Big Data Processing

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Outline

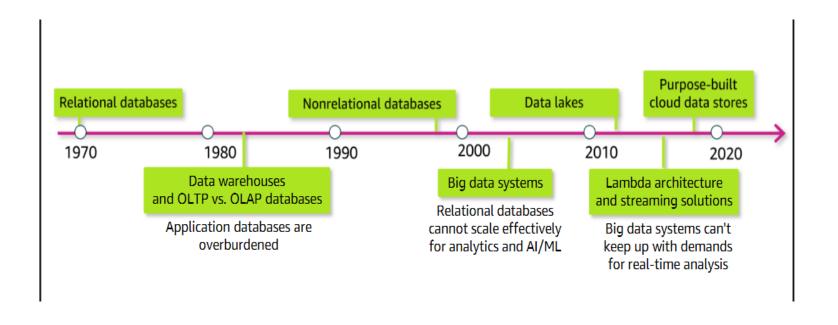
- Introduction to Big Data
- MapReduce Programming Model
- The Hadoop Framework

- Organizations are investing heavily in data and analytics services
 - Data driven
- There is an explosion of data in the current computing landscape
 - > 65% of the world population connected to the Internet
 - Web 2.0/3.0
 - IoT
 - . . .
- This explosion coincide with the availability of powerful compute resource with relatively cheaper cost
 - Cloud Computing

- Data pipeline
 - An infrastructure that supports data-driven decision
 - It basically involves:
 - Collection
 - Cleansing
 - Storage and processing
 - Make decisions based on the result

- Data can be generally:
 - Structured
 - Example: Relational databased
 - Semi-structured
 - Example: XML/JSON file
 - Unstructured
 - Example: collection of text files or sensor data
- The data explosion is unstructured data

- Organizations usually have large datasets collected from different sources
- Evolved through the years:
 - Relational DB → Data warehouses → non relational
 DB → Big data → data lakes → purpose-built data stores





Big Data

- Extremely large and diverse collections of structured, unstructured, and semi-structured data that continues to grow exponentially over time
- These datasets are large enough that traditional data processing software can't handle its storage and processing efficiently
 - Big data processing

- Big Data growth can be viewed as a threedimensional phenomenon:
 - 1. Implies an increased **volume** of data
 - Requires increased processing speed to process more data and produce more results – velocity
 - Involves a diversity of data sources and data types variety
- Additional three vs
 - 4. The accuracy or quality of the data veracity
 - 5. The worth of the data value
 - 6. The meaning of data changes constantly variability

- Some of the enduring challenges posed by Big Data are:
 - Develop a scalable data infrastructures capable of responding promptly to timing constraints of an application
 - Develop effective means of accommodating diversity in data management systems
 - Support comprehensive end-to-end processing and knowledge extraction from data
 - Develop convenient interfaces for a layman involved in data collection and analysis
 - Quality of data
 - Compliance and security

- Big data processing can be:
 - Batch data processing
 - Involves querying large amount of data
 - Cold data Infrequently accessed
 - Streaming data processing
 - Generated continuously by many sources
 - Example: IoT devices, sensors, financial data from trading floors, ...

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- Parallel Processing
 - Plays key role in big data processing
 - Different programming models:
 - Automatic parallelization
 - Parallelization made by the compiler
 - Writing of the parallel application from scratch
 - Example: MPI, OpenMP, CUDA
 - Using Parallel libraries
 - Example: Hadoop, GridGain
 - Flynn's taxonomy
 - Tasks can be:
 - Fine grained, Coarse grained, embarrassingly parallel

MapReduce

- Based on a very simple idea for parallel processing of data-intensive applications supporting arbitrarily divisible load sharing
- Inspired by the map and the reduce primitives of the Lisp programming language
- Developed by Google

Processes:

- split the data into blocks, assign each block to an instance/process and then run these instances in parallel map
- merge the partial results produced by individual instances reduce
- The data is stored using the Google File System (GFS)

Map function

- The large data are traced and segregated as key/value pairs
- The map function accepts the key/value pair and returns an intermediate key/value pair

