Hadoop & Map Reduce

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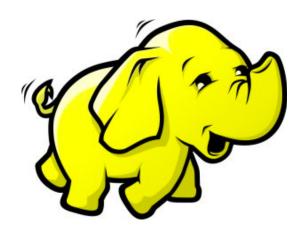
Big Data

What is big data?

- Various definitions:
 - Data that cannot be stored in more traditional relational databases
 - on the order of high millions to billions of records,
 high thousands to millions of columns, TB to PB

What is Hadoop?

- Hadoop is...
 - Hadoop Common
 - HDFS
 - YARN
 - MapReduce



HDFS + MapReduce

 "Apache Hadoop is a framework for running applications on large clusters of commodity hardware which implements the MapReduce computational paradigm and uses HDFS to store data among its nodes."

Hadoop = HDFS + MapReduce

- Hadoop Distributed Filesystem (HDFS)
 - Files sitting on different machines, but they behave like a single file system
 - This system is optimized for fault tolerance
- MapReduce
 - programming model for parallel processing
 - The implementation also makes it fault tolerant.

Distributed read/write capacity

- Distributed read/write capacity
 - Non-distributed hard drive processing is slow. 75MB/sec
 processing speed read 100TB = 16 days
 - Often, some data is more popular than other data. So some data are read more frequently. Thus most servers with data sit there unused. These resources can be more efficiently distributed.

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 - With multiple machines, server failure and error rates increase dramatically:
 - Hadoop expects these failures, and deals with them
 - Node recovery: nodes can get their act together and rejoin the party without full restart

- Distributed read/write capacity
- Deals with hardware failure
- Improves speed

- Distributed read/write capacity
- Deals with hardware failure
- Improves speed
 - read/write in parallel: 100 TB, 75MB/sec HD
 - 1 machine:
 - 75MB/sec → 16 days
 - 1000 machines:
 - 75,000MB/sec = 75 GB/sec \rightarrow 22 minutes

- Distributed read/write capacity
- Deals with hardware failure
- Improves speed
- Fault tolerance

- Distributed read/write capacity
- Deals with hardware failure
- Improves speed
- Fault tolerance
 - Even if a process fails, it's a small part and not the entire
 MapReduce job
 - Data recovery: one node can pick up workload of another
 - HDFS stores each data block on 3 machines by default

- Distributed read/write capacity
- Deals with hardware failure
- Improves speed
- Fault tolerance
- Cheaper

- Distributed read/write capacity
- Deals with hardware failure
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 - computer A has power X
 - computer B has power 4*X
 - $-\cos t(B) >> 4*\cos t(A)$

Hadoop is a framework

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- across a cluster of many computers
- that implements map and reduce functions
- using a distributed filesystem (HDFS)

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- master-slave architecture

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- data nodes (slave): deployed storage machines. MapReduce analyses happen here.
- secondary name node: performs periodic checkpoints (restart the name node from checkpoint in case of name node failure)

