# LASSO NN

### Cavalry

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```
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 -----
                                      ## 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))</pre>
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)</pre>
```

```
##
            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
                 -none-
                            numeric
```

```
## jerr
              1
                    -none-
                               numeric
## offset
                               logical
              1
                    -none-
## call
              7
                    -none-
                               call
## nobs
                               numeric
              1
                    -none-
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
                -0.1200926654
## FS1 1
## 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
               0.0020835779
## SAVEHABIT
## FRUGALITY
                -0.0011212105
## AUTOMATED 2 0.0008323304
## PROPPLAN 4
                 0.0002998221
## PROPPLAN 2
                 0.000000000
pred <- lassoPred(lasso, testx)</pre>
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
                                                    Pos Pred Value
##
            Sensitivity
                                  Specificity
              0.8016194
                                    0.6042899
                                                          0.2701228
##
         Neg Pred Value
##
                                    Precision
                                                             Recall
                                    0.2701228
                                                          0.8016194
##
              0.9434180
##
                     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)</pre>
auc.log <- WeightedAUC(tpr.fpr)</pre>
cat("AUC is", auc.log)
## AUC is 0.7029547
end <- Sys.time()</pre>
timedif <- end - begin
cat("Time for running","LASSO","is", timedif)
## Time for running LASSO is 0.531116
timedif
## Time difference of 0.531116 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</pre>
```

```
## [1] 0.001
## [1] 0.21
end <- Sys.time()</pre>
timedif <- end - begin
cat("Time for tuning", "SVM", "is", timedif)
## Time for tuning SVM is 21.75782
timedif
## Time difference of 21.75782 mins
SVM Train
begin <- Sys.time()</pre>
svm fit <- svm train(scale(as.matrix(trainx)), trainy, bestgamma, bestcost)</pre>
#svm_fit <- svm_train(scale(as.matrix(trainx)), trainy, 0.01, 0.11)</pre>
Step 5.1(b): Test SVM with tuning parameters
svm_pred <- svm_test(svm_fit, scale(as.matrix(testx)))</pre>
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
                                   Prevalence
                                                     Detection Rate
##
                      F1
##
              0.4033613
                                     0.1544715
                                                           0.1200750
## Detection Prevalence
                            Balanced Accuracy
              0.4409006
                                    0.6989450
tpr.fpr <- WeightedROC(svm.pred, testy, weight_test)</pre>
auc.svm <- WeightedAUC(tpr.fpr)</pre>
cat("AUC is",auc.svm)
## AUC is 0.698945
end <- Sys.time()</pre>
timedif <- end - begin
cat("Time for running", "SVM", "is", timedif)
## Time for running SVM is 2.897044
timedif
```

## Time difference of  $2.897044 \ \mathrm{mins}$ 

## Neural Net

```
source("./LASSO NN.R")
begin <- Sys.time()</pre>
nndata <- data.frame(</pre>
  scale(as.matrix(trainx)), HEALTH = trainy)
nnfit \leftarrow nnReg(nndata, hd = c(35, 25, 18, 13, 8, 5))
plot(nnfit)
end <- Sys.time()</pre>
timedif <- end - begin
cat("Time for training", "neuralnet", "is", timedif)
## Time for training neuralnet is 27.45296
timedif
## Time difference of 27.45296 secs
begin <- Sys.time()</pre>
pred_nn <- predict(nnfit, testx)</pre>
summary(pred_nn)
##
          V1
                               ٧2
## Min.
           :0.0000008
                         Min.
                                :0.0000000
## 1st Qu.:0.0000083
                         1st Qu.:0.0000000
## Median :0.9991259
                         Median: 0.0004897
## Mean
           :0.5476543
                                :0.4543838
## 3rd Qu.:0.9999998
                         3rd Qu.:0.9999977
## Max.
           :1.0000000
                         Max.
                                :0.9999998
nn.prediction <- rep(0, length(testy))</pre>
for(i in 1:length(nn.prediction)){
  if(pred_nn[i,1] < pred_nn[i,2]){</pre>
    nn.prediction[i] <- 0
  } else{
    nn.prediction[i] <- 1</pre>
}
summary(nn.prediction)
      Min. 1st Qu. Median
                               Mean 3rd Qu.
  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
                                                          0.42105263
##
             0.83600917
                                    0.14305365
##
                      F1
                                    Prevalence
                                                      Detection Rate
                                                          0.06504065
##
             0.21355236
                                    0.15447154
## Detection Prevalence
                            Balanced Accuracy
```

```
0.45465916
                                   0.48012691
##
tpr.fpr <- WeightedROC(nn.prediction, testy, weight_test)</pre>
auc.nn <- WeightedAUC(tpr.fpr)</pre>
cat("AUC is", auc.nn)
## AUC is 0.4801269
end <- Sys.time()</pre>
timedif <- end - begin
cat("Time for predicting using", "neuralnet", "is", timedif)
## Time for predicting using neuralnet is 0.03086591
timedif
## Time difference of 0.03086591 secs
Naive Bayies
begin <- Sys.time()</pre>
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
                                    0.6656805
                                                          0.2697900
##
              0.6761134
##
         Neg Pred Value
                                    Precision
                                                              Recall
##
              0.9183673
                                    0.2697900
                                                          0.6761134
##
                     F1
                                   Prevalence
                                                     Detection Rate
                                                          0.1044403
              0.3856813
                                    0.1544715
##
## Detection Prevalence
                          Balanced Accuracy
              0.3871169
                                    0.6708969
tpr.fpr <- WeightedROC(as.numeric(nbpred)-1, testy, weight_test)</pre>
auc.nb <- WeightedAUC(tpr.fpr)</pre>
cat("AUC is", auc.nb)
## AUC is 0.6708969
end <- Sys.time()</pre>
timedif <- end - begin
cat("Time for running", "Naive Bayies", "is", timedif)
## Time for running Naive Bayies is 0.8138459
timedif
```

## Time difference of 0.8138459 secs