The Building Blocks of Hadoop - HDFS, MapReduce, and YARN

INTRODUCING HADOOP

Overview

Understand the need for Distributed Computing

Understand the role of Hadoop in a distributed computing setup

Get introduced to core technologies which work on Hadoop

How Much Data Do These Organizations Deal With?

Facebook NSA Google

Facebook

Current storage = 300 petabytes

Processed per day = 600 terabytes

Users per month = 1 billion

Likes per day = 2.7 billion

Photos uploaded per day = 300 million

NSA

Current storage = ~5 exabytes

Processed per day = 30 petabytes

NSA touches 1.6% of internet traffic per day

Web searches, websites visited, phone calls, credit/debit card transactions, financial and health information

Google

Current storage = 15 exabytes

Processed per day = 100 petabytes

Number of pages indexed = 60 trillion

Unique search users per month > 1 billion

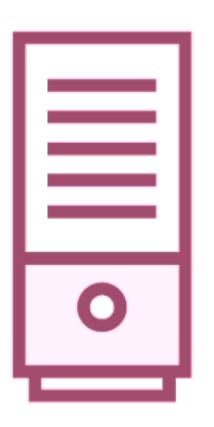
Searches per second = 2.3 million

Huge Data Set

Facebook NSA Google



Huge Data Set



Cannot meet big data requirements

Raw data

Store massive amounts of data

Store

Extract useful information

Store massive amounts of data

Process it in a timely manner

Store

Process

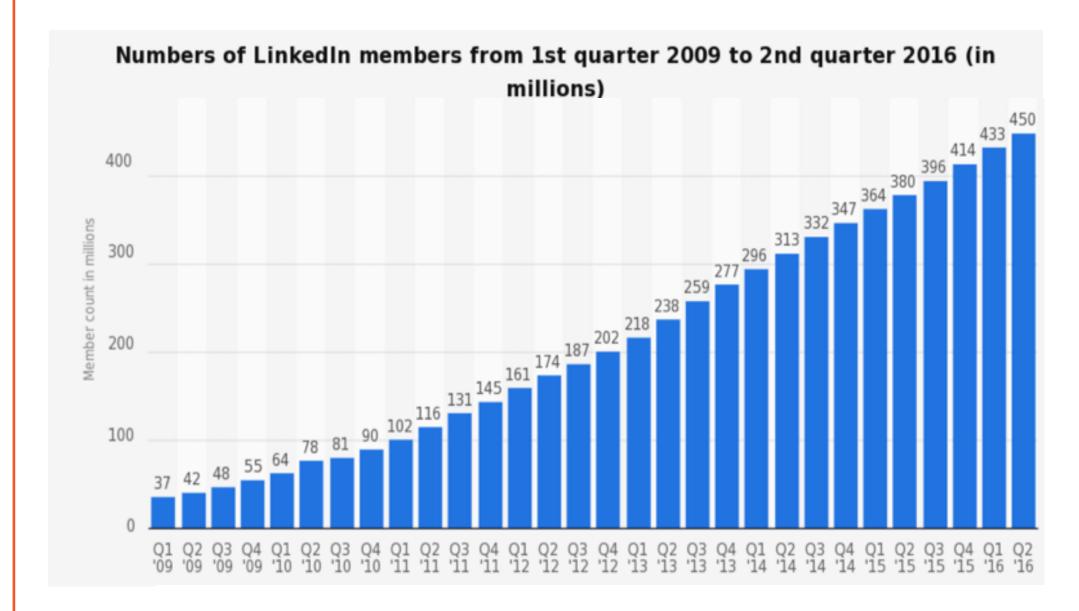


Store massive amounts of data

Process it in a timely manner

Store Process ?

Growing size of data



Store Process ?

