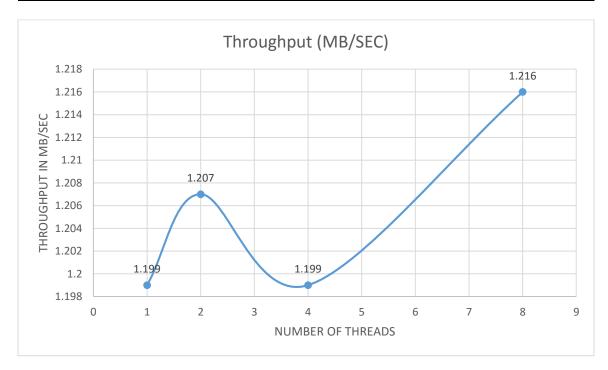
## 1) SHARED MEMORY SORT

10GB SORT - C3.Large

## Sort time elapsed in minutes

| 1 thread | 2 thread  | 4 thread  | 8 thread |
|----------|-----------|-----------|----------|
| 8340secs | 8280 secs | 8340 secs | 8220     |

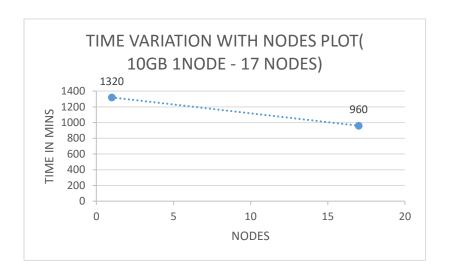
| 1 thread(MB/SEC) | 2 thread(MB/Sec) | 4 thread(MB/Sec) | 8 thread(MB/SEC) |
|------------------|------------------|------------------|------------------|
| 1.199            | 1.207            | 1.199            | 1.216            |



The maximum throughput is obtained in the case of 8 threads performing the shared memory sort. The file splitting time is constant in all the four scenarios and the difference is contributed by the variance in the sort time with multithreads.

## 2) Hadoop

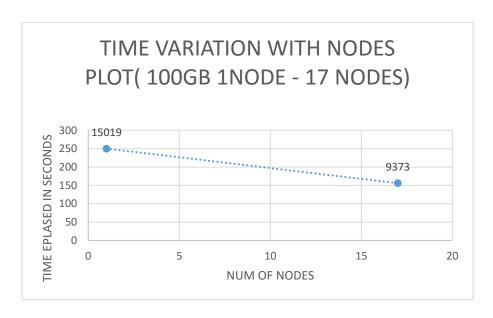
| Size | Number of nodes | Time elapsed(SECONDS) | Throughput(MB/SEC) |
|------|-----------------|-----------------------|--------------------|
| 10GB | 1               | 1320                  | 7.57575            |
| 10GB | 17              | 960                   | 10.41666           |



A) 100GB – 1 node - 4 hours 10mins 19 secs

100GB - 17 nodes - 2 hours 36 mins 13 sec

| Size   | Number of nodes | Time elapsed IN SECS | Throughput(MB/SEC) |
|--------|-----------------|----------------------|--------------------|
| 100 GB | 1               | 15019                | 6.658              |
| 100GB  | 17              | 9373                 | 10.668942          |



From the above two charts it could be found that with the number of nodes increase there is a considerable increase in performance

**HADOOP RESULTS** 

| Size  | Number of nodes | Time elapsed IN SECS | Throughput(MB/SEC) |
|-------|-----------------|----------------------|--------------------|
| 10GB  | 1               | 1320                 | 7.57575            |
| 100GB | 17              | 9373                 | 10.668942          |

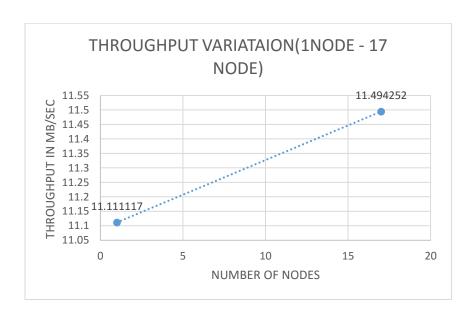
### **SPEED UP CALCULATION HADOOP**

Using the Throughput to calculate the baseline time and then speed up calculation.

