CS553 – Cloud Computing

Programming Assignment 2

Performance Evaluation

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1 Intro

Documentation and Hadoop done by – karan jeet singh (ksingh14@hawk.iit.edu)

SharedMemory and Swift done by – Kanwal preet singh (ksingh13@hawk.iit.edu)

2 System Specifications

All the tests are performed by using the Amazon's EC2 cluster services, out of which c3.large type of instances were employed to run the programs.

C3.large are compute optimized nodes from amazon, they host 7 ECUs, 2 vCPUs @ 2.8GHz each, Intel Xeon E5-2680v2, 3.75 GB memory, 2 x 16 GB SSD Storage Capacity.

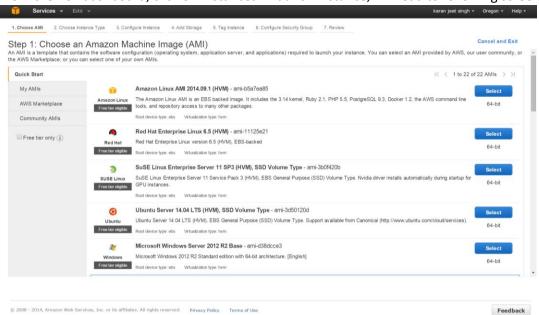
The tests were performed using Linux as its base OS, i.e. Ubuntu server 14.04 LTS (HVM).

JDK version used: 1.7

3 Virtual Cluster (1-node)

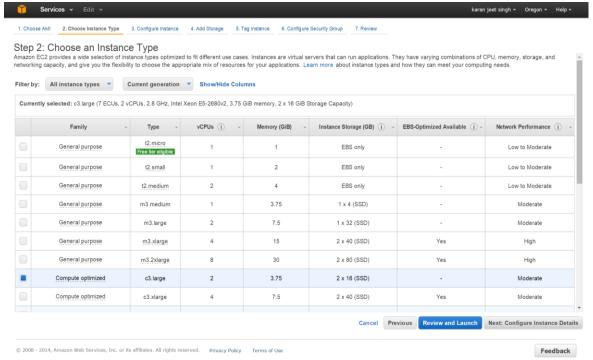
The process followed to create a single node machine is as follows:

1. From the EC2 dashboard, click on Instances > Launch Instance, will lead to following screen:

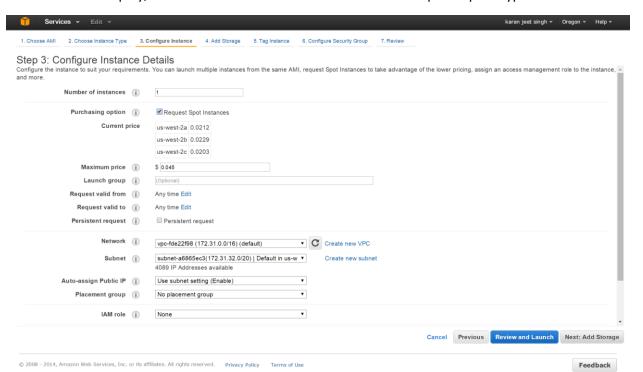


2. Select any OS, but for this assignment Ubuntu Server 14.04 LTS (HVM) was used. After selection, following screen will appear:

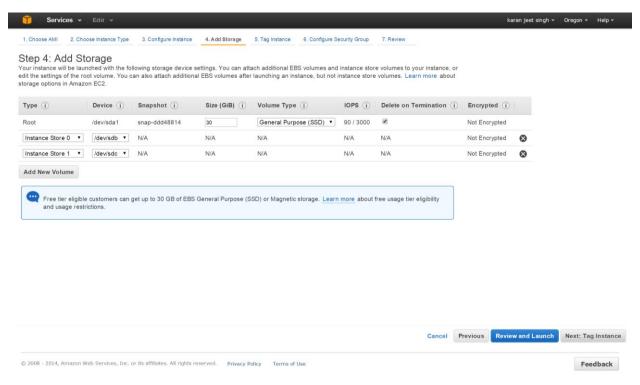
This Step lets the user choose type of instance to be deployed.



