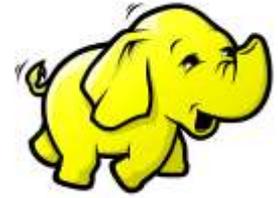


# Hadoop Ecosystem



BY RAHIM A.

# History of Hadoop



- Hadoop was created by Doug Cutting, the creator of Apache Lucene, the widely used text search library.
- Hadoop has its origins in Apache Nutch, an open source web search engine, itself a part of the Lucene project.
- In 2003, Google published a paper described Google's distributed file system GFS.
- In 2004, Nutch's developers implemented open source version on GFS, the Nutch Distributed File System (NDFS) – which evolved to HDFS
- In 2004, Google published the paper that introduced MapReduce
- In 2005, the Nutch developers implemented MapReduce in Nutch, and ported all the major Nutch algorithms to run using MapReduce and NDFS
- Thanks to Google, Hadoop including HDFS and MapReduce, was born in 2005

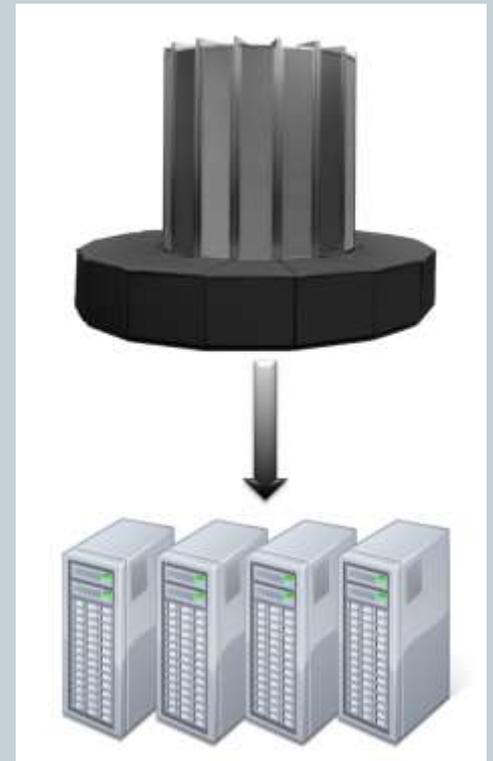
# The Motivation for Hadoop



- “In pioneer days they used oxen for heavy pulling, and when one ox couldn’t budge a log, we didn’t try to grow a larger ox. We shouldn’t be trying for bigger computers, but for *more systems* of computers.”

