

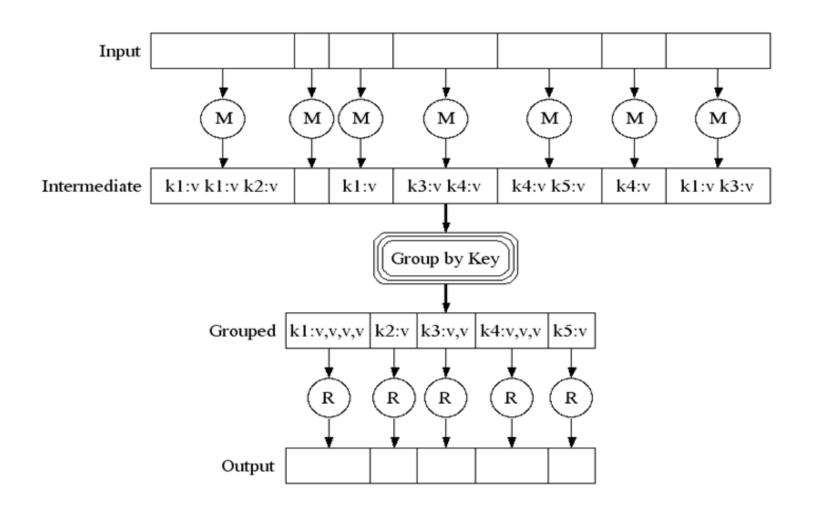
What is MapReduce?

- A computational model published in a paper by Google in 2004
- Based on distributed computation
- Complements Google's distributed file system (GFS)
- Works with key:value pairs

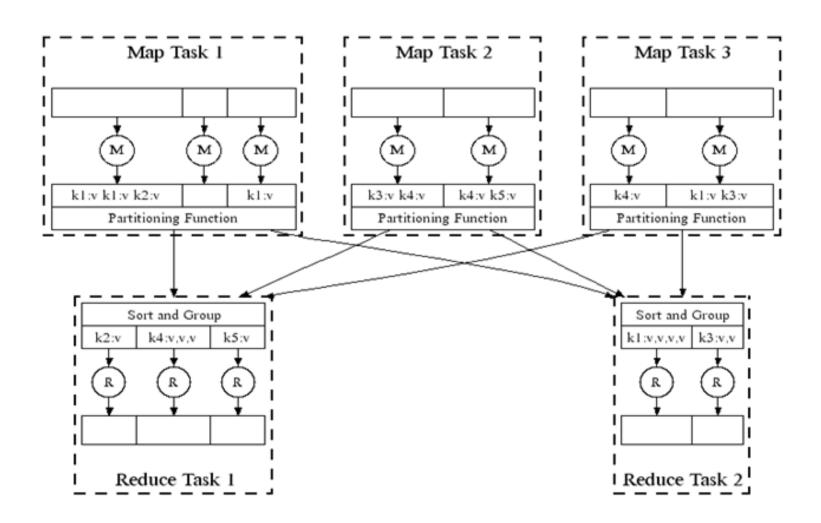
Why MapReduce?

- It is the ,answer' for Big Data problem
- Runs on commodity hardware
- Very scalable solution

