Mini-Project Report

<u>Title</u>: Data Analytics using Cloudera Hadoop



Dataset Used: Stocks.csv, Dividends.csv

<u>Technologies Used</u>: Hadoop, HDFS, Map-Reduce, Hive, Sqoop, Pig, Spark using Scala

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Objective:

The Mini-Project consists of 5 sub-projects which uses different technologies.

Part 1: Map-Reduce: Finding words present in different files

Objective: to find which words are present in how many files along with the file names using Map-Reduce Framework.

Part 2: Sqoop Task: Loading Data from RDBMS to HDFS

Objective: to load data from RDBMS (MySQL) to HDFS & then load data from HDFS to Hive tables using Apache Sqoop.

Part 3: Stocks Analysis using Hive

Objective: to run Hive queries on the stocks dataset loaded using Sqoop to understand it better and then perform analytics on it.

Part 4: Pig Analytics

Objective: to perform basic queries on 2 datasets: stocks and dividends and then joining them using Pig Latin to understand how Apache Pig Framework works.

Part 5: Twitter's Top 10 popular Hashtag Streaming per second using Apache Spark

Objective: to find Top 10 popular Hashtag on Twitter and perform Web Scraping using Spark & Scala to stream the data on per second basis.

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Introduction to Big Data

Big data is data sets that are so big and complex that traditional data-processing application software are inadequate to deal with them. Big data challenges include capturing data, data storage, data analysis, search, sharing, transfer, visualization, querying, updating, information privacy and data source. There are a number of concepts associated with big data: originally there were 3 concepts volume, variety, velocity. Other concepts later attributed with big data are veracity (i.e., how much noise is in the data) and value.

Lately, the term "big data" tends to refer to the use of predictive analytics, user behavior analytics, or certain other advanced data analytics methods that extract value from data, and seldom to a particular size of data set. "There is little doubt that the quantities of data now available are indeed large, but that's not the most relevant characteristic of this new data ecosystem." Analysis of data sets can find new correlations to "spot business trends, prevent diseases, combat crime and so on." Scientists, business executives, practitioners of medicine, advertising and governments alike regularly meet difficulties with large data-sets in areas including Internet search, fintech, urban informatics, and business informatics. Scientists encounter limitations in e-Science work, including meteorology, genomics, complex physics simulations, biology and environmental research.

Characteristics of Big Data

Big data can be described by the following characteristics:

Volume

The quantity of generated and stored data. The size of the data determines the value and potential insight, and whether it can be considered big data or not.

Variety

The type and nature of the data. This helps people who analyze it to effectively use the resulting insight. Big data draws from text, images, audio, video; plus, it completes missing pieces through data fusion.

Velocity

In this context, the speed at which the data is generated and processed to meet the demands and challenges that lie in the path of growth and development. Big data is often available in real-time.

Veracity

The data quality of captured data can vary greatly, affecting the accurate analysis.

Why is Big Data Important?

The importance of big data doesn't revolve around how much data you have, but what you do with it. You can take data from any source and analyze it to find answers that enable

- 1) cost reductions
- 2) time reductions
- 3) new product development and optimized offerings
- 4) smart decision making.

When you combine big data with high-powered analytics, you can accomplish business-related tasks such as:

- Determining root causes of failures, issues and defects in near-real time.
- Generating coupons at the point of sale based on the customer's buying habits.
- Recalculating entire risk portfolios in minutes.
- Detecting fraudulent behavior before it affects your organization.

Benefits of Big Data & Data Analytics

Big data makes it possible for you to gain more complete answers because you have more information.

More complete answers mean more confidence in the data—which means a completely different approach to tackling problems.

Big Data Challenges

While big data holds a lot of promise, it is not without its challenges.

- First, big data is...big. Although new technologies have been developed for data storage, data volumes are doubling in size about every two years.
 Organizations still struggle to keep pace with their data and find ways to effectively store it.
- But it's not enough to just store the data. Data must be used to be valuable and that depends on curation. Clean data, or data that's relevant to the client and organized in a way that enables meaningful analysis, requires a lot of work. Data scientists spend 50 to 80 percent of their time curating and preparing data before it can actually be used.
- Finally, big data technology is changing at a rapid pace. A few years ago, Apache Hadoop was the popular technology used to handle big data. Then Apache Spark was introduced in 2014. Today, a combination of the two frameworks appears to be the best approach. Keeping up with big data technology is an ongoing challenge

How Big Data works

Big data gives you new insights that open up new opportunities and business models. Getting started involves three key actions:

Integrate

Big data brings together data from many disparate sources and applications. Traditional data integration mechanisms, such as ETL (extract, transform, and load) generally aren't up to the task. It requires new strategies and technologies to analyze big data sets at terabyte, or even petabyte, scale.

During integration, you need to bring in the data, process it, and make sure it's formatted and available in a form that your business analysts can get started with.

Manage

Big data requires storage. Your storage solution can be in the cloud, on premises, or both. You can store your data in any form you want and bring your desired processing requirements and necessary process engines to those data sets on an on-demand basis. Many people choose their storage solution according to where their data is currently residing. The cloud is gradually gaining popularity because it supports your current compute requirements and enables you to spin up resources as needed.

Analyze

Your investment in big data pays off when you analyze and act on your data. Get new clarity with a visual analysis of your varied data sets. Explore the data further to make new discoveries. Share your findings with others. Build data models with machine learning and artificial intelligence. Put your data to work.

Introduction to Hadoop

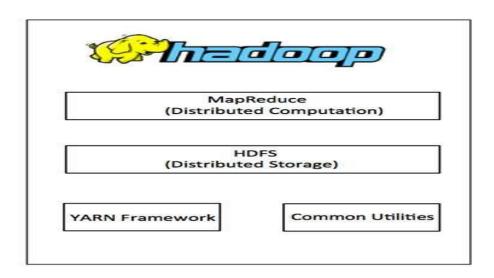
Hadoop is an Apache open source framework written in java that allows distributed processing of large datasets across clusters of computers using simple programming models. A Hadoop frame-worked application works in an environment that provides distributed storage and computation across clusters of computers. Hadoop is designed to scale up from single server to thousands of machines, each offering local computation and storage.

Hadoop Architecture

Hadoop framework includes following four modules:

- Hadoop Common: These are Java libraries and utilities required by other Hadoop modules. These libraries provide filesystem and OS level abstractions and contains the necessary Java files and scripts required to start Hadoop.
- Hadoop YARN: This is a framework for job scheduling and cluster resource management.
- Hadoop Distributed File System (HDFS): A distributed file system that provides high-throughput access to application data.

• **Hadoop MapReduce:** This is YARN-based system for parallel processing of large data sets.



Advantages of Hadoop

- Hadoop framework allows the user to quickly write and test distributed systems. It is efficient, and it automatic distributes the data and work across the machines and in turn, utilizes the underlying parallelism of the CPU cores.
- Hadoop does not rely on hardware to provide fault-tolerance and high availability (FTHA), rather Hadoop library itself has been designed to detect and handle failures at the application layer.
- Servers can be added or removed from the cluster dynamically and Hadoop continues to operate without interruption.
 Another big advantage of Hadoop is that apart from being open source, it is compatible on all the platforms since it is Java based.

Why is Hadoop important?



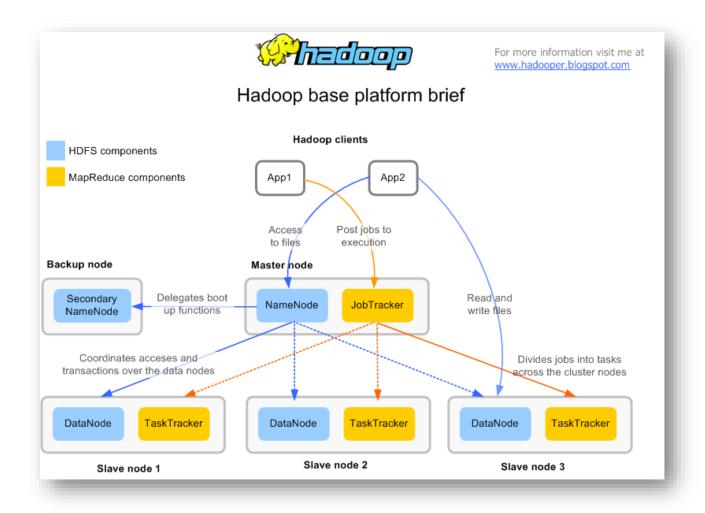
Hadoop Distributed File System

Hadoop can work directly with any mountable distributed file system such as Local FS, HFTP FS, S3 FS, and others, but the most common file system used by Hadoop is the Hadoop Distributed File System (HDFS).

The Hadoop Distributed File System (HDFS) is based on the Google File System (GFS) and provides a distributed file system that is designed to run on large clusters (thousands of computers) of small computer machines in a reliable, fault-tolerant manner.

