Aim: Write a Program to implement the basic matrix operations.

Theory:

Matrix operations involve mathematical operations performed on matrices, rectangular arrays of numbers arranged in rows and columns. Common operations include addition, subtraction, multiplication, and transposition, used in various fields such as linear algebra, statistics, computer graphics, and optimization, to solve systems of equations and analyze data. Multiplication, defined only when certain dimensions align, involves calculating the dot product between row vectors of the first matrix and column vectors of the second. Transposition simply flips rows and columns.

Source Code:

R-script

```
# Define two matrices matrix 1 < -matrix(c(1, 2, 3, 4, 5, 6), nrow =
2, byrow = TRUE) matrix2 <- matrix(c(7, 8, 9, 10, 11, 12), nrow
= 2, byrow = TRUE)
# Display the matrices
print("Matrix 1:")
print(matrix1)
print("Matrix 2:")
print(matrix2)
# Addition of matrices
addition result <- matrix1 + matrix2
print("Addition of matrices:")
print(addition result)
# Subtraction of matrices
subtraction result <- matrix1 - matrix2
print("Subtraction of matrices:")
print(subtraction result)
```

Multiplication of matrices

```
multiplication_result <- matrix1 %*% t(matrix2) # Matrix multiplication requires transpose of the second matrix
print("Multiplication of matrices:")
print(multiplication_result)

# Transpose of matrices
transpose_matrix1 <- t(matrix1)
transpose_matrix2 <- t(matrix2)
print("Transpose of Matrix 1:")
print(transpose_matrix1)
print("Transpose of Matrix 2:")
print("Transpose matrix2)
```

```
> source("c:\\Users\\ARYAN\\PycharmProjects\\gyrados\\new.r", encoding = "UTF-$
[1] "Matrix 1:"
[1] "Matrix 2:"
    [,1] [,2] [,3]
[1,]
[2,] 10 11 12
[1] "Addition of matrices:"
[1,] 8 10 12
[2,] 14 16 18
[1] "Subtraction of matrices:"
    [,1] [,2] [,3]
[1,] -6 -6 -6
[2,] -6 -6 -6
[1] "Multiplication of matrices:"
    [,1] [,2]
[1,] 50 68
[2,] 122 167
[1] "Transpose of Matrix 1:"
[2,]
[1] "Transpose of Matrix 2:"
     [,1][,2]
       7 10
            11
```

Aim: Write a Program to compute descriptive statistics such as mean, median, mode, standard deviation, and fractiles such as percentiles

Theory:

Mean: Average value of dataset, sensitive to outliers.

Median: Middle value when data is sorted, less affected by outliers.

Mode: Most frequent value in dataset.

Standard Deviation: Measure of data dispersion around mean.

Percentiles: Values dividing dataset into 100 equal parts, useful for analyzing data distribution.

Source Code:

R-script

```
# Define the dataset data <- c(23, 45, 67, 89, 12,
34, 56, 78, 90, 10, 22)
# Compute mean mean value
<- mean(data) # Compute
median median value <-
median(data)
# Compute mode
mode value <- names(sort(table(data), decreasing = TRUE))[1]
# Compute standard deviation
sd value <- sd(data)
# Compute percentiles (25th, 50th, and 75th
percentiles) percentile 25 <- quantile(data, 0.25)
percentile 50 <- quantile(data, 0.50) percentile 75 <-
quantile(data, 0.75) # Print the descriptive statistics
print("Descriptive Statistics:") print(paste("Mean:",
mean value)) print(paste("Median:", median value))
```

```
print(paste("Mode:", mode_value))
print(paste("Standard Deviation:", sd_value))
print("Percentiles:") print(paste("25th Percentile:",
percentile_25)) print(paste("50th Percentile (Median):",
percentile_50)) print(paste("75th Percentile:",
percentile_75))
```

```
> source("c:\Users\\ARYAN\\PycharmProjects\\gyrados\\new.r", encoding = "UTF-$
[1] "Descriptive Statistics:"
[1] "Mean: 47.8181818181818"
[1] "Median: 45"
[1] "Mode: 10"
[1] "Standard Deviation: 30.0260492966297"
[1] "Percentiles:"
[1] "25th Percentile: 22.5"
[1] "50th Percentile: 72.5"
```

Aim: Write a program to calculate Correlation Coefficient using R

Theory:

The correlation coefficient measures the strength and direction of the linear relationship between two variables. It ranges from -1 to 1, where 1 indicates a perfect positive correlation, -1 indicates a perfect negative correlation, and 0 indicates no correlation. Positive values signify direct relationships, while negative values signify inverse relationships

```
Source Code:
R-script
# Set working directory (optional, but recommended for organized projects)
# setwd("/path/to/your/project/directory") # Replace with your actual directory path
# The program evaluates Pearson's correlation coefficient
xdata < c(2,4.4,3,3,2,2.2,2,4)
ydata <- c(1,4.4,1,3,2,2.2,2,7)
# Covariance and correlation using base R
functions print('The covariance is:')
print(cov(xdata, ydata))
print('The correlation coefficient, rho, using cov function
is:') rho <- cov(xdata, ydata) / (sd(xdata) * sd(ydata))
print(rho)
# Correlation using cor function print('The
correlation using cor function is:')
print(cor(xdata, ydata))
```

Scatter plot plot(xdata, ydata, pch=13, cex=1.5, main = "Scatter Plot of xdata vs ydata")

Correlation matrix with corrplot package

Add main title

```
library(corrplot) corr_matrix <- cor(mtcars) # Store correlation
matrix for later use corrplot(corr_matrix, main = "Correlation
Matrix of mtcars") # Add main title
```

Correlation heatmap with PerformanceAnalytics package
library(PerformanceAnalytics) chart.Correlation(corr_matrix, main = "Correlation
Heatmap of mtcars") # Add main title

```
dev.new() # Clear any existing plots
plot(xdata, ydata, pch=13, cex=1.5)
dev.hold() # Overlap plots on the same device
```

Output:

```
> xdata <- c(2,4.4,3,3,2,2.2,2,4)
> ydata <- c(1,4.4,1,3,2,2.2,2,7)
 print('The covariance is:')
[1] "The covariance is:"
> print(cov(xdata,ydata))
[1] 1.479286
 print('The correlation coefficent, rho, using cov function is:')
[1] "The correlation coefficent, rho, using cov function is:"
> rho = cov(xdata,ydata)/(sd(xdata)*sd(ydata)) # Computes the correlation using
cov (covariance) # and sd (standard deviation) function
> print(rho)
[1] 0.7713962
> print('The correlation using cor function is:')
[1] "The correlation using cor function is:"
> print(cor(xdata,ydata)) # Computes the correlation coefficient using cor
function #We can plot these bivariate observations as a coordinate-based plot
(a scatterplot). #Executing the following gives figure between x and y data
[1] 0.7713962
[1] 32
```

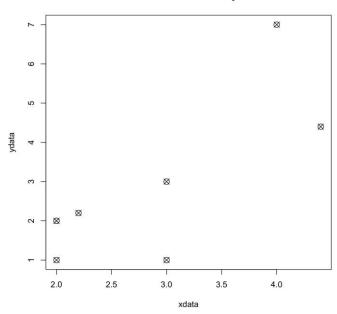
```
Correlation Matrix:
```

```
> print(correlation_matrix)
           [,1]
                      [,2]
                                 [,3]
[1,] 0.11216160 -0.09679700 0.11216160 -0.010755222 0.05070319 0.038411508
[2,] -0.21412670 0.18479428 -0.21412670
                                       0.020532697 -0.09679700
                                                             -0.073331062
[3,] -0.02379186  0.02053270 -0.02379186
                                       0.002281411 -0.01075522 -0.008147896
[4,] -0.02379186  0.02053270 -0.02379186
                                       0.002281411 -0.01075522 -0.008147896
[5,] 0.11216160 -0.09679700
                           0.11216160 -0.010755222
                                                  0.05070319
                                                              0.038411508
[6,] 0.08497091 -0.07333106
                           0.08497091 -0.008147896
                                                   0.03841151
                                                              0.029099628
    0.11216160 -0.09679700
                           0.11216160 -0.010755222
                                                   0.05070319
                                                              0.038411508
[8,] -0.15974532 0.13786240
                           [,7]
                      [,8]
[1,] 0.05070319
               -0.25658888
[2,] -0.09679700 0.48985149
[3,] -0.01075522 0.05442794
[4,] -0.01075522 0.05442794
[5,] 0.05070319 -0.25658888
[6,] 0.03841151 -0.19438551
    0.05070319 -0.25658888
[8,] -0.07221364 0.36544476
```

