

LAB 3: MAPREDUCE

Q1: How many times does the word Sherlock appear in the file?

The propose of part 1 is to statistic how many times a word occurs in the text. When we are using Hadoop to do the distributed calculation, what we should consider are the MAP job, REDUCE job and a MAIN function which invokes the program and configure the nodes. MAP is responsible to spilt all the word in the text and emit each word with a key-value pair, like (“word”, 1) and send word-iterationList like (“word”, 1, 1, 1...) to REDUCE nodes. REDUCE is responsible for aggregating the iteration list for each word and generating a (“word”, count) pair for each unique word. Notice that a word is only sent to a REDUCE node. So the result of the REDUCE is the word-wordCount pair. To find the times that “Sherlock” appears, we only need to check the file and find the count, which is 345.

In *TokenizerMapper.java*,

```
StringTokenizer itr = new StringTokenizer(value.toString(), "-- \t\n\r\f,.;?![]()'\",");
```

means spilt the text into a list if it meets the following syntax: “-- \t\n\r\f,.;?'\”,”.

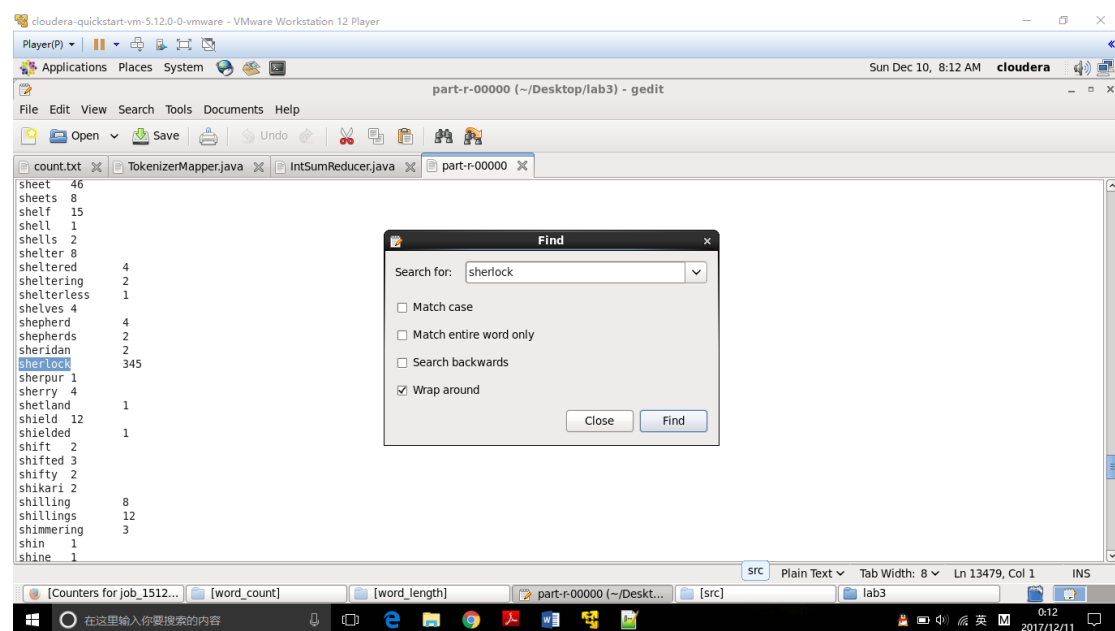
In *IntSumReducer.java*,

```
sum += value.get();
```

value.get() is to turn IntWritable type into int type; add the element of iterable list to sum.

```
context.write(key, result);
```

