# Exerise14\_AyachitMadhukar

## R Exercise 14

```
setwd("~/MadR/Workspaces/dsc520")
```

# Loading data from file

```
data=read.csv("data/binary-classifier-data.csv")
```

## Observations at first glance

```
#Normalization
summary(data)
```

```
##
      label
                                     у
## Min.
        :0.000 Min. : -5.20
                               Min.
                                     : -4.019
## 1st Qu.:0.000 1st Qu.: 19.77
                                1st Qu.: 21.207
## Median :0.000 Median : 41.76
                               Median: 44.632
## Mean :0.488 Mean : 45.07
                                Mean : 45.011
## 3rd Qu.:1.000 3rd Qu.: 66.39
                                3rd Qu.: 68.698
## Max. :1.000 Max. :104.58
                                Max. :106.896
```

# str(data)

```
## 'data.frame': 1498 obs. of 3 variables:
## $ label: int 0 0 0 0 0 0 0 0 0 0 ...
## $ x : num 70.9 75 73.8 66.4 69.1 ...
## $ y : num 83.2 87.9 92.2 81.1 84.5 ...
```

# head(data)

```
## 1 label x y
## 1 0 70.88469 83.17702
## 2 0 74.97176 87.92922
## 3 0 73.78333 92.20325
## 4 0 66.40747 81.10617
## 5 0 69.07399 84.53739
## 6 0 72.23616 86.38403
```

#### Randomization

```
set.seed(9850)
gp<-runif(nrow(data))</pre>
data<-data[order(gp),]</pre>
head(data)
##
       label
                     X
## 216
        0 12.006713 58.20435
## 405
           0 88.024357 13.26384
## 316
           0 7.993121 54.15258
## 804
          1 16.669075 73.98231
           0 40.565872 74.84798
## 103
## 20
           0 69.521713 89.94501
```

### Normalization

```
## Min. :5.200 Min. :4.019
## 1st Qu.:5.439 1st Qu.:4.255
## Median :5.650 Median :4.475
## Mean :5.681 Mean :4.478
## 3rd Qu.:5.885 3rd Qu.:4.700
## Max. :6.250 Max. :5.057
```

## Preparing training/testing models

```
r<-round(0.8*nrow(data.n))
l<-nrow(data.n)

data.train<-data.n[1:r,]
data.test<-data.n[r:1,]</pre>
```

```
data.train.target<-data[1:r,1]
data.test.target<-data[r:1,1]</pre>
```

# Logistic regression

```
library(caTools)
split<- sample.split(data,SplitRatio=.8)</pre>
train<-subset(data,split=="TRUE")</pre>
test<-subset(data,split=="FALSE")</pre>
GLM.1 <- glm(train$label ~ ., family=binomial(), data=train)</pre>
summary(GLM.1)
##
## glm(formula = train$label ~ ., family = binomial(), data = train)
## Deviance Residuals:
               1Q
                     Median
                                   3Q
                                           Max
       Min
## -1.3655 -1.1433 -0.9494 1.1749
                                        1.4434
##
## Coefficients:
##
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) 0.407395 0.142478 2.859 0.00424 **
              -0.001240 0.002250 -0.551 0.58164
## x
              -0.009435
                         0.002319 -4.069 4.72e-05 ***
## y
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
       Null deviance: 1382.4 on 997 degrees of freedom
## Residual deviance: 1363.2 on 995 degrees of freedom
## AIC: 1369.2
##
## Number of Fisher Scoring iterations: 4
res<-predict(GLM.1,test,type="response")</pre>
res<-predict(GLM.1,train,type="response")</pre>
conffmatrix<- table(actual_value=train$label,Predicted_value=res>0.5)
conffmatrix
##
               Predicted_value
## actual_value FALSE TRUE
                  296 220
              0
                  242 240
##
              1
```

```
acc<-(conffmatrix[1,1] + conffmatrix[2,2])/(conffmatrix[1,1] + conffmatrix[2,2] + conffmatrix[2,1]+ con</pre>
```

## **KNN** Classifire

Determining k value

```
k_value<-round(sqrt(nrow(data)))
```

Predicting knn and accuracy

```
library("class")
knn.39<- knn(train = data.train, test=data.test,cl=data.train.target, k=k_value)
ACC.39<-100 * sum( data.test.target == knn.39) / NROW(data.test.target)
table(data.test.target,knn.39)
##
                   knn.39
## data.test.target
                      0
                          1
##
                  0 156
                     1 142
##
ACC.39
## [1] 99.00332
knn.39<- knn(train = data.train, test=data.test,cl=data.train.target, k=k_value)
```

# Conclusion

a. What is the accuracy of the logistic regression classifier?

logistic regression shows close to 99.0033223 accuracy

b. How does the accuracy of the logistic regression classifier compare to the nearest neighbors algorithm?

 $logistic\ regression\ is\ way\ off\ from\ KNN\ whch\ is\ nearly\ 53.7074148\%\ vs\ KNN\ being\ 99.0033223\%$  accurate

c. Why is the accuracy of the logistic regression classifier different from that of the nearest neighbors?