

**M.Sc C.S - II**  
**SEM III**  
**Journal**

<b>Roll No.</b>	056
<b>Name</b>	OJHA ABHISHEK DEVMANI
<b>Subject</b>	Business Intelligence and Big Data Analytics - II



*Thakur Educational Trust's (Regd.)*

**THAKUR COLLEGE OF SCIENCE & COMMERCE**

AUTONOMOUS COLLEGE, PERMANENTLY AFFILIATED TO UNIVERSITY OF MUMBAI

NAAC Accredited Grade 'A' (3<sup>rd</sup> Cycle) & ISO 9001: 2015 (Certified)

**Best College Award by University of Mumbai for the Year 2018-2019**

**tcsc**

**CELEBRATING  
25 YEARS OF GLORY**

## CERTIFICATE

This is here to certify that Mr. OJHA ABHISHEK DEVMANI, Seat Number 056 of M.Sc. II Computer Science, has satisfactorily completed the required number of experiments prescribed by the **THAKUR COLLEGE OF SCIENCE & COMMERCE AUTONOMOUS COLLEGE, PERMANENTLY AFFILIATED TO UNIVERSITY OF MUMBAI** during the academic year 2021 - 2022.

Date: 23-09-2022

Place: Mumbai

Teacher In-Charge

Head of Department

External Examiner

# INDEX

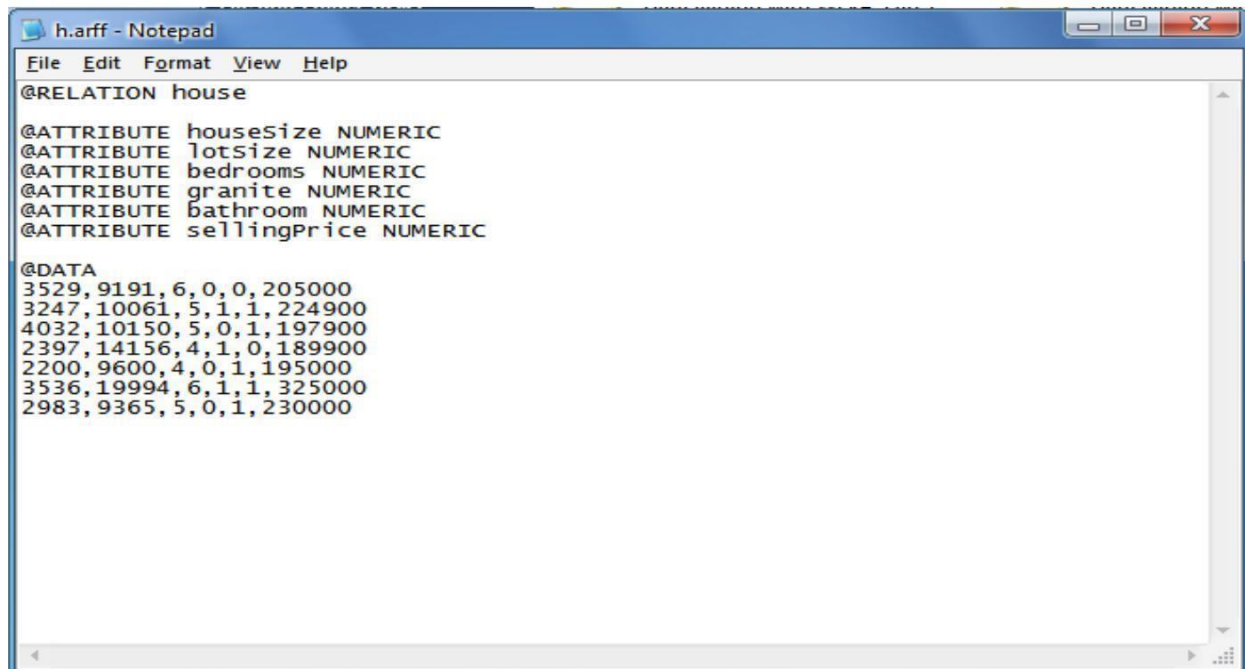
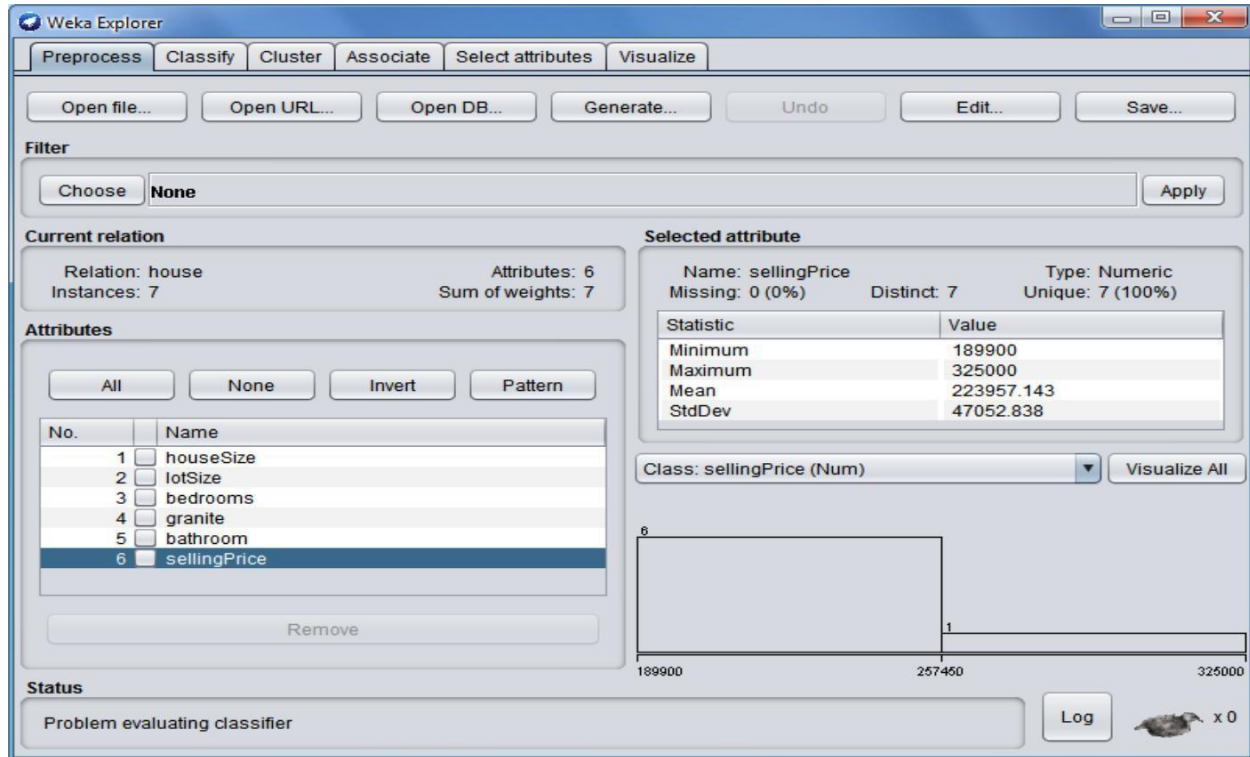
Sr. No.	Experiments	Page No	Date
1	Write a program to Generate Regression model and interpret the result for a given data set.	01-03	
2	Write a program to Generate forecasting model and interpret the result for a given data set.	04-06	
3	Write a map-reduce program to count the number of occurrences of each alphabetic character in the given dataset. The count for each should be case-insensitive(i.e include both upper-case and lower-case versions of the letter, ignore non-alphabetic characters)..	07-09	
4	Write a map-reduce program to count the number of occurrences of each word in the given dataset.(A word is defined as any string of alphabetic characters appearing between non-alphabetic characters like nature's is two words. The count should be case-insensitive. If a word occurs multiple times in a line, all should be counted).	10-12	
5	Write a program to construct different types of k-shingles for given document. Installation of required packages before executing program	13-14	
6	Write a program for measuring similarity among documents and detecting passages which have been reused.	15-19	
7	Write a program to compute the n-moment for a given stream where n is given	20-21	
8	Write a program to demonstrate the Alon-Matias-Szegedy Algorithm for second moments.	22-24	

