Big Data Processing and Analytics: Assignment 1

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In this assignment we are going to create an inverted index on a corpus including the complete works of William Shakespear, Mark Twain and Jane Austen using Hadoop's MapReduce framework. The source code is available on Github.

1 Setup

(a) System specifications

• Operating system:

Ubuntu 16.04 (Native)

• System specifications:

Model: Dell Inspiron 17R 5720

Processor: i5-3210M

Cores: 2 Threads: 4 Ram: 8 GB

Storage: 256GB SSD (MLC)

• Java version:

openidk version "1.8.0_121"

OpenJDK Runtime Environment (build 1.8.0_121-8u121-b13-0ubuntu1.16.04.2-b13)

OpenJDK 64-Bit Server VM (build 25.121-b13, mixed mode)

• Haddop version:

Hadoop 2.7.3

Subversion https://git-wip-us.apache.org/repos/asf/hadoop.git-r baa91f7c6bc9cb92be5982de4719c1c8af91ccff Compiled by root on 2016-08-18T01:41Z

Compiled with protoc 2.5.0

From source with checksum $2\mathrm{e}4\mathrm{c}e5\mathrm{f}957\mathrm{e}a4\mathrm{d}b193\mathrm{b}ce3734\mathrm{f}f29\mathrm{f}f4$

 $This\ command\ was\ run\ using\ /usr/local/hadoop/share/hadoop/common/hadoop-common-2.7.3. jar and the command was run using /usr/local/hadoop/share/hadoop/common/hadoop-common-2.7.3. jar and the command was run using /usr/local/hadoop/share/hadoop/common/hadoop-common-2.7.3. jar and the command was run using /usr/local/hadoop/share/hadoop/common/hadoop-common-2.7.3. jar and the common-2.7.3. jar and th$

Hadoop was installed using this tutorial and configured using this tutorial.

(b) Configuration

The configuration comes from the official documentation for a single Hadoop node cluster. The following configuration files allows Hadoop and YARN to run in a pseudo-distributed mode.

• core-site.xml:

1 <configuration>

2

3 </configuration>

• hdfs-site.xml:

```
1 <configuration>
2
3 </configuration>
```

• mapred-site.xml:

```
1 <configuration>
2
3 </configuration>
```

• yarn-site.xml:

```
1 < configuration>
2
3 </ configuration>
```

• Commands to set up HDFS and YARN:

```
1
         # Format the filesystem
2
         hdfs namenode -format
3
         # Start HDFS
4
          start-dfs.sh
5
6
         # Create directories to execute MapReduce jobs
         hdfs dfs -mkdir /user
7
8
          hdfs dfs -mkdir /user/louis
9
         # Put the data in HDFS
10
         hdfs dfs -put ~/dev/bdpa/a1/data data
11
12
13
         # Start YARN Ressource manager
14
          start-yarn.sh
```

(c) Bugs encountered

I encountered a bug while running the MapReduce jobs where I would be logged out and all my processes would be killed with no warning. After investigation, I found that this is a known bug referenced here and caused by /bin/kill in ubuntu 16.04.

I compiled propcps-3.3.10 from source to solve the problem as indicated in launchpad but to no avail.

```
# (1) download the sourcecode
1
2
    sudo apt-get source procps
3
    # (2) install dependency
4
    sudo apt-get build-dep procps
5
6
7
    # (3) compile procps
    cd procps -3.3.10
8
9
    sudo dpkg-buildpackage
```

I tried another solution from this stackoverflow thread but it didn't work either. The supposed fix was to add

- 1 [login]
- 2 KillUserProcesses=no
 - to /etc/systemd/logind.conf and restart.

As I could find a solution I had to cope with the problem and restart all the services every once in a while.

2 Inverted index

Workflow

In order to facilitate compiling the .java files into .jar files, I created the compile.sh script which takes as input parameter the name of the main class of the .java file.

Example: ./compile.sh StopWords.

compile.sh:

```
1 #!/bin/bash
2 rm -rf *.class *.jar;
3 hadoop com.sun.tools.javac.Main $1.java;
4 jar cf $1.jar $1*.class;
```

Assumptions

After investigating the results of the different map reduce jobs, I chose to define all of the following characters as words separators in addition to space

```
Separators: .,?!"'()[]$*-_;:|
```

Even if I lost some hyphenated words and contractions like "first-born" or "He's", the gain in coherence and clarity was worth the hassle.

