Hadoop: Past and Present

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About Me

- 2010: BS CS UVa
- 2012: MS CS UC Berkeley
 - AMP Lab alumni
 - Advised by Ion Stoica
- Now: HDFS team at Cloudera



Outline

 State of databases in 1999 Why is Hadoop displacing DB technology? Core stack HDFS and MapReduce Rest of the Hadoop ecosystem • HBase, Pig, Hive, Oozie, Zookeeper, Flume, Impala, ...







Search the we	
Google Search	I'm feeling lucky
More G	boogle!
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Indexing the Web

• Web is huge

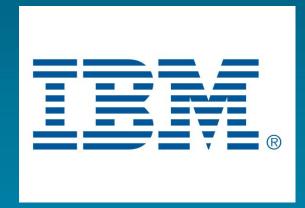
- Hundreds of millions of pages in 1999
- How do you index it?
 - Crawl all the pages
 - Rank pages based on relevance metrics
 - Build search index of keywords to pages
 - Do it in real-time!













Databases in 1999

- 1. Buy a really big machine
- 2. Install an expensive DBMS on it
- 3. Point your workload at it
- 4. Hope it doesn't fail
- 5. Ambitious: buy another really big machine as a backup



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Example: When precisely will the new millennium begin?				
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Database Limitations

 Didn't scale horizontally High marginal cost (\$\$\$) No real fault-tolerance story Vendor lock in (\$\$\$) SQL unsuited for search ranking Complex analysis (PageRank) Unstructured data



Google Does Something Different

Designed their own storage and processing infrastructure
Google File System and MapReduce
Goals: KISS

- Cheap
- Scalable
- Reliable



Google Does Something Different

- It worked!
- Powered Google Search for many years
- General framework for large-scale batch computation tasks
- Still used internally at Google to this day



The Original Google Storage

In 1996 Larry Page and Sergey Brin, then PhD students in Stanford CSD, working on the Digital Library Project, needed a large amount of diskspace to test their Pagerank ^{is} algorithm on actual worldwide-web data. At that time 4 GigaByte hard disks were the largest available, so they assembled 10 of these drives into a low-cost cabinet.

In Nov 1999, Google Inc, by then operating one of the primary search engines on the web, provided replacement storage capacity to the Digital Library project so that we could move this original storage assembly into our history displays.

As of September 2000, Google, now located in Mountain View, operated 5000 PCs for searching and web crawling, using the LINUX operating system.

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Google's messages from the future

Google was benevolent enough to publish

- 2003: Google File System (GFS) paper
- 2004: MapReduce paper
- Already mature technologies at this point



Google's messages from the future

Community didn't get it immediately

- DB people thought it was silly
- Non-Google weren't at the same scale yet
- Google had little interest in releasing GFS and MapReduce
 - Business was ads, not infrastructure



Birth of Hadoop

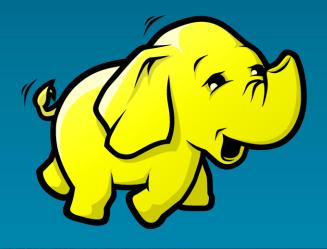
- Doug Cutting and Mike Cafarella
- Nutch
 - Open-source search platform
- Ran into scaling issues
 - 4 nodes
 - Hard to program
 - Hard to manage

Immediate application for GFS and MR



Birth of Hadoop

- 2004-2006: Implemented GFS/MR and ported Nutch to it
- 2006: Spun out into Apache Hadoop
- Name of Doug's son's stuffed elephant





Birth of Hadoop





Summary

The web is huge and unstructured

- Databases didn't fit the problem
 - Didn't scale, expensive, SQL limitations
- Google did their own thing: GFS + MR
- Hadoop is based on the Google papers





HDFS and MapReduce

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HDFS

Based on GFS

- Distributed, fault-tolerant filesystem
- Primarily designed for cost and scale
 - Works on commodity hardware
 - 20PB / 4000 node cluster at Facebook



HDFS design assumptions

• Failures are common

- Massive scale means more failures
- Disks, network, node
- Files are append-only
- Files are large (GBs to TBs)
- Accesses are large and sequential



Quick primers

- Filesystems
- Hard drives
- Datacenter networking



Quick filesystem primer

- Same concepts as the FS on your laptop
 - Directory tree
 - Create, read, write, delete files
- Filesystems store metadata and data
 - Metadata: filename, size, permissions, ...
 - Data: contents of a file
- Other concerns
 - Data integrity, durability, management



Quick disk primer

- Disk does a seek for each I/O operation
- Seeks are expensive (~10ms)
- Throughput / IOPS tradeoff
 - 100 MB/s and 10 IOPS
 - 10MB/s and 100 IOPS
- Big I/Os mean better throughput



