Introduction to Apache Spark



- Imagine we have a really big pile of data, like a mountain of information. Apache Spark is like a powerful machine that helps us dig through that mountain super-fast.
- Instead of using just one small shovel (like a regular computer), Spark gives you lots of shovels (like many computers working together). So, we can break up the mountain into smaller pieces and have all the shovels working at the same time to dig through it.
- Spark is really good at handling big data and doing all sorts of tasks with it, like analyzing, processing, and finding important stuff within it. It's like having a team of super-fast data detectives helping you make sense of all that information!

Spark v/s Hadoop?

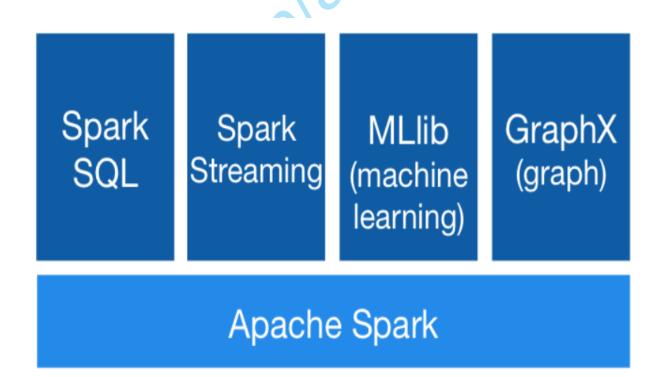
- Imagine we have a huge garden full of different types of plants. We want to water all the plants efficiently.
- Hadoop is like a sprinkler system. It's effective at covering large areas of the garden with water, but it might take a bit of time to reach all the plants, especially if they're spread out far apart.



- Apache Spark is like having a team of gardeners with hoses. Each gardener focuses on watering a specific section of the garden, and they all work together simultaneously. This way, the entire garden gets watered much faster than with just the sprinkler system.
- In this analogy, Hadoop provides a systematic approach to data processing, while Apache Spark offers parallel processing capabilities, allowing for faster and more efficient handling of large volumes of data.

What is Spark?

- Apache Spark is an open-source cluster computing framework. Its primary purpose is to handle the real-time generated data.
- Spark takes large data sets, break them down into smaller, manageable parts and then processes these parts across multiple computers at the same time.
- Spark was built on the top of the Hadoop MapReduce. It was optimized to run in memory whereas alternative approaches like Hadoop's MapReduce writes data to and from computer hard drives. So, Spark process the data much quicker than other alternatives.



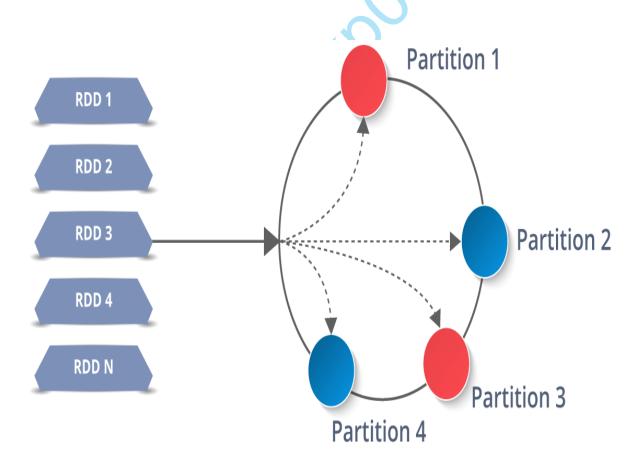
History of Apache Spark

- The Spark was initiated by Matei Zaharia at UC Berkeley's AMP Lab in 2009. It was open sourced in 2010 under a BSD license.
- In 2013, the project was acquired by Apache Software Foundation. In 2014, the Spark emerged as a Top-Level Apache Project.

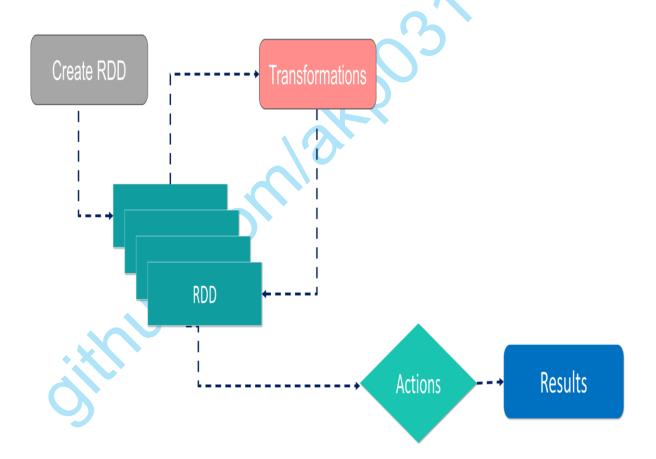


Resilient Distributed Datasets (RDD)

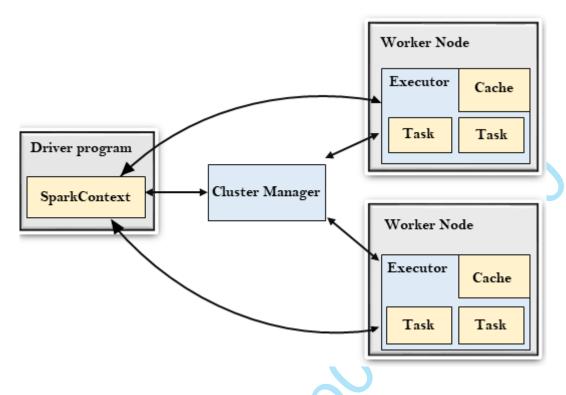
- RDDs are the building blocks of any Spark application. RDDs Stands for:
- Think of an RDD as a big, flexible bag that holds our data. This bag is special because it can be spread out over many computers (distributed) and it's resilient, meaning even if one of those computers goes stopped or malfunctioned, our data is safe and can be reconstructed from the other computers.
 - **Resilient:** If a computer holding part of the RDD fails, Spark can reconstruct that part using the data stored on other computers. It's like having multiple backups, so even if one fails, our data is still accessible.
 - Distributed: RDDs can be split into smaller chunks and stored on multiple computers. This allows Spark to process the data in parallel, making things faster.
 - o **Dataset:** It's just our data, whether it's numbers, text, or anything else. RDDs can hold any type of data we want to work with.



- Features of RDD:
 - o In-Memory
 - o Immutable
 - Lazy Evaluated
 - o Parallel
- With RDDs, we can perform two types of operations:
 - Transformations: They are the operations that are applied to create a new RDD.
 - **Actions:** They are applied on an RDD to instruct Apache Spark to apply computation and pass the result back to the driver.

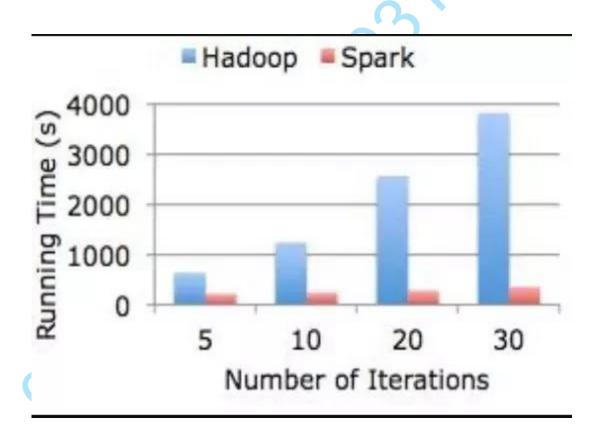


Spark Architecture



- In our master node, we have the **driver program**, which drives our application. The code we are writing behaves as a driver program or if we are using the interactive shell, the shell acts as the driver program.
- Inside the driver program, the first thing we do is, we create a **Spark Context**. Assume that the Spark context is a gateway to all the Spark functionalities. It is similar to our database connection. Any command we execute in our database goes through the database connection. Likewise, anything we do on Spark goes through Spark context.
- Now, this Spark context works with the cluster manager to manage various jobs. The driver program & Spark context takes care of the job execution within the cluster. A job is split into multiple tasks which are distributed over the worker node. Anytime an RDD is created in Spark context, it can be distributed across various nodes and can be cached there.
- Worker nodes are the slave nodes whose job is to basically execute the tasks. These tasks are then executed on the partitioned RDDs in the worker node and hence returns back the result to the Spark Context.

- Spark Context takes the job, breaks the job in tasks and distribute them to the worker nodes. These tasks work on the partitioned RDD, perform operations, collect the results and return to the main Spark Context.
- If we increase the number of workers, then we can divide jobs into more partitions and execute them parallelly over multiple systems. It will be a lot faster.
- With the increase in the number of workers, memory size will also increase & we can cache the jobs to execute it faster.



