



ILLINOIS INSTITUTE OF TECHNOLOGY

CLOUD COMPUTING  
CS 553

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Programming Assignment2 Part1 Report

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## 1. Introduction

As the sheer amount of data available in the world kept increasing by the day, we realized we needed a way to categorize, filter and sort it in a reliable and organized fashion. Another thing to note is that we also noticed the increasing need for large-scale storage, i.e., we'd reached a point where we just couldn't fit the entire dataset into memory. Thus, a need to devise an algorithm to store data onto high capacity disks and read from them whenever necessary. This forms the basis for **External Sorting**. To study more about big data processing as a part of cloud computing, we shall be implementing things in this programming assignment:

- External Merge Sort

In this assignment we are calculating the time taken for each sort to execute (excluding any extra time for the input to be loaded) and compare the results. In order to utilize all the available resource efficiently we are using multithreading. This assignment implementation of 2, 4 and 8 threads is done and corresponding processing time is calculated. The datasets used are of size **2 GB and 20 GB** we shall sort the data using external merge sort, which breaks it down into multiple chunks, sorts them separately by reading their respective portion(files) from disk and combines them into a single sorted output file

The program is executed on Neutron cluster with IP 216.47.142.38. The configuration for this cluster are given below:

No of CPU: 8

Core per Socket: 1

RAM: 2.09 GHz

## 2. External Merge Sort

### 2.1 Description

Simply put, external sorts are used when the data to be sorted does not fit into the main memory of the computer. It must be written to an external disk, broken down into chunks and loaded into memory whenever the data is needed. To attain maximum efficiency, the I/O operations as well as the sort mechanism itself need to be parallelized. The time taken by the external merge to completely and successfully sort the data is recorded and stored to be compared with the results of the Hadoop experiment.

### 2.2 Implementation

For implementation external sort, some data is passed to determine few of the parameter required for efficient performance and resource utilization. Argument 1 determine the file to be used for sorting, Argument 2 determine the chunks size in terms of records to be loaded in the RAM for sorting. Argument 3 determine the number of thread working concurrently in this chunk to get the sorted output. We are using the split command to partition the files. The split command is executed first on every node and included in the corresponding slurm files.

```
String sortFileSize=args[0];
int noOfThreads=Integer.parseInt(args[1]); //This represent the Number of thread to process each chunks
long maxRecordCount=Long.parseLong(args[2]); //This represents the maximum records of each chunk. Size of chunk=maxRecordCount*100
String filePath=null;
```

The Algorithm implementation is determined as follows:

1. Determine the number of chunks to sort the file

```
//System.out.println("Sorting of file size maximum of record file :"+maxRecordSize);
long size = 100;
File file = new File(filePath);
long fileSize=file.length();
long recordsCount = (fileSize/size);
System.out.println("Sorting of file size :"+sortFileSize);
logger.info("Sorting of file size :"+sortFileSize);
System.out.println("Total number of records(lines) in the file:"+recordsCount);
logger.info("Total number of records(lines) in the file:"+recordsCount);
//long maxRecordCount = 2500000 ;
long numberOfChunks = recordsCount/maxRecordCount;
```

2. Determine the number of files required to sort. (This is calculated by number of chunks \* number of threads)

```
System.out.println("Number of files = numberOfChunks * number of threads per each chunk");
logger.info("Number of files = numberOfChunks * number of threads per each chunk");
long numberOfFiles = (numberOfChunks * noOfThreads);
```

3. For each chunk, read the corresponding file and sort it and write to a temporary file. This is done by sorting each chunk parallelly.

The Logic implemented for sorting is as follows:

The actual “data-2GB” and “data-20GB” consists of records on each line. Each records corresponds to 100 bytes. Each records consists of key(unique and duplicate), ascii value in hexadecimal of the representing actual record position in the file. The aim was to sort the gensort input by the keys. We have overridden the comparator method for sorting the records. The First 10 byte of each records is the keys of the records and we have used it for sorting. Refer the below screenshot for reference.

```

public void run(){
    try {
        ArrayList<String> data = new ArrayList();
        String line = null;
        LineNumberReader reader;
        File inFile = new File(inputFile);
        long fileLength = (inFile.length())/100;
        reader = new LineNumberReader(new FileReader(inputFile));
        for(long i=0;i<fileLength;i++){
            line = reader.readLine();
            while(line == null && i < fileLength){
                line = reader.readLine();
            }
            data.add(line);
        }
        //System.out.println("Array size is " + data.size());
        reader.close();
        File f = new File(inputFile);
        f.delete();
        Collections.sort(data, new Comparator<String>() {
            public int compare(String str1, String str2){
                if(str1 == null || str2 == null){
                    return 0;
                }
                String substr1 = str1.substring(0,10);
                String substr2 = str2.substring(0,10);
                return substr1.compareTo(substr2);
            }
        });
        File tmpfile = new File(outputFile);
        FileWriter fileWriter = new FileWriter(tmpfile);
    }
}

