# **Assignment 18**

Limitations of map reduce-

MapReduce cannot handle:

- Interactive Processing
- Real-time (stream) Processing
- Iterative (delta) Processing
- In-memory Processing
- Graph Processing

**Slow Processing Speed** 

In Hadoop, with a parallel and distributed algorithm, MapReduce process large data sets. There are tasks that need to be performed: Map and Reduce and, MapReduce requires a lot of time to perform these tasks thereby increasing latency. Data is distributed and processed over the cluster in MapReduce which increases the time and reduces processing speed.

Support for Batch Processing only

Hadoop supports batch processing only, it does not process streamed data, and hence overall performance is slower. MapReduce framework of Hadoop does not leverage the memory of the Hadoop cluster to the maximum.

## No Caching

Hadoop is not efficient for caching. In Hadoop, MapReduce cannot cache the intermediate data in memory for a further requirement which diminishes the performance of Hadoop.

### **RDD**

Resilient Distributed Datasets (RDD) is a fundamental data structure of Spark. It is an immutable distributed collection of objects. Each dataset in RDD is divided into logical partitions, which may be computed on different nodes of the cluster.

RDDs are Immutable and partitioned collection of records, which can only be created by coarse grained operations such as map, filter, group by etc. By coarse grained operations, it means that the operations are applied on all elements in a datasets. RDDs can only be created by reading data from a stable storage such as HDFS or by transformations on existing RDDs.

In case of we lose some partition of RDD, we can replay the transformation on that partition—in lineage to achieve the same computation, rather than doing data replication across multiple nodes. This characteristic is biggest benefit of RDD, because it saves a lot of efforts in data management and replication and thus achieves faster computations.

## **RDD Operations**

RDD in Apache Spark supports two types of operations:

Transformation

**Actions** 

#### **Transformations**

Spark RDD Transformations are functions that take an RDD as the input and produce one or many RDDs as the output. They do not change the input RDD (since RDDs are immutable and hence one cannot change it), but always produce one or more new RDDs by applying the computations they represent e.g. Map(), filter(), reduceByKey() etc.

Transformations are lazy operations on an RDD in Apache Spark. It creates one or many new RDDs, which executes when an Action occurs. Hence, Transformation creates a new dataset from an existing one.

Certain transformations can be pipelined which is an optimization method, that Spark uses to improve the performance of computations. There are two kinds of transformations: narrow transformation, wide transformation.

An Action in Spark returns final result of RDD computations. It triggers execution using lineage graph to load the data into original RDD, carry out all intermediate transformations and return final results to Driver program or write it out to file system. Lineage graph is dependency graph of all parallel RDDs of RDD.

Actions are RDD operations that produce non-RDD values. They materialize a value in a Spark program. An Action is one of the ways to send result from executors to the driver. First(), take(), reduce(), collect(), the count() is some of the Actions in spark.