

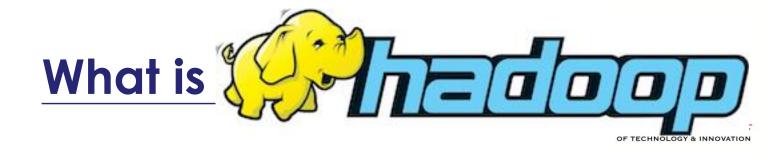
Data Management

Topic 7 – Hadoop

Topic & Structure of Lesson



- What is Hadoop?
- Hadoop Framework
- Hadoop's Architecture
- Hadoop in the Wild
- Data warehouse to Hadoop

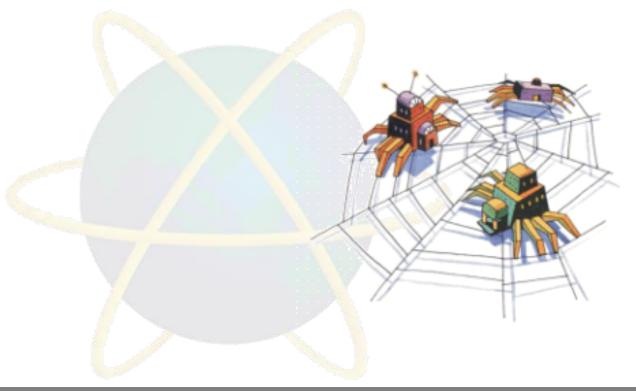


- Apache top level project, open-source implementation of frameworks for reliable, scalable, distributed computing, and data storage.
- It is a flexible and highly-available architecture for large scale computation and data processing on a network of commodity hardware.

Brief History of Hadoop

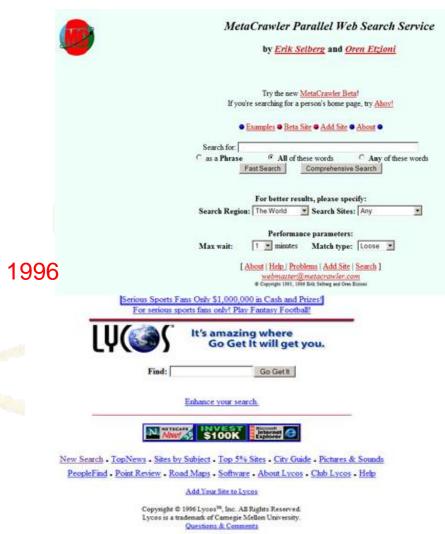


Designed to answer the question:
 "How to process big data with reasonable cost and time?"



Search engines in 1990s







1996



1996

1997

Google search engine







Copyright @1998 Google Inc.

1998

Google

2013

Google Search I'm Feeling Lucky

Google Origins



The Google File System

2003

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

(fay.jeff,sanjay,wilsonh,kerr,m3b,tushar,fikes,gruber)@google.com

Google, Inc.



igitable is a distributed storage system for managing nured data that is designed to scale to a very large petabytes of data across thousands of commodity is scaled to the state of the scale to a very large polyector of the scale of the scale of the scale of the ding web indexing, Google tore data in Bigitable, tiding web indexing, Google Earth, and Google Fice. These applications place very different demands ignable, both in terms of data size (from URLs to cases to satellite imasser) and better consistence of the constitution of the scale of t achieved scalability and high performance, but Big provides a different interface than such systems. Big does not support a full relational data model; inset provides clients with a simple data model that sup dynamic control over data layout and format, an lows clients to reason about the locality properties o data represented in the underlying storage. Data i dexed using row and column names that can be arbistrings. Bigtable also treats data as uninterpreted str

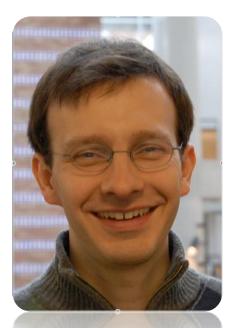


Hadoop's Developers





Doug Cutting



Michael J. Cafarella



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.



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
- 2015 Hadoop Yarn
- 2017 Hadoop 3.0 (Alpha)

What is Hadoop?

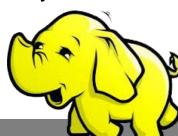


Hadoop:

 an open-source software framework that supports dataintensive distributed applications, licensed under the Apache v2 license.

