University Of Mumbai Institute of Distance & Open Learning



PRACTICAL JOURNAL BIG DATA ANALYTICS

SUBMITTED BY

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MASTER OF SCIENCE IN INFORMATION TECHNOLOGY PART-I SEMESTER II

ACADEMIC YEAR 2024-2025

INSTITUTE OF DISTANCE AND OPEN LEARNING IDOL BUILDING, VIDYANAGARI, SANTACRUZ (EAST), MUMBAI-400 098

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Certificate

This is to certify that

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Practical 1

Install, configure and run Hadoop and HDFS ad explore HDFS.

Step 1 Install Hadoop

Login to ubuntu

Apply following commands from ubuntu terminal

\$ sudo apt update

\$ sudo apt install default-jdk

\$ ava -version'

\$ wget https://hadoop.apache.org/release/3.2.2.html/hadoop-3.2.2.tar.gz

\$ tar xzvf hadoop-3.2.2.tar.gz

\$ sudo mv hadoop-3.2.2 /usr/local/hadoop

\$ readlink -f /usr/bin/java | sed "s:bin/java::"

#: Configuring Hadoop's Java Home; To begin, open hadoop-env.sh

\$ sudo nano /usr/local/hadoop/etc/hadoop/hadoop-env.sh

File will be opened. Add the following line at the end of .sh file

export JAVA HOME=/usr/lib/jvm/java-11-openjdk-amd64/

to save the changes in the file, press ctrl and x together.

then press Y

then press Enter key

then apply following commands:

\$ /usr/lib/jvm/java-11-openjdk-amd64/

Steps to get hdfs file system working:

1.In this section, we'll go over basic interactions with the distributed file system via the command line. It is assumed that these commands are performed on a client that can connect to a remote Hadoop cluster, or which is running a pseudo-distributed clusteron the localhost. It is also assumed that the hadoop command and other utilities from \$HADOOP HOME/bin are on the system path and can be found by the operating system

All of the usual file system operations are available to the user, such as creating directories; moving, removing, and copying files; listing directories; and modifying permissions of files on the cluster. To see the available commands in the fs shell, type:

hostname \$ hadoop fs -help

Usage: hadoop fs [generic options]

/usr/local/hadoop/bin/hadoop fs -help

As you can see, many of the familiar commands for interacting with the file system are there, specified as arguments to the hadoop fs command as flag arguments in the Java style—that is, as a single dash (—) supplied to the command. Secondary flags or options to the command are specified with additional Java style arguments delimited by spaces following the initial command. Be aware that order can matter when specifying such options.

To get started, let's copy some data from the local file system to the remote (dis-tributed) file system. To do this, use either the put or **copyFromLocal** commands. These commands are identical and write files to the distributed file system without removing the local copy. The **moveFromLocal** command is similar, but the local copy is deleted after a successful transfer to the distributed file system. In the /data directory of the GitHub repository for this book's code and resources, there is a shakespeare.txt file containing the complete works of William Shakespeare.

Download this file to your local working directory. After download, move the file to the distributed file system as follows:

hostname \$ hadoop fs -copyFromLocal shakespeare.txt shakespeare.txt

This example invokes the Hadoop shell command copyFromLocal with two arguments, <src> and <dst>, both of which are specified as relative paths to a file called shakespeare.txt. To be explicit about what's happening, the command searches your current working directory for the shakespeare.txt file and copies it to the /user/analyst/shakespeare.txt path on HDFS by first requesting information about that path from the NameNode, then directly communicating with the DataNodes to transfer the file. Because Shakespeare's complete works are less than 64 MB, it is not broken up into blocks. Note, however, that on both your local machine, as well as the remote HDFS system, relative and absolute paths must be taken into account. The preceding command is shorthand for:

hostname \$ hadoop fs -put /home/analyst/shakespeare.txt \

hdfs://localhost/user/analyst/shakespeare.txt

You'll note that there exists a home directory on HDFS that is similar to the home directory on POSIX systems; this is what the /user/analyst/ directory is—the home directory of the analyst user. Relative paths in reference to the remote file system treat the user's HDFS home directory as the current working directory. In fact, HDFS has a permissions model for files and directories that are very similar to POSIX. In order to better manage the HDFS file system, create a hierarchical tree of directories just as you would on your local file system:

hostname \$ hadoop fs -mkdir corpora

To list the contents of the remote home directory, use the ls command:

hostname \$ hadoop fs -ls.

drwxr-xr-x - analyst analyst 0 2015-05-04 17:58 corpora

-rw-r--r-- 3 analyst analyst 8877968 2015-05-04 17:52 shakespeare.txt

The HDFS file listing command is similar to the Unix ls –l command with some HDFS-specific features. Specified without any arguments, this command provides a listing of the user's HDFS home directory. The first column shows the permissions mode of the file. The second column is the replication of the file; by default, the replication is 3. Note that directories are not replicated, so this column is a dash (-) in that case. The user and group follow, then the size of the file in bytes (zero for directories). The last modified date and time is up next, with the name of the file appearing last. Other basic file operations like mv, cp, and rm will all work as expected on the remote file system. There is, however, no rmdir command; instead, use rm –R to recursively remove a directory with all files in it. Reading and moving files from the distributed file system to the local file system should be attempted with care, as the distributed file system is maintaining files that are extremely large. However, there are cases when files need to be inspected in detail by the user, particularly output files that are produced as the result of MapReduce jobs. Typically these are not read to the standard output stream but are piped to other programs like less or more. To read the contents of a file, use the cat command, then pipe the output to less in order view the contents of the remote file:

hostname \$ hadoop fs -cat shakespeare.txt | less

There is no similar hadoop fs -head command to inspect the first kilobyte of the file. Instead, it is efficient to hadoop fs -cat the file and pipe it to the local shell's head, as the head command terminates the remote stream before the entire file is read. However, using the shell's tail in this manner would be less efficient, as all of the data would have to be streamed from the remote file system to your local file system before the output could be computed. Instead the, hadoop fs —tail command seeks to the correct position in the remote file and returns only the required data over the network. To transfer entire files from the distributed file system to the local file system, use get or copyToLocal, which are identical commands. Similarly, use the moveToLocal command, which also removes the file from the distributed file system. Finally, the get merge command merges all files that match a given pattern or directory are copied and merged into a single file on the localhost. If files on the remote system are large, you may want to pipe them to a compression utility:

hostname \$ hadoop fs -get shakespeare.txt ./shakespeare.from-remote.txt

Comparing the original shakespeare.txt file should prove that it is identical to the shakespeare.from-remote.txt file. Hopefully we have demonstrated that the hadoop fs command is a fully featured command-line interface to HDFS and is used routinely when developing analytical jobs. Table 2-1 demonstrates other useful commands that are provided by hadoop fs.

hadoop fs -help <cmd> Provides information and flags specifically about the <cmd> in question.

hadoop fs -test <path> Answer various questions about <path> (e.g., exists, is directory, is file, etc.)

hadoop fs -count <path> Count the number of directories, files, and bytes under the paths that match the

specified file pattern.

hadoop fs -du -h <path> Show the amount of space, in bytes, used by the files that match the specified file pattern.

hadoop fs -stat <path> Print statistics about the file/directory at <path>.

hadoop fs -text <path> Takes a source file and outputs the file in text format. Currently Zip,

TextRecordInputStream, and Avro sources are supported

File Permissions in HDFS

As mentioned earlier, HDFS has POSIX-like file permissions. There are three types of permissions: read (r), write (w), and execute (x). These permissions define the access levels for the owner, the group, and any other system users. For directories, the exe- cute permission allows access to the contents of the directory; however, execute per-missions are ignored on HDFS for files.

