## Introduction to Spark

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# Big Data and Distributed Computing at Google

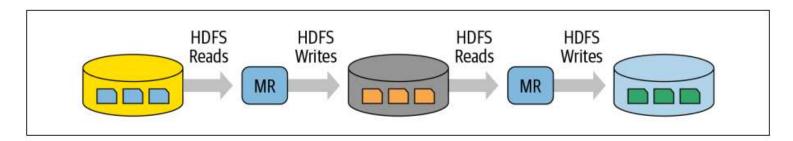
- Three important papers of Google for big data challenges of scale for its search engine
  - Google File System (GFS): a fault-tolerant and distributed file system across many commodity hardware servers in a cluster farm
  - Bigtable: a scalable storage of structured data across GFS
  - MapReduce (MR): a new parallel programming paradigm, based on functional programming, for large-scale processing of data distributed over GFS and Bigtable.

#### Hadoop at Yahoo!

- Google's GFS and MapReduce papers provided a blueprint for
  - the Hadoop File System (HDFS), and
  - MapReduce implementation as a framework for distributed computing
- Yahoo! donated HDFS and MapReduce implementation for Apache Software Foundation (ASF) in April 2006
- Apache Hadoop framework:
  - Hadoop Common,
  - MapReduce,
  - HDFS, and
  - Apache Hadoop YARN

#### Hadoop shortcomings

- The MapReduce framework on HDFS had a few shortcomings
  - It was complex to learn and hard to manage
  - MapReduce API was verbose (a lot of code) and low fault tolerance
  - Intermediate results are written to the local disk for the subsequent stage →
    not good for iterative computing jobs because high cost of I/O operations
  - It good for batch processing, but not good for combining other workloads such as machine learning, streaming, or interactive SQL-like queries
    - → Apache Hive, Storm, Impala, Giraph, Drill, Mahout, etc. → complexity



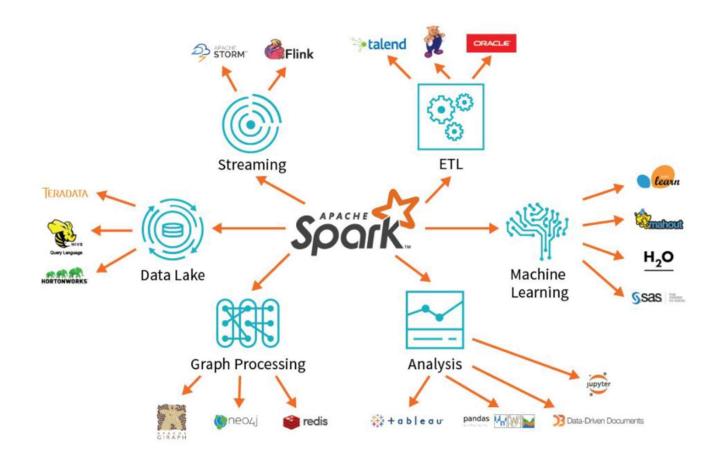
#### Spark's Early Years at AMPLab

- Researchers at UC Berkeley created Spark to deal with Hadoop shortcomings
  - Make it simpler, faster, and easier
  - started in 2009 at the RAD Lab → the AMPLab → RISELab
- 2010 Spark paper demonstrated that it was 10 to 20 times faster than Hadoop MapReduce for certain jobs
- 2013 Spark had gained widespread use and its original creators donated the Spark project to the ASF and formed a company called Databricks.
  - Spark original creators: Matei Zaharia, Ali Ghodsi, Reynold Xin, Patrick Wendell, Ion Stoica, and Andy Konwinski
- Spark 1.0 released in May 2014 (the first major release) by Databricks and the community of open source developers

#### Apache Spark

- A unified engine designed for large-scale distributed data processing, on premises in data centers or in the cloud.
- Spark provides in-memory storage for intermediate computations, making it much faster than Hadoop MapReduce.
- Spark's design philosophy centers around four key characteristics:
  - Speed
  - Ease of use
  - Modularity
  - Extensibility