## Business Intelligence and Big Data Analytics - II

### Practical No. 1

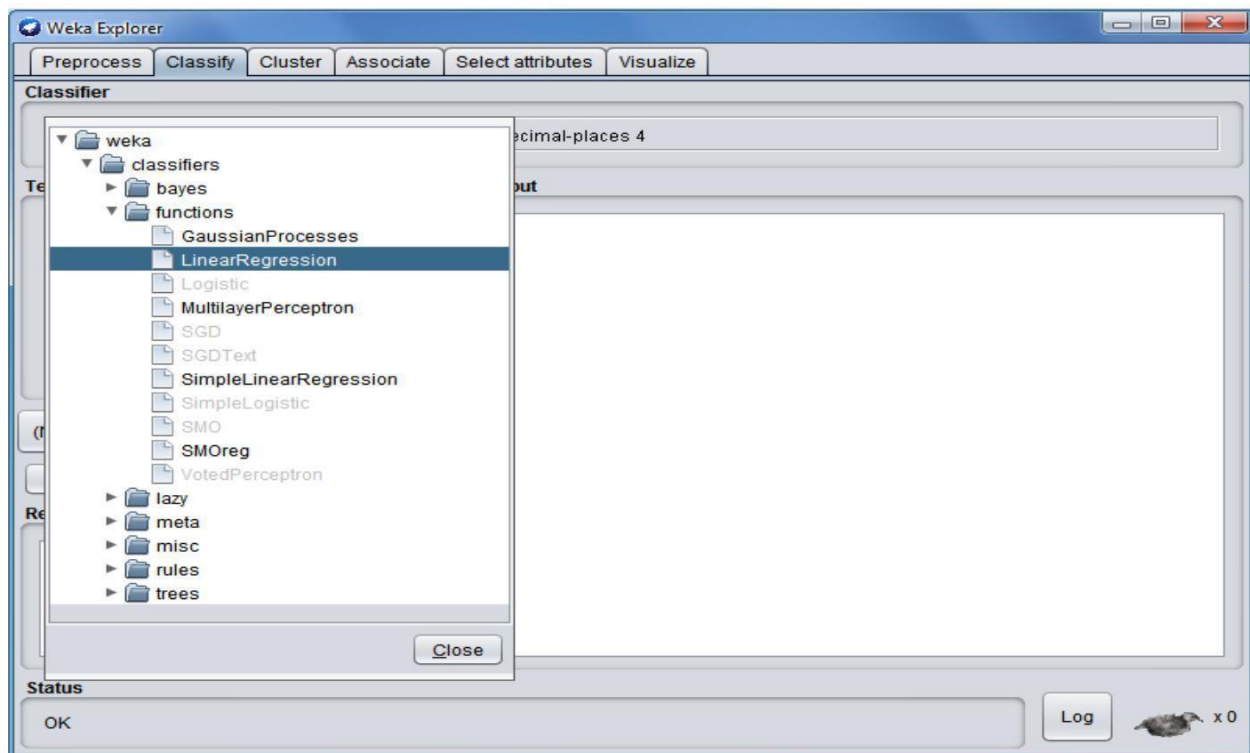
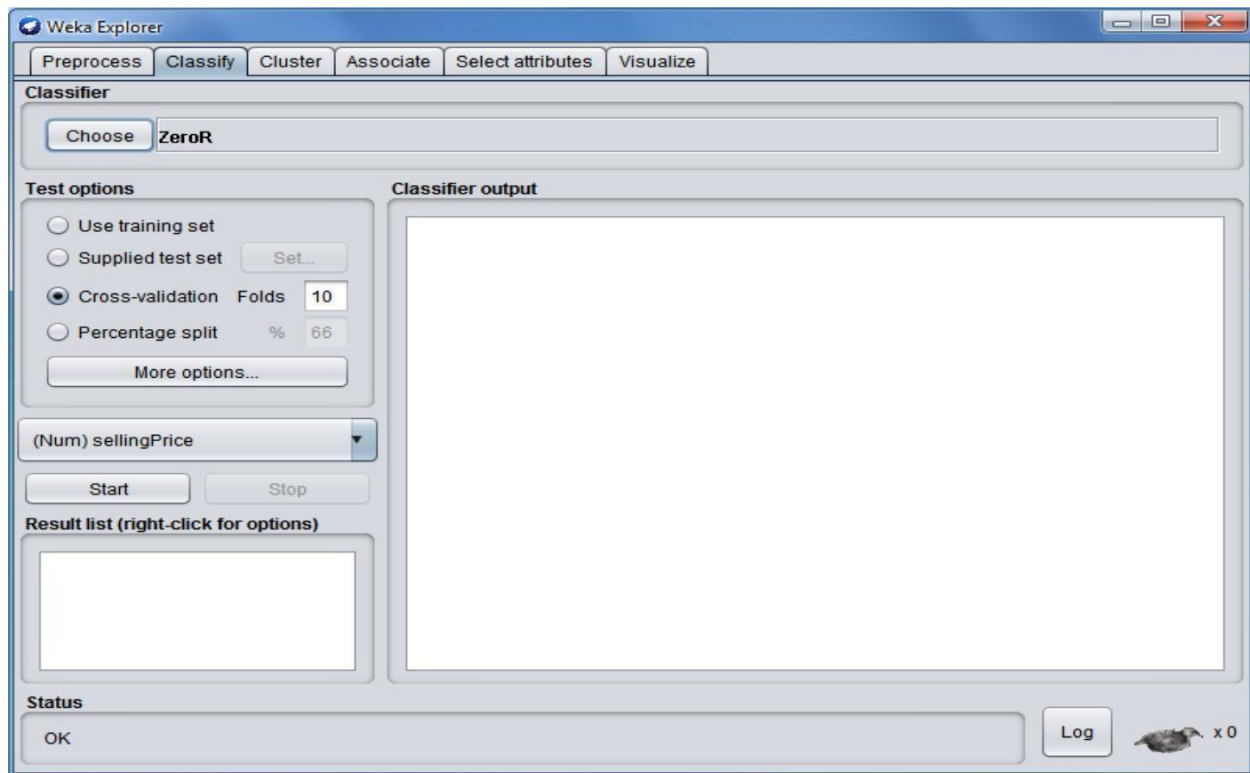
**Aim :** Generate Regression model and interpret the result for a given data set.