– Grace Hopper

- Main motivation for Hadoop is to utilize distributed systems for single job

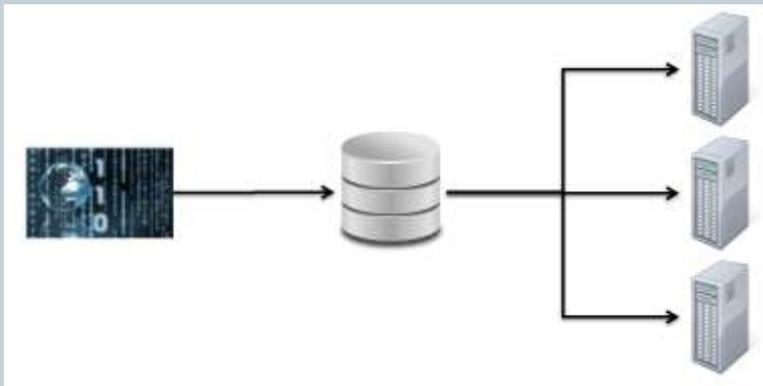


# Hadoop Vs. Traditional systems



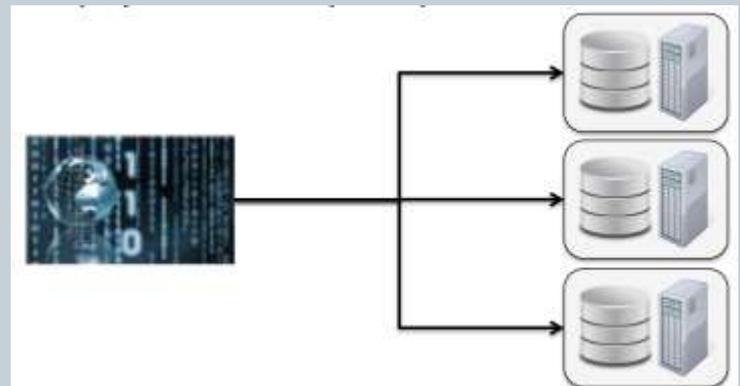
- **Traditional systems**

- Data is stored in a central location
- Data is copied to processors at runtime
- Fine for limited amounts of data

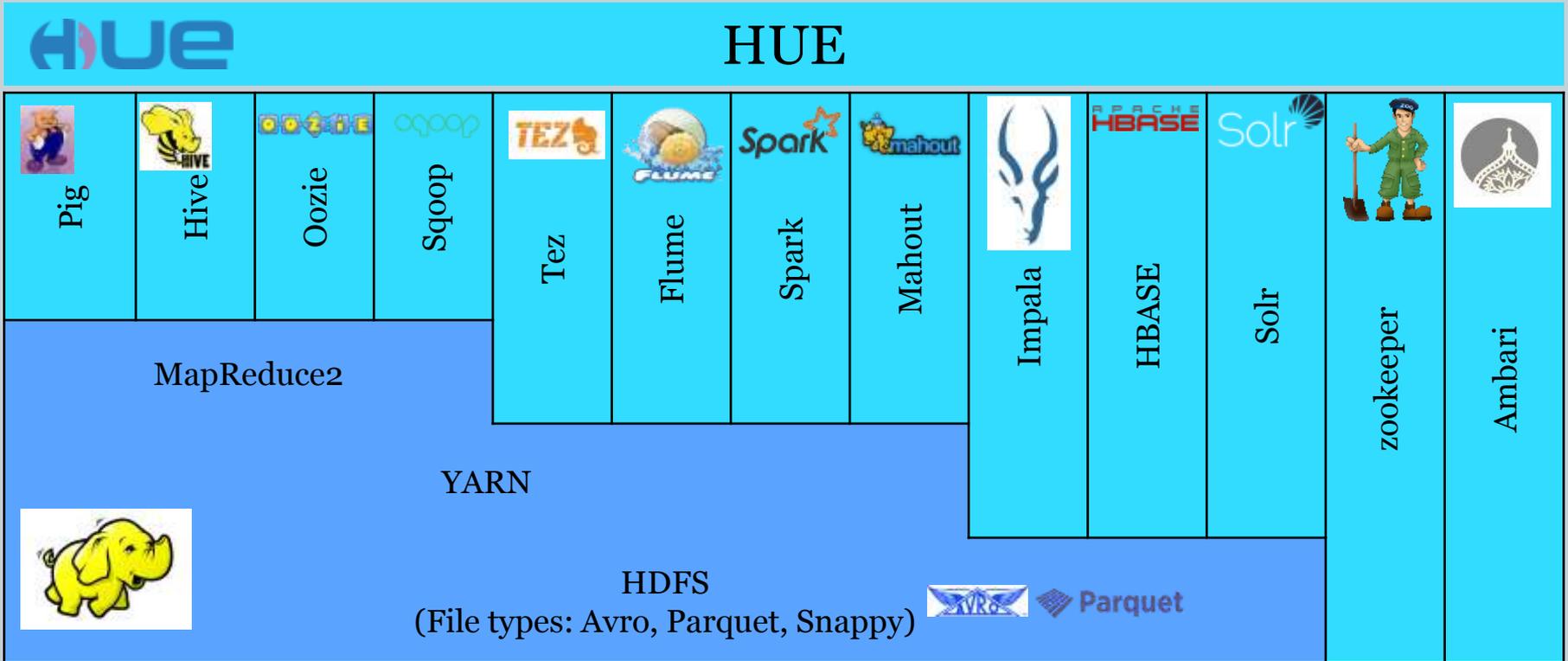


- **Hadoop**

- Distribute data when the data is stored
- Run computation where the data is - bring the computation to the data
- Data is replicated
- Hadoop is scalable and fault tolerant