### Apache Spark – A Unified Analytics Engine



#### Speed

- The framework is optimized to take advantage of memory, multiple cores, multithreading and parallel processing
- Spark builds its query computations as a directed acyclic graph (DAG)
  - computations is decomposed into tasks and executed in parallel
- Spark retained all the intermediate results in memory → huge performance boost

#### Ease of Use

- Spark achieves simplicity by providing a abstraction
  - Low level simple logical data structure: Resilient Distributed Dataset (RDD)
  - higher-level structured data abstractions: DataFrame and Dataset
- Spark offers a simple programming model by a set of transformations and actions as operations

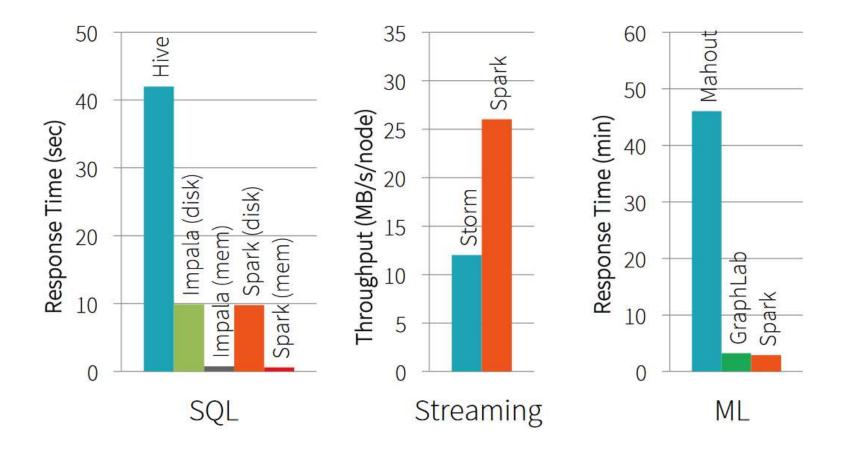
#### Modularity

- Spark operations can be applied across many types of workloads and expressed in many programming languages
  - Scala, Java, Python, SQL, and R
- Spark offers unified libraries with well-documented APIs
  - Spark SQL, Spark Structured Streaming, Spark MLlib, and GraphX

#### Extensibility

- Spark focuses on its fast, parallel computation engine rather than on storage
  - Hadoop included both storage and compute, Spark decouples the two
- Spark can read data stored in many sources and process it all in memory
  - Apache Hadoop, Cassandra, HBase, MongoDB, Hive, RDBMSs, ...
- Spark ecosystem is rich, the community of Spark developers is large

