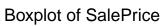
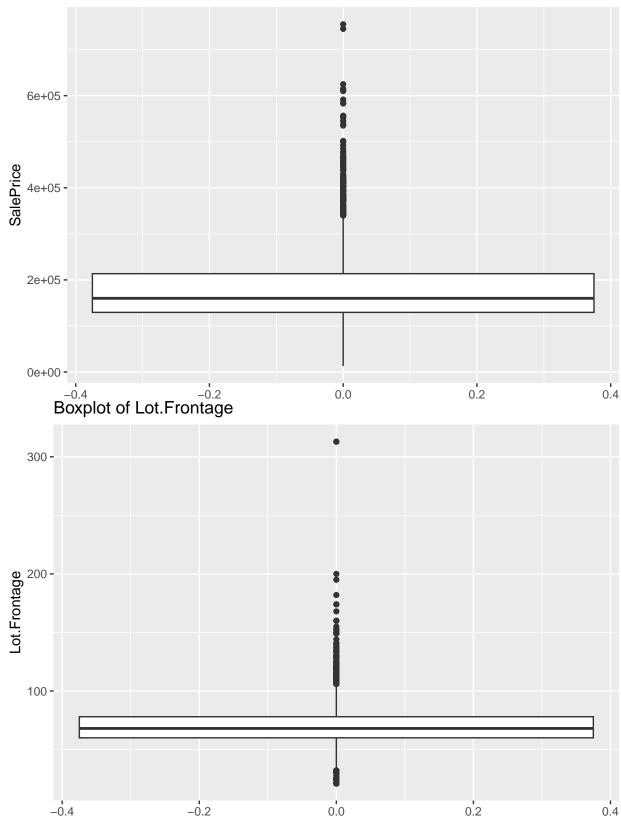
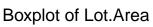
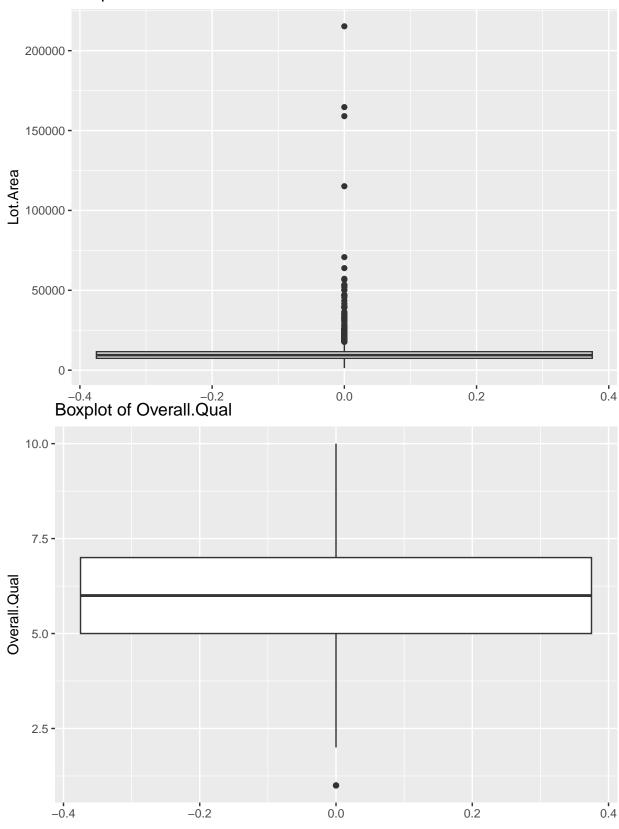
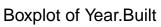
```
#Requirement 1: Preprocessing and Exploratory Data Analysis
housing <- read.table("AmesHousing.txt", header = TRUE, sep = "\t")
# For simplicity, impute missing values with the median (you may choose another method)
housing$Lot.Frontage[is.na(housing$Lot.Frontage)] <- median(housing$Lot.Frontage, na.rm = TRUE)
housing$Mas.Vnr.Area[is.na(housing$Mas.Vnr.Area)] <- median(housing$Mas.Vnr.Area, na.rm = TRUE)
housing$BsmtFin.SF.1[is.na(housing$BsmtFin.SF.1)] <- median(housing$BsmtFin.SF.1, na.rm = TRUE)
library(ggplot2)
library(corrplot)
## corrplot 0.92 loaded
library(carData)
library(lattice)
library(car)
library(caret)
library(Matrix)
library(glmnet)
## Loaded glmnet 4.1-8
# Boxplot for selected continuous variables
selected_vars <- c("SalePrice","Lot.Frontage", "Lot.Area", "Overall.Qual", "Year.Built", "Overall.Qual"</pre>
# Iterate through the selected variables and plot boxplots
for (var in selected_vars) {
  boxplot <- ggplot(housing, aes(y = .data[[var]])) +</pre>
    geom boxplot() +
    ggtitle(paste("Boxplot of", var))
  plot(boxplot)
```

