Data Science II Midterm

Megan Panier, Shiying Wu, and Rita Wang 2025-03-25

Libraries

```
library(readxl) # to import excel files
library(tidyverse)
library(ggplot2)
library(tidymodels)
library(glmnet)
library(caret)
library(splines)
library(mgcv)
library(pROC)
library(pROC)
library(pdp)
library(vip)
library(AppliedPredictiveModeling)
```

Importing and Organizing Data

```
load("./data/dat1.RData") #importing training data
    # Log-transformed antibody level (log_antibody) --> y
initial_training = dat1 #renaming the original training data name

load("./data/dat2.RData") #importing training data
initial_test = dat2 #renaming the original training data name

set.seed(2222)

# partition data into training and validation data sets
datSplit = initial_split(data = initial_training, prop = 0.8)
training = training(datSplit)
validation = testing(datSplit)
```

Linear Regression

```
# View the model summary
summary(model)
##
## Call:
## lm(formula = log_antibody ~ age + gender + race + smoking + height +
      weight + bmi + diabetes + hypertension + SBP + LDL + time,
##
      data = training)
##
## Residuals:
       Min
               1Q Median
                                       Max
                                30
## -2.14743 -0.35065 0.03211 0.37738 1.53018
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 28.2069787 2.6457948 10.661 < 2e-16 ***
             -0.0196829 0.0021607 -9.110 < 2e-16 ***
## age
## gender
             -0.0139482 0.0386090 -0.361
## race2
                                         0.7179
            -0.0080486 0.0218346 -0.369
## race3
                                          0.7124
## race4
             -0.0463573 0.0301577 -1.537
                                         0.1243
## smoking1
             0.0219875 0.0193608 1.136 0.2562
             ## smoking2
             -0.0919586  0.0154999  -5.933  3.23e-09 ***
## height
## weight
             -0.3264716 0.0471923 -6.918 5.32e-12 ***
## bmi
## diabetes
            0.0030653 0.0243426 0.126 0.8998
## hypertension -0.0287531 0.0290736 -0.989
                                         0.3227
## SBP
         0.0024700 0.0019002 1.300 0.1937
## LDL
             -0.0001017 0.0004518 -0.225 0.8219
             -0.0003804 0.0001988 -1.914 0.0557 .
## time
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.5471 on 3984 degrees of freedom
## Multiple R-squared: 0.147, Adjusted R-squared: 0.1438
## F-statistic: 45.78 on 15 and 3984 DF, p-value: < 2.2e-16
predictions_train = predict(model, newdata = validation)
# RMSE
rmse_train = sqrt(mean((predictions_train - validation$log_antibody)^2))
rmse train
## [1] 0.5639064
rsq train = 1 - sum((predictions train - validation$log antibody)^2) /
 sum((mean(training$log_antibody) - validation$log_antibody)^2)
rsq_train
```

[1] 0.1641537

```
generalization = predict(model, newdata = initial_test)
# Calculate RMSE for dat2
rmse_dat2 = sqrt(mean((generalization - initial_test$log_antibody)^2))
rmse_dat2
```

[1] 0.5662817

```
# Calculate R-squared for dat2
rsq_dat2 = 1 - sum((generalization - initial_test$log_antibody)^2) /
sum((mean(initial_test$log_antibody) - initial_test$log_antibody)^2)
rsq_dat2
```

[1] 0.06952672

```
ggplot(initial_training, aes(x = time, y = log_antibody)) +
geom_point() +
geom_smooth(method = "lm", se = FALSE) +
labs(title = "Log Antibody Levels Over Time Since Vaccination")
```

Log Antibody Levels Over Time Since Vaccination

