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**DEPARTMENT OF INFORMATION TECHNOLOGY** 



## **CERTIFICATE**

This is to certify that the Journal entitled **BIG DATA ANALYTICS** is bonafied work of **GOVIND SAINI** bearing Roll No: **07** submitted in partial fulfillment of the requirements for the award of degree of BACHELOR OF SCIENCE in INFORMATION TECHNOLOGY from University of Mumbai.

Date: 1/06/2021 Internal Guide:

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6	Write a Program showing implementation of Regression model.
7	Write a Program showing clustering.

Date Of Submission = 1/06/2021

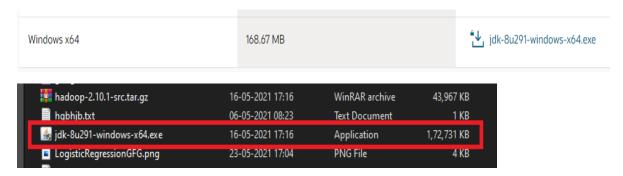
## PRACTICAL NO: 1

Aim: Install, configure and run Hadoop and HDFS

# **Description:**

Hadoop Installation.

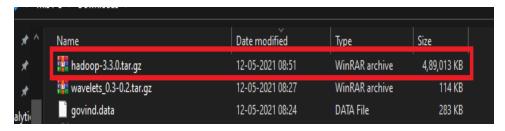
Step 1: downlaod java jdk first .the package size 168.67MB



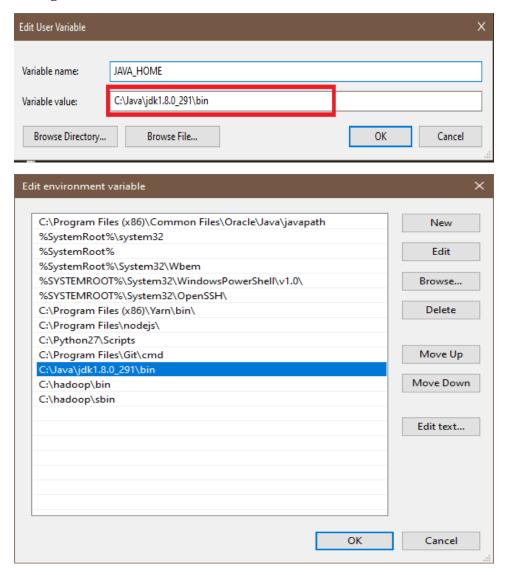
Step 2: download Hadoop binaries from the official website. The binary package size is about 342 MB.

Download								
Hadoop is released as source code tarballs with corresponding binary tarballs for convenience. The downloads are distributed via mirror sites and should be checked for tamperir using GPG or SHA-512.								
Version	Release date	Source download	Binary download	Release notes				
3.2.2	2021 Jan 9	source (checksum signature)	binary (checksum signature)	Announcement				
2.10.1	2020 Sep 21	source (checksum signature)	binary (checksum signature)	Announcement				
3.1.4	2020 Aug 3	source (checksum signature)	binary (checksum signature)	Announcement				
3.3.0	2020 Jul 14	source (checksum signature)	binary (checksum signature) binary-aarch64 (checksum signature)	Announcement				

Step 3: After finishing the file download, we should unpack the package using 7zip int two steps. First, we should extract the hadoop-3.2.1.tar.gz library, and then, we should unpack the extracted tar file:



Step 4: When the "Advanced system settings" dialog appears, go to the "Advanced" tab and click on the "Environment variables" button located on the bottom of the dialog.



Step 5: Check the version of java

```
C:\Windows\system32\cmd.exe
Microsoft Windows [Version 10.0.19041.928]
(c) Microsoft Corporation. All rights reserved.
C:\Users\hp>javac
Usage: javac <options> <source files>
where possible options include:
                           Generate all debugging info
 -g
 -g:none
                           Generate no debugging info
 -g:{lines,vars,source}
                           Generate only some debugging info
 -nowarn
                           Generate no warnings
                           Output messages about what the compiler is doing
 -verbose
                           Output source locations where deprecated APIs are used
 -deprecation
                           Specify where to find user class files and annotation process
 -classpath <path>
 -cp <path>
                           Specify where to find user class files and annotation process
                           Specify where to find input source files
 -sourcepath <path>
                           Override location of bootstrap class files
 -bootclasspath <path>
                           Override location of installed extensions
 -extdirs <dirs>
 -endorseddirs <dirs>
                           Override location of endorsed standards path
 -proc:{none,only}
                           Control whether annotation processing and/or compilation is
  -processor <class1>[,<class2>,<class3>...] Names of the annotation processors to run; t
C:\Users\hp>java -version
java version "1.8.0 291"
Java(TM) SE Runtime Environment (build 1.8.0_291-b10)
Java HotSpot(TM) 64-Bit Server VM (build 25.291-b10, mixed mode)
```

Step 6: Configuration core-site.xml

```
        container-executor.cfg
        07-07-2020 01:03
        CFG File

        core-site.xml
        19-05-2021 17:57
        XML File

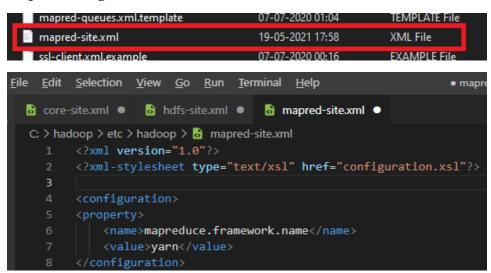
        hadoop-env.cmd
        19-05-2021 17:57
        Windows Comma...
```