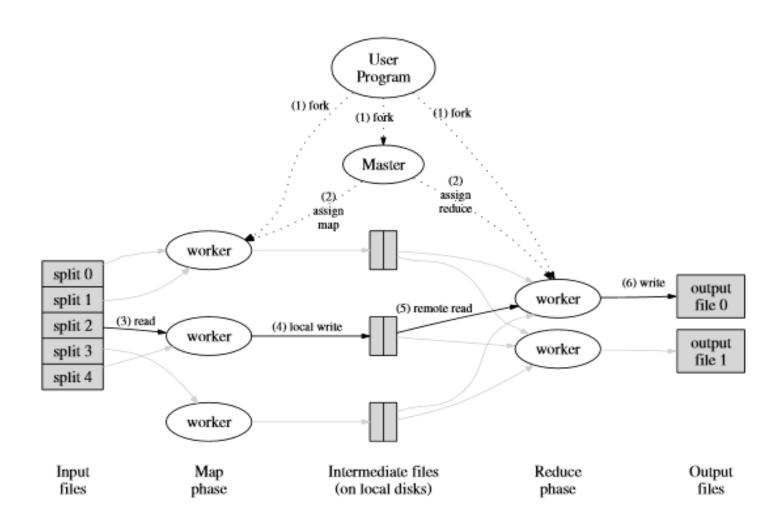
MapReduce model



MapReduce model



MapReduce model



MapReduce alternatives

- Hadoop Top-level Apache project
- Spark University of Berkeley project
- Disco Open source Nokia project
- MapReduce-MPI US Department of Energy project
- MARIANE academic project of University of Binghamton
- Phoenix University of Stanford project
- BashReduce MapReduce for std. Unix commands

MapReduce vs. RDBMS

	Traditional RDBMS	MadReduce
Data Size	Gigabytes	Petabytes
Access	Interactive and batch	Batch
Updates	Read and write many times	Write once, read many times
Structure	Static schema	Dynamic schema
Integrity	High	Low
Scaling	Nonlinear	Linear

Data Structure

- Structured Data data organized into entities that have a defined format.
 - Realm of RDBMS
- Semi-Structured Data there may be a schema, but often ignored; schema is used as a guide to the structure of the data.
- Unstructured Data doesn't have any particular internal structure.
- MapReduce works well with semi-structured and unstructured data.

What is Hadoop?

- Software platform that lets one easily write and run applications that process vast amounts of data
- Hadoop is most popular implementation of MapReduce so far

Why Hadoop?

- It has been Apache top-level project for a long time (6 years)
- Hadoop Ecosystem
- Hadoop exclusive technologies

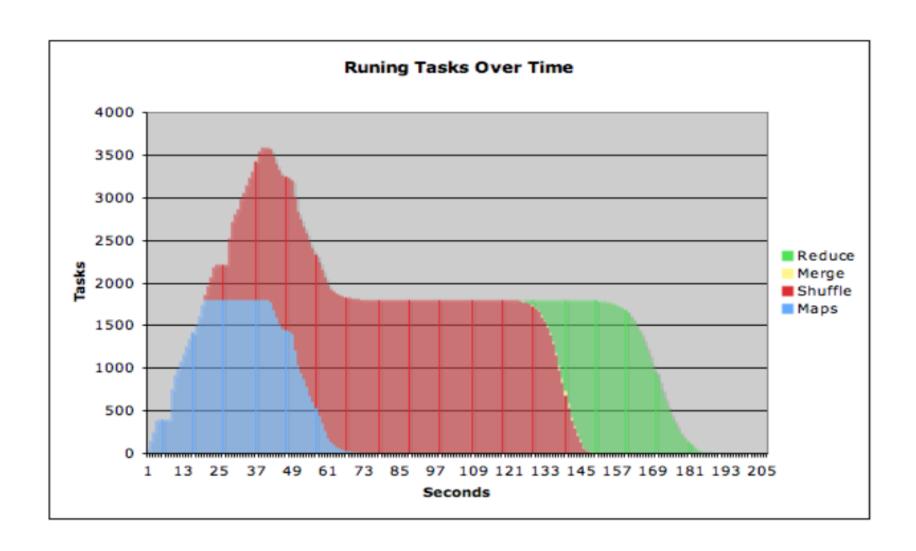
Why Hadoop?

