#Admission Dataset

#load packages

library("MLmetrics") #metric for model

library(readxl)#to read excel

library(plyr)#split and put data

library(DescTools) #For Mode

library(dplyr) #For Pipe Operator

library(ggplot2) #For QQ Plot

library(moments) #For Skewness & Kurtosis

library(rstatix) #For Welch ANOVA Test

library(Hmisc) #For rcorr Function & Missing Values Treatment

library(QuantPsyc) #For Im.beta function

library(ggpubr) #For advanced QQ Plots

library(aod) #For wald.test function in logistic regression

library(fmsb)#For NagelkerkeR2 function in logistic regression

library(caret) #For Data manupilation

library(imputeMissings) #For imputing missing Values

library(purrr) #For Missing Values

library(naivebayes) #For naive bayes

library(rpart.plot) #For DT Graph

library(psych) #descriptive stats

library(rpart) #for DT

library(e1071) #for SVM

library(knitr) #to save as pdf

library(gridExtra)#work with grid graphics

library(datasets)#dataset related operations

library(tinytex)# to convert to latex

```
8_Classification\\Admission data.xlsx")
str(AdmissionData)
#converting to dataframe
Add<-data.frame(AdmissionData)
str(Add)
#converting variables to appropriate types
Add$admit<-as.factor(AdmissionData$admit)
Add$rank<-as.factor(AdmissionData$rank)
Add$gre<-as.numeric(AdmissionData$gre)
Add$gpa<-as.numeric(AdmissionData$gpa)
str(Add)
# check for missing values
map(Add, ~sum(is.na(.)))
#Checking Skewness
sapply(Add[(2:3)], "skewness")
#boxplot to see outliers
boxplot(Add$gre)
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```
boxplot(Add$gpa)
#getting the outliers
outA<-boxplot(Add$gre)$out
outB<-boxplot(Add$gpa)$out
#removing the outliers
Addnew<- Add
Addnew<-Addnew[-which(Add$gre %in% outA),]
Addnew<-Addnew[-which(Add$gpa %in% outB),]
str(Addnew)
#categorizing gre
Addnew<- transform(Addnew,GRE_lvl = ifelse(gre<440,"low gre",ifelse(gre<580,"medium gre","high
gre")))
str(Addnew)
#summary stats
summary(Addnew)
describe(Addnew)
#normality of new dataset
skewness(Addnew$gre)
kurtosis(Addnew$gre)
skewness(Addnew$gpa)
kurtosis(Addnew$gpa)
#correlation of gpa and gre
ggplot(Addnew, aes(x =gpa, y = gre)) +
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geom_line() +
 labs(title = "GPA and GRA relation",
   x = "GPA",
   y = "GRA")
corr.test(Addnew$gpa,Addnew$gre,method = "pearson", use = "complete.obs")
# GRE_lvl distribution
ggplot(Addnew, aes(x = GRE_lvl)) +
 geom_density(fill = "indianred3") +
labs(title = "Participants by GRE_IvI")
# seeing the race distribution
plotdata <- Addnew %>%
count(Race) %>%
arrange(desc(Race)) %>%
 mutate(prop = round(n*100/sum(n), 1),
    lab.ypos = cumsum(prop) - 0.5*prop)
plotdata$label <- paste0(plotdata$Race, "\n",
             round(plotdata$prop), "%")
ggplot(plotdata,
   aes(x = "",
     y = prop,
      fill =Race)) +
 geom_bar(width = 1,
      stat = "identity",
      color = "red") +
 geom_text(aes(y = lab.ypos, label = label),
      color = "yellow") +
```

```
start = 0,
       direction = -1) +
 theme_void() +
 theme(legend.position = "FALSE") +
 labs(title = "Race Distribution")
# seeing GRE levels distribution by Rank
ggplot(Addnew,
   aes(x = rank,
      fill = GRE_lvl)) +
 geom_bar(position = "dodge")
#To plot GRE per Social economic level
plot(Addnew$ses,Addnew$gre, xlab = "Number of prospects", ylab = "duration per prospect", col =
"blue")
# Calculate test statistics
add_aov <- aov(Addnew$gre~factor(Addnew$ses))
summary(add aov)
#seeing which gender gets more admission
ggplot(Addnew, aes(x = Gender_Male, fill = admit)) +
 geom_bar(position = "fill") +
 theme_classic()+
 labs(x = "Gender",y = "Count",fill = "Admission")
# Seeing if high gpa leads to person taking the admission?
boxplot(Addnew$gpa~factor(Addnew$admit),
    xlab = "Did they take the admission?", ylab = "GPA")
```

coord_polar("y",

```
# Calculate test statistics
add_aov <- aov(Addnew$gpa~factor(Addnew$admit))</pre>
summary(add_aov)
str(Addnew)
#Data partition
set.seed(123)
Train <- createDataPartition(Addnew$admit, p=0.7, list=FALSE)
training <- Addnew[ Train, ]</pre>
testing <- Addnew[ -Train, ]
#Model1- Logistic Regression
#create the model using train()
Model1 <- train(admit~., data=training, method="glm", family="binomial")
summary(Model1)
#exponentiated coefficients
exp(coef(Model1$finalModel))
# prediction for test data
predadmit<-predict(Model1, newdata=testing)</pre>
#fitness metrics for validation
confusionMatrix(predadmit, testing$admit, mode = "everything", positive="1")
#predicting probability for new values
newvalues<-data.frame(gre=800,gpa=4,ses=2,GRE_lvl='medium
gre',ses=as.factor(2),Gender_Male=1,Race=1,rank=as.factor(1))
```

```
pred<-predict(Model1, newdata=newvalues,type="prob")</pre>
pred
#Model2-SVM
#create the model using train()
Model2<-svm(admit~.,data=training)
summary(Model2)
# prediction for test data
predadmit<-predict(Model2, newdata=testing)</pre>
#fitness metrics for validation
confusionMatrix(predadmit, testing$admit, mode = "everything", positive="1")
#predicting probability for new values
newvalues<-data.frame(gre=900,gpa=3.8,ses=2,GRE_lvl='high
gre',Gender_Male=1,Race=2,rank=as.factor(3))
pred<-predict(Model2, newdata=newvalues,type="prob")</pre>
pred
#Model3- Decision Tree
#create the model using train()
Model3 <- train(admit~., data=training, method="rpart",parms= list(split="information"))
summary(Model3)
```

```
# prediction for test data
predadmit<-predict(Model3, newdata=testing)</pre>
#fitness metrics for validation
confusionMatrix(predadmit, testing$admit, mode = "everything", positive="1")
#predicting probability for new values
newvalues<-data.frame(gre=650,gpa=4,ses=1,GRE_lvl='medium
gre',Gender_Male=1,Race=3,rank=as.factor(4))
pred<-predict(Model2, newdata=newvalues,type="prob")</pre>
pred
#converting to pdf
knitr::stitch('Admissiondata.r')
setwd("C:\\Users\\sujoydutta\\Desktop\\Data analysis\\R\\R scripts")
dir()
"Admissiondata.R"
knitr::stitch('Admissiondata.R')
```