Lab Report Ridgereg

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Here we instantiate the training and test data with the training data being 80% and test data being 20% of the data set.

We now fit the data to the models using linear regression and linear regression with forward selection on the covariates.

We evaluate the models with analyzing the RMSE and \mathbb{R}^2 values.

linMod\$results\$RMSE

[1] 54.07378

linFMod\$results\$RMSE

```
## [1] 57.07212 56.18481 54.11274
```

Analyzing the RMSE and the R^2 of the models entails some information. Lower RMSE value would indicate a 'tighter fit' of the data and a higher R^2 value indicates a measure of how well observed outcomes are replicated by the model, based on the proportion of total variation of outcomes explained by the model.

[1] 10

```
bestRMSE
```

```
## [1] 406.4993
```

We see from the code above that the best $\lambda = 0.5$ with the lowest RMSE of 1320.828.

```
# Acknowledgement to Eric Herwin and Albin Vasterlund
set.seed(theSeed)
fold_count <- 10</pre>
lambda \leftarrow seq(10, 20, by=1)
fitControl <- caret::trainControl(method = "repeatedcv",</pre>
                                   number = fold_count,
                                   ## repeated ten times
                                   repeats = fold_count)
ridgeMod <- caret::train(form,</pre>
                         data = trainDat,
                         method = ridgeMod,
                         trControl = fitControl)
## Warning in nominalTrainWorkflow(x = x, y = y, wts = weights, info =
## trainInfo, : There were missing values in resampled performance measures.
ridgeMod
## 406 samples
   13 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 10 times)
## Summary of sample sizes: 366, 365, 364, 366, 366, 366, ...
## Resampling results across tuning parameters:
##
##
     lambda RMSE
                       Rsquared
                                  MAE
##
     10
             410.1854 0.8887925 406.335
##
             410.1892 0.8887097 406.335
     11
##
     12
             410.1934 0.8886175 406.335
##
     13
             410.1980 0.8885163 406.335
##
     14
             410.2030 0.8884068 406.335
             410.2084 0.8882895 406.335
##
     15
##
     16
             410.2142 0.8881647 406.335
##
     17
             410.2202 0.8880330 406.335
##
             410.2266 0.8878946 406.335
     18
##
     19
             410.2333 0.8877500 406.335
##
     20
             410.2402 0.8875995 406.335
##
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was lambda = 10.
```

Last line says the best value for $\lambda = 17$. Which is the same as as we concluded above, though with different RMSE. The interval for $\lambda \in 10$: 20 is motivated by previous larger interval, but for output niceties we chose to limit lambda.

```
lin <- list(RMSE = linMod$results$RMSE, RSquared = linMod$results$Rsquared)
linF <- list(RMSE = linFMod$results$RMSE[3], RSquared = linFMod$results$Rsquared[3])
ridgeM <- list(RMSE = ridgeMod$results$RMSE[7], RSquared = ridgeMod$results$Rsquared[7])
temp <- cbind(lin,linF,ridgeM)
colnames(temp) <- c("Lin. Reg", "Lin. Reg. F", "Ridge Reg.")</pre>
```

temp

```
## RMSE 54.07378 Lin. Reg. F Ridge Reg.
## RSquared 0.8895151 0.8912715 0.8881647
```