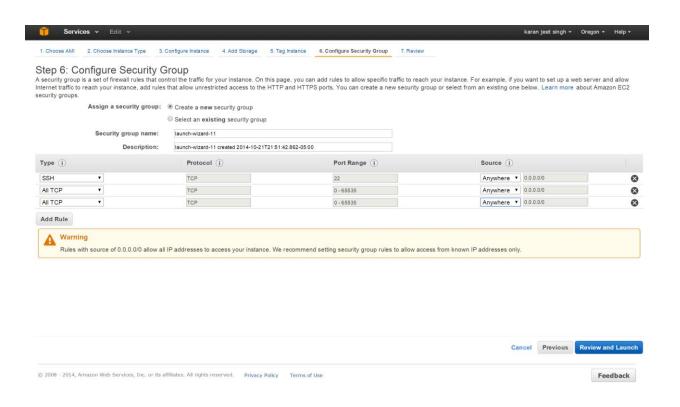
3. For this node setup c3.large was selected. After that we move on to configuring further details for this instance. The first screen we get is where we can specify the number of instances we want to deploy, whether we want On-Demand instances or a Spot-request type of instance.



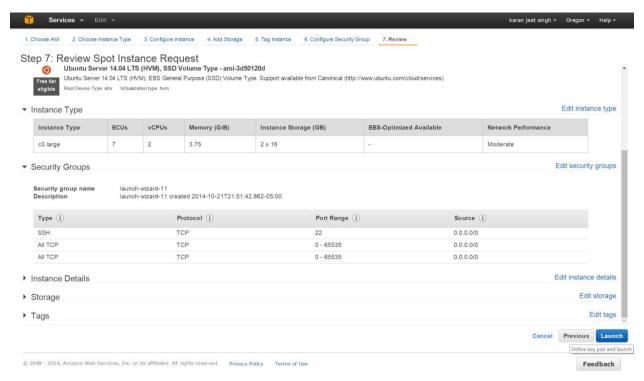
4. After that we go onto the screen where we specify storage details.



5. Next screen allows user to specify tags for the instances, which are being deployed for easier identification, we just skip that screen and move onto the next one which helps configure the security groups and network protocol settings.



6. After specifying the TCP and UDP forwarding details as specified in the screenshot above, we move onto the last screen for review, from where we hit "Launch" and the instance gets deployed after we select the ssh key-pair.



7. After initiation of node, we installed JDK version 1.7 in both shared memory word count and the Hadoop word count codes.

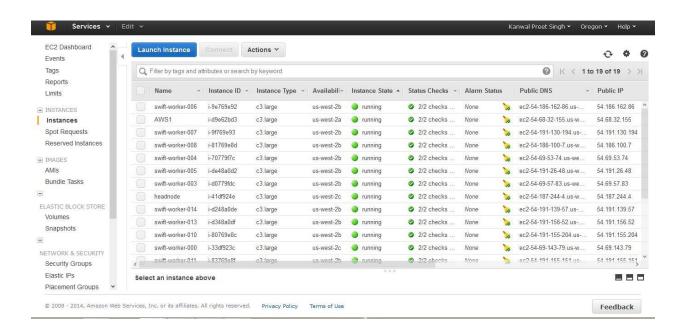
4 SHARED-MEMORY WORD COUNT

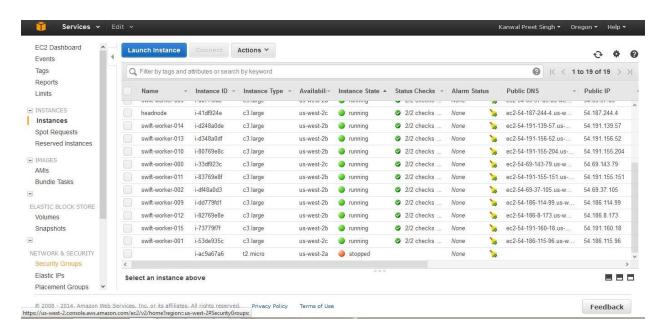
For the purpose of this exercise we developed a word count program using Java. We implemented the basic methodology behind map reduce using Java, the file is split into smaller size files which can be loaded onto the memory all at once.

JDK v1.7 was used to compile the code and create the executable jar files.

5 VIRTUAL CLUSTER (17-NODES)

A 17-node virtual cluster can be setup just like the way single node is set up. In the step 3 of configuration, we can specify the number of clusters to be deployed which would result in launch of 17 similar nodes.