Step 7: Configuration core-site.xml

hdfs-rbf-site.xml	07-07-2020 00:26	XML File
hdfs-site.xml	19-05-2021 17:58	XML File
httpfs-env.sh	07-07-2020 00:25	Shell Script

```
a core-site.xml ■ a hdfs-site.xml ■
C: > hadoop > etc > hadoop > 👶 hdfs-site.xml
  1 <?xml version="1.0" encoding="UTF-8"?>
       <?xml-stylesheet type="text/xsl" href="configuration.xsl"?>
       <configuration>
       property>
            <name>dfs.replication</name>
            <value>1</value>
       </property>
       property>
            <name>dfs.namenode.name.dir</name>
             <value>C:\hadoop\data\namenode</value>
       </property>
       property>
            <name>dfs.namenode.data.dir</name>
            <value>C:\hadoop\data\datanode</value>
       /property>
  17
       </configuration>
```

Step 8: Configuration core-site.xml

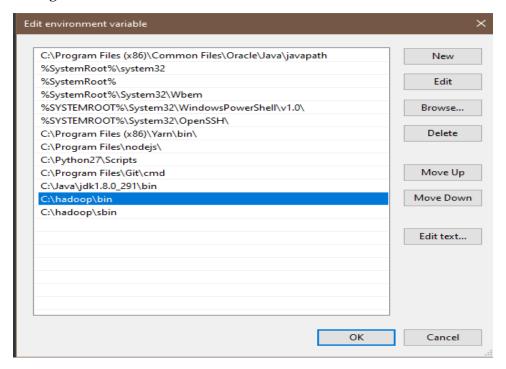


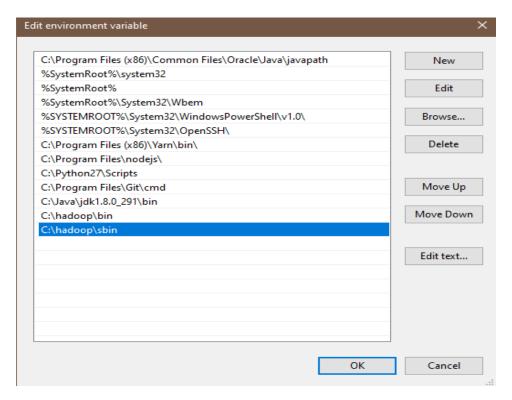
Step 9: Configuration core-site.xml

```
yarnservice-loq4j.properties 07-07-2020 01:03 PROPERTIES File

yarn-site.xml 19-05-2021 17:58 XML File
```

Step 10: When the "Advanced system settings" dialog appears, go to the "Advanced" tab and click on the "Environment variables" button located on the bottom of the dialog.





Step 11: let's check Hadoop install Successfully

```
C:\Windows\system32\cmd.exe
Java(TM) SE Runtime Environment (build 1.8.0_291-b10)
 Java HotSpot(TM) 64-Bit Server VM (build 25.291-b10, mixed mode)
 C:\Users\hp>hdfs namenode -format
 2021-05-23 17:17:11,111 INFO namenode.NameNode: STARTUP_MSG:
 STARTUP_MSG: Starting NameNode
 STARTUP_MSG: host = DESKTOP-VUUFK2Q/192.168.0.104
STARTUP_MSG:
                                                                    args = [-format]
                                                                     version = 3.3.0
classpath = C:\hadoop\etc\hadoop;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\had
STARTUP_MSG:
STARTUP_MSG:
  s-smart-1.2.jar;C:\hadoop\share\hadoop\common\lib\animal-sniffer-annotations-1.17
asm-5.0.4.jar;C:\hadoop\share\hadoop\common\lib\audience-annotations-0.5.0.jar;C:
7.7.jar;C:\hadoop\share\hadoop\common\lib\checker-qual-2.5.2.jar;C:\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\ha
   r;C:\hadoop\share\hadoop\common\lib\commons-configuration2-2.1.1.jar;C:\hadoop\sha
0.13.jar;C:\hadoop\share\hadoop\common\lib\commons-io-2.5.jar;C:\hadoop\share\hadoop\common\lib\commons-io-2.5.jar;C:\hadoop\share\hadoop\common\lib\commons-logging-1.1.3.jar;C:\hadoop\share\hadoop\common\lib\commons-net-3.6.jar;C:\hadoop\share\hadoop\common\lib\curator-client-4.2.0.jar;C:\hadoop\share\hadoop\common\lib\cura
e\hadoop\common\lib\curator-recipes-4.2.0.jar;C:\hadoop\share\hadoop\common\lib\dr
\common\lib\failureaccess-1.0.jar;C:\hadoop\share\hadoop\common\lib\gson-2.2.4.jar
va-27.0-jre.jar;C:\hadoop\share\hadoop\common\lib\hadoop-annotations-3.3.0.jar;C:\
auth-3.3.0.jar;C:\hadoop\share\hadoop\common\lib\hadoop-shaded-protobuf_3_7-1.0.0
 htrace-core4-4.1.0-incubating.jar;C:\hadoop\share\hadoop\common\lib\httpclient-4.
 ib\httpcore-4.4.10.jar;C:\hadoop\share\hadoop\common\lib\j2objc-annotations-1.1.j
```

```
Apache Hadoop Distribution
DEPRECATED: Use of this script to execute hdfs command is deprecated.
Instead use the hdfs command for it.
2021-05-23 17:19:33,116 INFO namenode.NameNode: STARTUP MSG:
 STARTUP_MSG: Starting NameNode
STARTUP MSG: host = DESKTOP-VUUFK20/192.168.0.104
STARTUP MSG: args = []
STARTUP MSG: version = 3.3.0
STARTUP_MSG: classpath = C:\hadoop\etc\hadoop;C:\hadoop\share\hadoop\common;C:\hadoop\share\hadoop\common\lib\accessor
s-smart-1.2.jar;C:\hadoop\share\hadoop\common\lib\animal-sniffer-annotations-1.17.jar;C:\hadoop\share\hadoop\common\lib\
asm-5.0.4.jar;C:\hadoop\share\hadoop\common\lib\audience-annotations-0.5.0.jar;C:\hadoop\share\hadoop\common\lib\avro-1.
7.7.jar;C:\hadoop\share\hadoop\common\lib\checker-qual-2.5.2.jar;C:\hadoop\share\hadoop\common\lib\commons-beanutils-1.9
.4.jar;C:\hadoop\share\hadoop\common\lib\commons-cli-1.2.jar;C:\hadoop\share\hadoop\common\lib\commons-codec-1.11.jar;C:
r;C:\hadoop\share\hadoop\common\lib\commons-configuration2-2.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-daemon-1.
0.13.jar;C:\hadoop\share\hadoop\common\lib\commons-io-2.5.jar;C:\hadoop\share\hadoop\common\lib\commons-lang3-3.7.jar;C:
\label{libcommons-logging-1.1.3.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\lib\commons-math3-3.1.1.jar;C:\hadoop\share\hadoop\common\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\share\hadoop\
adoop\share\hadoop\common\lib\commons-net-3.6.jar;C:\hadoop\share\hadoop\common\lib\commons-text-1.4.jar;C:\hadoop\share
 hadoop\common\lib\curator-client-4.2.0.jar;C:\hadoop\share\hadoop\common\lib\curator-framework-4.2.0.jar;C:\hadoop\shar
e\hadoop\common\lib\curator-recipes-4.2.0.jar;C:\hadoop\share\hadoop\common\lib\dnsjava-2.1.7.jar;C:\hadoop\share\hadoop
```

```
at com.ctc.wstx.sr.StreamScanner.throwParseError(StreamScanner.java:491)
at com.ctc.wstx.sr.StreamScanner.throwParseError(StreamScanner.java:475)
at com.ctc.wstx.sr.BasicStreamReader.reportWrongEndElem(BasicStreamReader.java:3365)
at com.ctc.wstx.sr.BasicStreamReader.readEndElem(BasicStreamReader.java:3292)
at com.ctc.wstx.sr.BasicStreamReader.nextFromTree(BasicStreamReader.java:2911)
at com.ctc.wstx.sr.BasicStreamReader.next(BasicStreamReader.java:1123)
at org.apache.hadoop.conf.Configuration$Parser.parseNext(Configuration.java:3347)
at org.apache.hadoop.conf.Configuration$Parser.parse(Configuration.java:3141)
at org.apache.hadoop.conf.Configuration.loadResource(Configuration.java:3034)
... 9 more
```