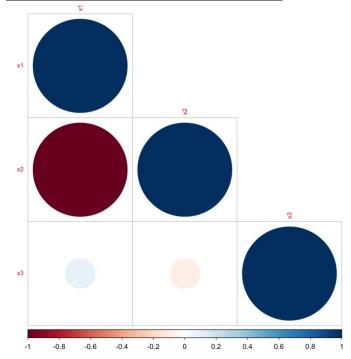
- + print("\nNo variables with absolute correlation greater than 0.5 found.")
- + }
- [1] "\nNo variables with absolute correlation greater than 0.5 found."

SCATTER PLOT

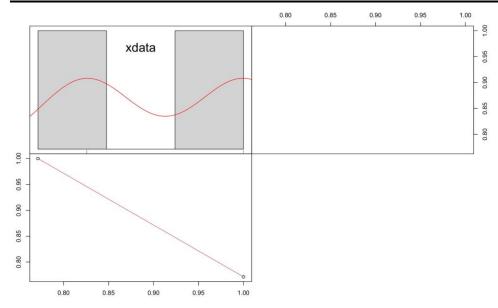
Scatter Plot of xdata vs ydata



CORRELATION MATRIX PLOT



CORRELATION CHART USING PERFORMANCE ANALYTICS



Aim: Write a program to present the data as a frequency table. Find the outliers in a dataset.

Theory:

A frequency table is a tabular summary of the frequency of occurrences of values or categories in a dataset, displaying the count or percentage of each value or category.

Outliers are data points that significantly deviate from the rest of the dataset, often indicating anomalies, errors, or rare events, and are typically identified using statistical methods such as the interquartile range.

Source Code:

```
R-script
# Sample data for age age <- c(25, 32, 28, 41,
20, 35, 23, 18, 45, 62)
# Frequency tables with different binning strategies
# Unequal intervals (adjust number of bins as needed) break interval <-
seq(from=min(age),to=max(age), length=9) age freq <- cut(age,
breaks=break interval,right=TRUE, include.lowest = TRUE) table(age freq) #
Print frequency table
# Equal intervals (adjust number of bins as needed) break interval2 <-
seq(from=min(age),to=max(age), length=5) age freq2 <- cut(age,
breaks=break interval2,right=TRUE, include.lowest = TRUE) table(age_freq2) #
Print frequency table
```

Histogram with breakpoints

hist(age, breaks=break interval)

lines(density(age), lwd=3, col="blue")

Density plot hist(age, freq=FALSE)

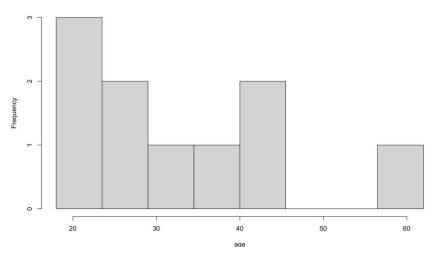
```
# Frequency table by group (replace with your own data structure)
# This example demonstrates by team and position team <-
c("A", "A", "A", "B", "B", "B", "C", "C", "C", "D")
position <- c("Manager", "Engineer", "Manager", "Sales", "Marketing",
"Director", "Analyst", "Analyst", "Intern")
df <- data.frame(team, position)
library(dplyr)
df %>% group by(team, position) %>% summarize(Freq=n())
# Boxplot for age distribution
boxplot(age, ylab="Age(years)")
# Identifying outliers (example with sample data) outliers
< c(0.6,-0.6,0.1,-0.2,-1.0,0.4,0.3,-1.8,1.1,6.0,-10)
plot(outliers,rep(0,length(outliers)),yaxt="n",ylab="",bty="n",cex=2,cex.axis=1.5,cex.lab=1.5
) abline(h=0,col="gray",lty=2)
arrows(length(outliers), 0.5, length(outliers) + 0.1, 0.1, lwd=2)
text(length(outliers)+0.2,0.7,labels="outlier",cex=3)
```

Frequency table by group

```
> df %-% group_by(team, position) %-% summarize(Freq=n())
`summarise()` has grouped output by 'team'. You can override using the
`.groups` argument.
# A tibble: 8 x 3
# Groups: team [4]
team position Fr
<chr> <chr> <ir
                                <int>
1 A
              Engineer
2 A
              Manager
3 B
4 B
5 B
              Manager
              Marketing
              Sales
6 C
7 C
8 D
              Analyst
                                       2
              Director
                                       1
              Intern
```