Different Ways to Create Resilient Distributed Datasets (RDD)

Parallelize Method	val data = Seq(("A", 1), ("B", 2), ("C", 3))					
	val r1 = spark.sparkContext.parallelize(data)					
	val r1 = spark.sparkContext.parallelize(Seq(("A",1),("B",2),("C",3)))					
	val r1 = spark.sparkContext.parallelize(List(("A", 1), ("B", 2), ("C", 3)))					
	val r1 = spark.sparkContext.parallelize(Array(("A", 1), ("B", 2), ("C", 3)))					
	<pre>val r1 = spark.sparkContext.parallelize(Vector(("A", 1), ("B", 2), ("C", 3)))</pre>					
	val r1 = spark.sparkContext.parallelize(1 to 100)					
	r1.foreach(println)					
	val data = List(("A", 1), ("B", 2), ("C", 3))					
	val r1 = spark.sparkContext.parallelize(data)					
Text File Method	val r2 = spark.sparkContext.textFile("C:/Spark/sparkfiles/g2.txt")					
	r2.collect()					
From Existing RDD	We can use transformations like map, flatmap, filter to create a new RDD from an existing one.					
	val r3 = r2.flatMap(split(","))					
	r3.collect()					

From existing	val r4 = spark.range(20).toDF().rdd
DataFrames and	
DataSet	val r4 = dataframe .rdd
	r4.collect()

Transformation in Spark RDD API

```
Transformation applies a function to each row in a Data Frame/Dataset and
map()
                       returns the new transformed Dataset.
                       val maprdd = sc.textFile("C:/Spark/sparkfiles/example.txt")
                       hadoop is slow
                        spark is fast
                       both are good
                       val maprdd2 = maprdd.map(r=>r.split(" "))
                       maprdd2.collect()
                        scala> val maprdd = sc.textFile("C:/Spark/sparkfiles/example.txt")
maprdd: org.apache.spark.rdd.RDD[String] = C:/Spark/sparkfiles/example.txt MapPartitionsRDD[1] at textFile at <consol
                        cala> maprdd.show()
console>:24: error: value show is not a member of org.apache.spark.rdd.RDD[String]
    maprdd.show()
                        .cala> val maprdd2 = maprdd.map(r=>r.split(" "))
aprdd2: org.apache.spark.rdd.RDD[Array[String]] = MapPartitionsRDD[2] at map at <console>:23
                        scala> maprdd2.collect()
res1: Array[Array[String]] = Array(Array(hadoop, is, slow), Array(spark, is, fast), Array(both, are, good))
                        cala>
                       map maintains a one-to-one relationship between input and output elements
```

RDD with some numbers, and we want to double each number using the map operation.

```
// Create an RDD with some numbers
val numbersRDD = sc.parallelize(List(1, 2, 3, 4, 5))

// Use map to double each number
val doubledNumbersRDD = numbersRDD.map(x => x * 2)

// Collect the results to print them
val doubledNumbers = doubledNumbersRDD.collect()

// Print the doubled numbers
doubledNumbers.foreach(println)
```

```
scala> :paste
// Entering paste mode (ctrl-D to finish)
// Create an RDD with some numbers
val numbersRDD = sc.parallelize(List(1, 2, 3, 4, 5))
// Use map to double each number
val doubledNumbersRDD = numbersRDD.map(x => x * 2)
// Collect the results to print them
val doubledNumbers = doubledNumbersRDD.collect()
// Print the doubled numbers
doubledNumbers.foreach(println)

// Exiting paste mode, now interpreting.

2
4
6
8
10
numbersRDD: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[3] at parallelize at <pastie>:24
doubledNumbersRDD: org.apache.spark.rdd.RDD[Int] = MapPartitionsRDD[4] at map at <pastie>:27
doubledNumbers: Array[Int] = Array(2, 4, 6, 8, 10)
scala> __
```

flatMap()

Transformation flattens the Data Frame/Dataset after applying the function on every element and returns a new transformed Dataset.

val flatmaprdd = sc.textFile("C:/Spark/sparkfiles/example.txt")

hadoop is slow spark is fast both are good

val flatmaprdd2 = flatmaprdd.flatMap(r=>r.split(" "))
flatmaprdd2.collect()

```
scala> :paste
// Entering paste mode (ctrl-D to finish)
val flatmaprdd = sc.textFile("C:/Spark/sparkfiles/example.txt")
val flatmaprdd2 = flatmaprdd.flatMap(r=>r.split(" "))
flatmaprdd2.collect()

// Exiting paste mode, now interpreting.
flatmaprdd: org.apache.spark.rdd.RDD[String] = C:/Spark/sparkfiles/example.txt MapPartitionsRDD[6] at textFile at <pastie>:23
flatmaprdd2: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[7] at flatMap at <pastie>:24
res3: Array[String] = Array(hadoop, is, slow, spark, is, fast, both, are, good)
scala> _
```

flatMap allows for a one-to-many relationship, and can generate zero or more output elements

Suppose we have an RDD with sentences, and we want to split each sentence into words using flatMap.

// Sample sentences

val sentencesRDD = sc.parallelize(List("Apache Spark is awesome", "It makes big data processing easy"))

// Apply flatMap to split each sentence into words

```
val wordsRDD = sentencesRDD.flatMap(sentence => sentence.split(" "))

// Collect and print the words
val wordsArray = wordsRDD.collect()
wordsArray.foreach(println)

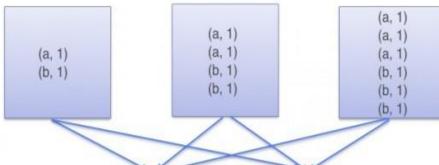
scala> :paste
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// Sample sentences
val sentencesRDD = sc.parallelize(List("Apache Spark is awesome", "It makes big data processing easy"))
```

```
// Apply flatMap to split each sentence into words
val wordsRDD = sentencesRDD.flatMap(sentence => sentence.split(" "))
// Collect and print the words
val wordsArray = wordsRDD.collect()
wordsArray.foreach(println)
// Exiting paste mode, now interpreting.
Apache
Spark
awesome
makes
big
data
processing
easy
sentencesRDD: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[8] at parallelize at <pastie>:24
wordsRDD: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[9] at flatMap at <pastie>:27
 ordsArray: Array[String] = Array(Apache, Spark, is, awesome, It, makes, big, data, processing, easy)
```

groupByKey() groupByKey shuffles the entire dataset across the network based on the key.





```
      (a, 1)
      (b, 1)

      (b, 1)
      (b, 6)

      (a, 1)
      (b, 1)

      (b, 1)
      (b, 1)

      (b, 1)
      (b, 1)
```

```
val rdd1 = sc.parallelize(Array(("a",1),("a",2),("b",1),("b",2)))
rdd1.groupByKey().collect()
```

```
scala> :paste
// Entering paste mode (ctrl-D to finish)
val rdd1 = sc.parallelize(Array(("a",1),("a",2),("b",1),("b",2)))
rdd1.groupByKey().collect()
// Exiting paste mode, now interpreting.
rdd1: org.apache.spark.rdd.RDD[(String, Int)] = ParallelCollectionRDD[10] at parallelize at <pastie>
res6: Array[(String, Iterable[Int])] = Array((a,CompactBuffer(1, 2)), (b,CompactBuffer(1, 2)))
scala>
```

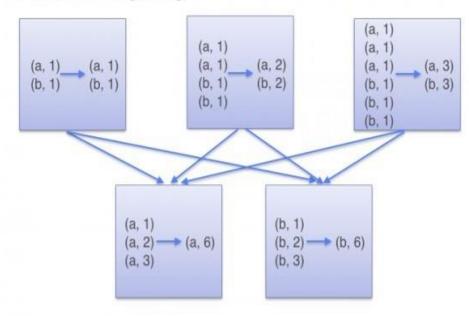
A simple example to understand groupBykey():

// Example RDD of books

```
val books = sc.parallelize(Seq(
 ("Fiction", "Book1"),
 ("Fiction", "Book2"),
 ("Non-Fiction", "Book3"),
 ("Fiction", "Book4"),
 ("Non-Fiction", "Book5")
))
// Applying reduceByKey to count the number of books in each genre
val bookCounts = books.map(genre => (genre._1, 1)).reduceByKey(_ + _)
// Collecting and printing the result
bookCounts.collect().foreach(println)
Output:
(Fiction,3)
(Non-Fiction, 2)
```

reduceByKey() reduceByKey will compute local sums for each key in each partition and combine those local sums into larger sums after shuffling.





```
val rdd2 = sc.parallelize(Array(("a",1),("a",2),("b",1),("b",2)))
rdd2.reduceByKey(_+_).collect()
```

```
scala> :paste
// Entering paste mode (ctrl-D to finish)
val rdd2 = sc.parallelize(Array(("a",1),("a",2),("b",1),("b",2)))
rdd2.reduceByKey(_+_).collect()
// Exiting paste mode, now interpreting.
rdd2: org.apache.spark.rdd.RDD[(String, Int)] = ParallelCollectionRDD[12] at parallelize at <pastie>:23
res7: Array[(String, Int)] = Array((a,3), (b,3))
scala> _
```

A simple example of reducing duplicate numbers in an RDD using reduceByKey.

