

Department of Computer Engineering

Experiment No.7

Data Visualization using Hive/PIG/R/Tableau/.

Date of Performance:04/09/2023

Date of Submission:11/09/2023



Department of Computer Engineering

Aim: Data Visualization using Hive/PIG/R/Tableau/.

Theory:

Data visualization is the technique used to deliver insights in data using visual cues such as graphs, charts, maps, and many others. This is useful as it helps in intuitive and easy understanding of the large quantities of data and thereby make better decisions regarding it. Data Visualization in R Programming Language

The popular data visualization tools that are available are Tableau, Plotly, R, Google Charts, Infogram, and Kibana. The various data visualization platforms have different capabilities, functionality, and use cases. They also require a different skill set. This article discusses the use of R for data visualization.

R is a language that is designed for statistical computing, graphical data analysis, and scientific research. It is usually preferred for data visualization as it offers flexibility and minimum required coding through its packages.

Consider the following *airquality* data set for visualization in R:

Ozone	Solar R.	Wind	Temp	Month	Day
41	190	7.4	67	5	1
36	118	8.0	72	5	2
12	149	12.6	74	5	3
18	313	11.5	62	5	4
NA	NA	14.3	56	5	5



Department of Computer Engineering

28	NA	14.9	66	5	6

1. Bar Plot

There are two types of bar plots- horizontal and vertical which represent data points as horizontal or vertical bars of certain lengths proportional to the value of the data item. They are generally used for continuous and categorical variable plotting. By setting the horiz parameter to true and false, we can get horizontal and vertical bar plots respectively.

Example 1:

Horizontal Bar Plot for

Ozone concentration in

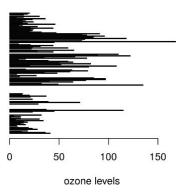
air barplot(airquality\$Ozone, main =

'Ozone Concenteration in air', xlab =

'ozone levels', horiz = TRUE)

Output:

Ozone Concenteration in air



Example 2:



Department of Computer Engineering

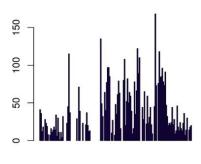
Vertical Bar Plot for

Ozone concentration in air

barplot(airquality\$Ozone, main = 'Ozone Concenteration in air',

xlab = 'ozone levels', col = 'blue', horiz = FALSE)

Ozone Concenteration in air



ozone levels

2. Histogram

A histogram is like a bar chart as it uses bars of varying height to represent data distribution. However, in a histogram values are grouped into consecutive intervals called bins. In a Histogram, continuous values are grouped and displayed in these bins whose size can be varied.

Example:

Histogram for Maximum Daily Temperature data(airquality)

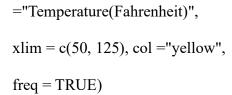
hist(airquality\$Temp, main ="La Guardia Airport's\

Maximum Temperature(Daily)",

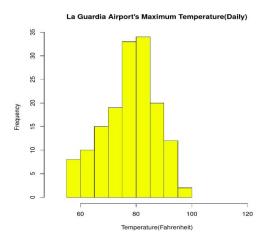
xlab



Department of Computer Engineering



Output:



3. Box Plot

The statistical summary of the given data is presented graphically using a boxplot. A boxplot depicts information like the minimum and maximum data point, the median value, first and third quartile, and interquartile range.

Example:

Box plot for average wind speed

data(airquality) boxplot(airquality\$Wind, main =

"Average wind speed\ at La Guardia Airport", xlab =

"Miles per hour", ylab =

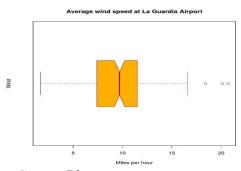
"Wind", col = "orange", border =

"brown", horizontal = TRUE, notch =

TRUE)



Department of Computer Engineering



4. Scatter Plot

A scatter plot is composed of many points on a Cartesian plane. Each point denotes the value taken by two parameters and helps us easily identify the relationship between them.

Example:.

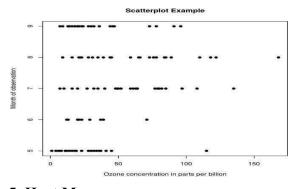
Scatter plot for Ozone Concentration per month

data(airquality) plot(airquality\$Ozone,

airquality\$Month, main ="Scatterplot Example",

xlab ="Ozone Concentration in parts per billion",

ylab =" Month of observation ", pch = 19)



5. Heat Map

Heatmap is defined as a graphical representation of data using colors to visualize the value of the matrix. heatmap() function is used to plot heatmap.