**Step 1:** Open Weka then open file h.arff in Weka Explorer.

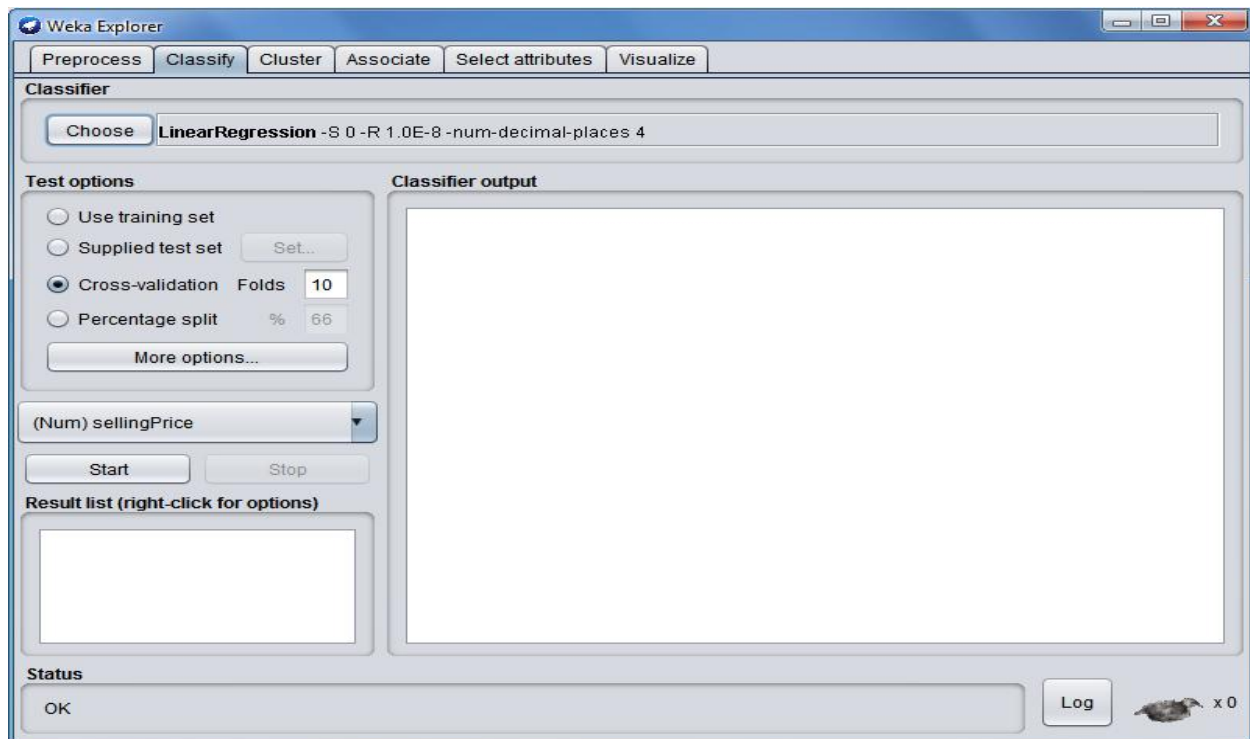


**Step 2:** Click on Classify, choose weka classifier function LinearRegression -S 0 -R 1.0E-8 -num-decimal-places 4.

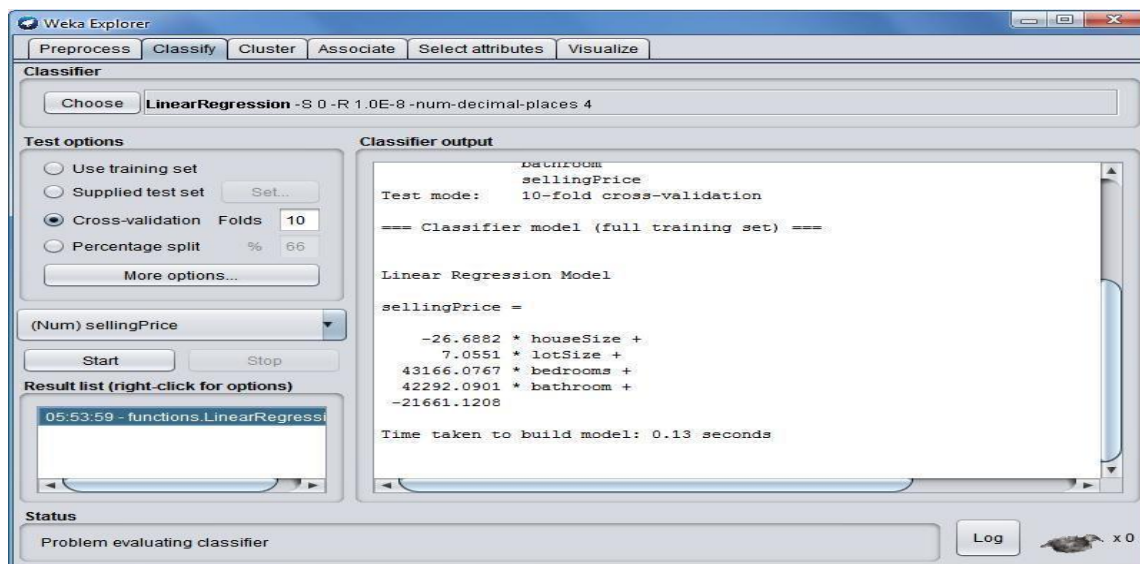
## Business Intelligence and Big Data Analytics - II