Step 12: Let check bin

```
C:\Windows\system32\cmd.exe

C:\Users\hp>cd C:\hadoop\sbin

C:\hadoop\sbin>start-all.cmd

This script is Deprecated. Instead use start-dfs.cmd and start-yarn.cmd starting yarn daemons

C:\hadoop\sbin>
```

### PRACTICAL NO: 2

Aim: Implement Decision tree classification techniques

## **Description:**

Decision tree builds classification or regression models in the form of a tree structure. It breaks down a dataset into smaller and smaller subsets while at the same time an associated decision tree is incrementally developed. The final result is a tree with **decision nodes** and **leaf nodes** 

Step 1: The package "party" has the function ctree() which is used to create and analyze decison tree.

```
> install.packages("party")
```

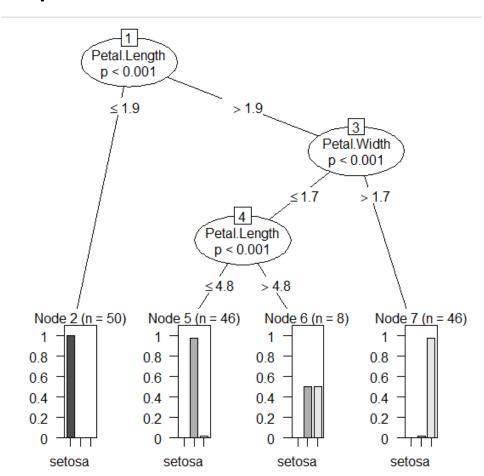
Step 2: Load the party package. It will automatically load other# dependent packages Print some records from data set readingSkills.

Step 3: Call function ctree to build a decision tree. The first parameter is a formula, which defines a target variable and a list of independent variables.

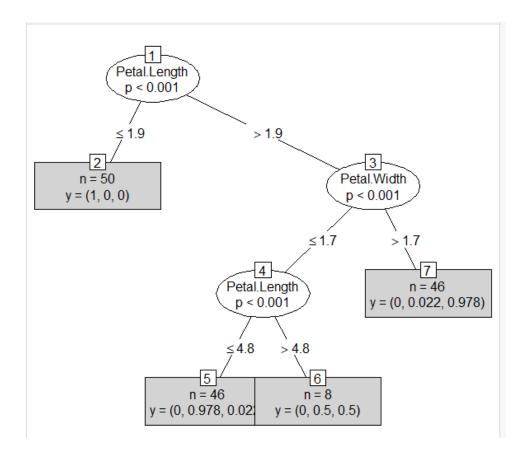
```
> library("party")
> str(iris)
'data.frame': 150 obs. of 5 variables:
$ sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
$ sepal.Width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
$ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
$ Petal.Width : num 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
$ Species : Factor w/ 3 levels "setosa", "versicolor", ..: 1 1 1 1 1 1 1 1
```

```
> iris_ctree <- ctree(Species ~ Sepal.Length + Sepal.Width + Petal.Length + Peta
1.Width, data=iris)
> print(iris_ctree)
         Conditional inference tree with 4 terminal nodes
Response: Species
Inputs: Sepal.Length, Sepal.Width, Petal.Length, Petal.Width
Number of observations: 150
1) Petal.Length <= 1.9; criterion = 1, statistic = 140.264
 2)* weights = 50
1) Petal.Length > 1.9
 3) Petal.Width <= 1.7; criterion = 1, statistic = 67.894
    4) Petal.Length <= 4.8; criterion = 0.999, statistic = 13.865
      5)* weights = 46
   4) Petal.Length > 4.8
      6)* weights = 8
  3) Petal.width > 1.7
    7)* weights = 46
> plot(iris_ctree)
```