(a) Stop words

(30) Run a MapReduce program to identify stop words (words with frequency > 4000) for the given document corpus. Store them in a single csv file on HDFS (stopwords.csv). You can edit the several parts of the reducers' output after the job finishes (with hdfs commands or with a text editor), in order to merge them as a single csv file.

Based on the wordcount example from the official documentation, we implement a MapReduce program that retrieves all the stopwords from a corpus, i.e. it retrieves the words with wordcount greater than 4000. The file is **StopWords.java**

• Mapper:

This mapper splits a string into words (tokens) and outputs one (key, value) pair for each word with the key being the word and the value equal to 1.

```
18
     public static class TokenizerMapper
          extends Mapper<Object, Text, Text, IntWritable>{
19
20
21
       private final static IntWritable one = new IntWritable(1);
22
       private Text word = new Text();
23
24
       public void map(Object key, Text value, Context context
                        ) throws IOException, InterruptedException {
25
26
         // Splits a string to tokens (here words)
27
         StringTokenizer itr = new StringTokenizer(value.toString(), "
             .,?!\"'()[]$*-_;:|");
         while (itr.hasMoreTokens()) {
28
29
           word.set(itr.nextToken().toLowerCase().trim());
30
            // Write one (key, value) pair to context
31
            context.write(word, one);
32
         }
33
       }
     }
34
```

• Reducer:

The reducers, simply counts the number of occurrences of each key writes it to the output file only if its count is greater than 4000.

```
public static class IntSumReducer
36
37
           extends Reducer<Text, IntWritable, Text, IntWritable> {
38
        private IntWritable result = new IntWritable (10);
39
        public void reduce (Text key, Iterable < IntWritable > values,
40
41
                            Context context
                            ) throws IOException, InterruptedException {
42
43
          int sum = 0;
44
          for (IntWritable val : values) {
45
            sum += val.get();
46
47
          // Only write the key value pair if its frequency is high enough
48
          if (sum > 4000) {
            result.set(sum);
49
50
            context.write(key, result);
51
52
        }
53
     }
```

• Job configuration:

The MapReduce task is set to write the output as a csv file. We can set the number of reducers, combiner and compression through command line arguments.

```
55
     public static void main(String[] args) throws Exception {
       Configuration conf = new Configuration();
56
       // Remove output folder if it exists
57
       Path output = new Path(args[1]);
58
59
       FileSystem hdfs = FileSystem.get(conf);
60
       // delete existing directory
       if (hdfs.exists(output)) {
61
62
            hdfs.delete(output, true);
63
       }
64
65
       // Set separator to write as a csv file
       conf.set("mapred.textoutputformat.separator", ", ");
66
67
       // Set compression
       if ((args.length >= 5) \&\& (Integer.parseInt(args[4]) == 1)) {
68
69
            conf.set("mapreduce.map.output.compress", "true");
70
       }
71
72
       Job job = Job.getInstance(conf, "stop words");
73
       job.setJarByClass(StopWords.class);
74
75
       // Set number of reducers and combiner through cli
76
       if (args.length >= 3) {
77
           job.setNumReduceTasks(Integer.parseInt(args[2]));
78
79
       if ((args.length >= 4) && (Integer.parseInt(args[3]) == 1)) {
80
           job.setCombinerClass(IntSumReducer.class);
81
82
       job.setMapperClass(TokenizerMapper.class);
83
       job.setReducerClass(IntSumReducer.class);
84
```

```
85
        job.setOutputKeyClass(Text.class);
86
        job.setOutputValueClass(IntWritable.class);
        FileInputFormat.addInputPath(job, new Path(args[0]));
87
88
        FileOutputFormat.setOutputPath(job, new Path(args[1]));
        long startTime = System.nanoTime();
89
90
        if (job.waitForCompletion(true)) {
91
          long endTime = System.nanoTime();
92
          float duration = (endTime - startTime);
93
          duration = 1000000000;
94
          System.out.println("***** Elapsed: " + duration + "s ****\n");
95
          System. exit(0);
96
97
        else {
98
          System.exit(1);
99
100
      }
    }
101
```

• Script running the experiments:

We run the set of experiments with the **run_stopwords.sh** script. This script executes the StopWords MapReduce task with different parameters and merges all the outputs into a csv file in HDFS.

```
1
   \#!/bin/bash
2
   ./compile.sh StopWords
3
4
   # Usage:
5
   # hadoop jar myfile.jar Class input_dir output_dir n_reducers combiner
       compression
6
   # 10 reducers no combiner
   hadoop jar StopWords.jar StopWords data/corpus/ out 10 0 0;
   hdfs dfs -getmerge out stopwords.csv;
10
11
   # 10 reducers, combiner
12
   hadoop jar StopWords.jar StopWords data/corpus/ out 10 1 0;
   hdfs dfs -getmerge out stopwords.csv;
13
14
15
   # 10 reducers, combiner, compression
   hadoop jar StopWords.jar StopWords data/corpus/ out 10 1 1;
17
   hdfs dfs -getmerge out stopwords.csv;
18
19
   # 50 reducers, combiner, compression
   hadoop jar StopWords.jar StopWords data/corpus/ out 50 1 1;
   hdfs dfs -getmerge out stopwords.csv;
```

\bullet **Results**: Here is an extract of the output csv:

```
10 with, 35591
11 good, 4632
12 from, 9677
13 has, 5190
14 its, 4569
15 man, 5267
16 made, 4046
17 not, 35143
18 said, 8434
19 have, 24625
```

20 my, 27231

i. (10) Use 10 reducers and do not use a combiner. Report the execution time.

The running time with 10 reducers is **126 seconds**.



ii. (10) Run the same program again, this time using a Combiner. Report the execution time. Is there any difference in the execution time, compared to the previous execution? Why?

The running time with 10 reducers and combiner is **105 seconds**. The running time is lower with the combiner. Indeed the combiner takes the output of each mapper separately and tries to reduce as many (key, value) pairs coming from this specific mapper as it can. Combining the data like this will lower the load for the reducers, and all the intermediary steps between the mappers and the reducers such as network transfer (does not occur for a pseudo-distributed cluster), sorting all the (key, value) pairs, assigning them to each reducer...



iii. (5) Run the same program again, this time compressing the intermediate results of map (using any codec you wish). Report the execution time. Is there any difference in the execution, time compared to the previous execution? Why?

Compression compresses the data between the mappers and the reducers. The running time with 10 reducers, combiner and compression is **125 seconds**. The running time increased. This can be explained by the fact that compression adds an additional overhead and its benefit is not used for a single node cluster! Indeed compression

is often used to reduce network transfer times, which does not occur here as all the mappers and reducers are on the same device.



iv. (5) Run the same program again, this time using 50 reducers. Report the execution time. Is there any difference in the execution time, compared to the previous execution? Why?

The running time with 50 reducers, combiner and compression is **126 seconds**. The running time is about the same as the previous experiment. We did not benefit from the additional number of reducers. This is logical because with a single node cluster, the processing power caps with the limited number of cores. No additional gain is achieved with this extra parallelization.



(b) (30) Implement a simple inverted index for the given document corpus, as shown in the previous Table, skipping the words of stopwords.csv.

The inverted index is implemented in **InvertedIndex.java**.

• readStopWords function:

First we need to read the stopwords from the csv file. We do it with the readStopWords function.

```
52 public static HashSet<String> readStopWords() {
53  // Read stopwords into a HashSet for fast membership testing
54  // The hashtable underlying structure provides O(1) membership testing
```

```
HashSet<String> stopWords = new HashSet<String>();
55
56
57
          // Read csv file: inspired from https://www.mkyong.com/java/how-to-
             read-and-parse-csv-file-in-java/
          String csvFile = "stopwords.csv";
58
59
          BufferedReader br = null;
          String line = "";
60
61
          String cvsSplitBy = ",";
62
63
          try {
              br = new BufferedReader(new FileReader(csvFile));
64
              while ((line = br.readLine()) != null) {
65
66
                  // use comma as separator
67
                  String[] splittedLine = line.split(cvsSplitBy);
68
                  stopWords.add(splittedLine[0]);
69
70
          } catch (FileNotFoundException e) {
71
              e.printStackTrace();
72
          } catch (IOException e) {
73
              e.printStackTrace();
          } finally {
74
              if (br != null) {
75
76
                  try {
77
                      br.close();
78
                  } catch (IOException e) {
79
                       e.printStackTrace();
80
81
82
83
         return stopWords;
84
     }
```

• Mapper:

Our mapper here outputs only words which are not in stopwords.csv as keys and the file from which they came from as values (formatted as a posting list). These posting lists are implemented using a custom class of MapWritable in order to have all the ComparableWritable properties that Hadoop's MapReduce needs (more details below). The posting list returned by the mapper is just a map containing one element which is a (key, value) pair of the form (filename, 1).

```
87
      public static class TokenizerMapper
88
           extends Mapper<Object, Text, Text, PostingListWritable >{
89
        private Text word = new Text();
90
        private Text doc = new Text():
        private IntWritable one = new IntWritable(1);
91
92
        private PostingListWritable postingList = new PostingListWritable();
93
        private HashSet<String> stopWords = InvertedIndex.readStopWords();
94
95
        public void map(Object key, Text value, Context context
96
                         ) throws IOException, InterruptedException {
97
           FileSplit fileSplit = (FileSplit) context.getInputSplit();
98
          String filename = fileSplit.getPath().getName();
99
          doc.set(filename);
100
          // Splits a string to tokens (here words)
101
102
          StringTokenizer itr = new StringTokenizer(value.toString(), "
              .,?!\"'()[]$*-_;:|");
103
          String token = new String();
```

```
104
           while (itr.hasMoreTokens()) {
105
             token = itr.nextToken().toLowerCase().trim();
106
             if (!stopWords.contains(token)) {
               word.set(token);
107
               // Output is a PostingListWritable containing the filename and the
108
                    value 1
109
               postingList.clear();
110
               postingList.put(doc, one);
111
               // Write one (key, value) pair to context
112
               context.write(word, postingList);
113
             }
          }
114
        }
115
116
      }
```

• Reducer:

Our reducer reduces all the posting lists it receives and also counts the number of occurences of each word for the following frequency part (more details below).

```
124
        public void reduce (Text word, Iterable < PostingListWritable > postingLists
125
                             Context context
126
                             ) throws IOException, InterruptedException {
127
          // We are going to aggregate all posting lists together
          postingList.clear();
128
          // Iterate through all posting lists comming from the mappers or the
129
              combiners
130
          for (PostingListWritable mw : postingLists) {
131
             // Iterate through the entries of one given posting list
             for (PostingListWritable.Entry<Writable, Writable> entry : mw.
132
                entrySet()) {
133
               Text doc = (Text) entry.getKey();
134
               IntWritable newValue = (IntWritable) entry.getValue();
135
               if (postingList.containsKey(doc)) {
                 IntWritable oldValue = (IntWritable) postingList.get(doc);
136
                 postingList.put(doc, new IntWritable(oldValue.get() + newValue.
137
                    get());
138
               } else {
                 postingList.put(doc, new IntWritable(newValue.get()));
139
140
            }
141
142
143
144
          isLastWordInUniqueDoc = (postingList.size() == 1);
145
          context.write(word, postingList);
146
      }
147
```

(c) (10) How many unique words exist in the document corpus (excluding stop words)? Which counter(s) reveal(s) this information? Define your own counter for the number of words appearing in a single document only. What is the value of this counter? Store the final value of this counter on a new file on HDFS.

The counter that counts the unique words in the documents is **TaskCounter.REDUCE_INPUT_GROUPS**. This counter counts the number of keys which is exactly the number of unique words.

}

166

In order to count the words appearing in a single document only, we implement the following counter:

```
31
      public static enum WordCounter {
        WORDS_IN_UNIQUE_DOC
32
33
    Which is incremented in the reducer's reduce function:
160
              if (isLastWordInUniqueDoc) {
                // Increment counter for word appearing in a single document only
161
                // We are guaranteed that this is the only time that the counter
162
                // will be incremented for this word because all the values from
163
                // a given key all go to the same reduce call.
164
165
                context.getCounter(WordCounter.WORDS_IN_UNIQUE_DOC).increment(1);
```

The counter gives us a count **36343 words appearing in a single document only** which seems pretty high given the number of total unique words 56491 (excluding about 140 stopwords).

After investigation, a lot of words appear only in pg3200.txt which are the Mark Twain works. This file is roughly 300.000 lines long, which is about three times longer than the other files. This might explained why it contains a lot more vocabulary specific than the rest.

The counter value is stored in a file in HDFS:

```
// Write the count of unique words to a file in HDFS
Path filePath = new Path("words_in_unique_file.txt");
if (hdfs.exists(filePath)) {
   hdfs.delete(filePath, true);
}

FSDataOutputStream fin = hdfs.create(filePath);
fin.writeUTF(message);
fin.close();
```

(d) (30) Extend the inverted index of (b), in order to keep the frequency of each word for each document. You are required to use a Combiner.

