# **CA675 Cloud Technologies**

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#### I. INTRODUCTION

Stack Exchange [SE] is a website where any users can post questions on various topics in any field. It is a network of questions and answers where anybody can answer to the queries posted on a particular topic. Many major websites are part of SE such as Stack Overflow, Server Fault, Ask Ubuntu, Super User, and many others. In this assignment, Analysis of Large dataset of Stack Overflow has been done using Big Data Technologies: Pig, Hive on Hadoop framework. MapReduce [1] is a programming technique that can be implemented to process and analyze Big datasets on a distributed, parallel computing cluster. Hadoop [2] is a framework comprised of open source software utilities built on the Java platform. The framework uses the MapReduce technique to perform parallel and distributed multiprocessing. Apache Pig [3] is an opensource technology that is built on Hadoop framework. Apache Hive [4] is a data warehouse infrastructure built on the Hadoop system which facilitates querying large datasets residing in the Hadoop storage. The rest of the report is structured in the following way: Section II presents the details of data acquisition, dataset explanation and preprocessing of data. Section III focuses on the roles of Pig and Hive technologies used in this work. Section IV presents a brief information of Google Dataproc in this work. Finally, Section V discusses over all conclusions.

#### II. **DATASET**

Stack Overflow's [5] data is extracted from SE by querying on the table POSTS. It has 22 columns and viewCount is one of the columns. We have analyzed 200000 records in this work. It is possible to fetch only 50000 records per query so, we obtained four separate csv files by running four queries to obtain the top 200000 records ordered by viewCount in descending order. The queries are provided along with this report in the file named "DataAcquisition.txt".

It is necessary to clean the data to load it into Pig. So, we made use of Python 3.7.1 [6] with pandas and re (regular expressions) libraries to clean the data. HTML tags, new-line characters, commas, and special characters are removed from Body and Title columns using Regular expressions. Four csv files are then merged to a single file named "final.csv". The source code for the same is submitted along with the report.

#### III. PIG AND HIVE USAGE

Apache Pig has a Java repository to support User Defined Functions (UDF) called Piggybank [7]. It is a jar file where useful functions for data loading has been written in Java, this Jar file should be registered by specifying the path of the directory where it resides. Once it is registered, we can use any functions defined in the Jar. We can register the jar files with the following syntax "REGISTER <path>". After registering the UDF, it is a good practice to define the function alias that we want to use. We can specify the alias name with the "DEFINE" keyword. Before we start loading the data into Pig it is mandatory to make sure all necessary Hadoop Daemons are up and running. We can use command "jps" to check if the daemons have started. Basic Hadoop commands are provided in a file named "HadoopCommands.txt".

The LOAD operator can be used to load the input data which is stored in the local or Hadoop directory. We have used CSVExcelStorage defined in Piggybank jar.

Null values in the Body and Title fields of the table are removed using the Pig FILTER operator. Once Null values are removed, we can select the necessary columns or fields required into a new relation and then it can be stored into local or Hadoop directory using STORE Keyword. The Pig script used for this work is submitted in the file named "PigScripts.txt". Below are the screenshots of Pig Scripts used in this work.

```
2019-03-05 12:35:31,341 [main] WARN org.apache.pig.PigServer - ATS is disabled since yarn.timeline-service.
[gruntx REGISTER file:/vagrant/pigygbank.jar
[gruntx DEFINE CSVExcelStorage org.apache.pig.pigygbank.storage.CSVExcelStorage;
gruntx pigLoad = LOAD 'hdfs://localhost:9000/final.csv' USING CSVExcelStorage(',','SKIP_INPUT_HEADER') AS
>> (Id:int, PostTypeId:int, AcceptedAnswerId:int,
>> ParentId:int, CreationDate:datetime,
>> DeletionDate:datetime, Score:int,
>> ViewCountint, Body:chararray, OmnerUserId:int,
>> OwnerDisplayName:chararray, LastEditorUserId:int,
>> LastEditOrDisplayName: chararray,
>> LastEditOrDisplayName: chararray, AnswerCount:int,
>> CommentCount:int, FavoriteCount:int,
>> CommentCount:int, FavoriteCount:int,
|>> ClosedDate:chararray, CommunityOwnedDate:datetime);
[gruntx pigSelected = FOREACH pigLoad GENERATE Id,Score,OwnerUserId,OwnerDisplayName,Body,Title;
[gruntx STORE pigSelected INTO 'hdfs://localhost:9000/pigOutput' USING PigStorage(',');
2019-03-05 12:37:06,679 [main] INFO org.apache.hadoop.conf.Configuration.deprecation - mapred.textoutputfor
```

Image 1. Pig Loading Screenshot

```
HadoopVersion PigVersion UserId StartedAt FinishedAt 2.9.0 0.17.0 vagrant 2019-03-08 16:17:15 2019-03-08 16:17:37
 Success!
Job Stats (time in seconds):
JobId Maps Reduces MaxMapTime
job_1551757644221_0063 2
utNulll,
                                                                     MinMapTime
                                                                                                 AvgMapTime
Input(s):
Successfully read 200001 records (188656296 bytes) from: "hdfs://localhost:9000/final.csv"
Output(s):
Successfully stored 195020 records (162888143 bytes) in: "hdfs://localhost:9000/pigOutputWithoutNulll"
Counters:
Total records written : 195020
Total bytes written : 162888143
Spillable Memory Manager spill count : 0
Total bags proactively spilled: 0
Total records proactively spilled: 0
 Job DAG:
job_1551757644221_0063
```