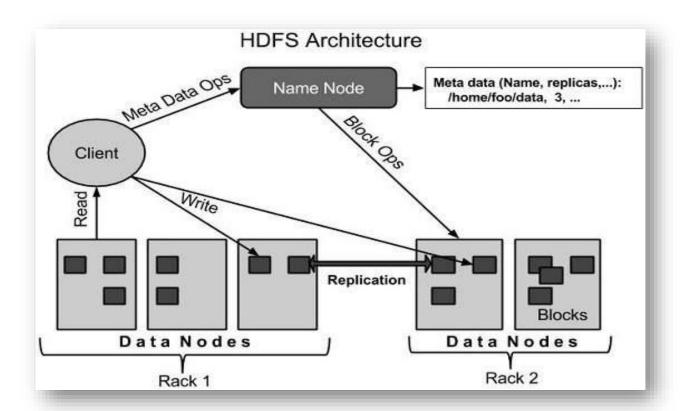
HDFS uses a master/slave architecture where master consists of a single NameNode that manages the file system metadata and one or more slave DataNodes that store the actual data.

A file in an HDFS namespace is split into several blocks and those blocks are stored in a set of DataNodes. The NameNode determines the mapping of blocks to the DataNodes. The DataNodes takes care of read and write operation with the file system. They also take care of block creation, deletion and replication based on instruction given by NameNode. HDFS provides a shell like any other file system and a list of commands are available to interact with the file system.



HDFS Architecture

HDFS follows the master-slave architecture and it has the following elements.



Namenode

The namenode is the commodity hardware that contains the GNU/Linux operating system and the namenode software. It is a software that can be run on commodity hardware. The system having the namenode acts as the master server and it does the following tasks:

- Manages the file system namespace.
- Regulates client's access to files.
- ➤ It also executes file system operations such as renaming, closing, and opening files and directories.

Datanode

The datanode is a commodity hardware having the GNU/Linux operating system and datanode software. For every node (Commodity hardware/System) in a cluster, there will be a datanode. These nodes manage the data storage of their system. Datanodes perform read-write operations on the file systems, as per client request.

They also perform operations such as block creation, deletion, and replication according to the instructions of the namenode.

Block

Generally, the user data is stored in the files of HDFS. The file in a file system will be divided into one or more segments and/or stored in individual data nodes. These file segments are called as blocks. In other words, the minimum amount of data that HDFS can read or write is called a Block. The default block size is 64MB, but it can be increased as per the need to change in HDFS configuration.

How Does Hadoop Work?

Stage 1

A user/application can submit a job to the Hadoop (a Hadoop job client) for required process by specifying the following items:

- > The location of the input and output files in the distributed file system.
- ➤ The java classes in the form of jar file containing the implementation of map and reduce functions.
- The job configuration by setting different parameters specific to the job.

• Stage 2

The Hadoop job client then submits the job (jar/executable etc.) and configuration to the JobTracker which then assumes the responsibility of distributing the software/configuration to the slaves, scheduling tasks and monitoring them, providing status and diagnostic information to the job-client.

Stage 3

The TaskTrackers on different nodes execute the task as per MapReduce implementation and output of the reduce function is stored into the output files on the file system.

Map-Reduce

MapReduce is a framework using which we can write applications to process huge amounts of data, in parallel, on large clusters of commodity hardware in a reliable manner.

MapReduce is a processing technique and a program model for distributed computing based on java. The MapReduce algorithm contains two important tasks, namely Map and Reduce. Map takes a set of data and converts it into another set of data, where individual elements are broken down into tuples (key/value pairs). Secondly, reduce task, which takes the output from a map as an input and combines those data tuples into a smaller set of tuples. As the sequence of the name MapReduce implies, the reduce task is always performed after the map job.

The major advantage of MapReduce is that it is easy to scale data processing over multiple computing nodes. Under the MapReduce model, the data processing primitives are called mappers and reducers. Decomposing a data processing application into mappers and reducers is sometimes nontrivial. But, once we write an application in the MapReduce form, scaling the application to run over hundreds, thousands, or even tens of thousands of machines in a cluster is merely a configuration change.

Map-Reduce: Finding words present in different files

mapper.py

```
File Edit View Search Tools Documents
reducer.py 💥 🖹 file2 💥 📄 file1 💥 📄 inputFilesPath 💥
                                                       mapper.py
                                                                       file1.txt 💥
#!/usr/bin/env python
from os import listdir
from os.path import isfile, join
import sys
for mypath in sys.stdin:
   mypath = mypath.rstrip()
   onlyFiles = [join(mypath, f) for f in listdir(mypath) if isfile(join(mypath, f))]
   fileNames = [f for f in listdir(mypath) if isfile(join(mypath, f))]
# print(onlyFiles)
   i = 0
   for file in onlyFiles:
   # print(file)
     fileName = fileSplit[-1]
        f = open(file, 'r')
       lines = f.readlines()
        for line in lines:
           for word in line.split():
               print fileNames[i] +" "+ word
        i += 1
```

reducer.py

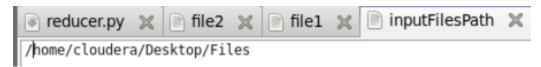
```
#!/usr/bin/env python
import sys
def removeDuplicates(words):
   removed = []
    for word in words:
       if word not in removed:
           removed.append(word)
   return removed
wordKeyFileValue = {}
final = []
fileAndWord = \{\}
words = []
for line in sys.stdin:
   line = line.strip()
   file, word = line.split()
   words.append(word)
   if file not in fileAndWord:
       fileAndWord[file] = []
       fileAndWord[file].append(word)
   else:
       if word not in fileAndWord[file]:
           fileAndWord[file].append(word)
   removed = removeDuplicates(words)
   # print(removed)
   # print(fileAndWord)
    for word in removed:
       for key in fileAndWord.keys():
           if word in fileAndWord[key]:
               final.append(key)
       wordKeyFileValue[word] = final
       final = []
    # print(wordKeyFileValue)
   words = sorted(wordKeyFileValue.keys())
   values = wordKeyFileValue.values()
for word, values in zip(words, values):
   print
   print word,
    for value in values:
       print value,
   print
   # print(fileAndWord)
```

file1.txt

file2.txt



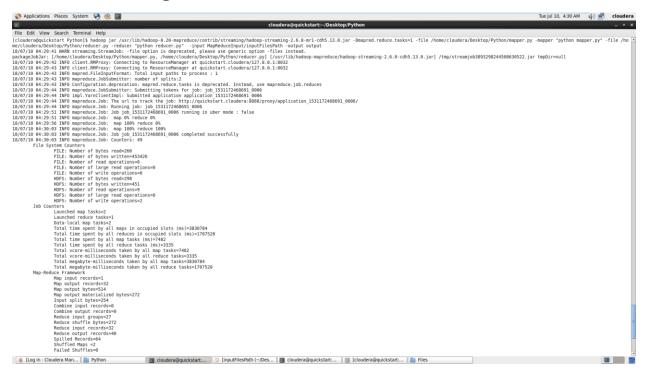
inputFilesPath



Command:



Command Running:



Output:

```
[cloudera@quickstart Python]$ hadoop fs -cat output/*0000

    file1.txt file2.txt

after file1.txt
and file1.txt file2.txt
based file1.txt
file file1.txt
has file2.txt
hey file1.txt file2.txt
is file1.txt file2.txt
map file2.txt
methods file1.txt
name file1.txt file2.txt
on file1.txt
output file1.txt file2.txt
project file2.txt
python file2.txt
reduce file1.txt file2.txt
the file1.txt
two file2.txt
upon file2.txt
word file2.txt
[cloudera@quickstart Python]$
```

Introduction to Sqoop

Apache Sqoop is a tool in Hadoop ecosystem which is designed to transfer data between HDFS (Hadoop storage) and relational database servers like MySQL, Oracle RDB, SQLite, Teradata, Netezza, Postgres etc. Apache Sqoop imports data from relational databases to HDFS, and exports data from HDFS to relational databases. It efficiently transfers bulk data between Hadoop and external datastores such as enterprise data warehouses, relational databases, etc.

This is how Sqoop got its name – "SQL to Hadoop & Hadoop to SQL".

Additionally, Sqoop is used to import data from external datastores into Hadoop ecosystem's tools like Hive & HBase.

Why Sqoop?

For Hadoop developer, the actual game starts after the data is being loaded in HDFS. They play around this data in order to gain various insights hidden in the data stored in HDFS.

So, for this analysis the data residing in the relational database management systems need to be transferred to HDFS. The task of writing MapReduce code for importing and exporting data from relational database to HDFS is uninteresting & tedious. This is where Apache Sqoop comes to rescue and removes their pain. It automates the process of importing & exporting the data.

Sqoop makes the life of developers easy by providing CLI for importing and exporting data. They just have to provide basic information like database authentication, source, destination, operations etc. It takes care of remaining part.

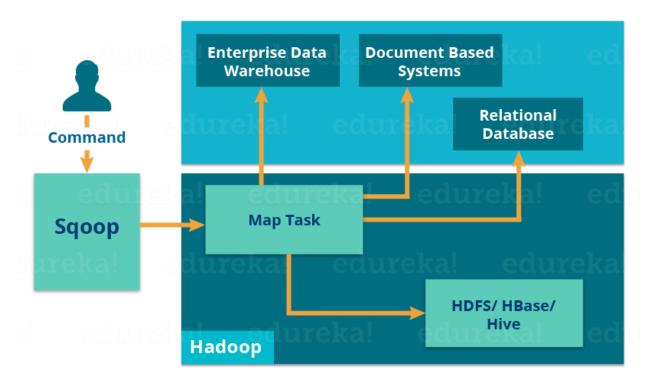
Sqoop internally converts the command into MapReduce tasks, which are then executed over HDFS. It uses YARN framework to import and export the data, which provides fault tolerance on top of parallelism.

