coalesce() Transformation

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In [1]: from pyspark.sql import SparkSession
        import pyspark
        spark = SparkSession \
                .builder \
                .master("local[*]") \
                .appName("coalesce Transformation") \
                .enableHiveSupport() \
                .getOrCreate()
        22/10/14 12:05:42 WARN Utils: Your hostname, Vaishalis-MacBook-Pro.local resolves to a loopback address: 127.0.
        0.1; using 192.168.0.105 instead (on interface en0)
        22/10/14 12:05:42 WARN Utils: Set SPARK_LOCAL_IP if you need to bind to another address
        Setting default log level to "WARN".
        To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
        22/10/14 12:05:43 WARN NativeCodeLoader: Unable to load native—hadoop library for your platform... using builti
        n-java classes where applicable
        22/10/14 12:05:44 WARN Utils: Service 'SparkUI' could not bind on port 4040. Attempting port 4041.
        22/10/14 12:05:44 WARN Utils: Service 'SparkUI' could not bind on port 4041. Attempting port 4042.
In [2]: def debug_a_partition(iterator):
            print("==begin-partition=")
            for x in iterator:
                print(x)
            #end-for
            print("==end-partition=")
        names_list =["ABC - 1", "DEF - 1", "GHI-1", "ABC - 2", "DEF - 2", "GHI -2", \
                     "ABC - 3", "DEF - 3", "GHI -3", "ABC - 4", "DEF - 4", "GHI -4"]
        names_rdd = spark.sparkContext.parallelize(names_list,4)
        print("From local[4] =",names_rdd.getNumPartitions())
        print("Repartition elements : ", names_rdd.foreachPartition(debug_a_partition))
        From local[4] = 4
        [Stage 0:>
                                                                             (0 + 4) / 4]
        Repartition elements: None
        ==begin-partition===begin-partition=
        ==begin-partition=
        ==begin-partition=ABC - 2
        DEF - 2ABC - 3
        ABC - 4
        DEF - 3
        DEF - 4GHI -3
        GHI -4GHI -2
        ==end-partition=
        ==end-partition===end-partition=
        ABC - 1
        DEF - 1
        GHI-1
        ==end-partition=
In [3]: #reparition shuffles the data completely
        repartition_rdd = names_rdd.repartition(3)
        print("coalesce elements : ", repartition_rdd.foreachPartition(debug_a_partition))
        [Stage 1:>
        coalesce elements : None
        ==begin-partition=
        ==end-partition=
        ==begin-partition=
        ABC - 2
        DEF - 2
        GHI -2
        ABC - 4
        DEF - 4
        GHI -4
        ==end-partition=
        ==begin-partition=
        ABC - 1
        DEF - 1
        GHI-1
        ABC - 3
        DEF - 3
        GHI -3
        ==end-partition=
        From the output above, we can see the reshuffling of the entire data from 4 partitions in names_rdd to 3 partitions in
        repartition_rdd. also, partition 1 is empty.
In [4]: #coalesce combines the partition close to it
        coalesce_rdd = names_rdd.coalesce(3)
        print("Repartition size : ", coalesce_rdd.foreachPartition(debug_a_partition))
        ==begin-partition=
        ==begin-partition=ABC - 1
        DEF - 1
        GHI-1
        ==begin-partition===end-partition=ABC - 2
        DEF - 2
        GHI -2
        ==end-partition=
        ABC - 3
        DEF - 3
        GHI -3
        ABC - 4
        DEF - 4
        GHI -4
        ==end-partition=
        Repartition size : None
        From the output above, we can see partitions 3 and 4 of names_rdd combined to form partition 3 in coalesce_rdd rather than
        shuffle around the entire data.
In [5]: #Reading data from a file on the local machine
        data_file_path = "/Users/vaishaliyasala/Desktop/Github/Spark/Exercise_Dependencies/SalesJan2009.csv"
        df = spark.read.csv(data_file_path, header = True )
        df1 = df.select(df["Name"],df["Country"]).repartition(4)
        print("Count in the original data=", df1.count())
        #Filter the names only from country United States
        filtered_rdd = df1.rdd.filter(lambda x: (x[1] == "United States"))
        filtered_rdd.toDF(["Name", "Country"]).show(5)
        print("Filtered data Count =", filtered_rdd.count())
        print("Number of Partitions =", filtered_rdd.getNumPartitions())
        Count in the original data= 998
        +----+
              Name|
                        Country|
            _____+
          Abikay|United States|
        |Christian|United States|
        | Alicja | United States|
        | Debora |United States|
        | Sandrine|United States|
        +----+
        only showing top 5 rows
        Filtered data Count = 463
        Number of Partitions = 4
In [6]: | #As the filtered data count increased significantly, we can reduce the number of partitions from 4 to 2
        #It doesn't change the resultant RDD as seen from the outputs of both before and after coalesce transformation
        #Only data changes in each partition
        names_coalesce_rdd = filtered_rdd.coalesce(2)
        names_coalesce_rdd.toDF(["Name", "Country"]).show(5)
        print("Number of Partitions =", names_coalesce_rdd.getNumPartitions())
              Name|
                        Country|
           -----+
          Abikay|United States|
        |Christian|United States|
        | Alicja |United States|
        | Debora |United States|
        | Sandrine|United States|
```

only showing top 5 rows

Number of Partitions = 2