Read and write permissions in the context of HDFS specify who can access the data and who can append to the file. Permissions are expressed during the directory listing command ls. Each mode has 10 slots. The first slot is a d for directories, otherwise a - for files. Each of the follow- ing groups of three indicates the rwx permissions for the owner, group, and other users, respectively.

There are several HDFS shell commands that will allow you to manage the permissions of files and directories, namely the familiar chmod, chgrp, and chown commands: hostname \$ hadoop fs —chmod 664 shakespeare.txt This command changes the permissions of shakespeare.txt to -rw-rw-r--. The 664 is an octal representation of the flags to set for the permission triple. Consider 6 in binary, 110—this means set the read and write flags but not the execute flag. Com- pletely permissible is 7, 111 in binary and read-only is 4, 100 in binary. The chgrp and chown commands change the group and owner of the files on the distributed file system.

A caveat with file permissions on HDFS: the identity of the client is determined by the username and groups of the process operating across HDFS, which means remote clients can create arbitrary users on the system. These permissions, therefore, should only be used to prevent accidental data loss and to share file system resources between known users, not as a security mechanism

Other HDFS Interfaces

Programmatic access to HDFS is made available to software developers through a Java API, and any serious data ingestion into a Hadoop cluster should consider utiliz- ing that API. There are also other tools for integrating HDFS with other file systems or network protocols—for example, FTP or Amazon S3. In Chapter 6, we'll focus more on data management issues and how to acquire and store data from a variety of sources into HDFS.

There are also HTTP interfaces to HDFS, which can be used for routine administra- tion of the cluster file system and programmatic access to HDFS with Python. HTTP access to HDFS comes in two primary interfaces: direct access through the HDFS daemons that serve HTTP requests, or via proxy servers that expose an HTTP inter- face then directly access HDFS on the client's behalf using the Java API.

Examples of proxies include HttpFS, Hoop, and WebHDFS, each of which allow RESTful network access to the Hadoop cluster, which is easily programmed against using Python. The NameNode also supplies direct, read-only access to HDFS over HTTP through an HTTP server that runs on port 50070. If running in pseudo-

distributed mode, simply open a browser and navigate to http://127.0.0.1:50070; otherwise, use the host name of the NameNode on your cluster. The NameNode Web UI provides a high

level overview of the cluster status, including the amount of storage available and used, the number of alive and dead DataNodes, and warning information about under-replicated blocks. The NameNode also allows users to browse the file system using a search and naviga- tion utility that is found under the Utilities drop-down tab. File meta information is listed similar to the command-line interface, and specific files may even be made available for download. DataNodes themselves can be directly browsed for informa- tion, accessing the DataNode host on port 50075; all active DataNodes are listed on the NameNode HTTP site.

By default, the direct HTTP interface is read-only. In order to provide write access to an HDFS cluster, a proxy such as WebHDFS must be used. WebHDFS secures the cluster via authentication with Kerberos. Accessing secure resources on Hadoop depends primarily on the specific configuration of the cluster, and if any third-party management tools are being used. Hadoop was designed to run on completely man- aged internal clusters without exposure to the outside world, and as a result security in Hadoop is not as mature as the platform itself—although this is one of the prime considerations of Hadoop development moving forward

Practical 2

Implement word count / frequency programs using MapReduce Map Reduce as two component Map and Reduce.

Java program:

```
Write program save as WordCount.java
```

```
import java.io.IOException;
import java.util.StringTokenizer;
import org.apache.hadoop.conf.Configuration;
import org.apache.hadoop.fs.Path;
import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.mapreduce.Job;
import org.apache.hadoop.mapreduce.Mapper;
import org.apache.hadoop.mapreduce.Reducer;
import org.apache.hadoop.mapreduce.lib.input.FileInputFormat;
import org.apache.hadoop.mapreduce.lib.output.FileOutputFormat;
public class WordCount {
public static class TokenizerMapper extends Mapper<Object, Text, Text, IntWritable>{
private final static IntWritable one = new IntWritable(1);
private Text word = new Text();
public void map(Object key, Text value, Context context
) throws IOException, InterruptedException {
StringTokenizer itr = new StringTokenizer(value.toString());
while (itr.hasMoreTokens()) {//"This is the output is the"
word.set(itr.nextToken());
context.write(word, one);
public static class IntSumReducer extends Reducer<Text,IntWritable,Text,IntWritable> {
private IntWritable result = new IntWritable();
public void reduce(Text key, Iterable<IntWritable> values, Context context) throws IOException,
InterruptedException
{//is.3}
int sum = 0;
for (IntWritable val : values) {
sum += val.get();
result.set(sum);
context.write(key, result);
}
}
public static void main(String[] args) throws Exception {
```

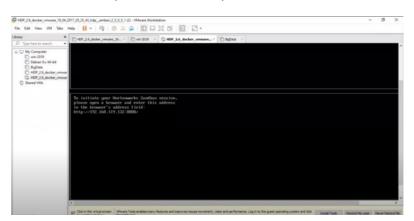
```
Configuration conf = new Configuration();
Job job = Job.getInstance(conf, "word count");
job.setJarByClass(WordCount.class);
job.setMapperClass(TokenizerMapper.class);
job.setCombinerClass(IntSumReducer.class);
job.setReducerClass(IntSumReducer.class);
job.setOutputKeyClass(Text.class);
job.setOutputValueClass(IntWritable.class);
FileInputFormat.addInputPath(job, new Path(args[0]));
FileOutputFormat.setOutputPath(job, new Path(args[1]));
System.exit(job.waitForCompletion(true)?0:1);
}
```

Text File:

Hello World

This is the output is the

Start the server (Horton Sandbox)



Open the terminal with 192.168.119.132/4200

Enter the login: root and the password and enter



Create a folder in local directory.

Command: mkdir mscitp2

Change the directory cd mscitp2

```
sandbox login: root
root@sandbox.hortonworks.com's password:
Last login: Thu Jul 1 13:38:19 2021 from 172.17.0.2
[root@sandbox ~]# mkdir mscitp2
[root@sandbox ~]# cd mscitp2
[root@sandbox mscitp2]# ls
[root@sandbox mscitp2]# |
```

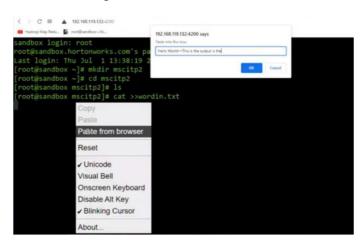
Now create input file

Command: cat >> wordin.txt

Paste the text by right clicking on terminal

Hello World

This is the output is the



```
Wedoop Map Reduce To root@sandbox.vbi.

sandbox login: root
root@sandbox.hortonworks.com's password:
Last login: Thu Jul 1 13:38:19 2021 from 172.17.0.2
[root@sandbox ~]# mkdir mscitp2
[root@sandbox ~]# cd mscitp2
[root@sandbox mscitp2]# ls
[root@sandbox mscitp2]# cat >>wordin.txt
Hello World

This is the output is the
```

To remove the extra space type command

vi wordin.txt

After removing the extra space check the content of the file

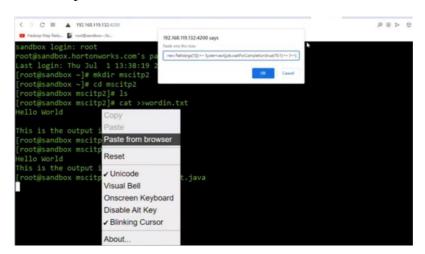
cat wordin.txt

```
| Theorem | Market |
```

Create another file wordcount.java

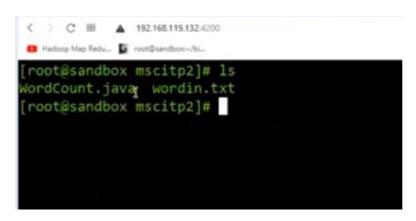
```
This is the output is the 
[root@sandbox mscitp2]# vi wordin.txt
[root@sandbox mscitp2]# cat wordin.txt
Hello World
This is the output is the
[root@sandbox mscitp2]# cat >>\TordCount.java
```

Paste the java code.



