# CS 553 Cloud Computing Programming Assignment 2

Boyang Li (A20314367) Xingtan Hu (A20304622) Zichen Zhang (A20307812)

## Team Member's Contributions

1. Virtual Cluster (1 node, 16 nodes): Boyang Li, Xingtan Hu

2. Shared-Memory WordCount : Boyang Li
3. Hadoop Configuration : Zichen Zhang
4. Hadoop WordCount : Zichen Zhang
5. Swift Configuration : Xingtan Hu

6. Swift WordCount : Boyang Li, Xingtan Hu

7. Performance Analysis: Boyang Li, Xingtan Hu, Zichen Zhang 8. Sort on Shared-Memory, Hadoop, Swift, and MPI: Boyang Li, Xingtan Hu, Zichen Zhang

9. Report: Xingtan Hu

# **Brief Description**

This programming assignment covers the WordCount application implemented in 3 different ways: Python, Hadoop, and Swift. We count the appearance of every words and count the appearance by using a 10 GB file. And the result txt is sorted by its appearance time.

## Methodology

For Shared-Memory WordCount, we use python and multi-thread technology to implement.

For Hadoop WordCount, we use Java based on a Hadoop framework and MapReduce.

For Swift WordCount, we use Swift language and MapReduce.

# Runtime Environment Settings

Amazon Web Services EC2

AMI - Ubuntu Server 14.04 LTS (HVM), SSD Volume Type - ami-3d50120d

Hadoop 1.2.1

Swift 0.95 RC6 Java 1.7.0 65

Shared-Memory:

EC2 instance type: c3-large

Hadoop part:

EC2 instance type: (master node)c3-large (slave node)m3-medium

Swift part:

EC2 instance type: (headnode)c3-large (worker node)m3-medium

MPI: Open MPI 3.0.4, starcluster

# Installation Steps of Virtual Cluster

- 1. 1 node
- 1) log in to the AWS, choose EC2
- 2) launch instance
- 3) choose Ubuntu Server 14.04 LTS (HVM), SSD Volume Type ami-3d50120d
- 4) choose t2.medium
- 5) choose spot instance
- 6) add storage to 12GB(enough for Hadoop and Swift WordCount)
- 7) create a new security group, with rules: SSH, All TCP(Anywhere), All ICMP(Anywhere)
- 2. 16 nodes
- 1) finish the step of 1 node
- 2) create an image of instance established in 1 node
- 3) create another 15 nodes use the AMI just created
- 4) use the same configuration

#### 3. difficulties

At first, we didn't notice that the convenient of AMI, we configure the same environment again and again for 16 nodes.

After we used the AMI, it saved a lot of times.

# **Hadoop Configuration**

Edit /etc/hosts file, put the following line in the file:

172.31.44.150 me1

172.31.44.149 me2

172.31.44.151 me3

172.31.44.197 me4

172.31.44.198 me5

172.31.44.208 me6

172.31.44.209 me7

172.31.44.206 me8

172.31.44.204 me9

172.31.44.199 me10

172.31.44.201 me11

172.31.44.205 me12

172.31.44.207 me13

172.31.44.200 me14

172.31.44.202 me15

172.31.44.203 me16

172.31.46.204 me17

Front is private IP, behind is node name.

For every mechine, edit ~/.ssh/config file, put the following line in the file:

Host me1

HostName 54.68.149.102

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me2

HostName 54.68.42.118

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me3

HostName 54.69.147.176

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me4

HostName 54.69.245.188

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me5

HostName 54.187.56.197

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me6

HostName 54.191.224.38

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me7

HostName 54.191.224.164

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me8

HostName 54.191.224.66

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me9

HostName 54.191.224.167

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me10

HostName 54.191.224.39

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me11

HostName 54.191.224.142

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me12

HostName 54.191.224.41

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me13

HostName 54.191.224.33

User ubuntu

IdentityFile ~/CS553-HW2.pem

Host me14

HostName 54.191.200.150
User ubuntu
IdentityFile ~/CS553-HW2.pem
Host me15
HostName 54.69.122.172
User ubuntu
IdentityFile ~/CS553-HW2.pem
Host me16
HostName 54.191.224.93
User ubuntu
IdentityFile ~/CS553-HW2.pem
Host me17
HostName 54.191.223.192
User ubuntu
IdentityFile ~/CS553-HW2.pem

HostName is the mechina public IP. IdentityFile is the key, here we use Amazon key pair.

After doing this, all mechines can ssh each other easily.

#### masters:

this file defines on which machines Hadoop will start secondary NameNodes in our multi-node cluster.

#### slaves:

this file lists the hosts, one per line, where the Hadoop slave daemons (DataNodes and TaskTrackers) will be run.