$\bullet \ PostingListWritable:$

As explained before we implemented a custom class extending MapWritable in order to have all the ComparableWritable properties that hadoop needs and to be able to use a combiner along with a reducer without losing the word frequencies during the transfer. The custom class allows pretty printing a MapWritable instance to a file.

```
36
     public static class PostingListWritable extends MapWritable {
       // Creates a custom class that overrides the toString method to pretty
37
38
       // print the posting list
       @Override
39
40
       public String toString() {
41
          String stringRepr = new String();
42
         for (PostingListWritable.Entry<Writable, Writable> entry: this.
             entrySet()) {
           stringRepr += entry.getKey() + "#" + entry.getValue() + ", ";
43
44
45
         // Remove last comma and space of string
46
         stringRepr = stringRepr.substring(0, stringRepr.length()-2);
47
         return stringRepr;
48
49
     }
```

• Combiner:

Different posting lists coming from different mappers or combiners can be reduced into a single one using the reduce function of the combiner.

```
public static class PostingListCombiner
119
120
           extends Reducer<Text, PostingListWritable, Text, PostingListWritable>
121
        private PostingListWritable postingList = new PostingListWritable();
122
        protected boolean isLastWordInUniqueDoc = false;
123
124
        public void reduce (Text word, Iterable < PostingList Writable > postingLists
125
                            Context context
126
                            ) throws IOException, InterruptedException {
127
          // We are going to aggregate all posting lists together
128
          postingList.clear();
          // Iterate through all posting lists comming from the mappers or the
129
              combiners
130
          for (PostingListWritable mw : postingLists) {
             // Iterate through the entries of one given posting list
131
132
             for (PostingListWritable.Entry<Writable, Writable> entry: mw.
                entrySet()) {
133
               Text doc = (Text) entry.getKey();
               IntWritable newValue = (IntWritable) entry.getValue();
134
               if (postingList.containsKey(doc)) {
135
                 IntWritable oldValue = (IntWritable) postingList.get(doc);
136
                 postingList.put(doc, new IntWritable(oldValue.get() + newValue.
137
                    get());
138
               } else {
                 postingList.put(doc, new IntWritable(newValue.get()));
139
140
            }
141
          }
142
143
144
          isLastWordInUniqueDoc = (postingList.size() == 1);
          context.write(word, postingList);
145
146
        }
      }
147
```

• Reducer:

The reducer shares the reduce function (it is a child of the combiner) with the combiner with the only exception that it will increment the counter for words in unique documents (which should only occur once per key and therefore only in the reducer).

```
150
      public static class PostingListReducer
           extends PostingListCombiner {
151
152
        // The only thing different between the reducer and the combiner is the
        // counter incrementation. It must be incremented only once per key,
153
            hence in
        // the reducer.
154
155
        @Override
        public void reduce (Text word, Iterable < PostingListWritable > postingLists
156
157
                             Context context
158
                            ) throws IOException, InterruptedException {
159
              super.reduce(word, postingLists, context);
              if (isLastWordInUniqueDoc) {
160
                //\ Increment\ counter\ for\ word\ appearing\ in\ a\ single\ document\ only
161
                // We are guaranteed that this is the only time that the counter
162
                // will be incremented for this word because all the values from
163
                // a given key all go to the same reduce call.
164
```

```
165 context.getCounter(WordCounter.WORDS_IN_UNIQUE_DOC).increment(1); 166 } 167 } 168 }
```

• Results: Here is an extract of the output inverted index csv:

```
enactment : pg3200.txt\#2
     enchant : pg100.txt#4, pg3200.txt#2
     enchanting: pg31100.txt#1, pg100.txt#6, pg3200.txt#58
1743
1744
     encircled: pg100.txt#1, pg3200.txt#5
     enclosed: pg31100.txt#9, pg100.txt#5, pg3200.txt#37
1745
     enclouded : pg100.txt\#1
1746
     encroach: pg31100.txt\#2, pg3200.txt\#2
1747
1748
     encyclopaedic: pg3200.txt#2
1749
     endeavored : pg3200.txt#12
     endeavors : pg31100.txt\#1, pg3200.txt\#10
1750
     endlich: pg3200.txt#1
1751
1752
     endue : pg100.txt#2
1753
     enforc : pg100.txt#15
     enforcing: pg31100.txt#2, pg3200.txt#1
1754
1755
     enfreedoming: pg100.txt#1
1756
     engineering: pg3200.txt#21
```

3 Conclusion

Inverted indexes are widely used for text retrieval or for search engines for the performance gain they provide. Creating an inverted index can however be a computing power intensive task but it can be highly parallelized. Hadoop's MapReduce framework is a perfect candidate for parallelizing this kind of task on a highly distributed cluster with no centralized data.