

Big data: architectures and data analytics

MapReduce Programming Paradigm and Hadoop – Part 2

Combiner

Combiner

- “Standard” MapReduce applications
 - The (key,value) pairs emitted by the Mappers are sent to the Reducers through the network
- Some “pre-aggregations” could be performed to limit the amount of network data

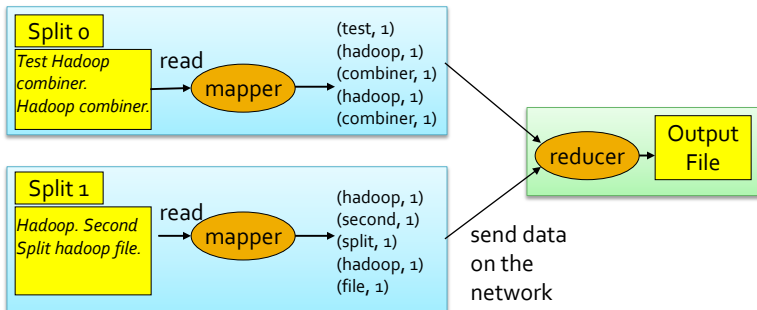
3

Combiner – Word count example

- Consider the standard word count problem
- Suppose the input file is split in two Input Splits
 - Hence, two Mappers are instanced (one for each split)

4

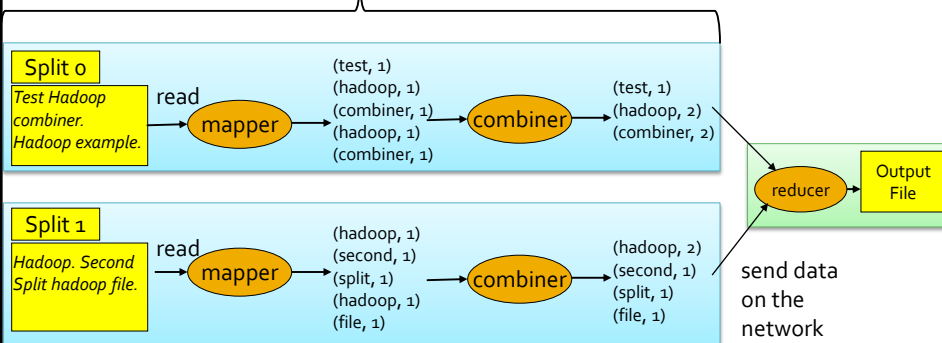
Word count example without combiner



5

Word count example with combiner

The combiner is **locally** called on the output (key, value) pairs of the mapper (works on data stored in main-memory)



Combiner

- MapReduce applications with combiners
 - The (key,value) pairs emitted by the Mappers are analyzed in main-memory and aggregated by the Combiners
 - The combiner performs some “pre-aggregations” to limit the amount of network data
 - Each combiner pre-aggregates the values associated with the same key emitted by a Mapper
- Works only if the reduce function is commutative and associative

7

Combiner

- The Combiner
 - Is an instance of the `org.apache.hadoop.mapreduce.Reducer` class
 - There is not a specific combiner-template class
 - “Implements” a pre-reduce phase
 - Is characterized by the `reduce(...)` method
 - Processes (key, [list of values]) pairs and emits (key, value) pairs
 - Runs on the cluster

8

MapReduce programs - Combiner

- The Combiner class extends the `org.apache.hadoop.mapreduce.Reducer` class
 - The `org.apache.hadoop.mapreduce.Reducer` class
 - Is a generic type/generic class
 - With four type parameters: input key type, input value type, output key type, output value type
- The designer/developer implements the `reduce(...)` method
 - That is automatically called by the framework for each (key, [list of values]) pair obtained by aggregating the output of a mapper
- i.e., Combiners and Reducers extend the same class

9

MapReduce programs - Combiner

- The Combiner class is specified by using the `job.setCombinerClass()` method in the `run` method of the Driver
 - i.e., in the job configuration part of the code

10

MapReduce Programming Paradigm and Hadoop – Part 2

Personalized Data Types

11

Personalized Data Types and Values

- Personalized Data Types are useful when the value of a key-value pair is a complex data type
- Personalized Data Types are defined by implementing the `org.apache.hadoop.io.Writable` interface
 - The following methods must be implemented
 - `public void readFields(DataInput in)`
 - `public void write(DataOutput out)`
 - To properly format the output of the job usually also the following method is “redefined”
 - `public String toString()`

12

Personalized Data Types - Example

- Suppose to be interested in “complex” values composed of two parts:
 - a counter (int)
 - a sum (float)
- An ad-hoc Data Type can be used to implement this complex data type in Hadoop

13

Personalized Data Types - Example (1)

```
package it.polito.bigdata.hadoop.combinerexample;
import java.io.DataInput;
import java.io.DataOutput;
import java.io.IOException;

public class SumAndCountWritable implements
    org.apache.hadoop.io.Writable {
    /* Private variables */
    private float sum = 0;
    private int count = 0;
```

14

Personalized Data Types - Example (2)

```
/* Methods to get and set private variables of the class */
public float getSum() {
    return sum;
}

public void setSum(float sumValue) {
    sum=sumValue;
}

public int getCount() {
    return count;
}

public void setCount(int countValue) {
    count=countValue;
}
```

15

Personalized Data Types - Example (3)

```
/* Methods to serialize and deserialize the contents of the
instances of this class */
@Override /* Serialize the fields of this object to out */
public void write(DataOutput out) throws IOException {
    out.writeFloat(sum);
    out.writeInt(count);
}

@Override /* Deserialize the fields of this object from in */
public void readFields(DataInput in) throws IOException {
    sum=in.readFloat();
    count=in.readInt();
}
```

16

Personalized Data Types - Example (4)

```
/* Specify how to convert the contents of the instances of this
class to a String
* Useful to specify how to store/write the content of this class
* in a textual file */
public String toString()
{
    String formattedString=
        new String("sum="+sum+",count="+count);

    return formattedString;
}
}
```

17

Personalized Data Types and Keys

- Personalized Data Types can be used also to manage complex keys
- In that case the Personalized Data Type must implement the `org.apache.hadoop.io.WritableComparable` interface
 - Because keys must be comparables

18

MapReduce Programming Paradigm and Hadoop – Part 2

Sharing parameters among Driver, Mappers, and Reducers

19

Sharing parameters among Driver, Mappers, and Reducers

- The configuration object is used to share the (basic) configuration of the Hadoop environment across the driver, the mappers and the reducers of the application/job
- It stores a list of (property-name, property-value) pairs
- Personalized (property-name, property-value) pairs can be specified in the driver
 - They can be used to share some parameters of the application with mappers and reducers

20

Sharing parameters among Driver, Mappers, and Reducers

- Personalized (property-name, property-value) pairs are useful to shared small (constant) properties that are available only during the execution of the program
 - The driver set them
 - Mappers and Reducers can access them
 - Their values cannot be modified by mappers and reducers

21

Sharing parameters among Driver, Mappers, and Reducers

- In the driver
 - `Configuration conf = this.getConf();`
 - Retrieve the configuration object
 - `conf.set("property-name", "value");`
 - Set personalized properties
- In the Mapper and/or Reducer
 - `context.getConfiguration().get("property-name")`
 - This method returns a String containing the value of the specified property

22