# Hadoop Related Project



\*\* Apart them there are also projects – **Storm, Kafka, Samza** for streaming and message processing; **Chukwa** – for data collection; **Drill** – for interactive data analysis; **Cascading, Flink, Whirl...**

# Hadoop core components



- **Hadoop common**
  - Common utilities to support other Hadoop modules
- **Hadoop Distributed File System (HDFS)**
  - Distributed file system
- **Yet Another Resource Negotiator (YARN)**
  - Framework for job scheduling and cluster resource manager
- **MapReduce2**
  - YARN based framework for parallel processing
- **MapReduce1**
  - Older and depreciated map reduce framework, based on own job scheduler –forget about it!

# HDFS



- **Properties:**
  - Sits on top of a native filesystem - *such as ext3, ext4 or xfs*
  - HDFS performs best with a 'modest' number of large files – *millions, rather than billions, of files, Each file typically 100MB or more*
  - Files in HDFS are 'write once' - *no random writes to files*
  - HDFS is optimized for large, streaming reads of files – *rather than random reads*
- **HDFS deemons**
  - Datanode – stores file blocks
  - Namenode – store filesystem's metadata (file name and path, number of blocks, number of block replicas, blocks locations)
  - Secondary namenode – bookkeeping and check points

# HDFS: file storage



**Metadata**

`/logs/031512.log`: B1, B2, B3  
`/logs/042313.log`: B4, B5

**B1**: A, B, D  
**B2**: B, D, E  
**B3**: A, B, C  
**B4**: A, B, E  
**B5**: C, E, D

**NameNode**

`/logs/031512.log`

1  
2  
3

`/logs/042313.log`

4  
5

**Node A**

1 3  
4

**Node B**

1 2  
3 4

**Node C**

3 5

**Node D**

1 5  
2

**Node E**

2 5  
4

`/logs/042313.log?`

B4, B5

**Client**

# YARN



- **Concepts**
  - YARN is cluster resource management system
  - Provides APIs for requesting and working with cluster resources – CPU and Memory
- **YARN daemons**
  - Resource manager (one per cluster) - to manage the use of resources across the cluster
  - Node managers - running on all the nodes in the cluster to launch and monitor containers. A **container** - executes an application-specific process with a constrained set of resources (memory, CPU)
  - Application's history server – only for map reduce jobs, stores job running histories

# YARN features



- **Scalability**

can run on larger clusters than MapReduce 1. It is designed to scale up to 10,000 nodes and 100,000 tasks.

