

LASSO_NN

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10/29/2020

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.0 --
## v ggplot2 3.3.2      v purrr  0.3.4
## v tibble  3.0.4      v dplyr  1.0.2
## 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 3.6.2
## Warning: package 'tibble' was built under R version 3.6.2
## Warning: package 'tidyr' was built under R version 3.6.2
## Warning: package 'readr' was built under R version 3.6.2
## Warning: package 'purrr' was built under R version 3.6.2
## Warning: package 'dplyr' was built under R version 3.6.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(readr)
library(caret)

## Loading required package: lattice
## Warning: package 'lattice' was built under R version 3.6.2
##
## Attaching package: 'caret'
## The following object is masked from 'package:purrr':
##
##   lift

library(WeightedROC)
source("../LASSO_NN.R")

## Warning: package 'glmnet' was built under R version 3.6.2
## Loading required package: Matrix
##
## Attaching package: 'Matrix'
## The following objects are masked from 'package:tidyr':
##
```

```

##      expand, pack, unpack
## Loaded glmnet 4.0-2
##
## Attaching package: 'neuralnet'
## The following object is masked from 'package:dplyr':
##
##      compute
## Warning: package 'e1071' was built under R version 3.6.2
datatrain = read_csv("../data/feature_train_balanced_data.csv")

##
## -- Column specification -----
## cols(
##   .default = col_double()
## )
## i Use `spec()` for the full column specifications.
datatest = read_csv("../data/feature_test_data.csv")

##
## -- Column specification -----
## cols(
##   .default = col_double()
## )
## i Use `spec()` for the full column specifications.
location = which(colnames(datatrain) == "HEALTH")
trainx = datatrain[,-location]
trainy = datatrain$HEALTH
testx = datatest[, -location]
testy = datatest$HEALTH

weight_test <- rep(NA, length(testy))
for (v in unique(testy)){
  weight_test[testy == v] = 0.5 * length(testy) / length(testy[testy == v])
}

```

LASSO method

```

begin <- Sys.time()
lasso <- lassoReg(trainx, trainy)
summary(lasso)

##      Length Class      Mode
## a0          1   -none-  numeric
## beta        50 dgCMatrix S4
## df           1   -none-  numeric
## dim          2   -none-  numeric
## lambda       1   -none-  numeric
## dev.ratio    1   -none-  numeric
## nulldev      1   -none-  numeric
## npasses      1   -none-  numeric

```

```
## jerr      1      -none-    numeric
## offset   1      -none-    logical
## call      7      -none-    call
## nobs      1      -none-    numeric
```

```
coef <- rbind("(intercept)" = lasso$a0, as.data.frame(as.matrix(lasso$beta))) %>%
  dplyr::arrange(desc(abs(s0))) %>% rename(Coef = s0)
```

```
coef
```

```
##              Coef
## (intercept)  0.5382698332
## FS1_1        -0.1200926654
## fpl          0.0737881831
## FWB2_1       -0.0621141873
## FS2_1        -0.0581033770
## FS1_4        -0.0557699516
## SWB_2         0.0528259504
## FWB2_4       -0.0458344678
## FS1_2        -0.0424773711
## FWB1_1        0.0398995276
## ACT1_1        0.0390377814
## SWB_1         0.0313843533
## SUBKNOWL1     0.0307676681
## FS1_7        -0.0303872325
## FWB1_3       -0.0293556904
## FS1_5        -0.0257466739
## ASK1_1       -0.0228183295
## FSscore       0.0221663563
## FS1_3        -0.0201078089
## FWB1_2        0.0194234243
## SUBNUMERACY1 0.0190647316
## FINGOALS      0.0184990470
## FS2_3         0.0181511710
## MANAGE1_4     -0.0169994161
## PROPPLAN_3    0.0151785494
## PROPPLAN_1   -0.0142079674
## SWB_3         0.0127998433
## ASK1_2        0.0125702918
## FWB1_6       -0.0123974355
## MANAGE1_3     -0.0123160248
## FWBscore     -0.0120046555
## sample       -0.0105703310
## FWB2_3       -0.0098978392
## FWB1_4        0.0092783613
## GOALCONF      0.0091858050
## FS1_6         0.0090480505
## CHANGEABLE   -0.0087448510
## LMscore      -0.0085703329
## MANAGE1_2     -0.0075003280
## FWB2_2        0.0074860885
## MANAGE1_1     0.0070861582
## AUTOMATED_1  -0.0055887392
## ACT1_2        0.0043000740
## FWB1_5       -0.0029075900
```

