

# Assignment 3

## R Markdown

```
prc <- read.csv("prostate_cancer.csv", stringsAsFactors = FALSE) #This command imports the required data set and saves it to the prc data frame.
```

```
#stringsAsFactors command helps to convert every character vector to a factor wherever it makes sense.
```

```
str(prc) #We use this command to see whether the data is structured or not.
```

```
## 'data.frame':    100 obs. of  10 variables:
## $ id           : int  1 2 3 4 5 6 7 8 9 10 ...
## $ diagnosis_result : chr  "M" "B" "M" "M" ...
## $ radius        : int  23 9 21 14 9 25 16 15 19 25 ...
## $ texture        : int  12 13 27 16 19 25 26 18 24 11 ...
## $ perimeter      : int  151 133 130 78 135 83 120 90 88 84 ...
## $ area           : int  954 1326 1203 386 1297 477 1040 578 520 476 ...
## $ smoothness     : num  0.143 0.143 0.125 0.07 0.141 0.128 0.095 0.119 0.127 0.1
19 ...
## $ compactness    : num  0.278 0.079 0.16 0.284 0.133 0.17 0.109 0.165 0.193 0.2
4 ...
## $ symmetry       : num  0.242 0.181 0.207 0.26 0.181 0.209 0.179 0.22 0.235 0.20
3 ...
## $ fractal_dimension: num  0.079 0.057 0.06 0.097 0.059 0.076 0.057 0.075 0.074 0.0
82 ...
```

```
View(prc)
```

```
prc <- prc[-1] # removes the first variable(id) from the data set.
```

```
table(prc$diagnosis_result) # it helps us to get the numbers of patients
```

```
##
##  B  M
## 38 62
```

```
prc$diagnosis <- factor(prc$diagnosis_result, levels = c("B", "M"), labels = c("Benign", "Malignant"))
```

```
View(prc$diagnosis)
```

```
round(prop.table(table(prc$diagnosis)) * 100, digits = 1) # it gives the result in the percentage form rounded off to 1 decimal place( and so it's digits = 1)
```

```
##  
##      Benign Malignant  
##      38          62
```

```
normalize <- function(x) {  
  return ((x - min(x)) / (max(x) - min(x))) } # Normalization function
```

```
prc_n <- as.data.frame(lapply(prc[2:9], normalize)) # Normalizing values in data set
```

```
View(prc_n)
```

```
summary(prc_n$radius) # summary function to get the mean, median, mode and other statistics
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
## 0.0000 0.1875 0.5000 0.4906 0.7500 1.0000
```

```

prc_train <- prc_n[1:65,]    # creating training data set
prc_test  <- prc_n[66:100,]  # creating test data set

prc_train_labels <- prc[1:65, 1]
prc_test_labels <- prc[66:100, 1] #This code takes the diagnosis factor in column 1 of the prc data frame and on turn creates prc_train_labels and prc_test_labels data frame.

#install.packages('class')
library(class)

View(prc_test_labels)
prc_test_pred <- knn(train = prc_train, test = na.omit(prc_test),cl = prc_train_labels, k=10) # Knn function to classify dataset

View(prc_test_pred)

#install.packages('gmodels')
library(gmodels)
CrossTable(x= na.omit(prc_test_labels), y = prc_test_pred,prop.chisq= FALSE) # check the accuracy of the predicted values in prc_test_pred as to whether they match up with the known values in prc_test_label using CrossTable function.