Key Features of Sqoop

Sqoop provides many salient features like:

- **Full Load**: Apache Sqoop can load the whole table by a single command. You can also load all the tables from a database using a single command.
- **Incremental Load:** Apache Sqoop also provides the facility of incremental load where you can load parts of table whenever it is updated.
- Parallel import/export: Sqoop uses YARN framework to import and export the data, which provides fault tolerance on top of parallelism.
- Import results of SQL query: You can also import the result returned from an SQL query in HDFS.
- **Compression:** You can compress your data by using deflate(gzip) algorithm with –compress argument, or by specifying –compression-codec argument. You can also load compressed table in Apache Hive.
- Connectors for all major RDBMS Databases: Apache Sqoop provides connectors for multiple RDBMS databases, covering almost the entire circumference.
- **Kerberos Security Integration:** Kerberos is a computer network authentication protocol which works on the basis of 'tickets' to allow nodes communicating over a non-secure network to prove their identity to one another in a secure manner. Sqoop supports Kerberos authentication.
- Load data directly into HIVE/HBase: You can load data directly into Apache Hive for analysis and also dump your data in HBase, which is a NoSQL database.
- **Support for Accumulo:** You can also instruct Sqoop to import the table in Accumulo rather than a directory in HDFS.

Sqoop Architecture



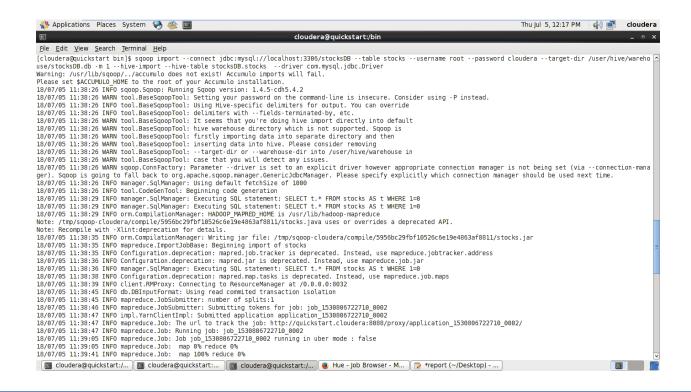
Sqoop Task: Loading Data from RDBMS to HDFS

Dataset used: stocks.csv

Task – to load data from RDBMS (MySQL) to HDFS & then load data from HDFS to Hive tables

Step 1 – Loading data from RDBMS (MySQL) to HDFS

Command: sqoop import --connect jdbc:mysql://localhost:3306/stocksDB
--table stocks --username root --password cloudera --target-dir
/user/hive/warehouse/stocksDB.db -m 1 --hive-import --hive-table
stocksDB.stocks --driver com.mysql.jdbc.Driver



What is Hive

Hive is a data warehouse infrastructure tool to process structured data in Hadoop. It resides on top of Hadoop to summarize Big Data and makes querying and analyzing easy.

Initially Hive was developed by Facebook, later the Apache Software Foundation took it up and developed it further as an open source under the name Apache Hive. It is used by different companies. For example, Amazon uses it in Amazon Elastic MapReduce.

Hive is not:

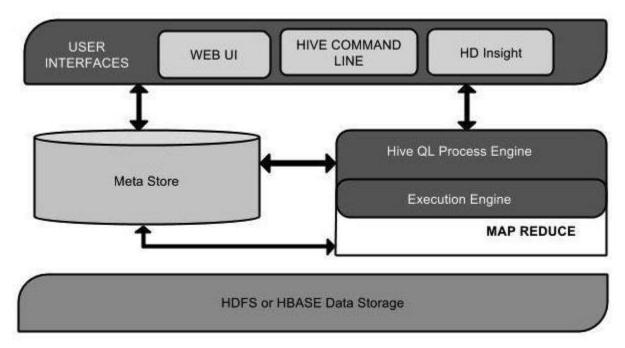
- A relational database
- A design for OnLine Transaction Processing (OLTP)
- A language for real-time queries and row-level updates

Features of Hive

- It stores schema in a database and processed data into HDFS.
- It is designed for OLAP.
- It provides SQL type language for querying called HiveQL or HQL.
- It is familiar, fast, scalable, and extensible

Architecture of Hive

The following component diagram depicts the architecture of Hive:



Stocks Analysis using Hive

Dataset used: Stocks.csv

Step 1 - Opening Hive2 or Thrift Server & make a connection with it.

```
[cloudera@quickstart Desktop]$ cd $HIVE_HOME/bin
[cloudera@quickstart bin]$ beeline
Beeline version 1.1.0-cdh5.13.0 by Apache Hive
beeline> !connect jdbc:hive2://localhost:10000/default
scan complete in 2ms
Connecting to jdbc:hive2://localhost:10000/default
Enter username for jdbc:hive2://localhost:10000/default: cloudera
Enter password for jdbc:hive2://localhost:10000/default: *******
Connected to: Apache Hive (version 1.1.0-cdh5.13.0)
Driver: Hive JDBC (version 1.1.0-cdh5.13.0)
Transaction isolation: TRANSACTION_REPEATABLE_READ
0: jdbc:hive2://localhost:10000/default> ■
```

Step 2 - Creating & using Database stocks_DB

Command - create database stocks_DB;

```
0: jdbc:hive2://localhost:10000/default> create database stocks DB;
INFO: Compiling\ command (queryId=hive\_20180709234747\_80246bbf-\overline{b}c30-4824-ad40-7a688b78926a):\ create\ database\ stocks\_DB
                  Semantic Analysis Completed
           : Returning Hive schema: Schema(fieldSchemas:null, properties:null)
            : Completed compiling command(queryId=hive_20180709234747_80246bbf-6c30-4824-ad40-7a688b78926a); Time taken: 0.016 seconds
INFO : Executing command(queryId=hive_20180709234747_80246bbf-6c30-4824-ad40-7a688b78926a): create database stocks_DB
TNFO
           : Starting task [Stage-0:DDL] in serial mode
INFO : Completed executing command(queryId=hive_20180709234747_80246bbf-6c30-4824-ad40-7a688b78926a); Time taken: 0.459 seconds
INFO : OK
No rows affected (0.503 seconds)
0: jdbc:hive2://localhost:10000/default> use stocks DB;
INFO : Compiling command(queryId=hive 20180709234747 e08a8c96-302f-43f8-8697-5ea94d0c8595): use stocks DB
INFO : Semantic Analysis Completed
INFO
                  Returning Hive schema: Schema(fieldSchemas:null, properties:null)
INFO: Completed compiling command (queryId=hive\_20180709234747\_e08a8c96-302f-43f8-8697-5ea94d0c8595); \ Time \ taken: 0.021 \ seconds \ taken: 0
INFO : Executing command(queryId=hive_0180709234747_e08a8c96-302f-43f8-8697-5ea94d0c8595): use stocks_DB
INFO : Starting task [Stage-0:DDL] in serial mode
           : Completed executing command(queryId=hive_20180709234747_e08a8c96-302f-43f8-8697-5ea94d0c8595); Time taken: 0.014 seconds
INFO
No rows affected (0.055 seconds)
```

Step 3 - Creating External table stocks

Command -

```
create external table if not exists stocks
exch STRING,
symbol STRING,
date STRING,
stocks price open FLOAT,
stocks price high FLOAT,
stocks price low FLOAT,
stocks price close FLOAT,
volume INT,
price adj close FLOAT
COMMENT 'This is External Table for Stocks'
ROW FORMAT DELIMITED
FIELDS TERMINATED BY',';
9: jdbc:hive2://localhost:10000/default> create external table if not exists stocks
 . . . . . . . . . . . . . . . exch STRING,
  . . . . . . . . . . . . . . . . . symbol STRING,
        . . . . . . . . . . . . . . date STRING,
                . . . . . . .> stocks_price_open FLOAT,
                       . .> stocks price high FLOAT,
. . . . . . . > stocks_price_close FLOAT, . . . . . . . > volume INT,
                    . . . .> price adj close FLOAT
```