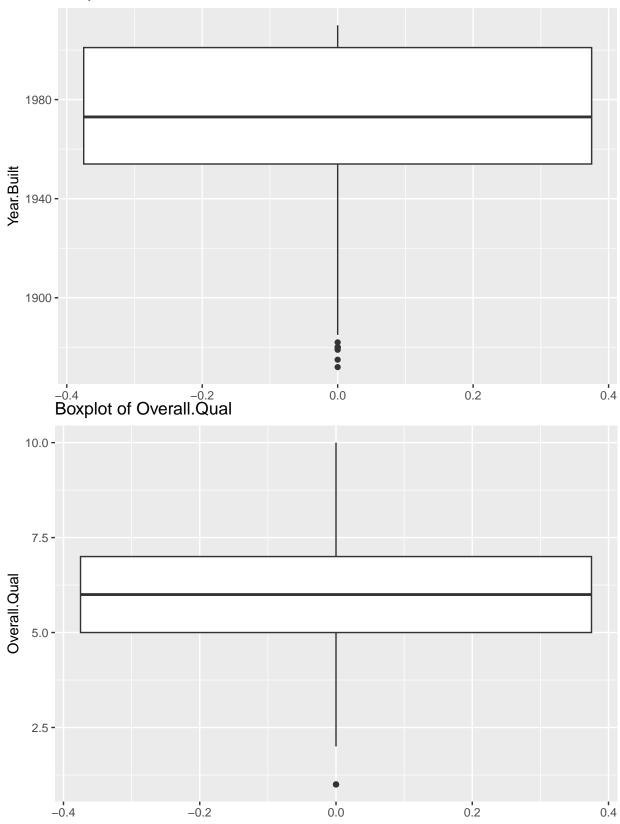




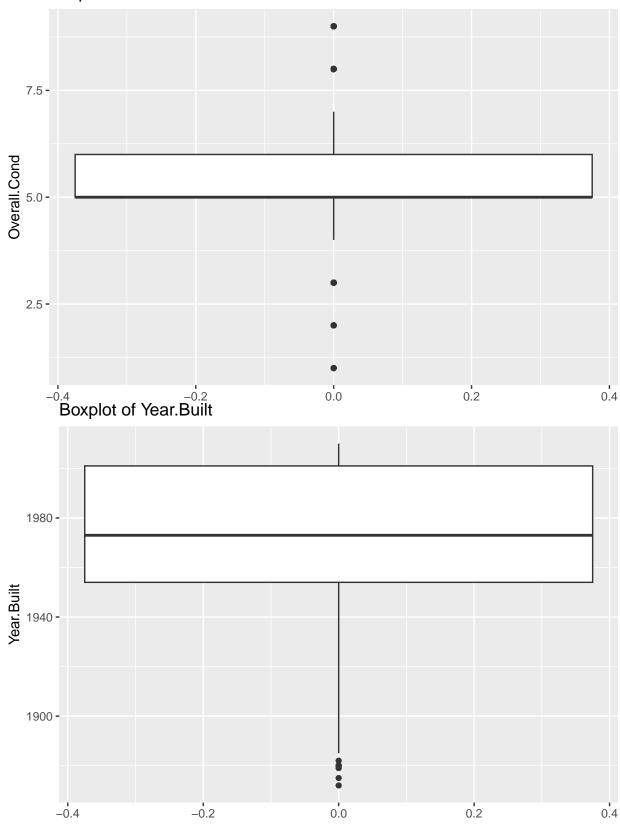


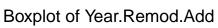


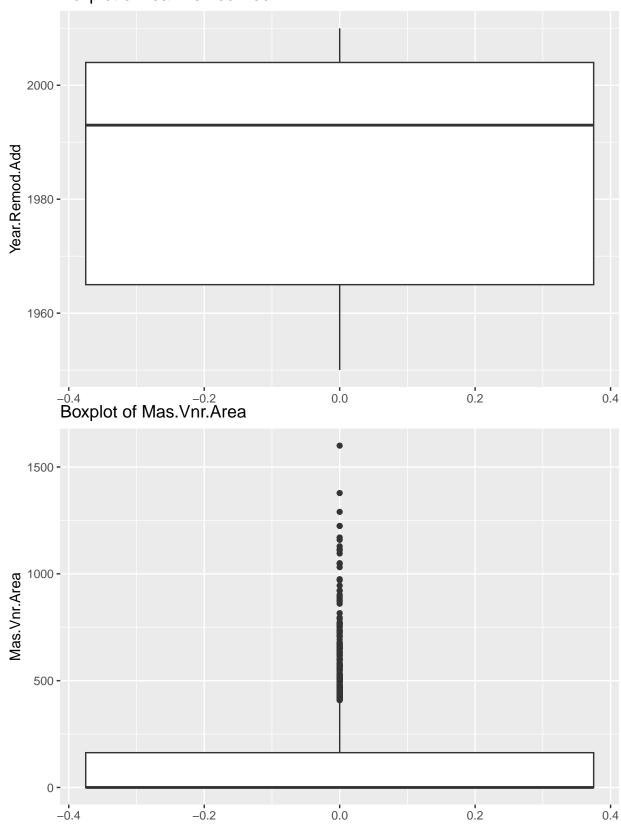




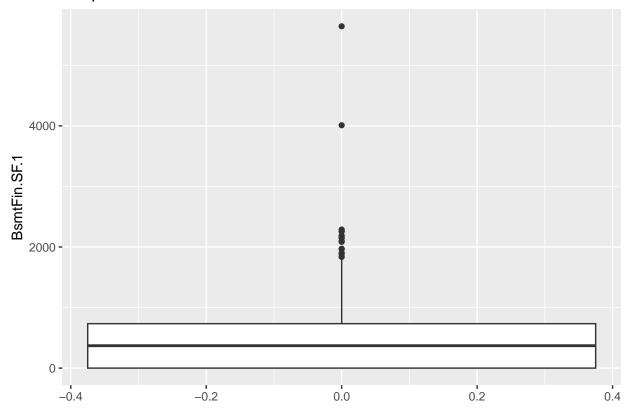