## Business Intelligence and Big Data Analytics - II



**Step 3:** Click on Start. You can see the linear regression on the input file.

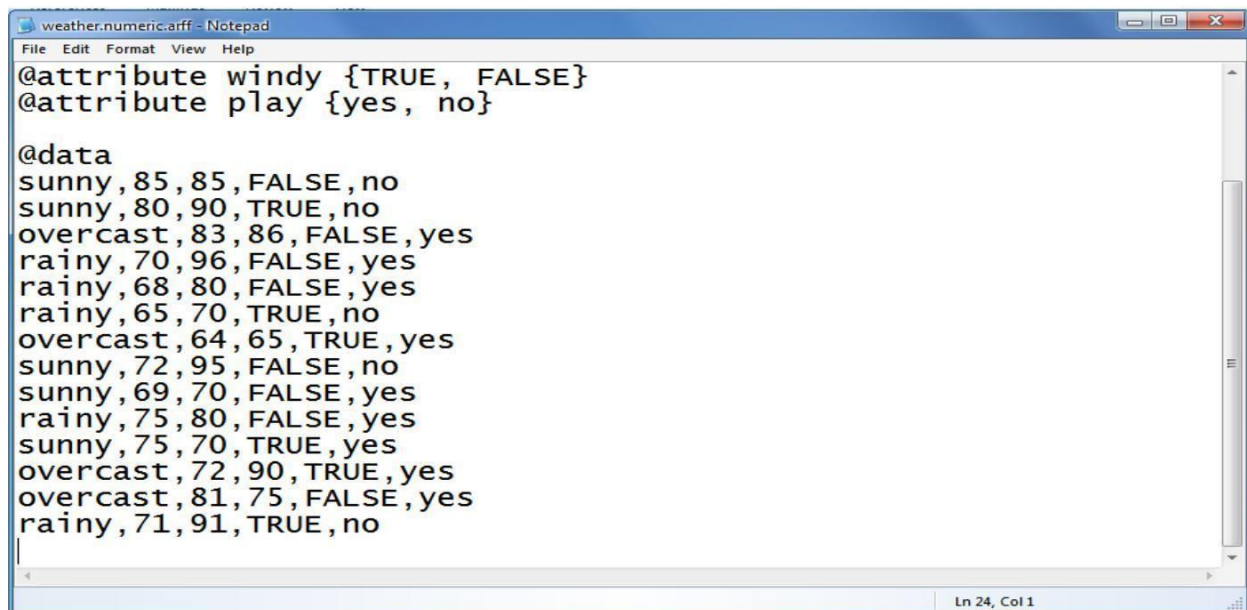
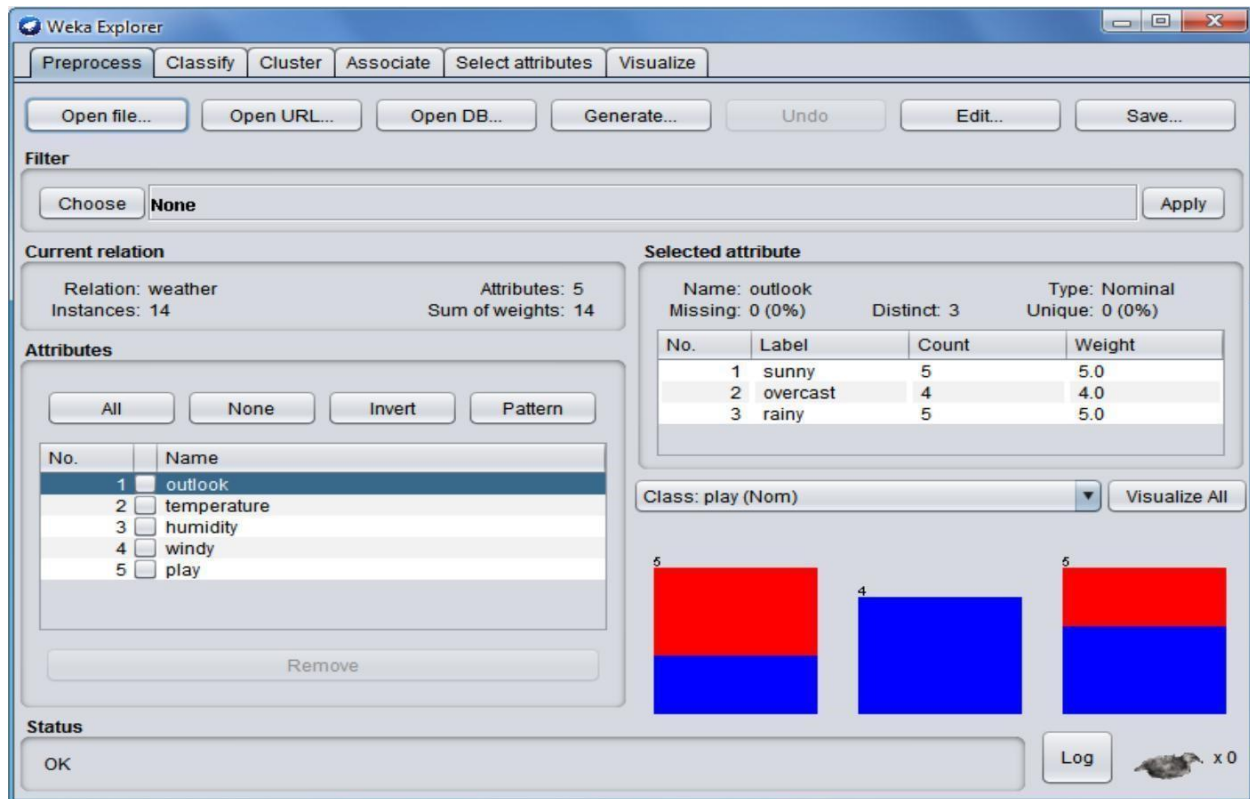


## Business Intelligence and Big Data Analytics - II

### Practical No. 2

**Aim :** Generate forecasting model and interpret the result for a given data set.

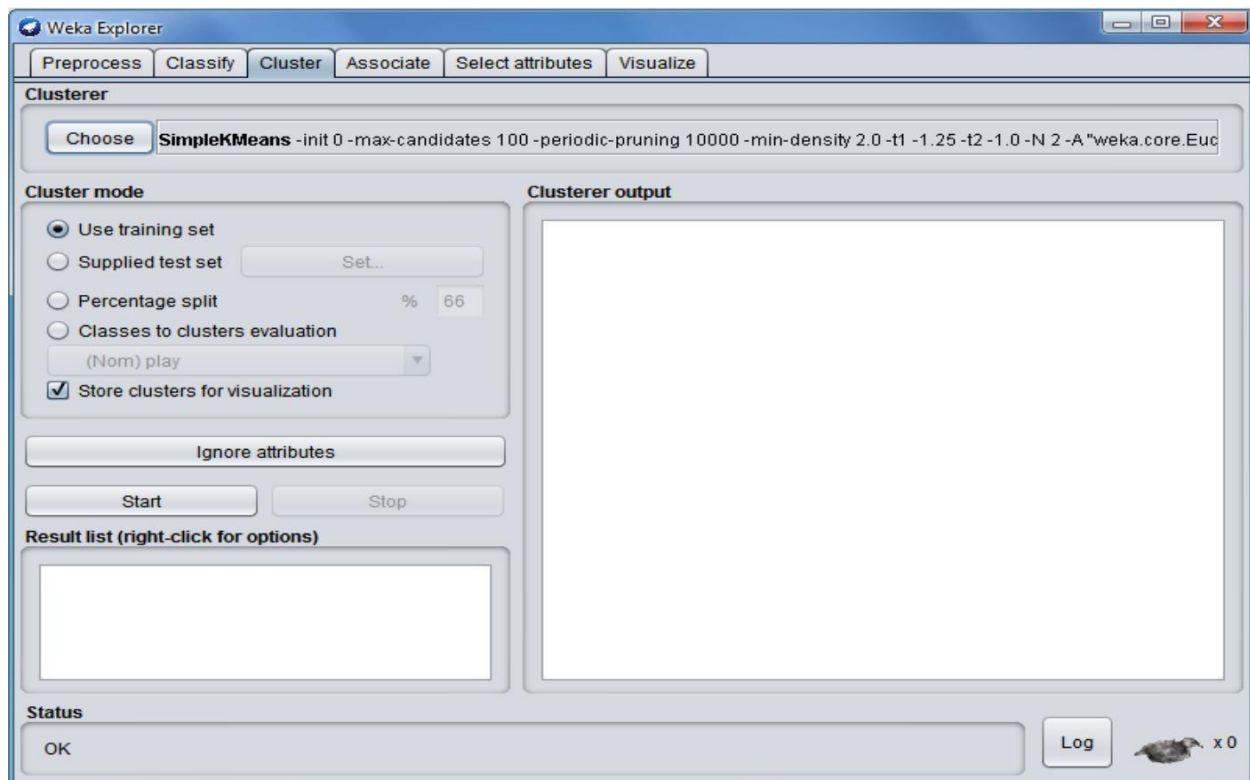
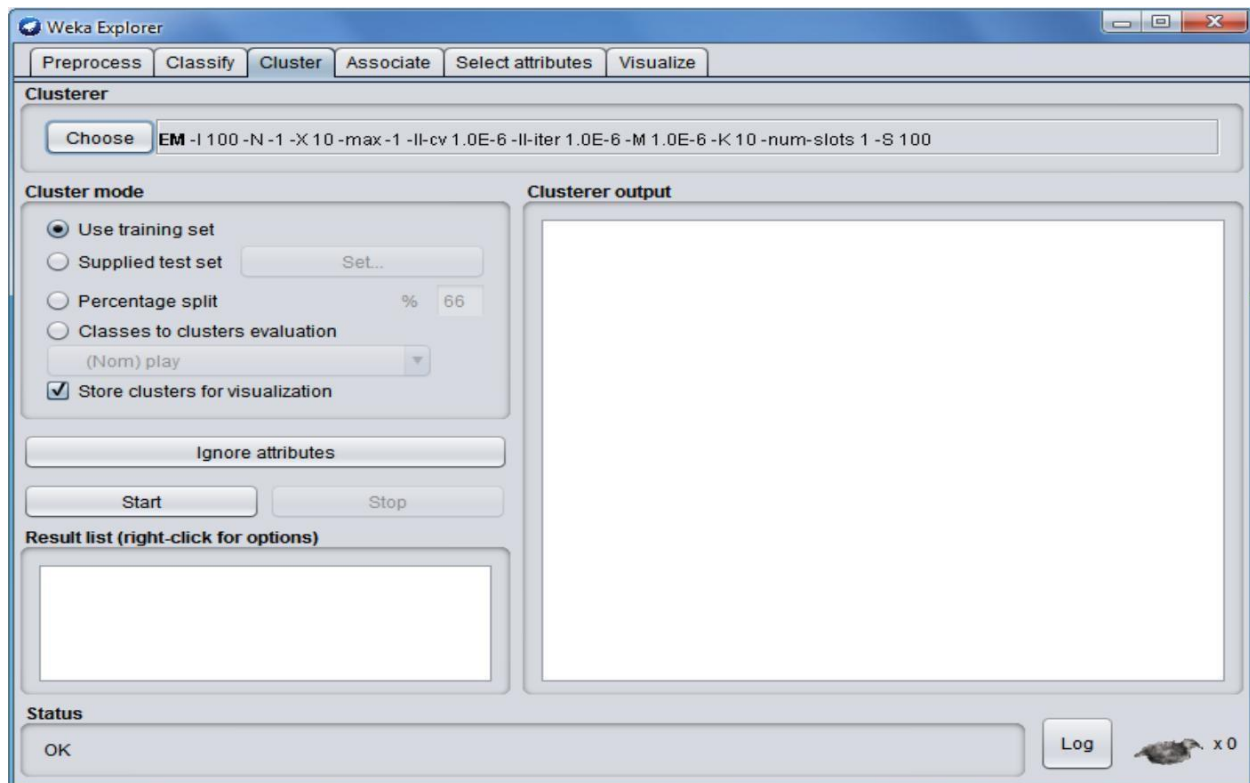
**Step 1:** Open Weka then open file Weather.arff in Weka Explorer.