Quick networking primer

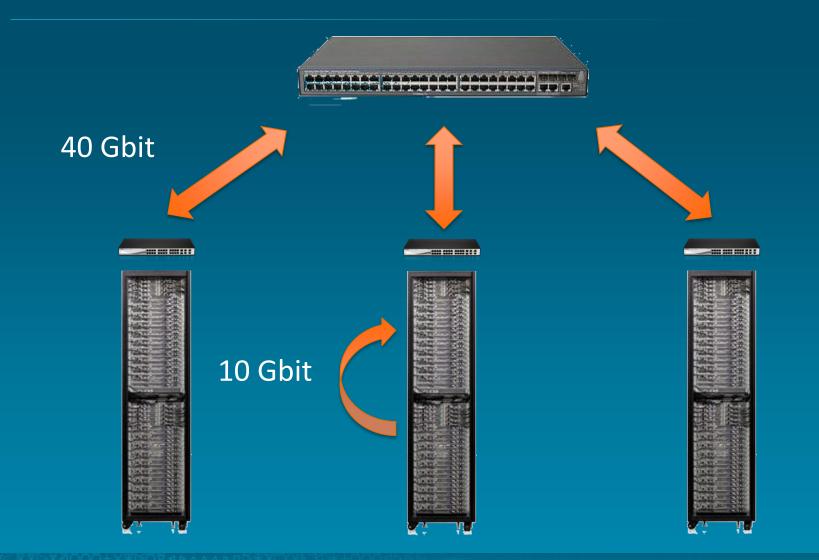


Top-of-rack switch



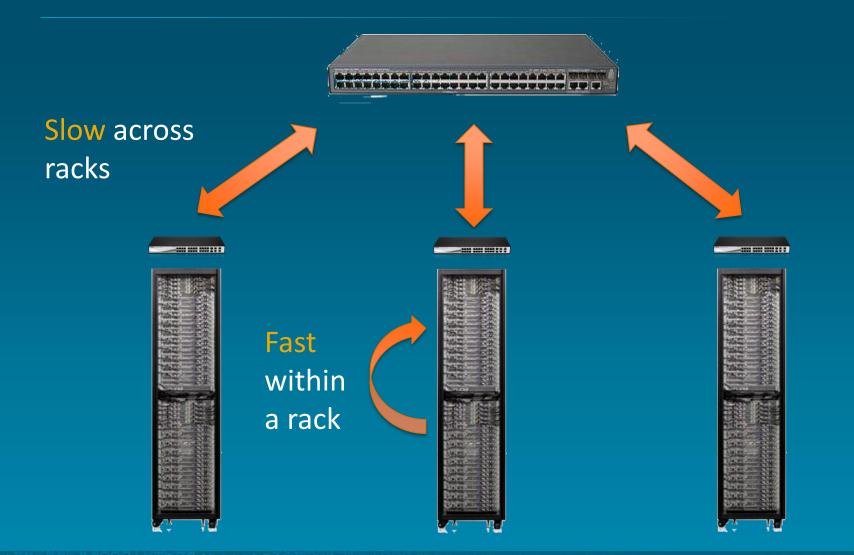


Quick networking primer





Quick networking primer





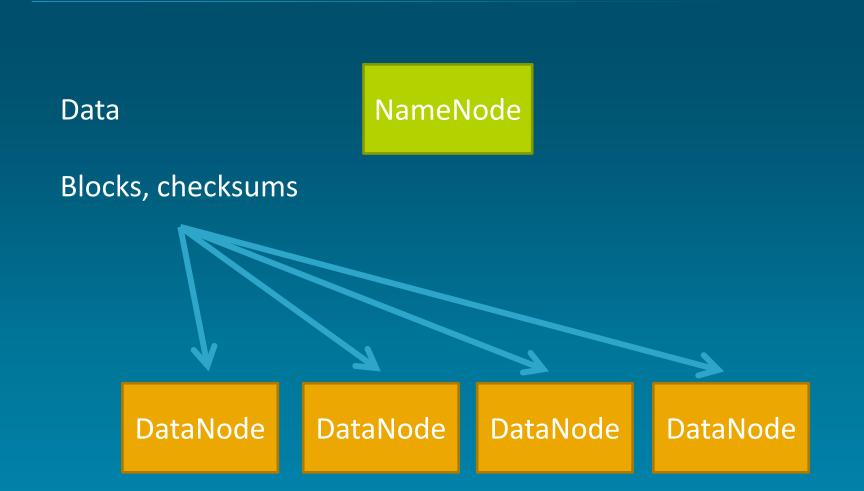
Metadata



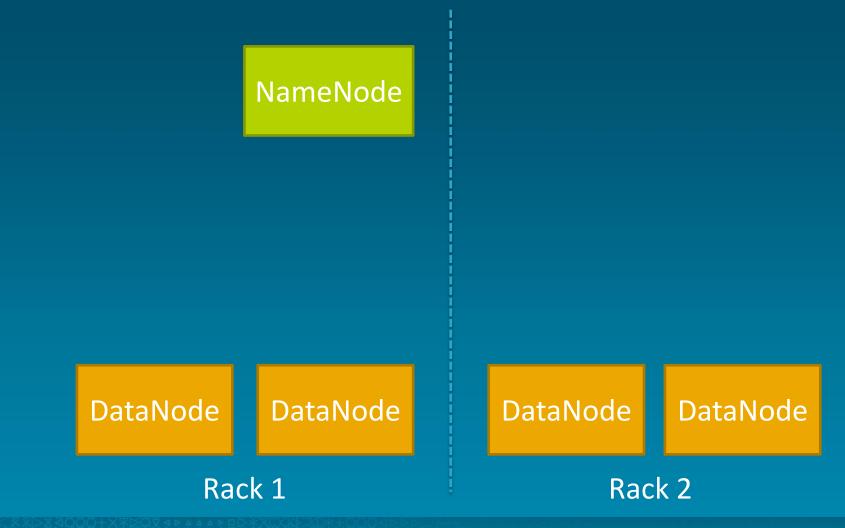
Paths, filenames, file sizes, block locations, ...