Press control d to save the file

Check both the files create with command ls



Now, to compile the java file

export HADOOP_CLASSPATH=\$(hadoop classpath)
mkdir classes (To keep the compile files)
javac -classpath \${HADOOP_CLASSPATH} -d classes WordCount.java

Check class files are created with command ls classes

```
WordCount.java wordin.txt
[root@sandbox mscitp2]# mkdir classes
[root@sandbox mscitp2]# javac -classpath ${HADOOP_CLASSPATH} -d classes WordCount.java
[root@sandbox mscitp2]# ls classes
WordCount.class WdrdCount$IntSumReducer.class WordCount$TokenizerMapper.class
[root@sandbox mscitp2]# []
```

Now we have to bind all the class into single jar file with below command

jar -cvf WordCount.jar -C classes/.

Run ls command we can see jar file is created.

```
[root@sandbox mscitp2]# ls
classes WordCount.jar WordCount.java wordin.txt
[root@sandbox mscitp2]# | ],
```

wordin.txt should be present in word directory of hdfs. So we need to upload wordin.txt file.

```
C 88 A 192.166.119.1324200

**Prot@sandbox mscitp2]# hdfs dfs -mkdir /p2
[root@sandbox mscitp2]# ls

classes WordCount.jar WordCount.java wordin.txt
[root@sandbox mscitp2]# hdfs dfs -put wordin.txt /p2
[root@sandbox mscitp2]# hdfs dfs -ls /p2
Found 1 items
-rw-r--r-- 1 root hdfs 38 2021-07-01 15:07 /p2/werdin.txt
[root@sandbox mscitp2]# |||
```

We need to put the final output p2output.

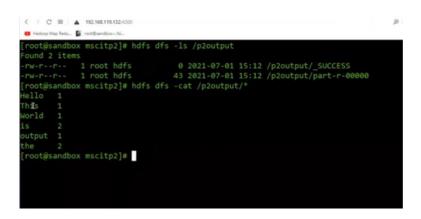
hadoop jar WordCount.jar WordCount /p2//p2output

```
Protop Models. B outdowninco.ds.

[root@sandbox mscitp2]# hadoop jar WordCount.jar WordCount /p2/ /p2output
21/87/01 15:11:44 INFO client.RMProxy: Connecting to ResourceManager at sandbox.hortonworks.com/
172.17.0.2:8032
21/07/01 15:11:44 INFO client.AHSProxy: Connecting to Application History server at sandbox.hortonworks.com/
172.17.0.2:10200
21/07/01 15:11:45 WARN mapreduce.JobResourceUploader: Hadoop command-line option parsing not per
formed. Implement the Tool interface and execute your application with ToolRunner to remedy this
.
.
21/07/01 15:11:46 INFO input.FileInputFormat: Total input paths to process: 1
21/07/01 15:11:47 INFO mapreduce.JobSubmitter: number of splits:1
21/07/01 15:11:47 INFO mapreduce.JobSubmitter: Submitting tokens for job: job_1625146513303_0004
21/07/01 15:11:48 INFO mapreduce.Job: The url to track the job: http://sandbox.hortonworks.com:8
088/proxy/application_1625146513303_0004
21/07/01 15:11:48 INFO mapreduce.Job: Running job: job_1625146513303_0004
```

Print the content of the output file

Command: hdfs dfs -cat /p2ouput/*



Ctrl + 1 to clear the screen.

vi filename.txt= this command will create/ open filename.txt

two modes of vi editor

- 1) Insert mode i (press i key)
- 2) Command mode esc key

: wq is to save and exit

Practical 3

Implement an MapReduce program that processes a weather dataset.

```
Java program:
MyMaxMin.java
// importing Libraries
import java.io.IOException;
import java.util.Iterator;
import org.apache.hadoop.fs.Path;
import org.apache.hadoop.io.LongWritable;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.mapreduce.lib.input.FileInputFormat;
import org.apache.hadoop.mapreduce.lib.output.FileOutputFormat;
import org.apache.hadoop.mapreduce.lib.output.TextOutputFormat;
import org.apache.hadoop.mapreduce.lib.input.TextInputFormat;
import org.apache.hadoop.mapreduce.Job;
import org.apache.hadoop.mapreduce.Mapper;
import org.apache.hadoop.mapreduce.Reducer;
import org.apache.hadoop.conf.Configuration;
public class MyMaxMin {
// Mapper
       /*MaxTemperatureMapper class is static
       * and extends Mapper abstract class
       * having four Hadoop generics type
       * LongWritable, Text, Text, Text.
       */
       public static class MaxTemperatureMapper extends
                     Mapper<LongWritable, Text, Text, Text> {
```

```
/**
       * @method map
       * This method takes the input as a text data type.
       * Now leaving the first five tokens, it takes
       * 6th token is taken as temp max and
       * 7th token is taken as temp min. Now
       * temp \max > 30 and temp \min < 15 are
       * passed to the reducer.
       */
// the data in our data set with
// this value is inconsistent data
public static final int MISSING = 9999;
@Override
       public void map(LongWritable arg0, Text Value, Context context)
                      throws IOException, InterruptedException {
       // Convert the single row(Record) to
       // String and store it in String
       // variable name line
       String line = Value.toString();
               // Check for the empty line
               if (!(line.length() == 0)) {
                      // from character 6 to 14 we have
                      // the date in our dataset
                      String date = line.substring(6, 14);
                      // similarly we have taken the maximum
                      // temperature from 39 to 45 characters
                      float temp Max = Float.parseFloat(line.substring(39, 45).trim());
```

```
// similarly we have taken the minimum
                             // temperature from 47 to 53 characters
                             float temp Min = Float.parseFloat(line.substring(47, 53).trim());
                             // if maximum temperature is
                             // greater than 30, it is a hot day
                             if (temp Max > 30.0) {
                                    // Hot day
                                    context.write(new Text("The Day is Hot Day:" + date),
                                    new Text(String.valueOf(temp Max)));
                             }
                             // if the minimum temperature is
                             // less than 15, it is a cold day
                             if (temp Min < 15) {
                                    // Cold day
                                    context.write(new Text("The Day is Cold Day:" + date),
                                                   new Text(String.valueOf(temp Min)));
                             }
                      }
       }
// Reducer
       /*MaxTemperatureReducer class is static
       and extends Reducer abstract class
       having four Hadoop generics type
       Text, Text, Text, Text.
       */
       //The Day is Cold Day :20150101 ,-21.8
```

```
public static class MaxTemperatureReducer extends
              Reducer<Text, Text, Text, Text> {
       /**
       * @method reduce
       * This method takes the input as key and
       * list of values pair from the mapper,
       * it does aggregation based on keys and
       * produces the final context.
       */
       public void reduce(Text Key, Iterator<Text> Values, Context context)
                      throws IOException, InterruptedException {
              // putting all the values in
              // temperature variable of type String
              String temperature = Values.next().toString();
              context.write(Key, new Text(temperature));
       }
}
* @method main
* This method is used for setting
* all the configuration properties.
* It acts as a driver for map-reduce
* code.
*/
public static void main(String[] args) throws Exception {
       // reads the default configuration of the
       // cluster from the configuration XML files
       Configuration conf = new Configuration();
```

```
// Initializing the job with the
// default configuration of the cluster
Job job = new Job(conf, "weather example");
// Assigning the driver class name
job.setJarByClass(MyMaxMin.class);
// Key type coming out of mapper
job.setMapOutputKeyClass(Text.class);
// value type coming out of mapper
job.setMapOutputValueClass(Text.class);
// Defining the mapper class name
job.setMapperClass(MaxTemperatureMapper.class);
// Defining the reducer class name
job.setReducerClass(MaxTemperatureReducer.class);
// Defining input Format class which is
// responsible to parse the dataset
// into a key value pair
job.setInputFormatClass(TextInputFormat.class);
// Defining output Format class which is
// responsible to parse the dataset
// into a key value pair
job.setOutputFormatClass(TextOutputFormat.class);
// setting the second argument
// as a path in a path variable
Path OutputPath = new Path(args[1]);
// Configuring the input path
// from the filesystem into the job
FileInputFormat.addInputPath(job, new Path(args[0]));
```

```
// Configuring the output path from

// the filesystem into the job

FileOutputFormat.setOutputPath(job, new Path(args[1]));

// deleting the context path automatically

// from hdfs so that we don't have

// to delete it explicitly

OutputPath.getFileSystem(conf).delete(OutputPath);

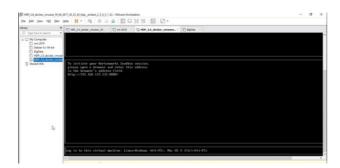
// exiting the job only if the

// flag value becomes false

System.exit(job.waitForCompletion(true) ? 0 : 1);

}
```

