# MAP REDUCE HADOOP (STREAMING AND CHAINING JOBS)

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#### Streaming in Hadoop

- One can use Java to write Hadoop programs
- Hadoop supports other languages
  - Via a generic API called Streaming
  - Useful for writing simple, short MapReduce scripts
  - Any scripting/programming language using non-Java libraries
- Hadoop Streaming
  - Interacts with programs using the Unix streaming paradigm
  - Inputs come in through STDIN
  - Outputs go to STDOUT
  - Data has to be text based and each line is considered a record

#### Streaming Data Flow

Similar to the following Unix command line:

```
cat [input file] | [mapper] | sort
                        | [reducer] >[output file]

    Only need to invoke the Streaming jar

bin/hadoop jar [hadoop-streaming-X.Y.Z.jar]
     -input [input file]
      -output [output files]
      -file [script to share]
     -mapper 'mappercommand'
      -reducer 'reducercommand'
```

...

#### Word count per split

```
bin/hadoop jar hadoop-streaming-X.Y.Z.jar
-input input.txt
-output count
-mapper 'wc -l'
-numReduceTasks 0
```

## Working principles

- Works with any executable script
  - Reads a line-oriented data stream from STDIN
  - Outputs to STDOUT with Hadoop Streaming
- Each mapper
  - Sees the entire stream of data
  - Takes on the responsibility of breaking the stream into (line-oriented) record
  - In Java the framework itself breaks input data
- Streaming works on key/value pairs just like in Java
  - The tab character to separate the key from the value in a record
  - When there's no tab character, the entire record is considered the key and the value is empty text

#### Streaming with key/value pairs

#### 1. The mapper

- Reads a split through STDIN
- Extracts each line as a record (can interpret each input record as a key/value pair or a line of text)

#### The framework ensures that

- Each line of your mapper's output is a key/ value pair
- Key/value pairs with the same key will end up at the same reducer

#### 3. The reducer

- key/value pairs are sorted by key
- is responsible for performing the grouping,
- reads one line at a time from STDIN and keeps track of the keys
- 4. The reducer output (STDOUT) is written to a file directly

#### Implement Max via script

```
./bin/hadoop jar hadoop-streaming-X.Y.Z.jar
     -input imput.txt
     -output out.txt
     -mapper 'cut -f1,3'
     -reducer 'maxGroupAttr.py'
# Equivalent to
cat imput.txt | cut -f1,3
     | sort | ./maxGroupAttr.py
```

#### maxGroupAttr.py

```
#!/usr/bin/python
                                  diversamente da
                                  java, qui non sono
import sys
                                  raggruppate per
                                  chiave, quindi
                                  quando la chiave
                                  cambia si inizializza
lastKey=None
                                  dinuovo il conto per
                                  il massimo locale
max = 0
for line in sys.stdin:
          (key, val) = line.strip().split()
          if lastKey and lastKey!=key:
                     print lastKey + "\t" + max
                     max=0
                     lastKey=None
          lastKey=key
          if (val > max):
                     max = val
print lastKey + "\t" + max
```

#### Streaming with the Aggregate package

- The library package called Aggregate
  - Simplifies obtaining aggregate statistics of a data set
  - Can simplify the writing of Java statistics collectors
    - Especially when used with Streaming
- Each line of the mapper's output looks like

function: key\tvalue

 The Aggregate reducer applies the function to the set of values for each key

## Standard aggregation functions

Value aggregator	Description
DoubleValueSum	Sums up a sequence of double values.
LongValueMax	Finds the maximum of a sequence of long values.
LongValueMin	Finds the minimum of a sequence of long values.
LongValueSum	Sums up a sequence of long values.
StringValueMax	Finds the lexicographical maximum of a sequence of string values.
StringValueMin	Finds the lexicographical minimum of a sequence of string values.
UniqValueCount	Finds the number of unique values (for each key).
ValueHistogram	Finds the count, minimum, median, maximum, average, and standard deviation of each value. (See text for further explanation.)