Step 4 - Executing Queries

Query 1: Finding top 10 companies according to number of stocks

Command - select symbol, count(symbol) as count from stocks group by symbol order by count desc limit 10;

```
0: jdbc:hive2://localhost:10000/default> select symbol,count(symbol) as count from stocks group by symbol order by count desc limit 10;
INFO : Compiling command(queryId=hive_20180710002323_ca41f7dc-9a7d-4864-bcdb-36e569775482): select symbol,count(symbol) as count from stocks group by symbol
 order by count desc limit 10
INFO : Semantic Analysis Completed
INFO : Returning Hive schema: Schema(fieldSchemas:[FieldSchema(name:symbol, type:string, comment:null), FieldSchema(name:count, type:bigint, comment:null)],
 properties:null)
INFO : Completed compiling command(queryId=hive_20180710002323_ca41f7dc-9a7d-4864-bcdb-36e569775482); Time taken: 0.273 seconds
INFO : Executing command(queryId=hive_20180710002323_ca41f7dc-0a7d-4864-bcdb-36e569775482): select symbol,count(symbol) as count from stocks group by symbol
 order by count desc limit 10
INFO : Query ID = hive 20180710002323 ca41f7dc-9a7d-4864-bcdb-36e569775482
       : Total jobs = 2
INFO : Launching Job 1 out of 2
INFO : Starting task [Stage-1:MAPRED] in serial mode
INFO : Number of reduce tasks not specified. Estimated from input data size: 1
      : In order to change the average load for a reducer (in bytes):
INFO
          set hive.exec.reducers.bytes.per.reducer=<number>
INFO : In order to limit the maximum number of reducers:
           set hive.exec.reducers.max=<number>
INFO : In order to set a constant number of reducers:
INFO
          set mapreduce.job.reduces=<number>
INFO : number of splits:1
INFO : Submitting tokens for job: job 1531159788380 0006
       : The url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159788380_0006/
INFO : Starting Job = job 1531159788380 0006, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0006/INFO : Kill Command = /usr/lib/hadoop/bin/hadoop job -kill job_1531159788380_0006
INFO : Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1
       : 2018-07-10 00:23:41,179 Stage-1 map = 0%, reduce = 0%
      : 2018-07-10 00:23:48,588 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 2.92 sec
: 2018-07-10 00:23:55,938 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 4.6 sec
INFO
INFO
       : MapReduce Total cumulative CPU time: 4 seconds 600 msec
       : Ended Job = job_1531159788380_0006
      : Launching Job 2 out of 2
: Starting task [Stage-2:MAPRED] in serial mode
TNFO
INFO
      : Number of reduce tasks determined at compile time: 1
       : In order to change the average load for a reducer (in bytes):
TNFO
          set hive.exec.reducers.bytes.per.reducer=<number>
INF0
      : In order to limit the maximum number of reducers:
      : set hive.exec.reducers.max=<number>
```

```
set hive.exec.reducers.bytes.per.reducer=<number>
INFO : In order to limit the maximum number of reducers:
INFO
      : set hive.exec.reducers.max=<number>
: In order to set a constant number of reducers:
INFO
INFO
           set mapreduce.job.reduces=<number>
          number of splits:1
INFO :
          Submitting tokens for job: job_1531159788380_0007
INFO
          The url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159788380_0007/
          Starting Job = job 1531159788380 0007, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0007/
Kill Command = /usr/lib/hadoop/bin/hadoop job -kill job_1531159788380_0007
TNFO
INFO
          Hadoop job information for Stage-2: number of mappers: 1; number of reducers: 1 2018-07-10 00:24:04,053 Stage-2 map = 0%, reduce = 0%
INFO
INF0
          2018-07-10 00:24:09,438 Stage-2 map = 100%, reduce = 0%, Cumulative CPU 1.13 sec 2018-07-10 00:24:16,793 Stage-2 map = 100%, reduce = 100%, Cumulative CPU 2.79 sec
INF0
INFO
INF0
          MapReduce Total cumulative CPU time: 2 seconds 790 msec
INFO
          Ended Job = job_1531159788380_0007
INFO
          MapReduce Jobs Launched:
       : Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 4.6 sec : Stage-Stage-2: Map: 1 Reduce: 1 Cumulative CPU: 2.79 sec : HDFS Read: 40998932 HDFS Write: 4994 SUCCESS : Stage-Stage-2: Map: 1 Reduce: 1 Cumulative CPU: 2.79 sec : HDFS Read: 10050 HDFS Write: 91 SUCCESS
INFO :
INFO
       : Total MapReduce CPU Time Spent: 7 seconds 390 msec
INF0
INFO : Completed executing command(queryId=hive 20180710002323 ca41f7dc-9a7d-4864-bcdb-36e569775482); Time taken: 43.735 seconds
INFO
| symbol | count |
               12109
| AA
  AEP
               10121
  AET
  AXP
               8290
  ASA
               8095
  AMR
               7590
  APA
               7091
  AVP
               7085
  AVT
               7084
  ALK
10 rows selected (44.14 seconds)
```

Query 2: Finding top 10 years according to average volume of stocks

Command - select year(date), avg(volume) as avgVol from stocks group by year(date) order by avgVol desc limit 10;

```
0: jdbc:hive2://localhost:10000/default> select year(date),avg(volume) as avgVol from stocks group by year(date) order by avgVol desc limit 10; INFO : Compiling command(queryId=hive_20180710002626_ba156c06-867f-45b0-9a1f-7744363c10b2): select year(date),avg(volume) as avgVol from stocks group by year(date) order by avgVol desc limit 10 INFO : Semantic Analysis Completed INFO : Returning Hive schema: Schema(fieldSchemas:[FieldSchema(name:_c0, type:int, comment:null), FieldSchema(name:avgvol, type:double, comment:null)], prop
erties:null)
INFO : Completed compiling command(queryId=hive 20180710002626 ba156c06-867f-45b0-9a1f-7744363c10b2); Time taken: 0.107 seconds
INFO : Executing command(queryId=hive_20180710002626 ba156c06-867f-45b0-9a1f-7744363c10b2): select year(date),avg(volume) as avgVol from stocks group by year(date) order by avgVol desc limit 10
INFO : Query ID = hive_20180710002626_ba156c06-867f-45b0-9a1f-7744363c10b2
INFO : Total jobs = 2
INFO : Launching Job 1 out of 2
INFO : Starting task [Stage-1:MAPRED] in serial mode
INFO : In order to change the average load for a reducer (in bytes):
INFO : In order to claim the maximum number of reducers:
 INF0
               : In order to limit the maximum number of reducers:
             : set hive.exec.reducers.max=<number>
: In order to set a constant number of reducers:
                   set mapreduce.job.reduces=<number>
number of splits:1
 TNFO
                  Submitting tokens for job: job_1531159788380_0008
 INFO
| INFO | Submitting tokens for job: job_1531159788380 0008 | The url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159788380_0008 | The url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159788380_0008 | INFO | Starting Job = job_1531159788380_0008, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0008 | INFO | Still Command = /usr/lib/hadoop/bin/hadoop job - kill job_1531159788380_0008 | INFO | Still Command = /usr/lib/hadoop/bin/hadoop job - kill job_1531159788380_0008 | INFO | 2018-07-10 00:26:43,971 Stage-1 map = 0%, reduce = 0% | Cumulative CPU 5.64 sec | INFO | 2018-07-10 00:26:52,518 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 7.37 sec | INFO | 2018-07-10 00:26:58,831 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 7.37 sec | INFO | Ended Job = job_1531159788380_0008 | INFO | Launching Job 2 out of 2 | INFO | Launching Job 2 out of 2 | INFO | INF
 INFO : Kill Command = /usr/lib/hadoop/bin/hadoop job -kill job 1531159788380 0009
 INFO : 2018-07-10 00:27:05,691 Stage-2 map = 0%, reduce = 0%
 INFO : 2018-07-10 00:27:12,189 Stage-2 map = 100%, reduce = 0%, Cumulative CPU 1.63 sec
  INFO : 2018-07-10 00:27:18,535 Stage-2 map = 100%,
                                                                                                                                                    reduce = 100%, Cumulative CPU 3.34 sec
  INFO : MapReduce Total cumulative CPU time: 3 seconds 340 msec
  INFO
                  : Ended Job = job 1531159788380 0009
  INFO : MapReduce Jobs Launched:
 INFO : Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 7.37 sec HDFS Read: 40999849 HDFS Write: 1468 SUCCESS INFO : Stage-Stage-2: Map: 1 Reduce: 1 Cumulative CPU: 3.34 sec HDFS Read: 6538 HDFS Write: 235 SUCCESS
  INFO : Total MapReduce CPU Time Spent: 10 seconds 710 msec
  INFO : Completed executing command(queryId=hive_20180710002626_ba156c06-867f-45b0-9a1f-7744363c10b2); Time taken: 43.49 seconds
  INFO : OK
                                      avgvol
  | _c0 |
      2009 | 2144999.0106877508
       2010
                        2009298.4872611465
       2008 | 1941244.2091383133
       2007
                           1337635.7611070648
       1983
                        1087647.1091521909
       2006
                         | 1059718.7874033398
                        1021882.4769433466
       1982
                             941810.9683610097
       2003
       2005
                             902292.056767634
      2002 | 899578.376947301
 10 rows selected (43.678 seconds)
```