## **Output:**



> plot(iris\_ctree, type="simple")



## PRACTICAL NO: 3

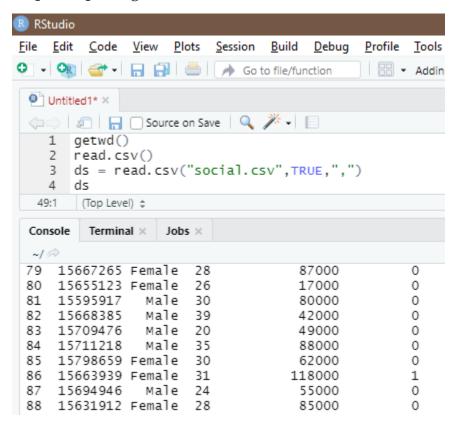
Aim: Classification using SVM

## **Description:**

A support vector machine (SVM) is a supervised machine learning model that uses classification algorithms for two-group classification problems. After giving an SVM model sets of labeled training data for each category, they're able to categorize new text

The implementation is explained in the following steps:

**Step 1: Importing the dataset** 



Step 2: Selecting columns 3-5

```
> ds = ds[3:5]
> ds[3:5]
Error in `[.data.frame`(ds, 3:5) : undefined
    Age EstimatedSalary Purchased
         19000
1
    19
2
     35
                20000
3
    26
                43000
                              0
4
    27
                57000
                              0
5
    19
                 76000
                              0
6
     27
                58000
                              0
7
                84000
                              0
    27
8
               150000
                             1
    32
9
    25
                33000
                             0
                65000
10
    35
11
    26
                80000
                             0
                52000
12
    26
                             0
```

#### Step 3: install package

```
> install.packages("caTools")
```

### Step 4: Splitting the dataset

```
> library(caTools)
 > set.seed(123)
 > split = sample.split(ds$Purchased, SplitRatio = 0.75)
 > training_set = subset(ds, split == TRUE)
 > test_set = subset(ds, split == FALSE)
 > ds
     Age EstimatedSalary Purchased
 1
      19
                  19000
                                 0
 2
                   20000
                                 0
      35
      26
 3
                  43000
                                 0
 4
     27
                  57000
                                 0
 5
     19
                  76000
                                 0
 6
     27
                  58000
                                 0
     27
 7
                  84000
                                 0
 8
     32
                                 1
                  150000
                                 0
 9
      25
                   33000
 10
      35
                   65000
                                 0
```

**Step 5: Feature Scaling** 

```
119000
332 48
                                              1
                         65000
333 42
                                             0
 [ reached 'max' / getOption("max.print") -- omitted 67 rows ]
> test_set[-3] = scale(test_set[-3])
> training_set[-3] = scale(training_set[-3])
> test_set[-3] = scale(test_set[-3])
> test_set[-3]
                 Age EstimatedSalary
    -0.30419063 -1.51354339
    -1.05994374
                          -0.32456026
    -1.81569686
5
                            0.28599864
    -1.24888202
                          -1.09579256
9
12 -1.15441288 -0.48523366
     0.64050076
                          -1.32073531
18
     0.73496990
                          -1.25646596
19
19  0.73496990  -1.25646596

20  0.92390818  -1.22433128

22  0.82943904  -0.58163769

29  -0.87100546  -0.77444577

32  -1.05994374  2.24621408

34  -0.96547460  -0.74231109

35  -1.05994374  0.73588415

38  -0.77653633  -0.58163769

45  -0.96547460  0.54307608

46  -1 43782030  -1 51354339
                            -1.51354339
46 -1.43782030
```

#### Step 6: Fitting SVM to the training set

Step 7: Predicting the test set result

```
> y_pred = predict(classifier, newdata = test_set[-3])
 y_pred
                        19
                            20
                                    29
                                                   38 45 46 48 52
                12
                    18
                                22
                                        32
                                            34
                                                35
                                                                        66
             0
                                 0
                                             0
                                                 0
     0
                     0
                         0
                             0
                                     0
                                         0
                                                    0
                                                        0
                                                            0
                    85
                        86
                           87
                                89 103 104 107 108 109 117 124 126 127 131
             0
                 0
                                             0
                     0
                         0
                             0
                                 0
                                    0
                                         1
                                                 0
                                                     0
                                                        0
                                                            0
134 139 148 154 156 159 162 163 170 175 176 193 199 200 208 213 224 226 228
             0
                 0
                     0
                         0
                             0
                                 0
                                         0
                                             0
                                                 0
                                                     0
229 230 234 236 237 239 241 255 264 265 266 273 274 281 286 292 299 302 305
                 0
                             1
                                0
                                     1
                                         1
                                             1
                                                1
                                                        0
307 310 316 324 326 332 339 341 343 347 353 363 364 367 368 369 372 373 380
                    1 0 1
383 389 392 395 400
 1
     0
        0
Levels: 0 1
```

```
> cm = table(test_set[, 3], y_pred)
> cm
    y_pred
        0 1
        0 57 7
        1 13 23
```