The infrastructure needs to keep up with the growing size of data

Store

Process

Accommodate changing needs

Store massive amounts of data

Process it in a timely manner

Scale easily as data grows

Store

Process

Scale

Store massive amounts of data

Process it in a timely manner

Scale easily as data grows



Traditional data technologies don't cut it anymore

Store Process Scale

Distributed Computing Frameworks like Hadoop were developed for exactly this

Two Ways to Build a System

Monolithic

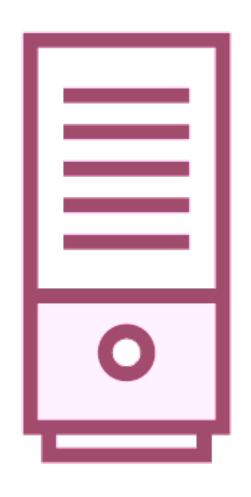
Two Ways to Build a Team



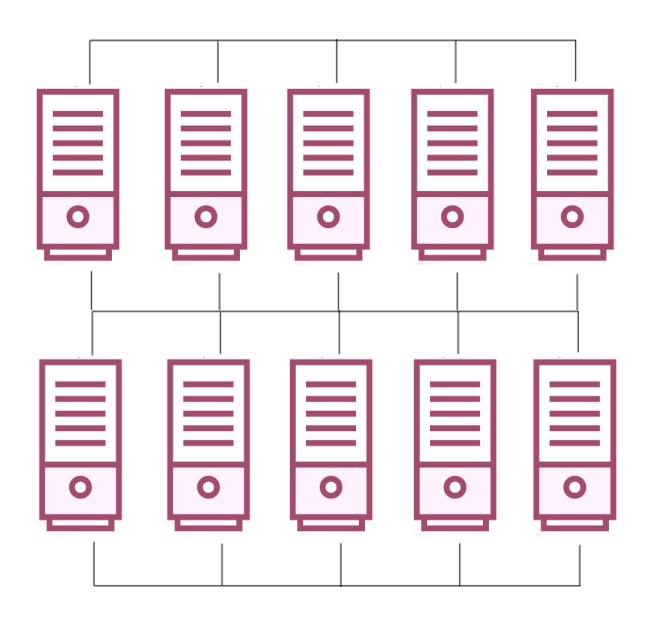