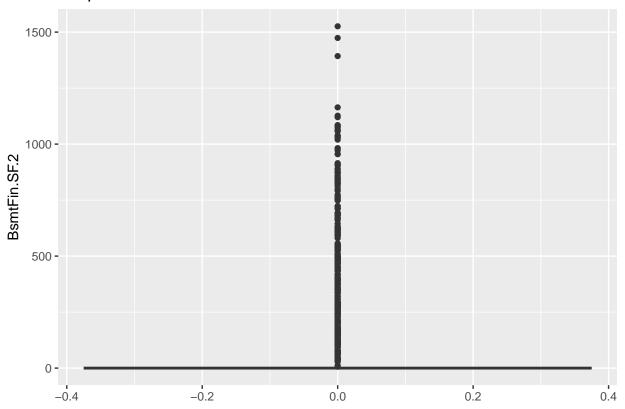


Boxplot of BsmtFin.SF.1



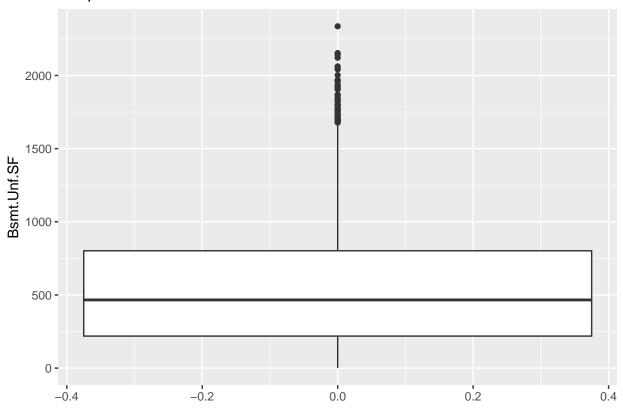
Warning: Removed 1 rows containing non-finite values (`stat_boxplot()`).

