# Spark High Performance Pattern(Sample Slide) By Mohit Kumar

## Spark Key-Values:Iterator-To-Iterator:Step-1

```
* Step 1. Map the rows to pairs of (value, column Index).
                             · All in the same key space
* For example:
* dataFrame:
* (0.0, 5.5, <u>7.7</u>, 5.0)
* (1.0, 5.5, 6.7, 6.0)
 * (2.0, 5.5, 1.5, 7.0)
* (3.0, 5.5, 0.5, 7.0)
 * (4.0, 5.5, 0.5, 8.0)
* The output RDD will be: Partition 1:
* (key:0.0 value:0),(key:5.5 value:1),(key:7.7 value:2),(key:5.0 value:3),(key:2.0 value:0),
* (key:5.5 value:1),(key:1.5 value:2),(key:7.0 value:3),(key:4.0 value:0),(key:5.5 value:1)
* Partition 2:
* (key:0.5 value:2),(key:8.0 value:3),(key:1.0 value:0),(key:5.5 value:1),(key:6.7 value:2),
* (key:6.0 value:3),(key:3.0 value:0),(key:5.5 value:1),(key:0.5 value:2),(key:7.0 value:3)
* @param dataframe dateframe of doubles
 * @return RDD of pairs (value, column Index)
private JavaPairRDD<Double, Integer> getValueColumnPairs(Dataset<Row> dataframe) {
   JavaPairRDD<Double, Integer> value ColIndex
           = dataframe.javaRDD().flatMapToPair((PairFlatMapFunction<Row, Double, Integer>) row -> {
               List<Double> rowList = (List<Double>) (Object) toList(row.toSeq());
               List<Tuple2<Double, Integer>> list = zipWithIndex(rowList);
                * Executed on:Partition(Locally)
                return list.iterator():
            });
```

## Spark Key-Values: Iterator-To-Iterator: Step-1

```
public Map<Integer, Iterable<Double>> findRankStatistics(Dataset<Row> dataframe, List<Long> targetRanks) {
    int numOfColumns = dataframe.schema().length();
   JavaPairRDD<Double, Integer> sortedValueColumnPairs=getValueColumnPairs(dataframe)
      .sortByKey()
        .persist(StorageLevel.MEMORY AND DISK());
        * Step-1
         * Executed on:Cluster/distributed
          sortedValueColumnPairs: Partition 1:
        * (key:0.0 value:0),(key:0.5 value:2),(key:0.5 value:2),(key:1.0 value:0),(key:1.5 value:2),
       * (key:2.0 value:0),(key:3.0 value:0),(key:4.0 value:0),(key:5.0 value:1),(key:5.5 value:3),
* (key:5.5 value:1),(key:5.5 value:1),(key:5.5 value:1)
        * (key:6.0 value:3),(key:6.7 value:2),(key:7.0 value:3),(key:7.0 value:3),(key:7.7 value:2),
        * (key:8.0 value:3)
        · Output after a total distributed sort.
      . The partitions are unbalanced. This is expected because all values in the same
      . When dealing with sensor data, facelty sensors often export deemong values.
```