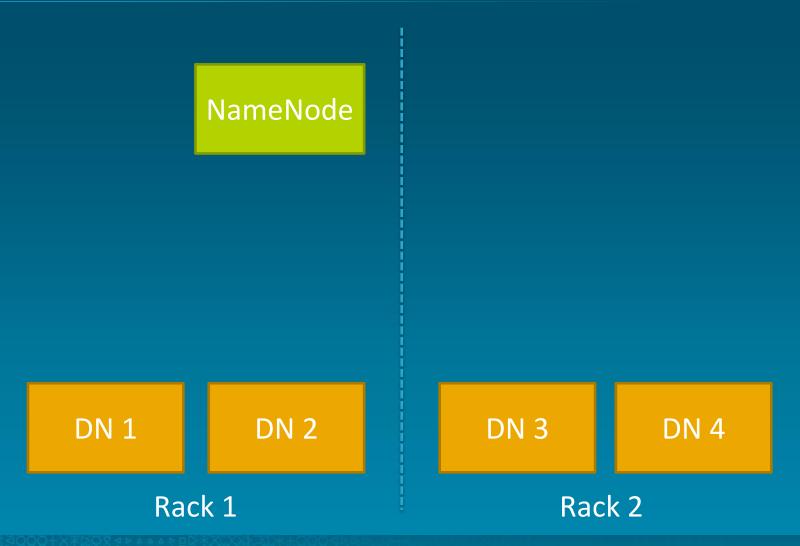






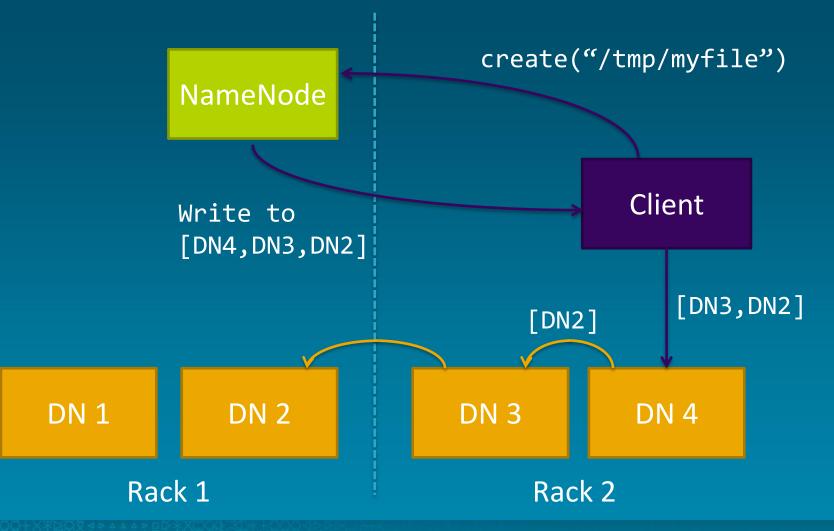






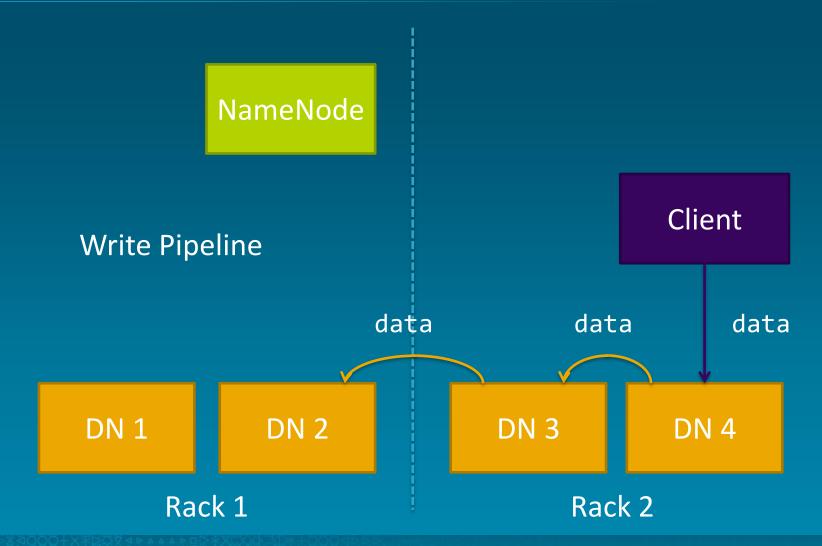


HDFS Write Path



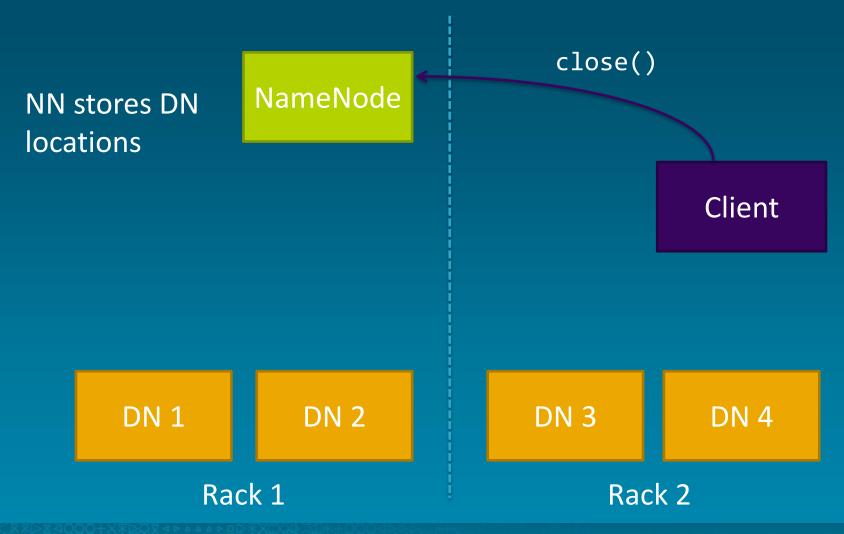


HDFS Write Path





HDFS Write Path





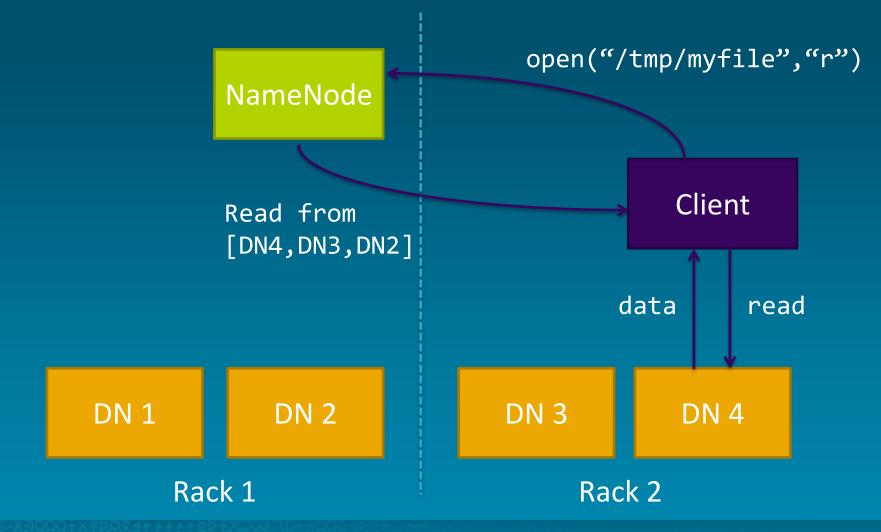
HDFS Write Path

Talk to NameNode

- Store metadata for new file
- Get topology-aware list of DataNodes
- Setup the write pipeline
- Stream data to pipeline
- Tell NameNode when done