Boxplot of BsmtFin.SF.2

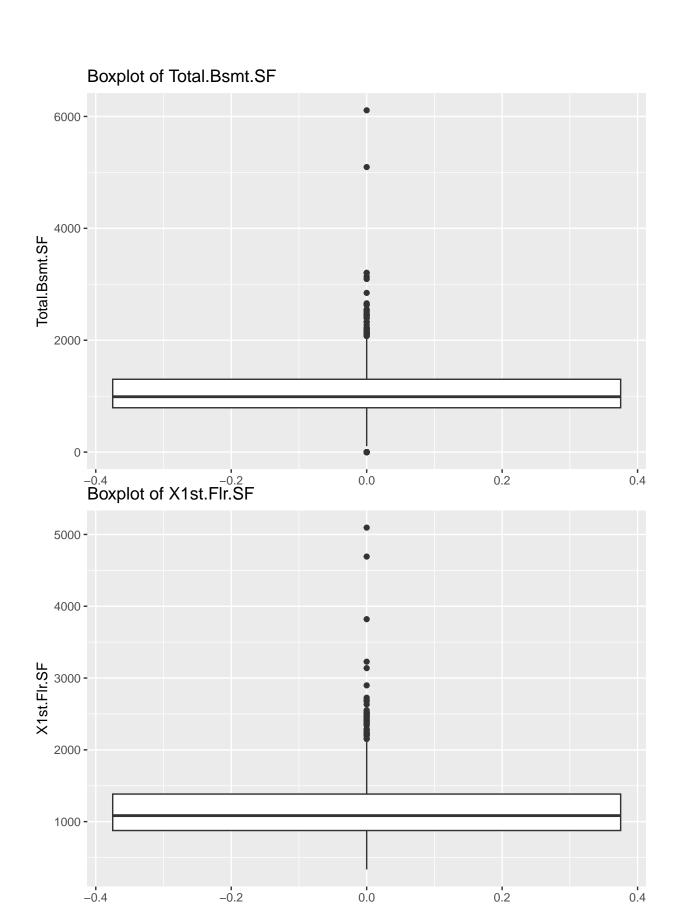


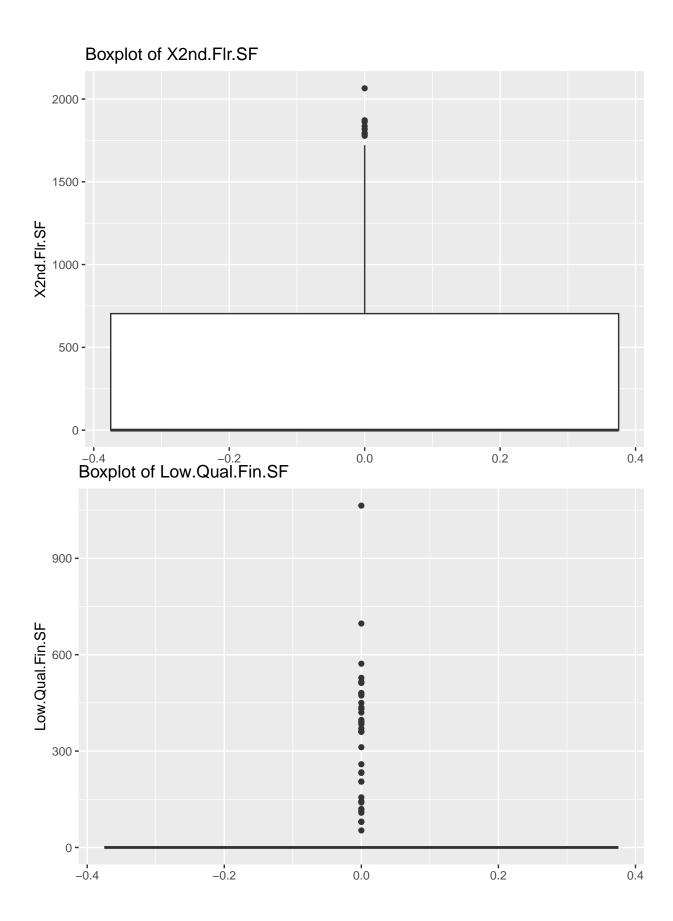
Warning: Removed 1 rows containing non-finite values (`stat_boxplot()`).

Boxplot of Bsmt.Unf.SF

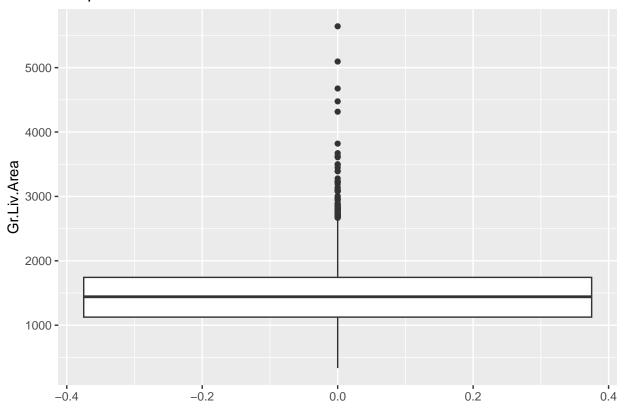


Warning: Removed 1 rows containing non-finite values (`stat_boxplot()`).