**Step 8:** Visualizing the Training set results

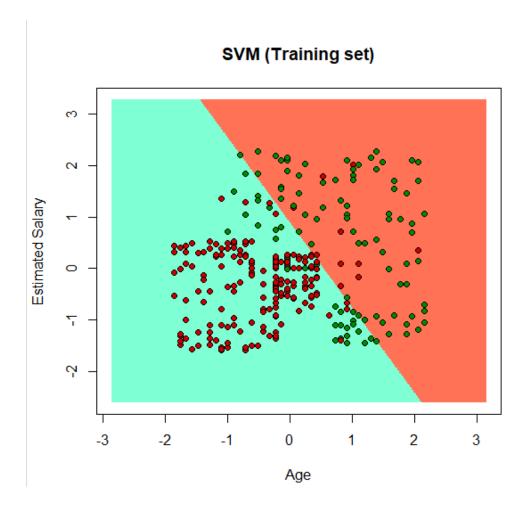
```
> set = training_set
> X1 = seq(min(set[, 1]) - 1, max(set[, 1]) + 1, by = 0.01)
> X2 = seq(min(set[, 2]) - 1, max(set[, 2]) + 1, by = 0.01)
```



```
> grid_set = expand.grid(X1, X2)
> colnames(grid_set) = c('Age', 'EstimatedSalary')
> y_grid = predict(classifier, newdata = grid_set)
 > plot(set[, -3],
+ main = 'SVM (Training set)',
          xlab = 'Age', ylab = 'Estimated Salary',
          xlim = range(X1), ylim = range(X2))
                                SVM (Training set)
      N
Estimated Salary
      0
      ۲
      Ņ
           -3
                      -2
                                -1
                                           0
                                                      1
                                                                2
                                                                           3
                                           Age
> contour(X1, X2, matrix(as.numeric(y_grid), length(X1), length(X2)), add = TRUE)
```

```
> contour(X1, X2, matrix(as.numeric(y_grid), length(X1), length(X2)), add = TRUE;
> points(grid_set, pch = '.', col = ifelse(y_grid == 1, 'coral1', 'aquamarine'))
> points(set, pch = 21, bg = ifelse(set[, 3] == 1, 'green4', 'red3'))
```

## **Output:**

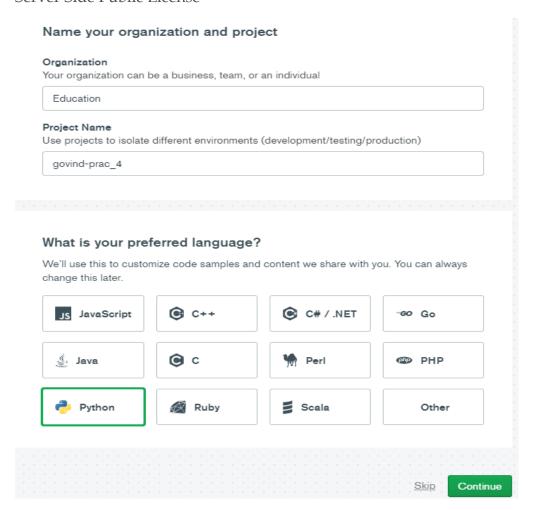


## PRACTICAL NO: 4

**Aim:** Implement an application that stores big data in Hbase / MongoDB and manipulate it using R / Python

# **Description:**

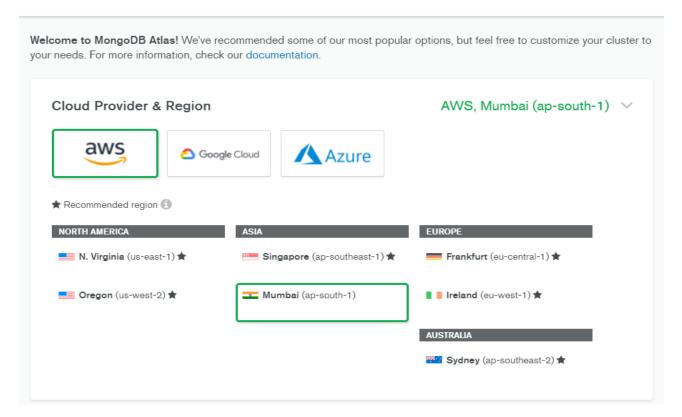
MongoDB is a source-available cross-platform document-oriented database program. Classified as a NoSQL database program, MongoDB uses JSON-like documents with optional schemas. MongoDB is developed by MongoDB Inc. and licensed under the Server Side Public License



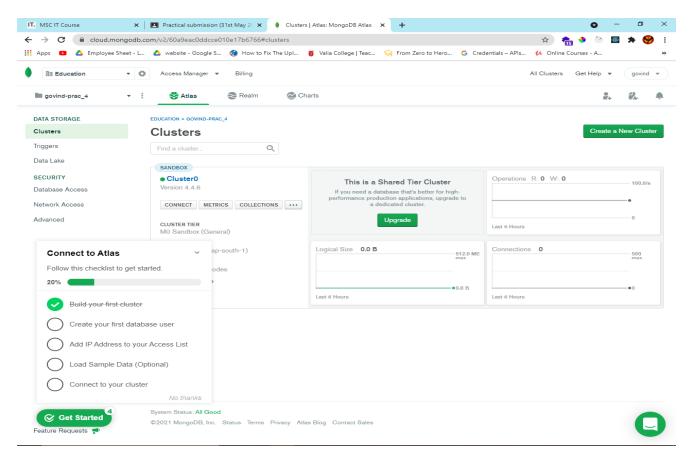
Step 1: Sign up and create a cluster.