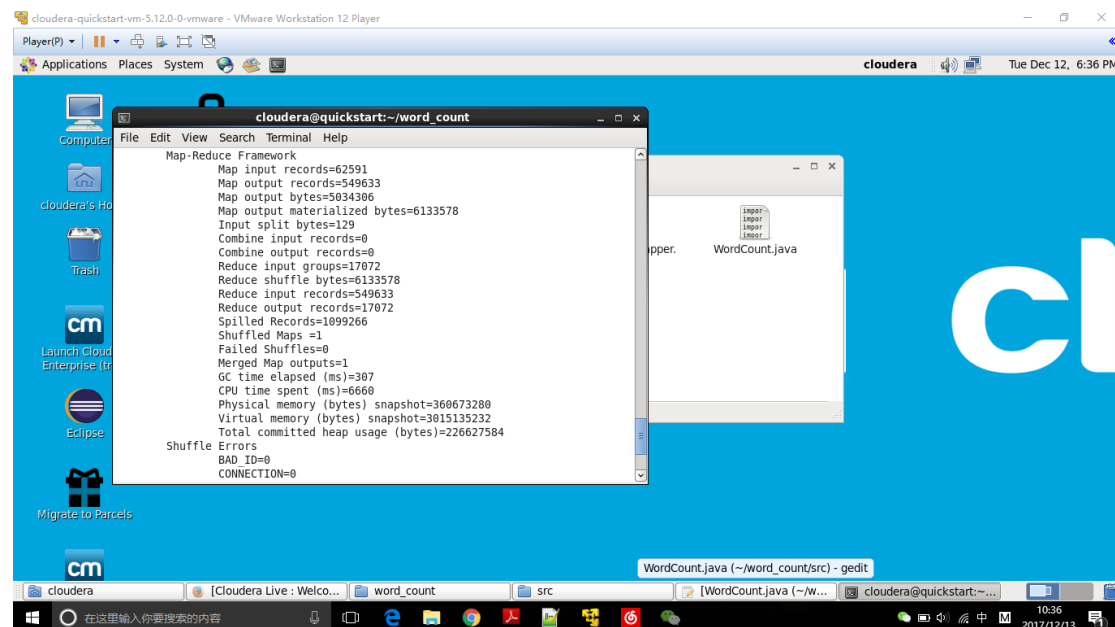
emit the (word, count) pair.



Q2: The Hadoop job outputs a set of statistics after the completion. What was the number of key-value pairs generated by the Mapper, the number of unique inputs to the Reducer, and the final number of records emitted by the Reducer? Based on these values, answer the following two questions: How many words has the document in total? How many unique words does the document have?

According to our MAP method, which emits (“words”, 1) for each word, the number of key-value

pairs generated by the Mapper is actually the total number of words in the text. From the screenshot, we can see the value is 549633. The number of unique inputs to the Reducer is the Reduce input group which is 17072. The final number of records emitted by the Reducer is 17072 since the reducer adds the occurrence times of each word, so it is also equal to the number of unique words. So base on the analysis above, the document has 549633 words in total. There are 17072 unique words in the document.



The screenshot shows a Cloudera Quickstart VM environment. A terminal window titled 'cloudera@quickstart:~/word_count' displays the output of a MapReduce job. The statistics are as follows:

```
Map-Reduce Framework
Map input records=62591
Map output records=549633
Map output bytes=5034306
Map output materialized bytes=6133578
Input split bytes=129
Combine input records=0
Combine output records=0
Reduce input groups=17072
Reduce shuffle bytes=6133578
Reduce input records=549633
Reduce output records=17072
Spilled Records=1099266
Shuffled Maps=1
Failed Shuffles=0
Merged Map outputs=1
GC time elapsed (ms)=307
CPU time spent (ms)=6660
Physical memory (bytes) snapshot=360673280
Virtual memory (bytes) snapshot=3015135232
Total committed heap usage (bytes)=226627584

Shuffle Errors
BAD_ID=0
CONNECTION=0
```

In the background, a file explorer window shows the 'WordCount.java' file in the 'src' directory. The taskbar at the bottom shows the Cloudera logo and the file explorer window.

Q3: How many Map and Reduce tasks does your MapReduce job have? Can you explain the difference with the Sherlock job in Part 1?