Start the server



Open the terminal with 192.168.119.132/4200

Enter the login: root and the password and enter



Create a folder in local directory.

Command: mkdir mscitp3

Change the directory cd mscitp3

```
The content of the co
```

Now create input file

Command: cat >> weatherin2.txt

Paste the weather dataset by right clicking on terminal

Ctrl d will save the file

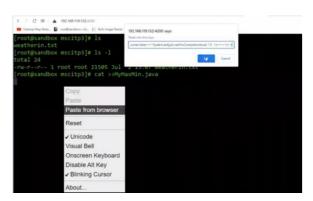
Run command Is to see the file.

Create java file

mkdir classes

Command: cat >>MyMaxMin.java

Paste the java code and ctrl d to save the file



export HADOOP_CLASSPATH=\$(hadoop classpath) ////compile and to create jar file

javac -classpath \${HADOOP CLASSPATH} -d classes MyMaxMin.java

After compile need to create a jar file

jar -cvf MyMaxMin.jar -C classes/.

```
### MINISTRATION | Description | Description
```

Now, put weatherin.txt in hdfs

Before that create a folder

Command: hdfs dfs -mkdir /p3input123

Then run command: hdfs dfs -put weatherin2.txt /p3input123

hadoop jar MyMaxMin.jar MyMaxMin /p3inputw /output123

Check outfile is created

Command: hdfs dfs -ls/output123

hdfs dfs -cat /output123/*

Practical 4

 $\label{lem:monopole} \textbf{Aim: Implement an application that stores big data in Hbase / MongoDB and manipulate it using R / Python \\$

Requirement

- a. PyMongo
- b. Mongo Database

Step A: Install Mongo database

- Step 1) Go to (https://www.mongodb.com/download-center/community) and Download MongoDB Community Server. We will install the 64-bit version for Windows.
- Step 2) Once download is complete open the msi file. Click Next in the start up screen
- Step 3) 1. Accept the End-User License Agreement 2. Click Next
- Step 4) Click on the "complete" button to install all of the components. The custom option can be used to install selective components or if you want to change the location of the installation.
- Step 5) 1. Select "Run service as Network Service user". make a note of the data directory, we'll need this later. 2. Click Next
- Step 6) Click on the Install button to start the installation
- Step 7) Installation begins. Click Next once completed
- Step 8) Click on the Finish button to complete the installation

Test Mongodb:

- Step 1) Go to "C:\Program Files\MongoDB\Server\4.0\bin" and double click on mongo.exe. Alternatively, you can also click on the MongoDB desktop item
- Create the directory where MongoDB will store it's files. From the command prompt run md \data\db . This is the default location. However, other locations can be specified using the --dbpath parameter. See the Mongo docs for more information. o C:\>md data o C:\md data\db o C:\Program Files\MongoDB\Server\4.05\bin>mongod.exe --dbpath "C:\data"
- Start the mongodb daemon by running C:\mongodb\bin\mongod.exe in the Command Prompt. Or by running, C:\path\to\mongodb\bin\mongod.exe
- Connect to MongoDB using the Mongo shell While the MongoDB daemon is running, from a different Command prompt window run C:\mongodb\bin\mongo.exe
- C:\Program Files\MongoDB\Server\4.05\bin>mongod.exe --dbpath "C:\data"
- C:\Program Files\MongoDB\Server\4.05\bin>mongo.exe

```
Step B: Install PyMongo
C:\Users\Your Name\AppData\Local\Programs\Python\Python36-32\Scripts>python -m pip install pymongo
Now you have downloaded and installed a mongoDB driver
Test PyMongo
demo mongodb test.py
Program 1: Creating a Database
import pymongo
myclient = pymongo.MongoClient("mongodb://localhost:27017/")
mydb = myclient["mybigdata"]
print(myclient.list database names())
Output:
Mybigdata
Progam 2: Creating a Collection
import pymongo
myclient = pymongo.MongoClient("mongodb://localhost:27017/")
mydb = myclient["mybigdata"]
mycol=mydb["student']
print(mydb.list collection names())
Output:
Student
Progam 3: Insert into Collection
import pymongo
myclient = pymongo.MongoClient("mongodb://localhost:27017/")
mydb = myclient["mybigdata"]
mycol=mydb["student']
mydict={"name":"Kaushal", "address":"Mumbai"}
x=mycol.insert one(mydict)
print(mydb.list collection names())
# insert one(containing the name(s) and value(s) of each field
```

Output:

Kaushal Mumbai

```
Program 4: Insert Multiple data into Collection
```

```
import pymongo
myclient = pymongo.MongoClient("mongodb://localhost:27017/")
mydb = myclient["mybigdata"]
mycol=mydb["student']
mylist=[{"name":"Kaushal", "address":"Mumbai"}, {"name":"A", "address":"Mumbai"}, {"name":"B", "address":"Pune"}, {"name":"C", "address":"Pune"},]
```

Steps and Output

x=mycol.insert many(mylist)

