## 580\_barrett\_quiz4.R

## Nick

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```
#AMS580 Quiz 4
#By Nicholas Barrett
library(tidyverse)
## Warning: package 'tidyverse' was built under R version 3.6.3
## -- Attaching packages --------------------- tidyverse 1.3.0 --
## v ggplot2 3.3.3 v purrr 0.3.3
## v tibble 3.1.0 v dplyr 1.0.4
## v tidyr 1.1.2
                    v stringr 1.4.0
## v readr 1.4.0 v forcats 0.5.0
## Warning: package 'ggplot2' was built under R version 3.6.3
## Warning: package 'tibble' was built under R version 3.6.3
## Warning: package 'tidyr' was built under R version 3.6.3
## Warning: package 'readr' was built under R version 3.6.3
## Warning: package 'dplyr' was built under R version 3.6.3
## Warning: package 'forcats' was built under R version 3.6.3
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(caret)
## Warning: package 'caret' was built under R version 3.6.3
## Loading required package: lattice
```

```
##
## Attaching package: 'caret'

## The following object is masked from 'package:purrr':
##
## lift

library(MASS)

## Warning: package 'MASS' was built under R version 3.6.3

##
## Attaching package: 'MASS'

## The following object is masked from 'package:dplyr':
##
## select
```

```
#01
data("birthwt", package = "MASS")
#this factoring method was taken from the reference in the homework
bwt <- with(birthwt, {</pre>
  race <- factor(race, labels = c("white", "black", "other"))</pre>
  ptd <- factor(ptl > 0)
  ftv <- factor(ftv)</pre>
  levels(ftv)[-(1:2)] <- "2+"
  data.frame(low = factor(low), age, lwt, race, smoke = (smoke > 0), ptd, ht = (ht >
0), ui = (ui > 0), ftv
options(contrasts = c("contr.treatment", "contr.poly"))
#Split Data
set.seed(123)
training.samples <- bwt$low %>%
  createDataPartition(p = 0.8, list = FALSE)
train.data <- bwt[training.samples, ]</pre>
test.data <- bwt[-training.samples, ]</pre>
#Testing function
test <- function(model, test.data) {</pre>
  probabilities <- model %>% predict(test.data, type = "response")
  predicted.classes <- ifelse(probabilities > 0.5, 1, 0)
  count.acc = 0
  count.tp = 0
  count.fp = 0
  count.fn = 0
  len = length(test.data[ ,1])
  for(i in 1:len) {
    if(predicted.classes[i] == test.data$low[i]) { #true positive and true negative
      if (predicted.classes[i] == 1) {count.tp = count.tp +1} #TP
      count.acc = count.acc + 1
    if(test.data$low[i]==0 && predicted.classes[i]==1){
      count.fp = count.fp +1 #false positive
    if(test.data$low[i] == 1 && predicted.classes[i] == 0) {
      count.fn = count.fn +1 #false negative
  }
  sen = count.tp/(count.tp+count.fn)
  count.tn = (count.acc-count.tp) #TN
  spe = (count.tn)/(count.tn + count.fp)
  acc = count.acc/len
  out = list(Acc = acc, Specif = spe, Sensit = sen)
  conf.mat = matrix(data = c(count.tn,count.fp,count.fn,count.tp),byrow = TRUE,nrow=
2)
  print(conf.mat)
  return(out)
```

```
#Full Model
model <- glm(train.data$low ~., data = train.data, family = binomial)
summary(model)</pre>
```

```
##
## Call:
## glm(formula = train.data$low ~ ., family = binomial, data = train.data)
##
## Deviance Residuals:
    Min 1Q
##
                  Median
                          3Q
                                      Max
## -2.0336 -0.7262 -0.4318 0.8866
                                   2.2005
##
## Coefficients:
##
            Estimate Std. Error z value Pr(>|z|)
## (Intercept) 1.741588 1.563566 1.114 0.26534
## age
            -0.040291 0.045521 -0.885 0.37611
## lwt
            ## raceblack 1.032835 0.639413 1.615 0.10625
## raceother 0.509939 0.526140 0.969 0.33244
## smokeTRUE 0.358188 0.491957 0.728 0.46656
            1.757537 0.559158 3.143 0.00167 **
## ptdTRUE
## htTRUE
            1.911114 0.885964 2.157 0.03100 *
## uiTRUE
            1.038787 0.537062 1.934 0.05309 .
## ftv1
            -0.864554 0.590892 -1.463 0.14343
## ftv2+
            0.265106 0.502134 0.528 0.59753
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 189.59 on 151 degrees of freedom
##
## Residual deviance: 149.67 on 141 degrees of freedom
## AIC: 171.67
## Number of Fisher Scoring iterations: 5
```

