







Important Spark Topics for Data Engineer

Important Spark operations and Transformations for Data Engineers.



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In this blog we will break down all the Important Spark Topics For data engineer to crack any data engineering interview and work in any spark related data engineering projects!



1. Spark Core Concepts - RDD Operations:

Transformation Operations:

- 1. `map`: Applies a function to each element of the RDD.
- 2. `filter`: Selects elements that satisfy a given condition.
- 3. `flatMap`: Similar to `map`, but each input item can be mapped to zero or more output items.
- 4. `distinct`: Removes duplicate elements from the RDD.
- 5. `union`: Combines two RDDs into one.
- 6. `groupByKey`: Groups elements by key in a key-value pair RDD.
- 7. `reduceByKey`: Combines values with the same key using a specified reduce function.
- 8. `sortByKey`: Sorts elements based on keys in a key-value pair RDD.
- 9. `mapPartitions`: Applies a function to each partition of the RDD.
- 10. `coalesce`: Reduces the number of partitions in the RDD.

Action Operations:

- 1. `count`: Returns the number of elements in the RDD.
- 2. `collect`: Retrieves all elements of the RDD as an array.
- 3. `reduce`: Aggregates the elements of the RDD using a specified function.
- 4. `first`: Retrieves the first element of the RDD.
- 5. `take`: Retrieves the first N elements of the RDD.
- 6. `foreach`: Applies a function to each element of the RDD.

- 7. `countByValue`: Counts the occurrences of each unique element in the RDD.
- 8. `saveAsTextFile`: Saves the RDD as a text file.
- 9. `saveAsSequenceFile`: Saves the RDD as a Hadoop SequenceFile.
- 10. `foreachPartition`: Applies a function to each partition of the RDD.

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2. DataFrames and Spark SQL Operations:

DataFrame Transformation Operations:

- 1. `select`: Selects specific columns from a DataFrame.
- 2. `filter`: Filters rows based on a specified condition.
- 3. `groupBy`: Groups the DataFrame by specified columns.
- 4. 'join': Joins two DataFrames based on a common column.
- 5. `orderBy`: Sorts the DataFrame based on one or more columns.
- 6. `distinct`: Removes duplicate rows from the DataFrame.
- 7. `withColumn`: Adds a new column or replaces an existing one.
- 8. `drop`: Drops specified columns from the DataFrame.
- 9. `agg`: Performs aggregation operations like sum, count, etc.
- 10. `na` (handling missing data): Provides methods for handling missing data, such as `drop` and `fill`.
- 11. `cache`: Caches the DataFrame in memory for faster access.

DataFrame Action Operations:

- 1. `show`: Displays the first N rows of the DataFrame.
- 2. `count`: Returns the number of rows in the DataFrame.
- 3. `collect`: Retrieves all rows of the DataFrame as an array.
- 4. `head`: Returns the first N rows as a list.
- 5. `first`: Returns the first row of the DataFrame.
- 6. `foreach`: Applies a function to each row of the DataFrame.
- 7. `write`: Writes the DataFrame to an external storage system.
- 8. `writeStream`: Writes the DataFrame as a streaming job.
- 9. `unpersist`: Removes the DataFrame from the cache.

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3. Spark Streaming Operations:

Streaming Operations:

- 1. `updateStateByKey`: Updates the state of the streaming application based on key-value pairs.
- 2. `window`: Performs operations over a sliding window of data.
- 3. `reduceByKeyAndWindow`: Combines values with the same key within a specified window using a reduce function.
- 4. `countByWindow`: Counts the number of elements in each window.
- 5. `foreachRDD`: Applies a function to each RDD in the DStream.
- 6. `transform`: Applies a transformation on the DStream.
- 7. `union`: Combines two DStreams.

8. `dstream` transformations (e.g., `map`, `filter`): Applies transformations on each batch of the DStream.

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4. Structured Streaming Operations:

Structured Streaming Operations:

- 1. `writeStream`: Writes the output of a streaming query to an external storage system.
- 2. `outputMode`: Specifies the output mode of the streaming query (e.g., `append`, `complete`, `update`).
- 3. `trigger`: Defines the trigger interval for the streaming query.
- 4. `format`: Specifies the output format for the streaming query (e.g., `parquet`, `json`).
- 5. `option`: Sets configuration options for the output format.
- 6. **start**: Initiates the execution of the streaming query.
- 7. awaitTermination: Waits for the termination of the streaming query.
- 8. isActive: Checks if the streaming query is actively processing data.
- 9. **recentProgress**: Retrieves information about recent progress of the streaming query.
- 10. stop: Stops the streaming query.

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5. Spark Performance Optimization Operations:

Optimization Techniques:

- 1. **Proper partitioning strategies:** Ensures data is distributed optimally across partitions for parallel processing.
- 2. **Efficient caching and persistence:** Stores intermediate results in memory to avoid recomputation.
- 3. **Effective use of broadcast variables:** Distributes read-only variables efficiently to worker nodes.
- 4. Accumulators for distributed computations: Performs efficient aggregations across distributed data.

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6. Cluster Management Operations:

Integration with Cluster Managers:

- 1. Apache Hadoop YARN integration: Utilizes YARN for resource management and job scheduling.
- 2. Apache Mesos integration: Integrates with Mesos, an open-source cluster manager, for resource sharing.
- 3. Configuring cluster resources for Spark applications: Adjusts resource allocation for optimal Spark application performance.

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7. Integration with Big Data Ecosystem Operations:

Hadoop and Hive Integration:

1. Reading and writing data to/from HDFS: Interacts with the Hadoop Distributed File System for data storage and retrieval.

2. **Interacting with Hive tables in Spark:** Accesses and manipulates data stored in Hive tables using Spark.

Kafka Integration:

1. Reading and writing data to/from Kafka: Connects Spark applications to Apache Kafka for real-time data streaming and processing.

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8. Deployment and Scaling Operations:

Deployment Strategies:

- 1. Local deployment: Runs Spark applications on a single machine for development and testing.
- 2. **Cluster deployment:** Deploys Spark applications on a cluster of machines for distributed processing.
- 3. **Dynamic resource allocation configuration:** Adapts resource allocation based on application needs, which improves efficiency.

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