- **Availability**

provide HA for the resource manager, then for YARN applications

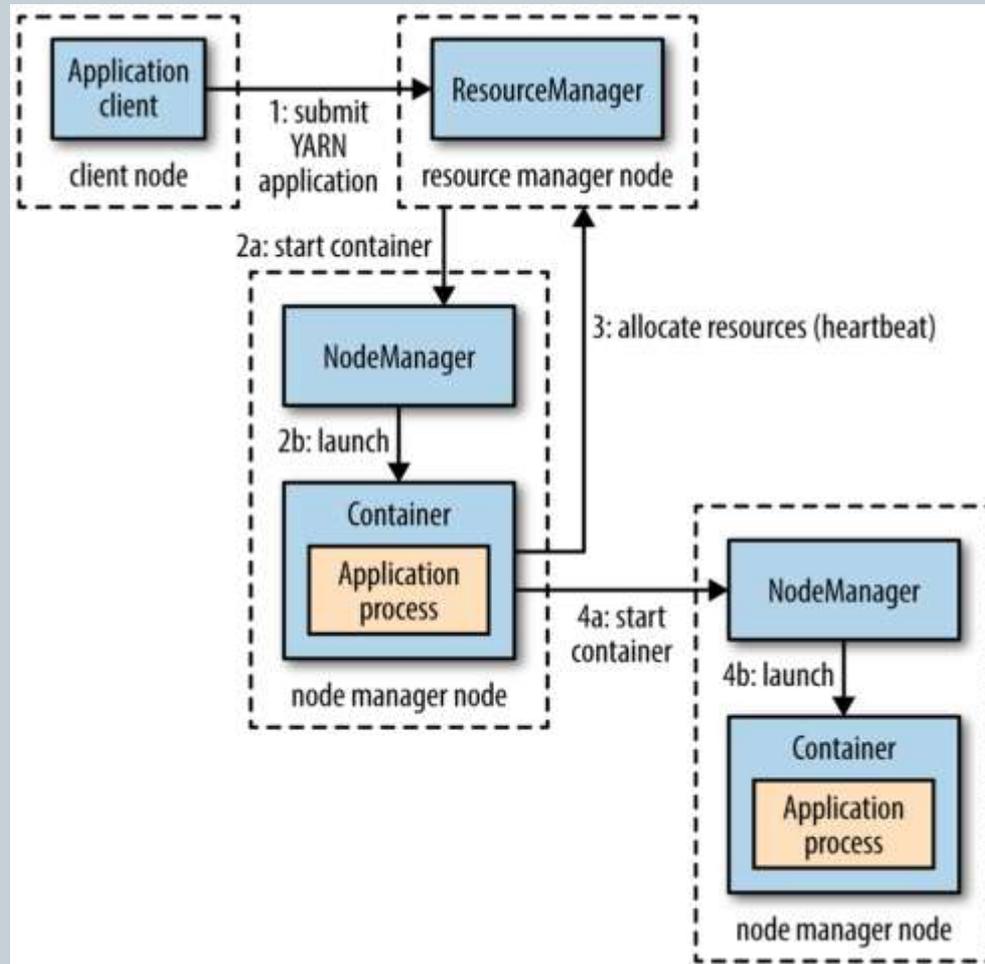
- **Utilization**

In YARN, a node manager manages a pool of resources, rather than a fixed number of designated slots

- **Multitenancy**

the biggest benefit of YARN is that it opens up Hadoop to other types of distributed application beyond MapReduce”

# YARN: how applications are run



# MapReduce



- MapReduce is a method for distributing a task across multiple nodes
- Each node processes data stored on that node – Where possible
- Consists of two phases:
  - Map
  - Reduce
- Features of MapReduce
  - Automatic parallelization and distribution
  - Fault tolerance
  - A clean abstraction for programmers
    - ✦ MapReduce programs are usually written in Java
    - ✦ Can be written in any language using *Hadoop Streaming*
  - MapReduce abstracts all the ‘housekeeping’ away from the developer

# MapReduce Key stages



- **The Mapper**

- Each Map task (typically) operates on a single HDFS block
- Map tasks (usually) run on the node where the block is stored

