Cordero week9.2

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```
library(foreign)
library(caTools)
```

Thoraric Surgery

```
ts_data <- read.arff("ThoraricSurgery.arff")</pre>
ts_split <- sample.split(ts_data, SplitRatio = 0.8)</pre>
ts_train <- subset(ts_data, ts_split == "TRUE")</pre>
ts_test <- subset(ts_data, ts_split == "FALSE")</pre>
```

1.b.i

```
ts_model1 <- glm(Risk1Yr ~ AGE + PRE17 + PRE19 + PRE30, data = ts_train, family = 'binomial')
summary(ts_model1)
##
## Call:
## glm(formula = Risk1Yr ~ AGE + PRE17 + PRE19 + PRE30, family = "binomial",
##
       data = ts_train)
##
## Coefficients:
                Estimate Std. Error z value Pr(>|z|)
## (Intercept) -2.47033 1.28297 -1.925 0.05417 .
## AGE
                -0.01067
                            0.01842 -0.579 0.56243
                            0.47241
## PRE17T
                                     3.087 0.00202 **
                 1.45824
## PRE19T
               -13.95361 1029.09687 -0.014 0.98918
## PRE30T
                 1.48188
                            0.61810 2.397 0.01651 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 311.20 on 359 degrees of freedom
## Residual deviance: 293.89 on 355 degrees of freedom
```

```
## AIC: 303.89
##
## Number of Fisher Scoring iterations: 14
```

1.b.ii

According to the summary, variables PRE17 and PRE30 had the greatest effect on survival rate based on their p-values

1.b.iii

```
ts_preds <- predict(ts_model1, ts_test, type = "response")

ts_preds <- predict(ts_model1, ts_train, type = "response")

ts_confmatrix <- table(Actual_Value=ts_train$Risk1Yr, Predicted_Value = ts_preds > 0.5)

ts_confmatrix

## Predicted_Value
## Actual_Value FALSE
## F 304
## T 56

ts_accuracy <- ts_confmatrix[1,1] / sum(ts_confmatrix)
ts_accuracy</pre>
```

Binary Classifier

[1] 0.8444444

2.b

```
bc_data <- read.csv("binary-classifier-data.csv")
bc_split <- sample.split(bc_data, SplitRatio = 0.8)
bc_train <- subset(bc_data, bc_split == "TRUE")
bc_test <- subset(bc_data, bc_split == "FALSE")
bc_model1 <- glm(label ~ x + y, data = bc_train, family = 'binomial')
summary(bc_model1)

##
## Call:
## glm(formula = label ~ x + y, family = "binomial", data = bc_train)
##
## Coefficients:</pre>
```

```
Estimate Std. Error z value Pr(>|z|)
##
## (Intercept) 0.415200 0.143553 2.892 0.003824 **
             ## y
             ## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
      Null deviance: 1382.9 on 997 degrees of freedom
## Residual deviance: 1367.6 on 995 degrees of freedom
## AIC: 1373.6
## Number of Fisher Scoring iterations: 4
bc_preds <- predict(bc_model1, bc_test, type = "response")</pre>
bc_preds <- predict(bc_model1, bc_train, type = "response")</pre>
bc_confmatrix <- table(Actual_Value= bc_train$label, Predicted_Value = bc_preds > 0.5)
bc_confmatrix
##
             Predicted_Value
## Actual_Value FALSE TRUE
##
            0
                289 222
##
            1
                202 285
bc_accuracy <- (bc_confmatrix[1,1] + bc_confmatrix[2,2]) / sum(bc_confmatrix)</pre>
2.b.i
bc_accuracy
```

[1] 0.5751503