A star who dribbles and shoots

A team of good players who know how to pass



A supercomputer



A cluster of decent machines that know how to parallelize

Two Ways to Build a System

Monolithic

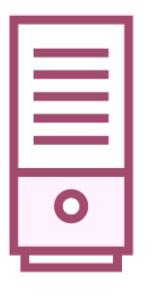
Monolithic



One star player

Monolithic

A single powerful server



2x Expense

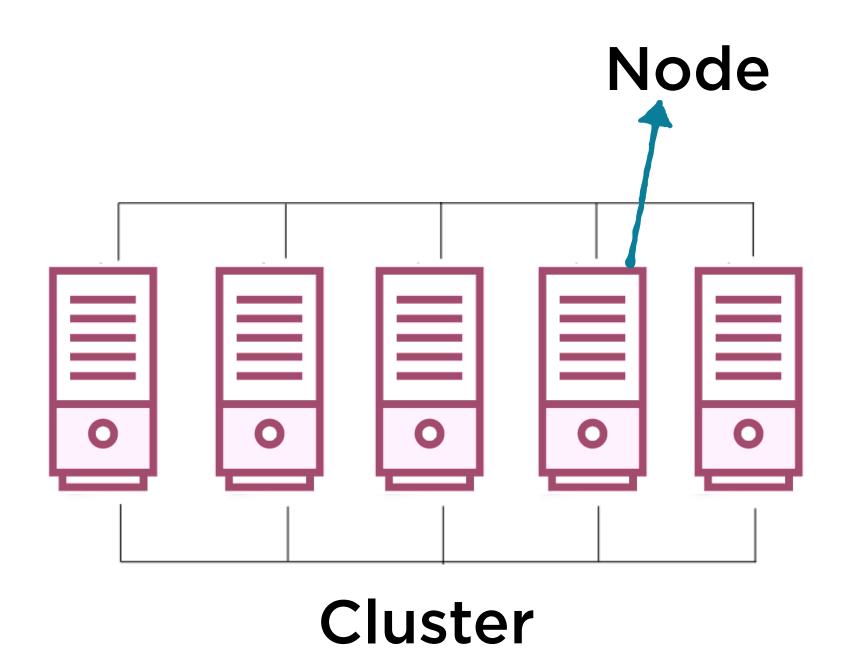
< 2x Performance

Two Ways to Build a System

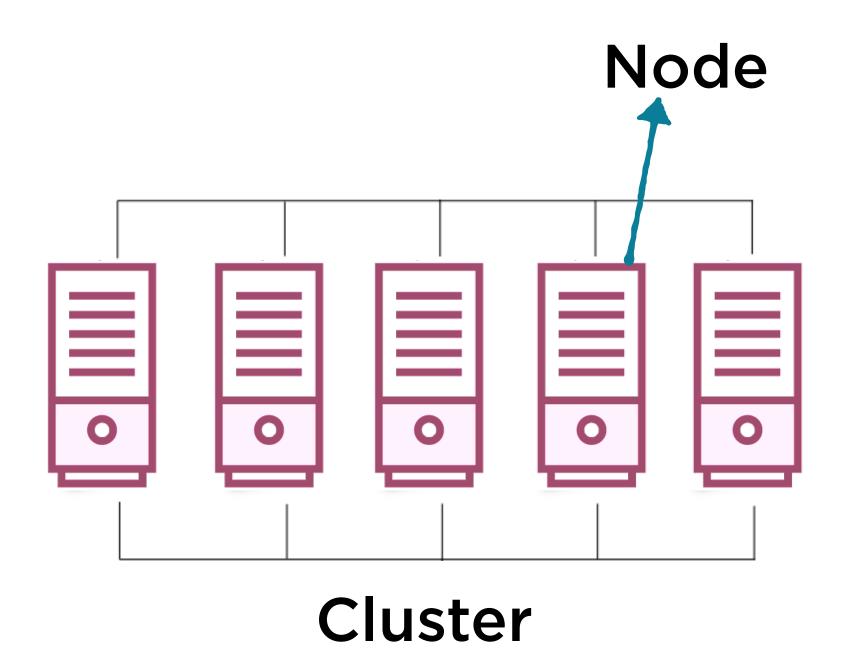
Monolithic



A team of good players who know how to pass



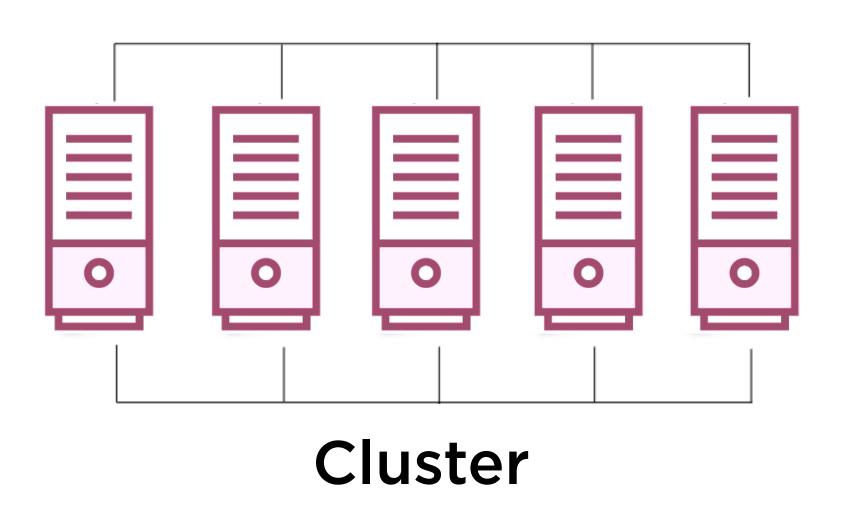
Distributed System



Many small and cheap computers come together...