Syntax: heatmap(data)

Parameters: data: It represent matrix data, such as values of rows and columns

Return: This function draws a heatmap.

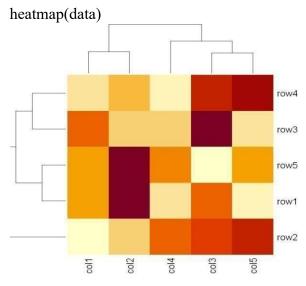


Department of Computer Engineering

- # Set seed for reproducibility
- # set.seed(110)
- # Create example data data <- matrix(rnorm(50, 0,
- 5), nrow = 5, ncol = 5) # Column names

rownames(data) <- paste0("row", 1:5)

Draw a heatmap



6. Map visualization in R

Here we are using maps package to visualize and display geographical maps using an R programming language.

- # Read dataset and convert it into
- # Dataframe data <-

read.csv("worldcities.csv") df <-

data.frame(data) # Load the

required libraries library(maps)

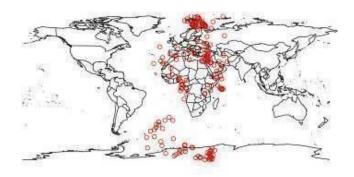


Department of Computer Engineering

map(database = "world") #

marking points on map

points(x = df lat[1:500], y = df lng[1:500], col = "Red") install.packages("maps")



7. 3D Graphs in R

Here we will use preps() function, This function is used to create 3D surfaces in perspective view. This function will draw perspective plots of a surface over the x-y plane.

Syntax: persp(x, y, z)

 $z \le -outer(x, y, cone)$

Parameter: This function accepts different parameters i.e. x, y and z where x and y are vectors defining the location along x- and y-axis. z-axis will be the height of the surface in the matrix z.

Return Value: persp() returns the viewing transformation matrix for projecting 3D coordinates (x, y, z) into the 2D plane using homogeneous 4D coordinates (x, y, z, t).

Adding Titles and Labeling Axes to Plot

```
cone <- function(x, y){ sqrt(x ^2 + y ^2)}

# prepare variables.

x <- y <- seq(-1, 1, length = 30)
```

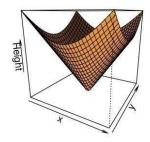


Department of Computer Engineering

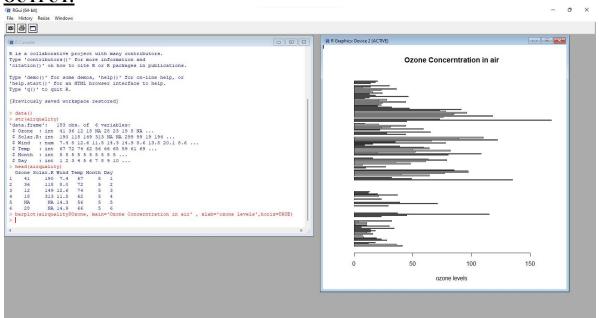
plot the 3D surface

Adding Titles and Labeling Axes to Plot persp(x, y, z, main="Perspective Plot of a Cone", zlab = "Height", theta = 30, phi = 15, col = "orange", shade = 0.4)

Perspective Plot of a Cone

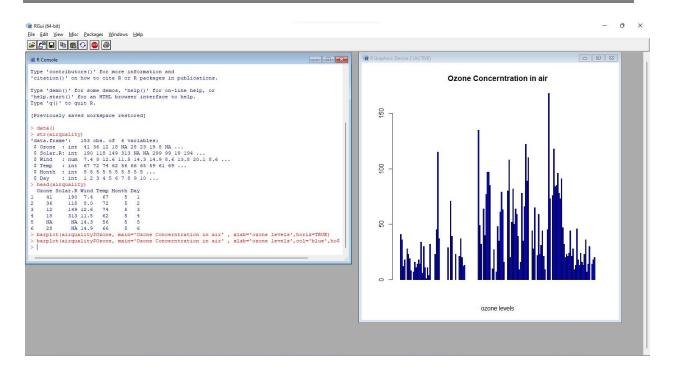


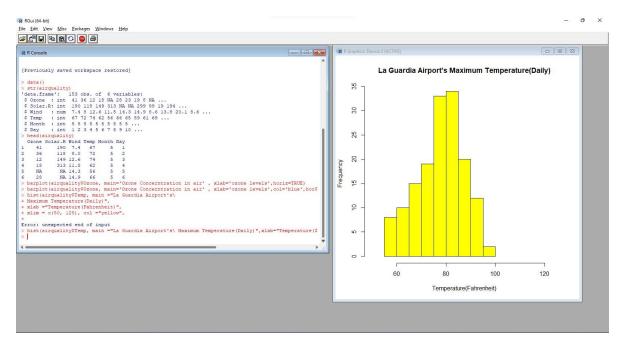
OUTPUT:





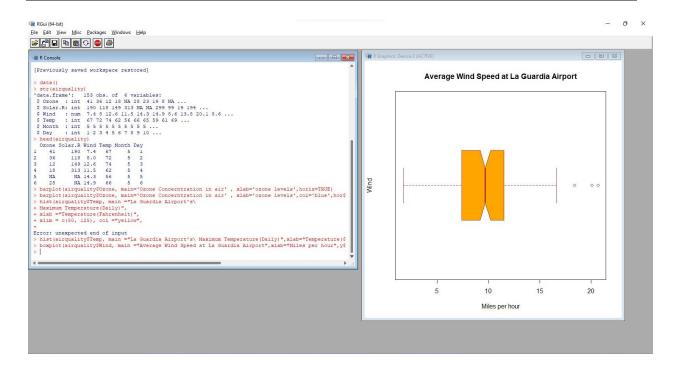
Department of Computer Engineering

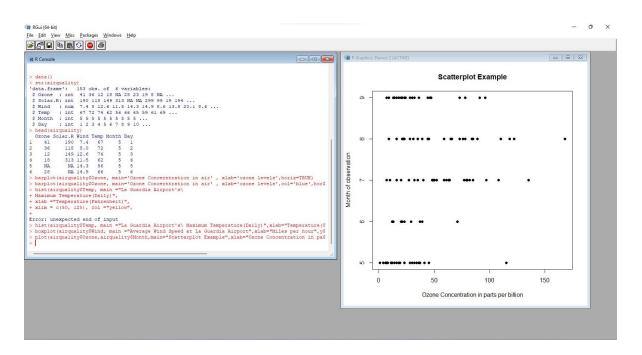






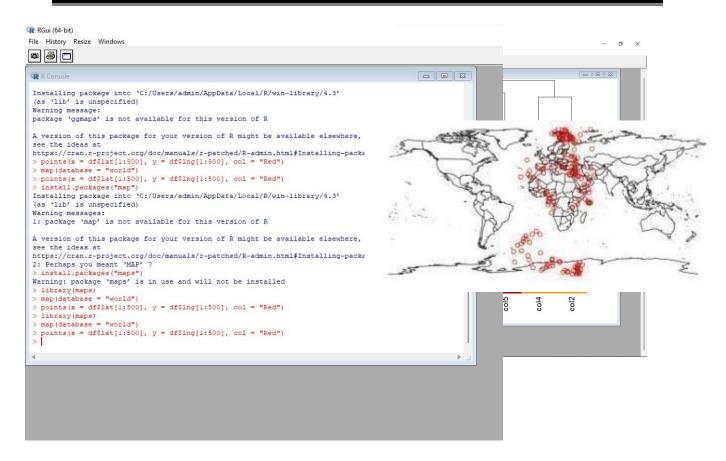
Department of Computer Engineering





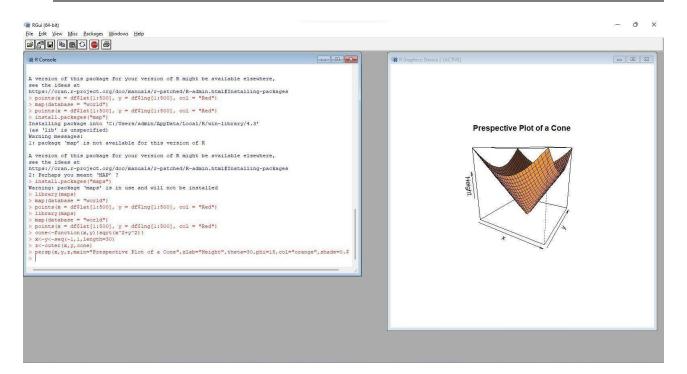


Department of Computer Engineering





Department of Computer Engineering



CONCLUSION:

The experiment demonstrating the Bloom filter highlighted its role as an efficient, memory-conserving data structure for approximate set membership queries. While effectively managing extensive datasets with minimal memory usage, it enables swift membership checks, albeit with controlled instances of false positives. Despite not supporting data deletion or precise counts, the Bloom filter finds practical use in applications like web caching, spell checkers, and network security, ultimately alleviating the demand on databases. Achieving optimal performance hinges on meticulous parameter selection. To sum it up, the Bloom filter proves its value in scenarios marked by limited memory resources and the need for rapid query processing, effectively balancing memory usage and false positive occurrences..