Big Data Processing and Analytics: Assignment 1

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February 13, 2017

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:

 ${\rm Hadoop}\ 2.7.3$

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 2e4ce5f957ea4db193bce3734ff29ff4

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

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:

```
6 </configuration>
 • hdfs-site.xml:
  <configuration>
      cproperty>
3
          <name>dfs.replication
4
          <value>1</value>
5
      6 </configuration>
 • mapred-site.xml:
  <configuration>
      cproperty>
3
          <name>mapreduce.framework.name</name>
4
          <value>yarn
5
      </configuration>
 • yarn-site.xml:
  <configuration>
1
2
      cproperty>
3
          <name>yarn.nodemanager.aux-services</name>
4
          <value>mapreduce_shuffle/value>
5
      6
7
8
      To prevent unhealthy nodes and jobs hanging due to low disk space
      9
         unhealthy-node-local-dirs-and-log-dirs-are-bad
10
11
      cproperty>
12
          <name>yarn.nodemanager.disk-health-checker.max-disk-utilization-per-
             disk-percentage</name>
13
          <value>98.5
14
      15
  </configuration>
 • Commands to set up HDFS and YARN:
1
        # Format the filesystem
2
        hdfs namenode -format
        # Start HDFS
3
4
        start-dfs.sh
5
6
        # Create directories to execute MapReduce jobs
7
        hdfs dfs -mkdir /user
8
        hdfs dfs -mkdir /user/louis
9
10
        # Put the data in HDFS
        hdfs dfs -put \sim /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 propeps-3.3.10 from source to solve the problem as indicated in launchpad but to no avail.

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

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.

```
public static class TokenizerMapper
18
19
          extends Mapper<Object, Text, Text, IntWritable>{
20
       private final static IntWritable one = new IntWritable(1);
21
22
       private Text word = new Text();
23
       public void map(Object key, Text value, Context context
24
25
                        ) throws IOException, InterruptedException {
26
          // Splits a string to tokens (here words)
27
         StringTokenizer itr = new StringTokenizer(value.toString(), "
             .,?!\"'()[]$*-_;:|");
28
         while (itr.hasMoreTokens()) {
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 occurences of each key writes it to the output file only if its count is greater than 4000.

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

```
public static void main(String[] args) throws Exception {
Configuration conf = new Configuration();

// Remove output folder if it exists
Path output = new Path(args[1]);
FileSystem hdfs = FileSystem.get(conf);
// delete existing directory
if (hdfs.exists(output)) {
```

```
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
68
        if ((args.length >= 5) \&\& (Integer.parseInt(args[4]) == 1)) {
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
        if ((args.length >= 4) \&\& (Integer.parseInt(args[3]) == 1)) {
79
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);
87
        FileInputFormat.addInputPath(job, new Path(args[0]));
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);
          duration = 10000000000;
93
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.

```
\#!/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
7
   hadoop jar StopWords.jar StopWords data/corpus/ out 10 0 0;
   hdfs dfs -getmerge out stopwords.csv;
9
10
11
  # 10 reducers, combiner
```

```
12 hadoop jar StopWords.jar StopWords data/corpus/ out 10 1 0;
13 hdfs dfs -getmerge out stopwords.csv;
14
15 # 10 reducers, combiner, compression
16 hadoop jar StopWords.jar StopWords data/corpus/ out 10 1 1;
17 hdfs dfs -getmerge out stopwords.csv;
18
19 # 50 reducers, combiner, compression
20 hadoop jar StopWords.jar StopWords data/corpus/ out 50 1 1;
21 hdfs dfs -getmerge out stopwords.csv;
```

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

```
10
    with, 35591
    good, 4632
12
   from, 9677
    has, 5190
13
14
    its, 4569
15
    man, 5267
    made, 4046
16
    \mathrm{not}\;,\;\;35143
17
    said, 8434
   have, 24625
19
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...



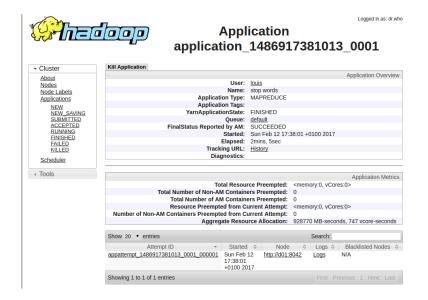
iii. (5) Run the same program again, this time compressing the intermediate results of map (using any codec you wish). Report the execution time. Is there any difference in the execution, time compared to the previous execution? Why?