HDFS Read Path





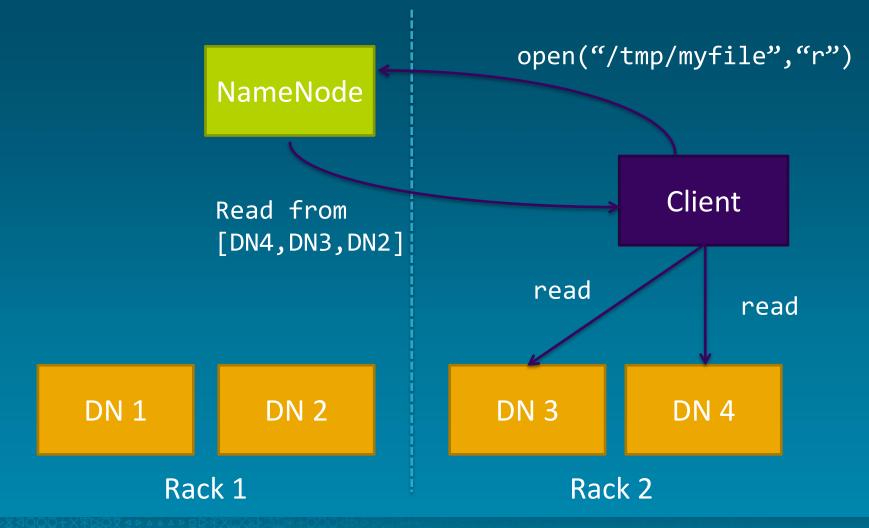
HDFS Fault-tolerance

Many different failure modes

- Disk corruption, node failure, switch failure
- Primary concern
 - Data is safe!!!
- Secondary concerns
 - Keep accepting reads and writes
 - Do it transparently to clients

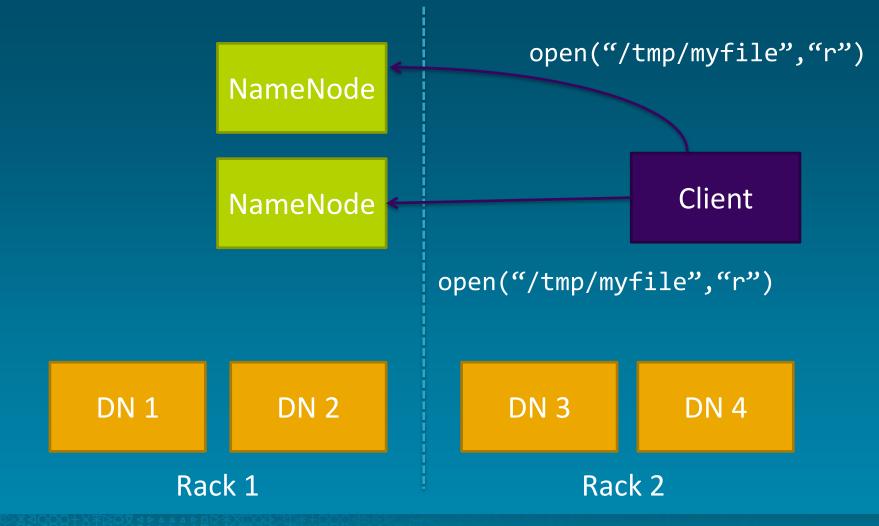


HDFS DataNode Failure





HDFS NameNode Failure





Other HDFS features

- NameNode federation
- Storage block pools
- Snapshots (new!)
- Future
 - Hierarchical storage management
 - Quality-of-Service
 - NameNode and DataNode scalability



MapReduce

Programming and execution framework

- Taken from functional programming
 - Map operate on every element
 - Reduce combine and aggregate results
- Abstracts storage, concurrency, execution
 - Just write two Java functions
 - Contrast with MPI