// Example RDD with numbers

```
val numbersRDD = sc.parallelize(Seq(
 (1, "one"),
 (2, "two"),
 (3, "three"),
 (1, "uno"),
 (2, "dos"),
 (3, "tres"),
 (1, "eins")
))
// Apply reduceByKey to count the occurrences of each number
val countRDD = numbersRDD.map(num => (num._1, 1)).reduceByKey(_ +
_)
// Collect and print the counts
val countArray = countRDD.collect()
countArray.foreach(println)
Output:
(1,3)
(2,2)
(3,2)
```

Action in Spark RDD API

Apache Spark RDD Actions are operations that trigger the execution of transformations and return results to the driver program or write data to external storage. Unlike transformations, which are lazily evaluated, actions are eagerly executed. Here are some common RDD actions in Apache Spark, along with examples:

collect()

This action retrieves all elements of the RDD and brings them to the driver program as an array.

val rdd = sc.parallelize(Seq(1, 2, 3, 4, 5))

val result = rdd.collect()

println(result.mkString(", "))

Output: 1, 2, 3, 4, 5

println(result.mkString("/"))

Output: 1/2/3/4/5

Note:

The **mkString(", ")** method is used in Scala to concatenate the elements of a collection into a single string, with each element separated by a specified delimiter.

For example, if we have a list of strings `["one", "two", "three"]` and we call `mkString(", ")`, it will join the elements of the list into a single string with each element separated by a comma and a space, resulting in `"one, two, three"`.

In the context of the examples provided earlier, 'mkString(", ")' is used to join the elements of an array into a single string with a comma and a space between each element, before printing it out. This is helpful for formatting the output when printing collections.

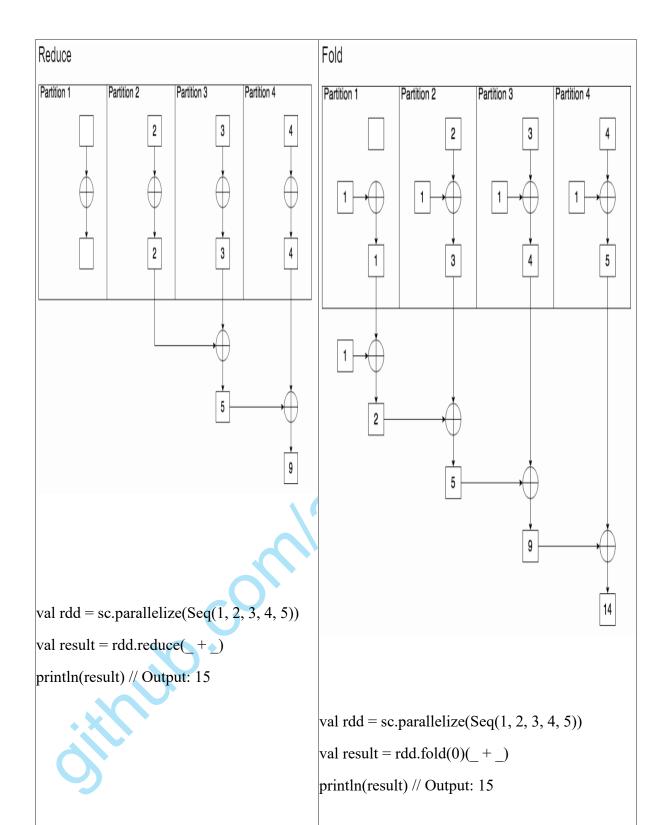
count()	This action counts the number of elements in the RDD.				
	val rdd = sc.parallelize(Seq(1, 2, 3, 4, 5))				
	val result = rdd.count()				
	println(result)				
take(n)	This action retrieves the first n elements of the RDD and returns them as an array.				
	val rdd = $sc.parallelize(Seq(1, 2, 3, 4, 5))$				
	val result = rdd.take(3)				
	println(result.mkString(", "))				
	Output: 1, 2, 3				
first()	This action returns the first element of RDD.				
	val rdd = $sc.parallelize(Seq(1, 2, 3, 4, 5))$				
	val result = rdd.first()				
	println(result)				
9,	Output: 1				
reduce(func)	This action aggregates the elements of the RDD using a specified binary function func.				
	val rdd = sc.parallelize(Seq(1, 2, 3, 4, 5))				

	val result = rdd.reduce(_ + _)
	println(result)
	Output, 15
	Output: 15
saveAsTextFile(path)	This action saves the RDD as a text file or files in the specified directory.
	val rdd = sc.parallelize(Seq("Hello", "World", "Apache", "Spark"))
	rdd.saveAsTextFile("C:/Spark/sparkfiles/output_dir")
	Note: output_dir must be created automatically, we shouldn't create
	by own.
max()	This action returns the maximum element in the RDD.
	val rdd = $sc.parallelize(Seq(1, 2, 3, 4, 5))$
	val result = rdd.max()
V	println(result)
	Output: 5
	Output. 3
min()	This action returns the minimum element in the RDD.
	val rdd = $sc.parallelize(Seq(1, 2, 3, 4, 5))$
	val result = rdd.min()
	println(result)

	Output: 1				
sum()	This action returns the sum of all elements in the RDD.				
	val rdd = sc.parallelize(Seq(1, 2, 3, 4, 5))				
	<pre>val result = rdd.sum() println(result)</pre>				
	Output: 15				
standardDeviation()	This action returns the standard deviation of the elements in the RDD.				
	val rdd = sc.parallelize(Seq $(1, 2, 3, 4, 5)$)				
	val result = rdd.stdev()				
	println(result)				
countByKey()	This action is specific to Pair RDDs (RDDs containing key-value pairs). It counts the number of occurrences of each key and returns the result as a Map.				
OH!	val rdd = sc.parallelize(Seq(("a", 1), ("b", 2), ("a", 3), ("c", 4), ("b", 5)))				
	<pre>val result = rdd.countByKey() println(result)</pre>				
	Output: Map(a -> 2, b -> 2, c -> 1)				

Difference between Reduce and Fold in Apache Spark

Reduce	Fold		
repeatedly applies it to pairs of elements in	fold also takes a binary function as input and repeatedly applies it to pairs of elements in the RDD until only a single value remains, similar to reduce.		
It combines elements pairwise to produce a single result.	However, fold additionally requires an initial value , which serves as the starting value for the aggregation.		
The result of reduce is the same type as the elements in the RDD.	It combines elements pairwise with the initial value to produce a single result.		
It doesn't require an initial value because it uses the first element of the RDD as the initial value for the first invocation of the binary function.	The result of fold can be a different type than the elements in the RDD.		



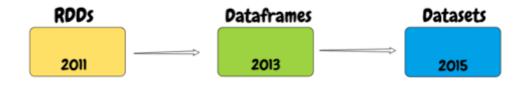
```
val rdd = sc.parallelize(Seq(1, 2, 3, 4, 5))
val result = rdd.reduce(_ + _)
println(result) // Output: 15

// Exiting paste mode, now interpreting.

scala> val rdd = sc.parallelize(Seq(1, 2, 3, 4, 5))
rdd: org.apache.spark.rdd.RDD[Int] = Parallelc
scala> val result = rdd.fold(1)(_ + _)
result: Int = 20

scala> println(result)
20
```

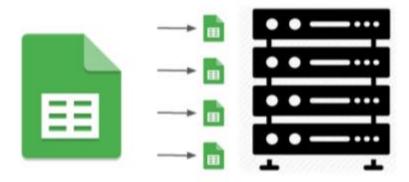
Difference between RDD, DataFrames, and Datasets



RDD RDDs or Resilient Distributed Datasets is the fundamental data structure of the Spark. It is the collection of objects which is capable of storing the data partitioned across the multiple nodes of the cluster and also allows them to do processing in parallel.

It is fault-tolerant if you perform multiple transformations on the RDD and then due to any reason any node fails. The RDD, in that case, is capable of recovering automatically.

RDD



- An RDD is like a box of assorted items.
- Just like a box can contain anything from books to toys, an RDD can contain any type of data, whether structured or unstructured.
- We can freely manipulate the contents of the box, moving items around or adding new ones as needed.
- However, because the contents are not organized, finding specific items might take some effort.

DataFrames

- DataFrame is a higher-level abstraction built on top of RDDs.
- It represents a distributed collection of data organized into named columns, similar to a table in a relational database.
- DataFrame provides a more structured and SQL-like interface for data processing.
- It uses a catalyst optimizer for optimization.
- DataFrame API is available in multiple languages like Scala, Java, Python, and R.

A DataFrame is like a well-organized bookshelf.

Each shelf represents a column, and each book represents a row of data.

The books are neatly arranged, making it easy to find and analyze information.

We can quickly query the bookshelf to find specific books or perform operations on entire rows or columns.

Additionally, if we add or remove books, the bookshelf adjusts itself automatically to maintain organization.

Datasets

- DataSet is another higher-level abstraction introduced in Spark 1.6, combining the benefits of RDDs and DataFrames.
- It provides the type-safety and object-oriented programming model of RDDs, along with the performance optimizations of DataFrames.
- DataSet allows us to work with strongly typed data structures, providing compile-time type checking and improved developer productivity.
- DataSet API is available in Scala and Java, but not in Python or R.

A DataSet is like a labelled storage system.

Imagine each item in the storage system is labelled with its type (e.g., book, toy, tool).