HADOOP

For the sake of this assignment we used Hadoop v1.2.1 to implement the word count algorithm.

We created following shell script to help us speed up the setup process of various nodes for Hadoop from scratch.



6.1 Installing Hadoop

In order to install Hadoop:

- 1. Install JDK v7
- 2. Setup hosts file to use the public DNS name rather than localhost
- 3. Edit .bashrc for environment variables.
- 4. Download and extract Hadoop-1.2.1
- 5. Make a local directory for hdfs to use.
- 6. Edit Hadoop-env.sh, hdfs-site.xml, and mapred-site.xml
- 7. Configure conf/masters, and conf/slaves file.

Screen shot for 10 node setup:

ec2-54-191-56-66 Hadoop Map/Reduce Administration

State: RUNNING Started: Wed Oct 22 14:01:49 UTC 2014 Version: 1.2.1, r1503152 Compiled: Mon Jul 22 15:23:09 PDT 2013 by mattf Identifier: 201410221401 SafeMode: OFF

Cluster Summary (Heap Size is 71 MB/889 MB)

Running Map Tasks	Running Reduce Tasks	Total Submissions	Nodes	Occupied Map Slots	Occupied Reduce Slots	Reserved Map Slots	Reserved Reduce Slots	Map Task Capacity	Reduce Task Capacity	Avg. Tasks/Node	Blacklisted Nodes	Graylisted Nodes	Excluded Nodes
20	1	1	<u>10</u>	20	1	0	0	20	20	4.00	0	Q	<u>Q</u>

Scheduling Information

Queue Name	State	Scheduling Information				
default	running	N/A				

Filter (Jobid, Priority, User, Name)

Example: 'user:smith 3200' will filter by 'smith' only in the user field and '3200' in all fields

Quick Link

7 HADOOP WORD COUNT

Hadoop word count employs the logic of map reduce to implement the word count. A file is split into hash tables, and then a sorted version of it is merged together to produce a final count. Hadoop lets the user use n number of nodes to create the hash tables, i.e. for mapping the input string.

7.1 CONFIGURE HADOOP

Configuring Hadoop requires changes in following 6 files:

- 1. Conf/masters
- Conf/slaves
- 3. Conf/Hadoop-env.sh
- 4. Conf/core-site.xml
- 5. Conf/hdfs-site.xml
- 6. Conf/mapred-site.xml

"masters" file hosts information about the master nodes in the cluster, i.e. the namenode and the secondary namenode.

"slaves" file in the master node hosts the address of all the slave nodes accessible via the master nodes which are then accessed and made part of the cluster, the slaves file in the slave node just has its own address in there so that the node knows that it is acting as a slave.

"Hadoop-env.sh" is used to set environment variables to be used by the Hadoop itself, for this exercise we define the path to the jdk in it.

"core-site.xml" it specifies the information about the master node, i.e. the namenode, and also the path to the directory which is used by Hadoop file system to create a shareable storage.

"hdfs-site.xml" specifies the count of node where the replication block for hdfs is to be created. This number basically determines how many nodes can connect to the master node.

"mapred-site.xml" specifies the path on which the map reduce job tasks are to be run, the address mentioned here should be accessible via all the slave nodes and the secondary namenode as well.

Hadoop map reduce work by dividing the task in hand to different workers, who are also known as slaves, and all the slaves are controlled by one primary node known as master node. When a file is read by master name node, it distributes the workload to the slaves, and then when they are done with their part of processing, the master node sorts the data by managing the data flow between the slaves. After slaves receive the sorted data, it is merged together into a single file which in result is the output file.

While on shared platform, different ports are used to access the file system, and these are configured in the files mentioned above:

- In core-site.xml, it is configured which port is used by the namenode for File System Metadata operations.
- In mapred-site.xml, it is configured which port is used by the JobTracker for Job submission, and Task Tracker heartbeats.

The number of mappers and reducers are also configurable by using below mentioned tags in the mapred-site.xml file:

In such a way we can set the number of map and reduce task to be 2 each for each node.

8 SWIFT

In this assignment we used Swift 0.95-RC6 to implement the word count algorithm.