Install mongo db

1. Open cmd

Start the mongodb daemon by running C:\mongodb\bin\mongod.exe in the

Command Prompt. Or by running, C:\path\to\mongodb\bin\mongod.exe

2. Open another cmd

Connect to MongoDB using the Mongo shell While the MongoDB daemon is

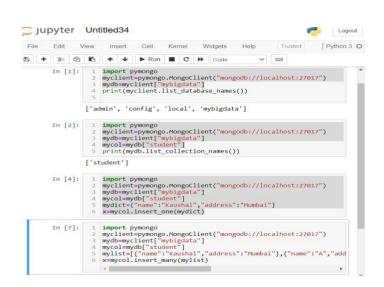
running, from a different Command prompt window

run C:\mongodb\bin\mongo.exe

3. Open Anaconda Prompt

Install pip install pymongo

Then in Anaconda Launch Jupyter Notebook

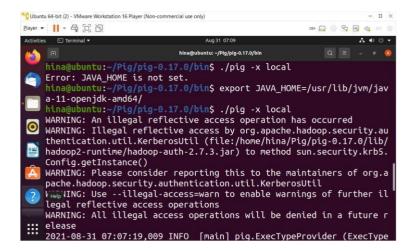


In MongoDB

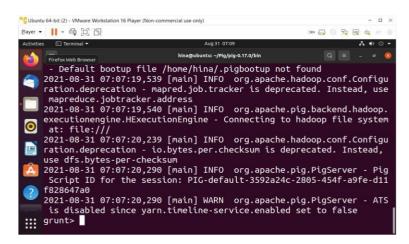
```
show dbs
dmin
       0.000GB
onfig
       0.000GB
       0.000GB
ocal
ybigdata 0.000GB
use mybigdata
witched to db mybigdata
show collections
tudent
db.collection name.find()
ncaught exception: SyntaxError: unexpected token: identifier :
(shell):1:14
student.collection name.find()
ncaught exception: SyntaxError: unexpected token: identifier :
(shell):1:19
db.collection
ybigdata.collection
```

Practical 5

Implement the program in practical 4 using Pig



create a directory and get into that directory



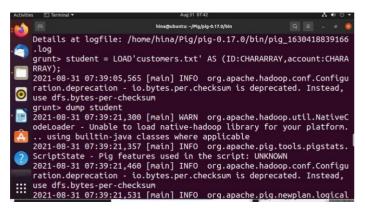
Create a file

Command: cat >>student.txt



Create a program file

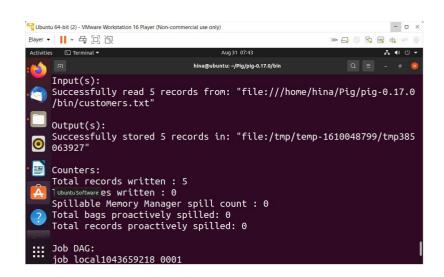
script start
student = LOAD 'student.txt' USING PigStorage(',')
as (id:int, firstname:chararray, lastname:chararray, age:int, phone:chararray, city:chararray);
student_order = ORDER student BY age DESC;
student_limit = LIMIT student_order 4;
Dump student_limit;
script end



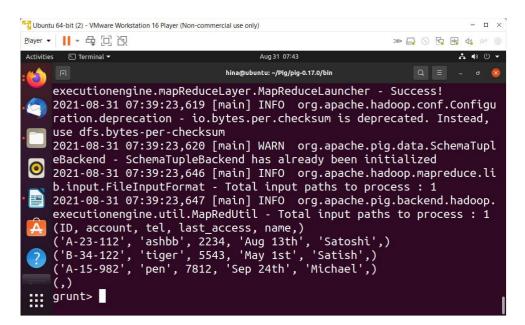
Upload student on hdfs

Command: hdfs dfs -put student.txt /user/root/

Run the pig program



OUTPUT:



Practical 6

Configure the Hive and implement the application in Hive.

Create a directory and get into that directory

Command: mkdir p6mscit

Create a file

Command: cat >>data.txt

Right click and paste the text

```
### Padeop Map Reface  
| Padeop Map Reface
```

Remove the space with vi editor

Command: vi student.txt and press i for insert mode

After editing: wq and enter

Print the content and see the text

Now start the hive terminal

Command: hive

Copy paste below command on hive and enter

create table

CREATE TABLE IF NOT EXISTS employee (eid int, fname String, lname String, age int, contact String, city String)
COMMENT 'Employee details'
ROW FORMAT DELIMITED
FIELDS TERMINATED BY ','
LINES TERMINATED BY '\n'
STORED AS TEXTFILE;

```
Description of the Company of the C
```

Run command: LOAD DATA LOCAL INPATH 'data.txt' OVERWRITE INTO TABLE employee;

```
hive> LOAD DATA LOCAL INPATH 'data.txt' OVERWRITE INTO TABLE employee;
Loading data to table default.employee
Table default.employee stats: [numFiles=1, numRows=0, totalSize=339, rawDataSize=0]
OK
Time taken: 1.296 seconds
hive> ;
```

Run the command like select * from employee;

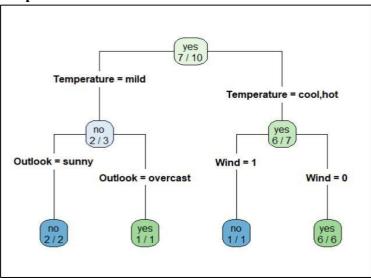
Practical 8

Implement the following:

a. Implement Decision tree classification techniques

```
Code:
library("rpart") # load libraries
library("rpart.plot")
play decision <- read.table ("C:/2020-21/IDOL/Big Data/Practical/DTdata.csv", header=TRUE, sep=",")
play decision
summary(play decision)
fit <- rpart(Play ~ Outlook+ Temperature +Humidity+ Wind,
       method="class",
       data=play decision,
       control=rpart.control(minsplit=1),
       parms=list(split='information'))
summary(fit)
rpart.plot(fit, type=4, extra=1)
rpart.plot(fit, type=4, extra=2, clip.right.labs=FALSE,
      varlen=0, faclen=0)
newdata <- data.frame(Outlook="rainy", Temperature="mild",
             Humidity="high", Wind=FALSE)
newdata
#predict(object, newdata = list(),type= c("vector", "prob", "Class","matrix"))
predict(fit,newdata=newdata,type="prob")
predict(fit,newdata=newdata,type="class")
```

Output:



b. Implement SVM classification techniques

```
Code: library(e1071)
# read the data into a table from the file
sample <- read.table("C:/2020-21/IDOL/Big Data/Practical/sample2.csv",header=TRUE,sep=",")
# define the data frames for the NB classifier
traindata<-as.data.frame(sample[1:14,])
testdata<-as.data.frame(sample[15,])
traindata
testdata
tprior<-table(traindata$Enrolls)
tprior
tprior<-tprior/sum(tprior)
tprior
ageCounts <- table(traindata[,c("Enrolls","Age")])
ageCounts
```

```
ageCounts <- ageCounts/rowSums(ageCounts)</pre>
ageCounts
incomeCounts <- table(traindata[,c("Enrolls", "Income")])</pre>
incomeCounts <- incomeCounts/rowSums(incomeCounts)</pre>
incomeCounts
jsCounts <- table(traindata[,c("Enrolls","JobSatisfaction")])
jsCounts <- jsCounts/rowSums(jsCounts)</pre>
isCounts
desireCounts <- table(traindata[,c("Enrolls","Desire")])</pre>
desireCounts <- desireCounts/rowSums(desireCounts)</pre>
desireCounts
#prob yes<-ageCounts["Yes",testdata[,c("Age")]]*</pre>
#incomeCounts["Yes",testdata[,c("Income")]]*
#jsCounts["Yes",testdata[,c("JobSatisfaction")]]*
#desireCounts["Yes",testdata[,c("Desire")]]*tprior["Yes"]
model <- naiveBayes(Enrolls ~Age+Income+JobSatisfaction+Desire,traindata)
model
results<- predict(model,testdata)
results
model1=naiveBayes(Enrolls ~., traindata, laplace=.01)
model1
results1 <- predict(model1,testdata)</pre>
results 1
Output:
```