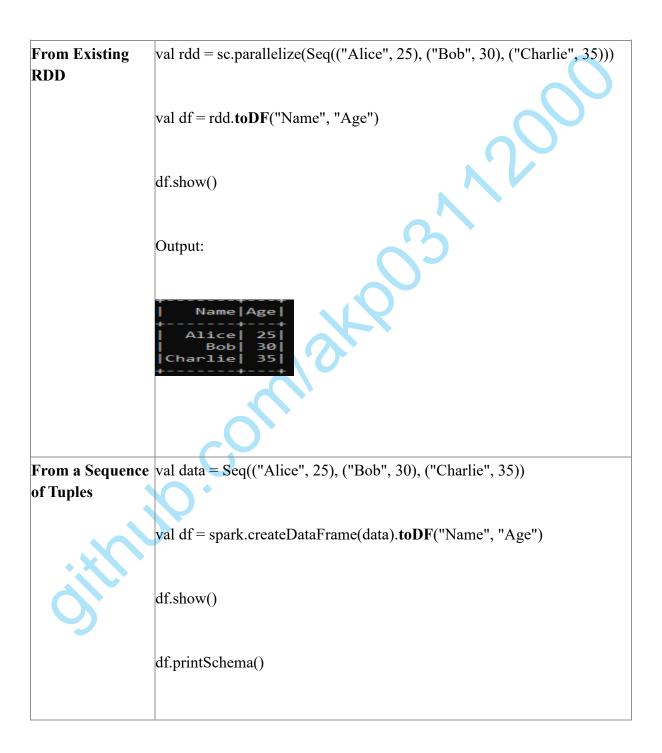
This labelling ensures that each item is handled appropriately and provides clarity on what each item represents.

We can confidently interact with the items, knowing their exact type and properties.

If we need to perform a specific task, such as finding all books, we can quickly locate and process them based on their labels.

Different Ways to Create DataFrames

import spark.implicits._



+	+			+	
_corrupt_record	petalLength	petalWidth	sepalLength	sepalWidth	species
+	+		+	 +	·
] [NULL	NULL	NULL	NULL	NULI
NULL	1.4	0.2	5.1	3.5	setosa
NULL	1.4	0.2	4.9	3.0	setosa
NULL	1.3	0.2	4.7	3.2	setosa
NULL	1.5	0.2	4.6	3.1	setosa
NULL	1.4	0.2	5.0	3.6	setosa
NULL	1.7	0.4	5.4	3.9	setosa
NULL	1.4	0.3	4.6	3.4	setosa
NULL	1.5	0.2	5.0	3.4	setosa
NULL	1.4	0.2	4.4	2.9	setosa
NULL	1.5	0.1	4.9	3.1	setosa
NULL	1.5	0.2	5.4	3.7	setosa
NULL	1.6	0.2	4.8	3.4	setosa
NULL	1.4	0.1	4.8	3.0	setosa
NULL	1.1	0.1	4.3	3.0	setosa
NULL	1.2	0.2	5.8	4.0	setosa
NULL	1.5	0.4	5.7	4.4	setosa
NULL	1.3	0.4	5.4	3.9	setosa
NULL	1.4	0.3	5.1	3.5	setosa
NULL	1.7	0.3	5.7	3.8	setosa
+	+	+	+	+	

val df4 = spark.read.csv("C:/Spark/sparkfiles/iris.csv")

df4.show()

	+	++			
_c0	_c1	_c2	_c3	_c4	_c
	+	++		+	
Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Specie
1	5.1	3.5	1.4	0.2	Iris-setos
2	4.9	3.0	1.4	0.2	Iris-setos
3	4.7	3.2	1.3	0.2	Iris-setos
4	4.6	3.1	1.5	0.2	Iris-setos
5	5.0	3.6	1.4	0.2	Iris-setos
6	5.4	3.9	1.7	0.4	Iris-setos
7	4.6	3.4	1.4	0.3	Iris-setos
8	5.0	3.4	1.5	0.2	Iris-setos
9	4.4	2.9	1.4	0.2	Iris-setos
10	4.9	3.1	1.5		Iris-setos
11	5.4	3.7	1.5	0.2	Iris-setos
12	4.8	3.4	1.6	0.2	Iris-setos
13	4.8	3.0	1.4	0.1	Iris-setos
14	4.3	3.0	1.1	0.1	Iris-setos
15	5.8	4.0	1.2	0.2	Iris-setos
16	5.7	4.4	1.5	0.4	Iris-setos
17	5.4	3.9	1.3	0.4	Iris-setos
18	5.1	3.5	1.4	0.3	Iris-setos
19	5.7	3.8	1.7	0.3	Iris-setos
	+	·			

val df4 = spark.read.text("C:/Spark/sparkfiles/iris.txt")

df4.show()

When reading a CSV file with Spark, we can specify whether the file has a header or not using the option method with the header parameter set to true or false. If the file has a header, setting header to true will instruct Spark to use the first line of the file as the header and skip it when reading the data.

```
val df = spark.read.option("header",
"true").csv("C:/Spark/sparkfiles/iris.csv")
```

val df = spark.read.option("header", "true").option("inferSchema",
"true").csv("C:/Spark/sparkfiles/iris.csv")

++					
Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Speci
1 1	5.1	3.5	1.4	0.2	Iris-seto
2	4.9	3.0	1.4		Iris-seto
3	4.7	3.2	1.3	0.2	Iris-seto
4	4.6	3.1	1.5	0.2	Iris-seto
5	5.0	3.6	1.4	0.2	Iris-seto
6	5.4	3.9	1.7	0.4	Iris-seto
7	4.6	3.4	1.4	0.3	Iris-seto
8	5.0	3.4	1.5	0.2	Iris-seto
9	4.4	2.9	1.4	0.2	Iris-seto
10	4.9	3.1	1.5	0.1	Iris-seto
11	5.4		1.5	0.2	Iris-seto
12	4.8	3.4	1.6	0.2	Iris-seto
13	4.8	3.0	1.4	0.1	Iris-seto
14	4.3	3.0	1.1	0.1	Iris-seto
15	5.8	4.0	1.2	0.2	Iris-seto
16	5.7	4.4	1.5	0.4	Iris-seto
17	5.4	3.9	1.3	0.4	Iris-seto
18	5.1	3.5	1.4	0.3	Iris-seto
19	5.7	3.8	1.7	0.3	Iris-seto
20	5.1	3.8	1.5	0.3	Iris-seto
++					
	_	1,0			

DataFrames API Queries

```
import org.apache.spark.sql.SparkSession

Create SparkSession

val spark = SparkSession.builder()

.appName("DataFrame API Example")

.master("local[*]")

.getOrCreate()

Synthetic data
```

```
val data = Seq(
 ("Alice", 25, "Sales"),
 ("Bob", 30, "Marketing"),
 ("Charlie", 35, "HR"),
 ("David", 40, "Sales"),
 ("Emma", 45, "Marketing"),
 ("Frank", 50, "HR")
Create DataFrame
val df = spark.createDataFrame(data).toDF("Name", "Age", "Department")
Show DataFrame
df.show()
 cala> df.show()
    Name | Age | Department |
                    Sales
           30 Marketing
           35|
 Charlie
           40
                    Sales
           45 Marketing
   Frank 50
```

import org.apache.spark.sql.{SparkSession, DataFrame}

```
Select
            Select specific columns
Columns
            df.select("col1", "col2")
            Select all columns
            df.select("*")
            Example-
            df.select("Name", "Age").show()
             scala> df.select("Name", "Age").show()
                 Name | Age |
                Alice | 25
                  Bob | 30 |
              Charlie 35
                David 40
Filter Rows Filter rows based on a condition
            df.filter($"col1" > 10)
             Filter rows using SQL expression
            df.filter("col1 > 10")
             Example-
            df.filter(\$"Age" > 30).show()
```

```
scala> df.filter($"Age" > 30).show()
+-----+
| Name|Age|Department|
+----+
|Charlie| 35| HR|
| David| 40| Sales|
| Emma| 45| Marketing|
| Frank| 50| HR|
```

Using Column Expression: We can construct the condition using column expressions. This involves using the col function to refer to the column:

```
scala> posDF.filter(col("Name") === "Alice").show()
+----+
| Name|Position|
+----+
|Alice| Manager|
+----+
scala> posDF.filter(col("Position") === "Executive").show()
+---+----+
|Name| Position|
+---+-----+
| Bob|Executive|
+----+------+
```

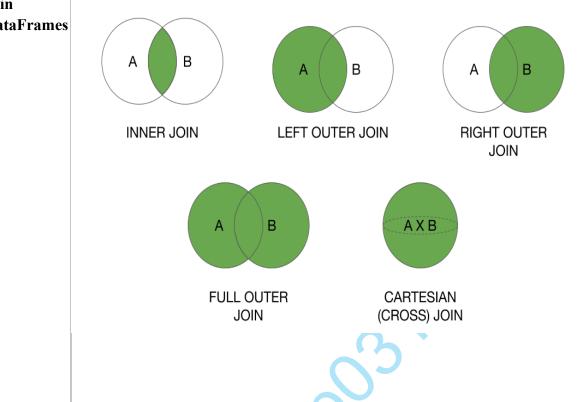
Group By and Aggregate

Group by a column and count the occurrences df.groupBy("col1").count()