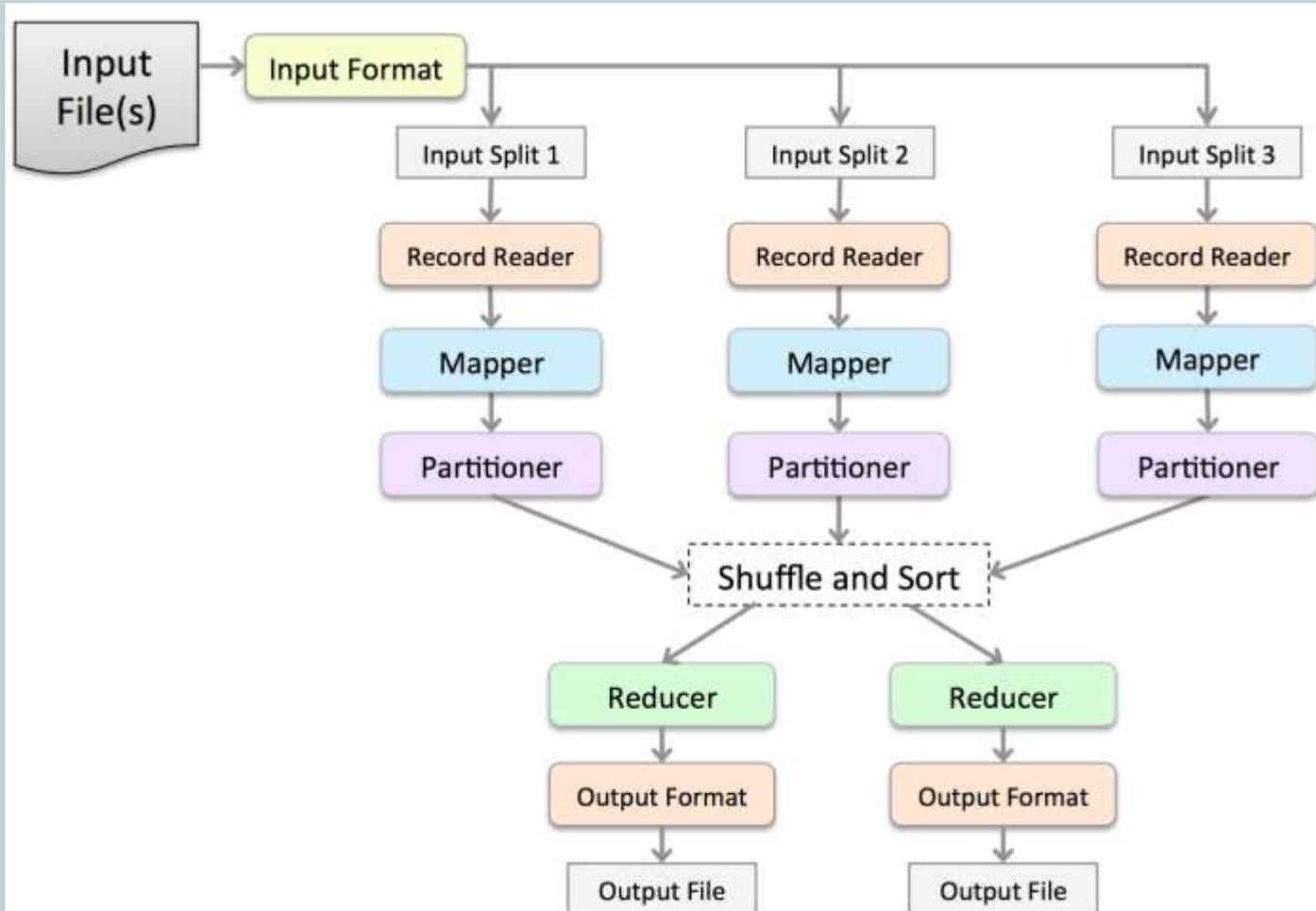
- **Shuffle and Sort**

- Sorts and consolidates intermediate data from all mappers
- Happens after all Map tasks are complete and before Reduce tasks start

- **The Reducer**

- Operates on shuffled/sorted intermediate data (Map task output)
- Produces final output

# MapReduce Workflow



# Motivation for Hadoop related projects



- Complexity of writing MapReduce jobs
- Not all tasks fits MapReduce
  - it is not good for complex directed-acyclic-graphs
  - It is not suitable for iterative algorithms
- Address business and data analyst needs
- Performance issues
- Data ingestion and integration with traditional systems
- Provide ready to use frameworks and system build on big data

# HIVE



- Apache Hive is a high level abstraction on top of MapReduce
- Uses an SQL like language called HiveQL
- Generates MapReduce jobs that run on the Hadoop cluster
- Originally developed by Facebook for data warehousing
- Hive Limitations
  - Not all 'standard' SQL is supported – No correlated subqueries
  - No support for UPDATE or DELETE
  - No support for INSERTing single rows

## HiveQL Statements

```
SELECT zipcode, SUM(cost) AS total
FROM customers JOIN orders
ON customers.id = orders.cid
WHERE zipcode LIKE '63%'
GROUP BY zipcode
ORDER BY total DESC;
```