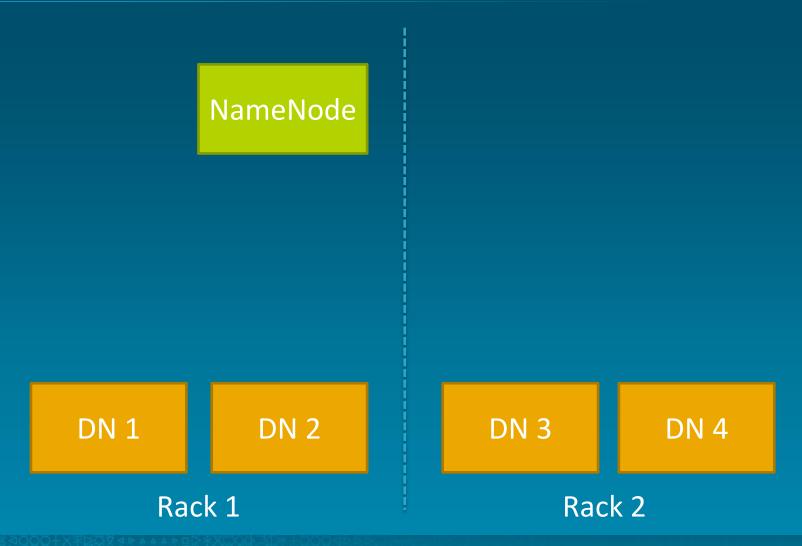


MapReduce

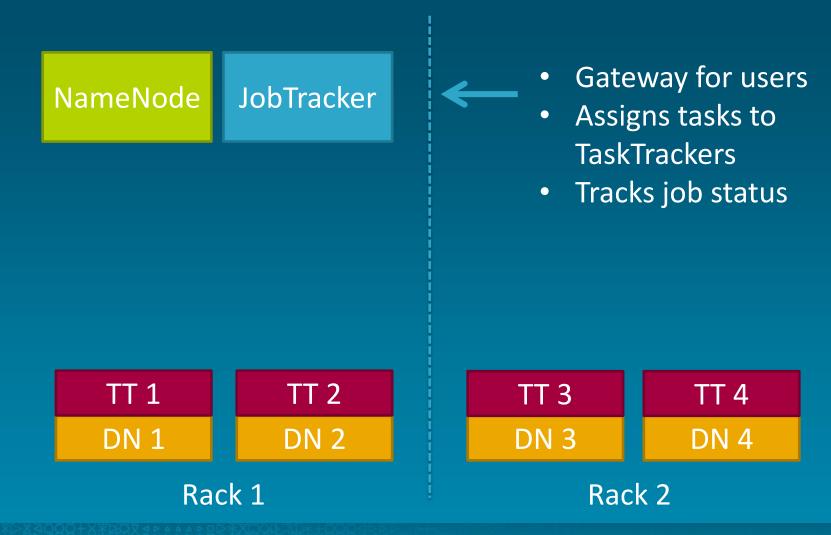
Constrained, but general

- Can do custom ML not possible in SQL
- Not as efficient as a DB for some queries
- No update in place
 - Take data in, transform, write new data out
 - Makes fault-tolerance easier

