## Exercise 10.2

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### Question 1

```
library(foreign)
surgery_df <- read.arff("data/ThoraricSurgery.arff")</pre>
surgery_glm <- glm(Risk1Yr ~ DGN + PRE4 + PRE5 + PRE6 + PRE7 + PRE8 + PRE9 + PRE10 + PRE11 + PRE14 + PR
summary(surgery_glm)
##
## Call:
  glm(formula = Risk1Yr ~ DGN + PRE4 + PRE5 + PRE6 + PRE7 + PRE8 +
##
       PRE9 + PRE10 + PRE11 + PRE14 + PRE17 + PRE19 + PRE25 + PRE30 +
##
       PRE32 + AGE, family = binomial, data = surgery_df)
##
## Deviance Residuals:
                 1Q
                     Median
                                   3Q
                                          Max
## -1.6084 -0.5439 -0.4199 -0.2762
                                        2.4929
##
## Coefficients:
                Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.655e+01 2.400e+03 -0.007 0.99450
## DGNDGN2
               1.474e+01 2.400e+03
                                      0.006 0.99510
## DGNDGN3
               1.418e+01 2.400e+03
                                      0.006 0.99528
## DGNDGN4
               1.461e+01 2.400e+03
                                      0.006 0.99514
## DGNDGN5
               1.638e+01 2.400e+03
                                      0.007
                                             0.99455
## DGNDGN6
               4.089e-01 2.673e+03
                                      0.000 0.99988
## DGNDGN8
               1.803e+01 2.400e+03
                                      0.008 0.99400
## PRE4
              -2.272e-01 1.849e-01 -1.229
                                             0.21909
## PRE5
              -3.030e-02 1.786e-02 -1.697
                                             0.08971 .
## PRE6PRZ1
              -4.427e-01 5.199e-01
                                     -0.852
                                             0.39448
## PRE6PRZ2
               -2.937e-01
                          7.907e-01
                                     -0.371
                                             0.71030
## PRE7T
               7.153e-01 5.556e-01
                                      1.288 0.19788
## PREST
               1.743e-01 3.892e-01
                                      0.448 0.65419
## PRE9T
               1.368e+00 4.868e-01
                                      2.811 0.00494 **
## PRE10T
               5.770e-01 4.826e-01
                                      1.196 0.23185
## PRE11T
               5.162e-01 3.965e-01
                                      1.302 0.19295
## PRE140C12
               4.394e-01 3.301e-01
                                      1.331 0.18318
## PRE140C13
               1.179e+00 6.165e-01
                                      1.913 0.05580 .
```

```
## PRE140C14
               1.653e+00 6.094e-01
                                     2.713 0.00668 **
## PRF17T
               9.266e-01 4.445e-01
                                     2.085 0.03709 *
              -1.466e+01 1.654e+03 -0.009 0.99293
## PRE19T
## PRE25T
              -9.789e-02 1.003e+00
                                    -0.098
                                            0.92227
## PRE30T
               1.084e+00 4.990e-01
                                     2.172
                                            0.02984 *
              -1.398e+01 1.645e+03 -0.008
## PRE32T
                                           0.99322
              -9.506e-03 1.810e-02 -0.525
## AGE
                                            0.59944
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##
      Null deviance: 395.61 on 469 degrees of freedom
## Residual deviance: 341.19 on 445 degrees of freedom
## AIC: 391.19
##
## Number of Fisher Scoring iterations: 15
```

According to the above summary, it appears that the variables "Pre9",those with Dyspnoea before the surgery, and the "OC14" variant of "Pre14" variable that list the size of the original tumor, have the greatest effect on survival rate. "Pre30" and "Pre17" also having a large impact as well, showing that these variables respectively tracking smoking and type 2 diabetes also have a large impact. The pvalues of "Pre9" at .005 and "Pre14OC14" at .007 indicate a high statistical significance. The pvalues of "Pre30" at .03 and "Pre17" at .04 also indicate statistical significance, but not nearly as much as "Pre9" and "Pre14OC14".

```
surgery_res <- predict(surgery_glm, type = "response")
surgeryPrediction <- table(Actual = surgery_df$Risk1Yr, Predicted = surgery_res > .5)
(surgeryPrediction[[1,1]] + surgeryPrediction[[2,2]]) / sum(surgeryPrediction)
```

```
## [1] 0.8361702
```

The accuracy of the model is approximately 84%.

# Question 2

```
binary_df <- read.csv("data/binary-classifier-data.csv")
head(binary_df)</pre>
```

```
##
     label
                   X
## 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
binary_glm <- glm(label ~ x+y, data = binary_df, family = binomial)</pre>
summary(binary_glm)
```

```
##
## Call:
## glm(formula = label ~ x + y, family = binomial, data = binary_df)
## Deviance Residuals:
       Min
                1Q
                     Median
                                   3Q
##
                                           Max
                              1.1646
## -1.3728 -1.1697 -0.9575
                                       1.3989
##
## Coefficients:
##
                Estimate Std. Error z value Pr(>|z|)
## (Intercept) 0.424809
                           0.117224
                                    3.624 0.00029 ***
                           0.001823 -1.411 0.15836
              -0.002571
## x
              -0.007956
                           0.001869 -4.257 2.07e-05 ***
## y
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 2075.8 on 1497 degrees of freedom
## Residual deviance: 2052.1 on 1495 degrees of freedom
## AIC: 2058.1
## Number of Fisher Scoring iterations: 4
binary_res <- predict(binary_glm, type = "response")</pre>
binaryPrediction <- table(Actual = binary_df$label, Predicted = binary_res > .5)
(binaryPrediction[[1,1]] + binaryPrediction[[2,2]]) / sum(binaryPrediction)
```

#### ## [1] 0.5834446

The accurracy of the logistic regression classifier is approximately 58%.