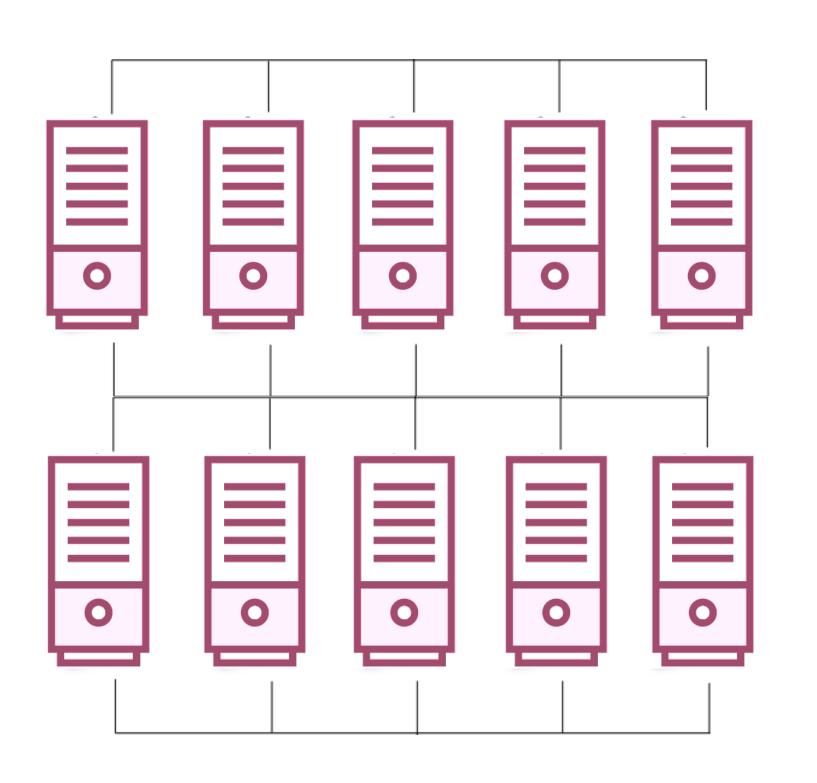
...to act as a single entity

Distributed System



Such a system can scale linearly

Distributed System



2x Nodes

2x Storage

~ 2x Speed

Server Farms



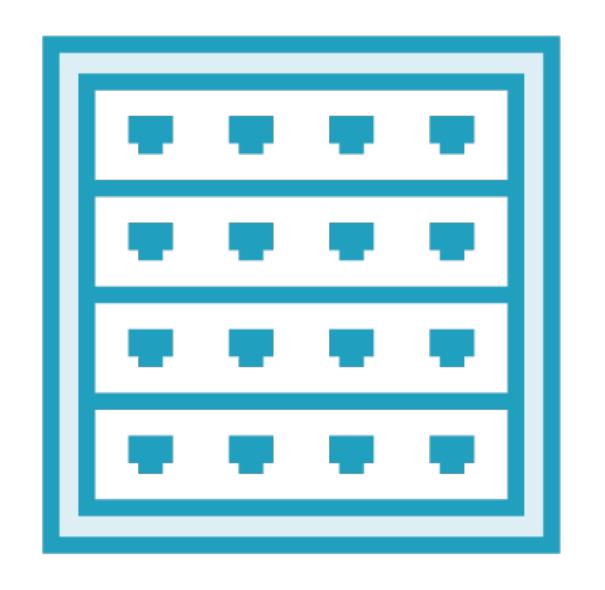
Companies like
Facebook, Google,
Amazon are building
vast server farms

Server Farms



These farms have 100s of 1000s of servers working in tandem to process complex data

Server Farms



All of these servers need to be co-ordinated by a single piece of software

Single Co-ordinating Software

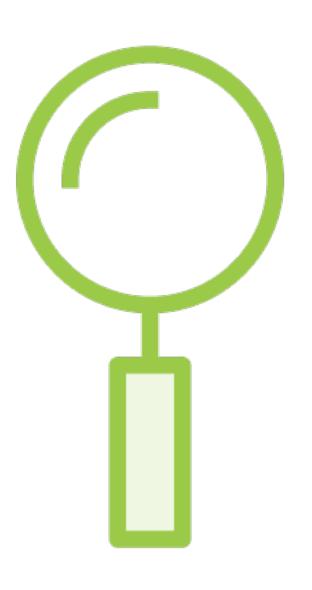


- Partition data
- Co-ordinate computing tasks
- Handle fault tolerance and recovery
- Allocate capacity to processes

