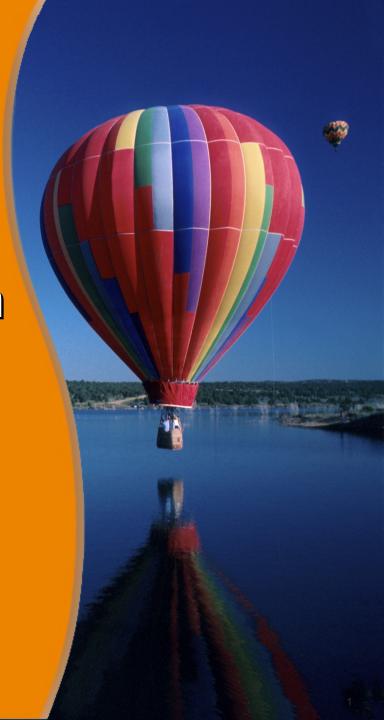
Hadoop Introduction

Sang Shin
JPassion.com
"Learn with Passion!"



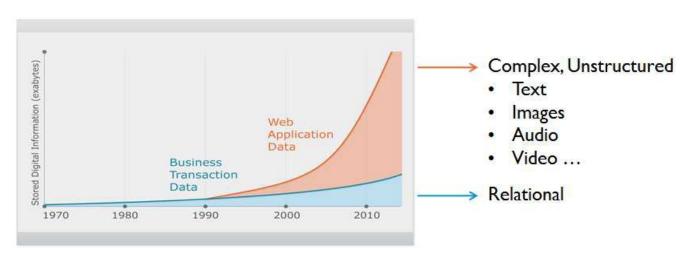
Topics

- Big Data Analytics
- What is and Why Hadoop?
- Comparing Hadoop with other technologies
- Hadoop architecture
- Hadoop ecosystem
- Hadoop usage examples

Big Data Analytics

What is Big Data?

- Big data is a collection of data sets so large and complex that it becomes difficult to process using traditional data processing technologies
 - How do you capture, store, search, share, transfer, analyze, and visualize big data?
- Unstructured data is exploding

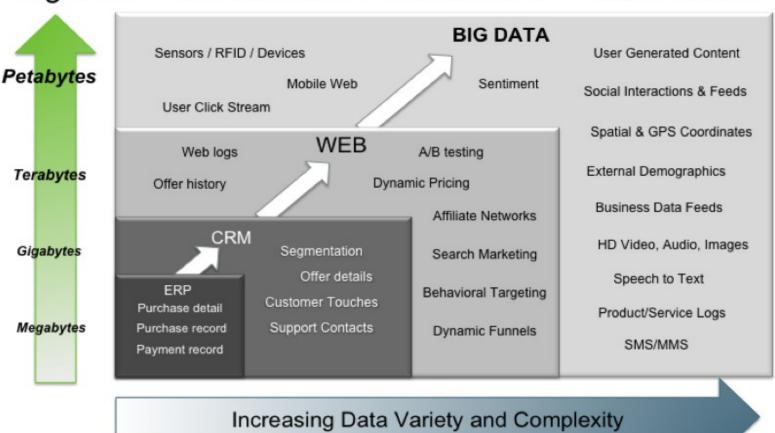


Big Data Examples

- Facebook
 - Has ~400 terabytes of data
 - > ~20 terabytes of new data per day
- New York Stock Exchange (NYSE)
 - Senerated 1 terabyte of trade data per day
- Internet Archive
 - Stores ~2 perabytes of data
 - Solution > Grows at a rate of 20 terabytes per month
 - http://archive.org/web/web.php
- Skybox imaging (satellite images)
 - > 1 terabyte per day
- Ancestry.com
 - > Stores around ~2.5 perabytes of data

Big Data Evolution

Big Data = Transactions + Interactions + Observations



Source: Contents of above graphic created in partnership with Teradata, Inc.