- 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

Brief history

- 2002 Project Nutch started (open source web search engine) Doug Cutting
- 2003 GFS (Google File System) paper published
- 2004 Implementation of GFS started
- 2004 Google published MapReduce paper
- 2005 Working implementations of MapReduce and GFS (NDFS)
- 2006 System applicable beyond realm of search
- 2006 Nutch moved to Hadoop project, Doug Cutting joins Yahoo!
- 2008 Yahoo!s production index generated by 10,000 core Hadoop cluster
- 2008 Hadoop moved under Apache Foundation
- April 2008 Hadoop broke world record fastest sorting of 1 TB of data (209 seconds, previously 297)
- November 2008 Google's implementation sorted 1 TB in 68 seconds
- May 2009 Yahoo! team sort 1 TB in 62 seconds

1TB sort by Hadoop



Who uses Hadoop?







The New Hork Times

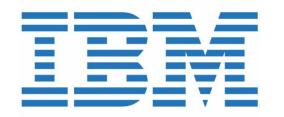
















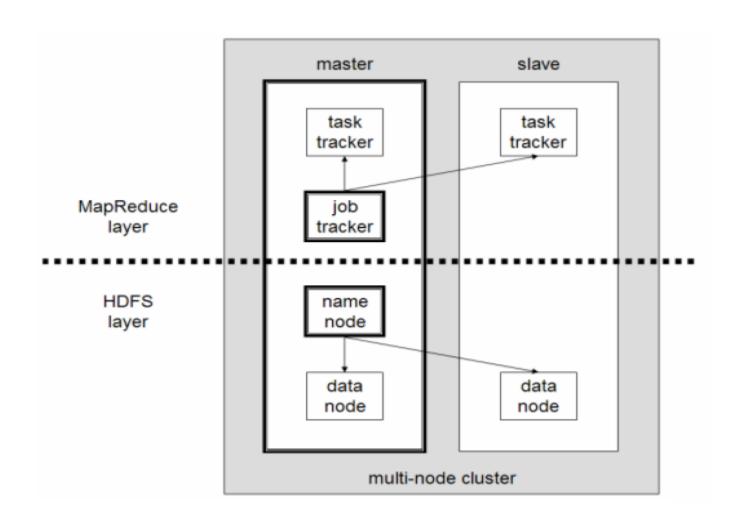
Assumptions

- Hardware will fail
- Processing will be run in batches. Thus there is an emphasis on high throughput as opposed to low latency
- Applications that run on HDFS have large data sets. A typical file in HDFS is gigabytes to terabytes in size
- It should provide high aggregate data bandwidth and scale to hundreds of nodes in a single cluster. It should support tens of millions of files in a single instance
- Applications need a write-once-read-many access model
- Moving Computation is Cheaper than Moving Data
- Portability is important

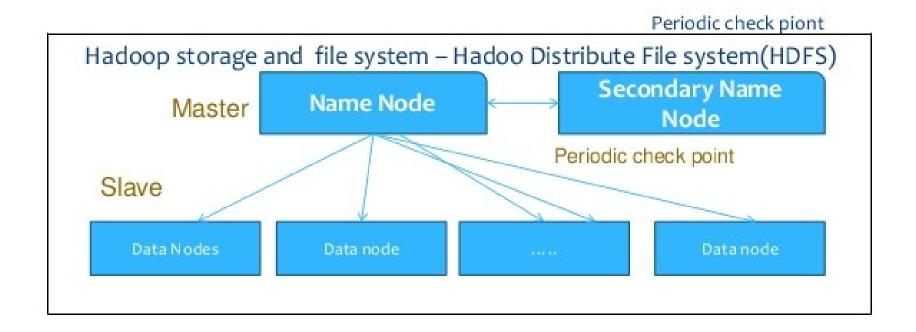
Hadoop modules

- Hadoop Common contains libraries and utilities needed by other Hadoop modules
- Hadoop Distributed File System (HDFS) a distributed file-system that stores data on the commodity machines, providing very high aggregate bandwidth across the cluster
- Hadoop MapReduce a programming model for large scale data processing
- Hadoop YARN a resource-management platform responsible for managing compute resources in clusters and using them for scheduling of users' applications
 - Provides base for MapReduce v2