```

We have implemented CountDownLatch instead of join to perform the fork operation. CountDownLatch acts like a synchronization aid that allows one or more threads to wait until a set of operations being performed in other threads completes. CountDownLatch initialze count is set to the number of threads for each chunk being processed. For each file succesfully sorted in the chunk the CountDownLatch is decreased ie. Thread has completed its exc. The awaits method will block until the current count reach to zero ie. All the files in the chunks are completely executed by the threads.

```

for(int i=0;i<numberOfChunks;i++){

    final CountDownLatch executionCompleted = new CountDownLatch(noOfThreads);
    //System.out.println("Sorting threads started");
    sortMultiThread threadSort[] = new sortMultiThread[noOfThreads];
    String tmpSortFilePath = "";
    for(int j=0;j<noOfThreads;j++){
        System.out.println("Sorting of files initialized!");
        logger.info("Sorting of files initialized!");
        if(tempIndex<10)
            tmpSortFilePath = "tmpFile0"+tempIndex;
        else
            tmpSortFilePath = "tmpFile" + tempIndex;
        threadSort[j] = new sortMultiThread(tmpSortFilePath,tmpSortFilePath+"sort",executionCompleted);
        threadSort[j].start();
        tempIndex = tempIndex + 1;
    }
}

```

4. Read all the sorted chunks file and perform K way merge to merge the sorted file. Each item(records) from the sorted run and pick the minimum and store it an output file. This is implemented by using Priority Queue as shown below.

```

public static void performKWayMergeSort(List<File> sortedFiles, File outputFile, int numberOfFiles) {
    PriorityQueue<DataBuffer> priorityQueue = new PriorityQueue<DataBuffer>((numberOfFiles + 1),
        new Comparator<DataBuffer>() {
            public int compare(DataBuffer data1, DataBuffer data2) {
                String sub1 = data1.getBufferString();
                String sub2 = data2.getBufferString();
                if (sub1 == null || sub2 == null) {
                    return 0;
                } else {
                    sub1 = sub1.substring(0,10);
                    sub2 = sub2.substring(0,10);
                    return sub1.compareTo(sub2);
                }
            }
        });

    try {
        for (File eachFile : sortedFiles) {
            DataBuffer fileBuffer = new DataBuffer(eachFile);
            priorityQueue.add(fileBuffer);
        }

        FileWriter fileWriter = new FileWriter(outputFile);
        while (priorityQueue.size() > 0) {
            DataBuffer fileBuffer = priorityQueue.poll();

```

### 3. Throughput

The number of I/O throughput is calculated by the number of reads and write operation performed to get the sorted output.

$$\text{Original file Size} + \text{file Size} * \text{Read} + \text{file Size} * \text{Write}$$

### 4. Results

The below mentioned Shared Memory Results are obtained as per Neutron cluster

Experiment	Shared Memory (1VM 2GB) 4 Threads	Linux Sort (1VM 2GB)	Shared Memory (1VM 20GB) 4 Threads	Linux Sort (1VM 20GB)
Compute Time (sec)	67	21.40	820	394.21
Data Read (GB)	6	2	60	20
Data Write (GB)	6	2	60	20
I/O Throughput (MB/sec)	59.70	191.40	48.78	103.40

Table 4.1 Performance Evaluation of Shared Memory Sort (using 4 Threads)

Experiment	Shared Memory (1VM 2GB) 2 Threads	Shared Memory (1VM 2GB) 8 Threads
Compute Time (sec)	86	68
Data Read (GB)	6	6
Data Write (GB)	6	6
I/O Throughput (MB/sec)	46.51	58.82

Table 4.1 Performance Evaluation of 2GB Shared Memory Sort

### 5. Analysis

For 2 GB input: From the result table we can conclude that execution time decreases as the number of threads operating on each chunk increases. Similarly, throughput increases as the number of threads increases. From the above analysis we can conclude that MySort works efficiently and follows real world scenarios