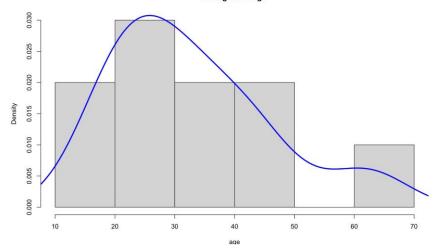
HISTOGRAM



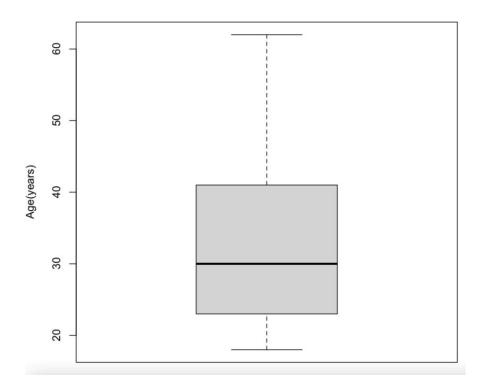


DENSITY PLOT

Histogram of age



BOX PLOT-



OUTLIERS-





Aim: Write a program to implement concepts of probability and distributions in R.

Theory:

Probability distributions describe the likelihood of outcomes in a dataset or random experiment. Common distributions include the normal distribution, representing symmetrical data; the Poisson distribution, modeling rare events; and the binomial distribution, for binary outcomes. They help analyze uncertainty and make predictions in various fields, including statistics and finance.

Source Code:

R-script

Binomial Distribution

```
# Probability of getting 5 successes in 8 trials with success probability 1/6
x prob binomial <- choose(8, 5) * (1/6)^5 * (1 - 1/6)^3 # Probability of
each outcome (0 to 8 successes) all probs <- sapply(0:8, function(x)
choose(8, x) * (1/6)^x * (1 - 1/6)^(8-x)
# Print all probabilities and their sum (should be close to
1) cat("Probabilities of each outcome (0-8 successes):\n")
print(all probs) cat("Sum of probabilities (should be close
to 1):\n") print(sum(all probs)) # Handles potential NAs
# Round probabilities to 3 decimal places and create a barplot
rounded probs <- round(all_probs, 3)</pre>
barplot(rounded probs, names.arg = 0.8, space = 0, xlab = "x", ylab = "Pr(X = x)", col =
"green")
# Cumulative probability of at most 3 successes p less than 3 <-
sum(all probs[1:4]) # Sum probabilities from 0 to 3 successes cat("Cumulative
probability (at most 3 successes):\n") print(p less than 3) # This represents
Pr(X \le 3)
```

Poisson Distribution

```
Probability of 3 occurrences with lambda = 3.22 lambda <- 3.22 prob_3_occurrences <- exp(-lambda) * (lambda^3) / factorial(3) Probabilities of 0 to 10 occurrences
```

```
(lambda^x) / factorial(x))
Round probabilities to 3 decimal places
rounded probs poisson <- round(all probs poisson, 3)
#Barplot with adjusted y-axis limits for better visualization
barplot(rounded probs poisson, ylim = c(0, 0.25), space = 0, names.arg = 0:10, ylab =
Pr(X=x), xlab = x, col = blue
# Cumulative probability of at most 2 occurrences p less than 2 <-
sum(all probs poisson[1:3]) # Sum probabilities from 0 to 2 occurrences cat("Cumulative
probability (at most 2 occurrences):\n") print(p less than 2) # This represents Pr(X \le 2)
Uniform Distribution min <- 0 max <- 100
# Sequence of x-values for the uniform distribution function
xpos \le seq(min, max, by = 0.5)
# Removed commented-out code for demonstration purposes
# Calculate the cumulative distribution function (CDF)
punif value <- ((xpos - min) / (max - min))</pre>
cat("Cumulative probability (X \le 15):\n")
print(punif value[31]) # Probability for X = 15 (index 31)
Normal Distribution
```

Generate sequence of numbers from -15 to +10 x <- seq(-15, 10, length = 101)