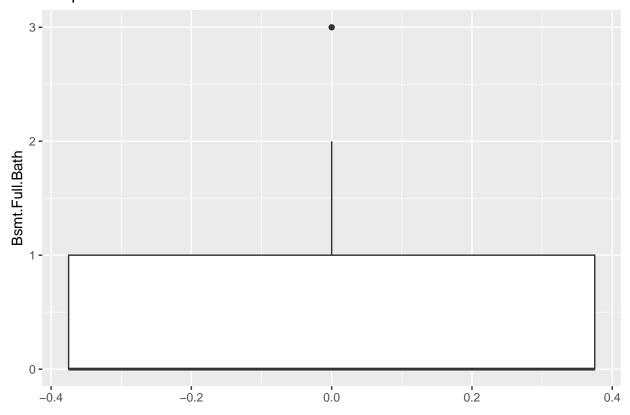


Boxplot of Gr.Liv.Area



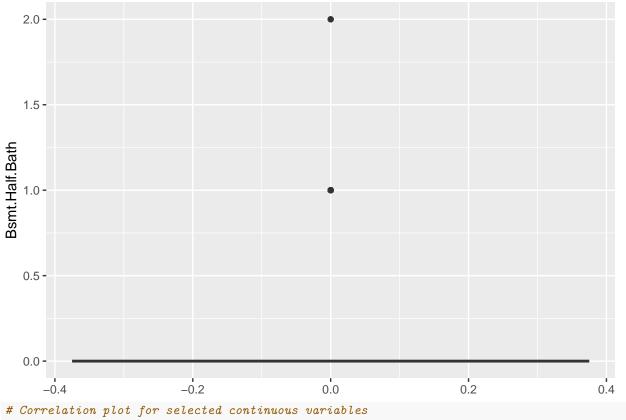
Warning: Removed 2 rows containing non-finite values (`stat_boxplot()`).

Boxplot of Bsmt.Full.Bath



Warning: Removed 2 rows containing non-finite values (`stat_boxplot()`).

Boxplot of Bsmt.Half.Bath



Correlation plot for selected continuous variables
continuous_data <- housing[, selected_vars]
corrplot(cor(continuous_data, use = "complete.obs"), method = "circle")</pre>

```
SalePrice
    Lot.Frontage
                                                                 0.8
        Lot.Area
                                                        Overall.Qual
                                                                 0.6
        Year.Built
                                                        Overall.Qual.1
                                                                 0.4
    Overall.Cond
      Year.Built.1
                                                        0.2
Year.Remod.Add
                                                        •
   Mas.Vnr.Area
                                                        •
                                                                  0
    BsmtFin.SF.1
                   BsmtFin.SF.2
                                                                  -0.2
    Bsmt.Unf.SF
                                                        Total.Bsmt.SF
                                                                  -0.4
      X1st.Flr.SF
     X2nd.Flr.SF
                                                                  -0.6
Low.Qual.Fin.SF
      Gr.Liv.Area
                                                                  -0.8
  Bsmt.Full.Bath
  Bsmt.Half.Bath
#Requirement 2: Regression Model and Diagnostics
# Split Data into Training and Testing Sets
set.seed(2023)
test_indices <- sample(nrow(housing), round(nrow(housing)/4))</pre>
train_set <- housing[-test_indices, ]</pre>
test_set <- housing[test_indices, ]</pre>
# Check for Highly Correlated Predictors
corr_matrix <- cor(train_set[, selected_vars], use = "complete.obs")</pre>
high_corr <- findCorrelation(corr_matrix, cutoff = 0.75)
high_corr_vars <- names(train_set[, selected_vars])[high_corr]</pre>
high_corr_vars
## [1] "SalePrice"
                       "Overall.Qual" "Total.Bsmt.SF" "Year.Built"
selected_high_corr_vars <- setdiff(selected_vars, c("SalePrice", "Overall.Qual", "Year.Built", "Total.B</pre>
# Fit Regression Model and Perform Diagnostics
lm_model <- lm(SalePrice ~ ., data = train_set[, selected_vars])</pre>
summary(lm_model)
##
## Call:
## lm(formula = SalePrice ~ ., data = train_set[, selected_vars])
##
## Residuals:
```

