What is Apache Spark?

Apache Spark is an open-source, distributed processing framework, which is used for big data processing and analytics. It utilizes in-memory caching, and optimized query execution for fast analytic queries against data of any size. It provides development APIs in Python, Java, Scala, and R, and supports code reuse across multiple workloads like machine learning, graph processing, interactive queries, real-time analytics, batch processing.

How does Apache Spark work?

Spark was created to address the limitations of MapReduce. A big challenge of MapReduce is the sequential multi-step process it takes to run a job. With each step, MapReduce reads data from the cluster, performs operations, and writes the results back to HDFS. As each step requires a disk read, and write, MapReduce jobs are slower due to the latency of disk I/O. So, Spark solves this issue by doing processing in-memory, reducing the number of steps in a job, and by reusing data across multiple parallel operations. With Spark, only one-step is needed where the data is read into memory, operations performed, and the results are written back, which results in a much faster execution. Spark also reuses data by using an in-memory cache to greatly speed up the machine learning algorithms that repeatedly call a function on the same dataset. This data reuse is achieved through the creation of Dataframes, an abstraction over Resilient Distributed Dataset (RDD), which is a collection of objects that is cached in memory, and reused in multiple Spark operations. This lowers the latency and makes Spark multiple times faster than MapReduce, especially when doing machine learning and interactive analytics.

Features of Apache Spark:

<u>In-Memory Processing(Speed):</u> Spark uses in-memory computing to store and process data in memory, resulting in significantly faster data processing compared to disk-based systems like Hadoop MapReduce.

<u>Distributed Computing:</u> Spark distributes data and computation across multiple nodes in a cluster, enabling parallel processing and efficient resource utilization. Then it automatically manages task scheduling, data partitioning, and fault tolerance, ensuring scalability and high availability.

<u>Broad Programming Language Support:</u> Spark supports multiple programming languages, including Scala, Python, Java, and R. This allows developers and data scientists to work with Spark using their preferred language.

<u>Integration with Big Data Ecosystem:</u> Spark integrates well with various data storage systems like, Hadoop Distributed File System(HDFS), Amazon S3, HBase. It can also read and write data from different sources.

Key concepts in Spark:

Resilient Distributed Datasets(RDD):

Resilient Distributed Datasets are the fundamental data structures in Spark. RDD is an immutable(read-only), fundamental collection of elements or items that can be operated on many devices at the same time i.e. spark parallel processing. Each dataset in an RDD can be divided into logical portions, which are then executed on different nodes of a cluster.

Dataframes and Datasets:

Spark introduced higher-level abstractions called DataFrames and Datasets, built on top of Resilient Distributed Datasets(RDD). DataFrames provide a structured and optimized way to work with structured data, while Datasets offer a type-safe and object-oriented API. Both DataFrames and Datasets support various data manipulation operations.

Components of Spark:

There are five components of Spark:

- Spark Core as the foundation for the platform
- Spark SQL for interactive queries
- Spark Streaming for real-time analytics
- Spark MLlib for machine learning
- Spark GraphX for graph processing

Spark Core:

Spark Core is the foundation of the platform. It is responsible for memory management, fault recovery, scheduling, distributing and monitoring jobs, and interacting with storage systems. Spark Core is exposed through APIs built for Java, Scala, Python and R. These APIs hide the complexity of distributed processing behind simple, high-level operators.

Spark SQL:

Spark SQL is a distributed query engine that provides low-latency, interactive queries upto 100x faster than MapReduce. It provides a programming interface for querying structured and semi structured data using SQL, HiveQL, or DataFrame APIs. It also enables integration with other Spark components and supports various data sources including HIVE, JSON, JDBC.

Spark Streaming:

Spark Streaming enables processing and analyzing real-time streaming data. It ingests the data in mini-batches and performs parallel processing on the live data stream. Spark streaming

integrates with other spark components, allowing integration of batch and real-time processing. Spark streaming supports data from Twitter(Now X), Kafka, Flume, HDFS, and many others.

Spark MLlib:

Spark includes MLlib, a library of algorithms to do machine learning on data at scale. ML models can be trained by data scientists with R or Python on any Hadoop data source, saved using MLlib, and imported into a Java or Scala-based pipeline. This library offers a wide range of algorithms and utilities for classification, regression, clustering, recommendation systems, and more.

Spark GraphX:

Spark's GraphX is a graph processing library that provides an API for graph computation and analysis. GraphX provides ETL(extract, transform, load), exploratory analysis, and iterative graph computation to enable users to interactively build, and transform a graph data structure at scale. It enables efficient graph parallel algorithms and integrates with the rest of the spark ecosystem.