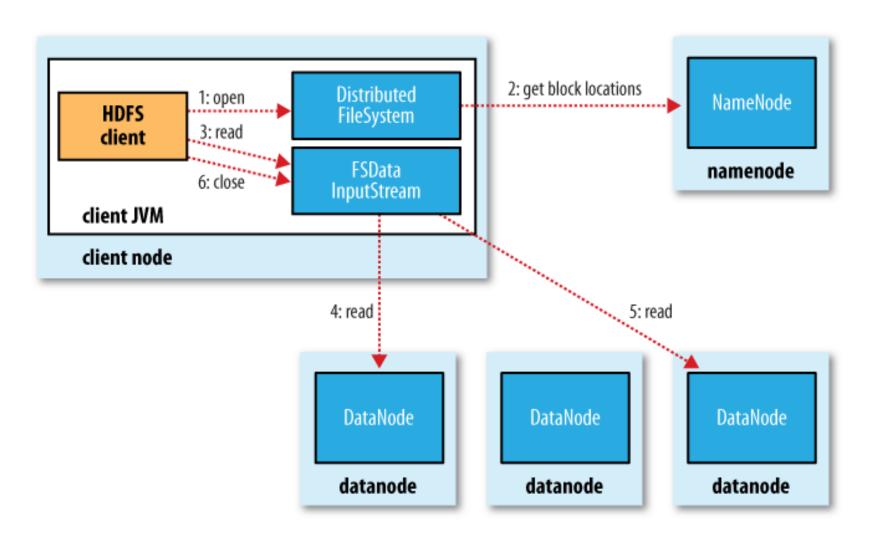
Hadoop architecture



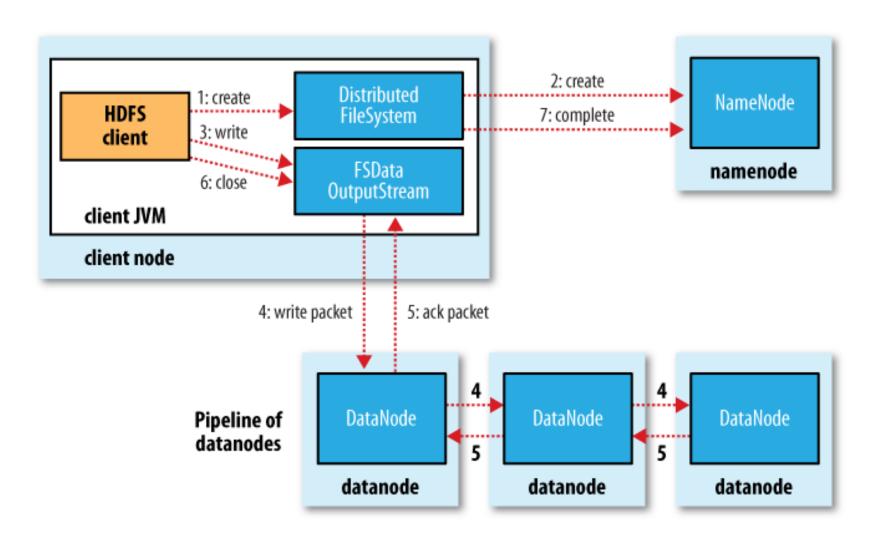
HDFS architecture



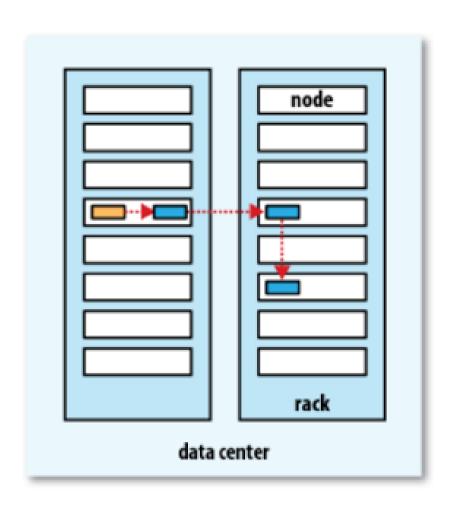
HDFS architecture (reading)