**Step 2:** click on Cluster, choose weka forecasting function SimpleKMean.



## Business Intelligence and Big Data Analytics - II



**Step 3:** Click on Start. You can see the SimpleKMean on the input file.



## Business Intelligence and Big Data Analytics - II

The screenshot shows the Weka Explorer Clusterer window. The 'Clusterer' tab is selected, and 'SimpleKMeans' is chosen. The 'Cluster mode' section has 'Use training set' selected. The 'Clusterer output' window displays the following information:

Initial starting points (random):

Cluster 0: rainy,75,80,FALSE,yes  
Cluster 1: overcast,64,65,TRUE,yes

Missing values globally replaced with mean/mode

Final cluster centroids:

Attribute	Full Data	Cluster# 0	Cluster# 1
	(14.0)	(9.0)	(5.0)
outlook	sunny	sunny	overcast
temperature	73.5714	75.8889	69.4
humidity	81.6429	84.1111	77.2
windy	FALSE	FALSE	TRUE
play	yes	yes	yes

The 'Result list' shows '06:10:41 - SimpleKMeans'. The 'Status' bar shows 'OK'.

The screenshot shows the Weka Explorer Clusterer window. The 'Clusterer' tab is selected, and 'SimpleKMeans' is chosen. The 'Cluster mode' section has 'Use training set' selected. The 'Clusterer output' window displays the following information:

Time taken to build model (full training data) : 0.01 second

=== Model and evaluation on training set ===

Clustered Instances

0	9 ( 64%)
1	5 ( 36%)

The 'Result list' shows '06:10:41 - SimpleKMeans'. The 'Status' bar shows 'OK'.

### Practical No. 3

**Aim :** Write a map-reduce program to count the number of occurrences of each alphabetic character in the given dataset. The count for each should be case-insensitive(i.e include both upper-case and lower-case versions of the letter, ignore non-alphabetic characters).

**Source Code:**

**Charcount.java(Driver Class)**

```
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.lib.input.FileInputFormat;
import org.apache.hadoop.mapreduce.lib.input.TextInputFormat;
import org.apache.hadoop.mapreduce.lib.output.FileOutputFormat;
import org.apache.hadoop.mapreduce.lib.output.TextOutputFormat;
public class Charcount { public static void main(String[] args) throws
Exception {

    // TODO Auto-generated method stub Configuration conf =
    new Configuration(); Job job = new Job(conf, "Charcount");
    job.setJarByClass(Charcount.class);
    job.setMapperClass(Charmap.class);
    job.setReducerClass(Charreduce.class);
    job.setInputFormatClass(TextInputFormat.class);
    job.setOutputFormatClass(TextOutputFormat.class);
    job.setMapOutputKeyClass(Text.class);
    job.setMapOutputValueClass(IntWritable.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);

}

}
```

## Business Intelligence and Big Data Analytics - II

### Charmap.java(Mapper Class)

```
import java.io.IOException; import
java.util.StringTokenizer;

import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.io.LongWritable;
import org.apache.hadoop.io.Text; import
org.apache.hadoop.mapreduce.Mapper;

public class Charmap extends Mapper<LongWritable, Text, Text, IntWritable> {
    public void map(LongWritable key, Text value, Context context)

        throws IOException, InterruptedException {

        String line = value.toString();
        char[] carr = line.toCharArray();

        for (char c : carr) {
            System.out.println(c);
            context.write(new Text(String.valueOf(c)), new IntWritable(1));
        }

    }
}
```

### Charreduce.java(Reducer Class)

```
import java.io.IOException;

import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.io.Text; import
org.apache.hadoop.mapreduce.Reducer;

public class Charreduce extends Reducer<Text, IntWritable, Text, IntWritable> { public
    void reduce(Text key,Iterable<IntWritable> values,Context context)throws
IOException,InterruptedException{
        int count = 0;

        IntWritable result = new IntWritable();
        for (IntWritable val : values) {
            count +=val.get();
            result.set(count);

        }
    }
}
```

## Business Intelligence and Big Data Analytics - II

```
String found = key.toString();
if (found.equals("a") || found.equals("t") || found.equals("c") || found.equals("g"))
{
    context.write(key, result);
}
}
```

## Business Intelligence and Big Data Analytics - II

### Practical No. 4

**Aim :** Write a map-reduce program to count the number of occurrences of each word in the given dataset.(A word is defined as any string of alphabetic characters appearing between non-alphabetic characters like nature's is two words. The count should be case-insensitive. If a word occurs multiple times in a line, all should be counted).

