STSCI4740HW4

Nick Gembs

10/30/2022

```
library(ISLR)
data=Default
#qlimpse(data)
#1
mylogit <- glm(default ~ income + balance, data = data, family = "binomial")</pre>
summary(mylogit)
##
## Call:
## glm(formula = default ~ income + balance, family = "binomial",
##
       data = data)
##
## Deviance Residuals:
       Min
                      Median
##
                 10
                                   30
                                           Max
## -2.4725 -0.1444 -0.0574 -0.0211
                                        3.7245
##
## Coefficients:
##
                 Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.154e+01 4.348e-01 -26.545 < 2e-16 ***
                2.081e-05 4.985e-06 4.174 2.99e-05 ***
## income
                5.647e-03 2.274e-04 24.836 < 2e-16 ***
## balance
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
       Null deviance: 2920.6 on 9999 degrees of freedom
## Residual deviance: 1579.0 on 9997
                                       degrees of freedom
## AIC: 1585
## Number of Fisher Scoring iterations: 8
#2
set.seed(1)
train = sample(length(data$default), length(data$default)/2)
```

```
training = data[train,]
testing = data[-train,]
mylogit <- glm(default ~ income + balance, data = training, family =
"binomial")
logit.pred=predict(mylogit, data=testing)
pred = (exp(logit.pred))/(1+exp(logit.pred))
for( i in 1:length(pred)){
  if (pred[i] > .5){
  pred[i] = "Yes"
} else {
  pred[i] = "No"
}
}
table(pred, testing$default)
##
## pred
          No Yes
     No 4090 132
##
    Yes 753
                25
mean(pred==testing$default)
## [1] 0.823
cat("Testing Error Rate is:" , 1-mean(pred==testing$default))
## Testing Error Rate is: 0.177
#3
set.seed(2)
train = sample(length(data$default), length(data$default)/2)
training = data[train,]
testing = data[-train,]
mylogit <- glm(default ~ income + balance, data = training, family =
"binomial")
logit.pred=predict(mylogit, data=testing)
pred = (exp(logit.pred))/(1+exp(logit.pred))
for( i in 1:length(pred)){
  if (pred[i] > .5){
  pred[i] = "Yes"
```

```
} else {
  pred[i] = "No"
}
}
table(pred, testing$default)
##
## pred
          No Yes
## No 4292 141
    Yes 545
##
                22
mean(pred==testing$default)
## [1] 0.8628
cat("Testing Error Rate is:" , 1-mean(pred==testing$default))
## Testing Error Rate is: 0.1372
set.seed(3)
train = sample(length(data$default), length(data$default)/2)
training = data[train,]
testing = data[-train,]
mylogit <- glm(default ~ income + balance, data = training, family =</pre>
"binomial")
logit.pred=predict(mylogit, data=testing)
pred = (exp(logit.pred))/(1+exp(logit.pred))
for( i in 1:length(pred)){
  if (pred[i] > .5){
  pred[i] = "Yes"
} else {
  pred[i] = "No"
}
}
table(pred, testing$default)
##
## pred
          No Yes
##
     No 3948 121
##
    Yes 897
                34
mean(pred==testing$default)
## [1] 0.7964
```

```
cat("Testing Error Rate is:" , 1-mean(pred==testing$default))
## Testing Error Rate is: 0.2036
print("After running the logistic regression on 3 different samples, the max
validation error rate was .2036 and the minimum was .1372. It appears that
the model is significantly better than random guessing, but does have
noticeable variance among trials")
## [1] "After running the logistic regression on 3 different samples, the max
validation error rate was .2036 and the minimum was .1372. It appears that
the model is significantly better than random guessing, but does have
noticeable variance among trials"
#4
set.seed(1)
options(contrasts = c("contr.treatment", "contr.helmert")) # dummy
train = sample(length(data$default), length(data$default)/2)
training = data[train,]
testing = data[-train,]
mylogit <- glm(default ~ income + balance + student, data = training, family
= "binomial")
logit.pred=predict(mylogit, data=testing)
pred = (exp(logit.pred))/(1+exp(logit.pred))
for( i in 1:length(pred)){
 if (pred[i] > .5){
 pred[i] = "Yes"
} else {
 pred[i] = "No"
}
}
table(pred, testing$default)
##
          No Yes
## pred
    No 4070 132
##
##
    Yes 773
               25
mean(pred==testing$default)
## [1] 0.819
cat("Testing Error Rate is:" , 1-mean(pred==testing$default))
```

```
## Testing Error Rate is: 0.181
options(contrasts = c("contr.treatment", "contr.helmert")) # dummy
set.seed(2)
train = sample(length(data$default), length(data$default)/2)
training = data[train,]
testing = data[-train,]
mylogit <- glm(default ~ income + balance + student, data = training, family
= "binomial")
logit.pred=predict(mylogit, data=testing)
pred = (exp(logit.pred))/(1+exp(logit.pred))
for( i in 1:length(pred)){
  if (pred[i] > .5){
  pred[i] = "Yes"
} else {
  pred[i] = "No"
}
}
table(pred, testing$default)
##
## pred
          No Yes
##
    No 4239 141
    Yes 598
               22
mean(pred==testing$default)
