Experiment -6

- 1. Use Principle component analysis for the given data set and find the better prediction model for quantitative data.
- 2. For the given data, based on the scree plot, find out the number of principle components and in principle component analysis.
- 3. Plot Principle Component Analysis output and explain the visualization part of the dimension reduction techniques.

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Program:
# Load Packages and prepare dataset
library(TH.data)
library(caret)
data("GlaucomaM", package = "TH.data")
trainData <- GlaucomaM
View(trainData)
dim(trainData)
head(trainData)
str(trainData)
set.seed(100)
options(warn = -1)
subsets <- c(1:5, 10, 15, 18)
ctrl <- rfeControl(functions = rfFuncs,
          method = "repeatedcv",
          repeats = 5,
          verbose = FALSE)
```

```
ImProfile <- rfe(x = trainData[, c(1:3, 5:13)], y=trainData$ozone_reading,</pre>
         sizes = subsets,
         rfeControl = ctrl)
ImProfile
input = read.csv("H:\\JGi Classes\\Dimentionality Reduction and Model Validation VII Sem 2018-22\\Lab
Programs\\data files\\iris.csv")
names(input)
str(input)
model = prcomp(input[,1:4], scale=TRUE)
model$sdev
model$rotation
model$center
model$scale
par(mfrow=c(2,2))
plot(model$x[,1], col=input[,5])
plot(model$x[,2], col=input[,5])
plot(model$x[,3], col=input[,5])
plot(model$x[,4], col=input[,5])
model$sdev^2 / sum(model$sdev^2)
plot(model)
```

```
## PCA without 'prcomp'
## Normalize the input feature.
input$sepal_len1 = (input$sepal_len - mean(input$sepal_len) )/sd(input$sepal_len)
input$sepal_wid1 = (input$sepal_wid - mean(input$sepal_wid))/sd(input$sepal_wid)
input$petal_len1 = (input$petal_len - mean(input$petal_len))/sd(input$petal_len)
input$petal_wid1 = (input$petal_wid - mean(input$petal_wid))/sd(input$petal_wid)

##Get the covarience matrix and eigen vector.
matrix_form = matrix(c(input$sepal_len1, input$sepal_wid1, input$petal_len1, input$petal_wid1),
ncol=4)

m = cov(matrix_form)
eigenV = eigen(m)
eigenV$vectors

Output:
```