Query 3: Finding top 10 month & year according to rise in stocks price

Command - select month(date) as month ,year(date) as year,
 (max(stocks_price_high) - min(stocks_price_low)) as stocksPriceRise
 from stocks group by month(date), year(date) order by stocksPriceRise
 DESC LIMIT 10;

```
10: jubs: street: 175.0/10 street: 18080/default> select month(date) as month ,year(date) as year, (max(stocks_price_high) - min(stocks_price_low)) as stocksPriceRise from stocks group by month(date), year(date) order by stocksPriceRise DESC LIMIT 10;

INFO : Compiling command(queryId=hive_20180710003131_efdf73b3-2b46-47ff-8cff-525a4040b137): select month(date) as month ,year(date) as year, (max(stocks_price_high) - min(stocks_price_low)) as stocksPriceRise from stocks group by month(date), year(date) order by stocksPriceRise DESC LIMIT 10

INFO : Semantic Analysis Completed

INFO : Returning Hive schema: Schema(fieldSchemas:[FieldSchema(name:month, type:int, comment:null), FieldSchema(name:year, type:int, comment:null)
 INFO : Completed compiling command(queryId=hive_20180710003131_efdf73b3-2b46-47ff-8cff-525a4040b137); Time taken: 0.122 seconds
INFO : Completed compiling command(queryId=hive_20180710003131_efdf73b3-2b46-47ff-8cff-525a4040b137); Time taken: 0.122 seconds
INFO : Executing command(queryId=hive_20180710003131_efdf73b3-2b46-47ff-8cff-525a4040b137): select month(date) as month ,year(date) as year, (max(stocks_price_high) - min(stocks_price_low)) as stocksPriceRise from stocks group by month(date), year(date) order by stocksPriceRise DESC LIMIT 10
INFO : Query ID = hive_20180710003131_efdf73b3-2b46-47ff-8cff-525a4040b137
INFO : Iotal_jobs = 2
INFO : Launching Job 1 out of 2
INFO : Starting task [Stage-1:MAPRED] in serial mode
INFO : Unwher of reduce tasks not specified. Estimated from input data size: 1
INFO : In order to change the average load for a reducer (in bytes):
INFO : set hive.exec.reducers.bytes.per.reducer=<number>
INFO : In order to limit the maximum number of reducers:
                : In order to limit the maximum number of reducers:
                      set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  TNFO
INFO : In order to set a constant number of reducers:
INFO : set mapreduce.job.reduces=<number>
INFO : unber of splits:1
INFO : Submitting tokens for job: job_1531159788380_0010
INFO : Starting Job = job_1531159788380_0010, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0010/
INFO : Starting Job = job_1531159788380_0010, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0010/
INFO : Kill Command = /usr/lib/hadoop/bin/hadoop job -kill job_1531159788380_0010
INFO : 2018-07-10 00:31:31,119 Stage-1 map = 0%, reduce = 0%
INFO : 2018-07-10 00:31:31,119 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 7.48 sec
INFO : 2018-07-10 00:31:49,259 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 9.72 sec
INFO : MapReduce Total cumulative CPU time: 9 seconds 720 msec
INFO : Ended Job = job_1531159788380_0010
INFO : Launching Job 2 out of 2
INFO : set hive.exec.reducers.bytes.per.reducer=<number>
                        set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
                        set hive.exec.reducers.max=<number>
In order to set a constant number of reducers:
  INFO
                      set mapreduce.job.reduces=<number>
  INFO
  INFO
INFO
  INFO
INFO
  INFO
  INFO
  INF0
                 . Ισται παριστατία την 11mm Spent: 12 seconds 600 msec
: Completed executing command(queryId=hive_20180710003131_efdf73b3-2b46-47ff-8cff-525a4040b137); Time taken: 46.419 seconds
: OK
      month | year | stockspricerise
                                 2007
                                                     465.7699890136719
                                                        444.5899963378906
444.5199890136719
       12
                                 2007
                                                        435.6399841308594
430.4900207519531
                                 2007
                                 2008
                                 2007
                                                         429.2699890136719
                                 2007
                                                        420.6000061035156
                                                        420.42999267578125
      10
                                 2007
                                                    | 418.9100036621094
| 415.760009765625
  10 rows selected (46.618 seconds)
```

Query 4: Finding top 10 max difference between stock price close & stock price adjustment close

Command - select price_adj_close-stocks_price_close as adj from
stocks order by adj desc limit 10;

```
0: jdbc:hive2://localhost:10000/default> select price_adj_close-stocks_price_close as adj from stocks order by adj_desc_limit_10;
INFO : Compiling_command(queryId=hive_20180710003838_13317352-ef6d-4663-8703-539d8af33595): select_price_adj_close-stocks_price_close as adj from stocks order by adj_desc_limit_10
INFO : Semantic Analysis Completed
INFO : Returning Hive schema: Schema(fieldSchemas:[FieldSchema(name:adj, type:float, comment:null)], properties:null)
INFO : Completed compiling_command(queryId=hive_20180710003838_13317352-ef6d-4663-8703-539d8af33595); rime_taken: 0.101 seconds
INFO : Executing_command(queryId=hive_20180710003838_13317352-ef6d-4663-8703-539d8af33595): select_price_adj_close-stocks_price_close as adj_from_stocks_order_by_adj_desc_limit_10
INFO : Completed_command(queryId=hive_20180710003838_13317352-ef6d-4663-8703-539d8af33595): select_price_adj_close-stocks_price_close as adj_from_stocks_order_by_adj_desc_limit_10
INFO : Completed_command(queryId=hive_20180710003838_13317352-ef6d-4663-8703-539d8af33595)
er by adj desc timit 10
INFO : Query ID = hive_20180710003838_13317352-ef6d-4663-8703-539d8af33595
INFO : Total jobs = 1
INFO : Launching Job 1 out of 1
INFO : Starting task [Stage-1:MAPRED] in serial mode
INFO : Number of reduce tasks determined at compile time: 1
INFO : In order to change the average load for a reducer (in bytes):
  INFO
                 : set hive.exec.reducers.bytes.per.reducer=<number>
: In order to limit the maximum number of reducers:
                set hive.exec.reducers.max=number>
: In order to set a constant number of reducers:
: set mapreduce.job.reduces=<number>
  INFO
             : set mapreduce.job.reduces=<number>
: number of splits:1
: Submitting tokens for job: job 1531159788380 0013
: The url to track the job: http://quickstart.cloudera:8088/proxy/application 1531159788380 0013/
: Starting Job = job 1531159788380 0013, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0013/
: Kill Command = /usr/lib/hadoop/bin/hadoop job -kill job_1531159788380_0013
: Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1
: 2018-07-10 00:38:24,570 Stage-1 map = 0%, reduce = 0%, cumulative CPU 5.97 sec
: 2018-07-10 00:38:33,894 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 7.66 sec
: MapReduce Total cumulative CPU time: 7 seconds 660 msec
: Ended Job = job 1531159788380 0013
: MapReduce Jobs Eaunched:
: Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 7.66 sec HDFS Read: 40999660 HDFS Write: 81 SUCCESS
 INFO
 INFO
  TNFO
 INFO : Stage-Stage-I Map: 1 Reduce: 1 Cumulative CPU: 7.66 sec HDFS Read: 40999660 HDFS Write: 81 SUCCESS
INFO : Total MapReduce CPU Time Spent: 7 seconds 660 msec
INFO : Completed executing command(queryId=hive_20180710003838_13317352-ef6d-4663-8703-539d8af33595); Time taken: 22.513 seconds
                                                          adj
  1843.7900390625
           1810.360107421875
         1800.4000244140625
         1786.3199462890625
            1780.81005859375
         1779.239990234375
           1779.239990234375
          1769.280029296875
          1769.280029296875
           1766.22998046875
 10 rows selected (22.688 seconds)
```

Query 5: Finding top 10 companies according to volume of stocks

Command - select symbol, max(volume) as MAXVOL from stocks group by
symbol order by MAXVOL desc limit 10;

```
0: jdbc:hive2://localhost:10000/default> select symbol,max(volume) as MAXVOL from stocks group by symbol order by MAXVOL desc limit 10;
INFO : Compiling command(queryId=hive_20180710004040_4340ed8b-1934-4883-b939-35bd2daf7fbf): select symbol,max(volume) as MAXVOL from stocks group by symbol order by MAXVOL desc limit 10
INFO : Compiling Command (queryId=hive_20180710004040_4340ed8b-1934-4883-b939-35bd2daf7fbf): select symbol,max(volume) as MAXVOL from stocks group by symbol order by MAXVOL desc limit 10
           : Semantic Analysis Completed
: Returning Hive schema: Schema(fieldSchemas:[FieldSchema(name:symbol, type:string, comment:null), FieldSchema(name:maxvol, type:int, comment:null)], p
roperties:null)
INFO : Completed compiling command(queryId=hive 20180710004040 4340ed8b-1934-4883-b939-35bd2daf7fbf); Time taken: 0.094 seconds
INFO : Executing command(queryId=hive 20180710004040 4340ed8b-1934-4883-b939-35bd2daf7fbf): select symbol,max(volume) as MAXVOL from stocks group by symbol order by MAXVOL desc limit 10
INFO : Query ID = hive 20180710004040 4340ed8b-1934-4883-b939-35bd2daf7fbf
INFO : Total jobs = 2
INFO : Launching Job 1 out of 2
INFO : Launching Job 1 out of 2
INFO : Starting task [Stage-1:MAPRED] in serial mode
INFO : Number of reduce tasks not specified. Estimated from input data size: 1
INFO : In order to change the average load for a reducer (in bytes):
INFO : set hive.exec.reducers.bytes.per.reducer=renumber>
                 set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
 INFO
                    set hive.exec.reducers.max=<number>
            : In order to set a constant number of reducers:
: set mapreduce.job.reduces=<number>
INFO: st mapreduce.job.reduces<number>
INFO: unmber of splits:1
INFO: unmber of splits:1
INFO: Submitting tokens for job: job_1531159788380_0014
INFO: The url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159788380_0014/
INFO: Starting Job = job_1531159788380_0014/, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0014/
INFO: Starting Job = job_1531159788380_0014/, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0014/
INFO: Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1
INFO: 2018-07-10_00:40:39,328_Stage-1 map = 0%, reduce = 0%, Cumulative CPU 3.52_sec
INFO: 2018-07-10_00:40:40:659_Stage-1 map = 100%, reduce = 0%, Cumulative CPU 5.1_sec
INFO: 2018-07-10_00:40:32_928_Stage-1 map = 100%, reduce = 100%, Cumulative CPU 5.1_sec
INFO: Ended Job = job_1531159788380_0014

INFO: Ended Job = job_1531159788380_0014
  INFO :
                         set hive.exec.reducers.bytes.per.reducer=<number>
  INFO : In order to limit the maximum number of reducers:
                        set hive.exec.reducers.max=<number>
  INFO
  INFO
                : In order to set a constant number of reducers:
  INFO
                        set mapreduce.job.reduces=<number>
  INFO : number of splits:1
  INFO : Submitting tokens for job: job 1531159788380 0015
INFO : The url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159788380_0015/
 INFO : Ine url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159/88380_0015/
INFO : Starting Job = job_1531159788380_0015, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0015/
INFO : Kill Command = /usr/lib/hadoop/bin/hadoop job -kill job_1531159788380_0015
INFO : Hadoop job information for Stage-2: number of mappers: 1; number of reducers: 1
INFO : 2018-07-10 00:40:59,769 Stage-2 map = 0%, reduce = 0%
INFO : 2018-07-10 00:41:106,043 Stage-2 map = 100%, reduce = 0%, Cumulative CPU 1.13 sec
INFO : 2018-07-10 00:41:12,626 Stage-2 map = 100%, reduce = 100%, Cumulative CPU 2.87 sec
  INFO : MapReduce Total cumulative CPU time: 2 seconds 870 msec
  INFO : Ended Job = job_1531159788380_0015
  INFO
                : MapReduce Jobs Launched:
  INFO : Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 5.1 sec HDFS Read: 40999017 HDFS Write: 5259 SUCCESS
INFO : Stage-Stage-2: Map: 1 Reduce: 1 Cumulative CPU: 2.87 sec HDFS Read: 10299 HDFS Write: 135 SUCCESS
  INFO : Total MapReduce CPU Time Spent: 7 seconds 970 msec
  INFO : Completed executing command(queryId=hive 20180710004040 4340ed8b-1934-4883-b939-35bd2daf7fbf); Time taken: 40.643 seconds
  INFO : OK
      symbol | maxvol
      AVP
                           | 318182200
      AA
                               242106500
      AMD
                               163101700
                               148878600
                               112834200
       ABK
      AXL
                               103526300
      AXP
                               90336900
      AMR
                               89947900
      ACE
                               74437100
                           67642400
 10 rows selected (40.809 seconds)
```

