## yacht\_hydrodynamics

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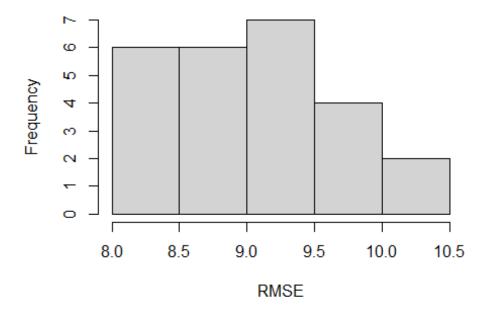
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```
## Installing the necessary packages for the problem ##
#install.packages('readr')
                               ## yacht_hydrodynamics.data is a Large
dataset. So, I used readr package in handling the data
                               ## To use machine Learning models, I used
#install.packages('caret')
caret to fit our model
#install.packages('qqplot2') ## used qqplot2 for better visualizations of
data
#install.packages('lattice') ## Lattice is used to implement the trellis
graphics for our data
# Loading the libraries
library(readr)
library(data.table)
library(caret)
## Loading required package: ggplot2
## Loading required package: lattice
library(ggplot2)
library(lattice)
# Reading the yacht hydrodynamics.data as the table without the header
yacht_hydrodynamics = read.table("https://archive.ics.uci.edu/ml/machine-
learning-databases/00243/yacht_hydrodynamics.data", header = F)
# Assigning the column names for our dataset
names(yacht hydrodynamics) = c("longitude", "Prismatic", "displacement", "beam-
draught", "beamlenght", "fraude", "residuary")
head(yacht_hydrodynamics)
     longitude Prismatic displacement beam-draught beamlenght fraude
residuary
## 1
          -2.3
                   0.568
                                 4.78
                                              3.99
                                                         3.17 0.125
0.11
## 2
          -2.3
                   0.568
                                 4.78
                                              3.99
                                                         3.17 0.150
0.27
## 3
          -2.3
                   0.568
                                 4.78
                                              3.99
                                                         3.17 0.175
0.47
## 4
          -2.3
                   0.568
                                 4.78
                                              3.99
                                                         3.17 0.200
0.78
                                 4.78
## 5
          -2.3
                   0.568
                                              3.99
                                                         3.17 0.225
```

```
1.18
## 6
          -2.3
                                 4.78
                                              3.99
                   0.568
                                                         3.17 0.250
1.82
# Creating the data partition for our data having 80% our data for the
training. So the rest 20% is for testing.
# I used the caret package to perform a 80/20 test-train split
cd = createDataPartition(y = yacht hydrodynamics$residuary , p = 0.8, list =
FALSE)
# Separating the dataset for the train data
train_data = yacht_hydrodynamics[cd,]
# Separating the test data without the output label data.
test data = yacht hydrodynamics[-cd,]
# Applying the linear regression model for the dataset
# Applying the multiple linear regression
lm1 = lm(yacht hydrodynamics$residuary~yacht hydrodynamics$longitude +
yacht hydrodynamics$Prismatic +
            yacht_hydrodynamics$displacement + yacht_hydrodynamics$`beam-
draught` + yacht hydrodynamics$`beam-draught` +
            yacht_hydrodynamics$displacement + yacht_hydrodynamics$fraude,
             data = train_data)
# creating a function for the mean square error
mse = function(y, yt){
  return (mean((y - yt)^2))
}
# Applying the mean square error for the residuary and the fitted values for
the linear regression model.
msee = mse(yacht hydrodynamics$residuary, lm1$fitted.values )
msee
## [1] 78.47651
cat("\n The MSE for the training data is = ", msee)
##
## The MSE for the training data is = 78.47651
cat("\n The Root mean square error for the train data is = ", sqrt(msee))
##
## The Root mean square error for the train data is = 8.858697
cat("\n The summary for the r-squared data for the linear model is =
", summary(lm1)$r.sq)
##
## The summary for the r-squared data for the linear model is = 0.6574487
```

```
# train control specify the resampling scheme
# I used the caret package to perform a bootstrap from the full sample
dataset with N=1000 samples
train = trainControl(method = "boot", number = 1000)
lm2 = train(residuary~., data = train_data, method = "lm" )
# summary of the model
summary(lm2$resample$RMSE)
##
     Min. 1st Qu.
                   Median
                             Mean 3rd Qu.
                                              Max.
            8,679
##
     8.243
                     9.306
                             9.109
                                     9.478 10.337
summary(1m2$resample$Rsquared)
##
      Min. 1st Qu.
                    Median
                              Mean 3rd Qu.
                                              Max.
## 0.5690 0.6301 0.6483 0.6457 0.6580
                                            0.7025
# Plotting a histogram for the resampled data and the root mean square error
hist(lm2$resample$RMSE, xlab = "RMSE", main = "Histogram of RMSE")
```

## **Histogram of RMSE**



```
# applying the mean for the resampled data as the mse2
mse2 = mean(lm2$resample$RMSE)^2
mse2
## [1] 82.96775
cat("\n Training MSE for the bootstrap model is = ", mse2)
```

```
##
## Training MSE for the bootstrap model is = 82.96775
cat("\n Training RMSE for the bootstrap model is ", mean(lm2$resample$RMSE))
##
## Training RMSE for the bootstrap model is
                                              9.108663
cat("\n Training Mean R-squared for the bootstrap model is
",mean(lm2$resample$Rsquared))
##
## Training Mean R-squared for the bootstrap model is 0.6457281
predVals_boot = predict(lm2,test_data)
cat("\n From the above observations, there is no difference in performance
between the original and bootstrap models.")
##
## From the above observations, there is no difference in performance
between the original and bootstrap models.
```