Max

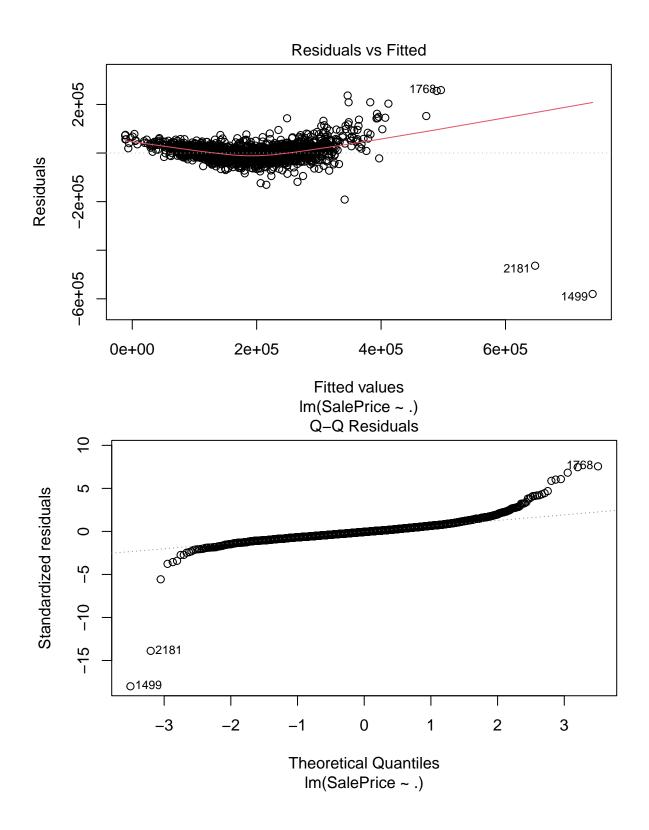
##

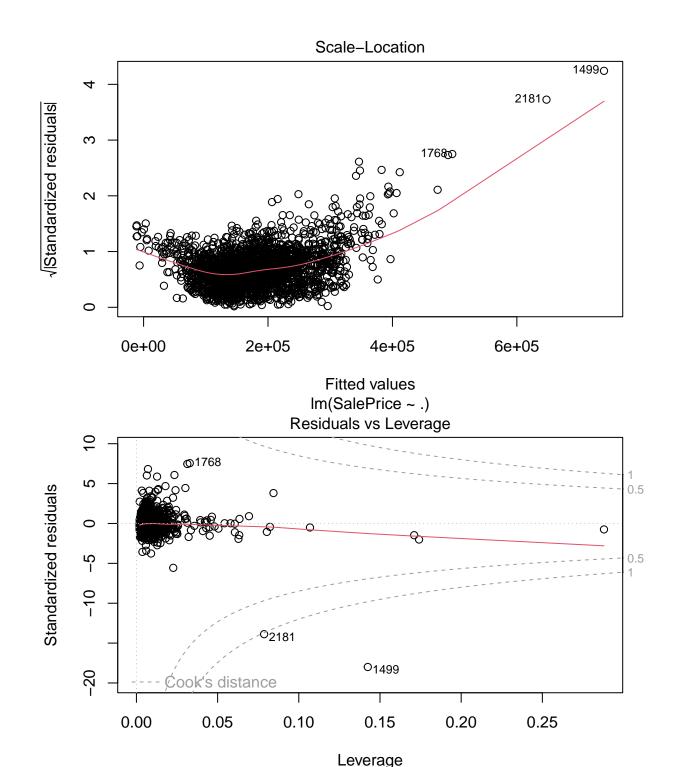
Min

1Q Median

3Q

```
## -580068 -16912 -2103
                            13834 258855
##
## Coefficients: (4 not defined because of singularities)
                    Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                   -1.282e+06 9.001e+04 -14.238 < 2e-16 ***
## Lot.Frontage
                   1.064e+02 3.959e+01
                                          2.688 0.007232 **
## Lot.Area
                   6.196e-01 9.524e-02
                                          6.506 9.55e-11 ***
## Overall.Qual
                   2.101e+04
                              8.669e+02
                                         24.233 < 2e-16 ***
## Year.Built
                   3.851e+02
                              4.039e+01
                                          9.534 < 2e-16 ***
## Overall.Qual.1
                           NA
                                     NA
                                              NA
                                                       NA
## Overall.Cond
                   4.685e+03
                              7.951e+02
                                          5.893 4.38e-09 ***
## Year.Built.1
                                             NA
                          NA
                                     NA
                                                       NA
## Year.Remod.Add
                   2.136e+02
                             5.168e+01
                                          4.132 3.72e-05 ***
## Mas.Vnr.Area
                                          6.029 1.93e-09 ***
                   2.862e+01 4.747e+00
## BsmtFin.SF.1
                   2.403e+01 3.618e+00
                                          6.643 3.87e-11 ***
## BsmtFin.SF.2
                   1.613e+01
                              5.513e+00
                                          2.925 0.003478 **
## Bsmt.Unf.SF
                   1.058e+01
                             3.216e+00
                                          3.289 0.001022 **
## Total.Bsmt.SF
                          NA
                                     NA
                                              NA
                                                       NA
## X1st.Flr.SF
                   6.077e+01
                              3.619e+00
                                         16.791
                                                 < 2e-16 ***
## X2nd.Flr.SF
                   5.246e+01
                              2.137e+00
                                         24.553 < 2e-16 ***
## Low.Qual.Fin.SF
                                           1.667 0.095749 .
                   2.733e+01
                              1.640e+01
## Gr.Liv.Area
                          NA
                                              NA
## Bsmt.Full.Bath
                   7.368e+03 2.056e+03