8.1 Installing Swift

In order to install swift on the cloud (18 node cluster):

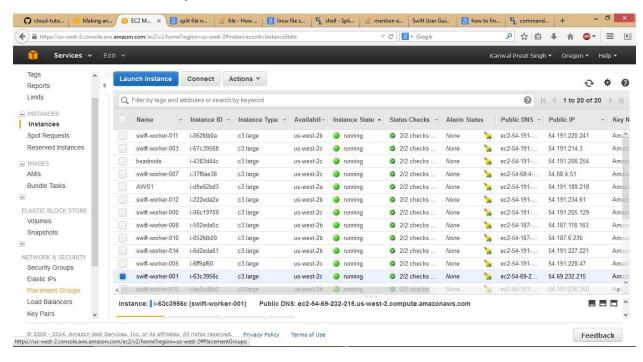
Setting up the Launchpad

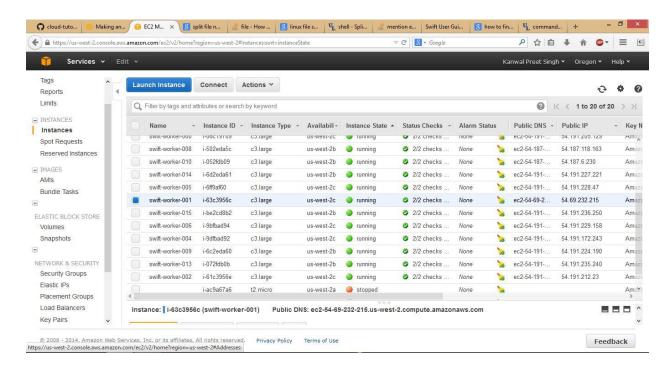
- 1. Install JDK v7
- 2. Install python
- 3. Install git
- 4. Download swift 0.95-RC6.tar.gz and extract it
- 5. Edit .bashrc for environment variables to add Java and swift
- 6. Clone the Github repository cloud-tutorials
- 7. In the EC2 folder of cloud-tutorials edit the configs files
- 8. Run setup.sh



The configs file mentions the headnode, no. of worker, the image to be used for creating the headnode and workers. The security group attached to the headnode and workers and the key details for the ssh connection.

Screen shot for 17 node setup:





9 SWIFT WORD COUNT

In swift we use the shell scripts to implement the word count logic.

First we connect to the Launchpad, from the Launchpad we run the 'setup.sh' script which initializes the workers and the headnode using the 'configs' file. When the headernode and workers are initialized we connect to the headernode. After connecting to the header node we copy the swift-cloud-tutorial folder to the root directory, run the 'setup.sh' file again.

For running the code on 1 node:

All the script files, input file and the swift file were moved to the part01 folder in the swift-cloud-tutorial folder. The files were first split using the splitfile.sh script to split the main file into 16 files which are then saved to the 'input' folder of the current location. The swift then takes input files from this 'input' folder then runs the word count script on individual file and dumps its output in the 'outwc' folder. The files in the 'outwc' contains the word count of individual files. These files are then merged to produce the final file which contains the word count of the original file. All the processing takes place on the head node as site.conf file in the part folder specifies to the swift to execute the process on the local site.

```
P
                                                                             ubuntu@ip-172-31-7-129: ~/swift-cloud-tutorial/part01
sites: local
site.local {
    filesystem {
        type: "local"
        URL: "localhost"
    execution {
        type: "local"
        URL: "localhost"
    workDirectory: "/tmp/"${env.USER}"/"
    maxParallelTasks: 32
    initialParallelTasks: 31
    app.ALL {executable: "*"}
lazyErrors: false
executionRetries: 0
keepSiteDir: true
providerStagingPinSwiftFiles: false
                                                                               Top
```

For running the code on 16 nodes:

All the script files, input file and the swift file were moved to the part05 folder in the swift-cloud-tutorial folder. The files were first split using the splitfile.sh script to split the main file into 16 files which are then saved to the 'input' folder of the current location. The swift then takes input files from this 'input'

folder then sends the individual file to every worker. The worker receives the input file runs the swift logic on it and returns the output file. The output files are dumped in the 'outwc' folder. The files in the 'outwc' contains the word count of individual files. These files are then merged to produce the final file which contains the word count of the original file. The site.conf file mentions the no. of workers and the site of execution of the tasks.