HDFS architecture (writing)



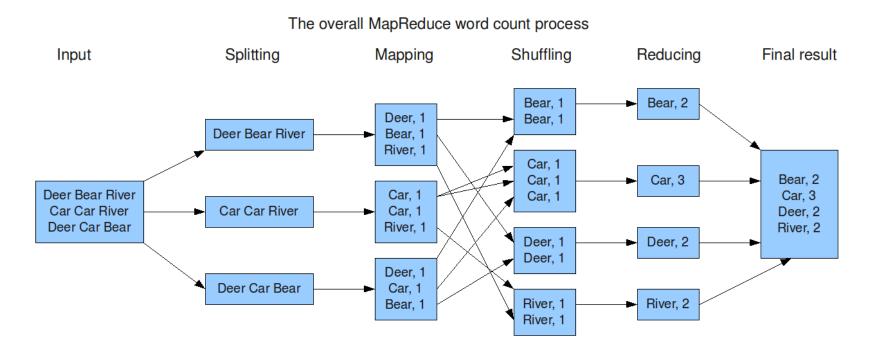
Data replication in HDFS



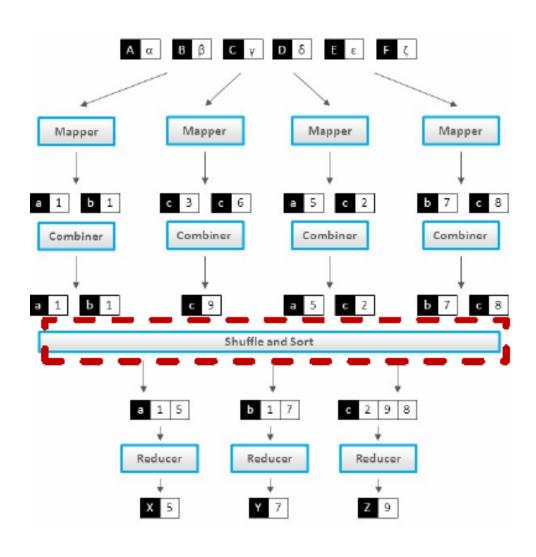
How to use Hadoop MapReduce?