Age Income JobSatis	sfaction Desire Enrolls	
1 <=30 High	No Fair No	
2 <=30 High	No Excellent No	
3 31 to 40 High	No Fair Yes	

traindata

4 >40 Medium	No Fair Yes
5 >40 Low	Yes Fair Yes
6 >40 Low	Yes Excellent No
7 31 to 40 Low	Yes Excellent Yes
8 <=30 Medium	No Fair No
9 <=30 Low	Yes Fair Yes
10 >40 Medium	Yes Fair Yes
11 <=30 Medium	Yes Excellent Yes
12 31 to 40 Medium	No Excellent Yes
13 31 to 40 High	Yes Fair Yes
14 >40 Medium	No Excellent No

```
> testdata
  Age Income JobSatisfaction Desire Enrolls
               Medium
                              Yes Fair
> tprior<-table(traindata$Enrolls)
> tprior
No Yes
 5 9
> tprior<- tprior/sum(tprior)</pre>
> tprior
   No
          Yes
0.3571429 0.6428571
> ageCounts <- table(traindata[,c("Enrolls","Age")])
> ageCounts
    Age
Enrolls <=30 >40 31 to 40
 No
       3 2
      2 3
 Yes
                4
> ageCounts <- ageCounts/rowSums(ageCounts)
> ageCounts
    Age
Enrolls
                  >40 31 to 40
         <=30
 No 0.6000000 0.4000000 0.0000000
 Yes 0.2222222 0.3333333 0.4444444
> incomeCounts <- table(traindata[,c("Enrolls", "Income")])
> incomeCounts <- incomeCounts/rowSums(incomeCounts)
> incomeCounts
    Income
Enrolls
         High Low Medium
 No 0.4000000 0.2000000 0.4000000
 Yes 0.2222222 0.3333333 0.4444444
```

```
> jsCounts <- table(traindata[,c("Enrolls","JobSatisfaction")])
 > jsCounts <- jsCounts/rowSums(jsCounts)
 > isCounts
     JobSatisfaction
 Enrolls
            No
                    Yes
   No 0.8000000 0.2000000
   Yes 0.3333333 0.6666667
 > desireCounts <- table(traindata[,c("Enrolls","Desire")])
 > desireCounts <- desireCounts/rowSums(desireCounts)
 > desireCounts
     Desire
 Enrolls Excellent
                     Fair
   No 0.6000000 0.4000000
   Yes 0.3333333 0.6666667
 > #prob yes<-ageCounts["Yes",testdata[,c("Age")]]*
 > #incomeCounts["Yes",testdata[,c("Income")]]*
 > #jsCounts["Yes",testdata[,c("JobSatisfaction")]]*
 > #desireCounts["Yes",testdata[,c("Desire")]]*tprior["Yes"]
 > model <- naiveBayes(Enrolls ~Age+Income+JobSatisfaction+Desire,traindata)
 > model
 Naive Bayes Classifier for Discrete Predictors
 naiveBayes.default(x = X, y = Y, laplace = laplace)
 A-priori probabilities:
 Y
     No
           Yes
 0.3571429 0.6428571
 Conditional probabilities:
     Age
         <=30
                  >40 31 to 40
  No 0.6000000 0.4000000 0.0000000
  Yes 0.2222222 0.3333333 0.4444444
     Income
 Y
         High
                 Low Medium
  No 0.4000000 0.2000000 0.4000000
  Yes 0.2222222 0.3333333 0.4444444
     JobSatisfaction
          No
                 Yes
  No 0.8000000 0.2000000
  Yes 0.3333333 0.6666667
     Desire
      Excellent
 Y
                  Fair
36
```

```
No 0.6000000 0.4000000
 Yes 0.3333333 0.6666667
> results<- predict(model,testdata)
> results
factor(0)
Levels:
> model1=naiveBayes(Enrolls ~., traindata, laplace=.01)
> model1
Naive Bayes Classifier for Discrete Predictors
Call:
naiveBayes.default(x = X, y = Y, laplace = laplace)
A-priori probabilities:
   No
          Yes
0.3571429 0.6428571
Conditional probabilities:
   Age
Y
        <=30
                >40 31 to 40
 No 0.6020000 0.4020000 0.0020000
 Yes 0.2233333 0.3344444 0.4455556
   Income
Y
              Low Medium
        High
 No 0.4020000 0.2020000 0.4020000
 Yes 0.2233333 0.3344444 0.4455556
   JobSatisfaction
Y
         No
               Yes
 No 0.8020000 0.2020000
 Yes 0.3344444 0.6677778
   Desire
    Excellent
               Fair
No 0.6020000 0.4020000
 Yes 0.3344444 0.6677778
> results1 <- predict(model1,testdata)</pre>
> results1
factor(0)
Levels:
```