Query 6: Finding top 10 times when price adjustment close and price close has positive difference

Command - select count(price_adj_close-stocks_price_close) as adj
from stocks where (price_adj_close-stocks_price_close) > 0 limit 10;

0: jdbc:hive2://localhost:10000/default> select count(price_adj_close-stocks_price_close) as adj from stocks where (price_adj_close-stocks_price_close) !0 li mit 10:

```
INFO : 2018-07-10 01:47:37,214 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 7.98 sec
INFO : MapReduce Total cumulative CPU time: 7 seconds 980 msec
INFO : Ended Job = job_1531165075887_0003
INFO : MapReduce Jobs Launched:
INFO : Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 7.98 sec HDFS Read: 81992182 HDFS Write: 8 SUCCESS
INFO : Total MapReduce CPU Time Spent: 7 seconds 980 msec
INFO : Completed executing command(queryId=hive_20180710014646_62180fb6-28ce-4a2a-b00a-fa1b14032a83); Time taken: 38.261 seconds
INFO : OK
+-----++
| adj | |
| adj | |
| ------++
| 1284850 | |
```

Query 7: Finding top 10 dates according to number of stocks of company AVP

Command - select date, volume from stocks where symbol = "AVP" order
by volume desc limit 10;

```
0: jdbc:hive2://localhost:10000/default> select date, volume from stocks where symbol = "AVP" order by volume desc limit 10;
INFO : Compiling command(queryId=hive_20180710004343_3b51aa20-1328-470e-8995-d9bb487853e8): select date, volume from stocks where symbol = "AVP" order by volume desc limit 10
INFO : Semantic Analysis Completed
INFO : Returning Hive schema: Schema(fieldSchemas:[FieldSchema(name:date, type:string, comment:null), FieldSchema(name:volume, type:int, comment:null)], pro
perties:null)
         : Completed compiling command(queryId=hive 20180710004343 3b51aa20-1328-470e-8995-d9bb487853e8); Time taken: 0.144 seconds
: Executing command(queryId=hive 20180710004343 3b51aa20-1328-470e-8995-d9bb487853e8): select date, volume from stocks where symbol = "AVP" order by vo
INFO : Executing command(queryId=hive_20180710004343_3b51aa20-1328-470e-8991  
INFO : Query ID = hive_20180710004343_3b51aa20-1328-470e-8995-d9bb48785368  
INFO : Total jobs = 1  
INFO : Launching Job 1 out of 1  
INFO : Starting task [Stage-1:MAPRED] in serial mode  
INFO : Number of reduce tasks determined at compile time: 1  
INFO : In order to change the average load for a reducer (in bytes):
           : set hive.exec.reducers.bytes.per.reducer=<number>
: In order to limit the maximum number of reducers:
: set hive.exec.reducers.max=<number>
INFO
          : In order to set a constant number of reducers:
: set mapreduce.job.reduces=<number>
TNFO
INFO
        : number of splits:1
: Submitting tokens for job: job_1531159788380_0016
: The url to track the job: http://quickstart.cloudera:8088/proxy/application_1531159788380_0016/
: Starting Job = job_1531159788380_0016, Tracking URL = http://quickstart.cloudera:8088/proxy/application_1531159788380_0016/
: Kill Command = /usr/lib/hadoop/bin/hadoop job -kill job_1531159788380_0016
: Hadoop job information for Stage-1: number of mappers: 1; number of reducers: 1
: 2018-07-10 00:43:42,249 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 3.74 sec
: 2018-07-10 00:43:45,555 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 5.53 sec
: MapReduce Total cumulative CPU time: 5 seconds 530 msec
: Ended Job = job_1531159788380_0016
: MapReduce_10sb_iaunched:
          : number of splits:1
INFO
INFO
         : Ented Job - Job 1931/9768369 0010
: MapReduce Jobs Launched:
: Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 5.53 sec HDFS Read: 40999895 HDFS Write: 202 SUCCESS
: Total MapReduce CPU Time Spent: 5 seconds 530 msec
: Completed executing command(queryId=hive_20180710004343_3b51aa20-1328-470e-8995-d9bb487853e8); Time taken: 22.903 seconds
INFO
                                              date
                                                               volume
                ------
        1988-05-12 | 318182200
        1988-02-12
                                               165642600
        1991-03-14
                                                 93343200
        1987-11-13
                                                 87786600
        1988-02-08
                                                         82144000
        1989-05-03
                                                         79389800
        1988-02-09
                                                   77893300
        1989-05-10
                                                   57950900
        1989-05-18
                                                   46672000
                                                  38488800
        1989-05-19
         ------
 10 rows selected (23.111 seconds)
```

Step 3 - Creating External table stocks

```
Command -
```

```
create table stocks partition (
exch STRING,
symbol STRING,
date STRING,
stocks price close FLOAT,
stocks price high FLOAT,
stocks price low FLOAT,
stocks price open FLOAT,
volume INT,
price adj close FLOAT)
PARTITIONED BY (exch name STRING, yr STRING, sym STRING)
ROW FORMAT DELIMITED FIELDS TERMINATED BY ',';
0: jdbc:hive2://localhost:10000/default> create table stocks partition(
. . . . . . . . . . . . . . . . . . stocks_price_high FLOAT,
|0: jdbc:hive2://localhost:10000/default> set hive.exec.dynamic.partition = true;
No rows affected (0.013 seconds)
0: jdbc:hive2://localhost:10000/default> set hive.exec.max.dynamic.partitions = 2000;
```

0: jdbc:hive2://localhost:10000/default> set hive.exec.max.dynamic.partitions.pernode = 1000;

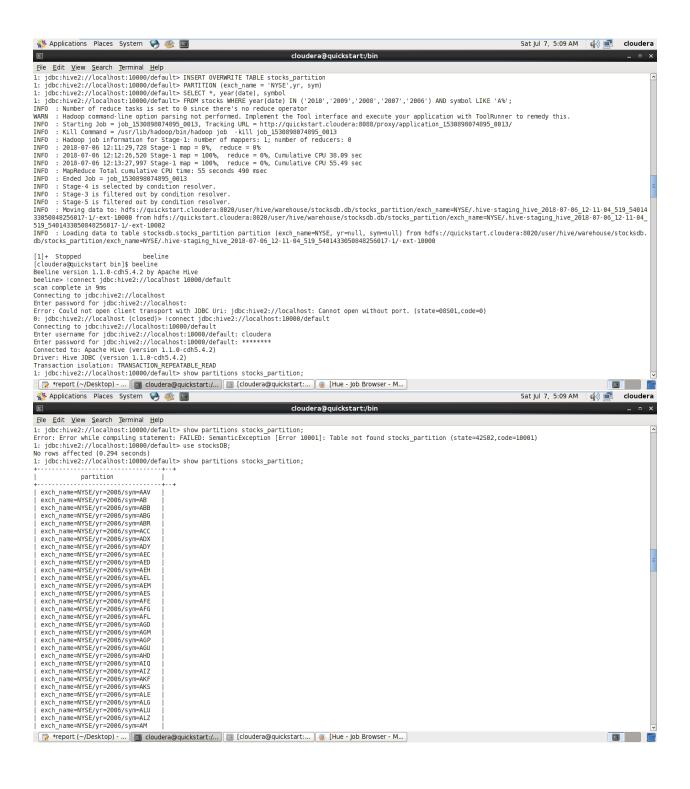
Query 8: Loading data into partitions

No rows affected (0.005 seconds)

No rows affected (0.004 seconds)

Command -

```
INSERT OVERWRITE TABLE stocks_partition
PARTITION (exch_name = 'NYSE',yr, sym)
SELECT *, year(date), symbol
FROM stocks WHERE year(date) IN ('2010','2009','2008','2007','2006')
AND symbol LIKE 'A%';
```



Query 9: Finding maximum stock price in AVP company partition

Command -

```
SELECT MAX(stocks_price_high) as max_high FROM stocks_partition
WHERE sym = 'AVP';
```

Query 10: Finding minimum stock price in AVP company partition

Command -

SELECT MIN(stocks_price_low) as min_low FROM stocks_partition
WHERE sym = 'AVP';

What is Apache Pig?