## core-site.xml

The name of the default file system. A URI whose scheme and authority determine the FileSystem implementation. The uri's scheme determines the config property (fs.SCHEME.impl) naming the FileSystem implementation class. The uri's authority is used to determine the host, port, etc. for a filesystem.

Add the following snippets between the <configuration> ... </configuration> tags in the respective configuration XML file:

mapred-site.xml

The host and port that the MapReduce job tracker runs at. If "local", then jobs are run in-process as a single map and reduce task.

hdfs-site.xml

Default block replication.

The actual number of replications can be specified when the file is created.

The default is used if replication is not specified in create time.

cproperty>

<name>dfs.replication</name>

<value>1</value>

<description>Default block replication.

The actual number of replications can be specified when the file is created.

The default is used if replication is not specified in create time.

</description>

From 1 to 16 nodes, copy all three files (core-site.xml, mapred-site.xml, hdfs-site.xml) in the file / hadoop-1.2.1/conf/ in all nodes.

For master node, put mechina name in the file (me1), this is the master node. (only need to modify master node)

For slave node, put the slave nodes that you want in the slaves file. If you want 4 slaves, then put four mechine names in the file. (only need to modify master node)

Formatting hdfs: ~/hadoop-1.2.1/bin/hadoop namenode -format

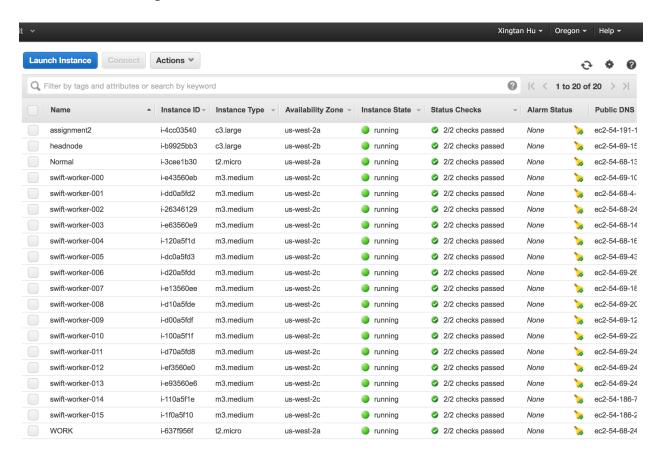
start cluster: ~/hadoop-1.2.1/bin/start-all.sh

create input file: bin/hadoop dfs -mkdir /zzc/input

copy dataset in the file: bin/hadoop dfs -copyFromLocal /home/ubuntu/tmp/wiki10gb /zzc/input

run wordcount: bin/hadoop jar wordcount.jar org.wordcount/ZzcWordCount /zzc/input /zzc/ output

# Swift Set&Configuration



ubuntu@ip-172-31-24-22:~/swift/run001\$ ls
cf run001.log swift.out wordcount-run001.d

You can find all the config files in all submitted file by using ReadMe.

# **MPI** Configuration

```
>>> Configuring SGE...
>>> Configuring NFS exports path(s):
/opt/sge6
>>> Mounting all NFS export path(s) on 1 worker node(s)
>>> Configuring cluster took 1.943 mins
>>> Starting cluster took 3.751 mins
The cluster is now ready to use. To login to the master node
as root, run:
    $ starcluster sshmaster mpicluster
If you're having issues with the cluster you can reboot the
instances and completely reconfigure the cluster from
scratch using:
    $ starcluster restart mpicluster
When you're finished using the cluster and wish to terminate
it and stop paying for service:
    $ starcluster terminate mpicluster
Alternatively, if the cluster uses EBS instances, you can use the 'stop' command to shutdown all nodes and put them
into a 'stopped' state preserving the EBS volumes backing
the nodes:
    $ starcluster stop mpicluster
WARNING: Any data stored in ephemeral storage (usually /mnt)
will be lost!
You can activate a 'stopped' cluster by passing the -x
option to the 'start' command:
    $ starcluster start -x mpicluster
This will start all 'stopped' nodes and reconfigure the
cluster.
ubuntu@ip-172-31-40-35:~$
```

# Data Analysis

## 1. Performance

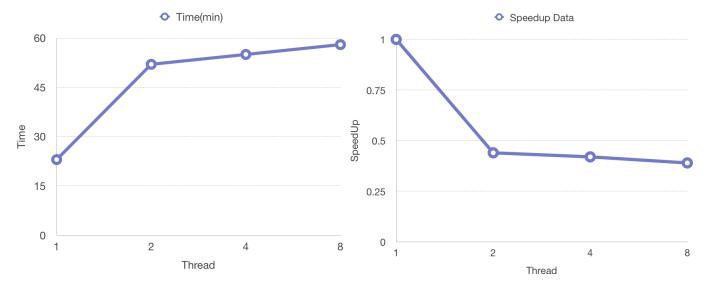
## 1) Shared-Memory WordCount Performance

