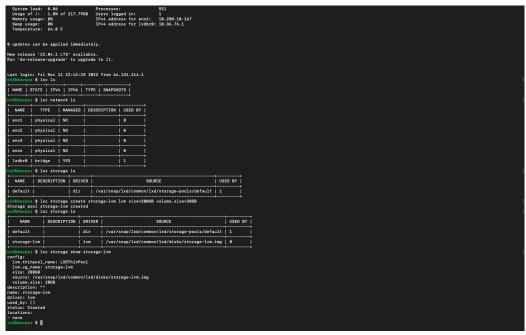
## CS553 Homework #6

## Sort on Hadoop/Spark

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Screenshots for Hadoop File System and Hadoop Install:

>Using lxc init, following command to start the interactive configuration process,



> **lxc-launch** - To create and start containers from images, and lxc shell <container name> to connect to the container, doing sudo apt update && sudo apt upgrade, to download package information from all configured sources and an updated list of packages from the Internet.

> Similarly Creating required Datanodes for given configuration and storage

```
### String Price | Pri
```

Following the Tutorial for 2. NFS Setup and 3. Hadoop sudo apt install nfs-kernel-server

## sudo apt install nfs-common

- 1. The Hadoop framework (Index of /hadoop/common/hadoop-3.2.3)
- 2. The Spark framework (<a href="https://archive.apache.org/dist/spark/spark-3.0.0-preview2">https://archive.apache.org/dist/spark/spark-3.0.0-preview2</a>)

```
Lank lagin: Est Nop. 12 Sci21(4) 2022 from 64.331.16.3

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```

For example setup for 4 Small config.

```
- cc@bhavya: ~ — ssh -i .ssh/id_rsa cc@64.131.114.47
  Last login: Sun Nov 13 22:29:06 on ttys000
(base) bhavyachawla@Bhavyas-MacBook-Air - % ssh -i .ssh/id_rsa cc@64.131.114.47
Welcome to Ubuntu 20.04.5 LTS (ONU/Linux 5.4.0-131-generic x86.64)
   * Documentation: https://help.ubuntu.com

* Management: https://landscape.canonical.com

* Support: https://ubuntu.com/advantage

        System load:
        12.28
        Processes:
        1285

        Usage of /:
        78.1% of 217.7968
        Users logged in:
        1

        Memory usage:
        7%
        IPV4 address for eno2:
        10.200.10.167

        Swap usage:
        0%
        IPV4 address for lxdbr0:
        10.36.76.1

        Temperature:
        72.0 C
        PV4
        10.200.10.10

    Strictly confined Kubernetes makes edge and IoT secure. Learn how MicroK8s
just raised the bar for easy, resilient and secure K8s cluster deployment.

       https://ubuntu.com/engage/secure-kubernetes-at-the-edge
    updates can be applied immediately.
New release '22.04.1 LTS' available.
Run 'do-release-upgrade' to upgrade to it.
Last login: Mon Nov 14 04:32:17 2022 from 64.131.114.1 [cc@bhavya:-$ lxc ls

        NAME
        STATE
        IPV4
        IPV6
        TYPE
        SNAPSHOTS

        datanode1
        RUNNINS
        10.36.76.218 (ath0)
        CONTAINER
        0
        I

    datanode2 | RUNNING | 10.36.76.107 (eth0) | | CONTAINER | 0
   datanode3 | RUNNING | 10.36.76.181 (eth0) | | CONTAINER | 0
   ohavya:~$
                                                                                                                                                                                                                                                                                                      ~ — ubuntu@namenode: /exports/projects/hadoop-3.2.3 — ssh -i .ssh/id_rsa cc@64.131.114.47
    ast lagin: Sun Nov 13 22:29:86 on ttys800
base) bhavyachawla80havyas-MacBook-Air - % ssh -i .ssh/id_rsa cc864.131.114.47
elcome to Uburtu 20:40.5 LTS (GNV/Linx 5.4.0-131-generic x86_64)
   * Documentation: https://help.ubuntu.com