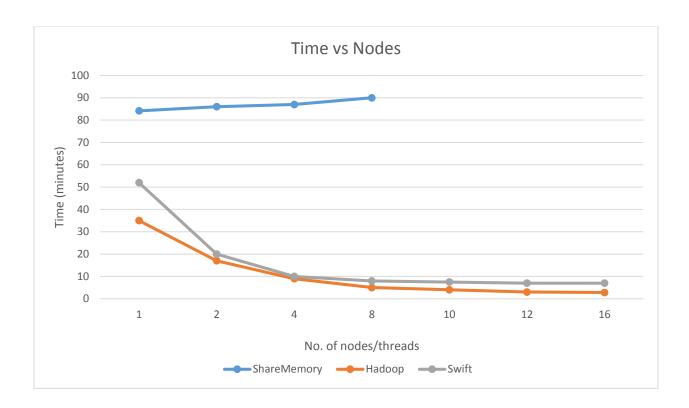
10 Performance Evaluation

Scaling experiments were performed on word count program in different environments to look for the best possible approach to do the same.

SharedMemory-wordcount was implemented using java and was scaled from 1 thread run to max of 8 threads on the c3.large instance.

Wordcount-hadoop was implemented using mapreduce on Hadoop and this was scaled from 1 node processing to the max of 16 nodes.

Wordcount-swift was implemented using swift/T and was scaled from 1 node processing to the max of 16 nodes.



10.1 SHARED MEMORY

The shared memory word count program was made using Java. The ConcurrentHashMaps functionality was used to maintain a single hash table for all the threads, the effects of which are shown in our scaling experiments.

Even when we increase the number of threads from 1 to 8, the performance doesn't change much because the ConcurrentHashMaps become a bottle neck for us.

The speed up graphs also depict the same verdict that there is no change in performance as we jump from 1 thread to more on the c3.large instance

10.2 HADOOP

The word count in Hadoop is implemented by using map reduce functionality.

Increase in number of nodes show clear effect on the performance of the application, the Hadoop system proves out to be very scalable and highly efficient even as we increase the number of nodes.

The speed up graph shows a very little change in the trajectory of the performance of Hadoop, i.e. adding more nodes to the scenario would definitely improve the speed at an equal rate.

10.3 SWIFT/T

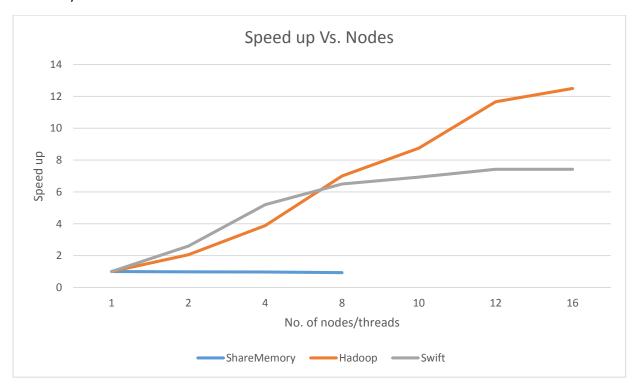
The word count in swift is implemented by combining shell scripting with swift code. The shell scripting is used to split the input file into smaller files, which are then fed to the slaves. At the end, when slaves are done processing, the output files are merged into one using scripts.

In swift, increase in the number of nodes show great signs of improvements, the time it takes to process the file using 1 node is significantly higher than it takes in 16.

The speed up graph shows that the increase in nodes improve the system drastically but even so the change in performance as we go from 8 nodes to 16 nodes is not as much as 1 to 2 nodes.

10.4 COMPARISON

The performance comparison of SharedMemory, Hadoop, and Swift prove that Hadoop is the best when it comes to scalability. The Speed up for it is much more linear and shows drastic change in performance with every node added.



CONCLUSION

Upon evaluating all the above mentioned data and experiments, we notice that both Hadoop and Swift turn out to be better than Shared Memory application.

At single node, Hadoop was the best performer. The only factor that causes the difference in performance between Hadoop and swift at single node is that Hadoop sorts its data before it merges them into a hash table. This helps in quicker formation of hash table and hence is the game changer.

At 16 nodes, Hadoop again turns out to be the best performer. The sorting again shows its effect on the result and also the fact that the different nodes in Hadoop make use of hdfs to pick and store data, this helps in easier and quicker data management between the nodes.

Extrapolating the results to 100 or even more number of nodes reveal that Hadoop will continue to be the best performer.