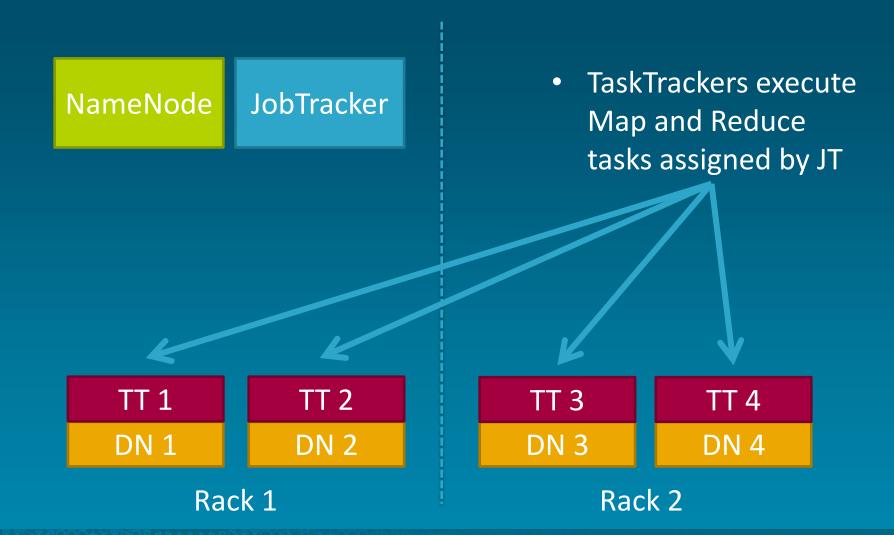






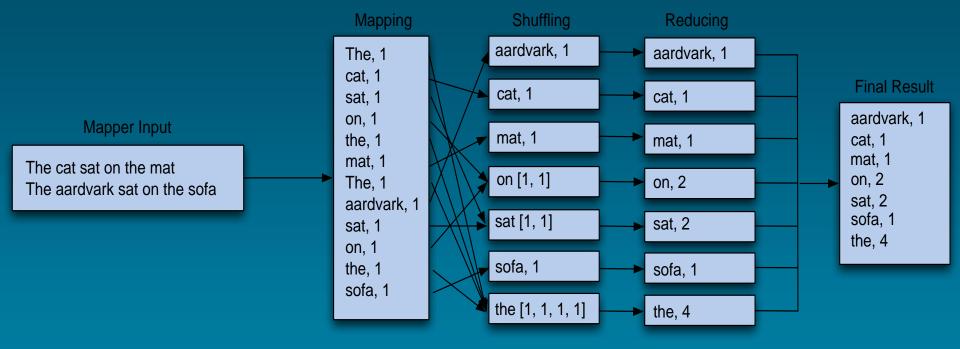




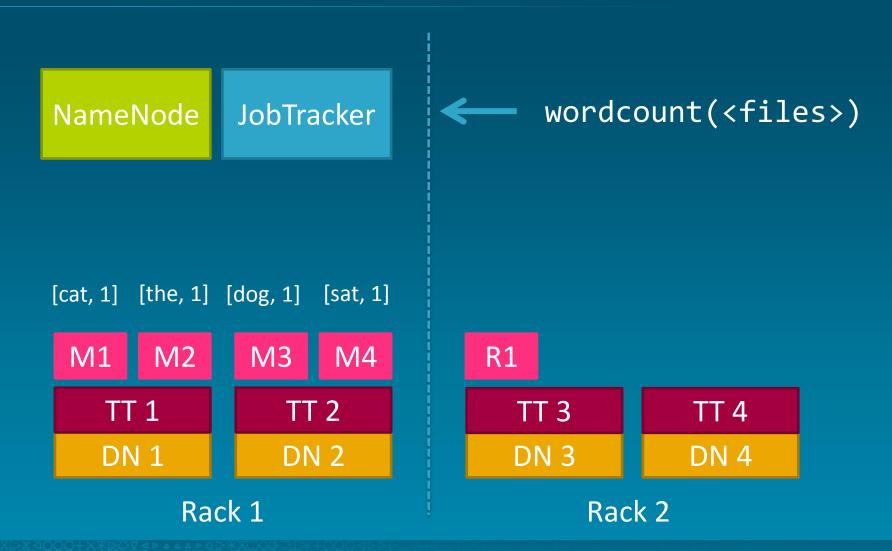




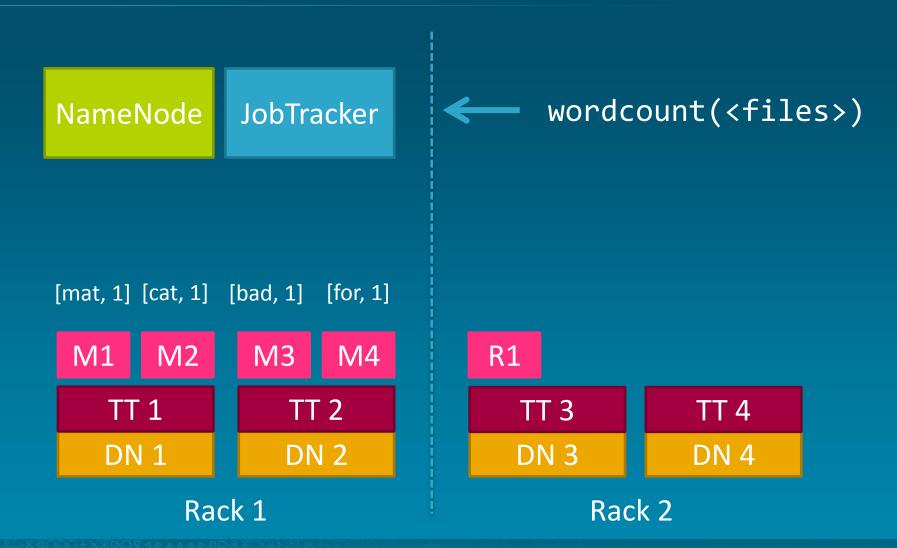
Word Count Example



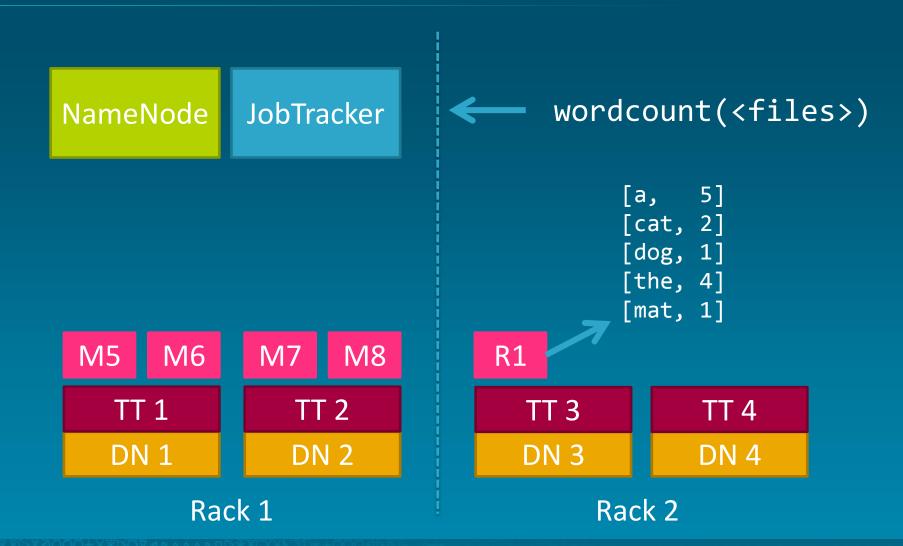




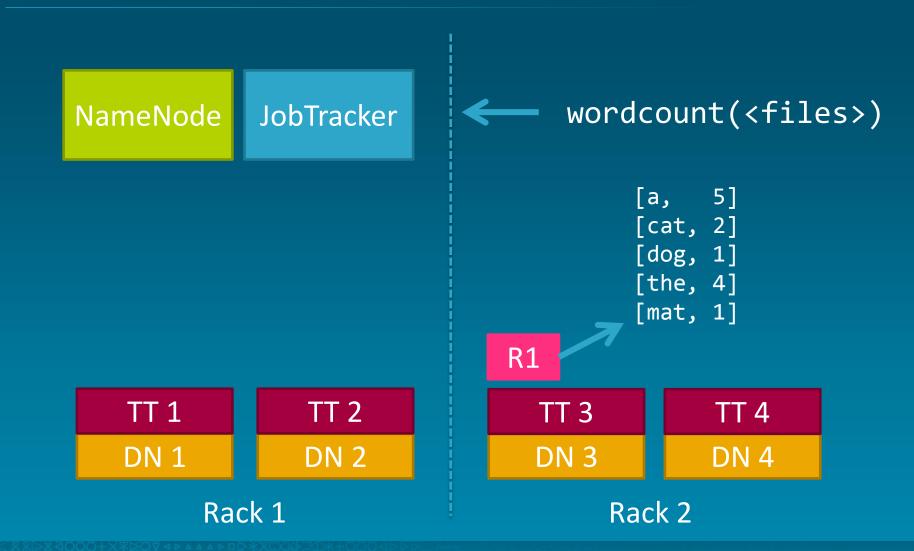




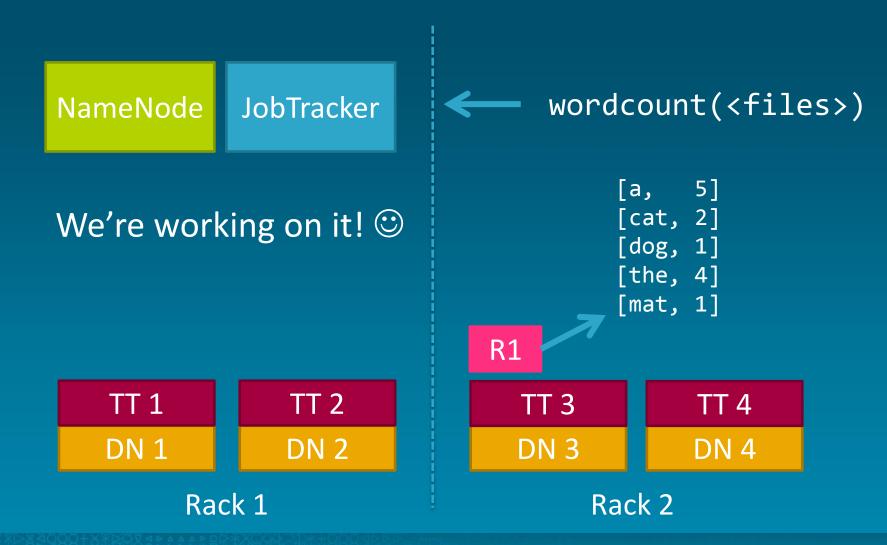














Summary

GFS and MR co-design

- Cheap, simple, effective at scale
- Fault-tolerance baked in
 - Replicate data 3x
 - Incrementally re-execute computation
 - Avoid single points of failure
- Held the world sort record (0.578TB/min)



