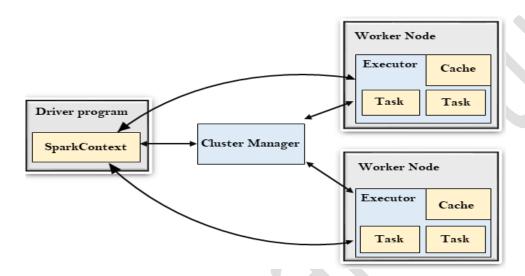
Master Spark Concepts Zero to Big Data Hero:

Detailed Notes on Spark Architecture and Execution Flow

Apache Spark is a powerful open-source distributed computing system that enables fast and efficient data processing. Here's a quick overview of its architecture to help you understand how it works:



Key Components of Spark Architecture

Driver Program

Description: The central coordinator that converts user code into tasks.

Role: Manages the execution of the Spark application and maintains information about the status of tasks.

Cluster Manager

Description: Manages resources across the cluster.

Types: Standalone, YARN, Mesos, Kubernetes.

Role: Allocates resources and schedules tasks on the cluster.

Executors

Description: Workers that run the tasks assigned by the driver program.

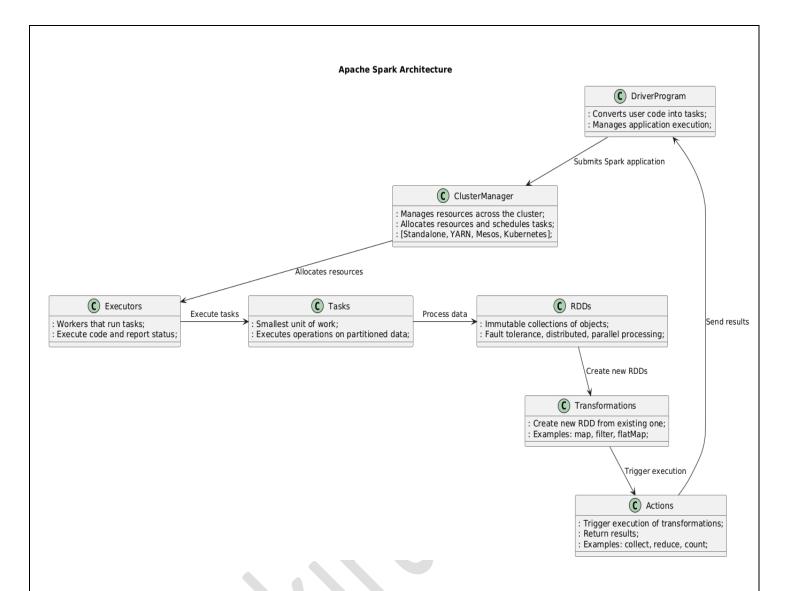
Role: Execute code and report the status of computation and storage.

Tasks

Description: The smallest unit of work in Spark.

Role: Executes individual operations on the partitioned data.





Data Processing in Spark

RDDs (Resilient Distributed Datasets)

Description: Immutable collections of objects that can be processed in parallel. **Features:** Fault tolerance, distributed processing, and parallel execution.

Transformations

Description: Operations that create a new RDD from an existing one. **Examples:** map, filter, flatMap.

Actions

Description: Operations that trigger the execution of transformations and return a result.

Examples: collect, reduce, count.

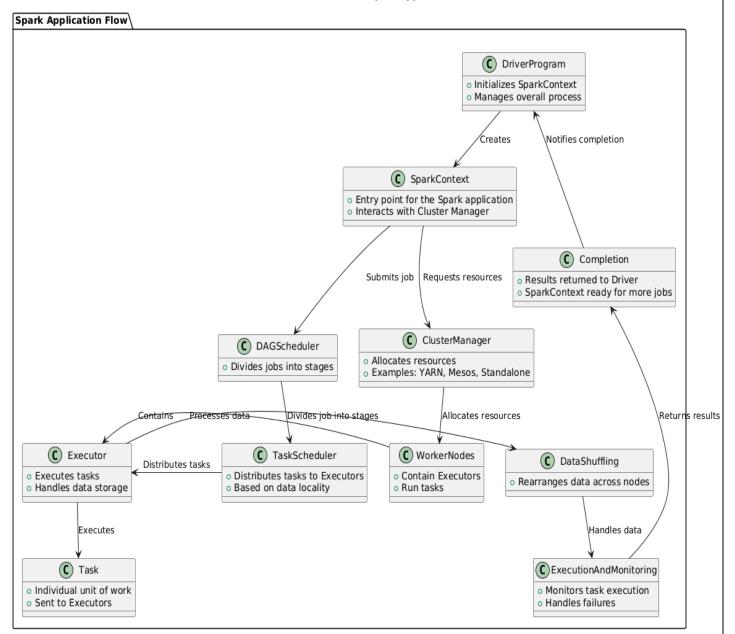


Example Workflow

- 1. The Driver Program submits a Spark application.
- 2. The Cluster Manager allocates resources across the cluster.
- 3. Executors run on worker nodes, executing tasks on the data.
- 4. Transformations create new RDDs, and Actions trigger the execution.
- 5. Results are sent back to the Driver Program.

Execution Flow on Spark Application

Execution Flow of a Spark Application





The execution flow in Apache Spark outlines how data processing occurs from the initial job submission to the final result collection. Here's a step-by-step breakdown:

1. SparkContext Creation

- The driver program starts and initializes the SparkContext.
- This context serves as the main entry point for Spark functionality and manages the entire Spark application.

2. Cluster Manager Interaction

- The SparkContext interacts with the Cluster Manager (e.g., YARN, Mesos, or Standalone).
- o It requests resources (CPU, memory) needed for the application to run.

3. Job Submission

- The user defines transformations and actions within the Spark application.
- A job is created when an action is called (e.g., collect, count).
- At this point, Spark begins to build a logical execution plan.

4. DAG Scheduler

- The Directed Acyclic Graph (DAG) Scheduler takes the job and breaks it down into stages.
- Each stage corresponds to a set of transformations that can be executed in parallel.

5. Task Scheduler

- The Task Scheduler takes the stages defined by the DAG Scheduler and converts them into tasks.
- It distributes these tasks to the executors based on data locality, optimizing resource usage by minimizing data transfer.

6. Executor Execution

- The tasks are executed on the worker nodes by the executors.
- Executors process the data according to the tasks assigned and handle intermediate data storage as required.



7. Data Shuffling

- o If operations like reduceByKey or join are performed, data shuffling may occur.
- This involves redistributing data across the cluster, which can be resourceintensive.

8. Execution and Monitoring

- The driver program continuously monitors the execution of tasks.
- It handles any failures that may occur during processing, re-executing tasks if necessary.

9. Completion

- o Once all tasks are completed, the results are sent back to the driver program.
- The SparkContext is now ready to process more jobs, maintaining a seamless workflow.

Conclusion

Apache Spark's architecture is designed to handle large-scale data processing efficiently and effectively. Understanding its components and workflow can help you leverage its full potential for your big data projects.

Important Interview Question from previous post

- 1. Explain Hadoop Architecture?
- 2. How MapReduce Works?
- 3. Difference between MapReduce and Spark?
- 4. Why Spark is better than MapReduce?
- 5. What are the components of Spark?
- 6. What you mean by JVM in Spark and what are its component?
- 7. What is use of Driver node and Work Node?
- 8. Explain Spark Architecture?
- 9. Explain the flow of execution in Spark?