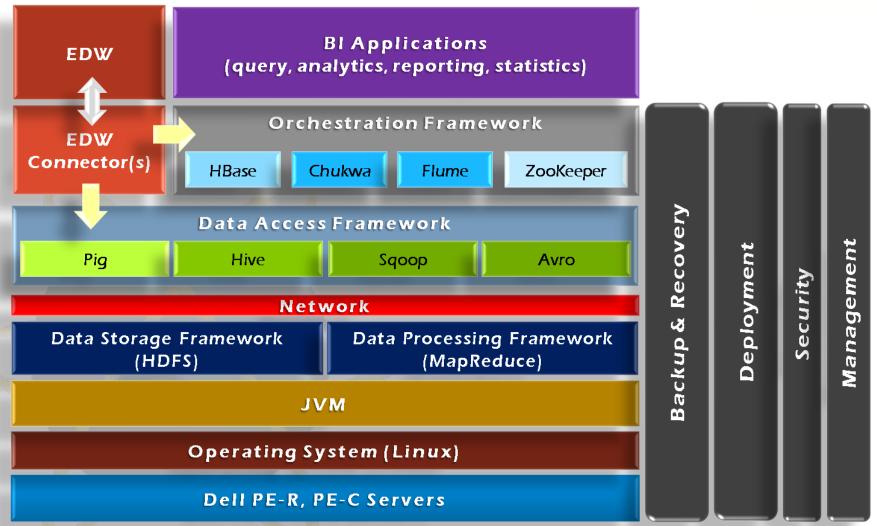
Goals / Requirements:

- Abstract and facilitate the storage and processing of large and/or rapidly growing data sets
 - Structured and non-structured data
 - Simple programming models
- High scalability and availability
- Use commodity (cheap!) hardware with little redundancy
- Fault-tolerance
- Move computation rather than data



Hadoop Framework Tools

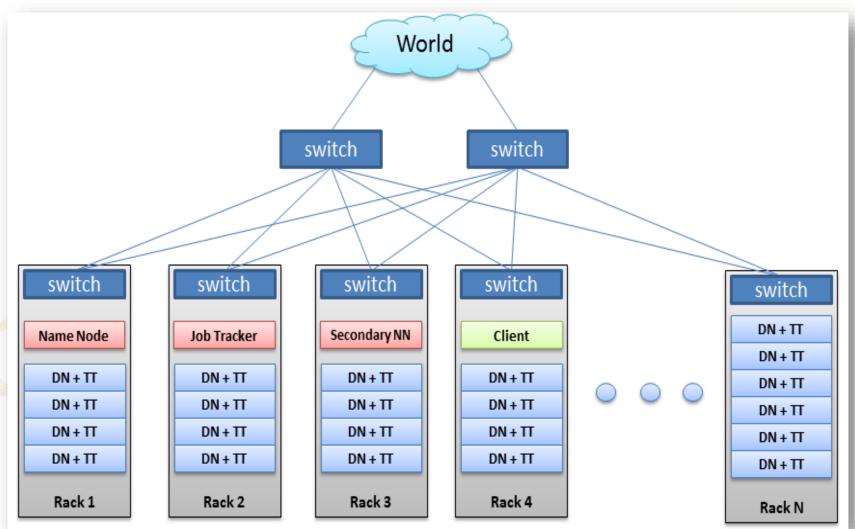






- Distributed, with some centralization
- Main nodes of cluster are where most of the computational power and storage of the system lies
- Main nodes run TaskTracker to accept and reply to MapReduce tasks, and also DataNode to store needed blocks closely as possible
- Central control node runs NameNode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







- <u>H</u>adoop <u>D</u>istributed <u>F</u>ilesystem
- 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

Title of Slides



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



DataNode:

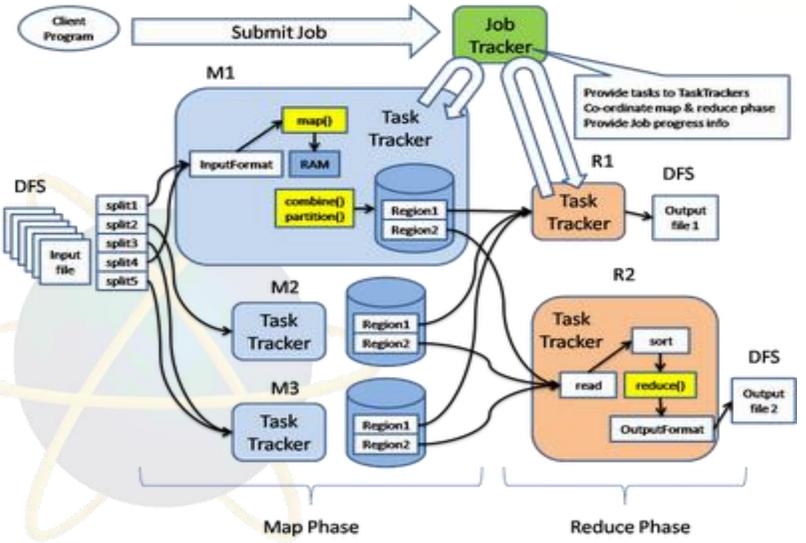
- Stores the actual data in HDFS
- Can run on any underlying filesystem (ext3/4, NTFS, etc)
- Notifies NameNode of what blocks it has
- NameNode replicates blocks 2x in local rack, 1x elsewhere

Module Code and Module Title

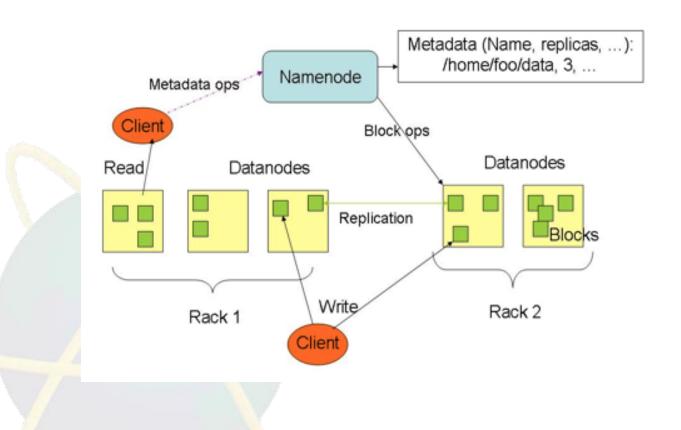
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Hadoop's Architecture: MapReduce Engine





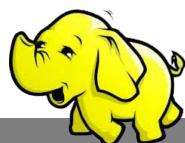






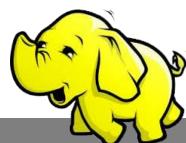
MapReduce Engine:

- JobTracker & TaskTracker
- JobTracker splits up data into smaller tasks("Map") and sends it to the TaskTracker process in each node
- TaskTracker reports back to the JobTracker node and reports on job progress, sends data ("Reduce") or requests new jobs





- None of these components are necessarily limited to using HDFS
- Many other distributed file-systems with quite different architectures work
- Many other software packages besides Hadoop's MapReduce platform make use of HDFS





- 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)



- Non-realtime large dataset computing:
 - NY Times was dynamically generating PDFs of articles from 1851-1922
 - Wanted to pre-generate & statically serve articles to improve performance
 - Using Hadoop + MapReduce running on EC2 / S3,
 converted 4TB of TIFFs into 11 million PDF articles in 24 hrs

Hadoop in the Wild: Facebook

Messages



- Design requirements:
 - Integrate display of email, SMS and chat messages between pairs and groups of users
 - Strong control over who users receive messages from
 - Suited for production use between 500 million people immediately after launch
 - Stringent latency & uptime requirements

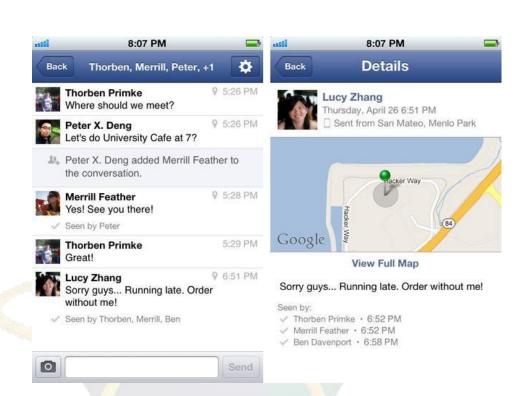




System requirements



- Cheap, elastic storage
- Low latency
- High consistency (within a single data center good enough)
- Disk-efficient sequential and random read performance