CLUSTERS > CREATE A SHARED CLUSTER

#### Create a Shared Cluster



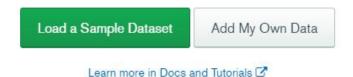
This is the home page of mongoDB Atlas.



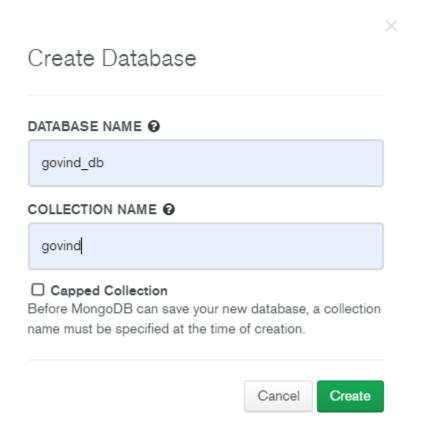
Step 2: Click on collections to create and view existing databases.



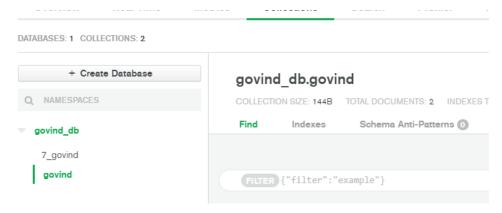
- Find: run queries and interact with documents
- Indexes: build and manage indexes
- Aggregation: test aggregation pipelines
- Search: build search indexes



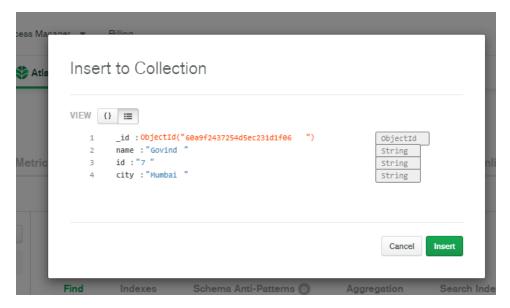
Step 3: Click on 'Add My Own Data' to create a database.



Step 4: Click on insert document to add records.



Since MongoDB is a No-SQL database, so you can add 'n' number of columns for any row/record.



### Perform updating data

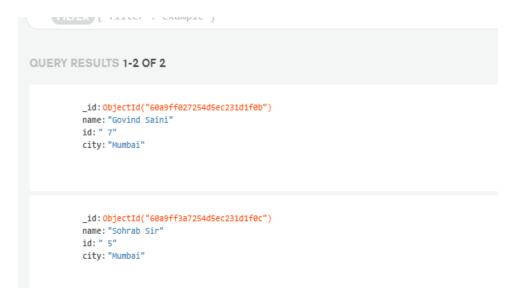
### Performing deleting data

```
_id: ObjectId("60a9f2437254d5ec231d1f06")
name: "Govind Saini"
id: "7"
city: "Mumbai"

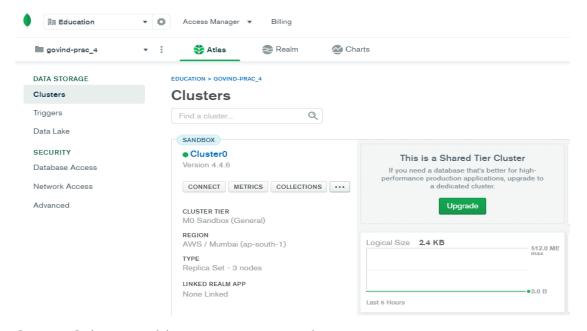
_id: ObjectId("60a9f4917254d5ec231d1f07")
name: "Sayali Mam"
id: "8"
city: "Mumbai"

Deleting Document.
```

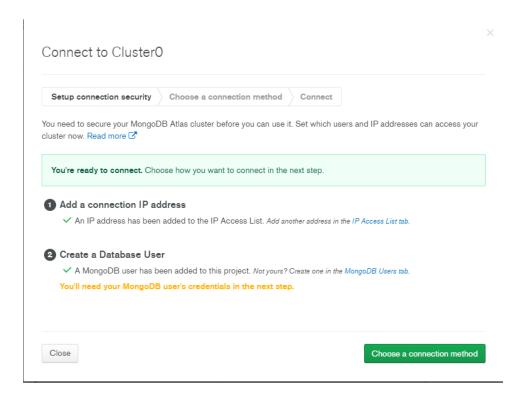
### **Performing Insert data**



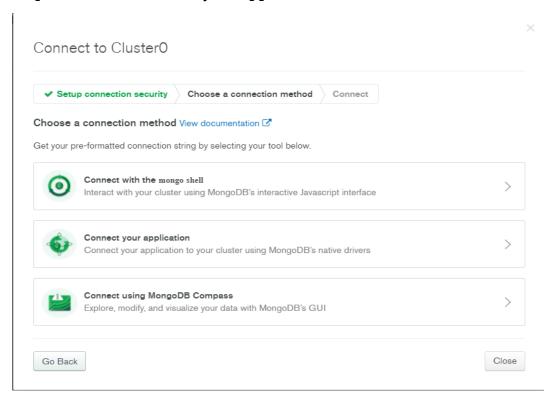
Step 5: To start with the connection click on Overview, and then click on Connect.