## [1] 0.8522
cat("Testing Error Rate is:" , 1-mean(pred==testing$default))
## Testing Error Rate is: 0.1478
options(contrasts = c("contr.treatment", "contr.helmert")) # dummy
set.seed(3)
train = sample(length(data$default), length(data$default)/2)
training = data[train,]
testing = data[-train,]
mylogit <- glm(default ~ income + balance +student, data = training, family =
"binomial")
```

```
logit.pred=predict(mylogit, data=testing)
pred = (exp(logit.pred))/(1+exp(logit.pred))
for( i in 1:length(pred)){
  if (pred[i] > .5){
  pred[i] = "Yes"
} else {
  pred[i] = "No"
}
table(pred, testing$default)
##
## pred
          No Yes
##
     No 3903 120
    Yes 942
##
               35
mean(pred==testing$default)
## [1] 0.7876
cat("Testing Error Rate is:" , 1-mean(pred==testing$default))
## Testing Error Rate is: 0.2124
print("After running the logistic regression on 3 different samples, the max
validation error rate was .2124 and the minimum was .1478. It appears that
the model is significantly better than random guessing, but does have
noticeable variance among trials. It does not appear that the student
variable was effective in predicting default. Including a dummy variable for
student does not lead to a reduction in the test error rate")
## [1] "After running the logistic regression on 3 different samples, the max
validation error rate was .2124 and the minimum was .1478. It appears that
the model is significantly better than random guessing, but does have
noticeable variance among trials. It does not appear that the student
variable was effective in predicting default. Including a dummy variable for
student does not lead to a reduction in the test error rate"
#5
library(caret)
## Loading required package: ggplot2
## Loading required package: lattice
library(tidyverse)
```

```
## — Attaching packages
## -
## tidyverse 1.3.2 —
## √ tibble 3.1.8
                       √ dplyr
                                   1.0.10

√ stringr 1.4.1

## √ tidyr 1.2.1
## √ readr
             2.1.3

√ forcats 0.5.2

## √ purrr
             0.3.5
## — Conflicts —
tidyverse_conflicts() —
## X dplyr::filter() masks stats::filter()
## X dplyr::lag()
                     masks stats::lag()
## X purrr::lift()
                     masks caret::lift()
ctrl <- trainControl(method = "cv", number = 5)</pre>
options(contrasts = c("contr.treatment", "contr.helmert")) # dummy
mylogit <- train(default ~ income + balance, data = data, method = "glm",
family = "binomial", trControl = ctrl)
print(mylogit)
## Generalized Linear Model
##
## 10000 samples
##
       2 predictor
##
       2 classes: 'No', 'Yes'
##
## No pre-processing
## Resampling: Cross-Validated (5 fold)
## Summary of sample sizes: 8000, 8000, 8001, 8000, 7999
## Resampling results:
##
##
     Accuracy
                Kappa
##
     0.9732997 0.4282539
logit.pred=predict(mylogit, data=testing)
table(pred, testing$default)
##
## pred
           No Yes
##
     No 3903 120
    Yes 942
##
                35
mean(pred==testing$default)
## [1] 0.7876
cat("Testing Error Rate is:" , 1-mean(pred==testing$default) , "\n")
```

```
## Testing Error Rate is: 0.2124
mylogit <- train(default ~ income + balance + student, data = data, method =</pre>
"glm", family = "binomial", trControl = ctrl)
print(mylogit)
## Generalized Linear Model
##
## 10000 samples
       3 predictor
       2 classes: 'No', 'Yes'
##
##
## No pre-processing
## Resampling: Cross-Validated (5 fold)
## Summary of sample sizes: 8000, 8000, 8000, 8001, 7999
## Resampling results:
##
##
                Kappa
     Accuracy
##
     0.9730001 0.4206439
logit.pred=predict(mylogit, data=testing)
table(pred, testing$default)
##
## pred
           No Yes
##
     No 3903 120
    Yes 942
               35
##
mean(pred==testing$default)
## [1] 0.7876
cat("Testing Error Rate is:" , 1-mean(pred==testing$default))
## Testing Error Rate is: 0.2124
print("5-fold cross-validation yields the same results, adding dummy variable
student does not reduce test error in predicting default.")
## [1] "5-fold cross-validation yields the same results, adding dummy
variable student does not reduce test error in predicting default."
#LOOCV
ctrl <- trainControl(method = "LOOCV")</pre>
options(contrasts = c("contr.treatment", "contr.helmert")) # dummy
mylogit <- train(default ~ income + balance, data = data, method = "glm",
family = "binomial", trControl = ctrl)
```

```
print(mylogit)
logit.pred=predict(mylogit, data=testing)
table(pred,testing$default)
mean(pred==testing$default)

cat("Testing Error Rate is:" , 1-mean(pred==testing$default) , "\n")

mylogit <- train(default ~ income + balance + student, data = data, method =
"glm", family = "binomial", trControl = ctrl)

print(mylogit)
logit.pred=predict(mylogit, data=testing)
table(pred,testing$default)
mean(pred==testing$default)

cat("Testing Error Rate is:" , 1-mean(pred==testing$default))</pre>
```