```
test(model,test.data)
```

```
## [,1] [,2]
## [1,] 21 5
## [2,] 9 2
```

```
## $Acc
## [1] 0.6216216
##
## $Specif
## [1] 0.8076923
##
## $Sensit
## [1] 0.1818182
```

```
#Q2.1
step <- stepAIC(model)</pre>
```

```
## Start: AIC=171.67
## train.data$low ~ age + lwt + race + smoke + ptd + ht + ui + ftv
##
##
        Df Deviance AIC
## - smoke 1 150.20 170.20
## - race 2 152.38 170.38
## - age 1 150.47 170.47
## - ftv 2 153.15 171.15
## <none>
            149.66 171.66
## - ui 1 153.41 173.41
         1 154.64 174.64
## - ht
## - lwt 1 156.11 176.11
## - ptd 1 160.41 180.41
##
## Step: AIC=170.2
## train.data$low ~ age + lwt + race + ptd + ht + ui + ftv
##
    Df Deviance AIC
##
## - race 2 152.42 168.42
## - age 1 151.08 169.08
## <none>
           150.20 170.20
## - ftv 2 154.97 170.97
## - ui 1 153.95 171.95
## - ht 1 154.99 172.99
## - lwt 1 157.05 175.05
## - ptd 1 163.12 181.12
##
## Step: AIC=168.42
## train.data$low ~ age + lwt + ptd + ht + ui + ftv
## Df Deviance AIC
## - age 1 154.21 168.21
## <none>
            152.42 168.42
## - ftv 2 157.48 169.48
## - ui 1 156.06 170.06
## - ht 1 157.41 171.41
## - lwt 1 159.16 173.16
## - ptd 1 165.99 179.99
##
## Step: AIC=168.21
## train.data$low ~ lwt + ptd + ht + ui + ftv
##
##
    Df Deviance AIC
## <none> 154.21 168.21
## - ui 1 157.82 169.82
## - ftv 2 160.07 170.07
## - ht 1 159.55 171.55
## - lwt 1 162.11 174.11
## - ptd 1 166.70 178.70
```

```
step$anova
```

```
## Stepwise Model Path
## Analysis of Deviance Table
##
## Initial Model:
## train.data$low ~ age + lwt + race + smoke + ptd + ht + ui + ftv
##
## Final Model:
## train.data$low ~ lwt + ptd + ht + ui + ftv
##
##
##
      Step Df Deviance Resid. Df Resid. Dev
                            141 149.6654 171.6654
## 2 - smoke 1 0.530663
                            142 150.1961 170.1961
                            144 152.4163 168.4163
## 3 - race 2 2.220214
## 4 - age 1 1.790531
                            145 154.2068 168.2068
```

```
test(step,test.data)
```

```
## [,1] [,2]
## [1,] 23 3
## [2,] 9 2
```

```
## $Acc
## [1] 0.6756757
##
## $Specif
## [1] 0.8846154
##
## $Sensit
## [1] 0.1818182
```