Aggregate with custom aggregation functions import org.apache.spark.sql.functions._
df.groupBy("col1").agg(avg("col2"), sum("col3"))

```
Example-
           df.groupBy("Department").count().show()
            scala> df.groupBy("Department").count().show()
             Department | count |
                             2
                  Sales
                             2
                     HR |
              Marketing
Sort Data
           Sort data by one or more columns
           df.sort("col1")
           Sort data in descending order
           df.sort(desc("col1"))
           Example-
           df.sort("Age").show()
            scala> df.sort("Age").show()
                Name | Age | Department |
               Alice 25
                 Bob | 30 |
                           Marketing
             Charlie 35
                                   HR
               David 40
                                Sales
                      45
                Emma
                           Marketing
               Frank 50
```

Join **DataFrames**



Inner join with another DataFrame

df.join(otherDf, df("col1") === otherDf("col1"), "inner")

Left outer join

df.join(otherDf, Seq("col1"), "left_outer")

import org.apache.spark.sql.{SparkSession, DataFrame}

Example-

```
val otherData = Seq(
 ("Alice", "Manager"),
 ("Bob", "Executive"),
 ("Charlie", "Analyst")
```

```
val otherDf = spark.createDataFrame(otherData).toDF("Name", "Position")
           otherDf.show()
           scala> otherDf.show()
                Name | Position
               Alice
                      Manager
                 Bob Executive
             Charlie
                       Analyst
           df.join(otherDf, Seq("Name"), "left_outer").show()
           scala> df.join(otherDf, Seq("Name"),
                                                   "left_outer").show()
                Name | Age | Department | Position |
                               Sales
               Alice 25
                                       Manager
                 Bob | 30 |
                          Marketing Executive
             Charlie | 35
                                  HR
                                        Analyst
               David 40
                               Sales
                                           NULL
                      45
                                           NULL
                Emma
                          Marketing
               Frank 50
                                  HR |
                                           NULL
Union
           Union two DataFrames
DataFrames
           df.union(otherDf)
           Example-
           df.union(df).show()
```

```
scala> df.union(df).show()
               Name | Age | Department |
              Alice 25
                              Sales
                Bob 30 Marketing
            Charlie 35
              David
                     40
                              Sales
               Emma | 45 | Marketing
              Frank
                      50
                                  HR
              Alice 25
                              Sales
                Bob
                      30 | Marketing
            Charlie | 35|
                                  HRI
              David 40
                              Sales
               Emma 45
                          Marketing
              Frank 50
                                  HR
           Pivot DataFrame based on a column
Pivot
           df.groupBy("col1").pivot("col2").agg(sum("col3"))
           Example-
           df.groupBy("Department").pivot("Age").count().show()
            scala> df.groupBy("Department").pivot("Age").count().show()
            Department
                          25
                               30
                                     35
                                          40
                  Sales | 1 | NULL | NULL | 1 | NULL | NULL |
                     HR|NULL|NULL| 1|NULL|NULL|
             Marketing NULL
                                1|NULL|NULL|
                                                 1 | NULL |
Window
           Calculate row number over a window
Functions
           import org.apache.spark.sql.expressions.Window
```

```
val windowSpec = Window.orderBy("col1")
df.withColumn("row_number", row_number().over(windowSpec))
Example-
import org.apache.spark.sql.expressions.Window
import org.apache.spark.sql.functions.
val windowSpec = Window.orderBy("Age")
df.withColumn("row_number", row_number().over(windowSpec)).show()
 cala> import org.apache.spark.sql.expressions.Window
 import org.apache.spark.sql.expressions.Window
 scala> import org.apache.spark.sql.functions._
 mport org.apache.spark.sql.functions._
 cala> val windowSpec = Window.orderBy("Age")
 uindowSpec: org.apache.spark.sql.expressions.WindowSpec = org.apache.spark.sql.expressions.WindowSpec@70ac2eb2
scala> df.withColumn("row_number", row_number().over(windowSpec)).show()
24/04/26 20:18:41 WARN WindowExec: No Partition Defined for Window operation! Moving all data to a single partition, this can cause serious performance degradation
 24/04/26 20:18:41 MARN WindowExec: No Partition Defined for Window operation! Moving all data to a single partition, this can cause serious performance degradation
24/04/26 20:18:41 WARN WindowExec: No Partition Defined for Window operation! Moving all data to a single partition, this can cause serious performance degradation
 14/04/26 20:18:41 WARN WindowExec: No Partition Defined for Window operation! Moving all data to a single partition, this can cause serious performance degradation
 4/04/26 20:18:41 WARN WindowExec: No Partition Defined for Window operation! Moving all data to a single partition, this can cause serious performance degradation
   Name | Age | Department | row_number
    Bob 30 Marketing
  harlie 35
   David 40
               Sales
   Emma | 45 | Marketing
Get distinct values of a column
df.select("col1").distinct()
Example-
```

df.select("Department").distinct().show()

Distinct

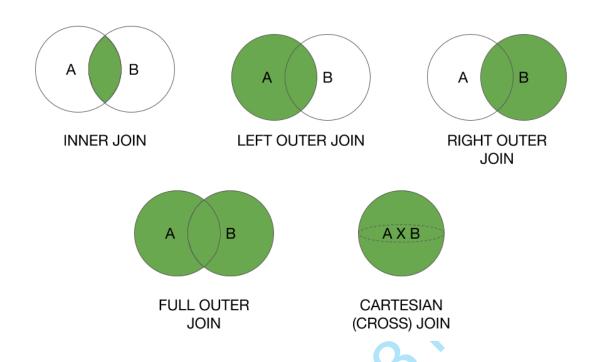
Values

```
scala> df.select("Department").distinct().show()
             Department
                  Sales
              Marketing
           Display first n rows
Show Data
           df.show()
           Display only selected columns
           df.select("col1", "col2").show()
           Example-
           df.show()
            scala> df.show()
                Name | Age | Department |
               Alice 25
                               Sales
                 Bob | 30 | Marketing
             Charlie | 35|
                                   HR
               David 40
                               Sales
                Emma | 45 | Marketing
               Frank 50
```

```
Column
        scala> val sampleData = Seq(
Rename
                             "P", 9),
                ("A", 5, 7,
                ("B", 3, 4,
                             "P", 7),
                             "C"
                ("C", 2, 1,
                             "C", 2),
"C", 4),
                ("D", 5, 6,
                ("E", 6, 3,
                ("F", 1, 7,
                             "H"
                ("G", 4, 1, "H", 7)
        sampleData: Seq[(String, Int, Int, String, I
        scala> val sampleDF = spark.createDataFrame(
        sampleDF: org.apache.spark.sql.DataFrame = [
        scala> sampleDF.show()
        |Name| N1| N2|City| R1|
            А
                 5
                     7
                           PΙ
                               9
            ВΙ
                 3|
                     4
                           PΙ
                               7
                 2
            C
                     1
                           C
                               4
            DΙ
                 5
                     6
                           C
                               2
            E
                 6
                     3
                          C
                               4
            FΙ
                 11
                     7
                           нΙ
                               9
                 4
                               7
            G
                     1
                           ΗI
        scala> sampleDF.groupBy("City").count().alias("Total").show()
        |City|count|
                2
           c
                3
           H
                2
```

```
scala> val totalDF = sampleDF.groupBy("City").count().withColumnRenamed("count", "Total").show()
|City|Total|
       2|
3|
2|
totalDF: Unit = ()
scala>
val totalDF =
sampleDF.groupBy("City").count().withColumnRenamed("count", "Total")
totalDF.show()
sampleDF.withColumnRenamed("R1", "P1").show()
scala> sampleDF.withColumnRenamed("R1", "P1").show()
 Name N1 N2 City P1
          3
                4
          2
                           4
                1
          5
                           2
                6
          6
                3|
                      C
                           4
                7
                           9
                1
                            7
```

Joins



Before Running Queries Import:

import org.apache.spark.sql.{SparkSession, DataFrame}

```
Sample data for DataFrame 1

val data1 = Seq(

(1, "Alice"),

(2, "Bob"),

(3, "Charlie")

Create DataFrame 1

val df1 = spark.createDataFrame(data1).toDF("id", "name")
```

```
scala> df1.show()
                    id
                           name
                         Alice
                            Bob
                     3|Charlie
                 Sample data for DataFrame 2
                 val data2 = Seq(
                  (1, "Sales"),
                  (3, "Marketing"),
                  (4, "HR")
                 Create DataFrame 2
                 val df2 = spark.createDataFrame(data2).toDF("id", "department")
                  scala> df2.show()
                    id department
                             Sales
                        Marketing
                 val innerJoinDF: DataFrame = df1.join(df2, "id", "inner")
Inner Join
                 innerJoinDF.show()
```

```
scala> innerJoinDF.show()
                          name department
                        Alice
                     3 | Charlie | Marketing |
                 val leftJoinDF: DataFrame = df1.join(df2, Seq("id"), "left outer")
Left Outer Join
                 leftJoinDF.show()
                                      Sales
                         Alice
                            Bob
                     3|Charlie|
Right Outer Join val rightJoinDF: DataFrame = df1.join(df2, Seq("id"), "right_outer")
                 rightJoinDF.show()
                  scala> rightJoinDF.show()
                       Charlie
                                 Marketing
Full Outer Join
                 val fullOuterJoinDF: DataFrame = df1.join(df2, Seq("id"), "outer")
                 fullOuterJoinDF.show()
```

```
scala> fullOuterJoinDF.show()
                         name department
                   1
                       Alice
                                   Sales
                   2
                         Bob
                                    NULL
                   3 | Charlie | Marketing |
                        NULL
Inner Join
                val crossJoinDF: DataFrame = df1.crossJoin(df2)
                crossJoinDF.show()
                scala> crossJoinDF.show()
                  id
                        name| id|department
                                1
                       Alice
                                        Sales
                   1
                       Alice
                                3|
                                   Marketing
                                4
                   1
                       Alice
                   2
                          Bob
                                1
                                        Sales
                   2
                                31
                                   Marketing
                         Bob
                         Bob
                                4
                   3 | Charlie |
                                1
                                        Sales
                                   Marketing
                   3|Charlie|
                                3
                                4
                   3 Charlie
```