- Implement 2 basic functions:
 - Map
 - Reduce
- Implement Driver class

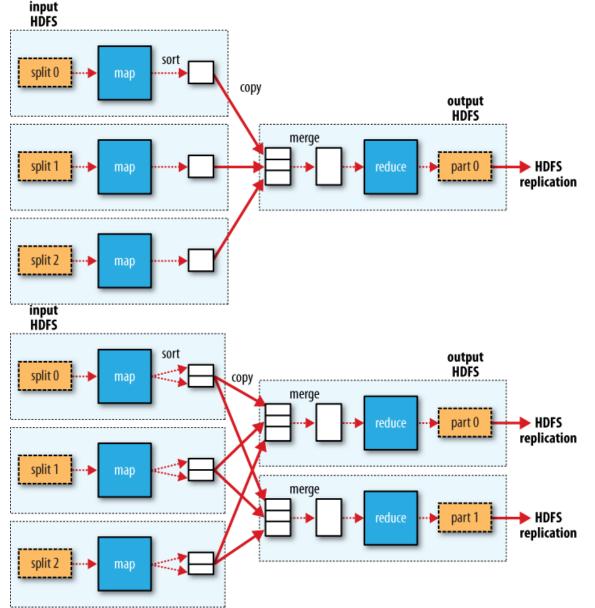
MapReduce structure



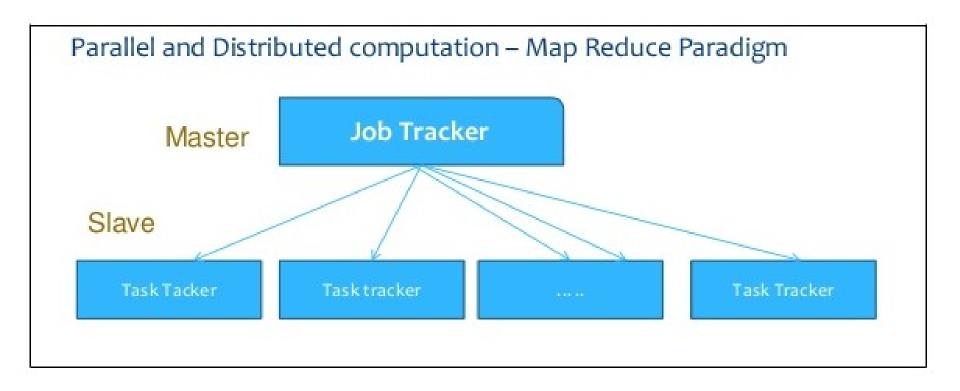
MapReduce structure



MapReduce structure



Job submission (MapReduce v1)



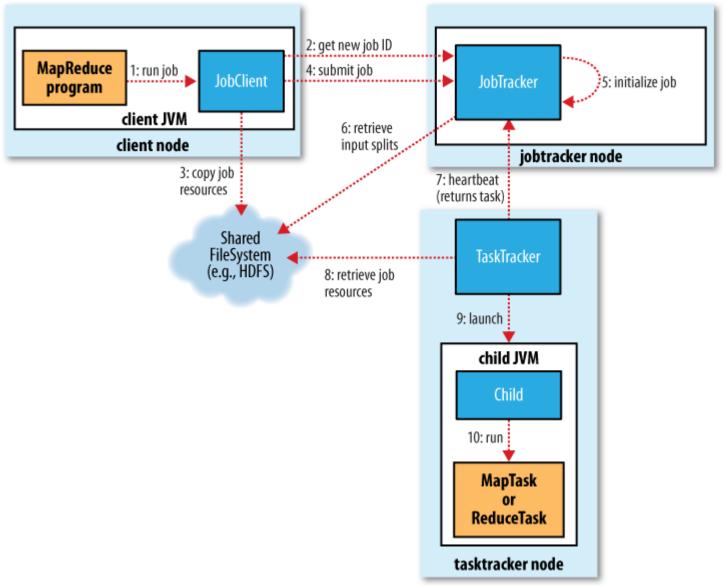
Job submission (MapReduce v1)

- Client applications submit jobs to the Job tracker
- The JobTracker talks to the NameNode to determine the location of the data
- The JobTracker locates TaskTracker nodes with available slots at or near the data
- The JobTracker submits the work to the chosen TaskTracker nodes.
- The TaskTracker nodes are monitored. If they do not submit heartbeat signals often enough, they are deemed to have failed and the work is scheduled on a different TaskTracker

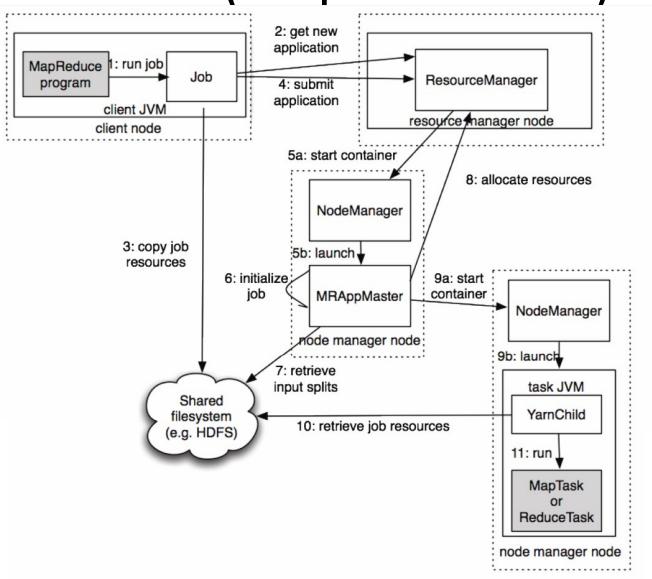
Job submission (MapReduce v1)