## Generic aggregator example

tipo di valore

```
./bin/hadoop jar hadoop-streaming-X.Y.Z.jar
  -input input.txt
  -output out.txt
                                                             Lo 0 indica la
  -mapper 'funtionMapper.py LongValueSum 0'
                                                             colonna che viene
                                                             influenzata al conto
 -reducer aggregate
./bin/hadoop jar hadoop-streaming-X.Y.Z.jar
  -input input.txt
 -output out.txt
  -mapper 'funtionMapper.py DoubleValueSum 0 2'
 -reducer aggregate
                                                       In questo caso
                                                       vengono considerate
                                                       le colonne 0 e 2.
                                                       dove 0 è la chiave e
       aggregate è la
                                                       2 è il value
       funzione in base al
```

#### functionMapper.py

```
#!/usr/bin/python
import sys
function = (sys.argv[1])
keyIdx = int(sys.argv[2])
if len(sys.argv) > 3:
        valIdx = int(sys.argv[3])
else:
        valIdx = -1
for line in sys.stdin:
        fields = line.strip().split()
        if fields[keyIdx] and not fields[keyIdx].startswith("#"):
                if valIdx>0:
                        print function + ":" +fields[keyIdx] + "\t" + fields[valIdx]
                else:
                        print function + ":" +fields[keyIdx] + "\t" + "1"
```

```
Usage: $HADOOP_HOME/bin/hadoop jar hadoop-streaming.jar [options]
Options:
  -input
                  <path> DFS input file(s) for the Map step.
  -output
                  <path> DFS output directory for the Reduce step.
                  <cmd|JavaClassName> Optional. Command to be run as mapper.
 -mapper
 -combiner
                  <cmd|JavaClassName> Optional. Command to be run as combiner.
 -reducer
                  <cmd|JavaClassName> Optional. Command to be run as reducer.
                  <file> Optional. File/dir to be shipped in the Job jar file.
 -file
                  Deprecated. Use generic option "-files" instead.
                  <TextInputFormat(default)|SequenceFileAsTextInputFormat|JavaClassName>
  -inputformat
                  Optional. The input format class.
  -outputformat
                  <TextOutputFormat(default)|JavaClassName>
                  Optional. The output format class.
                  <JavaClassName> Optional. The partitioner class.
  -partitioner
  -numReduceTasks <num> Optional. Number of reduce tasks.
  -inputreader
                  <spec> Optional. Input recordreader spec.
 -cmdenv
                  <n>=<v> Optional. Pass env.var to streaming commands.
  -mapdebug
                  <cmd> Optional. To run this script when a map task fails.
  -reducedebug
                  <cmd> Optional. To run this script when a reduce task fails.
  -io
                  <identifier> Optional. Format to use for input to and output
                  from mapper/reducer commands
                  Optional. Lazily create Output.
  -lazyOutput
  -background
                  Optional. Submit the job and don't wait till it completes.
 -verbose
                  Optional. Print verbose output.
 -info
                  Optional. Print detailed usage.
                  Optional. Print help message.
  -help
```

## Chaining MapReduce jobs (1)

- Many complex tasks need to be broken down
  - Each accomplished by an individual MapReduce job
- Chaining MapReduce jobs in a sequence
  - Like:

```
mapreduce-1 | mapreduce-2 | ...
```

Just schedule every next job after the other in the main

```
Job.waitForCompletion(true) ? 0 : 1
```

To wait until preceding job completes!!

#### Chaining MapReduce jobs (2)

- Complex chains using the ChainMapper class
  - Like:

alcune volte si fanno anche più operazioni di map prima di un reducing

```
[map]+ | reduce [| map]*
```

pre-processing | standard computation | post-processing

```
ChainMapper.addMapper(job, map1.class, key.class, val.class)
ChainMapper.addMapper(job, map2.class, key.class, val.class)
ChainMapper.setReducer(job, red.class, key.class, val.class)
ChainMapper.addMapper(job, map3.class, key.class, val.class)
...
```

Job.waitForCompletion(true) ? 0 : 1

#### Chaining MapReduce jobs (3)

- Enforce specific dependencies
  - Specify like a chain of dependencies
  - Using JobControl and ControlledJob classes

```
ControlledJob cJ1 = new ControlledJob(job1)
ControlledJob cJ2 = new ControlledJob(job2)
cJ1.addDependingJob(cJ2) cj1 deve finire per runnare cj2
JobControl jC = new JobControl("A chain")
jC.addJob(cJ1)
jC.addJob(cJ1)
jC.run()
```

## LET'S START

(open, configure and run examples)

#### Required Software

- Eclipse
  - www.eclipse.org
- Hadoop
  - Download it from <a href="http://hadoop.apache.org/">http://hadoop.apache.org/</a>
  - Obtain everything it via Maven (<a href="https://mvnrepository.com/artifact/org.apache.hadoop">https://mvnrepository.com/artifact/org.apache.hadoop</a>)
- Small examples do not need a cluster
  - Running Hadoop in Standalone mode
- Terminal app
- Few python scripts
- Few data files