Week9_Assignment9.2_KoppulaVeera

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##Fit a Logistic Regression Model to Thoracic Surgery Binary Dataset. #For this problem, you will be working with the thoracic surgery data set from the University of California Irvine machine learning repository. This dataset contains information on life expectancy in lung cancer patients after surgery. The underlying thoracic surgery data is in ARFF format. This is a text-based format with information on each of the attributes. You can load this data using a package such as foreign or by cutting and pasting the data section into a CSV file.

##Assignment Instructions: #Fit a binary logistic regression model to the data set that predicts whether or not the patient survived for one year (the Risk1Y variable) after the surgery. Use the glm() function to perform the logistic regression. See Generalized Linear Models for an example. Include a summary using the summary() function in your results.

```
##
## Call:
  glm(formula = Risk1Yr ~ ., family = binomial, data = thoracic)
##
## Deviance Residuals:
       Min
##
                  1Q
                       Median
                                     3Q
                                             Max
   -2.4929
             0.2762
                       0.4199
                                0.5439
                                          1.6084
##
##
  Coefficients:
##
                 Estimate Std. Error z value Pr(>|z|)
                2.604e+01
                            2.333e+03
                                         0.011 0.991093
## (Intercept)
## DGNDGN2
                -5.557e-01
                            4.128e-01
                                        -1.346 0.178199
## DGNDGN4
               -4.278e-01
                            4.733e-01
                                        -0.904 0.366122
## DGNDGN6
                 1.377e+01
                            1.178e+03
                                         0.012 0.990671
## DGNDGN5
               -2.201e+00
                            6.113e-01
                                        -3.600 0.000318 ***
## DGNDGN8
                -3.852e+00
                            1.550e+00
                                        -2.485 0.012959
## DGNDGN1
                 1.418e+01
                            2.400e+03
                                         0.006 0.995285
## PRE4
                 2.272e-01
                            1.849e-01
                                         1.229 0.219094
## PRE5
                 3.030e-02
                            1.786e-02
                                         1.697 0.089715
## PRE6PRZ1
                 1.490e-01
                            5.783e-01
                                         0.258 0.796647
## PRE6PRZ0
                -2.937e-01
                            7.907e-01
                                        -0.371 0.710303
## PRE7F
                 7.153e-01
                            5.556e-01
                                         1.288 0.197884
## PRE8F
                            3.892e-01
                 1.743e-01
                                         0.448 0.654188
## PRE9F
                 1.368e+00
                            4.868e-01
                                         2.811 0.004942 **
## PRE10F
                 5.770e-01
                            4.826e-01
                                         1.196 0.231855
## PRE11F
                 5.162e-01
                            3.965e-01
                                         1.302 0.192948
## PRE140C14
                -1.653e+00
                            6.094e-01
                                        -2.713 0.006675 **
## PRE140C12
                -4.394e-01
                            3.301e-01
                                        -1.331 0.183177
## PRE140C13
               -1.179e+00
                            6.165e-01
                                        -1.913 0.055799
## PRE17F
                9.266e-01
                            4.445e-01
                                         2.085 0.037092 *
```

```
## PRE19F
              -1.466e+01 1.654e+03 -0.009 0.992928
## PRE25F
              -9.789e-02 1.003e+00 -0.098 0.922273
## PRE30F
               1.084e+00 4.990e-01
                                      2.172 0.029840 *
## PRE32F
              -1.398e+01 1.645e+03 -0.008 0.993219
## AGE
               9.506e-03 1.810e-02
                                      0.525 0.599442
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 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 summary, which variables had the greatest effect on the survival rate? Variables that have greatest effect on survival rate are - AGE, Type 2 DM - diabetes mellitus(PRE17), PAD - peripheral arterial diseases(PRE25)

#To compute the accuracy of your model, use the dataset to predict the outcome variable. The percent of correct predictions is the accuracy of your model. What is the accuracy of your model?

[1] 0.2281786

Accuracy prediction for my model is about 23%

##Fit a Logistic Regression Model

ResourceSelection 0.3-5 2019-07-22

```
##
## -- Column specification -------
## cols(
##
    label = col_double(),
    x = col_double(),
##
    y = col_double()
##
## )
## spec tbl df [1,498 x 3] (S3: spec tbl df/tbl df/tbl/data.frame)
   $ label: num [1:1498] 0 0 0 0 0 0 0 0 0 0 ...
         : num [1:1498] 70.9 75 73.8 66.4 69.1 ...
          : num [1:1498] 83.2 87.9 92.2 81.1 84.5 ...
   - attr(*, "spec")=
##
    .. cols(
##
##
        label = col_double(),
    . .
         x = col_double(),
##
##
         y = col_double()
##
```

#Fit a logistic regression model to the binary-classifier-data.csv dataset

```
##
## Call:
## glm(formula = label ~ x, family = "binomial", data = binclass)
## Deviance Residuals:
##
     Min
               1Q Median
                                30
                                       Max
## -1.246 -1.159 -1.065
                            1.184
                                     1.293
##
## Coefficients:
##
                Estimate Std. Error z value Pr(>|z|)
## (Intercept) 0.137369
                           0.095119
                                       1.444
                                               0.1487
               -0.004119
                           0.001775 -2.321
                                               0.0203 *
## x
## ---
## 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: 2070.4 on 1496 degrees of freedom
## AIC: 2074.4
##
## Number of Fisher Scoring iterations: 3
#The dataset (found in binary-classifier-data.csv) contains three variables; label, x, and y. The label variable
is either 0 or 1 and is the output we want to predict using the x and y variables.
##
## Call:
## glm(formula = label ~ x + y, family = "binomial", data = binclass)
## Deviance Residuals:
##
                      Median
       Min
                 1Q
                                    30
                                            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
                                     -1.411 0.15836
                           0.001823
## x
               -0.007956
                           0.001869 -4.257 2.07e-05 ***
## y
## ---
## 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 the logistic regression classifier?
## [1] 0.5732267
```

Accuracy using ROC model is 57% - This is the accuracy of the logical regression classifier #Keep this assignment handy, as you will be comparing your results from this week to next week.