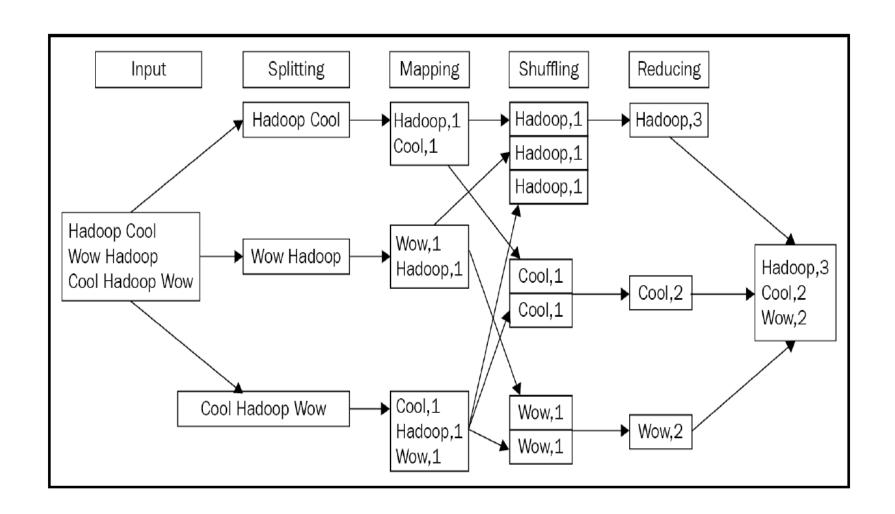
 $map(key_1, value_1) \rightarrow List(Key_2, value_2)$

Reduce function

- Accepts the output of the map function (list of key/value pairs) as an input
- A sort and merge operations are applied on key/value pairs

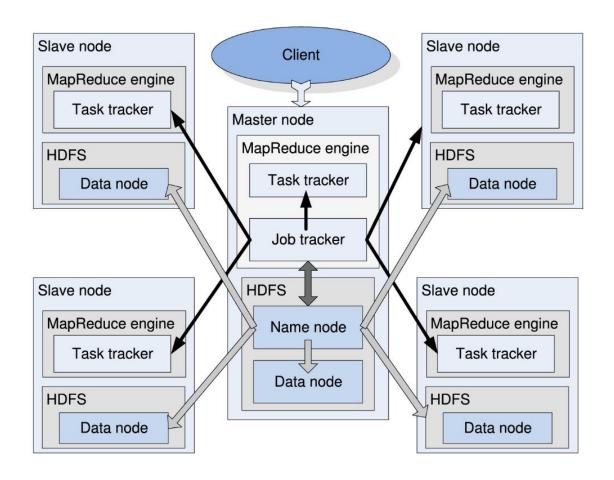
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reduce(List(key_2, value_1) \rightarrow List(value_3))
```

