

Class: SY-MCA

# Progressive Education Society's

# Modern College of Engineering, Pune MCA Department

# A.Y.2022-23

Shift / Div: A

Roll Number: 52040

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*****************************				
Q1.We have four things: grape, green bean, nuts and orange with two characteristics: sweetness $(8, 3, 3, 7)$ and Crunchiness $(5, 7, 6, 3)$ . Among them two are fruits, one is protein and one is vegetable. Suppose we wanted to classify tomatoes into one of the classes. Is tomato a fruit, vegetable or protein? Tomato has the following characteristics; sweetness=6, crunchiness = 4. Let's add Carrots with characteristics sweetness = 4 and crunchiness = 9 keep k=1. Try for k=4 also.				
Program:				
things <- data.frame(ingredient = c("grape", "green bean", "nuts", "orange"),				
	sweetness = c(8,3,3	,7),		
	crunchiness = c(5,7,	6,3),		
	class = c("fruit", "ve	getable", "protein", "fruit	'))	
things				

unknown

#install.packages("dplyr")

unknown <- data.frame(ingredient = "tomato",

sweetness = 6,

crunchiness = 4,

class="unknown")

```
#install.packages("descr")
#install.packages("ggplot2")
library(dplyr)
library(descr)
library(ggplot2)
ggplot(bind_rows(things, unknown)) +
 geom_point(aes(x=sweetness, y=crunchiness, color=class),size=10) +
 geom_label(aes(x=sweetness, y=crunchiness, label=ingredient), hjust = 0, nudge_x = 0.25)+
 xlim(2,9) + ylim(3,8)
library(class) #contains knn function
pred <- knn(select(things, sweetness, crunchiness),</pre>
      select(unknown,sweetness, crunchiness), things$class, k=1)
pred
unknown <- data.frame(ingredient = c("tomato", "carrot"),
            sweetness = c(6,4),
            crunchiness = c(4,9),
            class=c("unknown", "unknown"))
unknown
pred <- knn(select(things, sweetness, crunchiness),</pre>
      select(unknown,sweetness, crunchiness), things$class, k=1)
pred
```

pred <- knn(select(things, sweetness, crunchiness),</pre>

select(unknown,sweetness, crunchiness), things\$class, k=4)

pred

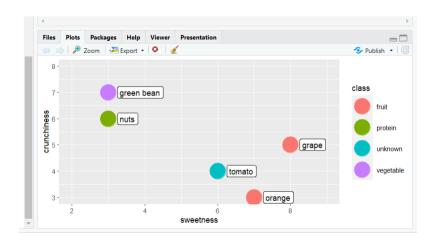
#### Output:

```
> yyyror(urmu_roms(urmys, unknown)) +
+ geom_point(aes(x=sweetness, y=crunchiness, color=class),size=10) +
+ geom_label(aes(x=sweetness, y=crunchiness, label=ingredient), hjust = 0, nudge_x = 0.25)+
+ xlim(2,9) + ylim(3,8)
> library(class) #contains knn function
> pred <- knn(select(things, sweetness, crunchiness),
+ select(unknown,sweetness, crunchiness), things$class, k=1)
> pred
[1] fruit
Levels: fruit protein vegetable
> unknown <- data.frame(ingredient = c("tomato", "carrot"),
+ sweetness = c(6,4),
+ crunchiness = c(4,9),
+ crunchiness = c(4,9),
+ unknown
ingredient sweetness crunchiness class
1 tomato 6 4 unknown
2 carrot 4 9 unknown
> pred <- knn(select(things, sweetness, crunchiness),
+ select(unknown,sweetness, crunchiness),
+ pred
[1] fruit vegetable
Levels: fruit protein vegetable
> pred <- knn(select(things, sweetness, crunchiness),
+ select(unknown,sweetness, crunchiness),
+ select
```

#### For k=1

```
> pred <- knn(select(things, sweetness, crunchiness),
+ select(unknown,sweetness, crunchiness), things$class, k=1)
> pred
[1] fruit vegetable
Levels: fruit protein vegetable
```

#### For k=4



Q2.Using Titanic.CSV file predict which people are more likely to survive after the collision with the iceberg using Decision Trees.

```
Program:
library(caret)
library(FSelector)
library(rpart)
library(rpart.plot)
library(dplyr)
library(xlsx)
library(data.tree)
library(caTools)
df
#Sys.setenv(JAVA_HOME='C:\\Program Files\\Java\\jdk1.8.0_202')
summary(df)
#Titanic <- Titanic[,c('Class','Age','Sex','Survived')]</pre>
df <- select(df, Survived, Class, Sex, Age)</pre>
df <- na.omit(df)
df
df <- mutate(df, Survived = factor(Survived), Class = as.numeric(Class), Age =
       as.numeric(Age))
set.seed(123)
```

```
sample = sample.split(df$Survived, SplitRatio = .70)

train = subset(df, sample==TRUE)

test = subset(df, sample == FALSE)

#Training the decision tree classifier

tree <- rpart(Survived ~., data = train)

#Predictions

tree.survived.predicted <- predict(tree, test, type = 'class')

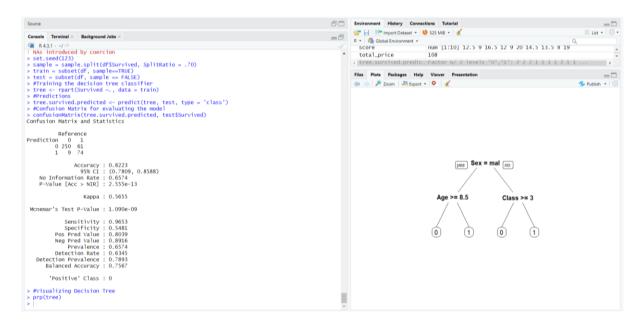
#Confusion Matrix for evaluating the model

confusionMatrix(tree.survived.predicted, test$Survived)

#Visualizing Decision Tree

prp(tree)</pre>
```

## Output:



Q3.Load the tissue gene expression dataset. Run a k-means clustering on the data with K=7. Make a table comparing the identified clusters to the actual tissue types. Run the algorithm several times to see how the answer changes.

### Program:

```
#install.packages("dslabs")
library(dslabs)
data("tissue_gene_expression")
df <- data.frame(tissue_gene_expression)
df
cl <- kmeans(tissue_gene_expression$x, centers = 7)
table(cl$cluster, tissue_gene_expression$y)</pre>
```

## Output:

#### 1st run

# 2nd run

Q4.Plot the distribution of distances between data points and their fifth nearest neighbors using the KNNdistplot function from the dbscan package. Examine the plot and find a tentative threshold at which distances start increasing quickly. On the same plot, draw a horizontal line at the level of the threshold (use Iris dataset)

```
Program:
#install.packages("dbscan")

df <- iris[, -ncol(iris)]

df <- scale(df)

df <- as.data.frame(df)

library(dbscan)

kNNdistplot(df, k = 5)

abline(h = 0.8, col = "red")
```

# Output:

