```
### Installing packages and loading library
library(readxl)
### Setting up the working directory
setwd("D:/great learning/4. Predictive Modeling/Project-4")
mydata = read_xlsx("Cellphone.xlsx", sheet = "Data")
View(mydata)
str(mydata)
summary(mydata)
### Check for multi co-linearity
library(corrplot)
mydata_1 = mydata[-c(1)]
mydata_1
mydata\_corr = cor(mydata)
corrplot(mydata_corr, method = "number")
plot(mydata$DataPlan~mydata$DataUsage, col = "blue")
plot(mydata$DataPlan~mydata$MonthlyCharge, col = "green")
plot(mydata$MonthlyCharge~mydata$DataUsage, col = "red")
plot(mydata$Churn~mydata$DataUsage, col = "orange")
plot(mydata$Churn~mydata$MonthlyCharge)
### Treatment of multi co-linearity (Factor analysis)
```

```
#install.packages("nFactors")
library(nFactors)
mydata1 = mydata[c(4,5,9)]
                                # Considering only correlated variables DataPlan, DataUsage,
MonthlyCharge
names(mydata1)
mydata1corr = cor(mydata1)
mydata1corr
ev = eigen(cor(mydata1))
ev
EigenValue = ev$values
EigenValue
solution = fa(r = mydata1, nfactors = 1, rotate = "none", fm = "pa")
solution
solution1 = fa(r = mydata1, nfactors = 1, rotate = "varimax", fm = "pa")
solution1
solution1$scores
### Binding factor with other attributes
mydata\_bind = cbind(mydata[c(1,2,3,6,7,8,10,11)],solution1$scores)
names(mydata_bind) = c("Churn", "AccountWeeks", "ContractRenewal", "CustServCalls",
"DayMins", "DayCalls", "OverageFee", "RoamMins", "ServiceUsage")
names(mydata_bind)
attach(mydata_bind)
### Building Logistic Regression Model
```

Splitting the PCA/FA treatement data into train data set and test data set

```
library(caTools)
Sample = sample.split(mydata_bind,SplitRatio = 0.7)
train_mydata_bind = subset(mydata_bind, Sample ==T)
test_mydata_bind = subset(mydata_bind, Sample ==F)
dim(train_mydata_bind)
dim(test_mydata_bind)
# Step-1 Log likelyhood ratio test
install.packages("lmtest")
library(lmtest)
logit =
glm(Churn~AccountWeeks+ContractRenewal+CustServCalls+DayMins+DayCalls+OverageFee+Ro
amMins+ServiceUsage, data = train_mydata_bind, family = binomial)
lrtest(logit)
# Step-2 Mcfaden R square computation
install.packages("pscl")
library(pscl)
pR2(logit)
# Step-3 Test of individual coefficient
summary(logit)
# Step-4 Odds ration
```

```
odds = exp(coef(logit))
odds
Probability = odds/(1+odds)
Probability
# Step-5 Confusion matrix for measurement of predictive accuracy
predict(logit, type = "response")
pred = fitted(logit)
data.frame(train_mydata_bind$Churn, pred)
gg1 = floor(pred + 0.5)
gg1
table(Actual = train_mydata_bind$Churn, prediction = gg1)
# Step-6 ROC plot
#install.packages("Deducer")
#library(Deducer)
#rocplot(logit)
library(ROCR)
pred1 = prediction(gg1, train_mydata_bind$Churn)
pred1
perf1 = performance(pred1, "tpr", "fpr")
#plot(perf1)
plot(perf1, col = "Blue", main = "ROC Plot")
abline(0, 1, lty =8, col = "red")
```

```
auc = performance(pred1, "auc")
auc = auc@y.values
auc
### Building KNN model
# Exploratory data analysis
norm = function(x)\{(x - min(x))/(max(x) - min(x))\}\
norm.data = as.data.frame(lapply(mydata_bind[,-1], norm))
View(data)
View(norm.data)
# Data partitioning
usable.data = cbind(Churn,norm.data)
names(usable.data)
View(usable.data)
Sample1 = sample.split(usable.data,SplitRatio = 0.7)
train = subset(usable.data, Sample1 == T)
test = subset(usable.data, Sample1 == F)
dim(train)
dim(test)
### Use KNN classifier
library(class)
```

```
pred1 = knn(train[,-1],test[,-1],train[,1], k=19)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=17)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=15)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=13)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=11)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=9)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=7)
table.knn = table(test[,1],pred1)
table.knn
```

```
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=5)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=3)
table.knn = table(test[,1],pred1)
table.knn
sum(diag(table.knn))/sum(table.knn)
pred1 = knn(train[,-1],test[,-1],train[,1], k=1)
table.knn = table(test[,1],pred1)
table.knn
### Building Naive Bayes Model
# Naive Bayes Classifier
library(e1071)
View(train)
NB = naiveBayes(Churn~., data = train)
NB
predNB = predict(NB,test,type = "class")
predNB
table.NB = table(test[,1],predNB)
table.NB
sum(diag(table.NB))/sum(table.NB)
```

Fold validation and model comparison

```
install.packages("caret")
install.packages("klaR")
library(caret)
library(klaR)
### Folds are created on the basis of target variable
folds = createFolds(factor(train$Churn, K=10, list = FALSE))
folds
### Define train control for k fold cross validation
train_control = trainControl(method = "cv", number = 10)
### Fit KNN
modelKNN = train('mydata_bind[,1]'~., data = train, trcontrol = train_control, method = "knn")
summary(modelKNN)
### Fit Naive Bayes Model
modelNB = train('mydata_bind[,1]'~., data = train, trcontrol = train_control, method = "nb")
summary(modelNB)
### Collect resamples
```

results = resamples(list(KNN = modelKNN, NB = modelNB)) summary(results) bwplot(results)