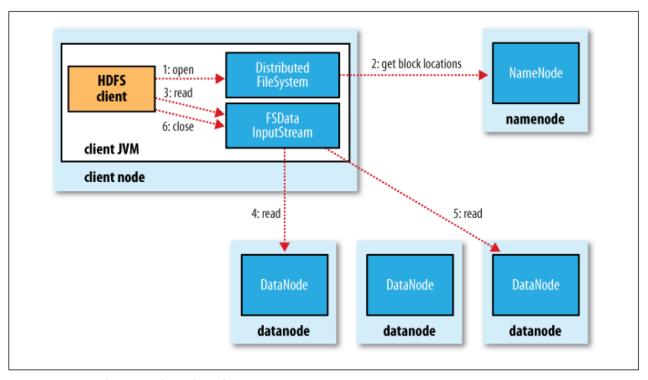


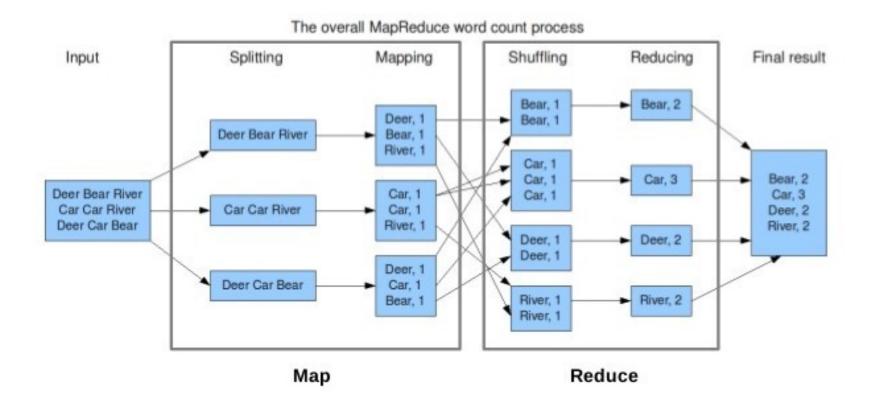
Figure 3-2. A client reading data from HDFS

MapReduce = Map + Reduce functions

- MapReduce = Map + Reduce functions
 - MAP function
 - Input: raw data
 - Output: (key, value) pairs
 - REDUCE function
 - Input: all (key, value) pairs with a certain key
 - Output: summarized result for that key

- Compute word counts for a very large book.
 - MAP function
 - Input: raw text data
 - Output: For every word ("the"), return ("the", 1) key-value pair
 - REDUCE function
 - Input: list of all ("the", 1) pairs
 - Output: count of all "the"s in the book ("the", 2343)

- Mean precipitation over time, across cities
 - MAP function
 - Input: large table of precipitation over time, location data
 - Output: key, value pairs are like: (San Francisco, 2 cm)
 - REDUCE function
 - Input: list of all (city, precipitation) pairs for a given city
 - Output: average precipitation for that city
 - (San Francisco, 0.51 cm)



- Master-slave architecture
 - Master: jobtracker
 - coordinator, scheduler
 - reassigns failed jobs
 - Slave: tasktracker
 - run tasks, send reports to jobtracker
 - lack of report → failure

- Data locality
 - "Push the computation to the data"
 - Mapper code is sent to all data nodes and run locally
 - Thus no data is moved over the network → faster.

- map function: turns data into (key, value) pairs
- reduce function: reduce all values for key to one value or set

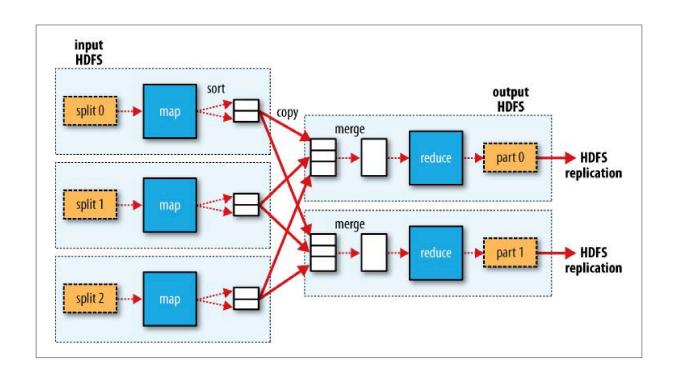
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- map function: turns data into (key,value) pairs
- partition function: assign keys to reducer servers
- reduce function: reduce all values for key to one value or set
- output writer: write output to file system

- input reader: gets data from file system
- map function: turns data into (key,value) pairs
- compare function: sort keys, collecting pairs with the same keys
- partition function: assign keys to reducer servers
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- input reader: gets data from file system
- map function: turns data into (key, value) pairs
- compare function: sort keys, collecting pairs with the same keys
- combiner function: reduce network traffic
- partition function: assign keys to reducer servers
- reduce function: reduce all values for key to one value or set
- output writer: write output to file system



Hadoop environment

- core:
 - Hadoop Common
 - HDFS
 - YARN
 - MapReduce

Hadoop environment

- core: Common, HDFS, YARN, MapReduce
- Hive: takes over management of HDFS storage, interprets & runs jobs via SQL queries
- Pig: write tasks more easily (for MapReduce+HDFS)
- HBase: take HDFS and add some good stuff from relational databases (make hadoop more SQL-like)
- Sqoop: import SQL → HDFS
- Mahout machine learning for Hadoop (err... Spark)

Apache Spark

- Popularized in ~ 2013-2014
 - load all the data into memory
 - up to ~100 times faster than MapReduce
 - iterative algorithms become reasonable!
- Popular configuration:
 - HDFS / S3
 - YARN / Mesos
 - MapReduce Spark