- Classic alternatives
 - These requirements typically met using large MySQL cluster & caching tiers using Memcached
 - Content on HDFS could be loaded into MySQL or Memcached if needed by web tier
- Problems with previous solutions
 - MySQL has low random write throughput... BIG problem for messaging!
 - Difficult to scale MySQL clusters rapidly while maintaining performance
 - MySQL clusters have high management overhead, require more expensive hardware



- Facebook's solution
 - Hadoop + HBase as foundations
 - Improve & adapt HDFS and HBase to scale to FB's workload and operational considerations
 - Major concern was availability: NameNode is SPOF & failover times are at least 20 minutes
 - Proprietary "AvatarNode": eliminates SPOF, makes HDFS safe to deploy even with 24/7 uptime requirement
 - Performance improvements for realtime workload: RPC timeout. Rather fail fast and try a different DataNode

Hadoop Highlights



- Distributed File System
- Fault Tolerance
- Open Data Format
- Flexible Schema
- Queryable Database

Module Code and Module Title

Title of Slides

Why use Hadoop?



- Need to process Multi Petabyte Datasets
- Data may not have strict schema
- Expensive to build reliability in each application
- Nodes fails everyday
- Need common infrastructure
- Very Large Distributed File System
- Assumes Commodity Hardware
- Optimized for Batch Processing
- Runs on heterogeneous OS

DataNode



- A Block Sever
 - Stores data in local file system
 - Stores meta-data of a block checksum
 - Serves data and meta-data to clients
- Block Report
 - Periodically sends a report of all existing blocks to NameNode
- Facilitate Pipelining of Data
 - Forwards data to other specified DataNodes

Block Placement



- Replication Strategy
 - One replica on local node
 - Second replica on a remote rack
 - Third replica on same remote rack
 - Additional replicas are randomly placed
- Clients read from nearest replica

Data Correctness



- Use Checksums to validate data CRC32
- File Creation
 - Client computes checksum per 512 byte
 - DataNode stores the checksum
- File Access
 - Client retrieves the data and checksum from DataNode
 - If validation fails, client tries other replicas

Data Pipelining



- Client retrieves a list of DataNodes on which to place replicas of a block
- Client writes block to the first DataNode
- The first DataNode forwards the data to the next DataNode in the Pipeline
- When all replicas are written, the client moves on to write the next block in file

Hadoop MapReduce



- MapReduce programming model
 - Framework for distributed processing of large data sets
 - Pluggable user code runs in generic framework
- Common design pattern in data processing
 - cat * | grep | sort | uniq -c | cat > file
 - input | map | shuffle | reduce | output

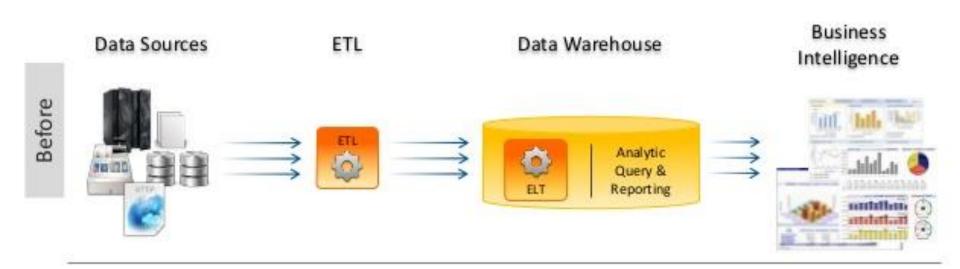
MapReduce Usage

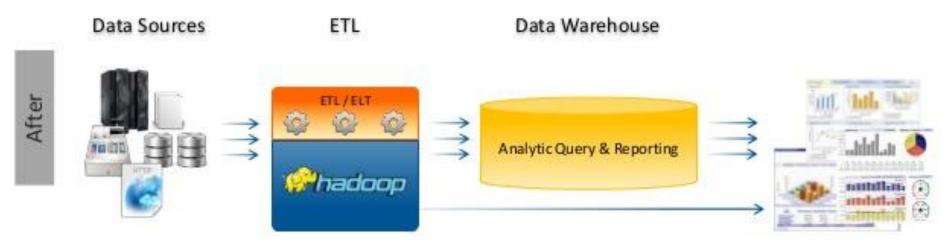


- Log processing
- Web search indexing
- Ad-hoc queries



Offloading The Data Warehouse to Hadoop





Question & Answer Session