#### Performance vs Specialized Systems



#### Spark Community

• Over 1000 deployments, clusters up to 8000 nodes







































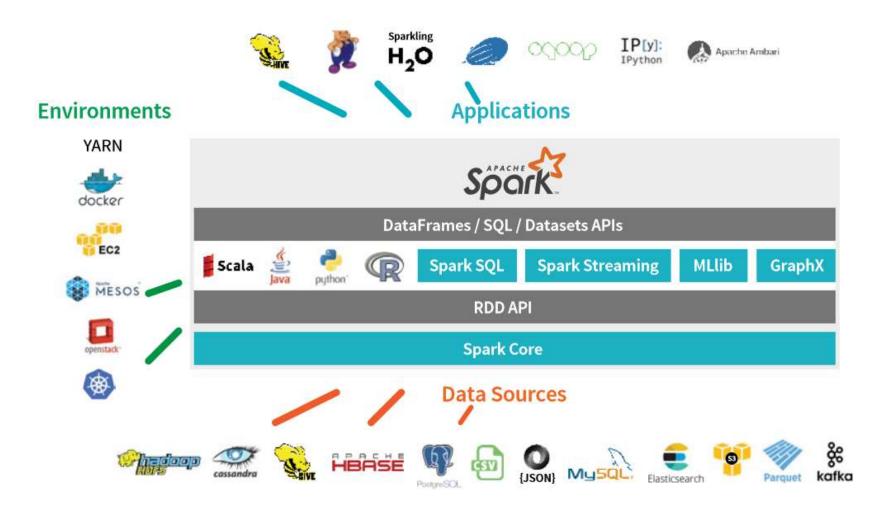








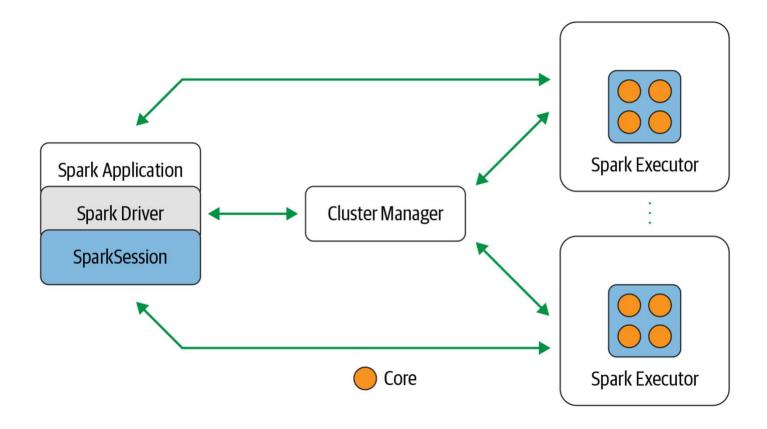
#### Apache Spark's ecosystem



## Apache Spark components and API stack

Spark SQL and Spark Streaming Graph Machine Learning (Structured DataFrames + Processing MLlib **Datasets** Streaming) Graph X Spark Core and Spark SQL Engine R Scala Python SQL Java

## Spark components and architecture

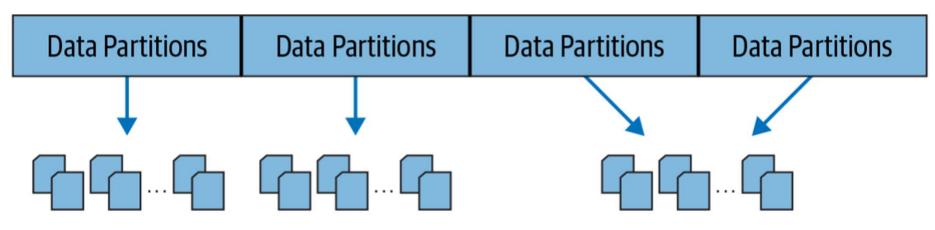


## Spark deployment modes

| Mode              | Spark driver                                       | Spark executor  | Cluster manager  |
|-------------------|--|---|--|
| Local             | Runs on a single JVM, like a laptop or single node | Runs on the same JVM as the driver                        | Runs on the same host  |
| Standalone        | Can run on any node in the cluster                 | Each node in the cluster will launch its own executor JVM | Can be allocated arbitrarily to any host in the cluster  |
| YARN (client)     | Runs on a client, not part of the cluster          | YARN's NodeManager's container                            | YARN's Resource Manager works<br>with YARN's Application Master to<br>allocate the containers on<br>NodeManagers for executors |
| YARN<br>(cluster) | Runs with the YARN Application<br>Master           | Same as YARN client mode                                  | Same as YARN client mode   |
| Kubernetes        | Runs in a Kubernetes pod                           | Each worker runs within its own pod                       | Kubernetes Master  |