Challenges in Big Data (Using Traditional Data Processing System)

- Slow to process
 - > Takes 11 days to read ~100TB on a single computer
- Slow to move data
 - Moving big data over the network is slow
- Can't scale
 - Scaling up vertically (more memory, more powerful hardware) has limitation
- Hard drive capacity is limited
 - A single hard drive cannot accommodate the size of big data
- Unreliable
 - > Failures in computing devices are inevitable

Big Data Characteristics

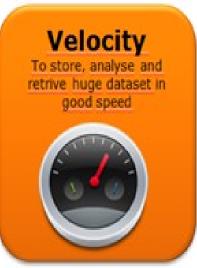
- Very large, distributed aggregations of loosely structured (or unstructured) data
- Petabytes/exabytes of data
- Flat schemas with few complex interrelationships
- Often involving time-stamped events
- Often made up of incomplete data

Challenges in Big Data Handling

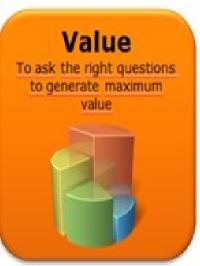
- Moving large amount of data from storage cluster to computation cluster is not feasible (cost and performance wise)
 - > Instead, moving computing logic to the data is more efficient
- In a large cluster, failure is expected
 - Machines, networks break down
- It is expensive to build reliability in each application
 - Reliability should be handled by the framework

Things to Consider in Big Data Analytics









Examples of Public Data Sets

- U.S. Government
 - http://usgovxml.com/
- Amazon
 - http://aws.amazon.com/public-data-sets/
- Weather data from NCDC
 - http://www.ncdc.noaa.gov/data-access
- Million Song Dataset
 - http://labrosa.ee.columbia.edu/millionsong/

What is and Why Hadoop?

What is Hadoop?

- Apache Hadoop is a framework that allows for the distributed processing of large data sets across clusters of commodity computers using a simple programming model
- It is designed to scale up from a single node to thousands of nodes, each providing computation and storage

Hadoop Historical Background

- Started as a sub-project of Apache Nutch
 - > Nutch is for indexing the web and expose it for searching
 - Open source alternative to Google
 - Started by Doug Cutting
- In 2004, Google publishes Google File System (GFS) and Map Reduce framework papers
- Doug Cutting and Nutch team implemented GFS and Map Reduce in Nutch
- In 2006, Yahoo! hires Doug Cutting to work on Hadoop
- In 2008, Hadoop became Apache Top Level project
 - http://hadoop.apache.org

Why is Hadoop?

- Scalable by simply adding new nodes to the cluster
- Economical by using commodity machines
- Efficient by running tasks where data is located
- Flexible by handling schema-less, non-structured data
- Fault tolerant by self-discovering failed nodes and self-healing, by redistributing failed jobs and data reliability by replication
- Simple by using simple programming model
- Evolving by having vibrant ecosystem

Hadoop Design Principles

- Store and process large amounts of data
- Scales on performance, storage, processing
- "Computing" ("code") moves to "Data" (Not the other way around)
- Self-heals recovery from failures is built-in
- Designed to run on commodity hardware

When to use or not to use Hadoop?

- Hadoop is good for
 - Indexing data
 - > Log analysis
 - Image manipulation
 - Sorting large scale data
 - > Data mining
- Hadoop is NOT good for
 - > Real time processing (Hadoop is batch oriented)
 - > Random access (Hadoop is not database)
 - Computation-intensive tasks with little data
- Some limitations of Hadoop are addressed by Hadoop ecosystem technologies, however

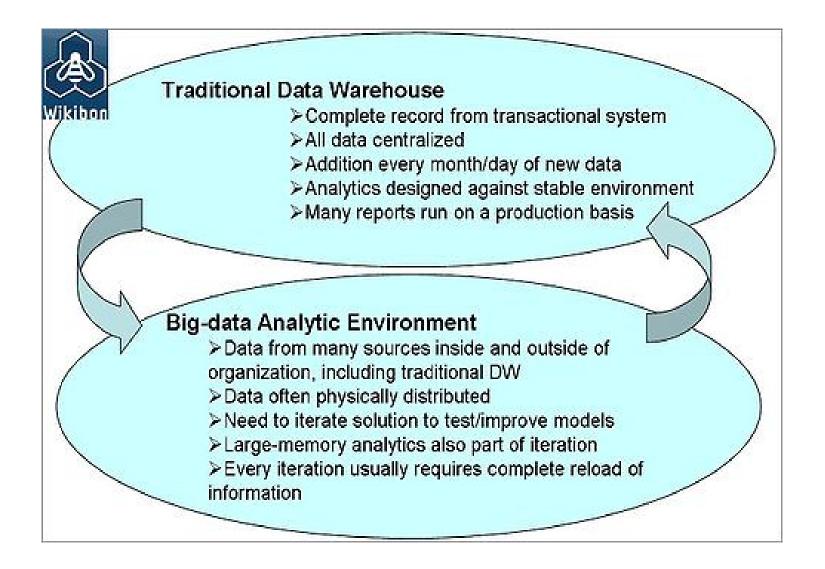
Comparing Hadoop with Other Technologies