```

```
##
##
##   Cell Contents
## |-----|
## |                N |
## |      N / Row Total |
## |      N / Col Total |
## |      N / Table Total |
## |-----|
##
##
## Total Observations in Table:  35
##
##
##               | prc_test_pred
## na.omit(prc_test_labels) |      B |      M | Row Total |
## -----|-----|-----|-----|
##               B |      8 |     11 |      19 |
##               |    0.421 |    0.579 |    0.543 |
##               |    0.889 |    0.423 |          |
##               |    0.229 |    0.314 |          |
## -----|-----|-----|-----|
##               M |      1 |     15 |      16 |
##               |    0.062 |    0.938 |    0.457 |
##               |    0.111 |    0.577 |          |
##               |    0.029 |    0.429 |          |
## -----|-----|-----|-----|
##           Column Total |      9 |     26 |      35 |
##               |    0.257 |    0.743 |          |
## -----|-----|-----|-----|
##
##
```

There were 8 cases of Benign which are accurately predicted as Benign which constitutes 22.9%. But there were 27 cases of Malignant of which 16 were correctly predicted which constitutes 45.7% as Malignant but 11 cases were predicted incorrectly which constitutes to 31.4%. Hence the total accuracy of the model is 68.6% which shows that there are chances to improve the model performance.

```
#install.packages("caret")
```

```
library(caret)
```

```
## Loading required package: lattice
```

```
## Loading required package: ggplot2
```

setwd("C:/Users/Meghana Nadig/Desktop/Knn/") *#Using this command, we've imported the 'Prostate\_Cancer.csv' data file. This command is used to point to the folder containing the required file. Do keep in mind, that it's a common mistake to use "\" instead of "/" after the setwd command.*

```
p <- read.csv("C:/Users/Meghana Nadig/Desktop/Knn/prostate_cancer.csv", stringsAsFactors = FALSE)
```

```
str(p)
```

```
## 'data.frame': 100 obs. of 10 variables:
## $ id : int 1 2 3 4 5 6 7 8 9 10 ...
## $ diagnosis_result : chr "M" "B" "M" "M" ...
## $ radius : int 23 9 21 14 9 25 16 15 19 25 ...
## $ texture : int 12 13 27 16 19 25 26 18 24 11 ...
## $ perimeter : int 151 133 130 78 135 83 120 90 88 84 ...
## $ area : int 954 1326 1203 386 1297 477 1040 578 520 476 ...
## $ smoothness : num 0.143 0.143 0.125 0.07 0.141 0.128 0.095 0.119 0.127 0.119 ...
## $ compactness : num 0.278 0.079 0.16 0.284 0.133 0.17 0.109 0.165 0.193 0.24 ...
## $ symmetry : num 0.242 0.181 0.207 0.26 0.181 0.209 0.179 0.22 0.235 0.203 ...
## $ fractal_dimension: num 0.079 0.057 0.06 0.097 0.059 0.076 0.057 0.075 0.074 0.082 ...
```

```
p <- p[-1]
```

```
set.seed(1)
```

```
TrainingSet <- createDataPartition(p$diagnosis_result,p =.66, list=FALSE)
```

```
TrainingData <- p[TrainingSet,]
```

```
Test <- p[-TrainingSet,]
```

```
View(TrainingData)
```

```
typeof(TrainingSet)
```

```
## [1] "integer"
```

```
TrainingLabel <- TrainingData[1]
```

```
TestLabel <- Test[1]
```

```
typeof(TrainingLabel)
```

