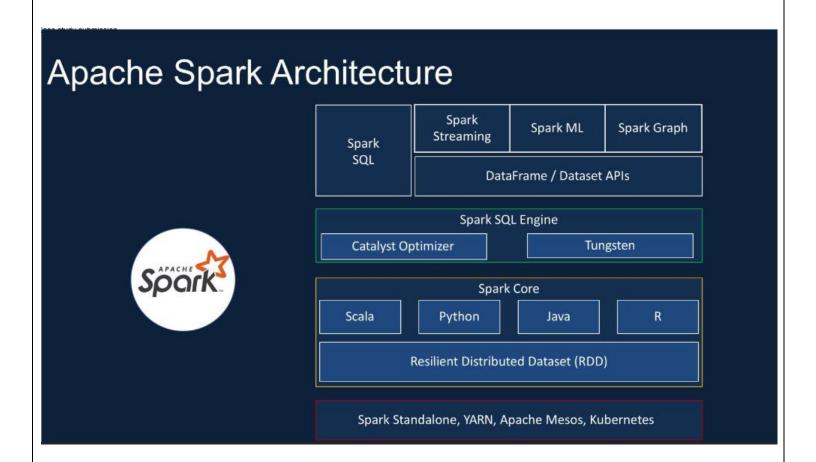
Spark Architecture

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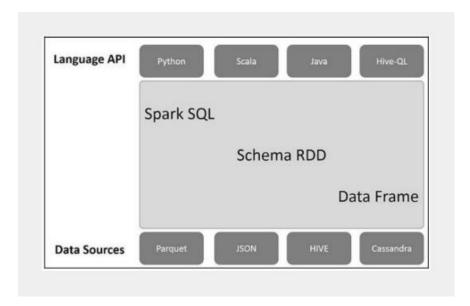
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Spark SQL Overview

- Spark SQL is a module in Apache Spark for structured data processing.
- Introduced in **Spark 1.0** (May 2014), initially developed by **Michael Armbrust and Reynold Xin** from Databricks.
- It adds a programming abstraction called DataFrame and serves as a distributed SQL query engine.

Spark SQL Architecture:



Language API

- Spark SQL supports multiple programming languages, including Python, Scala, Java, and HiveQL (SQL-like query language).
- This compatibility enables developers to work in their preferred language for structured data processing.

Schema RDD

- Spark Core introduces a specialized data structure called RDD (Resilient Distributed Dataset).
- Spark SQL extends RDDs to work with **schemas, tables, and records**, making structured data processing easier.
- SchemaRDD can serve as a temporary table for executing SQL queries.
- Over time, SchemaRDD evolved into the more advanced **DataFrame** abstraction.

Data Sources

- Spark Core typically works with data sources like text files and Avro files.
- Spark SQL, however, supports structured data sources such as Parquet files, JSON documents, Hive tables, and Cassandra databases.

Key Features of Spark SQL

1. Integrated

- Allows mixing of SQL queries with Spark programs seamlessly.
- Queries structured data as distributed datasets (RDDs) using APIs for Python, Scala, and Java.
- o Enables running SQL queries alongside complex analytical algorithms.

2. Unified Data Access

- Supports loading and querying data from diverse sources like Hive tables, Parquet files, JSON files, and more.
- o Provides a single interface (SchemaRDD/DataFrame) for structured data processing.

3. Hive Compatibility

- Executes unmodified Hive queries on existing Hive warehouses.
- o Reuses the Hive frontend and MetaStore for compatibility.
- Can be installed alongside Hive to support Hive queries and User-Defined Functions (UDFs).

4. Standard Connectivity

- o Offers industry-standard connectivity through JDBC and ODBC.
- o Includes a server mode for external applications to execute SQL queries.

5. Scalability

- Supports interactive and long-running queries using the same engine.
- Leverages RDD-based fault tolerance for mid-query recovery and scalability to handle large datasets.

Spark RDD (Resilient Distributed Dataset)

- Fundamental data structure in Spark, representing **immutable distributed collections** of objects.
- RDDs can store data in **memory or on disk** and are distributed across cluster nodes.
- Key Features:
 - o **Partitioned Data**: Divides datasets into partitions for parallel computation.
 - Parallel Transformations: Supports operations like map, filter, and more.
 - o Fault Tolerance: Automatically rebuilds partitions in case of failures.
- RDDs are created by:
 - o Parallelizing collections in the driver program.
 - Referencing external data in storage systems like HDFS, HBase, or shared files.
- Enables faster and more efficient MapReduce operations.

DataFrame and Dataset

DataFrame:

- A distributed collection of data organized into named columns, similar to a relational table.
- o Sources: Hive tables, structured files, external databases, or RDDs.

Dataset:

o A distributed collection of strongly typed data.

Features of DataFrames

- Handles data sizes ranging from KBs to PBs across clusters.
- Supports various data formats (e.g., Avro, CSV, Cassandra) and storage systems (e.g., HDFS, MySQL).
- Uses the Catalyst Optimizer for advanced query optimization and code generation.
- Offers APIs in Python, Scala, Java, and R.
- Integrates with big data tools and frameworks via Spark Core.

SchemaRDD and Data Sources

SchemaRDD:

- A specialized RDD that Spark SQL uses for schema-based operations.
- Also referred to as a DataFrame.

Data Sources:

- Spark Core sources: Text files, Avro files, etc.
- Spark SQL sources: Parquet files, JSON documents, Hive tables, Cassandra databases, etc.