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.

```
public static HashSet<String> readStopWords() {
52
53
          // Read stopwords into a HashSet for fast membership testing
54
          // The hashtable underlying structure provides O(1) membership testing
55
         HashSet<String> stopWords = new HashSet<String>();
56
          // Read csv file: inspired from https://www.mkyong.com/java/how-to-
57
             read-and-parse-csv-file-in-java/
          String csvFile = "stopwords.csv";
58
59
          BufferedReader br = null;
          String line = "";
60
          String cvsSplitBy = ",";
61
62
63
          \mathbf{try}
64
              br = new BufferedReader(new FileReader(csvFile));
65
              while ((line = br.readLine()) != null) {
                  // use comma as separator
66
67
                  String [] splittedLine = line.split(cvsSplitBy);
68
                  stopWords.add(splittedLine[0]);
69
70
          } catch (FileNotFoundException e) {
              e.printStackTrace();
71
72
          } catch (IOException e) {
73
              e.printStackTrace();
74
          } finally {
75
              if (br != null) {
76
                  try {
77
                      br.close();
                  } catch (IOException e) {
78
79
                      e.printStackTrace();
80
              }
81
```

• 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 >{
         \mathbf{private} \ \mathrm{Text} \ \mathrm{word} = \mathbf{new} \ \mathrm{Text}();
 89
 90
         private Text doc = new Text();
 91
         private IntWritable one = new IntWritable(1);
 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 {
           FileSplit fileSplit = (FileSplit) context.getInputSplit();
 97
 98
           String filename = fileSplit.getPath().getName();
99
           doc.set (filename);
100
101
           // Splits a string to tokens (here words)
           StringTokenizer itr = new StringTokenizer(value.toString(), "
102
               .,?!\"'()[]$*-_;:|");
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();
               postingList.put(doc, one);
110
               // Write one (key, value) pair to context
111
               context.write(word, postingList);
112
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).

```
public void reduce(Text word, Iterable < PostingListWritable > postingLists

Context context

125

throws IOException, InterruptedException {

// We are going to aggregate all posting lists together

postingList.clear();

// Iterate through all posting lists comming from the mappers or the combiners

for (PostingListWritable mw : postingLists) {
```

```
// Iterate through the entries of one given posting list
131
132
             for (PostingListWritable.Entry<Writable, Writable> entry: mw.
                entrySet()) {
              Text doc = (Text) entry.getKey();
133
              IntWritable newValue = (IntWritable) entry.getValue();
134
               if (postingList.containsKey(doc)) {
135
136
                 IntWritable oldValue = (IntWritable) postingList.get(doc);
137
                 postingList.put(doc, new IntWritable(oldValue.get() + newValue.
                    get());
138
                else {
139
                 postingList.put(doc, new IntWritable(newValue.get()));
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.

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

```
31 public static enum WordCounter {
32 WORDS_IN_UNIQUE_DOC
33 };
```

Which is incremented in the reducer's reduce function:

```
if (isLastWordInUniqueDoc) {

// Increment counter for word appearing in a single document only

// We are guaranteed that this is the only time that the counter

// will be incremented for this word because all the values from

// a given key all go to the same reduce call.

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);
```

227 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.

• 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.

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

```
119
      public static class PostingListCombiner
120
           extends Reducer<Text, PostingListWritable, Text, PostingListWritable>
121
        private PostingListWritable postingList = new PostingListWritable();
122
        protected boolean isLastWordInUniqueDoc = false;
123
        public void reduce (Text word, Iterable < Posting List Writable > posting Lists
124
125
                            Context context
126
                            ) throws IOException, InterruptedException {
127
          // We are going to aggregate all posting lists together
          postingList.clear();
128
129
          // Iterate through all posting lists comming from the mappers or the
              combiners
          for (PostingListWritable mw : postingLists) {
130
131
            // Iterate through the entries of one given posting list
132
            for (PostingListWritable.Entry<Writable, Writable> entry: mw.
                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());
              } else {
138
                 postingList.put(doc, new IntWritable(newValue.get()));
139
```

• 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
151
            extends PostingListCombiner {
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 < Posting List Writable > posting Lists
156
157
                            Context context
                            ) throws IOException, InterruptedException {
158
159
              super.reduce(word, postingLists, context);
              if (isLastWordInUniqueDoc) {
160
161
                // Increment counter for word appearing in a single document only
                // 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
                context.getCounter(WordCounter.WORDS_IN_UNIQUE_DOC).increment(1);
165
166
           }
167
168
         }
```

 \bullet $\mathbf{Results}:$ Here is an extract of the output inverted index csv:

```
enactment: pg3200.txt#2
1741
1742
     enchant : pg100.txt\#4, pg3200.txt\#2
1743
     enchanting : pg31100.txt\#1, pg100.txt\#6, pg3200.txt\#58
     encircled : pg100.txt#1, pg3200.txt#5
1744
1745
     enclosed: pg31100.txt#9, pg100.txt#5, pg3200.txt#37
     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
1751
     endlich : pg3200.txt#1
     endue : pg100.txt#2
1752
1753
     enforc: pg100.txt#15
1754
     enforcing : pg31100.txt#2, pg3200.txt#1
1755
     enfreedoming: pg100.txt#1
     engineering: pg3200.txt#21
1756
```

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.