

Experiment No. : ..... 1 .....

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1. Write a program to implement Random forest using any data sets

data(iris)

str(iris)

install.packages ("caTools")

install.packages ("randomForest")

library(caTools)

library(randomForest)

split ← sample.split (iris, SplitRatio = 0.7)

split

train ← subset(iris, split == "TRUE")

test ← subset (iris, split == "FALSE")

set.seed(120)

classifier\_RF = randomForest(x = train[ -5 ],  
y = train\$species,  
ntree = 500)

classifier\_RF

~~classifier\_RF, newdata = test[ -5 ]~~

y\_pred = predict ( classifier\_RF, newdata = test [ -5 ] )

confusion\_mtx = table (test[, 5], y\_pred)

confusion\_mtx

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plot(classifier - RF)

importance(classifier - RF)

varImp Plot(classifier - RF)

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```
> classifier_RF  
call:  
randomForest(x = train[-5], y = train$Species, ntree = 500)  
Type of random forest: classification  
Number of trees: 500  
No. of variables tried at each split: 2
```

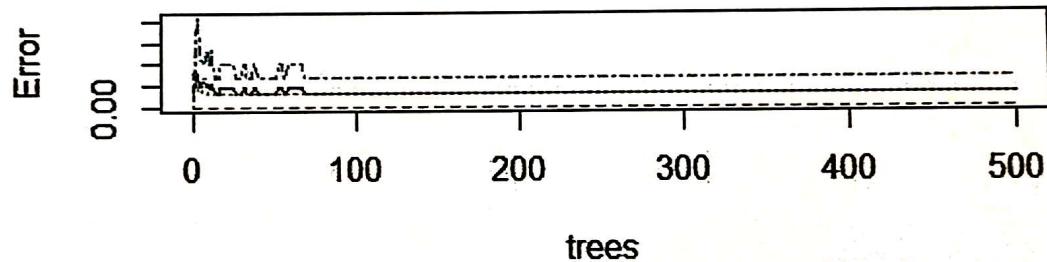
OOB estimate of error rate: 3.33%

Confusion matrix:

	setosa	versicolor	virginica	class.error
setosa	30	0	0	0.00000000
versicolor	0	29	1	0.03333333
virginica	0	2	28	0.06666667

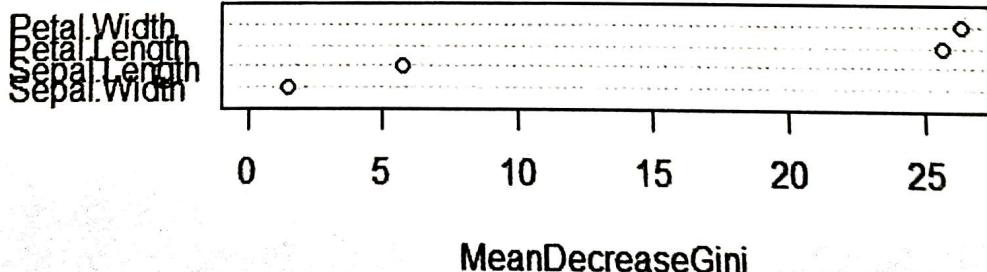
	setosa	versicolor	virginica
setosa	20	0	0
versicolor	0	20	0
virginica	0	3	17

## classifier\_RF



```
> importance(classifier_RF)  
MeanDecreaseGini  
Sepal.Length      5.772467  
Sepal.Width       1.419426  
Petal.Length     25.728012  
Petal.Width      26.374384
```

## classifier\_RF



3. Implement any clustering technique

Hierarchical clustering