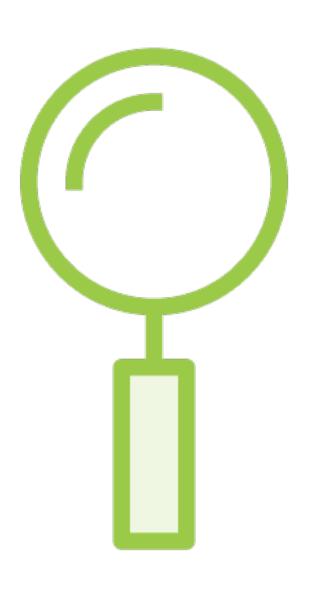
Store Process Scale

Distributed Computing makes for a lot of complexity

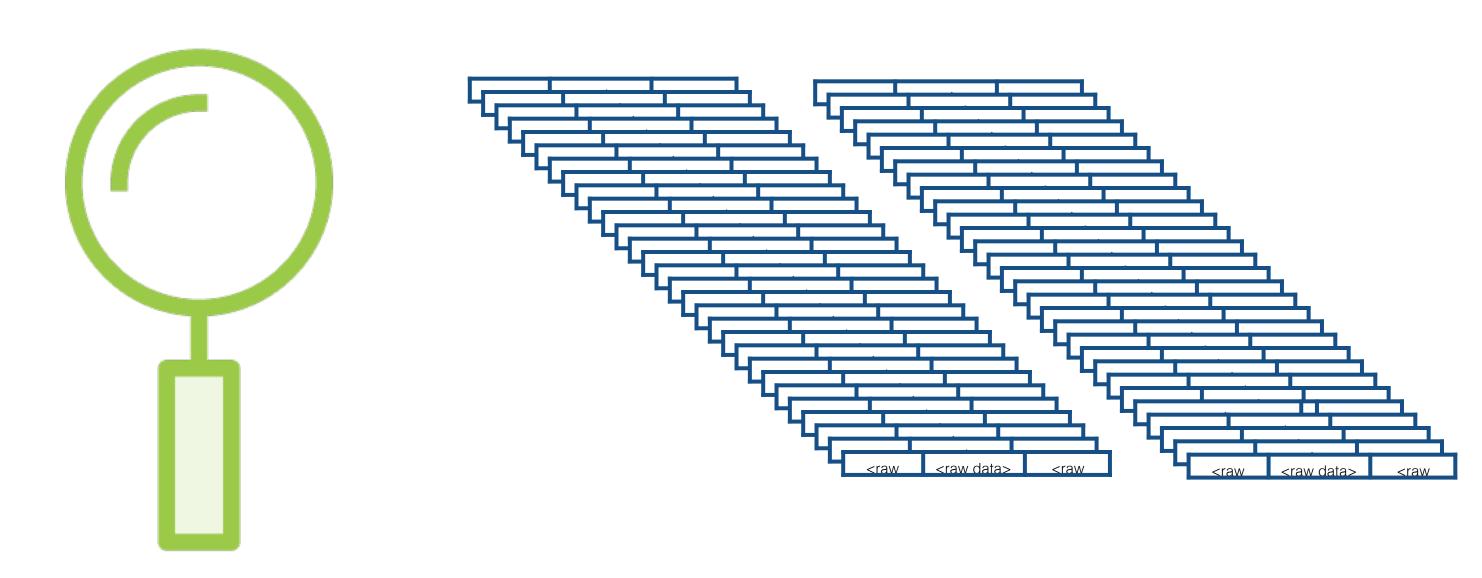
Single Co-ordinating Software



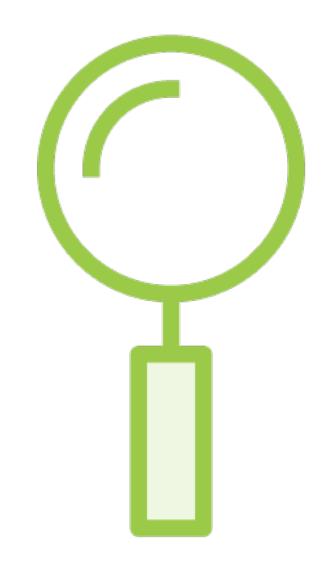
Back in the early 2000s Google realized that web search requires something completely new