                                          3.583 0.000347 ***
## Bsmt.Half.Bath -4.544e+03 3.355e+03 -1.354 0.175735
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 34810 on 2180 degrees of freedom
     (2 observations deleted due to missingness)
## Multiple R-squared: 0.8062, Adjusted R-squared: 0.8048
## F-statistic: 604.5 on 15 and 2180 DF, p-value: < 2.2e-16
# Diagnostics: Check for assumptions, outliers, and multicollinearity
plot(lm_model)
```





#Requirement 3: Model Fitting and Comparison
Assuming 'selected_vars' contains the names of your predictors and excludes 'SalePrice'
Ensure the training set does not have missing values in the selected variables
train_set_clean <- na.omit(train_set[, c(selected_vars, "SalePrice")])
Create the model matrix for predictors</pre>

Im(SalePrice ~ .)

```
x <- model.matrix(~ ., data = train_set_clean[, selected_vars])</pre>
# Create the response vector
y <- train_set_clean$SalePrice
# Check if the number of rows in x matches the length of y
nrow(x) == length(y)
## [1] TRUE
# Fit AIC model
aic_model <- step(lm(SalePrice ~ ., data = train_set[, selected_vars]), direction = "both", criterion =
## Start: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Qual.1 + Overall.Cond + Year.Built.1 + Year.Remod.Add +
       Mas.Vnr.Area + BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF +
##
##
       Total.Bsmt.SF + X1st.Flr.SF + X2nd.Flr.SF + Low.Qual.Fin.SF +
##
       Gr.Liv.Area + Bsmt.Full.Bath + Bsmt.Half.Bath
##
##
## Step: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Qual.1 + Overall.Cond + Year.Built.1 + Year.Remod.Add +
##
##
       Mas.Vnr.Area + BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF +
       Total.Bsmt.SF + X1st.Flr.SF + X2nd.Flr.SF + Low.Qual.Fin.SF +
##
       Bsmt.Full.Bath + Bsmt.Half.Bath
##
##
##
## Step: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
##
       Overall.Qual.1 + Overall.Cond + Year.Built.1 + Year.Remod.Add +
##
       Mas.Vnr.Area + BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF +
##
       X1st.Flr.SF + X2nd.Flr.SF + Low.Qual.Fin.SF + Bsmt.Full.Bath +
##
       Bsmt.Half.Bath
##
##
## Step: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Qual.1 + Overall.Cond + Year.Remod.Add + Mas.Vnr.Area +
##
       BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF +
       X2nd.Flr.SF + Low.Qual.Fin.SF + Bsmt.Full.Bath + Bsmt.Half.Bath
##
##
##
## Step: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Cond + Year.Remod.Add + Mas.Vnr.Area + BsmtFin.SF.1 +
##
       BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF + X2nd.Flr.SF +
##
       Low.Qual.Fin.SF + Bsmt.Full.Bath + Bsmt.Half.Bath
##
##
                     Df Sum of Sq
                     1 2.2226e+09 2.6434e+12 45945
## - Bsmt.Half.Bath
## <none>
                                   2.6412e+12 45946
## - Low.Qual.Fin.SF 1 3.3649e+09 2.6445e+12 45946
```