#### **WordCount.java(Driver Class)**

```
import org.apache.hadoop.fs.Path;
import org.apache.hadoop.conf.*;
import org.apache.hadoop.io.*;
import org.apache.hadoop.mapred.*;
import org.apache.hadoop.util.*;

public class WordCount extends Configured implements Tool{ public
    int run(String[] args) throws Exception

    {

        //creating a JobConf object and assigning a job name for identification purposes

        JobConf conf = new JobConf(getConf(), WordCount.class);
        conf.setJobName("WordCount");

        //Setting configuration object with the Data Type of output Key and Value
        conf.setOutputKeyClass(Text.class);
        conf.setOutputValueClass(IntWritable.class);

        //Providing the mapper and reducer class names
        conf.setMapperClass(WordCountMapper.class);
        conf.setReducerClass(WordCountReducer.class);

        //We wil give 2 arguments at the run time, one in input path and other is output
path
        Path inp = new Path(args[0]);
        Path out = new Path(args[1]);

        //the hdfs input and output directory to be fetched from the command line
        FileInputFormat.addInputPath(conf, inp); FileOutputFormat.setOutputPath(conf, out);
        JobClient.runJob(conf);
        return 0;
    }
}
```

## Business Intelligence and Big Data Analytics - II

```
public static void main(String[] args) throws Exception
{
    // this main function will call run method defined above.

    int res = ToolRunner.run(new Configuration(), new WordCount(),args);
    System.exit(res);

}
}
```

### WordCountMapper.java(Mapper Class)

```
import java.io.IOException;
import java.util.StringTokenizer;

import org.apache.hadoop.io.*;
import org.apache.hadoop.mapred.*;

public class WordCountMapper extends MapReduceBase implements
Mapper<LongWritable, Text, Text, IntWritable> { //hadoop supported data types

    private final static IntWritable one = new IntWritable(1); private
    Text word = new Text();

    //map method that performs the tokenizer job and framing the initial key value pairs

    // after all lines are converted into key-value pairs, reducer is called.

    public void map(LongWritable key, Text value, OutputCollector<Text, IntWritable> output,
Reporter reporter) throws IOException

    {

        //taking one line at a time from input file and tokenizing the same String
        line = value.toString();

        StringTokenizer tokenizer = new StringTokenizer(line);
        //iterating through all the words available in that line and forming the key value
        pair while (tokenizer.hasMoreTokens())
```

## Business Intelligence and Big Data Analytics - II

```
{
    word.set(tokenizer.nextToken());
    ;

    //sending to output collector which inturn passes the same to reducer
    output.collect(word, one);

}
}
```

### WordCountReducer.java(Reducer Class)

```
import java.io.IOException;
import java.util.Iterator;

import org.apache.hadoop.io.*; import org.apache.hadoop.mapred.*; public class
WordCountReducer extends MapReduceBase implements Reducer<Text,
IntWritable, Text, IntWritable>
{

    //reduce method accepts the Key Value pairs from mappers, do the aggregation based on
    keys and produce the final out put

    public void reduce(Text key, Iterator<IntWritable> values, OutputCollector<Text,
    IntWritable> output, Reporter reporter) throws IOException

    { int sum = 0;

        /*iterates through all the values available with a key and add them together and give the
        final result as the key and sum of its values*/ while
        (values.hasNext()) { sum += values.next().get();

        }
        output.collect(key, new IntWritable(sum));
    }
}
```



## Business Intelligence and Big Data Analytics - II

### Practical No. 5

**Aim:** Write a program to construct different types of k-shingles for given document.

**Installation of required packages before executing program:-**

```
install.packages("tm")
require("tm")
install.packages("devtools")
```

```
readinteger <- function()
{
    n      <- readline(prompt="Enter value of k-1:
") k<-as.integer(n)

    u1 <- readLines(E:/BA/Hadoop.txt")
    Shingle<-0

    i <-0
    while(i<nchar(u1)-k+1)
    {
        Shingle[i] <- substr(u1, start=i, stop=i+k)
        print(Shingle[i])

        i=i+1
    }
}
if(interactive()) readinteger()
```

**OutPut:-**

```
>      if(interactive()) readinteger()
Enter value of k-1: 2 character(0)
[1] "thi"
[1] "his"
[1] "is "
[1] "s i"
[1] " is"
[1] "is "
[1] "s a"
[1] " a "
[1] "a t"
[1] " te"
```

## Business Intelligence and Big Data Analytics - II

```
[1] "tex"  
[1] "ext"  
[1] "xt."
```

### OutPut:-

```
> if(interactive()) readinteger()  
Enter value of k-1: 3 character(0)  
[1] "this"  
[1] "his "  
[1] "is i"  
[1] "s is"  
[1] " is "  
  
[1] "is a"  
[1] "s a "  
[1] " a t"  
[1] "a te"
```

### OutPut:-

```
> if(interactive()) readinteger()  
Enter value of k-1: 4 character(0)  
[1] "this "  
[1] "his i"  
[1] "is is"  
[1] "s is "  
[1] " is a"  
[1] "is a "  
[1] "s a t"  
[1] " a te"  
[1] "a tex"  
[1] " text"  
[1] "text."  
[1] "ext. "
```

## Business Intelligence and Big Data Analytics - II

### Practical No. 6

**Aim : Write a program for measuring similarity among documents and detecting passages which have been reused.**

Installation of required packages before executing program:-

```
install.packages("tm")  
require("tm")
```

```
install.packages("ggplot2")
```

```
install.packages("textreuse")
```

```
install.packages("devtools") Source Code
```

```
l:my.corpus <- Corpus(DirSource("c:/msc/r-  
corpus"))
```

```
my.corpus <- tm_map(my.corpus, removeWords, stopwords("english"))  
my.tdm <- TermDocumentMatrix(my.corpus)  
#inspect(my.tdm)  
my.dtm <- DocumentTermMatrix(my.corpus, control = list(weighting =
```

```
weightTfIdf, stopwords = TRUE))  
#inspect(my.dtm)  
my.df <- as.data.frame(inspect(my.tdm))  
my.df.scale <- scale(my.df)
```

```
d <-  
dist(my.df.scale,method="euclidean") fit  
<hclust(d, method="ward") plot(fit)
```

**OutPut:-**

```
<<TermDocumentMatrix (terms: 69, documents: 6)>>
```

```
Non-/sparse entries: 97/317
```

```
Sparsity : 77%
```

```
Maximal term length: 12
```

```
Weighting : term frequency (tf)
```