RDBMS vs. Hadoop

- Small data (comp. to Hadoop)
- Strictly structured
- Schema on write
- Relational
- Transaction-oriented
- ACID
- Real-time query
- Frequent write/update
- Centralized
- Limited scalability (vertical)

- Big data
- Loosely structured
- Schema on read
- Non relational
- Analytics-oriented
- Non-ACID
- Batch-oriented
- Initial write, No update/delete
- Distributed
- High scalability (horizontal)

Data Warehouse vs. Hadoop



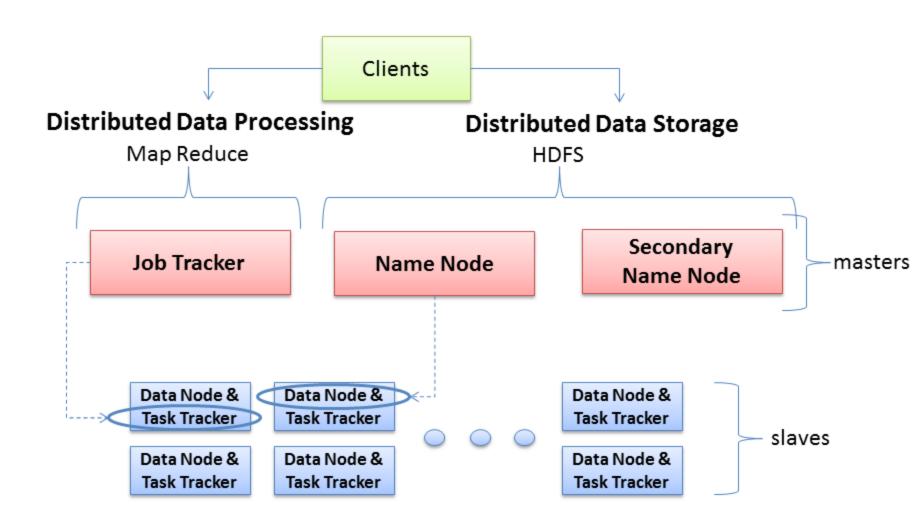
Hadoop Architecture

Hadoop Architecture

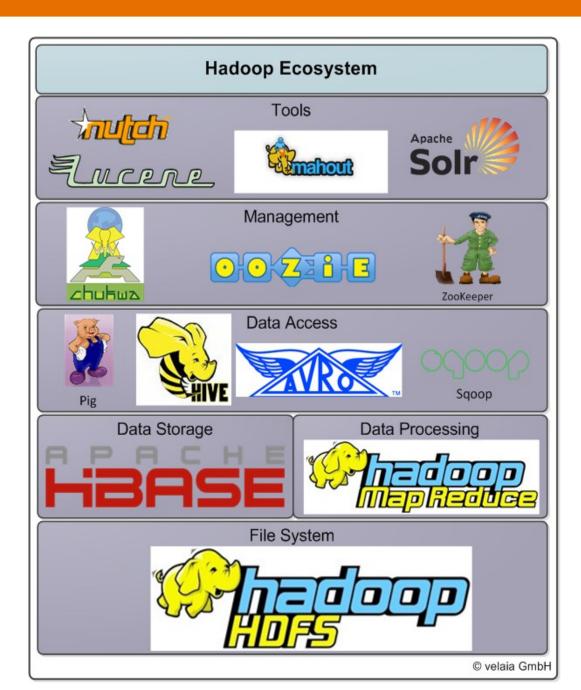
- Hadoop is built on two core technologies: "Map Reduce" for processing and "HDFS" for distributed file system
 - Map Reduce is a framework for performing high performance distributed data processing using the "divide and aggregate" programming paradigm
 - > HDFS is a reliable distributed file system that provides highthroughput access to data
- Hadoop has master/slave architecture for both "Map Reduce" and "HDFS"
 - Map Reduce has "Job Tracker" as a master and multiple "Task Trackers" as slaves
 - HDFS has "Name node" as a master and multiple "Data nodes" as slaves