Increased number of points for smoother curve # Calculate the probability

density function (PDF) assuming unknown mean and standard deviation

all probs poisson <- sapply(0:10, function(x) exp(-lambda) *

```
pdf_normal <- (1 / (sqrt(2 * pi) * sd(x))) * exp(-((x - mean(x))^2) / (2 * (sd(x))^2))

# Plot the PDF

plot(x, pdf_normal, type = "l", lwd = 2, xlab = "x", ylab = "Density", main = "Normal Distribution PDF")

# Calculate the CDF using an empirical approach cdf_normal <- sapply(x, function(val) sum(x <= val) / length(x))

# Plot the CDF

plot(x, cdf_normal, type = "l", lwd = 2, xlab = "x", ylab = "Cumulative Probability", main = "Normal Distribution CDF (empirical)")

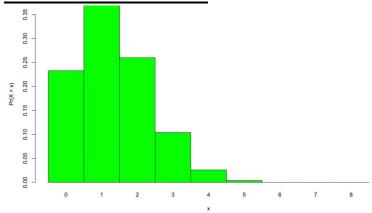
# QQ-Plot for normality check

norm_samp <- rnorm(100) qq
```

BINOMIAL DISTRIBUTION

```
> #### Binomial Distribution
> # Probability of getting 5 successes in 8 trials with success probability 1/6
> x_prob_binomial <- choose(8, 5) * (1/6)^5 * (1 - 1/6)^3</pre>
> # Probability of each outcome (0 to 8 successes)
> all_probs <- sapply(0:8, function(x) choose(8, x) * (1/6)^x * (1 - 1/6)^(8-
> # Print all probabilities and their sum (should be close to 1)
> cat("Probabilities of each outcome (0-8 successes):\n")
Probabilities of each outcome (0-8 successes):
> print(all_probs)
[1] 2.325680e-01 3.721089e-01 2.604762e-01 1.041905e-01 2.604762e-02 [6] 4.167619e-03 4.167619e-04 2.381497e-05 5.953742e-07
> cat("Sum of probabilities (should be close to 1):\n")
Sum of probabilities (should be close to 1):
> print(sum(all_probs)) # Handles potential NAs
[1] 1
successes
> cat("Cumulative probability (at most 3 successes):\n")
Cumulative probability (at most 3 successes):
> print(p_less_than_3) # This represents Pr(X <= 3)
[1] 0.9693436
```

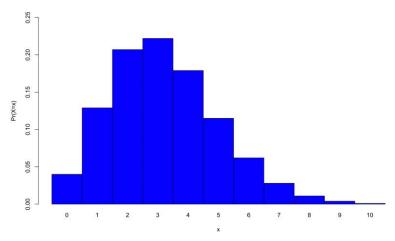
BAR PLOT BINOMIAL-



POISSON DISTRIBUTION-

```
> # Cumulative probability (at most 2 occurrences)
> p_less_than_2 <- sum(all_probs_poisson[1:3])
> 
> # Print cumulative probability
> cat("Cumulative probability (at most 2 occurrences):\n")
Cumulative probability (at most 2 occurrences):
> print(p_less_than_2)
[1] 0.3757454
```

BAR PLOT POISSON DISTRIBUTION-

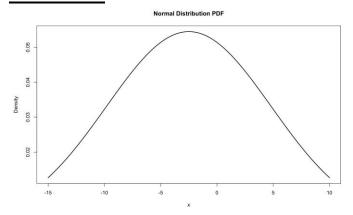


UNIFORM DISTRIBUTION-

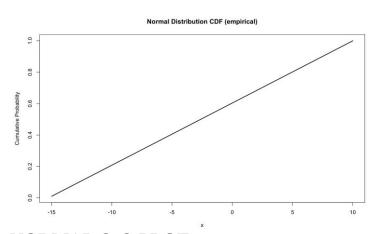
```
> # Sequence of x-values for the uniform distribution function
> xpos <- seq(min, max, by = 0.5)
> # Removed commented-out code for demonstration purposes
> # Calculate the cumulative distribution function (CDF)
> punif_value <- ((xpos - min) / (max - min))
> cat("Cumulative probability (X <= 15):\n")
Cumulative probability (X <= 15):
> print(punif_value[31]) # Probability for X = 15 (index 31 [1] 0.15
```

NORMAL DISTRIBUTION-

PDF PLOT-



CDF PLOT-



NORMAL Q-Q PLOT