Apache Pig is an abstraction over MapReduce. It is a tool/platform which is used to analyze larger sets of data representing them as data flows. Pig is generally used with Hadoop; we can perform all the data manipulation operations in Hadoop using Apache Pig.

To write data analysis programs, Pig provides a high-level language known as Pig Latin. This language provides various operators using which programmers can develop their own functions for reading, writing, and processing data.

To analyze data using Apache Pig, programmers need to write scripts using Pig Latin language. All these scripts are internally converted to Map and Reduce tasks. Apache Pig has a component known as Pig Engine that accepts the Pig Latin scripts as input and converts those scripts into MapReduce jobs.

Why Do We Need Apache Pig?

Programmers who are not so good at Java normally used to struggle working with Hadoop, especially while performing any MapReduce tasks. Apache Pig is a boon for all such programmers.

- Using Pig Latin, programmers can perform MapReduce tasks easily without having to type complex codes in Java.
- Apache Pig uses multi-query approach, thereby reducing the length of codes. For example, an operation that would require you to type 200 lines of code (LoC) in Java can be easily done by typing as less as just 10 LoC in Apache Pig. Ultimately Apache Pig reduces the development time by almost 16 times.
- Pig Latin is SQL-like language and it is easy to learn Apache Pig when you are familiar with SQL.
- Apache Pig provides many built-in operators to support data operations like joins, filters, ordering, etc. In addition, it also provides nested data types like tuples, bags, and maps that are missing from MapReduce.

Features of Pig

Apache Pig comes with the following features -

• **Rich set of operators** – It provides many operators to perform operations like join, sort, filer, etc.

- **Ease of programming** Pig Latin is similar to SQL and it is easy to write a Pig script if you are good at SQL.
- **Optimization opportunities** The tasks in Apache Pig optimize their execution automatically, so the programmers need to focus only on semantics of the language.
- **Extensibility** Using the existing operators, users can develop their own functions to read, process, and write data.
- UDF's Pig provides the facility to create User-defined Functions in other programming languages such as Java and invoke or embed them in Pig Scripts.
- Handles all kinds of data Apache Pig analyzes all kinds of data, both structured as well as unstructured. It stores the results in HDFS.

Apache Pig Vs Hive

Both Apache Pig and Hive are used to create MapReduce jobs. And in some cases, Hive operates on HDFS in a similar way Apache Pig does. In the following table, we have listed a few significant points that set Apache Pig apart from Hive.

Apache Pig	Hive
Apache Pig uses a language called Pig Latin . It was originally created at Yahoo .	Hive uses a language called HiveQL . It was originally created at Facebook .
Pig Latin is a data flow language.	HiveQL is a query processing language.
Pig Latin is a procedural language and it fits in pipeline paradigm.	HiveQL is a declarative language.
Apache Pig can handle structured, unstructured, and semi-structured data.	Hive is mostly for structured data.

Pig Analytics tasks

Dataset used: NYSE_stocks.csv, NYSE_dividends.csv

Step 1: Opening Pig in Local mode

Step 2: Loading Data from NYSE_stocks.csv dataset into Pig

```
grunt> stocks = LOAD '/home/cloudera/NYSE_stocks.csv' USING PigStorage(',') as (Exch: CHARARRAY, Symbol:CHARARRAY, Date:DATETIME, Stocks_Open:FLOAT, Stocks_H OAT)
>> ;
```

Step 3: Loading Data from NYSE_dividends.csv dataset into Pig

```
grunt> dividends = LOAD '/home/cloudera/NYSE_dividends.csv' USING PigStorage(',') AS (EXCH: CHARARRAY,Symbol:CHARARRAY,Date:DATETIME,Dividents:FLOAT);
```

```
Step 4: Grouping Stocks Data by symbol
```

```
grunt> grp_sym = GROUP stocks BY Symbol;
```

Step 5: Generating Percentage return per stock

grunt> return = FOREACH grp_sym GENERATE group, MAX(stocks.Stocks_Close)-MIN(stocks.Stocks_Open) / MIN(stocks.Stocks_Open) * 100 as PCT_RETURN;

```
Step 6: Ordering the generated data
```

```
grunt> order_return = ORDER return BY PCT_RETURN DESC;
```

Step 7: Putting limit to number of generated outputs to show

```
grunt> limit_return = LIMIT order_return 10;
```

Step 8: Showing output

```
grunt> dump limt_return;
```

Step 9: Joining the two datasets

```
grunt> joining = JOIN stocks BY (Symbol,Date), dividends BY (Symbol,Date);
2018-07-09 01:55:41,864 [main] WARN org.apache.pig.PigServer - Encountered Warning IMPLICIT_CAST_TO_FLOAT 1 time(s).
grunt> DESCRIBE joining;
joining: {stocks::Exch: chararray,stocks::Symbol: chararray,stocks::Date: datetime,stocks::Stocks_Open: float,stocks::Stocks_High: float,dividends::Exch: chararray,dividends::Stocks_Upen: float}
```

What are the challenges of using Hadoop?

- MapReduce programming is not a good match for all problems. It's good for simple information requests and problems that can be divided into independent units, but it's not efficient for iterative and interactive analytic tasks. MapReduce is file-intensive. Because the nodes don't intercommunicate except through sorts and shuffles, iterative algorithms require multiple map-shuffle/sort-reduce phases to complete. This creates multiple files between MapReduce phases and is inefficient for advanced analytic computing.
- There's a widely acknowledged talent gap. It can be difficult to find entry-level programmers who have sufficient Java skills to be productive with MapReduce. That's one reason distribution providers are racing to put relational (SQL) technology on top of Hadoop. It is much easier to find programmers with SQL skills than MapReduce skills. And, Hadoop administration seems part art and part science, requiring low-level knowledge of operating systems, hardware and Hadoop kernel settings.
- Data security. Another challenge centers around the fragmented data security issues, though new tools and technologies are surfacing. The Kerberos authentication protocol is a great step toward making Hadoop environments secure.
- Full-fledged data management and governance. Hadoop does not have easy-to-use, full-feature tools for data management, data cleansing, governance and metadata. Especially lacking are tools for data quality and standardization.

Apache Spark

Apache Spark is a lightning-fast cluster computing technology, designed for fast computation. It is based on Hadoop MapReduce and it extends the MapReduce model to efficiently use it for more types of computations, which includes interactive queries and stream processing. The main feature of Spark is its in-

memory cluster computing that increases the processing speed of an application.

Spark is designed to cover a wide range of workloads such as batch applications, iterative algorithms, interactive queries and streaming. Apart from supporting all these workload in a respective system, it reduces the management burden of maintaining separate tools.

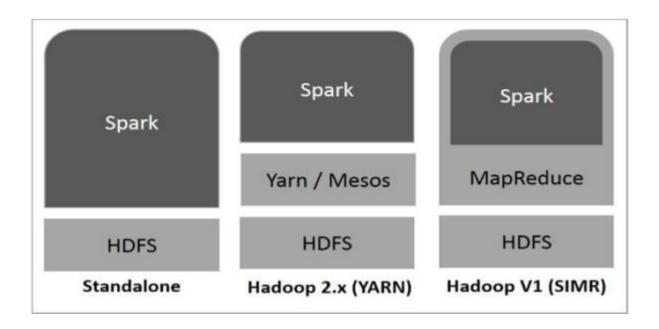
Features of Apache Spark

Apache Spark has following features.

- Speed Spark helps to run an application in Hadoop cluster, up to 100 times faster in memory, and 10 times faster when running on disk. This is possible by reducing number of read/write operations to disk. It stores the intermediate processing data in memory.
- Supports multiple languages Spark provides built-in APIs in Java, Scala, or Python. Therefore, you can write applications in different languages. Spark comes up with 80 high-level operators for interactive querying.
- Advanced Analytics Spark not only supports 'Map' and 'reduce'. It also supports SQL queries, Streaming data, Machine learning (ML), and Graph algorithms.

Spark Built on Hadoop

The following diagram shows three ways of how Spark can be built with Hadoop components.