Running SQL Queries Using Spark SQL

Querying structured data using Spark SQL allows you to leverage SQL queries to interact with DataFrames and perform various operations.

import org.apache.spark.sql.SparkSession

Sample data for DataFrame

```
val data = Seq(

("Alice", 25, "Sales", 50000),

("Bob", 30, "Marketing", 60000),

("Charlie", 35, "HR", 55000),

("David", 40, "Sales", 52000),

("Emma", 45, "Marketing", 62000),

("Frank", 50, "HR", 58000)
)
```

Create DataFrame

```
val df = spark.createDataFrame(data).toDF("name", "age", "department", "salary")
```

Register DataFrame as a temporary table

df.createOrReplaceTempView("employee")

Select all Columns

```
val result1 = spark.sql("SELECT * FROM employee")
result1.show()
```

Select Specific Columns

val result2 = spark.sql("SELECT name, age FROM employee")
result2.show()

```
scala> result2.show()
+----+
| name|age|
+----+
| Alice| 25|
| Bob| 30|
|Charlie| 35|
| David| 40|
| Emma| 45|
| Frank| 50|
```

Filter Rows With WHERE

val result3 = spark.sql("SELECT * FROM employee WHERE age > 30") result3.show()

Count Rows

val result4 = spark.sql("SELECT COUNT(*) AS total_employees FROM employee")
result4.show()

Using LIKE Operator

val result5 = spark.sql("SELECT * FROM employee WHERE department LIKE 'Sa%'")
result5.show()

```
scala> result5.show()
+----+---+
| name|age|department|salary|
+----+---+
|Alice| 25| Sales| 50000|
|David| 40| Sales| 52000|
+----+
```

Group By with Aggregate Functions

val result6 = spark.sql("SELECT department, AVG(salary) AS avg_salary FROM employee GROUP BY department")

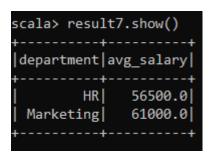
result6.show()

```
scala> result6.show()
+------+
|department|avg_salary|
+-----+
| Sales| 51000.0|
| HR| 56500.0|
| Marketing| 61000.0|
```

Having Clause

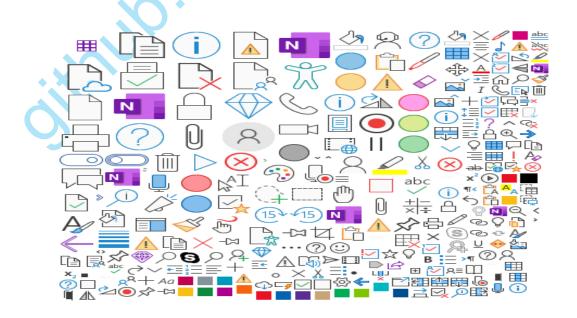
val result7 = spark.sql("SELECT department, AVG(salary) AS avg_salary FROM employee GROUP BY department HAVING AVG(salary) > 55000")

result7.show()



Managing Spark Partitions with Coalesce and Repartition

Repartition	Coalesce
Can Both Increase and Decrease number of partitions	 Used To decrease number of partitions
 Does not Worry about amount of shuffle 	Tries to avoid shuffle
Slow than Coalesce	Better Performance than Repartition
Tries to create partitions of similar size	Output Partition can be uneven in size



```
Data
```

```
val x = (1 \text{ to } 10).\text{toList}
val numberDf = x.toDF("number")
numberDf.show()
numberDf.rdd.partitions.size
numberDf.write.csv("C:/Spark/sparkfiles/partinfo")
Coalesce (Merging)
val numberDf1 = numberDf.coalesce(3)
numberDf1.rdd.partitions.size
numberDf1.write.csv("C:/Spark/sparkfiles/partinfo1")
Repartition (Dividing)
val numberDf2 = numberDf.repartition(2)
numberDf2.rdd.partitions.size
numberDf2.write.csv("C:/Spark/sparkfiles/partinfo2")
val numberDf3 = numberDf.repartition(6)
numberDf3.rdd.partitions.size
numberDf3.write.csv("C:/Spark/sparkfiles/partinfo3")
```

Feature Transformation in Spark

```
Metho from pyspark.sql import SparkSession
d-1
          spark = SparkSession.builder.appName("Name anything")getOrCreate()
           data =
           spark.read.csv('C:/Spark/sparkfiles/transformation data.csv',header=True,inferSche
          ma=True)
          data.show()
           from pyspark.ml.feature import StringIndexer
          indexer = StringIndexer(inputCol = 'color', outputCol = "color indexed")
           indexer model = indexer.fit(data)
           indexed data = indexer model.transform(data)
          indexed data.show()
             id color
              from pyspark.ml.feature import StringIndexer
indexer = StringIndexer(inputCol = 'color', outputCol = "color_indexed")
indexer_model = indexer_fit(data)
indexed_data = indexer_model.transform(data)
indexed_data.show()
             id| color|color indexed|
                 blue
```

Suppose you have a dataset with a categorical feature like "Color" with values "Red", "Blue", and "Green". Before applying a machine learning algorithm to predict some outcome based on this data, you need to convert these string values into numerical indices. This is where StringIndexer comes into play, transforming "Red" to 0, "Blue" to 1, and "Green" to 2, for example, allowing the algorithm to work with the data effectively.

Metho d-2 Feature Scaling

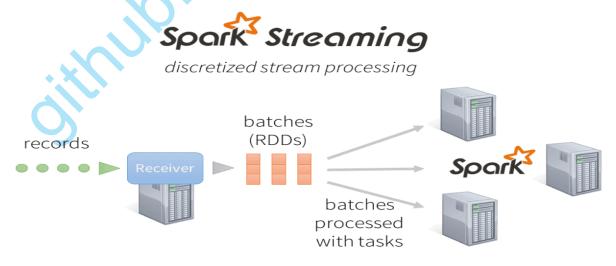
Spark Streaming



Apache Spark Streaming is a component of the Apache Spark ecosystem that enables scalable, high-throughput, fault-tolerant stream processing of real-time data.

Imagine we have a continuous stream of data coming in, like tweets from Twitter or sensor data from IoT devices. Spark Streaming allows us to process this data in real-time, meaning we can analyze and react to it as soon as it arrives.

How it Works:



records processed in batches with short tasks each batch is a RDD (partitioned dataset)

Micro-Batching: Spark Streaming divides the incoming stream of data into small batches or micro-batches. Each batch represents a short interval of time, such as a few seconds.

Processing: Once the data is divided into batches, Spark processes each batch using the same powerful distributed computing engine used for batch processing in Apache Spark.

Transformation and Analysis: Within each batch, we can apply transformations, filters, aggregations, and machine learning algorithms to analyze the data and derive insights from it.

Output: After processing, we can store the results in various data stores, generate alerts, update dashboards, or trigger actions based on the analyzed data.

Benefits

Real-Time Insights: Spark Streaming enables us to analyze and respond to data in real-time, providing timely insights and actionable information.

Scalability: It leverages the scalability and fault-tolerance of Apache Spark, allowing us to process large volumes of data efficiently.

Integration: It integrates seamlessly with other components of the Spark ecosystem, such as Spark SQL, MLlib, and GraphX, enabling us to perform complex analytics and machine learning on streaming data.