Hadoop Architecture

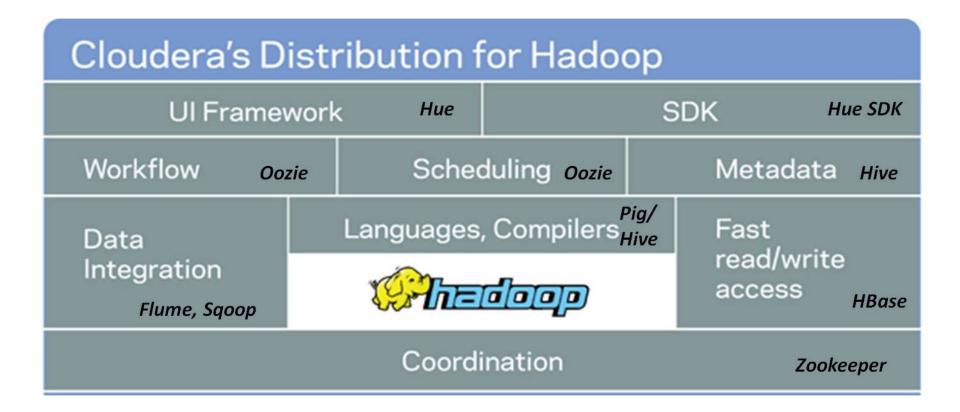
Hadoop = Map Reduce + HDFS



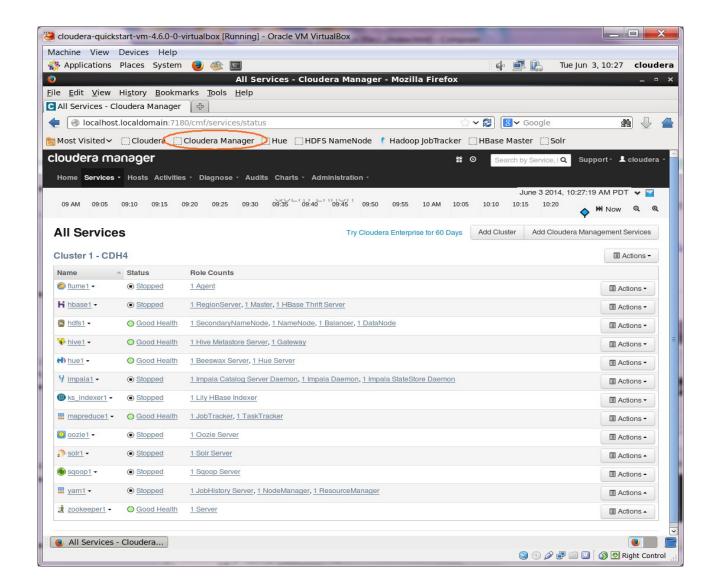
Hadoop Ecosystem



Cloudera Hadoop Distribution



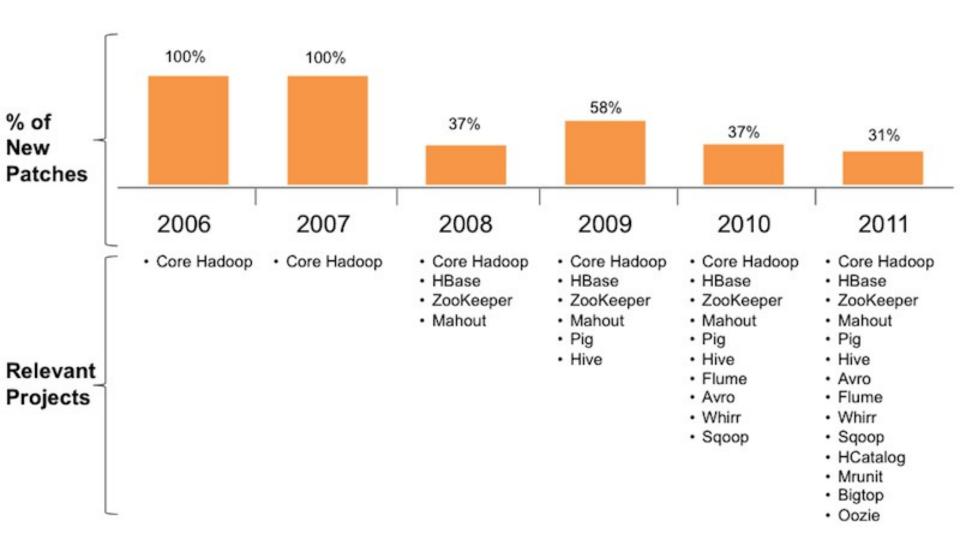
Cloudera Manager



For CDH 4.x.x: Automatically Started

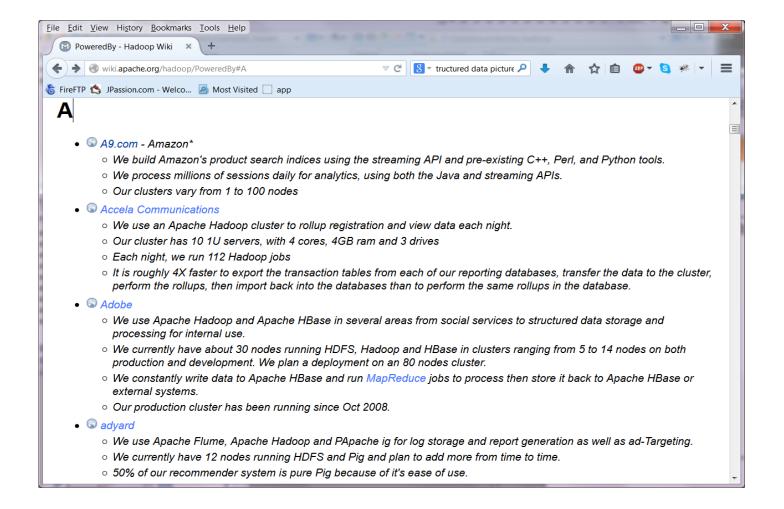
For CDH 5.x.x Needs to be started manually

Hadoop Ecosystem Development



Hadoop Usage Examples

http://wiki.apache.org/hadoop/PoweredBy



Facebook

- We use Apache Hadoop to store copies of internal log and dimension data sources and use it as a source for reporting/analytics and machine learning.
- > Currently we have 2 major clusters:
 - A 1100-machine cluster with 8800 cores and about 12 PB raw storage
 - > A 300-machine cluster with 2400 cores and about 3 PB raw storage
 - > Each (commodity) node has 8 cores and 12 TB of storage
- > We are heavy users of both streaming as well as the Java APIs
- We have built a higher level data warehousing framework using Hive
- > We have also developed a FUSE implementation over HDFS.

- EBay
 - > 532 nodes cluster (8 * 532 cores, 5.3PB).
 - Heavy usage of Java MapReduce, Apache Pig, Apache Hive, Apache HBase
 - Using it for Search optimization and Research

Twitter

- We use Apache Hadoop to store and process tweets, log files, and many other types of data generated across Twitter
- > We store all data as compressed LZO files.
- We use both Scala and Java to access Hadoop's MapReduce APIs
- We use Apache Pig heavily for both scheduled and ad-hoc jobs, due to its ability to accomplish a lot with few statements.
- We employ committers on Apache Pig, Apache Avro, Apache Hive, and Apache Cassandra, and contribute much of our internal Hadoop work to opensource (see hadoop-lzo)

- Yahoo
 - > More than 100,000 CPUs in >40,000 computers running Hadoop
 - Our biggest cluster: 4500 nodes (2*4cpu boxes w 4*1TB disk & 16GB RAM)
 - Used to support research for Ad Systems and Web Search
 - Also used to do scaling tests to support development of Apache Hadoop on larger clusters
 - > Our Blog Learn more about how we use Apache Hadoop.
 - > >60% of Hadoop Jobs within Yahoo are Apache Pig jobs.

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