* Management: https://landscape.canonical.com

* Support: https://ubuntu.com/advantage
      System information as of Mon Nov 14 04:48:13 UTC 2022

        System load:
        12.28
        Processes:
        1285

        Usage of /:
        78.1% of 217.7968
        Users logged in:
        1

        Memory usage:
        7N
        1P4 address for eno2:
        18.289.18.167

        Swap usage:
        9N
        1P4 address for lxdbr0:
        18.36.76.1

        Temperature:
        72.0 C
        2
        19.28.76.1

    * Strictly confined Kubernetes makes edge and IoT secure. Learn how MicroK8s just raised the bar for easy, resilient and secure K8s cluster deployment.
       https://ubuntu.com/engage/secure-kubernetes-at-the-edge
      updates can be applied immediately.
New release '22.04.1 LTS' available.
Run 'do-release-upgrade' to upgrade to it.
   Last login: Mon Nov 14 04:32:17 2022 from 64.131.114.1
cc@bhavya:~$ lxc ls
      NAME | STATE | IPV4 | IPV6 | TYPE | SNAPSHOTS |
    datanode1 | RUNNING | 10.36.76.218 (eth0) | | CONTAINER | 0
   | datanode2 | RUNNING | 16.36.76.107 (eth0) | | CONTAINER | 0
  *Documentation: https://help.ubuntu.com