Streaming

).awaitTermination()

```
import org.apache.spark.sql.{SparkSession}
import org.apache.spark.sql.functions._
val spark = SparkSession.builder().master("local").appName("mysource").getOrCreate()
val initDF = (spark.readStream.format("rate").option("rowsPerSecond",1).load())
println("Streaming DataFrame: " +initDF.isStreaming)

val resultDF = initDF.withColumn("result",col("value")+lit(1))
```

resultDF.writeStream.outputMode("append").option("truncate",false).format("console").start(

Structures Streaming - Socket Source

```
//Powershell-1
// starting neat server
ncat -lk 9999
//Powershell-2
// Import Libraries
import org.apache.spark.sql.SparkSession
import org.apache.spark.sql.functions.
// Create Spark Session
val spark = SparkSession
 .builder()
 .master("local")
 .appName("Socket Source")
 .getOrCreate()
// Define host and port number to Listen.
val host = "127.0.0.1"
val port = "9999"
// Create Streaming DataFrame by reading data from socket.
val initDF = (spark)
 .readStream
 .format("socket")
 .option("host", host)
 .option("port", port)
 .load())
```

```
// Perform word count on streaming DataFrame
val wordCount = (initDF
 .select(explode(split(col("value"), " ")).alias("words"))
                                              .groupBy("words")
 .count()
 )
wordCount
 .writeStream
 .outputMode("complete")
 .format("console")
 .start()
 .awaitTermination()
Batch: 6
     words | count |
                                      Administrator: Windows PowerShell
   Kerala
              1 | 4 | 2 | 2 | 4 | 1 | 1 | 2 | 5 | 2 | 2 | 1 | 2 | 2 | 1 | 2 |
                                      Copyright (C) Microsoft Corporation. All rights reserved.
      name
                                      Install the latest PowerShell for new features and improvements! https://aka.ms/PSWindow
       how
   Mumbai
      your
    Chatra
     Tamil|
       my
     Nadu
      Agra
     what
     Delhi
       are
 Rajasthan
   Aakash
```



Apache Kafka is an open-source distributed event streaming platform used for building real-time data pipelines and streaming applications. It is designed to handle high-throughput, fault-tolerant, and scalable ingestion, storage, and processing of real-time data streams.

Imagine we're at a busy post office where letters are constantly arriving and need to be sorted, processed, and delivered to the right destinations. The post office has several departments working together efficiently to handle this continuous flow of mail. Apache Kafka works similarly but with data instead of letters.

- **Post Office:** Think of Kafka as a modern post office. Just like the post office receives, processes, and delivers mail, Kafka receives, processes, and delivers data.
- **Data Pipelines:** Data pipelines are like conveyor belts in the post office, moving mail from one department to another. Kafka acts as a reliable conveyor belt for your data, ensuring it gets from one system to another without loss or delay.

Why Kafka is Used:

Real-Time Data Processing: Kafka enables organizations to process and react to data in real-time, allowing them to make timely decisions and provide immediate responses.

Scalability: Kafka is designed to handle large volumes of data and can scale horizontally by adding more servers to the cluster, making it suitable for high-throughput applications.

Fault Tolerance: Kafka is fault-tolerant, meaning it can continue to function even if some of its servers fail. This ensures that data is not lost and processing can continue uninterrupted.

Data Integration: Kafka serves as a central hub for integrating different systems and applications, allowing data to flow seamlessly between them.

Event Streaming: Kafka facilitates event-driven architectures where applications can react to events in real-time, enabling use cases like real-time analytics, monitoring, and fraud detection.

Producer	Consumer	
A Producer is a component or application responsible for sending data (messages) to Kafka topics.	A Consumer is a component or application responsible for reading data (messages) from Kafka topics.	
It's like a publisher or sender of messages.	It's like a subscriber or receiver of messages.	
Producers are typically used to push data from various sources into Kafka topics.	 Consumers retrieve data from Kafka topics and process it according to their requirements. 	
• For example, imagine a sensor sending temperature readings to a Kafka topic every second. In this scenario, the sensor acts as a Producer, continuously sending temperature data to Kafka.	• For example, imagine an analytics application that analyzes temperature trends. This application would consume temperature data from a Kafka topic, process it to detect trends, and possibly visualize the results.	

Commands	Start Zookeeper
	1. cd C:\kafka_2.12-3.7.0
	2\bin\windows\zookeeper-server-start.bat .\config\zookeeper.properties
	Start Server
0)	1. cd C:\kafka_2.12-3.7.0
	2\bin\windows\kafka-server-start.bat .\config\server.properties
	Go to new powershell window
	1. jps

Create Topic	.\bin\windows\kafka-topics.batbootstrap-server 127.0.0.1:9092create replication-factor 1partitions 1topic bigdata
Topic List	.\bin\windows\kafka-topics.batlistbootstrap-server localhost:9092
Producer	.\bin\windows\kafka-console-producer.batbroker-list localhost:9092topic bigdata
Consumer	.\bin\windows\kafka-console-consumer.batbootstrap-server localhost:9092 topic bigdatafrom-beginning
Describe	.\bin\windows\kafka-topics.batdescribebootstrap-server localhost:9092 topic bigdata
Delete	.\bin\windows\kafka-topics.batbootstrap-server localhost:9092delete topic bigdata

```
PS C:\WINDOwS\system32> jps
4384 3ps
3880 QuorumPeerMain
1886 4f 1876 2 12-3 7.8

S C:\Wafka_2.12-3.7.8

S C:\Wafka_2.12-3.7.8

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S C:\Wafka_2.12-3.7.8

S C:\Wafka_2.12-3.7.8

No Nin\windows\wafka-console-producer.bat --broker-list localhost:9892

No C:\Wafka_2.12-3.7.8

No Nin\windows\wafka-console-producer.bat --broker-list localhost:9892 --topic bigdata

Nin\windows\wafka-console-producer.bat --broker-list localhost:9892 --topic bigdata

Nin\windows\windows\wafka-console-producer.bat --broker-list localhost:9892 --topic bigdata

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```

```
S C:\WINDOWS\system32> cd C:\kafka_2.12-3.7.0
S C:\kafka_2.12-3.7.0> .\bin\windows\kafka-console-consumer.bat --bootstrap-server localhost:9092 --topic bigdata --from-beginning
by Name Name Is Aakash Kumar
hat is Your Name
Belongs to Jharkhand
     rom Where Do You Belong
    low Are You
lero
     ero
   123
Install the latest PowerShell for new features and improvements! https://aka.ms/PSWindows
 C:\WINDOWS\system32> cd C:\kafka 2.12-3.7.0
c:\kafka 2.12-3.7.0> \bin\windows\zookeeper.server-start.bat .\config\zookeeper.properties
go24-94-27 go2:12:20,211] INFO Reading configuration from: .\config\zookeeper.properties (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
go24-94-27 go2:12:20,211] INFO clientPortAddress is 0.0.0 go2:2181 (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
go24-04-27 go2:12:20,211] INFO secureClientPort is not set (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
go24-04-27 go2:12:20,211] INFO observerMasterPort is not set (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
go24-04-27 go2:12:20,212] INFO metricsProvider.className is org.apache.zookeeper.metrics.impl.DefaultMetricsProvider (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
       1024-04-27 02:12:20,227] INFO autopurge.snapRetainCount set to 3 (org.apache.zookeeper.server.DatadirCleanupManager)
1024-04-27 02:12:20,227] INFO autopurge.snapRetainCount set to 3 (org.apache.zookeeper.server.DatadirCleanupManager)
1024-04-27 02:12:20,227] INFO autopurge.purgeInterval set to 0 (org.apache.zookeeper.server.DatadirCleanupManager)
1024-04-27 02:12:20,227] INFO purge task is not scheduled. (org.apache.zookeeper.server.DatadirCleanupManager)
1024-04-27 02:12:20,227] INFO purge task is not scheduled. (org.apache.zookeeper.server.DatadirCleanupManager)
1024-04-27 02:12:12:0,227] INFO log4j 1.2 jmx support not found; jmx disabled. (org.apache.zookeeper.jmx.ManagedUtil)
1024-04-27 02:12:0,227] INFO log4j 1.2 jmx support not found; jmx disabled. (org.apache.zookeeper.jmx.ManagedUtil)
1024-04-27 02:12:12:0,227] INFO clientFortAddress is 0.8 0.8 0:2181 (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
1024-04-27 02:12:0,227] INFO observerMasterPort is not set (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
1024-04-27 02:12:20,227] INFO observerMasterPort is not set (org.apache.zookeeper.server.quorum.QuorumPeerConfig)
        024-04-27 02:12:20,227] INFO Starting server (org.apache.zookeeper.server.ZooKeeperServerMain)
024-04-27 02:12:20,258] INFO ServerMetrics initialized with provider org.apache.zookeeper.metrics.impl.DefaultMetricsProvider@617c74e5 (org.apache.zookeeper.server.Se

      8924-82-7 02:12:20,258]
      INFO ServerMetrics Initialized with provider org.apache.Zookeeper.metrics.impliberauthetrics.rovidergol.com.com.go.go.ne.eremetrics.