- 1) Speed up Hadoop = Time elapsed in single node /Time elapsed in 16 nodes 10.668942/7.57575 = 1.408
- 2) Speed up Shared Memory = Time elapsed in shared memory sort / Time elapsed 16 nodes **10.668/1.199= 8.8974**

#### 3) Spark

| Size  | Number of nodes | Time elapsed(SECONDS) | Throughput(MB/Sec) |
|-------|-----------------|-----------------------|--------------------|
| 10GB  | 1               | 900                   | 11.11111 MB/SEC    |
| 100GB | 17              | 8700                  | 11.494252 MB/SEC   |



### **Speed Up Calculation for Spark:**

Since we are using different data sets in 1 node and 16 node set up , using the Throughput to calculate the baseline time and the speed up calculation . From throughput the

1) Speed up spark = Time elapsed in single node /Time elapsed in 16 nodes 11.494/11.111 = 1.03

2) Speed up Shared Memory = Time elapsed in shared memory sort / Time elapsed 16 nodes 11.494/1.199= 9.58

### **Speed up comparison Hadoop and Spark**

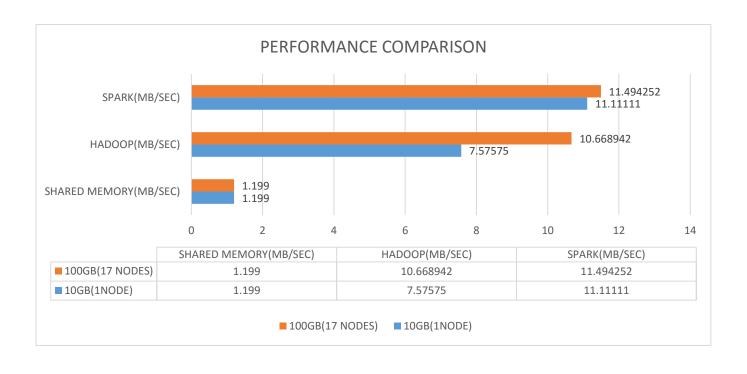
Speed up Hadoop = 1.404;

Speed up Spark = 1.03;

Speed up of Hadoop is better compared to the speed up of Spark

From the speed up shared memory values for Hadoop and spark, it could be inferred that the performance of Shared memory sort is weak compared to Hadoop and spark

# 3) Shared Memory / Hadoop /Spark – Performance comparison



## **APACHE SPARK COMPARISON**

| Tera Sort Competition                      | Spark sort implementation  |
|--|--|
| 100 TB in 1,406 seconds                    | 100 GB in 8700 seconds > 100TB should be done in approximately 8700000 seconds |
| 207 Amazon EC2 i2.8xlarge nodes x          | 17 Amazon instances C3.large   |
| 32 vCores - 2.5Ghz , Intel Xeon E5-2670 v2 | 2 vCores - 2.8Ghz, Intel Xeon E5-2680 v2                                       |
| 244GB memory                               | 3.75GB Memory  |
| 8x800 GB SSD                               | 2 x 16 SSD   |

### **HADOOP COMPARISON**

| Hadoop Competition            | Hadoop implementation   |
|-------------------------------|---|
| 102.5 TB in 4,328 seconds     | 100 GB in 9373 seconds > 100 TB should be done in approximately 9373000 seconds |
| 2100 nodes x                  | 17 Amazon instances C3.large  |
| 2 2.3Ghz hexcore Xeon E5-2630 | 2 vCores - 2.8Ghz, Intel Xeon E5-2680 v2  |
| 64 GB memory,                 | 3.75GB Memory   |
| 12x3TB disks                  | 2 x 16 SSD  |

The number of threads, cores, available memory, hard drive capacity and most importantly an optimized code could help achieve better results. Decision on the number of nodes proportionately with the size of data to be sorted is important in performance improvement.

### **Cloud Sort Bench Mark**

This paper tries to address the shortcomings of the existing sort benchmarks and proposes a total cost of ownership benchmark. For deciding on a particular cloud with various vendors available for public cloud, it becomes extremely important to have a benchmark on cost. External sorting becomes an ideal candidate for this as it is high IO intensive process which would cover network, storage, compute benchmarking. This would be a major step in deciding whether to go for a public cloud or a private cloud as well. This would definitely drive innovation on the public cloud sphere to provide best possible performance.

### **CONCLUSION**

What conclusions can you draw? Which seems to be best at 1 node scale? How about 16 nodes? Can you predict which would be best at 100 node scale? How about 1000 node scales?

Data points are collected for Shared Memory, Spark, and Hadoop for 1 node and 16 node scenarios. With the throughput calculated in MB/sec for the three systems, it is found that Spark has a better performance in both 1node and 16 node cases. As Spark performs better in 1 node and 16 node cases, it could be assumed for a 100 node scale as well Spark is going to perform better when compared to Shared Memory and Hadoop.