Step 6: Select on add your current IP and create a MongoDB user.

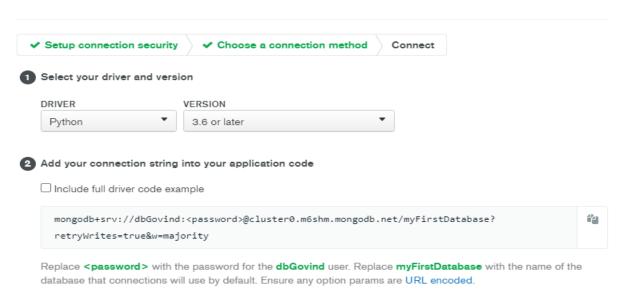


Step 7: Click on 'Connect your application'.



Step 8 : Select the driver as 'Python' and version as '3.6 or later'. (Select the version as 3.6 or later only if your Python's version is 3.6 or later.)

#### Connect to ClusterO



Having trouble connecting? View our troubleshooting documentation

#### Step 9: Write the code given below in a Python file.

```
import pymongo
from pymongo import MongoClient
client = pymongo.MongoClient("mongodb+srv://dbGovind:GmongoDB123@cdb = client.get_database('govind_db')
records = db.govind
db = client.test
print(records.count_documents({}))
print(list(records.find()))
```

## Output:

## PRACTICAL NO: 5

Aim: write program in R of Naive baye's theorem

# **Description:**

Naive Bayes is a Supervised Non-linear classification algorithm in R Programming. Naive Bayes classifiers are a family of simple probabilistic classifiers based on applying Baye's theorem with strong(Naive) independence assumptions between the features or variables

#### # Loading data

```
> data(iris)
> str(iris)
'data.frame': 150 obs. of 5 variables:
$ sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
$ sepal.width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
$ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
$ Petal.width : num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
$ species : Factor w/ 3 levels "setosa", "versicolor", ..: 1 1 1 1
```

### # Installing Packages

```
> install.packages("e1071")
> install.packages("caTools")
> install.packages("caret")
```

#### # Loading package

```
> library(e1071)
> library(caTools)
> library(caret)
Loading required package: lattice
Loading required package: ggplot2
```

# Splitting data into train and test data

```
> split <- sample.split(iris, SplitRatio = 0.7)
> train_cl <- subset(iris, split == "TRUE")
> test_cl <- subset(iris, split == "FALSE")
> train_scale <- scale(train_cl[, 1:4])</p>
> test_scale <- scale(test_cl[, 1:4])</pre>
> set.seed(120) # Setting Seed
> classifier_cl <- naiveBayes(Species ~ ., data = train_cl)
> classifier_cl
Naive Bayes Classifier for Discrete Predictors
naiveBayes.default(x = X, y = Y, laplace = laplace)
A-priori probabilities:
   setosa versicolor virginica
 0.3333333 0.3333333 0.3333333
Conditional probabilities:
          Sepal.Length
              [,1]
                         [,2]
          5.046667 0.3848272
 setosa
 versicolor 5.963333 0.5268536
 virginica 6.553333 0.6693967
           Sepal.Width
              [,1]
                          [,2]
 setosa
            3.413333 0.4256705
 versicolor 2.823333 0.3470897
 virginica 2.956667 0.3136914
           Petal.Length
              [,1]
            1.466667 0.1561019
 setosa
 versicolor 4.320000 0.4759020
 virginica 5.496667 0.5738457
           Petal.Width
                [,1]
                           [,2]
 setosa 0.2766667 0.1135124
 versicolor 1.3533333 0.1960530
 virginica 2.0433333 0.2568823
```

#### # Predicting on test data'

#### # Model Evauation

#### > confusionMatrix(cm)

Confusion Matrix and Statistics

y\_pred

setosa versicolor virginica setosa 20 0 0 versicolor 0 19 1 virginica 0 2 18

Overall Statistics

Accuracy : 0.95 95% CI : (0.8608, 0.9896)

No Information Rate: 0.35 P-Value [Acc > NIR] : < 2.2e-16

Kappa : 0.925

Mcnemar's Test P-Value : NA

Statistics by Class:

	Class: setosa Class:	versicolor Class:	virginica
Sensitivity	1.0000	0.9048	0.9474
Specificity	1.0000	0.9744	0.9512
Pos Pred Value	1.0000	0.9500	0.9000
Neg Pred Value	1.0000	0.9500	0.9750
Prevalence	0.3333	0.3500	0.3167
Detection Rate	0.3333	0.3167	0.3000
Detection Prevalence	0.3333	0.3333	0.3333
Balanced Accuracy	1.0000	0.9396	0.9493

# **PRACTICAL NO: 6**

**Aim:** Write a Program showing implementation of Regression model.

## **Description:**

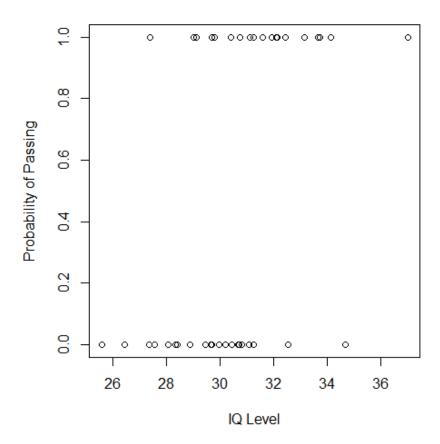
Regression is a method to mathematically formulate relationship between variables that in due course can be used to estimate, interpolate and extrapolate. Suppose we want to estimate the weight of individuals, which is influenced by height, diet, workout, etc. Here, *Weight* is the **predicted** variable

