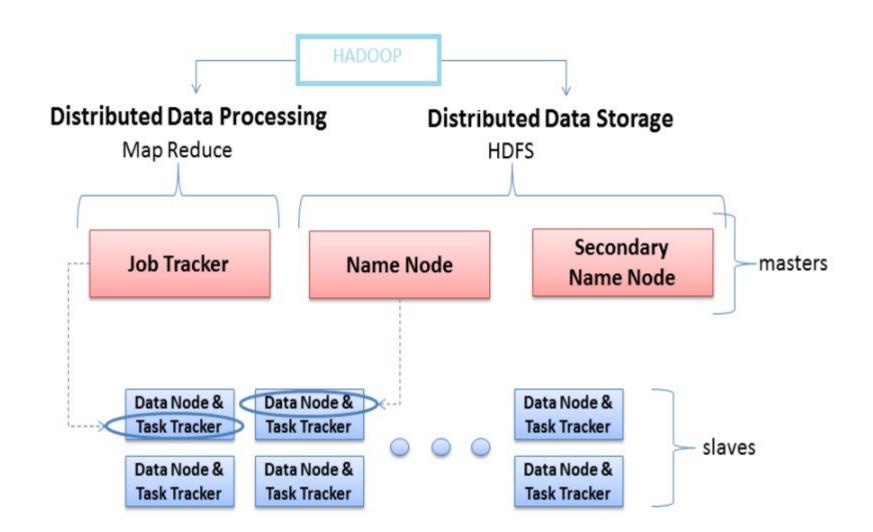
Hadoop at Facebook

Production cluster

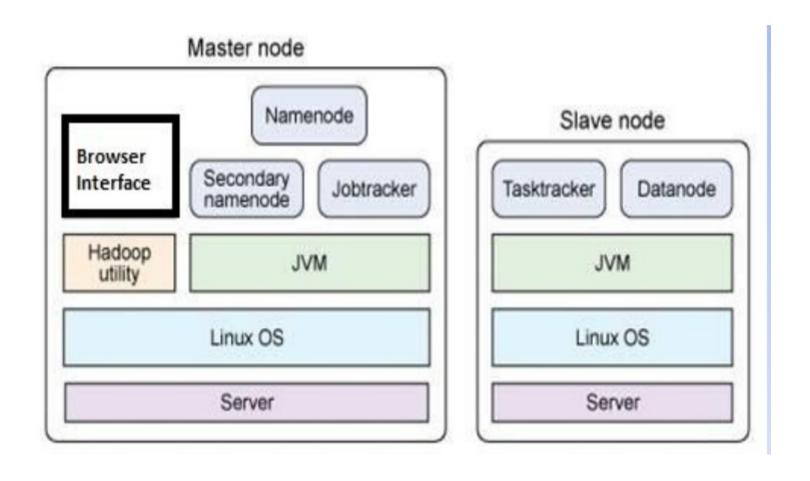
- 4800 cores, 600 machines, 16GB per machine April 2009
- 8000 cores, 1000 machines, 32 GB per machine July 2009
- 4 SATA disks of 1 TB each per machine
- 2 level network hierarchy, 40 machines per rack
- Total cluster size is 2 PB, projected to be 12 PB in Q3 2009

Test cluster

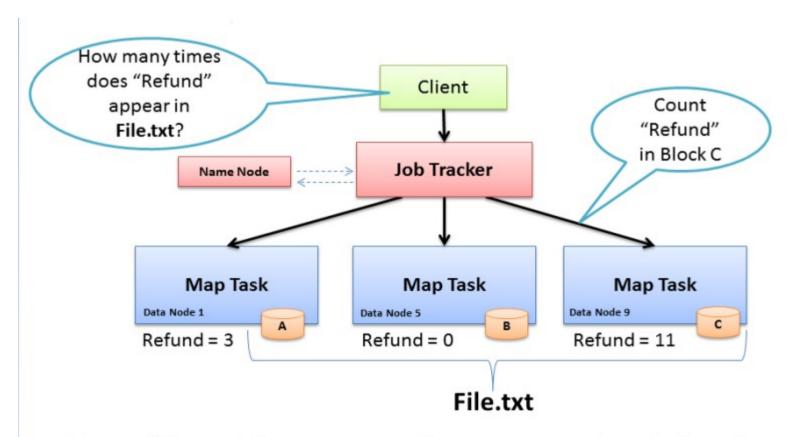
800 cores, 16GB each



HDFS Cluster Architecture

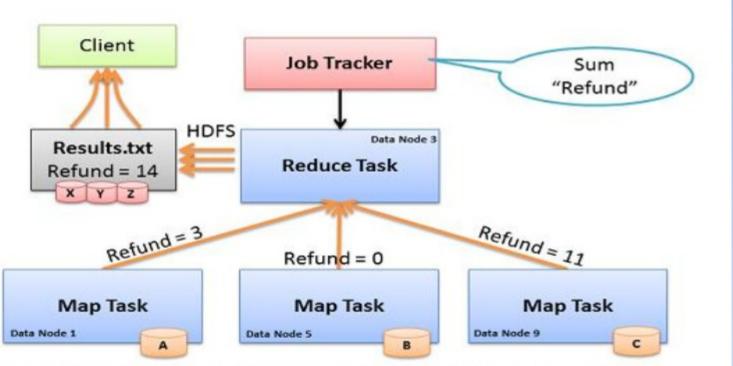


MAP Task



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

Reduce Task



- 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