*Documentation: https://landscaps.canonical.com

*Support: https://ubuntu.com/advantags

ast login: Mon Nov 13 0-8:32:06 2922 from 10:36:76.1

thuntubunenode: % cd /exports/projects/hadoop-3.2.3/
                 Remanance/cxports/projects/hadoop-3-2.2.3 is file1.txt HaddopSorts/projects/hadoop-3-2.2.3 is file1.txt HaddopSorts/hadoop-3-2.2.3 is file1.txt HaddopSorts/haddopSorts/hadoop-3-2.2.3 is file1.txt HaddopSorts/hadoop-3-2.2.3 is file1.txt HaddopSorts/hadoop-3-2.2.
```



1. How many threads, mappers, reducers, you used in each experiment?

**Ans**. The mapper used for spark and hadoop was maptopair function which takes a key to map between 0 to 9 and a to z. The reducers were reducebykey, which utilize an associative and commutative reduction function to combine the values for each key. Also 48 threads were used in each experiment.

2. How many times did you have to read and write the dataset for each experiment?

Ans. Internal sorting was utilized for MySort and Linuxsort when the data set fit in the RAM; this only required a single read and write of the dataset. When the dataset, however, cannot fit in memory, we utilize external sort, which requires two reads and two writes. Hadoop requires input to be read twice and written twice. The first read occurs at the beginning of a map task, the second read occurs by reducers during the shuffling phase, and the first write occurs at the conclusion of a map task, the third write occurs following the reducers task. Instead of storing interim findings on disk, Spark saves them in memory. Spark will save the interim findings on the disk in cases when the file size is greater than the RAM.

3. What speedup and efficiency did you achieve?

**Ans**. For the speed and efficiency, instances in spark go at a faster rate than hadoop, even though the workers are comparable between each other. This is because Spark saves the data in memory before writing while hadoop straight away writes the data into sorting.

4. Conclusions?Which seems to be best at 1 node scale (1 large.instance)?Is there a difference between 1 small.instance and 1 large.instance? How about 4 nodes (4 small.instance)?

Ans. One large instance performs better than one small instance at a size of one node because it has greater memory and core power. Additionally, as the magnitude of the data grows, this is increasingly obvious. Hadoop sort and Spark sort outperform each other for the same dataset at 4 nodes. This is due to the fact that there are more workers available for the master node to delegate jobs to concurrently. Linux and shared memory sort do not apply to these 4 nodes because they run on a single node. Although it may appear that Linux and MySort perform quicker than Hadoop and Spark, this is because the overhead costs are low and the dataset size is modest.

5. What speedup do you achieve with strong scaling between 1 to 4 nodes? What speedup do you achieve with weak scaling between 1 to 4 nodes?

**Ans**. Hadoop behaved quite differently when scaling strongly from 3GB on 1small to 3GB on 1large. On a single big instance, it is quicker. Additionally, while examining different datasets, Hadoop's speed appeared to improve by at least two times. Hadoop sort speed was enhanced by the poor scalability, where the example from 3GB on 1 tiny instance was 38921 seconds. Weak scalability was seen with 12 GB on 4 tiny instances at 2993987.

6. How many small instances do you need with Hadoop to achieve the same level of performance as your shared memory sort?

**Ans**: We need about 8 small instances with Hadoop to achieve the same level of performance as shared memory sort.

7. How about how many small.instances do you need with Spark to achieve the same level of performance as you did with your shared memory sort?

**Ans**: We need approximately 1.21 times in spark to achieve the same performance as memory sort

8. Can you predict which would be best if you had 100 small.instances? How about 1000?