Practical 9a Logistic Regression

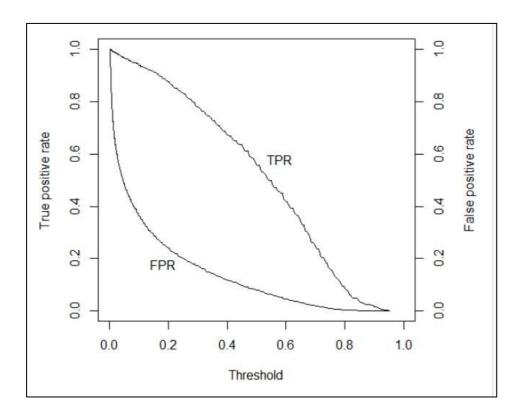
A wireless telecommunications company wants to estimate the probability that a customer will churn (switch to a different company) in the next six months. With a reasonably accurate prediction of a person's likelihood of churning, the sales and marketing groups can attempt to retain the customer by offering various incentives. Data on 8,000 current and prior customers was obtained. The variables collected for each customer follow:

```
o Age (years)
o Married(true/false)
o Duration as a customer (years)
o Churned con tact s (count)-Number of the customer's contacts that have churned (count)
o Churned (true/false)-Whether the customer churned
Code
churn input = as.data.frame(read.csv("C:/2020-21/IDOL/Big Data/Practical/churn.csv"))
head(churn input)
sum(churn input$Churned)
Churn logistic1 <- glm (Churned~Age + Married + Cust years + Churned contacts,
                   data=churn input, family=binomial(link="logit"))
summary(Churn_logistic1)
Churn logistic2 <- glm (Churned~Age + Married + Churned contacts,
              data=churn input, family=binomial(link="logit"))
summary(Churn logistic2)
Churn logistic3 <- glm (Churned~Age + Churned contacts,
              data=churn input, family=binomial(link="logit"))
summary(Churn logistic3)
# Deviance and the Log Likelihood Ratio Test
# Using the residual deviances from Churn logistics2 and Churn logistic3
# determine the signficance of the computed test statistic
summary(Churn logistic2)
```

```
pchisq(.9, 1, lower=FALSE)
# Receiver Operating Characteristic (ROC) Curve
#install.packages("ROCR")
                              #install, if necessary
library(ROCR)
pred = predict(Churn logistic3, type="response")
predObj = prediction(pred, churn input$Churned )
rocObj = performance(predObj, measure="tpr", x.measure="fpr")
aucObj = performance(predObj, measure="auc")
plot(rocObj, main = paste("Area under the curve:", round(aucObj@y.values[[1]],4)))
# extract the alpha(threshold), FPR, and TPR values from rocObj
alpha <- round(as.numeric(unlist(rocObj@alpha.values)),4)
fpr <- round(as.numeric(unlist(rocObj@x.values)),4)</pre>
tpr <- round(as.numeric(unlist(rocObj@y.values)),4)
# adjust margins and plot TPR and FPR
par(mar = c(5,5,2,5))
plot(alpha,tpr, xlab="Threshold", xlim=c(0,1), ylab="True positive rate", type="l")
par(new="True")
plot(alpha,fpr, xlab="", ylab="", axes=F, xlim=c(0,1), type="l")
axis(side=4)
mtext(side=4, line=3, "False positive rate")
text(0.18,0.18,"FPR")
text(0.58,0.58,"TPR")
i \le \text{-which(round(alpha,2)} == .5)
paste("Threshold="\ , (alpha[i])\ , "\ TPR="\ , tpr[i]\ , "\ FPR="\ , fpr[i])
i \leftarrow which(round(alpha,2) == .15)
paste("Threshold=", (alpha[i]), "TPR=", tpr[i], "FPR=", fpr[i])
```

```
Output:
# Deviance and the Log Likelihood Ratio Test
> # Using the residual deviances from Churn logistics2 and Churn logistic3
> # determine the signficance of the computed test statistic
> summary(Churn logistic2)
Call:
glm(formula = Churned ~ Age + Married + Churned contacts, family = binomial(link = "logit"),
   data = churn input)
Deviance Residuals:
   Min
           10 Median
                           3Q
                                 Max
-2.4476 -0.5178 -0.1972 -0.0723 3.3776
Coefficients:
           Estimate Std. Error z value Pr(>|z|)
(Intercept)
              3.472062  0.132107  26.282  <2e-16 ***
Age
            -0.156635 0.004088 -38.318 <2e-16 ***
              0.066430 0.068299 0.973 0.331
Married
Churned contacts 0.381909 0.027302 13.988 <2e-16 ***
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' '1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 8387.3 on 7999 degrees of freedom
Residual deviance: 5358.3 on 7996 degrees of freedom
AIC: 5366.3
Number of Fisher Scoring iterations: 6
> pchisq(.9, 1, lower=FALSE)
[1] 0.3427817
> # Receiver Operating Characteristic (ROC) Curve
>
> #install.packages("ROCR")
                               #install, if necessary
> library(ROCR)
> pred = predict(Churn logistic3, type="response")
> predObj = prediction(pred, churn input$Churned)
> rocObj = performance(predObj, measure="tpr", x.measure="fpr")
> aucObj = performance(predObj, measure="auc")
> plot(rocObj, main = paste("Area under the curve:", round(aucObj@y.values[[1],4)))
> # extract the alpha(threshold), FPR, and TPR values from rocObj
> alpha <- round(as.numeric(unlist(rocObj@alpha.values)),4)
```

```
> fpr <- round(as.numeric(unlist(rocObj@x.values)),4)
> tpr <- round(as.numeric(unlist(rocObj@y.values)),4)
> # adjust margins and plot TPR and FPR
> par(mar = c(5,5,2,5))
> plot(alpha,tpr, xlab="Threshold", xlim=c(0,1), ylab="True positive rate", type="l")
> par(new="True")
> plot(alpha,fpr, xlab="", ylab="", axes=F, xlim=c(0,1), type="l")
> axis(side=4)
> mtext(side=4, line=3, "False positive rate")
> \text{text}(0.18, 0.18, "FPR")
> text(0.58,0.58,"TPR")
> i <- which(round(alpha,2) == .5)
> paste("Threshold=", (alpha[i]), "TPR=", tpr[i], "FPR=", fpr[i])
[1] "Threshold= 0.5004 TPR= 0.5571 FPR= 0.0793"
> i <- which(round(alpha,2) == .15)
> paste("Threshold=", (alpha[i]), "TPR=", tpr[i], "FPR=", fpr[i])
[1] "Threshold= 0.1543 TPR= 0.9116 FPR= 0.2869" "Threshold= 0.1518 TPR= 0.9122 FPR= 0.2875"
[3] "Threshold= 0.1479 TPR= 0.9145 FPR= 0.2942" "Threshold= 0.1455 TPR= 0.9174 FPR= 0.2981"
```



Practical 10

Solve the Following:

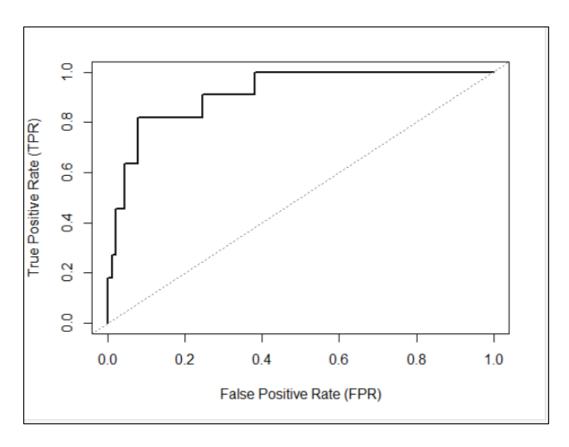
a. CLASSIFICATION MODEL a. Install relevant package for classification. b. Choose classifier for classification problem. c. Evaluate the performance of classifier.

```
Code:
```

```
library(ROCR)
 library(e1071)
 ## Read the data
 setwd("C:/2020-21/IDOL/Big Data/Practical")
 # training set
 banktrain <- read.table("bank-sample.csv",header=TRUE,sep=",")
 # drop a few columns
 drops <- c("balance", "day", "campaign", "pdays", "previous", "month")
 banktrain <- banktrain [,!(names(banktrain) %in% drops)]
 # testing set
 banktest <- read.table("bank-sample-test.csv",header=TRUE,sep=",")
 banktest <- banktest [,!(names(banktest) %in% drops)]
 # build the naïve Bayes classifier
 nb model <- naiveBayes(subscribed~., data=banktrain)
 # perform on the testing set
 nb prediction <- predict(nb model,
                # remove column "subscribed"
                 banktest[,-ncol(banktest), type='raw')
 score <- nb prediction[, c("yes")]
 actual class <- banktest$subscribed == 'yes'</pre>
 pred <- prediction(score, actual class)</pre>
 perf <- performance(pred, "tpr", "fpr")</pre>
 plot(perf, lwd=2, xlab="False Positive Rate (FPR)",
42
```

```
ylab="True Positive Rate (TPR)")
abline(a=0, b=1, col="gray50", lty=3)
## corresponding AUC score
auc <- performance(pred, "auc")</pre>
auc <- unlist(slot(auc, "y.values"))</pre>
auc
Output:
Copyright (C) 2020 The R Foundation for Statistical Computing
Platform: x86 64-w64-mingw32/x64 (64-bit)
R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.
R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.
Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.
Microsoft R Open 4.0.2
The enhanced R distribution from Microsoft
Microsoft packages Copyright (C) 2020 Microsoft Corporation
Using the Intel MKL for parallel mathematical computing (using 4 cores).
Default CRAN mirror snapshot taken on 2020-07-16.
See: https://mran.microsoft.com/.
[Workspace loaded from ~/ClassificaBAyes/.RData]
Loading required package: ROCR
> library(ROCR)
> library(e1071)
> ## Read the data
> setwd("C:/2020-21/IDOL/Big Data/Practical")
> # training set
> banktrain <- read.table("bank-sample.csv",header=TRUE,sep=",")
> # drop a few columns
> drops <- c("balance", "day", "campaign", "pdays", "previous", "month")</pre>
> banktrain <- banktrain [,!(names(banktrain) %in% drops)]
```

```
> # testing set
> banktest <- read.table("bank-sample-test.csv",header=TRUE,sep=",")
> banktest <- banktest [,!(names(banktest) %in% drops)]
> # build the naïve Bayes classifier
> nb model <- naiveBayes(subscribed~.,
                data=banktrain)
> # perform on the testing set
> nb prediction <- predict(nb model,
                 # remove column "subscribed"
+
                 banktest[,-ncol(banktest)],
                 type='raw')
> score <- nb prediction[, c("yes")]
> actual class <- banktest$subscribed == 'yes'
> pred <- prediction(score, actual class)
> perf <- performance(pred, "tpr", "fpr")
> plot(perf, lwd=2, xlab="False Positive Rate (FPR)",
     ylab="True Positive Rate (TPR)")
> abline(a=0, b=1, col="gray50", lty=3)
> ## corresponding AUC score
> auc <- performance(pred, "auc")
> auc <- unlist(slot(auc, "y.values"))</pre>
> auc
[1] 0.9152196
```



