

Spark-Homework: Transformation pipeline

Input Tuples --> Cells

- Splitting each record with (value, (Singleton type))

```
datasets.flatMap, columns.map, cells.reduce,  
groupedCellsTuples .toDF
```

Cells --> Cache-based Preaggregation (optional when repeatedly occurring cells with values)

- Should reduce network load
- Groups by the value, and the singleton type

```
-----  
.groupBy (value) .agg (collect_set (singletonType )
```

Cache-based Preaggregation --> Global partitioning

- Reordering the cells among the workers of the cluster through hashing
- *We need an appropriate function to map each different value to unique cell*

$$p(v) \stackrel{def}{=} \text{hash}(v) \bmod n.$$

- Therefore: cells with the same value are on the same worker

```
-----  
  
Done by spark
```

Spark-Homework: Transformation pipeline

Global partitioning --> Attribute Sets

- Grouping all cells by their values
- Aggregating attribute sets using union operator

Attribute Sets --> Inclusion Lists

- Set with n attributes = n inclusion lists (all possible combinations)

Inclusion List --> Partition

- Group by the first attribute

Partition --> Aggregate

- Intersection with preaggregation
- Ends with attributes with empty sets; no (n,0)

Aggregate --> INDs

- Disassembling into INDs

```
.select("aggregatedAttributeSet"), .distinct()
```

```
-----
.select(explode(aggregatedAttributeSet),
.map(row => (row._1, row._2.toList.filter(_ != row._1)))
```

```
-----
Done by spark
-----
```

```
.groupBy(firstAttribute).agg(collect_set(inclusionArray))
.map(row => (row._1, row._2.reduce(_ .intersect(_)))
.filter(row => row._2.nonEmpty)
.sort(firstAttribute)
```

```
-----
.collect()
.foreach(row => println(row._1 + " < " + row._2.reduce(_ +
", " + _)) )
```