```

##install.packages ("dplyr")
##install.packages ("ggplot2")
##install.packages ("ggfortify")
library ("ggplot2")
library ("dplyr")
library ("ggfortify")

summary (iris)
head(iris)
data <- select (iris, c(1:4))

wssplot <- function(data, nc=15, seed=1234) {
  wss <- (nrow(data)-1) * sum(apply(data, 2, var))
  for (i in 2:nc) {
    set.seed(seed)
    wss[i] <- sum(kmeans(data, centers=i)$withinss)
  }
  plot(1:nc, wss, type="b", xlab="No. of clusters",
       ylab="within groups sum of squares")
  wss
}

```

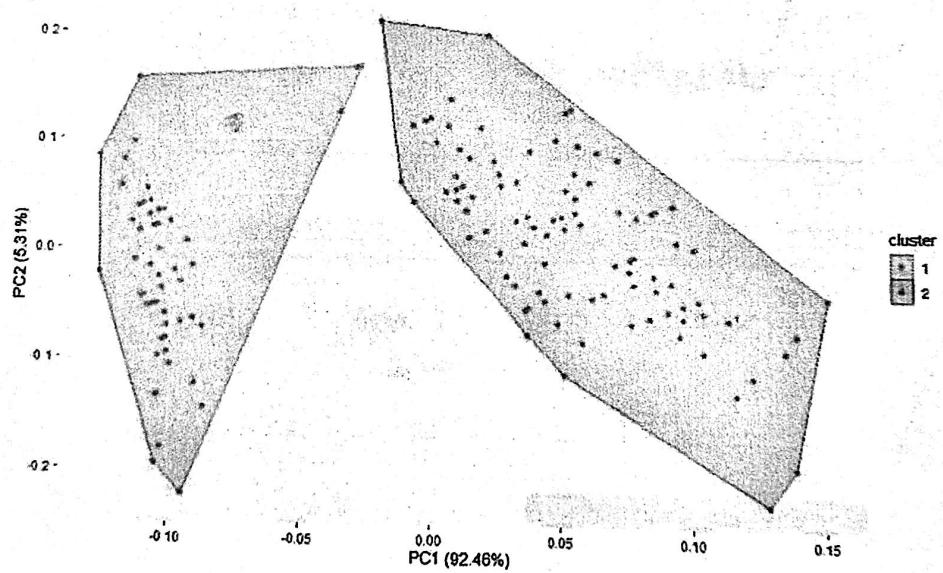
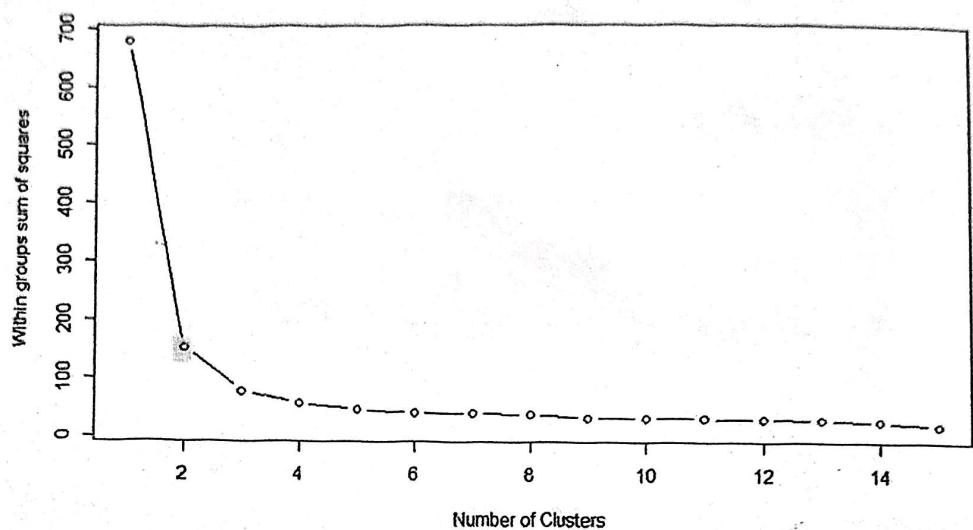
kmean <- kmeans (data, 2)

kmean\$centers

autoplot (kmean, data, frame = TRUE),

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## 4. Implement linear and logistic regression

## Logistic Regression