## Hive Interpreter / Execution Engine

- Parse HiveQL
- Make optimizations
- Plan execution
- Generate MapReduce jobs
- Submit job(s) to Hadoop
- Monitor progress



## MapReduce Jobs



# PIG



- Apache Pig is a platform for data analysis and processing on Hadoop
- Offers an alternative to writing MapReduce code directly
- Originally developed at Yahoo
- Pig Goals: flexibility, productivity, and maintainability
- Main components:
  - The data flow language (Pig Latin)
  - The interactive shell (Grunt)
  - The Pig interpreter and execution engine

## Pig Latin Script

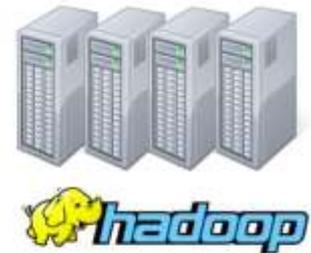
```
AllSales = LOAD 'sales'  
           AS (cust, price);  
BigSales = FILTER AllSales  
           BY price > 100;  
STORE BigSales INTO 'myreport';
```

## Pig Interpreter / Execution Engine

- Preprocess and parse Pig Latin
- Check data types
- Make optimizations
- Plan execution
- Generate MapReduce jobs
- Submit job(s) to Hadoop
- Monitor progress



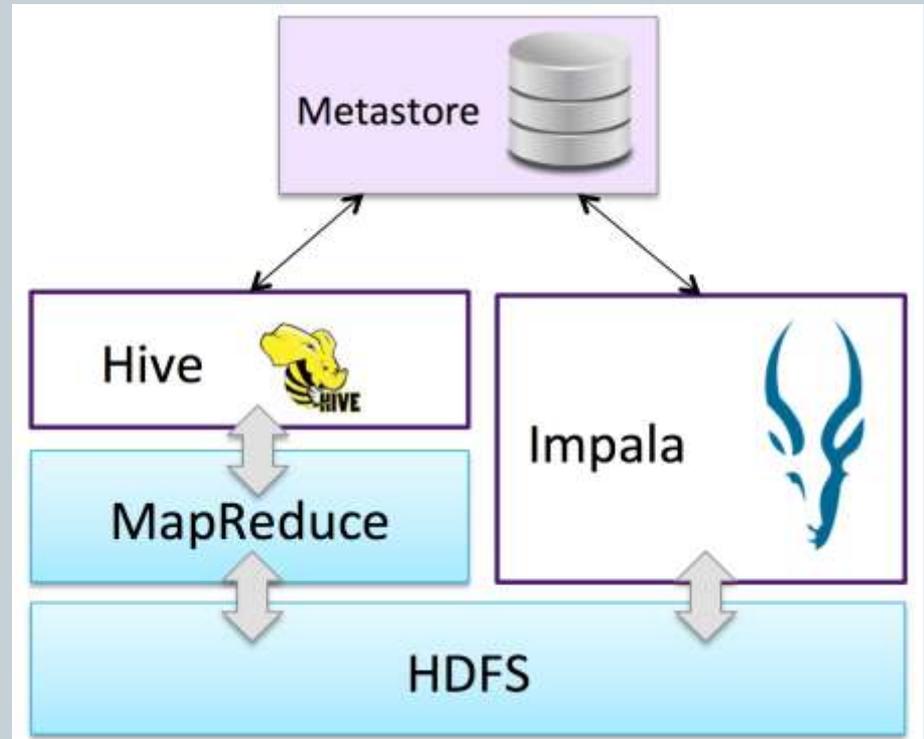
## MapReduce Jobs



# Impala



- High performance SQL engine for vast amounts of data
  - Similar query language to HiveQL
  - 10 to 50+ Times faster than Hive, Pig, or MapReduce
- Developed by Cloudera
- MapReduce is not optimized for interactive queries -High latency – even trivial queries can take 10 seconds or more
- Impala does not use MapReduce
- Uses the same Metastore as Hive



# SQOOP and Flume



- **Sqoop**

- Sqoop stands for “SQL for Hadoop”
- Imports tables from an RDBMS into HDFS and vice-versa
  - ✦ Just one table
  - ✦ All tables in a database
  - ✦ Sqoop supports a WHERE clause
- Uses MapReduce to actually import the data to RDBMS
- Uses a JDBC interface
- Supports Oracle Database (connector developed with Quest Software)

- **Flume**

- Flume is a distributed, reliable, available service for efficiently moving large amounts of data as it is produced
- Ideally suited for file ingestions (web logs, network device logs etc.)