Image 2. Pig Storage Successful Screenshot

Apache Hive is SQL like querying language which facilitates querying of large datasets. In this work, the data stored in the Hadoop directory which is the output of Pig explained earlier is loaded directly into the Hive table while creating. Once the data is loaded into hive tables, we can start querying on the tasks mentioned as a part of this work. Below are the screenshots of Task 1, Task 2 and Task 3 output.

```
2019-03-05 12:52:42,179 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 4.02 sec
MapReduce Total cumulative CPU time: 4 seconds 20 msec
Ended Job = job_1551757644221_0014
MapReduce Jobs Launched:
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 4.02 sec HDFS Read: 177720989 HDFS
Total MapReduce CPU Time Spent: 4 seconds 20 msec
OK
11227809
927358 19181
2003505 14790
                                                                  22648 Why is it faster to process a sorted array than an unsorted arr How do I undo the most recent commits in Git?
How do I delete a Git branch both locally and remotely?
What is the difference between 'git pull' and 'git fetch'?
What is the correct JSON content type?
What does the "yield" keyword do?
What is the "-->" operator in C++?
How to undo isit add! before commit?
 292357 10769
477816 9544
231767 8991
1642028 8152
                                   10769
3.8170 7932 What is the "-->" operator in C++?
3.8170 7932 How to undo 'git add' before commit?
503093 7733 How do I redirect to another webpage?
179123 7676 How to modify existing unpushed commits?
Time taken: 19.246 seconds, Fetched: 10 row(s)
hive>
```

Image 3. Task 1 Output

```
2019-03-05 22:22:54,858 Stage-2 map = 0%, reduce = 0%, Cumulative CPU 2.58 sec
2019-03-05 22:23:00,070 Stage-2 map = 100%, reduce = 0%, Cumulative CPU 3.71 sec
MapReduce Total cumulative CPU time: 3 seconds 710 msec
Ended Job = job_1551757644221_0033
MapReduce Jobs Launched:
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 6.84 sec HDFS Read: 174075383 HDFS Write: 2524948 SUCCESS
Stage-Stage-2: Map: 1 Reduce: 1 Cumulative CPU: 3.71 sec HDFS Read: 2530382 HDFS Write: 324 SUCCESS
Total MapReduce CPU Time Spent: 10 seconds 550 msec
OK
 OK
32585
  22862
                        4883
  21585
                        6068
 19851
16684
                     89904
51816
 15753
15486
                       49153
95592
  14954
                       63051
  14927
  Time taken: 44.078 seconds, Fetched: 10 row(s)
  hive>
```

Image 4. Task 2 Output

```
2019-03-05 13:04:48,146 Stage-1 map = 0%, reduce = 0%
2019-03-05 13:04:54,364 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 3.59 sec
2019-03-05 13:04:59,561 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 4.88 sec
MapReduce Total cumulative CPU time: 4 seconds 880 msec
Ended Job = job_1551757644221_0020
MapReduce Jobs Launched:
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 4.88 sec HDFS Read: 177722852 HDFS Write: 103 SUCCESS Total MapReduce CPU Time Spent: 4 seconds 880 msec
OK
175
Time taken: 17.548 seconds, Fetched: 1 row(s)
hive>
```

Image 5. Task 3 Output

As the result of Task two we will get the top ten users by post score and for each of these users we need to find out the top 10 terms used by each one of them by calculating tf-idf [8], This is the Task four. A new table called TFIDF is created to store the results of Task two which will give the top 10 users. By selecting each user from the TFIDF table, Body and Title content of each post of each of the top ten users is extracted and stored into text file <userId>.txt format where userId represents each user. So, we have now ten text files of the top ten users containing Body and Title data.

These ten text files are provided as inputs for tf-idf calculation using MapReduce. Python 2.7.12 is used to calculate tf-idf. Four mapper functions, three reducer functions are used for this task, the Source code for this task is submitted along with this report and it can also be found on GitHub [9]. Below is the screenshot of the top ten terms used by the user 4883.

Image 6. Task 4. Top Ten terms used by the user 4883(ownerUserId)

The Hive Queries used to perform all of these tasks are submitted in the file named "*HiveQueries.txt*". Since we have page limit for this report only one screenshot of tf-idf and cropped screenshots of other tasks are attached. Detailed and rest of the screenshots are submitted in a folder named "*Screenshots*".

# IV. Google Dataproc

Google Cloud platform offers a public service for the analysis of Big Data called Cloud Dataproc [10]. This platform comprises of a wide variety of public offerings for running Apache Hadoop clusters and Apache Spark. It is simple, cost-effective and quick solutions for Big Data analysis where anybody can create and launch clusters within a few minutes. It provides many open source technologies such as Hive, Pig, Spark, and others for Extract, Transforms and Load processes. In this work, I created a single node cluster consisting of 4 vCPUs and a storage capacity of 20GB. Task one, two and three are executed in the Dataproc. Due to the limitation of space, screenshots are not attached in this report. Screenshots are provided along with report separately.

### V. CONCLUSION

The task of Big Data analysis was challenging in earlier days before the cloud solutions. Public Cloud services offered today for Big Data analysis makes the whole process simple, quick and cheap. In this work, I have analyzed the data on both locally(vagrant) and on the cloud platform, working locally on a virtual machine with Hadoop framework, Hive and Pig technologies made me understand the Hadoop ecosystem but, it also has its disadvantages such as time-consuming, manual setup of each technology separately. On the other side, if we use cloud platforms, it has its advantages where the user need not worry about the setup and configuration also it is quick. Overall, whether it is a local platform or a cloud we see the major benefits of using the MapReduce programming technique for the Big Data analysis makes the whole process fast.

## **REFERENCES**

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