```
BIC <- stepAIC(model, k=log(nrow(bwt)))
```

```
## Start: AIC=207.32
## train.data$low ~ age + lwt + race + smoke + ptd + ht + ui + ftv
##
##
        Df Deviance AIC
## - race 2 152.38 199.56
## - ftv 2 153.15 200.32
## - smoke 1 150.20 202.61
## - age 1 150.47 202.88
## - ui 1 153.41 205.83
         1 154.64 207.06
## - ht
            149.66 207.32
## <none>
## - lwt 1 156.11 208.53
## - ptd 1 160.41 212.83
##
## Step: AIC=199.56
## train.data$low ~ age + lwt + smoke + ptd + ht + ui + ftv
##
##
        Df Deviance AIC
## - ftv 2 156.98 193.67
## - smoke 1 152.42 194.35
## - age 1 154.18 196.11
## - ui 1 156.03 197.97
## - ht
         1 157.41 199.34
## <none>
            152.38 199.56
## - lwt 1 159.09 201.02
## - ptd 1 165.06 206.99
##
## Step: AIC=193.67
## train.data$low ~ age + lwt + smoke + ptd + ht + ui
## Df Deviance AIC
## - smoke 1 157.48 188.94
## - age 1 159.49 190.94
## - ui 1 161.25 192.70
## - ht 1 161.81 193.26
            156.98 193.67
## <none>
## - lwt 1 162.39 193.84
## - ptd 1 167.33 198.78
##
## Step: AIC=188.93
## train.data$low ~ age + lwt + ptd + ht + ui
##
## Df Deviance AIC
## - age 1 160.07 186.28
## - ui 1 161.78 187.98
## - ht 1 162.06 188.27
## <none> 157.48 188.94
## - lwt 1 162.93 189.14
## - ptd 1 168.95 195.16
##
## Step: AIC=186.28
```

```
## train.data$low ~ lwt + ptd + ht + ui
##
##
       Df Deviance AIC
## - ui 1 164.21 185.17
## - ht 1 165.08 186.05
## <none> 160.07 186.28
## - lwt 1 166.98 187.95
## - ptd 1 170.21 191.17
##
## Step: AIC=185.17
## train.data$low ~ lwt + ptd + ht
##
##
       Df Deviance AIC
## - ht 1 168.43 184.16
## <none> 164.21 185.17
## - lwt 1 173.03 188.76
## - ptd 1 175.90 191.63
##
## Step: AIC=184.16
## train.data$low ~ lwt + ptd
##
##
      Df Deviance AIC
## <none> 168.43 184.16
## - lwt 1 176.05 186.53
## - ptd 1 180.69 191.18
```

## BIC\$anova

```
## Stepwise Model Path
## Analysis of Deviance Table
## Initial Model:
## train.data$low ~ age + lwt + race + smoke + ptd + ht + ui + ftv
## Final Model:
## train.data$low ~ lwt + ptd
##
##
      Step Df Deviance Resid. Df Resid. Dev
##
## 1
                            141 149.6654 207.3246
## 2 - race 2 2.7173201
                            143 152.3827 199.5585
## 3 - ftv 2 4.5981331
                            145 156.9809 193.6731
## 4 - smoke 1 0.5032579
                            146 157.4841 188.9346
## 5 - age 1 2.5871158
                            147 160.0712 186.2800
## 6
      - ui 1 4.1332841
                           148 164.2045 185.1715
## 7
                           149 168.4304 184.1557
      - ht 1 4.2258991
```

```
test(BIC, test.data)
```

```
[,1] [,2]
## [1,] 24 2
        9
## [2,]
## $Acc
## [1] 0.7027027
## $Specif
## [1] 0.9230769
## $Sensit
## [1] 0.1818182
#Q2.2
library(leaps)
## Warning: package 'leaps' was built under R version 3.6.3
library(bestglm)
## Warning: package 'bestglm' was built under R version 3.6.3
library(dummies)
## dummies-1.5.6 provided by Decision Patterns
library(tidyverse)
bwt.move<-bwt[,-1]</pre>
bwt.move$low<-bwt$low
race = data.frame(dummy(bwt$race)[,c(1,2)])
## Warning in model.matrix.default(\sim x - 1, model.frame(\sim x - 1), contrasts = FALSE):
## non-list contrasts argument ignored
ftv = data.frame(dummy(bwt$ftv)[,c(2,3)])
## Warning in model.matrix.default(\sim x - 1, model.frame(\sim x - 1), contrasts = FALSE):
## non-list contrasts argument ignored
```

```
bwt.dummy = bwt[,-c(1,4,9)]
low = bwt$low
bwt.dummy = cbind(bwt.dummy, race, ftv, low)
BIC.sub = bestglm(bwt.dummy,IC="BIC",family=binomial)
## Morgan-Tatar search since family is non-gaussian.
test(BIC.sub$BestModel,test.data)
## [,1] [,2]
## [1,] 24 2
## [2,] 8
               3
## $Acc
## [1] 0.7297297
## $Specif
## [1] 0.9230769
##
## $Sensit
## [1] 0.2727273
```

#The subset variable BIC seems to be the best, with the stepwise BIC just behind it