assignment_2

2024-10-16

```
library(caTools)
## Warning: package 'caTools' was built under R version 4.3.3
library(ggplot2)
```

Problem 1. Regression

```
data <- read.csv("qsar_aquatic_toxicity.csv", sep = ";", header = FALSE)
names(data) <- c(
    "TPSA",
    "SAacc",
    "H050",
    "ML0GP",
    "RDCHI",
    "GATS1p",
    "nN",
    "C040",
    "LC50"
)</pre>
```

```
##
      TPSA
            SAacc H050 MLOGP RDCHI GATS1p nN C040 LC50
## 1
      0.00
            0.000 0 2.419 1.225 0.667 0
                                             0 3.740
## 2
     0.00 0.000 0 2.638 1.401 0.632 0
                                             0 4.330
      9.23 11.000 0 5.799 2.930 0.486 0
                                             0 7.019
## 4
      9.23 11.000
                  0 5.453 2.887 0.495 0
                                             0 6.723
     9.23 11.000
                    0 4.068 2.758 0.695 0
                                             0 5.979
## 6 215.34 327.629
                    3 0.189 4.677 1.333 0
                                             4 6.064
```

a. Split the data into a training and test set

```
# Use 70% of dataset as training set and remaining 30% as testing set
sample <- sample.split(data$LC50, SplitRatio = 0.7)
train <- subset(data, sample == TRUE)
test <- subset(data, sample == FALSE)</pre>
```

```
dim(train)
## [1] 382
dim(test)
## [1] 164
# Fit linear regression model on training data
model <- lm(LC50 ~ ., data=train)</pre>
summary(model)
##
## Call:
## lm(formula = LC50 \sim ., data = train)
##
## Residuals:
##
      Min
               1Q Median
                              3Q
                                    Max
## -4.3666 -0.7729 -0.0625 0.6028 5.0378
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.462240 0.301036 8.179 4.54e-15 ***
## TPSA
              ## SAacc
             ## H050
              0.095604 0.072157 1.325 0.18600
## MLOGP
             0.488611
                         0.077813 6.279 9.47e-10 ***
## RDCHI
              -0.450424   0.187565   -2.401   0.01682 *
## GATS1p
## nN
             -0.147462   0.057226   -2.577   0.01035 *
## CO40
              0.026590 0.090711 0.293 0.76959
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.202 on 373 degrees of freedom
## Multiple R-squared: 0.4954, Adjusted R-squared: 0.4846
## F-statistic: 45.77 on 8 and 373 DF, p-value: < 2.2e-16
# Predict on training and test datasets
pred_train <- predict(model, newdata=train)</pre>
pred_test <- predict(model, newdata=test)</pre>
# Adding predictions columns to the datasets
train$predicted_LC50 <- pred_train</pre>
test$predicted_LC50 <- pred_test</pre>
# Evaluate model: calculate MSE, RMSE, and R-squared for training and test sets
mse_train <- mean((train$LC50 - train$predicted_LC50)^2)</pre>
rmse train <- sqrt(mse train)</pre>
r2_train <- 1 - (sum((train$LC50 - train$predicted_LC50)^2) / sum((train$LC50 - mean(train$LC50))^2))
```

```
mse_test <- mean((test$LC50 - test$predicted_LC50)^2)</pre>
rmse_test <- sqrt(mse_test)</pre>
r2_{test} \leftarrow 1 - (sum((test_{LC50} - test_{predicted_LC50})^2) / sum((test_{LC50} - mean(test_{LC50}))^2))
# Print evaluation metrics
cat("Training Metrics:\n")
## Training Metrics:
cat("MSE (Train): ", mse_train, "\n")
## MSE (Train): 1.410874
cat("RMSE (Train): ", rmse_train, "\n")
## RMSE (Train): 1.187802
cat("R-squared (Train): ", r2_train, "\n\n")
## R-squared (Train): 0.4953916
cat("Test Metrics:\n")
## Test Metrics:
cat("MSE (Test): ", mse_test, "\n")
## MSE (Test): 1.499417
cat("RMSE (Test): ", rmse_test, "\n")
## RMSE (Test): 1.224507
cat("R-squared (Test): ", r2_test, "\n")
## R-squared (Test): 0.4450487
# Combine data for plotting
train$Dataset <- 'Train'</pre>
test$Dataset <- 'Test'</pre>
plot_data <- rbind(train, test)</pre>
# Plot observed vs predicted LC50 values
ggplot(plot_data, aes(x = LC50, y = predicted_LC50, color = Dataset)) +
  geom_point(alpha = 0.7) +
  geom_abline(intercept = 0, slope = 1, linetype = "dashed") +
  labs(title = "Observed vs Predicted LC50", x = "Observed LC50", y = "Predicted LC50") +
  theme minimal() +
  facet_wrap(~Dataset)
```

Observed vs Predicted LC50