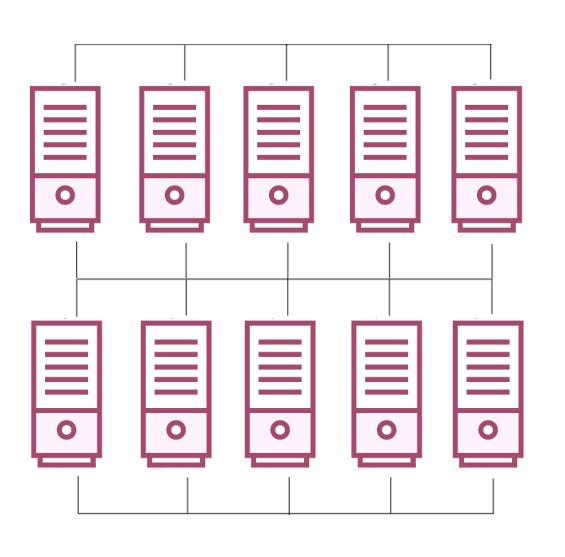


Google developed proprietary software to run on these distributed systems

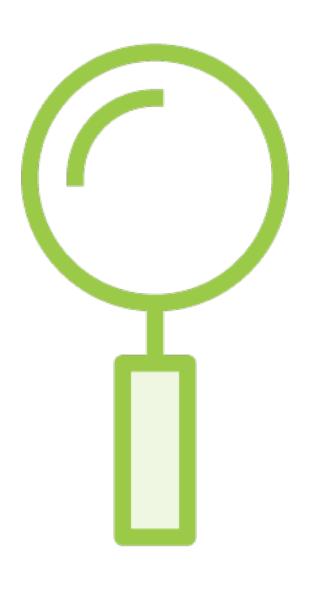


First: store millions of records on multiple machines





Second: run processes on all these machines to crunch data