## Spark Key-Values: Iterator-To-Iterator: Step-2

```
Step 2. Find the number of elements for each column in each partition.
                                                                                    · All five values of
 * For Example: Input
 * sortedValueColumnPairs: Partition 1:
 * (key:0.0 value:0),(key:0.5 value:2),(key:0.5 value:2),(key:1.0 value:0),(key:1.5 value:2), the column are there
 * (key:2.0 value:0),(key:3.0 value:0),(key:4.0 value:0),(key:5.0 value:3),(key:5.5 value:1),
 * (key:5.5 value:1),(key:5.5 value:1),(key:5.5 value:1),(key:5.5 value:1)
                                                                                   in partition-1.
 * Partition 2:
 * (key:6.0 value:3),(key:6.7 value:2),(key:7.0 value:3),(key:7.0 value:3),(key:7.7 value:2),
                                                                                     For column - 2 3 values
  (kev:8.0 value:3)
                                                                                  carne prom parelition-1
 * numOfColumns: 4
 * The output will be: \{(0, (5, (5, (3), (1))), (1, (0, 0, (2), (4)))\}
                                                                                  And 2 bean paretitions
 * Oparam sortedValueColumnPairs - sorted RDD of (value, column Index) pairs
 * @param numOfColumns the number of columns
                                                                                 · Only column freq in
 * @return Array that contains (partition index, number of elements from
 * every column on this partition)
                                                                                                      memory
private List<Tuple2<Integer, List<Long>>> getColumnsFreqPerPartition(JavaPairRDD<Double, Integer> sortedValueColumn
   List<Tuple2<Integer, List<Long>>> columsFreqPerPartition
          = sortedValueColumnPairs.mapPartitionsWithIndex((partitionIndex, valueColumnPairs) -> {
                                                                    · Length of columns so not
              Long[] freg = new Long[numOfColumns];
              AtomicInteger ai=new AtomicInteger();
              Arrays.fill(freq, 0l);
                                                                a problem in the memory.
              while (valueColumnPairs.hasNext())
                 int colIndex = valueColumnPairs.next(). 2;
                                                                  · Herator-to-Iterator transform,
                 freq[colIndex] = freq[colIndex] + 1;
                 ai.incrementAndGet();
                                                               one value column pair at a
                                                                times.
```

## Spark Key-Values:Iterator-To-Iterator:Step-2

```
* Not Collecting in a collection as that would require huge memory
             * The above "StreamSupport.stream" create an stream from iterator
             * and we return back another iterator after processing the stream.
             * The Trick(Executed on:Partition(Locally)) works because stream is also lazily processed.
             * 1 valueColumnPairs at a time.
            //System.out.println("Element per partition:"+partitionIndex+" :"+ai.get());
            <u>List<Long></u> freqList = Arrays.asList(freq);
           List<Tuple2<Integer, List<Long>>> partitionList = Arrays.asList(new Tuple2<>(partitionIndex, freqLi
                                                              . One array per partition,
shown in previous slide.
            return partitionList.iterator();
        }, true).collect();
* Iterator to Iterator transform.
* Powerful and streaming but works because sort already done
 * during shuffle sort phase.
* Executed on:Partition(Locally(ITOI))
return columsFreqPerPartition;
```

## Spark Key-Values: Iterator-To-Iterator: Step-3

```
* Step 3: For each Partition determine the index of the elements that are
 * desired rank statistics
 * For Example: targetRanks: 2,4
 * partitionColumnsFreq: [(0,[5, 5, 3, 1]), (1,[0, 0, 2, 4])]
 * numOfColumns: 4
 * The output will be:
 * ranksLocations: [(0, [(0,2), (0,4), (1,2), (1,4), (2,2)]), (1, [(2,1), (3,1), (3,3)])]
 * (partition,[(columnindx,elementnumberinthatpartition)]
 * @param partitionColumnsFreq Array of (partition index, columns)
 * frequencies per this partition)
                                                                                   ils under is -2.
 * @return Array that contains (partition index, relevantIndexList where
 * relevantIndexList(i) = the index of an element on this partition that
 * matches one of the target ranks)
 * Executed on:Driver/client
private List<Tuple2<Integer, List<Tuple2<Integer, Long>>>> getRanksLocationsWithinEachPart(List<Long> targetRanks,
       List<Tuple2<Integer, List<Long>>> partitionColumnsFreq, int numOfColumns) {
   //running sum to track ranks
```

long[] runningTotal = new long[numOfColumns];

= partitionColumnsFreq

List<Tuple2<Integer, List<Tuple2<Integer, Long>>>> ranksLocations

· Rank 2 of 4 for column-1 occur in partition-land their index are 2, 4.

· Same as above

· For column index-3, rank-2 is in the first partition and

· Kank-4 is in the second partition and its index is -1

· Completely excutes on