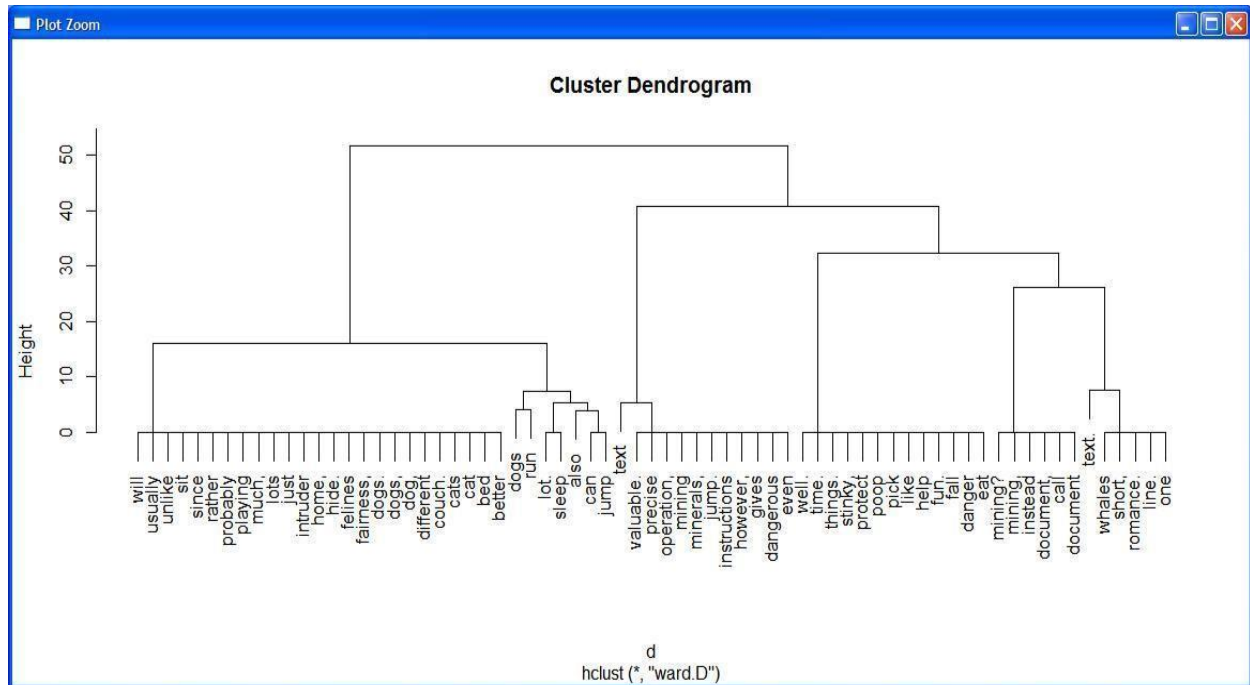
```
Docs
```

Terms	File1.txt	File2.txt	File3.txt	File4.txt	File5.txt	File6.txt
also	0	1	1	1	0	0
bed	0	0	0	1	0	0
better	0	0	0	1	0	0
call	0	1	0	0	0	0
can	0	0	1	1	0	0
cat	0	0	0	1	0	0
cats	0	0	0	1	0	0

## Business Intelligence and Big Data Analytics - II

```
couch.          0          0          0          1          0          0
> barplot(as.matrix(my.tdm))
> my.df.scale <- scale(my.df)
> d <- dist(my.df.scale,method="euclidean")
> fit <- hclust(d, method="ward")
```

The "ward" method has been renamed to "ward.D"; note new "ward.D2" >  
plot(fit)



Source code 2 (using bar plot with and without color):my.corpus

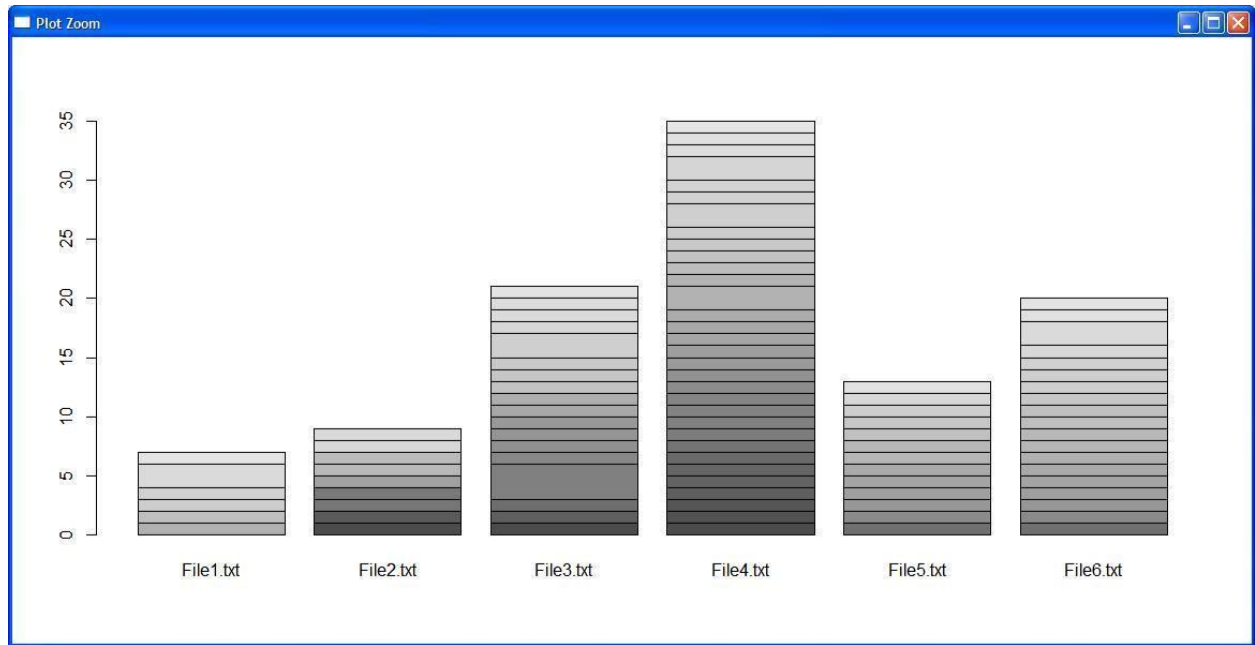
```
<- Corpus(DirSource("c:/msc/r-corpus"))

my.corpus <- tm_map(my.corpus, removeWords, stopwords("english"))
my.tdm <- TermDocumentMatrix(my.corpus)

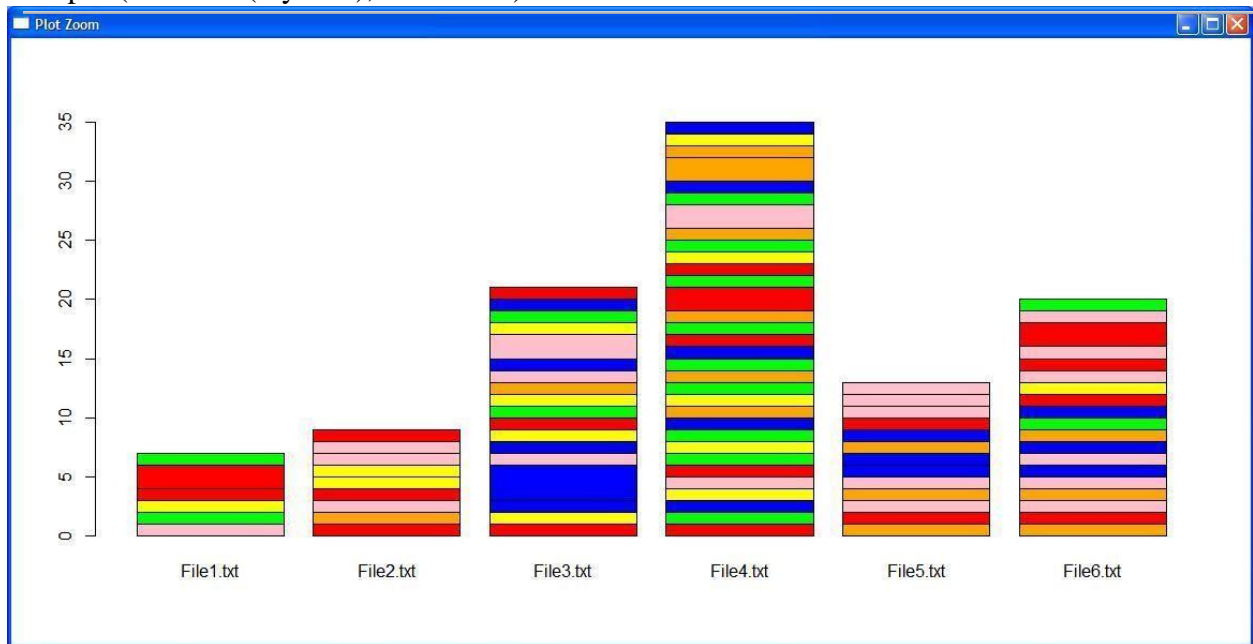
inspect(my.tdm) my.df <-
as.data.frame(inspect(my.tdm))
barplot(as.matrix(my.tdm))
#barplot(as.matrix(my.tdm),col = color)
```

## Business Intelligence and Big Data Analytics - II

OutPut:-



`barplot(as.matrix(my.tdm),col = color)`



Jaccard similarity

## Business Intelligence and Big Data Analytics - II

### Similarity of asymmetric binary attributes[[edit](#)]

Given two objects,  $A$  and  $B$ , each with  $n$  binary attributes, the Jaccard coefficient is a useful measure of the overlap that  $A$  and  $B$  share with their attributes. Each attribute of  $A$  and  $B$  can either be 0 or 1. The total number of each combination of attributes for both  $A$  and  $B$  are specified as follows:

represents the total number of attributes where  $A$  and  $B$  both have a value of 1.

represents the total number of attributes where the attribute of  $A$  is 0 and the attribute of  $B$  is 1.

represents the total number of attributes where the attribute of  $A$  is 1 and the attribute of  $B$  is 0.

represents the total number of attributes where  $A$  and  $B$  both have a value of 0.