Google File System

To solve distributed storage

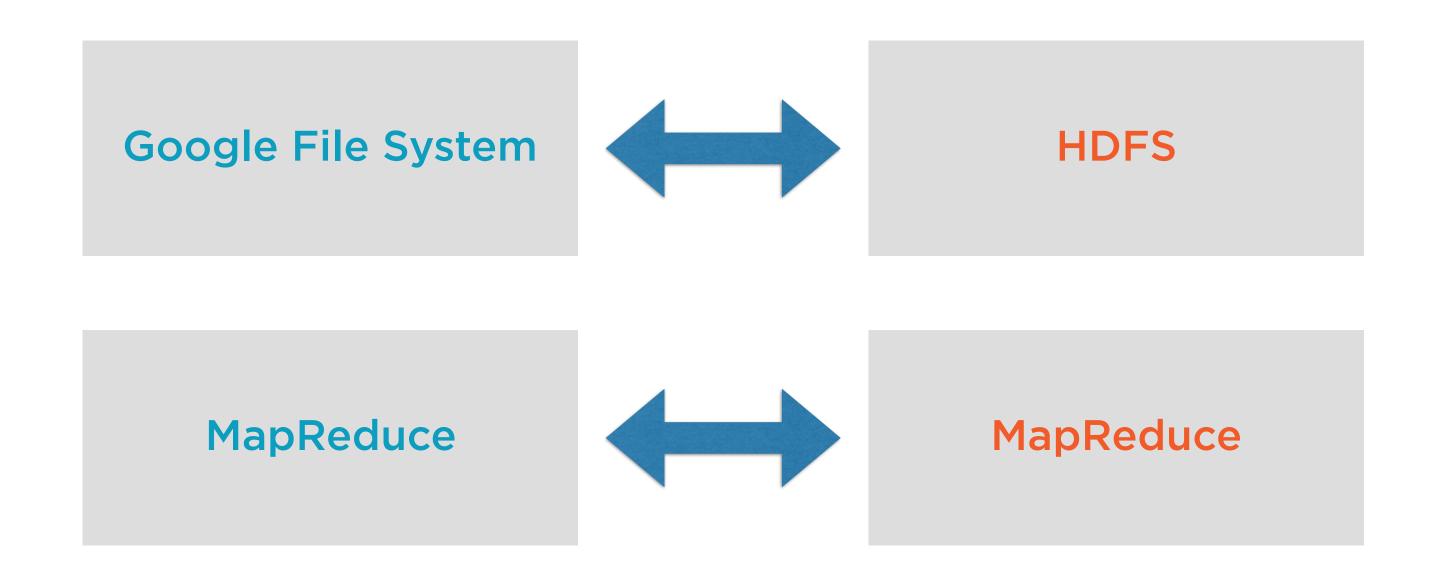
MapReduce

To solve distributed computing

Google File System

MapReduce

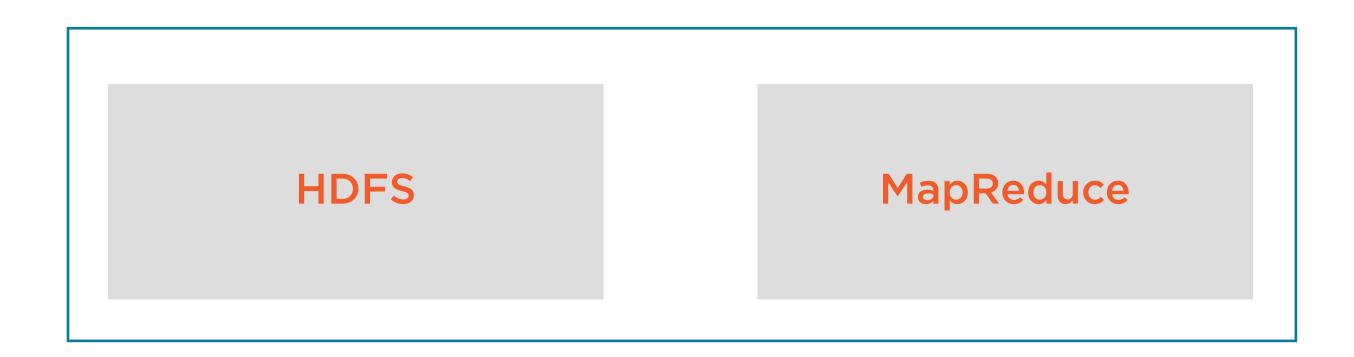
Apache developed open source versions of these technologies