Lets implementation of Regression Model some Example:

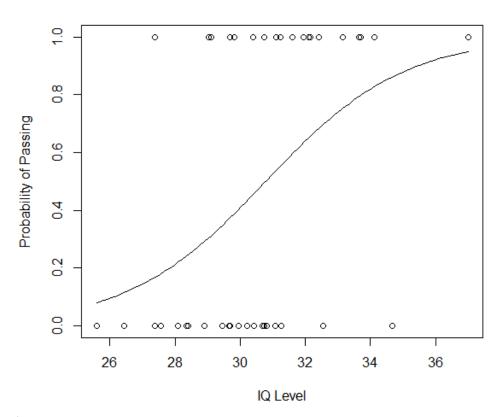
```
> IQ <- rnorm(40, 30, 2)
|> IQ <- sort(IQ)

> result <- c(0, 0, 0, 1, 0, 0, 0, 0, 0, 1, + 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, + 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, + 1, 1, 1, 0, 1, 1, 1, 1, 0, 1)</pre>
```

```
> df <- as.data.frame(cbind(IQ, result))</pre>
> print(df)
         IQ result
1 25.58824 0
               0
2 26.43200
              0
3 27.37083
               1
4 27.37898
               0
5 27.56671
6 28.08275
              0
7 28.35637
8 28.41538
9 28.89752
              0
10 29.03158
               1
               1
11 29.12386
               0
12 29.46181
13 29.66945
               0
               0
14 29.68934
               1
15 29.69886
               1
16 29.80735
17 29.95326
               0
18 30.21428
              0
               1
19 30.39298
20 30.43421
              0
              0
21 30.67802
              0
22 30.72653
23 30.74974
               1
24 30.82265
25 31.07116 0
26 31.11633 1
27 31.24740 1
28 31.25662 0
29 31.60194 1
30 31.93038 1
> png(file="LogisticRegressionGFG.png")
> plot(IQ, result, xlab = "IQ Level",
+ ylab = "Probability of Passing")
> g = glm(result~IQ, family=binomial, df)
```



- > curve(predict(g, data.frame(IQ=x), type="resp"), add=TRUE)
  > points(IQ, fitted(g), pch=30)



```
> summary(q)
glm(formula = result ~ IQ, family = binomial, data = df)
Deviance Residuals:
              1Q Median
                                   3Q
                                            мах
-1.9877 -0.9804 -0.4502
                             0.9731
                                         1.8898
Coefficients:
             Estimate Std. Error z value Pr(>|z|)
                       5.8835 -2.463 0.0138 * 0.1922 2.450 0.0143 *
(Intercept) -14.4934
               0.4708
ΙQ
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
Null deviance: 55.352 on 39 degrees of freedom Residual deviance: 47.090 on 38 degrees of freedom
AIC: 51.09
Number of Fisher Scoring iterations: 4
> dev.off()
null device
```

## PRACTICAL NO: 7

Aim: Write a Program showing clustering.

## **Description:**

# In this Program we understand about K-Mean Clustering #

What Does K-Means Clustering Mean?

- K-means clustering is a simple unsupervised learning algorithm that is used to solve clustering problems.
- It follows a simple procedure of classifying a given data set into a number of clusters, defined by the letter "k," which is fixed beforehand.
- The clusters are then positioned as points and all observations or data points are associated with the nearest cluster, computed, adjusted and then the process starts over using the new adjustments until a desired result is reached.

#### We Understand in different Steps:

Step 1: Apply kmeans to *newiris*, and store the clustering result in kc. The cluster number is set to 3.

```
> newiris <- iris
> newiris$Species <- NULL
> (kc <- kmeans(newiris, 3))
K-means clustering with 3 clusters of sizes 38, 62, 50
Cluster means:
 Sepal.Length Sepal.Width Petal.Length Petal.Width
    6.850000 3.073684 5.742105 2.071053
5.901613 2.748387 4.393548 1.433871
2
    5.006000 3.428000
3
                      1.462000
                                 0.246000
Clustering vector:
 [103] 1 1 1 1 2 1 1 1 1 1 1 2 2 1 1 1 1 2 2 1 2 1 2 1 2 1 2 1 1 1 2 2 1 1 1 1 1 2 1 1
[137] 1 1 2 1 1 1 2 1 1 1 2 1 1 2
Within cluster sum of squares by cluster:
[1] 23.87947 39.82097 15.15100
(between_SS / total_SS = 88.4 %)
Available components:
           "centers"
[1] "cluster"
                         "totss"
                                     "withinss"
[5] "tot.withinss" "betweenss"
                                      "iter"
                          "size"
[9] "ifault"
```

Step 2: Compare the Species label with the clustering result

```
> table(iris$Species, kc$cluster)

1 2 3
setosa 0 0 50
versicolor 2 48 0
virginica 36 14 0
```

Step 3: Plot the clusters and their centres. Note that there are four dimensions in the data and that only the first two dimensions are used to draw the plot below.

```
> plot(newiris[c("Sepal.Length", "Sepal.Width")], col=kc$cluster)
```

0 **4** 0 0 0 0 Ю 000 Sepal.Width ത് 00 00 000 0 00 0 0 3.0 00 000 0000 00 00 0 0 00 0 2.5 Ö 0 000 0 2.0 0 4.5 5.0 5.5 6.0 6.5 7.0 7.5 8.0

Step 4: Some black points close to the green centre (asterisk) are actually closer to the black centre in the four dimensional space.

> points(kc\$centers[,c("Sepal.Length", "Sepal.Width")], col=1:3, pch=8, cex=2)

Sepal.Length