      1924-84-27 02:12:20,258]
      INFO ACL digest algorithm is: SHA1 (org.apache.zookeeper.server.auth.DigestAuthenticationProvider)

      1924-84-27 02:12:20,258]
      INFO zookeeper.DigestAuthenticationProvider.enabled = true (org.apache.zookeeper.server.auth.DigestAuthenticationProvider)

      1924-80-27 02:12:20,258]
      INFO zookeeper.DigestAuthenticationProvider.enabled = true (org.apache.zookeeper.server.auth.DigestAuthenticationProvider)

      1924-80-27 02:12:20,3051
      INFO (org.apache.zookeeper.server.perserver)

      1924-90-27 02:12:20,3051
      INFO (org.apache.zookeeper.server.Zookeeperserver)

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                                                                                                          stall the latest PowerShell for new features and improvements! https://aka.ms/PSWindows
         C:\WINDOWS\system32> cd C:\kafka_2.12-3.7.0
C:\kafka_2.12-3.7.0> .\bin\windows\kafka-server-start.bat .\config\server.properties
024-04-27 02:12:37,458] INFO Registered kafka:type=kafka.LogdjController MBean (kafka.utils.Log4jControllerRegistration$)
024-04-27 02:12:38,345] INFO Setting -D jdk.tls.rejectClientInitiatedRenegotiation=true to disable client-initiated TLS renegotiation (org.apache.zookeep
    UTII)
2024-04-27 02:12:38,503] INFO starting (kafka.server.KafkaServer)
2024-04-27 02:12:38,504] INFO Connecting to zookeeper on localhost:2181 (kafka.server.KafkaServer)
2024-04-27 02:12:38,535] INFO [ZooKeeperClient Kafka server] Initializing a new session to localhost:2181. (kafka.zooKeeper.ZooKeeperClient)
2024-04-27 02:12:38,558] INFO Client environment:zooKeeper.version=3.8.3-6ad6d364c7c0bcf0de45zd54ebefa3058098ab56, built on 2023-10-05 10:34 UTC (org.apach
 2024-04-7 02:12:38,555] INFO Client environment:zookeeper.version=3.8.3-6ad6d364C70bcf0de452d54ebefa3058098ab56, built on 2023-10-05 10:34 UTC (org.apach cookeeper) 2024-04-7 02:12:38,559] INFO Client environment:java.version=1.8.0_401 (org.apach zookeeper.Zookeeper) 2024-04-7 02:12:38,559] INFO Client environment:java.version=1.8.0_401 (org.apach zookeeper.Zookeeper) 2024-04-7 02:12:38,550] INFO Client environment:java.version=1.8.0_401 (org.apach zookeeper.Zookeeper.Zookeeper) 2024-04-7 02:12:38,560] INFO Client environment:java.version=1.8.0_401 (org.apach zookeeper.Zookeeper.Zookeeper) 2024-04-7 02:12:38,560] INFO Client environment:java.class.path-c:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.1.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\kafka_2.12-3.7.0\libs\activation=1.1.0.jar;C:\ka
```

Create a Java project with name MyProject to configure the consumer in Apache Kafka. Link that consumer with producer via command line and show the communication. (Using Eclipse IDE)

Step-Create a New Java Project:

01

- Open Eclipse IDE.
- Go to File > New > Java Project.
- Enter "MyProject" as the project name.
- Click "Finish" to create the project.

Step-Add Kafka Dependencies:

02

- Right-click on the project in the Project Explorer.
- Navigate to Build Path > Configure Build Path.
- In the Libraries tab, click on "Add External JARs...".
- Add the Kafka client JARs (e.g., kafka-clients-x.x.x.jar, slf4j-api-x.x.x.jar, etc.) from the Kafka installation directory.
- Click "Apply and Close".

Step-Create a Kafka Consumer Class:

03

- Right-click on the src folder in the project.
- Go to New > Class.
- Name the class "KafkaConsumerExample" and click "Finish".
- Add code to configure and run the Kafka consumer:

import org.apache.kafka.clients.consumer.ConsumerConfig;

import org.apache.kafka.clients.consumer.ConsumerRecord;

import org.apache.kafka.clients.consumer.ConsumerRecords;

import org.apache.kafka.clients.consumer.KafkaConsumer;

```
import java.time.Duration;
import java.util.Collections;
import java.util.Properties;
public class KafkaConsumerExample {
  public static void main(String[] args) {
    Properties props = new Properties();
    props.put(ConsumerConfig.BOOTSTRAP SERVERS CONFIG,
"localhost:9092");
    props.put(ConsumerConfig.GROUP ID CONFIG, "test-group");
    props.put(ConsumerConfig.KEY DESERIALIZER CLASS CONFIG,
"org.apache.kafka.common.serialization.StringDeserializer");
    props.put(ConsumerConfig.VALUE DESERIALIZER CLASS CONFIG,
"org.apache.kafka.common.serialization.StringDeserializer");
    KafkaConsumer<String, String> consumer = new KafkaConsumer<>(props);
    consumer.subscribe(Collections.singletonList("my-topic"));
    while (true) {
      ConsumerRecords<String, String> records =
consumer.poll(Duration.ofMillis(100));
      for (ConsumerRecord<String, String> record : records) {
         System.out.printf("Received message: key = %s, value = %s%n",
record.key(), record.value());
```

Step-Run Producer and Consumer via Command Line: 04

- Start Zookeeper and Kafka server if not already running.
- Create a topic named "my-topic" using the following command:

\bin\windows\kafka-topics.bat --bootstrap-server 127.0.0.1:9092 --create --replication-factor 1 --partitions 1 --topic my-topic

• Run a Kafka producer in one Command Prompt window:

\bin\windows\kafka-console-producer.bat --broker-list localhost:9092 --topic my-topic

- Enter some messages in the producer terminal.
- Run the Kafka consumer class in eclipse IDE

```
Step—

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```

Create a Java project with name "MyProject" to configure the producer in Apache Kafka. Link that Producer with Consumer via command line and show the communication. (Using Eclipse)

Step- 01	Create a New Java Project in Eclipse:
	Open Eclipse IDE.
	• Go to File > New > Java Project.
	Enter "MyProject" as the project name.
	Click "Finish" to create the project.
Step-	Add Kafka Dependencies:
	Right-click on the project in the Project Explorer.
	Navigate to Build Path > Configure Build Path.
	• In the Libraries tab, click on "Add External JARs".
	• Add the Kafka client JARs (e.g., kafka-clients-x.x.x.jar, slf4j-api-x.x.x.jar, etc.) from the Kafka installation directory.
	Click "Apply and Close".
	Create a Kafka Producer Class:
03	% .
	Right-click on the src folder in the project.
	Go to New > Class.
	Name the class "KafkaProducerExample" and click "Finish".
	Add the following code to configure and run the Kafka producer:
	Code1:
	import org.apache.kafka.clients.producer.*;

```
import java.util.Properties;
public class KafkaProducerExample {
  public static void main(String[] args) {
    Properties props = new Properties();
    props.put("bootstrap.servers", "localhost:9092");
    props.put("key.serializer",
'org.apache.kafka.common.serialization.StringSerializer");
    props.put("value.serializer",
"org.apache.kafka.common.serialization.StringSerializer");
    Producer < String > producer = new KafkaProducer <> (props);
    String topic = "my-topic";
    for (int i = 0; i < 10; i++) {
       String key = "key" + i;
       String value = "value" \pm i;
       ProducerRecord < String > record = new ProducerRecord <> (topic, key,
value);
       producer.send(record, new Callback() {
         @Override
         public void onCompletion(RecordMetadata metadata, Exception exception) {
            if (exception != null) {
              exception.printStackTrace();
            } else {
              System.out.printf("Sent message: topic = \%s, partition = \%d, offset =
\%d, key = \%s, value = \%s\%n",
                   metadata.topic(), metadata.partition(), metadata.offset(), key,
value);
            }
```

```
}
       });
    producer.flush();
    producer.close();
Code-2
import org.apache.kafka.clients.producer.*;
import java.util.Properties;
import java.util.Scanner;
public class ConsoleInputProducer {
  public static void main(String[] args) {
     Properties props = new Properties();
    props.put("bootstrap.servers", "localhost:9092");
    props.put("key.serializer",
'org.apache.kafka.common.serialization.StringSerializer");
    props.put("value.serializer",
"org.apache.kafka.common.serialization.StringSerializer");
     Producer<String, String> producer = new KafkaProducer<>(props);
     String topic = "my-topic";
```

```
// Create a scanner to read input from console
    Scanner scanner = new Scanner(System.in);
    // Continuously read input from console and send it to Kafka
    while (true) {
       System.out.print("Enter message (or type 'exit' to quit): ");
       String message = scanner.nextLine();
       // If user types 'exit', break the loop and close the producer
       if (message.equalsIgnoreCase("exit")) {
         break;
       }
       // Create a producer record with the input message
       ProducerRecord < String > record = new ProducerRecord <> (topic,
message);
       // Send the record to Kafka
       producer.send(record, new Callback() {
         @Override
         public void onCompletion(RecordMetadata metadata, Exception exception) {
            if (exception != null) {
              exception.printStackTrace();
            } else {
              System.out.printf("Sent message: topic = %s, partition = %d, offset =
%d, value = %s%n'',
                   metadata.topic(), metadata.partition(), metadata.offset(), message);
            }
         }
```

```
});
}

// Close the producer
producer.close();
scanner.close();
}
```

Step-Run Producer and Consumer via Command Line: 04

- Start Zookeeper and Kafka server if not already running. Navigate to the Kafka installation directory in the Command Prompt.
- Create a topic named "my-topic" using the following command:

.\bin\windows\kafka-topics.bat --bootstrap-server localhost:9092 --create --replication-factor 1 --partitions 1 --topic my-topic

• Run the Kafka consumer in one Command Prompt window:

.\bin\windows\kafka-console-consumer.bat --bootstrap-server localhost:9092 --topic my-topic --from-beginning

• Run the Kafka producer code in Eclipse IDE