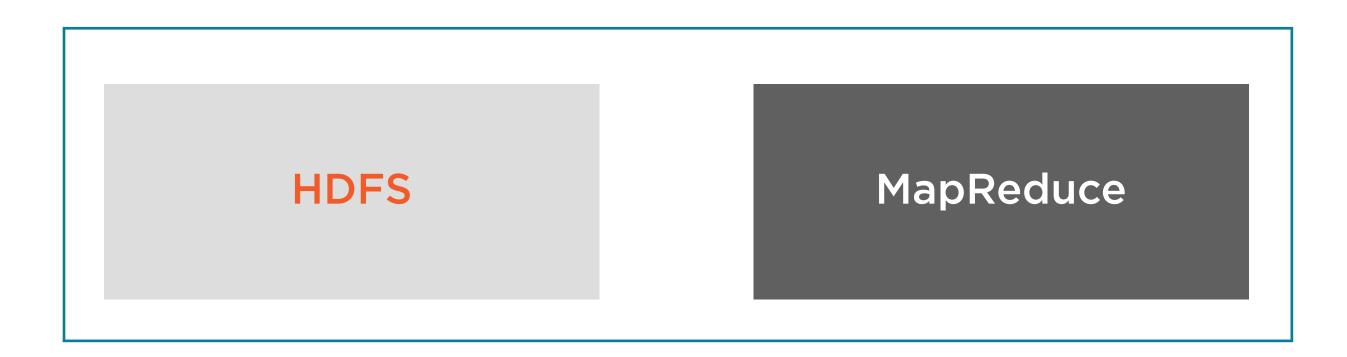
HDFS MapReduce

A file system to manage the storage of data

A framework to process data across multiple servers



In 2013, Apache released Hadoop 2.0



MapReduce was broken into two separate parts

HDFS MapReduce YARN

A framework to define a data processing task

A framework to run the data processing task



Each of these components have corresponding configuration files

Co-ordination Between Hadoop Blocks

MapReduce

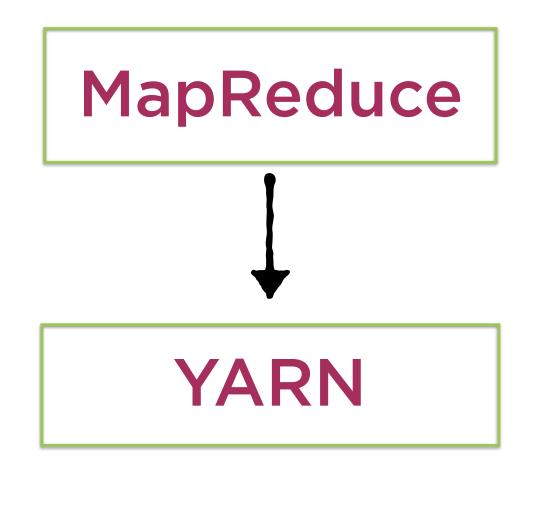


User defines map and reduce tasks using the MapReduce API

YARN

HDFS

Co-ordination Between Hadoop Blocks

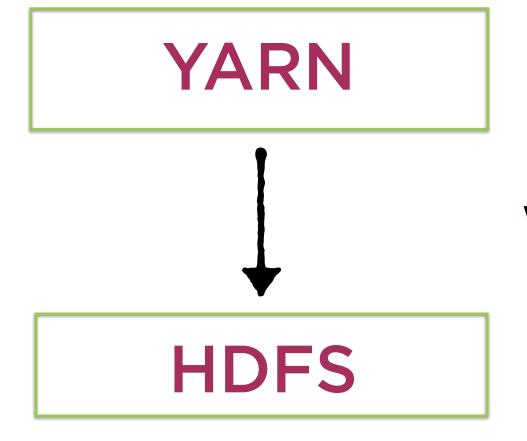


A job is triggered on the cluster

HDFS

Co-ordination Between Hadoop Blocks

MapReduce



YARN figures out where and how to run the job, and stores the result in HDFS

Hadoop Ecosystem

Hadoop

An ecosystem of tools have sprung up around this core piece of software

Hadoop Ecosystem

Pig Hive **HBase** Hadoop Flume/Sqoop Spark Oozie

Hadoop Ecosystem

Pig Hive **HBase** Flume/Sqoop Spark Oozie Hive

Provides an SQL interface to Hadoop

The bridge to Hadoop for folks who don't have exposure to OOP in Java

HBase

A database management system on top of Hadoop

Integrates with your application just like a traditional database

Pig

A data manipulation language

Transforms unstructured data into a structured format

Query this structured data using interfaces like Hive

Spark

A distributed computing engine used along with Hadoop

Interactive shell to quickly process datasets

Has a bunch of built in libraries for machine learning, stream processing, graph processing etc.

Oozie

A tool to schedule workflows on all the Hadoop ecosystem technologies Flume/Sqoop

Tools to transfer data between other systems and Hadoop

Summary

Understood the need for Distributed Computing

Understood the role of Hadoop in a distributed computing setup

Overview of basic technologies which exist in the Hadoop eco-system