Since we added "*job.setNumReduceTasks(2);*" in our "main class", which means we set two Reduce nodes to do the job. So we have 1 Map task and 2 Reduce tasks. Notice that the task in 2 Reduce nodes is executed parallel and each node is allocated about half of the total output of Map job. So compared with Sherlock job which only has 1 Reduce node, the Reduce job is faster in time in the case of the same Reduce input.

cloudera-quickstart-vm-5.12.0-0-vmware - VMware Workstation 12 Player

Player(P) Thu Dec 7, 8:11 PM cloudera

Applications Places System

Map Tasks for job_1512703268494_0001 - Mozilla Firefox

Cloudera Live : Welcom... x Map Tasks for job_1512... x

quickstart.cloudera:19888/jobhistory/tasks/job_1512703268494_0001/m

Cloudera Hue Hadoop HBase Impala Spark Solr Oozie Cloudera Manager Getting Started

Logged in as: drwho

Map Tasks for job_1512703268494_0001

Application

- Job
 - Overview
 - Counters
 - Configuration
 - Map tasks
 - Reduce tasks
- Tools

Show 20 entries

Task					Successful Attempt		
Name	State	Start Time	Finish Time	Elapsed Time	Start Time	Finish Time	Elapsed Time
task_1512703268494_0001_m_000000	SUCCEEDED	Thu Dec 7 19:53:04 -0800 2017	Thu Dec 7 19:54:49 -0800 2017	1mins, 45sec	Thu Dec 7 19:53:04 -0800 2017	Thu Dec 7 19:54:49 -0800 2017	1mins, 45sec

Showing 1 to 1 of 1 entries

word_count

Map Tasks for job_151... [lab3] [cloudera@quickstart:... cloudera word_count cloudera@quickstart:...]

12:11 2017/12/8

Applications Places System Thu Dec 7, 8:11 PM cloudera

Reduce Tasks for job_1512703268494_0001 - Mozilla Firefox

Cloudera Live : Welcom... x Reduce Tasks for job_15... x

quickstart.cloudera:19888/jobhistory/tasks/job_1512703268494_0001/r

Cloudera Hue Hadoop HBase Impala Spark Solr Oozie Cloudera Manager Getting Started

Logged in as: drwho

Reduce Tasks for job_1512703268494_0001

Task					Successful Attempt							
State	Start Time	Finish Time	Elapsed Time	Start Time	Shuffle Time	Shuffle Finish Time	Merge Finish Time	Finish Time	Elapsed Time Shuffle	Elapsed Time Merge	Elapsed Time Reduce	Elapsed Time
.0001_r_000000	SUCCEEDED	Thu Dec 7 19:54:53 -0800 2017	Thu Dec 7 19:55:48 -0800 2017	54sec	Thu Dec 7 19:54:53 -0800 2017	Thu Dec 7 19:55:17 -0800 2017	Thu Dec 7 19:55:33 -0800 2017	Thu Dec 7 19:55:48 -0800 2017	23sec	16sec	14sec	54sec
.0001_r_000001	SUCCEEDED	Thu Dec 7 19:54:54 -0800 2017	Thu Dec 7 19:55:43 -0800 2017	48sec	Thu Dec 7 19:54:54 -0800 2017	Thu Dec 7 19:55:16 -0800 2017	Thu Dec 7 19:55:26 -0800 2017	Thu Dec 7 19:55:43 -0800 2017	21sec	10sec	16sec	48sec