Each attribute must fall into one of these four categories, meaning that

$$M_{11} + M_{01} + M_{10} + M_{00} = n.$$

The Jaccard similarity coefficient,  $J$ , is given as

$$J = \frac{M_{11}}{M_{01} + M_{10} + M_{11}}.$$

The Jaccard distance,  $d_J$ , is given as

$$d_J = \frac{M_{01} + M_{10}}{M_{01} + M_{10} + M_{11}} = 1 - J.$$

		<b>A</b>	
		0	1
<b>B</b>	0	$M_{00}$	$M_{10}$
	1	$M_{01}$	$M_{11}$

## Business Intelligence and Big Data Analytics - II

Source code 3 (using minhash and jaccard similarity):-

library(textreuse) **Source Code:** minhash <-

```
minhash_generator(200, seed = 235)
```

```
ats <- TextReuseCorpus(dir = "c:/msc/r-corpus", tokenizer = tokenize_ngrams, n = 5,  
minhash_func = minhash)
```

```
buckets <- lsh(ats, bands = 50, progress = interactive())
```

```
candidates <- lsh_candidates(buckets)
```

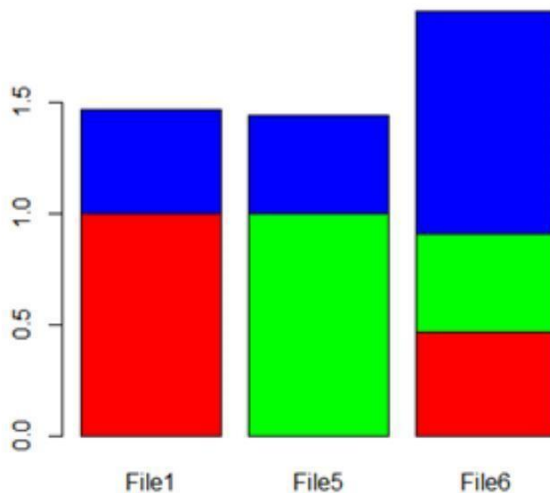
```
scores <- lsh_compare(candidates, ats, jaccard_similarity, progress =
```

```
FALSE) scores color <- c("red","green","blue","orange","yellow","pink")
```

```
barplot(as.matrix(scores),col = color)
```

### Output:

```
   a      b      score  
  <chr> <chr>  <dbl>  
1 File 1 File 6 0.4651163  
2 File 5 File 6 0.4418605
```





**Business Intelligence and Big Data Analytics - II**  
**Practical No. 7**

**Aim :** Write a program to compute the n-moment for a given stream where n is given.

Source Code:

```
import java.io.*;
import java.util.*;

class n_moment2
{
    public static void main(String args[])
    {
        int n=15;
        String stream[]={"a","b","c","b","d","a","c","d","a","b","d","c","a","a","b"};

        int zero_moment=0,first_moment=0,second_moment=0,count=1,flag=0;
        ArrayList<Integer> arrlist=new ArrayList();

        System.out.println("Arraylist elements are :: ");
        for(int i=0;i<15;i++)
        {
            System.out.print(stream[i]+" ");
        }

        Arrays.sort(stream);

        //Calculate Zeroth moment(calculates unique elements-raised to zero)
        for(int i=1;i<n;i++)
        {
            if(stream[i]==stream[i-1])
            { count++; }

            else
            {
                arrlist.add(count);
                count=1;
            }
        }
        arrlist.add(count);

        zero_moment=arrlist.size();
        System.out.println("\n\n\nValue of Zeroth moment for given stream ::"+zero_moment);
    }
}
```

## Business Intelligence and Big Data Analytics - II

//Calculate First moment(Calculate length of the stream-raised to one)

```
for(int i=0;i<arrlist.size();i++)  
{ first_moment+=arrlist.get(i);  
}
```

System.out.println("\n\nValue of First moment for given stream ::"+first\_moment);

//Calculate Second moment(raised to two)

```
for(int i=0;i<arrlist.size();i++)  
{  
  
    int j=arrlist.get(i);  
    second_moment+=(j*j);  
}
```

System.out.println("\n\nValue of Second moment for given stream ::"+second\_moment);

```
}  
}
```

### Output :

Arraylist elements are :: a  
b c b d a c d a b d c a a b

Value of Zeroth moment for given stream ::4

Value of First moment for given stream ::15

Value of Second moment for given stream ::59

**Business Intelligence and Big Data Analytics - II**  
**Practical No. 8**

**Aim : Write a program to demonstrate the Alon-Matias-Szegedy Algorithm for second moments.**

**Source Code:**

```
import java.io.*;
import java.util.*;
class AMSA
{ public static int findCharCount(String stream,char XE,int random,int n)

    { int countOccurance=0; for(int

        i=random;i<n;i++)

        { if(stream.charAt(i)==XE)

            { countOccurance++;//System.out.println(countOccurance+" "+i);

                }

        } return

        countOccurance;

    }

    public static int estimateValue(int XV1,int n)

    { int ExpValue;

        ExpValue=n*(2*XV1-1);

        return ExpValue;

    }

    public static void main(String args[])

    {

        int n=15;
```

## Business Intelligence and Big Data Analytics - II

```
String stream="abcbdacdabdcaab"; int

random1=3,random2=8,random3=13; char

XE1,XE2,XE3; int XV1,XV2,XV3; int

ExpValuXE1, ExpValuXE2, ExpValuXE3;

int apprSecondMomentValue; XE1=stream.charAt(random1-
1);
XE2=stream.charAt(random2-1);

XE3=stream.charAt(random3-1);

//System.out.println(XE1+" "+XE2+" "+XE3);

XV1=findCharCount(stream,XE1,random1-1,n);

XV2=findCharCount(stream,XE2,random2-1,n);

XV3=findCharCount(stream,XE3,random3-1,n);

System.out.println(XE1+"="+XV1+" "+XE2+"="+XV2+" "+XE3+"="+XV3);

ExpValuXE1=estimateValue(XV1,n);

ExpValuXE2=estimateValue(XV2,n);

ExpValuXE3=estimateValue(XV3,n);

System.out.println("Expected value for "+XE1+" is :: "+ExpValuXE1);

System.out.println("Expected value for "+XE2+" is :: "+ExpValuXE2);

System.out.println("Expected value for "+XE3+" is :: "+ExpValuXE3);

apprSecondMomentValue=(ExpValuXE1+ExpValuXE2+ExpValuXE3)/3;

System.out.println("Approximate Second moment value using
Alon-Matias-Szegedy is :: "+apprSecondMomentValue);

    }
}
```

**Output:**

## **Business Intelligence and Big Data Analytics - II**

c=3    d=2    a=2

Expected value for c is :: 75

Expected value for d is :: 45

Expected value for a is :: 45

Approximate Second moment value using Alon-Matias-Szegedy is :: 55