10. b CLUSTERING MODEL

a. Clustering algorithms for unsupervised classification. b. Plot the cluster data using R visualizations.

```
Code:
 library (plyr)
 library(ggplot2)
 library(cluster)
 library(lattice)
 library(graphics)
 library(grid)
 #library(gridExtra)
 #library("cowplot")
 grade input = as.data.frame(read.csv("C:/2020-21/IDOL/Big Data/grades km input.csv"))
 kmdata orig = as.matrix(grade input[,c("Student","English","Math","Science")])
 kmdata <- kmdata orig[,2:4]
 kmdata[1:10,]
 wss <- numeric(15)
 for (k \text{ in } 1:15) \text{ wss}[k] \le \text{sum } (k \text{means}(k \text{mdata}, c \text{enters} = k, n \text{start} = 25) \text{withinss})
 plot(1:15, wss, type="b", xlab="Number of Clusters ", ylab="Within Sum of Squares")
 km = kmeans(kmdata,3, nstart=25)
 km
 c(wss[3], sum(km$withinss))
 df= as.data.frame(kmdata orig[,2:4])
 df\$cluster = factor(km\$cluster)
 centers=as.data.frame(km$centers)
 g1= ggplot(data=df, aes(x=English, y=Math, color=cluster)) +
  geom point() + theme(legend.position="right") +
  geom point(data=centers,
          aes(x=English,y=Math, color=as.factor(c(1,2,3))), size=10, alpha=.3, show.legend=FALSE)
 g2 = ggplot(data=df, aes(x=English, y=Science, color=cluster)) +
  geom point()+
45
```

```
geom point(data=centers,
         aes(x=English,y=Science, color=as.factor(c(1,2,3))),size=10, alpha=.3, show.legend=FALSE)
 g3 = ggplot(data=df, aes(x=Math, y=Science, color=cluster)) +
  geom point() +
  geom point(data=centers,
         aes(x=Math,y=Science, color=as.factor(c(1,2,3))), size=10, alpha=.3, show.legend=FALSE)
 tmp = ggplot gtable(ggplot build(g1))
 grid.arrange(arrangeGrob(g1 + theme(legend.position="none"), g2 + theme(legend.position="none"), g3 +
 theme(legend.position="none"), top ="High School Student Cluster Analysis", ncol=1))
 Output:
 > library (plyr)
 > library(ggplot2)
 > library(cluster)
 > library(lattice)
 > library(graphics)
 > library(grid)
 > #library(gridExtra)
 > #library("cowplot")
 > grade input = as.data.frame(read.csv("C:/2020-21/IDOL/Big Data/grades km input.csv"))
 > kmdata orig = as.matrix(grade input[,c("Student","English","Math","Science")])
 > kmdata <- kmdata orig[,2:4]
 > kmdata[1:10,]
     English Math Science
         99 96
                   97
  [1,]
  [2,]
         99 96
                   97
         98 97
  [3,]
                   97
         95 100
                   95
  [4,]
                   96
  [5,]
         95 96
  [6,]
         96 97
                   96
        100 96
  [7,]
                   97
         95 98
                   98
  [8,]
  [9,]
         98 96
                   96
         99 99
                   95
 [10,]
 > wss <- numeric(15)
 > for (k in 1:15 ) wss[k] <- sum (kmeans(kmdata, centers=k, nstart=25)$withinss)
 > plot(1:15, wss, type="b", xlab="Number of Clusters ", ylab="Within Sum of Squares")
 > km = kmeans(kmdata,3, nstart=25)
 > km
 K-means clustering with 3 clusters of sizes 244, 158, 218
46
```

```
Cluster means:
 English
      Math Science
1 85.84426 79.68033 81.50820
2 97.21519 93.37342 94.86076
3 73.22018 64.62844 65.84862
Clustering vector:
 222222
2222222
21112111
11111111
11111111
1111133
3 3 3 3 3 3 3 3
3 3 3 3 3 3 3 3
3 3 3 3 3 3 3 3
Within cluster sum of squares by cluster:
[1] 22984.131 6692.589 34806.339
(between SS / total SS = 76.5\%)
Available components:
[1] "cluster"
        "centers"
             "totss"
                   "withinss"
                         "tot.withinss" "betweenss" "size"
                                            "iter"
[9] "ifault"
> c( wss[3], sum(km$withinss) )
[1] 64483.06 64483.06
> df= as.data.frame(kmdata orig[,2:4])
> df\$cluster = factor(km\$cluster)
> centers=as.data.frame(km$centers)
> g1= ggplot(data=df, aes(x=English, y=Math, color=cluster)) +
+ geom_point() + theme(legend.position="right") +
+ geom point(data=centers,
+
     aes(x=English,y=Math, color=as.factor(c(1,2,3))), size=10, alpha=.3, show.legend=FALSE)
> g2 =ggplot(data=df, aes(x=English, y=Science, color=cluster)) +
+ geom point()+
+ geom point(data=centers,
47
```

```
+ aes(x=English,y=Science, color=as.factor(c(1,2,3))),size=10, alpha=.3, show.legend=FALSE)
> g3 = ggplot(data=df, aes(x=Math, y=Science, color=cluster)) +
+ geom_point() +
+ geom_point(data=centers,
+ aes(x=Math,y=Science, color=as.factor(c(1,2,3))), size=10, alpha=.3, show.legend=FALSE)
> tmp = ggplot_gtable(ggplot_build(g1))
> grid.arrange(arrangeGrob(g1 + theme(legend.position="none"), g2 + theme(legend.position="none"), g3 + theme(legend.position="none"), top = "High School Student Cluster Analysis", ncol=1))
```

Output