Showing 1 to 2 of 2 entries

word_count

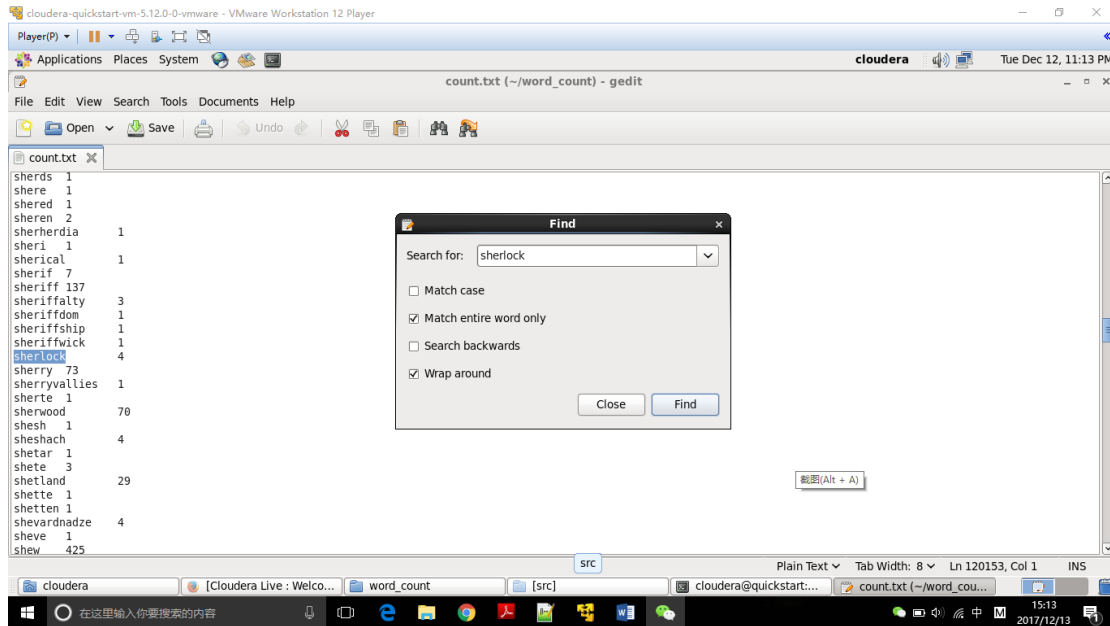
Reduce Tasks for job_1... [lab3] [cloudera@quickstart:... cloudera word_count cloudera@quickstart:...]

Q4: Can you see a clear pattern on how Hadoop partitions the keys among multiple reducers? Does it make it easier or harder the problem of manually retrieving information about a specific key? E.g. Answering questions such as "How many times the word Sherlock appears in the text?"

I can't see a clear pattern on how Hadoop partitions the keys. All pairs of a key are allocated to one reducer. It is the internal mechanism of Hadoop to handle which key is distributed to which Reduce node and it should be balance for each node.

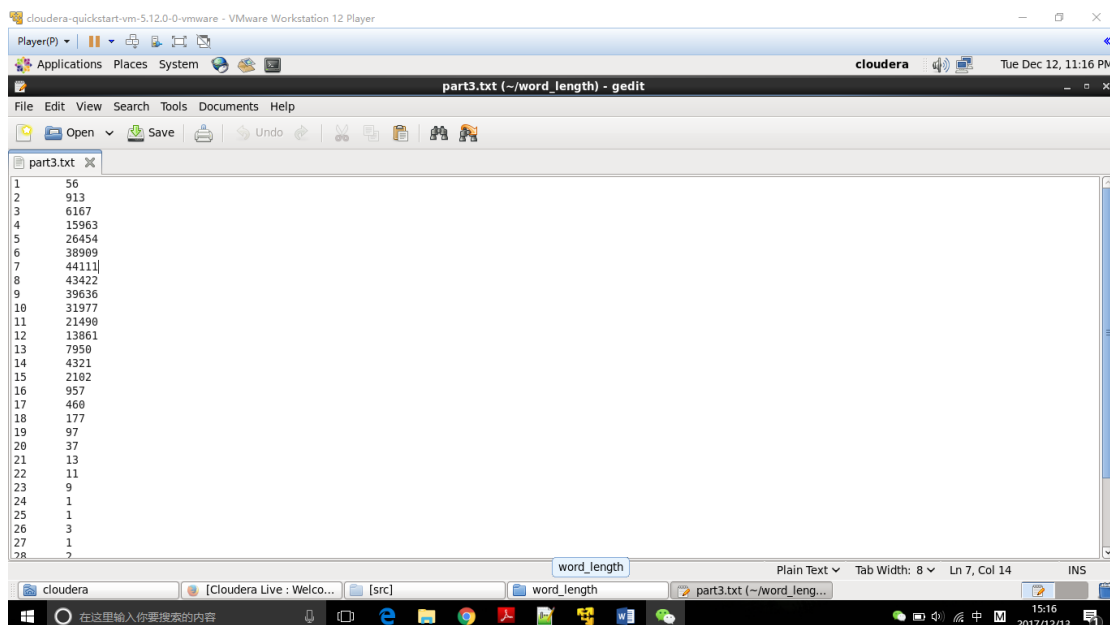
Since there are two Reduce nodes to handle the aggregation process and the Reduce tasks is parallel, I think it make it easier to handle the problem.