- A TaskTracker will notify the JobTracker when a task fails. The JobTracker decides what to do then: it may resubmit the job elsewhere, it may mark that specific record as something to avoid, and it may may even blacklist the TaskTracker as unreliable
- When the work is completed, the JobTracker updates its status
- Client applications can poll the JobTracker for information

Job flow (MapReduce v1)



Job flow (MapReduce v2)



Hadoop Ecosystem

- Apache Avro (serialization system for persistent data)
- Apache Pig (high-level dataflow querying language)
- Apache Hive (data warehouse infrastructure)

Hadoop Ecosystem

- Apache HBase (database for real-time access)
- Apache Sqoop (tool for moving data from SQL to Hadoop or opposite)
- Apache ZooKeeper (distributed coordination service providing high availability) – library for building distributed systems

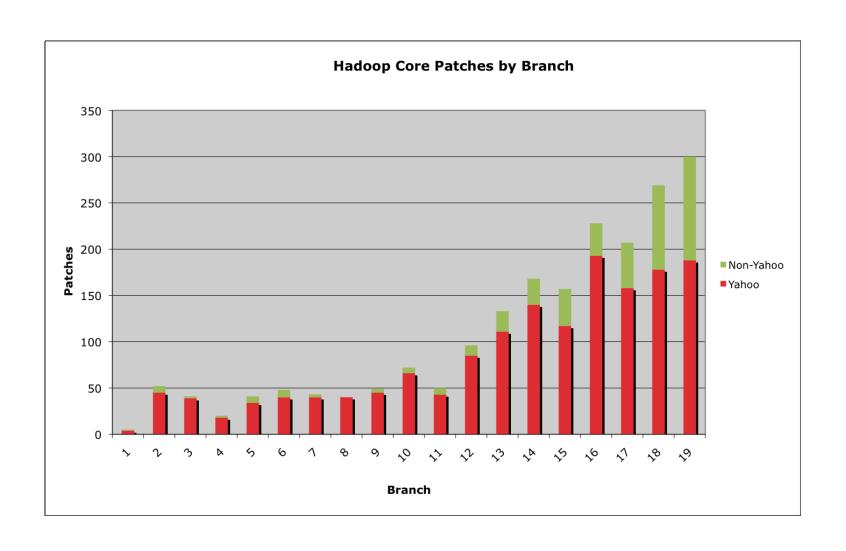
Hadoop exclusive technologies

- YARN Yet Another Resource Negotiator
- HDFS federation possibility of partitioning namespace across several namenodes to support high number of files
- HDFS high-availability techniques for for removing the namenode as the single point of failure

Examples of production use

- Yahoo!: More than 100,000 CPUs in ~20,000 computers running Hadoop; biggest cluster: 2000 nodes (2*4cpu boxes with 4TB disk each); used to support research for Ad Systems and Web Search
- Facebook: To store copies of internal log and dimension data sources and use it as a source for reporting/analytics and machine learning; 320 machine cluster with 2,560 cores and about 1.3 PB raw storage

Size of releases



Hadoop

- + Framework for applications on large clusters
- + Built for commodity hardware
- + Provides reliability and data motion
- + Implements a computational paradigm named Map/Reduce
- + Very own distributed file system (HDFS) (very high aggregate bandwidth across the cluster)
- + Failures handles automatically

Hadoop

- Time consuming development
- Documentation sufficient, but not the most helpful
- HDFS is complicated and has plenty issues of its own
- Debugging a failure is a "nightmare"
- Large clusters require a dedicated team to keep it running properly
- Writing a Hadoop job becomes a software engineering task rather than a data analysis task