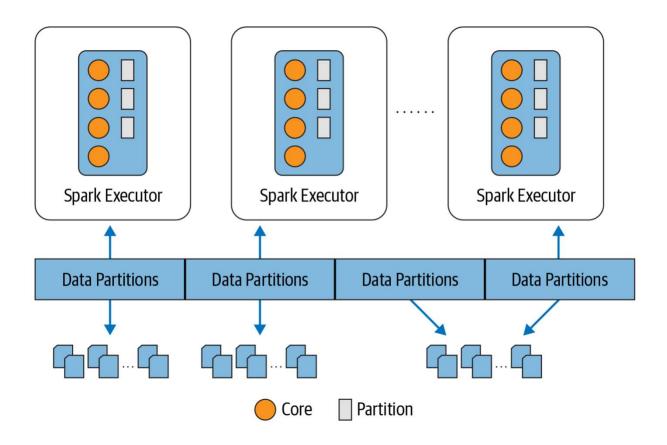
#### Data is distributed across physical machines

#### **Logical Model Across Distributed Storage**



Files Across S3, Azure Blob, or HDFS

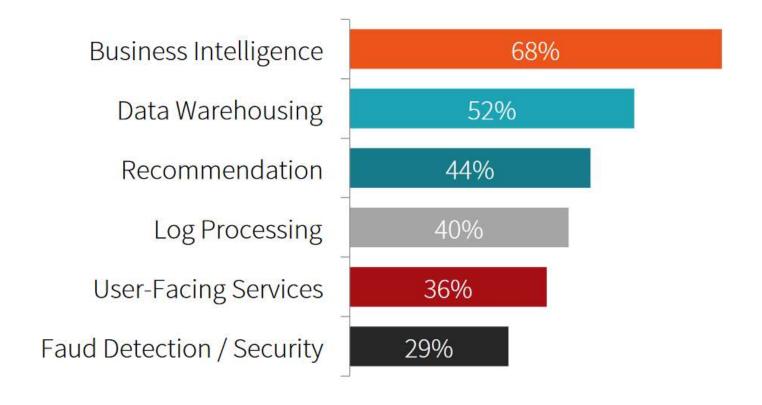
#### Each executor's core gets a partition of data to work on



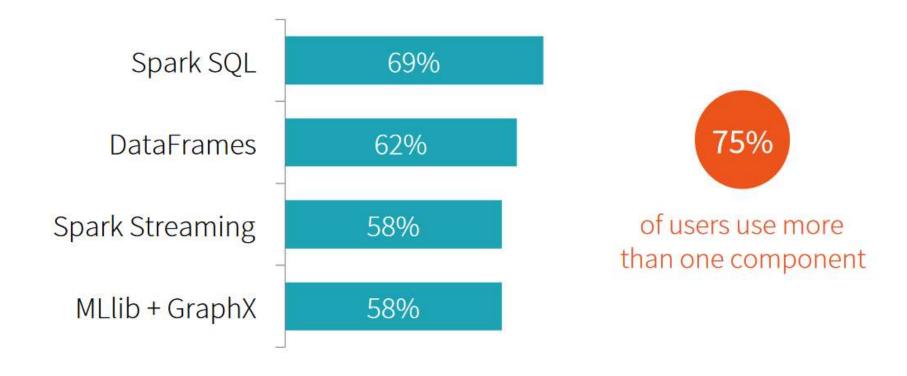
#### Popular Spark use cases

- Processing in parallel large data sets distributed across a cluster
- Performing ad hoc or interactive queries to explore and visualize data sets
- Building, training, and evaluating machine learning models using MLlib
- Implementing end-to-end data pipelines from myriad streams of data
- Analyzing graph data sets and social networks

## Top Applications



## Spark Components Used



#### References

- Jules S. Damji, Brooke Wenig, Tathagata Das & Denny Lee, *Learning Spark: Lightning-Fast Data Analytics*, 2<sup>nd</sup> Edition, O'Reilly, 2020.
- Matei Zaharia, Making Big Data Processing Simple with Spark, ACM Techtalks, December 17, 2015, "Making Big Data Processing Simple with Spark," Matei Zaharia - YouTube