Sherlock appears 4 times in the text.



Q5: What is the most common word length for the onehundredM.txt dataset? Briefly explain the Mapper and Reducer code that you have implemented.

According to the screenshot below, we find that the most common word length for the onehundredM.txt dataset is 7.



The Mapper code is shown below:

```
import java.io.IOException;
import java.util.StringTokenizer;
import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.io.LongWritable;
import org.apache.hadoop.mapreduce.Mapper;

public class TokenToLengthMapper extends Mapper<Object, Text, IntWritable, IntWritable> {
    private final IntWritable word_length = new IntWritable();
    public void map(Object key, Text value, Context context) throws IOException, InterruptedException {
        //This map method is called for every line in the text file
        StringTokenizer itr = new StringTokenizer(value.toString(), "\t");
        while(itr.hasMoreTokens())
        {
            String[] components = itr.nextToken().toString().split("\t");
            word_length.set(components[0].length());
            context.write(word_length, one);
        }
    }
}
```

The input of Map is the document: count.txt, which generated in part 2. So the input data will be $\langle \text{Object}, \text{Text} \rangle$. Then we want to emit pair for each unique word, which is $\langle \text{word_length}, 1 \rangle$. So the output key and value are both *IntWritable*. The format of the file has each pair in a line and the key (unique word) and value (how many times the word occurs) is divided by “\t”. After we scan the whole file, the value is the whole text. Then we spilt each line through *StringTokenizer* and for each line, we spilt it to a string array. In this case, components[0] represents the word, components[1] represents the times of the words occur. Then we acquire the length of the word and emit the pair (*word_length*, 1).

The Reduce code is shown below:

```
import java.io.IOException;
import java.util.Iterator;
import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.mapreduce.Reducer;
import org.apache.hadoop.mapreduce.Reducer;

public class IntSumReducer extends Reducer<IntWritable, IntWritable, IntWritable, IntWritable> {
    //The above <Test, IntWritable, Text IntWritable> tells Java the type of <KEY-IN, VALUE-IN, KEY-OUT, VALUE-OUT>.
    private IntWritable result = new IntWritable(); //a int value in Madoop - where sum result should be stored

    public void reduce(IntWritable key, Iterable<IntWritable> values, Context context) throws IOException, InterruptedException {
        //This reduce method will be called for each key (i.e. each word). Values will contain all the values that have been recorded for that word...
        int sum = 0;

        for (IntWritable value : values) {
            sum += value.get();
        }
        //complete code here! HINT: You need to go through each value and sum up the result!

        result.set(sum); //sets result to value of sum
        context.write(key, result);
        //complete code here! HINT: You need to emit the outputs!
    }
}
```

Since the type that Map emit is $\langle \text{IntWritable}, \text{IntWritable} \rangle$, the input key should be *IntWritable* and the value is a *IntWritable* list. The output of Reduce is also pairs $\langle \text{IntWritable}, \text{IntWritable} \rangle$. Since

Reduce method is called for each key, what we need to do is to add the 1 of the list and emit the pair (word_length, sum_of_uniqueWord_has_the_same_length).