**Shared-Memory WordCount Performance** 

Shared-Memory SpeedUp Data

THREAD NO.	TIME(MIN)	
1	23	
2	52	
4	55	
8	58	

THREAD NO.	SPEEDUP	
1	1	
2	0.44	
4	0.42	
8	0.39	



The speedup of shared-memory wordcount is less than 1, which means with more than 1 thread, the program execute slower. The execution-time graph reveals the same thing as speedup graph, that is more threads running concurrently will delay the execution time. The reason is that all threads have to read the file from the disk, which will cause a competition situation, so all these I/O operation will definitely delay the execution time.

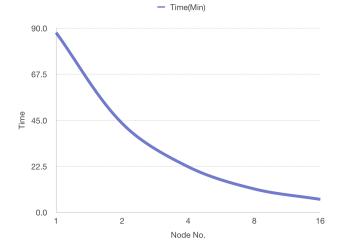
## 2) Hadoop WordCount Performance

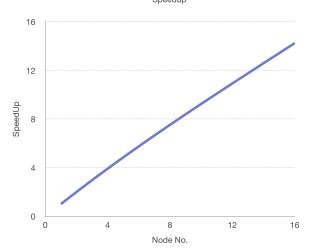
## Hadoop WordCount Performance

NODE NO.	TIME		
1	87 min 50 sec		
2	43 min 26 sec		
4	22 min 9 sec		
8	11 min 32 sec		
16	6 min 10 sec		

## Hadoop WordCount SpeedUp Data

Node No.	Speedup		
1	1		
2	2.01		
4	3.97		
8	7.62		
16	14.24		
<ul> <li>Speedup</li> </ul>			





According to the execution-time graph and SpeedUp graph, the Hadoop wordcount program is quite obeying the practical phenomenon. When we run the Hadoop wordcount with 1 or 2 threads, the execution time is quite large. Hadoop needs to set up and configure many things but this level of concurrency is not able to make up the time consumption. However, when number of threads become larger, the execution time becomes much less because high level of concurrency.

## 3) Swift WordCount Performance

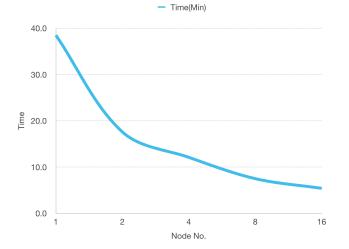
16

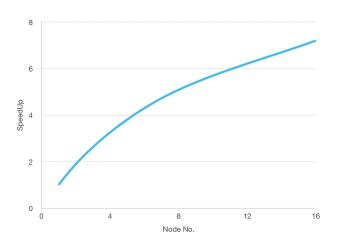
### **Swift WordCount Performance**

NODE NO.	TIME	
1	38 min 24 sec	
2	17 min 41 sec	
4	12 min 6 sec	
8	7 min 32 sec	

## Swift WordCount SpeedUp Data

Node No.	Speedup		
1	1		
2	2.02		
4	3.17		
8	5.09		
16	16 7.20		





The performances of swift and Hadoop programs are quite familiar. They both map and reduce the file automatically. So when the concurrency level is low, the running time could be large because many configurations to be done. But as the concurrency level becoming higher, the running time will reduce significantly because the amortized time on each thread is small with more nodes.

5 min 20 sec

## 2. Comparison of Three Methods(Shared-Memory, Swift, Hadoop WordCount)

### Performance Comparison

THREAD(NODE) NO.	SHARED-MEMORY TIME(MIN)	HADOOP TIME(MIN)	SWIFT TIME(MIN)
1	23	87 min 50 sec	38 min 24 sec
16(node only for hadoop & swift)	N/A	6 min 10 sec	5 min 20 sec

For 1 thread, shared-memory takes less time than Hadoop and Swift. As we mentioned above, shared-memory has the highest efficiency.

For 16 threads, swift has a better performance than Hadoop.

From all the performance data we collected, Swift(16 nodes) has the best performance, which compared to Hadoop(16 nodes) and shared-memory(1 thread).

# Conclusion

After several weeks' hard-work, we finally get used to use Hadoop and Swift. After taking long time configuration, we can implement MapReduce on AWS EC2 cluster by using Hadoop and Swift. Through the comparison of performance, we can find a balance between time and cost on build virtual cluster, which makes calculations more effective.