# Sample Workflow



Sessionize Web Log Data with Pig



Import Transaction Data from RDBMS



Sentiment Analysis on Social Media with Hive



Hadoop Cluster with Impala



Analyst using Impala shell for ad hoc queries

Total sales:	\$10,000,000
Last quarter:	\$7,800,000
Top Departments:	
Electronics	\$3,200,000
Audio	\$1,750,000

Generate Nightly Reports using Pig, Hive, or Impala

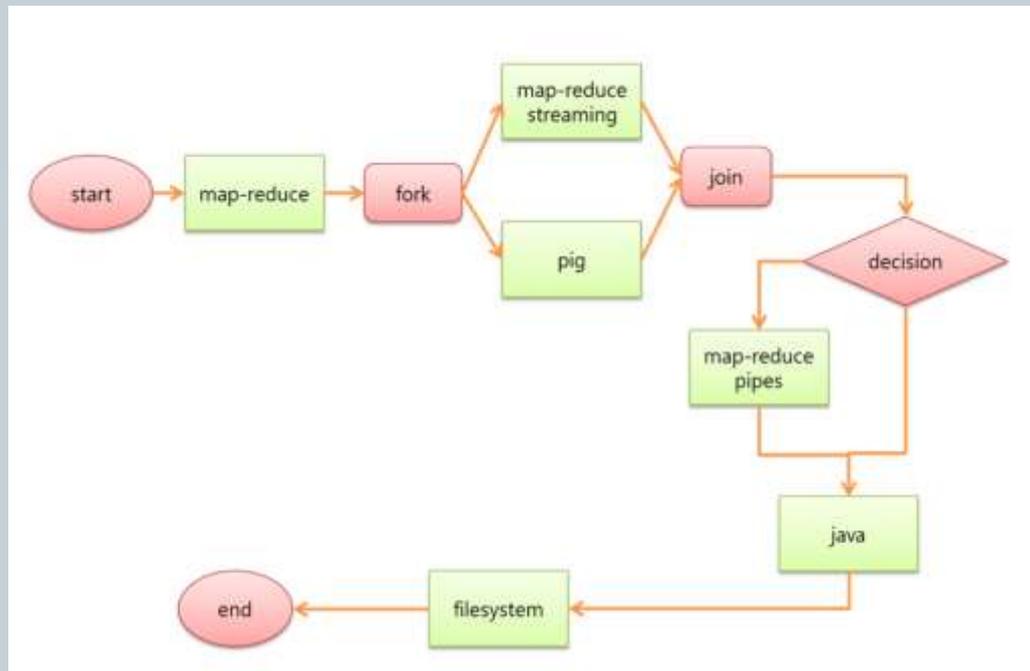


Analyst using Impala via BI tool

# Oozie



- Oozie is a ‘workflow engine’
- Runs on a server, typically outside the cluster
- Runs workflows of Hadoop jobs
  - Including Pig, Hive, Sqoop jobs
  - Submits those jobs to the cluster based on a workflow definition
- Workflow definitions are in XML and submitted via HTTP
- Jobs can be scheduled
  - One-off or recurring jobs



# HBASE



- HBase is the Hadoop database - A 'NoSQL' datastore
- Developed as open source version of Google's BigTable
- Can store massive amounts of data – Petabytes+
- High write throughput Scales to 100K inserts per second
- Handles sparse data well – No wasted spaces for empty columns in a row
- Limited access model
  - Optimized for lookup of a row by key rather than full queries
  - No transactions: single row operations only
  - Only one column (the 'row key') is indexed