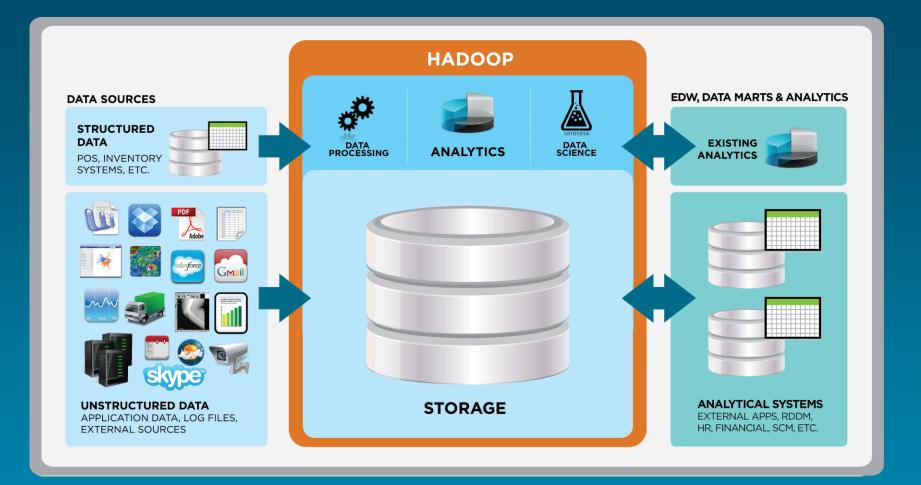


Hadoop ecosystem

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Data Processing Pipeline

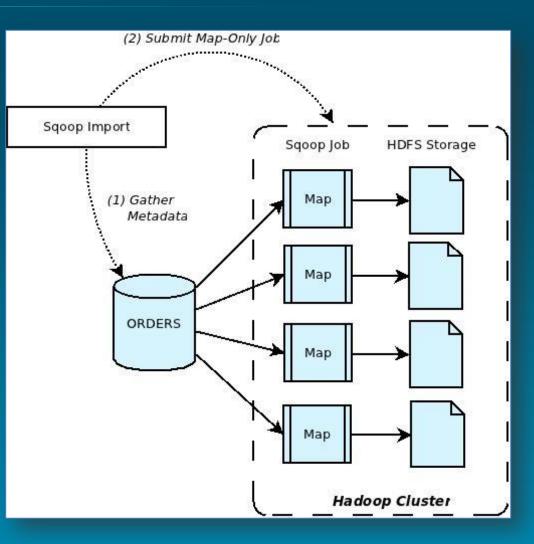




Sqoop



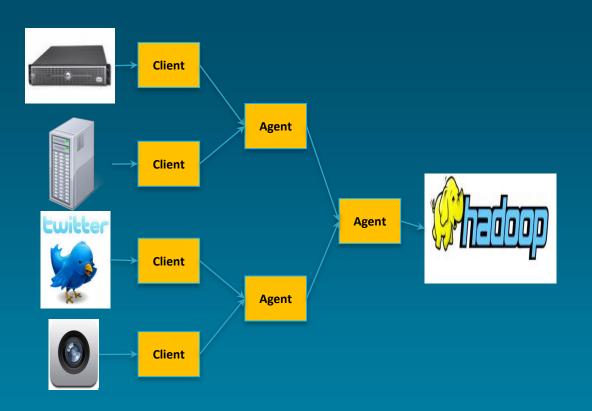
Performs bidirectional data transfers between Hadoop and almost any SQL database with a JDBC driver





Flume

A streaming data collection and aggregation system for massive volumes of data, such as RPC services, Log4J, Syslog, etc.





Hive



- Relational database
 abstraction using a SQL like
 dialect called HiveQL
- Statements are executed as one or more MapReduce Jobs

SELECT s.word, s.freq, k.freq FROM shakespeare JOIN ON (s.word= k.word) WHERE s.freq >= 5;

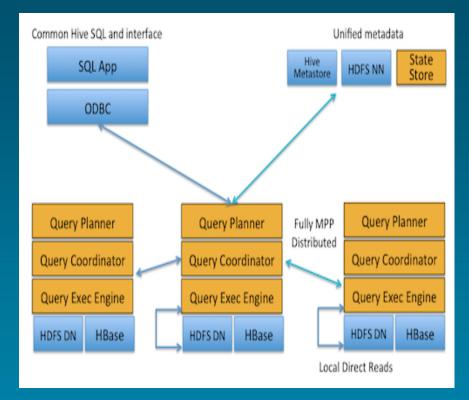


Impala

Modern MPP database built on top of HDFS

Really fast! Written in C++

10-100x faster than Hive





Pig



- High-level scripting language for for executing one or more MapReduce jobs
- Created to simplify authoring of MapReduce jobs
- Can be extended with user defined functions

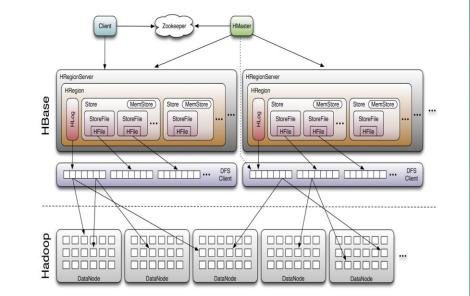
emps = LOAD 'people.txt' AS
(id,name,salary);
rich = FILTER emps BY salary >
200000;
sorted_rich = ORDER rich BY
salary DESC;
STORE sorted_rich INTO
'rich_people.txt';