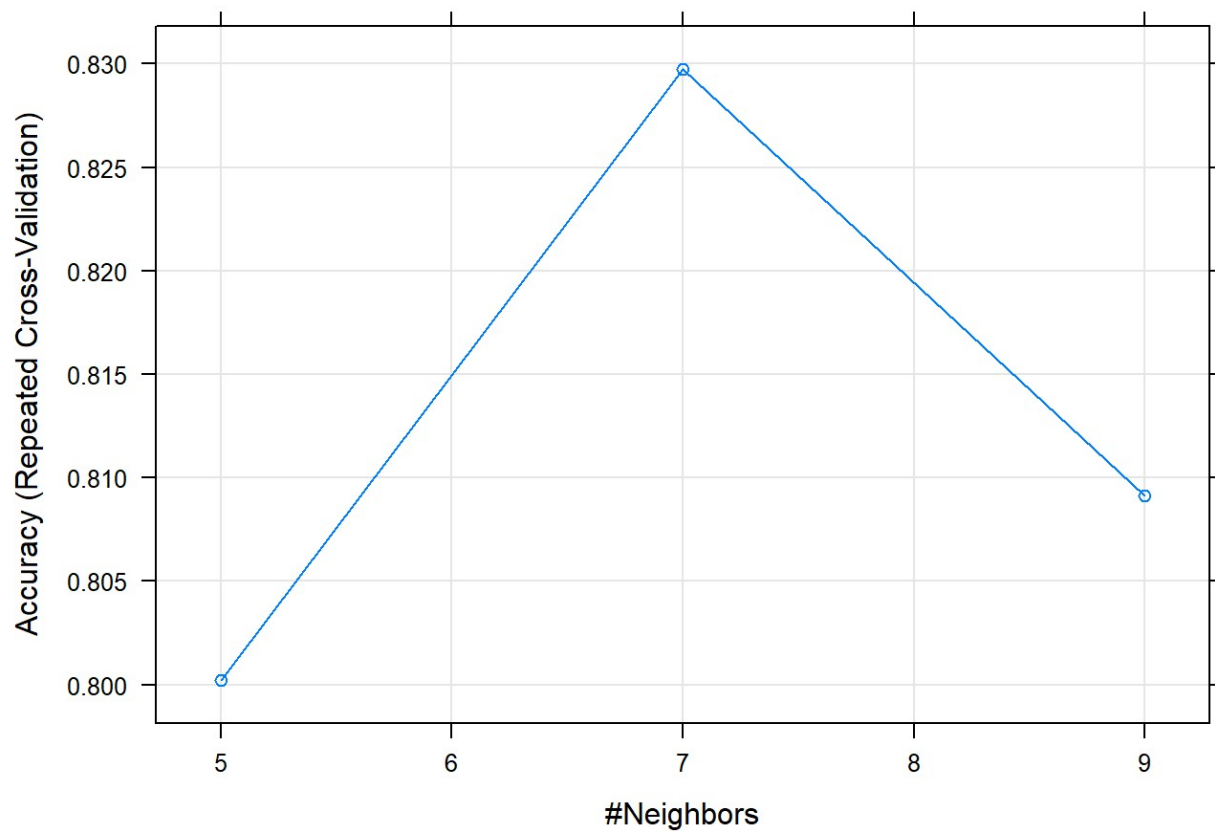
```
## [1] "list"
```

```
View(TrainingLabel)

#preprocessing
preProcValues <- preProcess(x = TrainingData,method = c("center", "scale"))

set.seed(400)
ctrl <- trainControl(method="repeatedcv",repeats = 3)
knnFit <- train(diagnosis_result ~ ., data = TrainingData, method = "knn", trControl
= ctrl, preProcess = c("center","scale"))

plot(knnFit)
```



```
knnPredict <- predict(knnFit, Test )
View(Test)
View(knnPredict)
```

```
#Get the confusion matrix to see accuracy value and other parameter values
confusionMatrix(knnPredict, Test$diagnosis_result)
```

```

## Confusion Matrix and Statistics
##
##           Reference
## Prediction  B   M
##           B   9   2
##           M   3  19
##
##           Accuracy : 0.8485
##           95% CI : (0.681, 0.9489)
##           No Information Rate : 0.6364
##           P-Value [Acc > NIR] : 0.006669
##
##           Kappa : 0.6667
##           McNemar's Test P-Value : 1.000000
##
##           Sensitivity : 0.7500
##           Specificity : 0.9048
##           Pos Pred Value : 0.8182
##           Neg Pred Value : 0.8636
##           Prevalence : 0.3636
##           Detection Rate : 0.2727
##           Detection Prevalence : 0.3333
##           Balanced Accuracy : 0.8274
##
##           'Positive' Class : B
##

```

```

#install.packages('e1071', dependencies=TRUE)

```