ASSIGNMENT 10.3

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```
## Load the `data/binary-classifier-data` to
class_df <- read.csv("/Users/siddharthabhaumik/Documents/GitHub/dsc520/data/binary-classifier-data.csv"</pre>
## Viewing Sample data
head(class_df)
##
     label
## 1
        0 70.88469 83.17702
## 2
        0 74.97176 87.92922
        0 73.78333 92.20325
## 4
        0 66.40747 81.10617
## 5
        0 69.07399 84.53739
         0 72.23616 86.38403
# Fit a linear model
class.glm <- glm(label ~ x + y, data = class_df, family = "binomial")</pre>
# Print the model
class.glm
##
## Call: glm(formula = label ~ x + y, family = "binomial", data = class_df)
## Coefficients:
## (Intercept)
      0.424809
                  -0.002571
                               -0.007956
##
## Degrees of Freedom: 1497 Total (i.e. Null); 1495 Residual
## Null Deviance:
                        2076
## Residual Deviance: 2052 AIC: 2058
# View the summary of your model
summary(class.glm)
##
## Call:
## glm(formula = label ~ x + y, family = "binomial", data = class_df)
## Deviance Residuals:
##
       Min
                 1Q
                     Median
                                    3Q
                                            Max
```

```
## -1.3728 -1.1697 -0.9575 1.1646
                                       1.3989
##
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
##
## (Intercept) 0.424809 0.117224
                                    3.624 0.00029 ***
              -0.002571
                          0.001823 -1.411 0.15836
## x
              -0.007956  0.001869  -4.257  2.07e-05 ***
## v
## ---
## Signif. codes: 0 '***' 0.001 '**' 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
## What is the accuracy of your model?
set.seed(1234)
#load necessary packages
library(caTools)
library(caret)
## Loading required package: ggplot2
## Loading required package: lattice
library(InformationValue)
##
## Attaching package: 'InformationValue'
## The following objects are masked from 'package:caret':
##
##
       confusionMatrix, precision, sensitivity, specificity
library(ISLR)
# 70:30 data split into training and validation
data_split = sample.split(class_df, SplitRatio = 0.7)
training_data = subset(class_df, data_split==TRUE)
test_data = subset(class_df, data_split==FALSE)
# Print data frames
print(dim(training_data))
## [1] 999
print(dim(test_data))
## [1] 499
# Fit a linear model
new_class_glm = glm( label ~ . , family="binomial", data = training_data)
```

```
#Summary
summary(new_class_glm)
##
## Call:
## glm(formula = label ~ ., family = "binomial", data = training_data)
## Deviance Residuals:
      Min
           1Q
                    Median
                                          Max
## -1.3733 -1.1714 -0.9584
                                       1.3962
                             1.1637
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) 0.426096
                                    2.973 0.002946 **
                          0.143309
## x
                          0.002229 -1.172 0.241028
              -0.002613
              -0.007897
                          0.002279 -3.465 0.000531 ***
## y
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
      Null deviance: 1384.4 on 998 degrees of freedom
## Residual deviance: 1368.6 on 996 degrees of freedom
## AIC: 1374.6
##
## Number of Fisher Scoring iterations: 4
#Prediction based on test data
predict_data = predict(new_class_glm, newdata = test_data, type = "response")
# Confusion matrix on test set
confusionMatrix(test_data$label,predict_data)
      0
##
          1
## 0 143 96
## 1 113 147
#calculate sensitivity: The "true positive rate"
sensitivity(test_data$label,predict_data)
## [1] 0.6049383
#calculate specificity: The "true negative rate"
specificity(test_data$label,predict_data)
## [1] 0.5585938
#calculate total mis-classification error rate
misClassError(test_data$label,predict_data)
## [1] 0.4188
## The total misclassification error rate is 41.88% for this model which indicates its not a accurate m
```