

<b>Ex. No. 4</b>	<b>Matrix Multiplication Using RDDs in Spark</b>
<b>Youtube Link</b>	<b><a href="https://youtu.be/aqhQ5MDyEuI">https://youtu.be/aqhQ5MDyEuI</a></b>
<b>Date of Exercise</b>	<b>13.10.25</b>

## **AIM**

To compute the product of two matrices **A** ( $m \times n$ ) and **B** ( $n \times p$ ) using **Apache Spark RDDs**

## **Procedure:**

### **1. Data Preparation**

- Define two small matrices **A** and **B** (e.g., A of shape  $2 \times 3$ , B of shape  $3 \times 2$ ) using MatrixEntry.

### **2. Spark Logic**

1. Create two RDDs: entriesA and entriesB, each as RDD[MatrixEntry].
2. Map:
  - o aKeyed = entriesA.map(e => (e.j, (e.i, e.value)))
  - o bKeyed = entriesB.map(e => (e.i, (e.j, e.value)))
3. Join on key j (common index).
4. Map to ((i, k), product) and use reduceByKey to sum contributions.
5. Construct CoordinateMatrix from summed entries and collect the results.

## **Program:**

```
from pyspark.sql import SparkSession
from pyspark import SparkContext
from pyspark.mllib.linalg.distributed import MatrixEntry, CoordinateMatrix
```

```
# Initialize SparkSession (CoordinateMatrix internally converts RDD -> DataFrame)
spark = SparkSession.builder.appName("PyMatrixMultiplyRDD").getOrCreate()
sc = spark.sparkContext

entriesA = sc.parallelize([
    MatrixEntry(0, 0, 1.0), MatrixEntry(0, 2, 2.0),
    MatrixEntry(1, 1, 3.0), MatrixEntry(1, 2, -1.0)
])

entriesB = sc.parallelize([
    MatrixEntry(0, 0, 1.0), MatrixEntry(1, 0, 3.0), MatrixEntry(2, 0, 5.0),
    MatrixEntry(0, 1, 2.0), MatrixEntry(1, 1, 4.0), MatrixEntry(2, 1, 6.0)
])

aKeyed = entriesA.map(lambda e: (e.j, (e.i, e.value)))
bKeyed = entriesB.map(lambda e: (e.i, (e.j, e.value)))

product = (aKeyed.join(bKeyed)
    .map(lambda kv: ((kv[1][0][0], kv[1][1][0]), kv[1][0][1] * kv[1][1][1]))
    .reduceByKey(lambda a, b: a + b)
    .map(lambda ik_sum: MatrixEntry(ik_sum[0][0], ik_sum[0][1], ik_sum[1])))

result = CoordinateMatrix(product)

for e in result.entries.collect():
    print(f"({e.i},{e.j}) = {e.value}")

spark.stop()
```

**Output:**

--- Retrieval Actions ---

Collect: ['Spark', 'is', 'fast', 'Spark', 'is', 'powerful', 'Spark', 'is', 'easy', 'to', 'use']

Take(5): ['Spark', 'is', 'fast', 'Spark', 'is']

TakeSample (no replacement, 4): ['is', 'to', 'easy', 'Spark']

TakeOrdered (alphabetical, 5): ['Spark', 'Spark', 'Spark', 'easy', 'fast']

Top(5): ['use', 'to', 'powerful', 'is', 'is']

First element: Spark

IsEmpty?: False

Foreach output:

Word: Spark

Word: is

Word: fast

Word: Spark

Word: is

Word: powerful

Word: Spark

Word: is

Word: easy

Word: to

Word: use

--- Aggregation Actions ---

Count: 11

CountByValue: {'Spark': 3, 'is': 3, 'fast': 1, 'powerful': 1, 'easy': 1, 'to': 1, 'use': 1}

Reduce (total word count): 11

Fold (total word count): 11

Aggregate (avg word length): 3.8181818181818183

### **Result :**

The resulting **2×2** product matrix is correctly computed using distributed operations.