```
## FS2_2          -0.0026010412
## SUBNUMERACY2  -0.0025140813
## SAVEHABIT     0.0020835779
## FRUGALITY     -0.0011212105
## AUTOMATED_2   0.0008323304
## PROPPLAN_4    0.0002998221
## PROPPLAN_2    0.0000000000

pred <- lassoPred(lasso, testx)
pred <- ifelse(pred > mean(pred), 1, 0)

cat("MSE is", mean((pred - testy)^2), ". Accuracy is", mean(pred==testy), ".")

## MSE is 0.3652283 . Accuracy is 0.6347717 .

confusionMatrix(factor(pred), factor(testy))$byClass

##           Sensitivity      Specificity      Pos Pred Value
##           0.8016194      0.6042899      0.2701228
##           Neg Pred Value      Precision      Recall
##           0.9434180      0.2701228      0.8016194
##           F1      Prevalence      Detection Rate
##           0.4040816      0.1544715      0.1238274
## Detection Prevalence      Balanced Accuracy
##           0.4584115      0.7029547

tpr.fpr <- WeightedROC(pred, testy, weight_test)
auc.log <- WeightedAUC(tpr.fpr)

cat("AUC is", auc.log)

## AUC is 0.7029547

end <- Sys.time()
timedif <- end - begin
cat("Time for running", "LASSO", "is", timedif)

## Time for running LASSO is 0.520165

timedif

## Time difference of 0.520165 secs
```

SVM

SVM Tune

```
begin <- Sys.time()
source("./svm.R")
# Tune SVM
set.seed(2020)
opt.svm <- svm_tune(scale(as.matrix(trainx)), trainy)
bestgamma = opt.svm$best.parameters$gamma
bestcost = opt.svm$best.parameters$cost

# tune svm is very time consuming, takes about 5 hours to run
# tuned result is gamma=0.01, cost = 0.21
bestgamma; bestcost
```

```
## [1] 0.001
## [1] 0.21
end <- Sys.time()
timedif <- end - begin
cat("Time for tuning","SVM","is", timedif)
```

```
## Time for tuning SVM is 21.8611
timedif

## Time difference of 21.8611 mins
```

SVM Train

```
begin <- Sys.time()
svm_fit <- svm_train(scale(as.matrix(trainx)), trainy, bestgamma, bestcost)
#svm_fit <- svm_train(scale(as.matrix(trainx)), trainy, 0.01, 0.11)
```

Step 5.1(b): Test SVM with with tuning parameters

```
svm_pred <- svm_test(svm_fit, scale(as.matrix(testx)))

svm.pred <- ifelse(svm_pred > mean(svm_pred), 1, 0)

# Calculate Accuracy
cat("MSE is",mean((svm.pred-testy)^2),
    "and accuracy is",mean(svm.pred==testy))
```

```
## MSE is 0.355222 and accuracy is 0.644778
```

```
confusionMatrix(factor(svm.pred), factor(testy))$byClass
```

##	Sensitivity	Specificity	Pos Pred Value
##	0.7773279	0.6205621	0.2723404
##	Neg Pred Value	Precision	Recall
##	0.9384787	0.2723404	0.7773279
##	F1	Prevalence	Detection Rate
##	0.4033613	0.1544715	0.1200750
##	Detection Prevalence	Balanced Accuracy	
##	0.4409006	0.6989450	