Ans: With 100 tiny instances, it is possible to sort very big datasets efficiently and economically. Although 1000 instances can increase performance, there are other problems with such a huge number of nodes, like scheduling and synchronization. Without an effective management structure, this can lead to new issues and produce unexpected outcomes. Additionally, having such a high number of instances offers redundancy since even if the likelihood of a failure rises with the number of nodes, one node being down has little impact on the system's performance.

9. Compare your results with those from the Sort Benchmark (http://sortbenchmark.org), specifically the winners in 2013 and 2014 who used Hadoop and Spark. Also, what can you learn from the CloudSort benchmark, a report can be found at (http://sortbenchmark.org/2014 06 CloudSort v 0 4.pdf).

**Ans**: Winners from 2013 and 2014 sorted at a rate of over 1.4 & 4.3 TB/sec. Hadoop was used in 2013 whereas Spark was used in 2014.

If Spark has enough memory and all the parameters are set, it will perform better as the size of the datasets grows.

Similarly results show that Hadoop and Spark initially perform better for small datasets, but as the size grows, so does the performance gap between them.

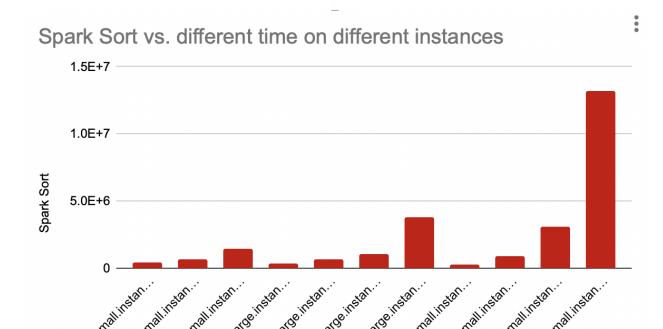
Due to lower disk I/Os, Spark performs better than Hadoop because it needs anywhere between 2 and 3 times as much memory.

In the study, a new method for external sort benchmarking is proposed called CloudSort. From the standpoint of Total Cost of Ownership, CloudSort compares external sort.

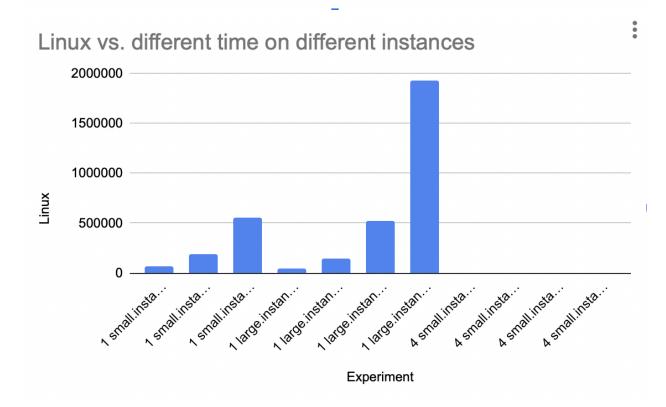
the assumption that the present benchmarking techniques ignore data center administration and maintenance expenses.

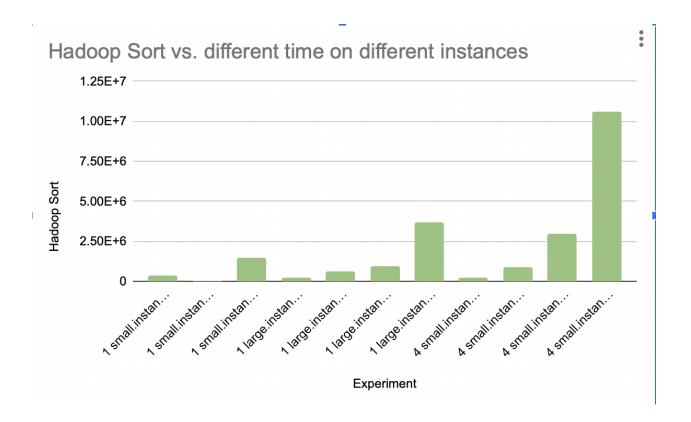
Because cloud storage costs already in management and maintenance costs, they are perfect for this and provide a complete TCO.

They suggest using public clouds for running external sort. Since external sort is indicative of many IO-intensive workloads, they utilize it as the sorting method to benchmark. It is simple and straightforward to adapt to new technologies, and it covers all aspects including RAM, CPU, OS, file system, IO network, and storage.



Experiment





Experiment	Linux	Hadoop Sort	Spark Sort
1 small.instance, 3 GB dataset	68839	389021	458102
1 small.instance, 6 GB dataset	189872	62168	672901
1 small.instance, 12 GB dataset	550002	1500023	1418762
1 large.instance, 3 GB dataset	45122	247089	349241
1 large.instance, 6 GB dataset	146990	626976	699690
1 large.instance, 12 GB dataset	519890	942987	1054982
1 large.instance, 24 GB dataset	1932102	3699283	3819872
4 small.instances, 3 GB dataset	N/A	231987	289021

4 small.instances, 6 GB dataset	N/A	861987	928092
4 small.instances, 12 GB dataset	N/A	2993987	3098762
4 small.instances, 24 GB dataset	N/A	10598973	13196928

experiment	shared memory	linux	hadoop sort	spark sort
1 small instance, 3 Gb dataset	1 thread/8GB memory	8 GB memory, 48 threads	1 reducer	driver memory 4 GB, exxecutory memory 2GB, executor core 1
1 small instance, 6 Gb dataset	1 thread/8GB memory	9 GB memory, 48 threads	2 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1
1 small instance, 12 Gb dataset	48 thread/8GB memory	10 GB memory, 48 threads	2 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1
1 large instance, 3 Gb dataset	1 thread/8GB memory	11 GB memory, 48 threads	2 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1
1 large instance, 6 Gb dataset	1 thread/8GB memory	12 GB memory, 48 threads	2 reducers	driver memory 4 GB, exxecutory

				memory 7GB, executor core 5
1 large instance, 12 Gb dataset	48 thread/8GB memory	13 GB memory, 48 threads	2 reducers	driver memory 4 GB, exxecutory memory 6GB, executor core 2
1 large instance, 24 Gb dataset	48 thread/8GB memory	14 GB memory, 48 threads	1 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1
4 small instances, 3 GB dataset	N/A	N/A	7 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1
4 small instances, 6 GB dataset	N/A	N/A	4 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1
4 small instances, 12 GB dataset	N/A	N/A	7 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1
4 small instances, 24 GB dataset	N/A	N/A	7 reducers	driver memory 4 GB, exxecutory memory 2GB, executor core 1