# HBASE vs RDBMSs

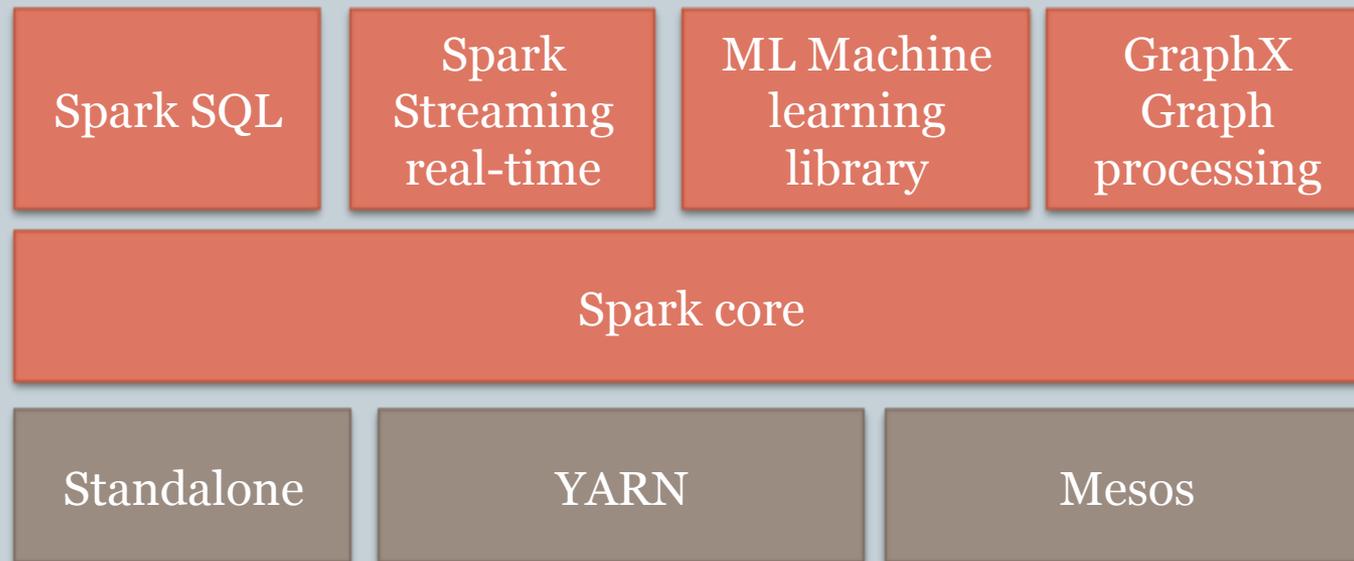


	<b>RDBMS</b>	<b>HBASE</b>
Data layout	Row oriented	Column oriented
Transaction	Supported	Single row only
Query language	SQL	Put/get/scan
Security	Buit-in Authentication- Authorization	Kerberos
Indexes	Any column	Row key only
Max data size	TBs	PB+
Read-write throughput	Thousands	Millions

# Spark



- Spark is an open source project, started in 2009 as a research project in the UC Berkeley
- Spark is a cluster computing platform designed to be fast and general-purpose
- 10-20 times faster than MapReduce – so it is MapReduce killer
- Written in Scala, using Akka framework
- Spark can run as standalone cluster, on YARN, and on Apache Mesos cluster
- Spark components:



# Spark components



- **Spark Core**
  - task scheduling, memory management, fault recovery, interacting with storage systems, and more.
  - API that defines resilient distributed datasets (RDDs), -
    - ✦ RDDs represent a collection of items distributed across Spark cluster nodes that can be manipulated in parallel.
- **Spark SQL**
  - Spark's package for working with structured data
- **Spark Streaming**
  - Spark component that enables processing of live streams of data
- **MLlib**
  - Provides classification, regression, clustering, collaborative filtering
- **GraphX**
  - Library for manipulating graphs and performing graph-parallel computations