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HBase

- Low-latency, distributed, columnar key-value store
- Based on BigTable
- Efficient random reads/writes on HDFS
- Useful for frontend applications



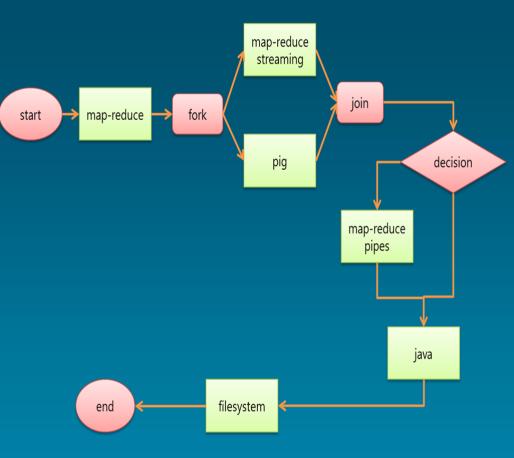


cloudera

Oozie



A workflow engine and scheduler built specifically for large-scale job orchestration on a Hadoop cluster





Hue

- Hue is an open source web-based application for making it easier to use Apache Hadoop.
- Hue features
 - File Browser for HDFS
 - Job Designer/Browser for MapReduce
 - Query editors for Hive, Pig and Cloudera Impala
 - Oozie

Hue 💱 💈 🔌	🕼 🧕 🖩 😌 🔹
Query Editor My Queries	Saved Queries History Tables Settings
HIVE SETTINGS	Hive Query
FILE RESOURCES	Query
Add	select * from mydata;
Add	
PARAMETERIZATION	
EMAIL NOTIFICATION	Exercite Save as Explain or create a New query

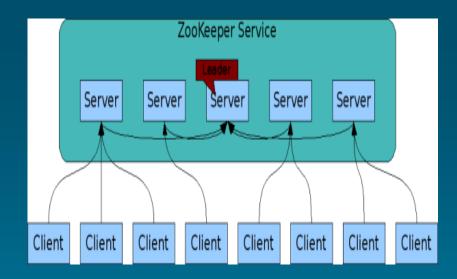




Zookeeper



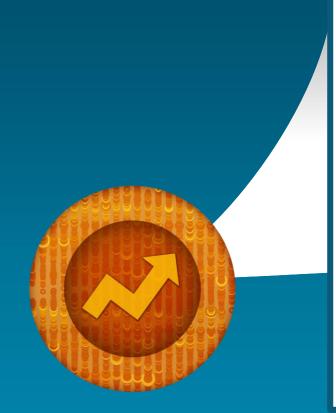
- Zookeeper is a distributed consensus engine
- Provides well-defined concurrent access semantics:
 - Leader election
 - Service discovery
 - Distributed locking / mutual exclusion
 - Message board / mailboxes





Cloudera Manager

End-to-End Administration for CDH



Manage

Easily deploy, configure & optimize clusters



Monitor

Maintain a central view of all activity



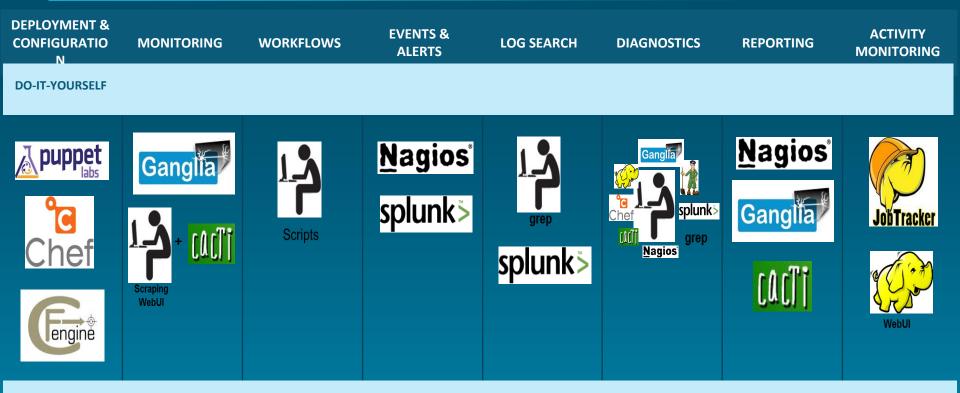
Diagnose Easily identify and resolve issues



Integrate Use Cloudera Manager with existing tools



Cloudera Manager



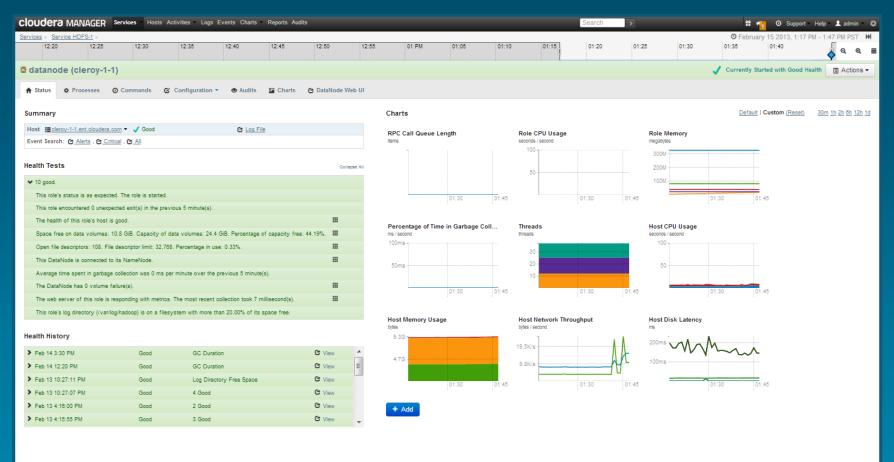
WITH CLOUDERA





View Service Health & Performance

Cloudera Manager Key Features







Thank You!

andrew.wang@cloudera.com @umbrant