```
tpr.fpr <- WeightedROC(svm.pred, testy, weight_test)
auc.svm <- WeightedAUC(tpr.fpr)
cat("AUC is",auc.svm)
```

```
## AUC is 0.698945
```

```
end <- Sys.time()
timedif <- end - begin
cat("Time for running","SVM","is", timedif)
```

```
## Time for running SVM is 2.912898
timedif
```

```
## Time difference of 2.912898 mins
```

Neural Net

```
source("./LASSO_NN.R")
begin <- Sys.time()
nndata <- data.frame(
  scale(as.matrix(trainx)), HEALTH = trainy)

nnfit <- nnReg(nndata, hd = c(35, 25, 18, 13, 8, 5))

plot(nnfit)

end <- Sys.time()
timedif <- end - begin
cat("Time for training", "neuralnet", "is", timedif)

## Time for training neuralnet is 27.70749
timedif

## Time difference of 27.70749 secs
begin <- Sys.time()

pred_nn <- predict(nnfit, testx)
summary(pred_nn)

##           V1           V2
## Min.      :0.0000008   Min.      :0.0000000
## 1st Qu.:0.0000083     1st Qu.:0.0000000
## Median :0.9991259     Median :0.0004897
## Mean     :0.5476543     Mean     :0.4543838
## 3rd Qu.:0.9999998     3rd Qu.:0.9999977
## Max.     :1.0000000     Max.     :0.9999998

nn.prediction <- rep(0, length(testy))

for(i in 1:length(nn.prediction)){
  if(pred_nn[i,1] < pred_nn[i,2]){
    nn.prediction[i] <- 0
  } else{
    nn.prediction[i] <- 1
  }
}
summary(nn.prediction)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.0000 0.0000  1.0000  0.5453  1.0000  1.0000

confusionMatrix(factor(nn.prediction), factor(testy))$byClass

##           Sensitivity           Specificity           Pos Pred Value
##           0.42105263           0.53920118           0.14305365
##           Neg Pred Value           Precision           Recall
##           0.83600917           0.14305365           0.42105263
##           F1           Prevalence           Detection Rate
##           0.21355236           0.15447154           0.06504065
## Detection Prevalence           Balanced Accuracy
```

```
##           0.45465916           0.48012691
tpr.fpr <- WeightedROC(nn.prediction, testy, weight_test)
auc.nn <- WeightedAUC(tpr.fpr)

cat("AUC is", auc.nn)

## AUC is 0.4801269
end <- Sys.time()
timedif <- end - begin
cat("Time for predicting using","neuralnet","is", timedif)

## Time for predicting using neuralnet is 0.03090096
timedif

## Time difference of 0.03090096 secs
```

Naive Bayies

```
begin <- Sys.time()

nbfit = nbReg(datatrain)
nbpred = predict(nbfit, testx)

mean((as.numeric(nbpred)-1 - testy)^2)

## [1] 0.3327079

confusionMatrix(nbpred, factor(testy))$byClass

##           Sensitivity           Specificity           Pos Pred Value
##           0.6761134           0.6656805           0.2697900
##           Neg Pred Value           Precision           Recall
##           0.9183673           0.2697900           0.6761134
##           F1           Prevalence           Detection Rate
##           0.3856813           0.1544715           0.1044403
## Detection Prevalence           Balanced Accuracy
##           0.3871169           0.6708969

tpr.fpr <- WeightedROC(as.numeric(nbpred)-1, testy, weight_test)
auc.nb <- WeightedAUC(tpr.fpr)

cat("AUC is", auc.nb)

## AUC is 0.6708969
end <- Sys.time()
timedif <- end - begin
cat("Time for running","Naive Bayies","is", timedif)

## Time for running Naive Bayies is 0.837626
timedif

## Time difference of 0.837626 secs
```