```
map (String key, String value):
      // key:document name
     // value:document contents
   for each word w in value:
       EmitIntermediate(w,"1");
reduce (String key, Iterator values):
       // key:a word
       // values:a list of counts
   int result= 0;
   for each v in values:
       result+=ParseInt(v);
   Emit (AsString (result));
```



- Apache Hadoop
 - An open-source software framework for distributed storage and distributed processing based on the MapReduce programming model
 - Java-based
 - Supports distributed applications handling extremely large volumes of data

- The Apache Hadoop framework has the following modules:
 - Common
 - Contains libraries and utilities needed by all Hadoop modules
 - Distributed File System (HDFS)
 - A distributed file-system that stores data on commodity machines, providing very high aggregate bandwidth across the cluster
 - YARN Yet Another Resource Negotiator
 - A resource-management platform responsible for managing computing resources in clusters and using them for scheduling of users' applications
 - MapReduce
 - An implementation of the MapReduce programming model



- Distributed File System (HDFS)
 - Sits on top of the native filesystem
 - Divides the files into blocks (64 MB, 128 MB, ...)
 - Distributed processing
 - HDFS has two main components:
 - The NameNode
 - The DataNode

The NameNode

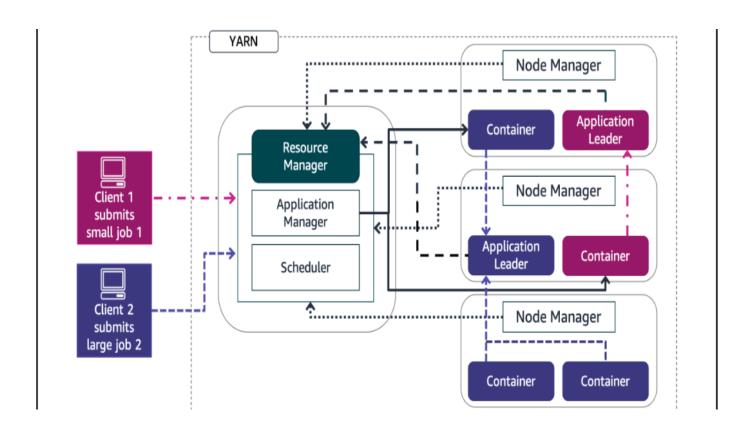
- Contains all the metadata of all content of the filesystem:
 - Filenames, file permissions, and the location of each block of each file
 - The most important machine in HDFS

DataNodes

- Connect to the NameNode and store the blocks within HDFS
- They rely on the NameNode for all metadata information regarding the content in the filesystem

YARN

- Yet Another Resource Negotiator
- A resource management system supplying CPU cycles, memory, and other resources needed by a single job or to a DAG of MapReduce applications
- Has the following components:
 - Resource Manager
 - Scheduler and Application Manager components
 - Node Manager
 - Application leader
 - Containers





- Advantages
 - Scalability
 - Fault tolerance
- Limitations of Hadoop
 - Slower due to the latency of disk I/O
 - Difficult to handle interactive analytical tasks
 - Does not have many tools for:
 - Data Management and governance
 - Data quality and standardization
 - Security

- Hive
 - An open-source system for data-warehousing supporting queries expressed in an SQL-like declarative language
 - Hive Query Language (Hive QL)
 - Must be implemented in the MapReduce Java API
 - Runs over Hadoop
 - Can creates a virtual table over S3 or HDFS
 - Schema-on-read

- Presto (or PrestoDB)
 - An open source, distributed, in-memory SQL query engine, designed from the ground up for fast analytic queries against data of any size
 - Runs on Hadoop
 - Supports:
 - Non-relational sources
 - The Hadoop Distributed File System (HDFS) MongoDB, and HBase
 - Relational data sources such as
 - MySQL, PostgreSQL

- Apache Spark.
 - General-purpose distributed processing framework used for big data workloads
 - Opensource in-memory distributed data processing
 - Has the following components:
 - Spark Core
 - Responsible for memory management, fault tolerance, scheduling and management of jobs, . . .
 - Spark SQL
 - Distributed query engine
 - Spark Graphx
 - Distributed graph processing framework run over spark
 - Spark Streaming
 - Real-time streaming analytics
 - Spark MLlib
 - A library of algorithms for ML

HBase

- A non-relational or NoSQL database in the Hadoop ecosystem
- Opensource and written in Java
- HBase has master and region servers
- Uses Zookeeper to provide centralized highperformance coordination between nodes or region servers
- Compaction
 - HBase keeps track of the changes as change files and then merges them periodically
 - Can be major or minor

Apache Flink

- Streaming data engine that processes real-time stream data for big data applications
- Supports batch and streaming APIs and some SQL

Impala

- A query engine exploiting a shared-nothing parallel database architecture
- Written in C++ and Java and designed to use standard Hadoop components (HDFS, HBase, Metastore, YARN)

References

- MapReduce: Simplified Data Processing on Large Clusters (2004)
- Spark: Cluster Computing with Working Sets (2010)
- Alla, Sridhar. Big Data Analytics with Hadoop 3: Build Highly Effective Analytics Solutions to Gain Valuable Insight into Your Big Data, Packt Publishing, Limited, 2018.