```
library(dplyr)
```

```
summary(mtcars)
```

```
install.packages("ROCR")
```

```
library(caret)
```

```
library(ROCR)
```

```
split ← sample.split(mtcars, splitratio = 0.8)  
split
```

```
train_reg ← subset(mtcars, split == "TRUE")
```

```
test_reg ← subset(mtcars, split == "FALSE")
```

```
logistic_model ← glm(vs ~ wt + disp, data = train_reg,  
family = "binomial")
```

```
logistic_model
```

```
summary(logistic_model)
```

```
predict_reg ← predict(logistic_model, test_reg,  
type = "response")
```

```
predict_reg
```

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~~predict\_reg <- ifelse(predict\_reg > 0.5, 1, 0)~~

~~table(test\_reg\$vs, predict\_reg)~~

~~missing\_classerr <- mean(predict\_reg != test\_reg\$vs)~~  
~~print(paste('Accuracy = ', 1 - missing\_classerr))~~

~~ROCPred <- prediction(predict\_reg, test\_reg\$vs)~~

~~ROCPer <- performance(ROCPred, measure = "tpr",~~  
~~x.measure = "fpr")~~

~~AUC <- performance(ROCPred, measure = "auc")~~

~~AUC <- AUC@y.values[[1]]~~

~~AUC~~

~~plot(ROCPer)~~

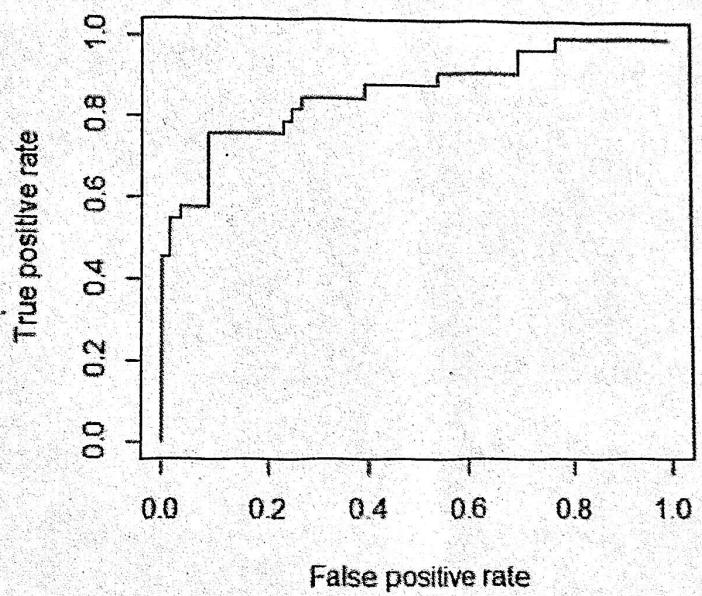
~~plot(ROCPer, colorize = TRUE,~~  
~~print.cutoffs.at = seq(0.1, by = 0.1),~~  
~~main = "ROC CURVE")~~

~~abline(a = 0, b = 1)~~

~~AUC <- round(AUC, 4)~~

~~legend(0.6, 0.4, AUC, title = "AUC", aux = 1)~~

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## # Linear Regression

```
Advertising ← read.csv("c:/users/ABC/downloads/  
Advertising.csv")
```

```
view(Advertising)
```

```
summary(Advertising)
```

```
head(Advertising)
```

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

```
library(ggplot2)
```

```
ggplot(Advertising, aes(x=TV, y=sales)) +  
  geom_point() + stat_smooth()
```

```
cor(Advertising$sales, Advertising$TV)
```

```
model ← lm(sales~TV, data = Advertising)
```

```
model
```

```
ggplot(Advertising, aes(TV, sales)) + geom_point() +  
  stat_smooth(method = lm)
```

```
summary(model)
```

```
confint(model)
```

```
Sigma(model) * 100 / mean(Advertising$sales)
```



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### **Height & Weight Regression**