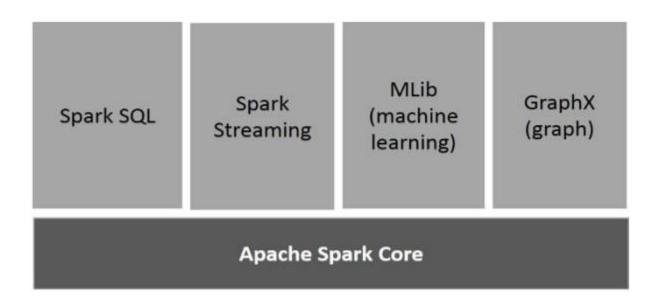
Standalone – Spark Standalone deployment means Spark occupies the place on top of HDFS (Hadoop Distributed File System) and space is allocated for HDFS, explicitly. Here, Spark and MapReduce will run side by side to cover all spark jobs on cluster.

Hadoop Yarn – Hadoop Yarn deployment means, simply, spark runs on Yarn without any pre-installation or root access required. It helps to integrate Spark into Hadoop ecosystem or Hadoop stack. It allows other components to run on top of stack.

Spark in MapReduce (SIMR) – Spark in MapReduce is used to launch spark job in addition to standalone deployment. With SIMR, user can start Spark and uses its shell without any administrative access.

Components of Spark

The following illustration depicts the different components of Spark.



Apache Spark Core

Spark Core is the underlying general execution engine for spark platform that all other functionality is built upon. It provides In-Memory computing and referencing datasets in external storage systems.

Spark SQL

Spark SQL is a component on top of Spark Core that introduces a new data abstraction called SchemaRDD, which provides support for structured and semi-structured data.

Spark Streaming

Spark Streaming leverages Spark Core's fast scheduling capability to perform streaming analytics. It ingests data in mini-batches and performs RDD (Resilient Distributed Datasets) transformations on those mini-batches of data.

MLlib (Machine Learning Library)

MLlib is a distributed machine learning framework above Spark because of the distributed memory-based Spark architecture. It is, according to benchmarks, done by the MLlib developers against the Alternating Least Squares (ALS) implementations. Spark MLlib is nine times as fast as the Hadoop disk-based version of Apache Mahout (before Mahout gained a Spark interface).

GraphX

GraphX is a distributed graph-processing framework on top of Spark. It provides an API for expressing graph computation that can model the user-defined graphs by using Pregel abstraction API. It also provides an optimized runtime for this abstraction.

Resilient Distributed Datasets (RDD)

Resilient Distributed Datasets (RDD) is a fundamental data structure of Spark. It is an immutable distributed collection of objects. Each dataset in RDD is divided into logical partitions, which may be computed on different nodes of the cluster. RDDs can contain any type of Python, Java, or Scala objects, including user-defined classes.

Formally, an RDD is a read-only, partitioned collection of records. RDDs can be created through deterministic operations on either data on stable storage or other RDDs. RDD is a fault-tolerant collection of elements that can be operated on in parallel.

There are two ways to create RDDs – parallelizing an existing collection in your driver program, or referencing a dataset in an external storage system, such as a shared file system, HDFS, HBase, or any data source offering a Hadoop Input Format.

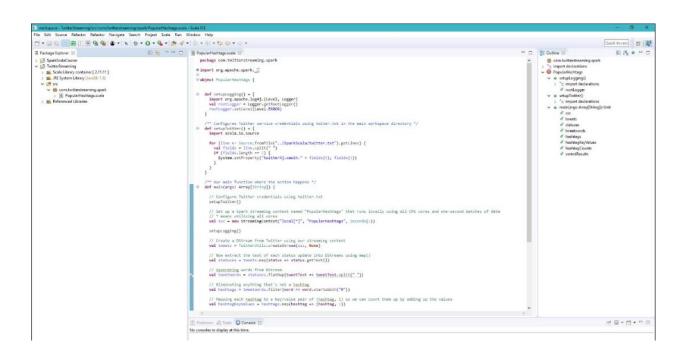
Spark makes use of the concept of RDD to achieve faster and efficient MapReduce operations. Let us first discuss how MapReduce operations take place and why they are not so efficient.

Twitter's Top 10 popular Hashtag Streaming per second using Apache Spark

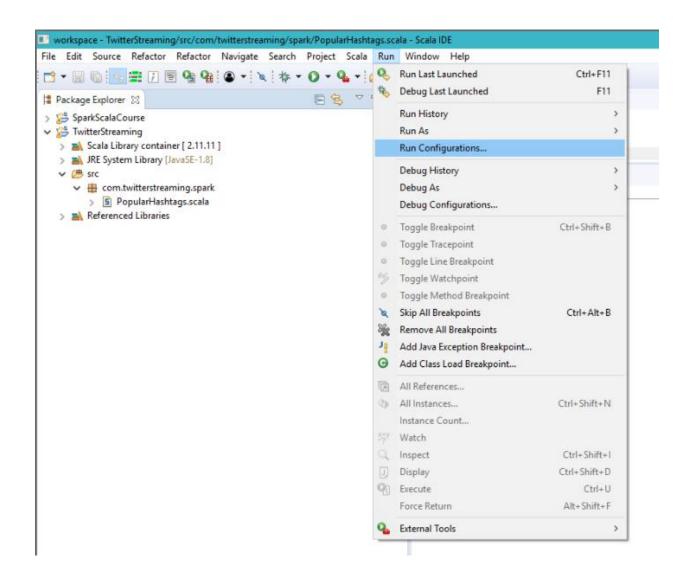
Technologies used -

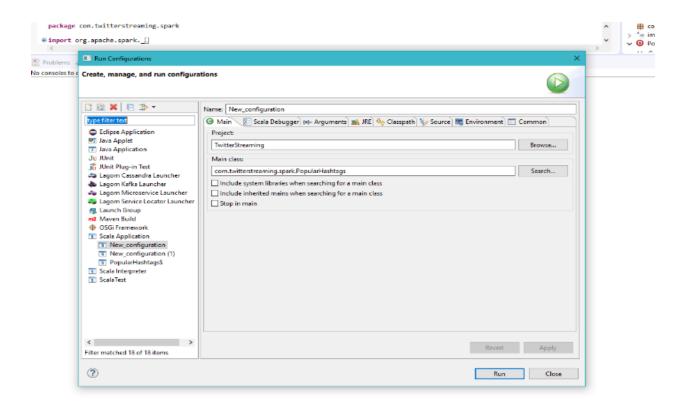
- Spark
- Scala 2.11.11
- Libraries used: dsstream-twitter_2.11-0.1.0-SNAPSHOT, twitter4j-core-4.0.4, twitter4j-stream-4.0.4
- Eclipse Scala IDE

Scala Code:



Running Script:





Output:

```
New configuration [Scale Application] COProgram Files/Davaydet/B_017N/bin/phow.ee (10-bil-2018, 225-46 AM)
Using Spark's default logid profile: erg/apach/Spark/Logid-defaults.properties
12/07/18 02:25:47 JNRO SparkContect: Bunning Spark version 1.0.0
12/07/19 02:25:47 JNRO SecurityPhonager: Changing view acls to: Mayank
12/07/19 02:25:47 JNRO SecurityPhonager: Changing view acls groups to:
12/07/19 02:25:47 JNRO SecurityPhonager: Changing view acls groups to:
12/07/19 02:25:47 JNRO SecurityPhonager: Changing view acls groups to:
12/07/19 02:25:48 JNRO Sparkfor: Registering application of the securityPhonager: Changing view acls groups to:
12/07/19 02:25:48 JNRO Sparkfor: Registering BlockforangerHaster
12/07/19 02:25:48 JNRO Sparkfor: Registering BlockforangerHaster
12/07/19 02:25:48 JNRO Sparkfor: Registering BlockforangerHaster
12/07/19 02:25:48 JNRO Sparkfor: Registering OptotyPutTracker
12/07/19 02:25:48
```

```
(#PachaMama,1)
(#happygrandpa,1)
(#weareacmilan,1)
Time: 1531169768000 ms
(#Brexit!,1)
(#ForzaMilan,1)
(#deixaelastrabalhar,1)
(#LunesDeGanarSeguidores,1)
(#Ty,1)
(#Dominos,1)
(طِفْ_الورد_لدم, 1_#)
(#PachaMama,1)
(#happygrandpa,1)
(#weareacmilan,1)
Time: 1531169769000 ms
(#MasterChef,2)
(#Brexit!,1)
(#ForzaMilan,1)
(#deixaelastrabalhar,1)
(#LunesDeGanarSeguidores,1)
(#Ty,1)
(#Dominos,1)
(طف_الورد_للدعمر1_#)
(#PachaMama,1)
(#happygrandpa,1)
Time: 1531169770000 ms
(#MasterChef,2)
(#Brexit!,1)
(#ForzaMilan,1)
(#deixaelastrabalhar,1)
(#9999,1)
(#LunesDeGanarSeguidores,1)
(#Ty,1)
(#Dominos,1)
```

Twitter.txt file:

```
File Edit Format View Help

consumerKey 061iS5x7CNLBUNINGTT, j.T.v

consumerSecret iaWAan80KC4CgFFgF UM02a-26bc8VWbwerFYVDIKAyCfovs8Y
accessToken 500050250 55cfoF19 H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H0004-H00
```