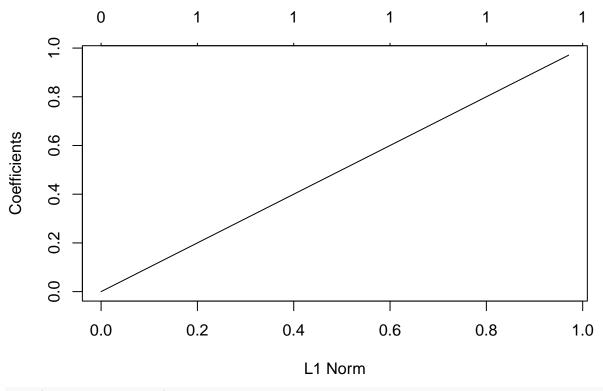
```
## - Lot.Frontage
                      1 8.7570e+09 2.6499e+12 45951
## - BsmtFin.SF.2
                      1 1.0367e+10 2.6515e+12 45952
## - Bsmt.Unf.SF
                      1 1.3105e+10 2.6543e+12 45955
## - Bsmt.Full.Bath
                      1 1.5556e+10 2.6567e+12 45957
## - Year.Remod.Add
                      1 2.0690e+10 2.6619e+12 45961
## - Overall.Cond
                      1 4.2075e+10 2.6832e+12 45978
## - Mas.Vnr.Area
                      1 4.4035e+10 2.6852e+12 45980
## - Lot.Area
                      1 5.1280e+10 2.6924e+12 45986
## - BsmtFin.SF.1
                      1 5.3463e+10 2.6946e+12 45988
## - Year.Built
                      1 1.1014e+11 2.7513e+12 46033
## - X1st.Flr.SF
                      1 3.4160e+11 2.9828e+12 46211
## - Overall.Qual
                      1 7.1146e+11 3.3526e+12 46467
## - X2nd.Flr.SF
                      1 7.3035e+11 3.3715e+12 46480
##
## Step: AIC=45945.48
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Cond + Year.Remod.Add + Mas.Vnr.Area + BsmtFin.SF.1 +
##
##
       BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF + X2nd.Flr.SF +
##
      Low.Qual.Fin.SF + Bsmt.Full.Bath
##
                                          RSS
##
                     Df Sum of Sq
                                                ATC:
                                   2.6434e+12 45945
## <none>
## + Bsmt.Half.Bath
                      1 2.2226e+09 2.6412e+12 45946
## - Low.Qual.Fin.SF 1 3.3661e+09 2.6468e+12 45946
## - BsmtFin.SF.2
                      1 9.0867e+09 2.6525e+12 45951
## - Lot.Frontage
                      1 9.1020e+09 2.6525e+12 45951
## - Bsmt.Unf.SF
                      1 1.3132e+10 2.6565e+12 45954
## - Year.Remod.Add
                      1 2.0769e+10 2.6642e+12 45961
## - Bsmt.Full.Bath
                    1 2.1068e+10 2.6645e+12 45961
## - Overall.Cond
                      1 4.0931e+10 2.6843e+12 45977
## - Mas.Vnr.Area
                      1 4.3803e+10 2.6872e+12 45980
## - Lot.Area
                      1 5.0522e+10 2.6939e+12 45985
## - BsmtFin.SF.1
                      1 5.1292e+10 2.6947e+12 45986
## - Year.Built
                      1 1.0959e+11 2.7530e+12 46033
## - X1st.Flr.SF
                      1 3.4457e+11 2.9880e+12 46213
## - Overall.Qual
                      1 7.1269e+11 3.3561e+12 46468
## - X2nd.Flr.SF
                      1 7.3397e+11 3.3773e+12 46482
# Fit BIC model
bic_model <- step(lm(SalePrice ~ ., data = train_set[, selected_vars]), direction = "both", criterion =</pre>
## Start: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
##
       Overall.Qual.1 + Overall.Cond + Year.Built.1 + Year.Remod.Add +
##
       Mas.Vnr.Area + BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF +
##
       Total.Bsmt.SF + X1st.Flr.SF + X2nd.Flr.SF + Low.Qual.Fin.SF +
##
       Gr.Liv.Area + Bsmt.Full.Bath + Bsmt.Half.Bath
##
##
## Step: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Qual.1 + Overall.Cond + Year.Built.1 + Year.Remod.Add +
      Mas.Vnr.Area + BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF +
##
##
      Total.Bsmt.SF + X1st.Flr.SF + X2nd.Flr.SF + Low.Qual.Fin.SF +
##
      Bsmt.Full.Bath + Bsmt.Half.Bath
```

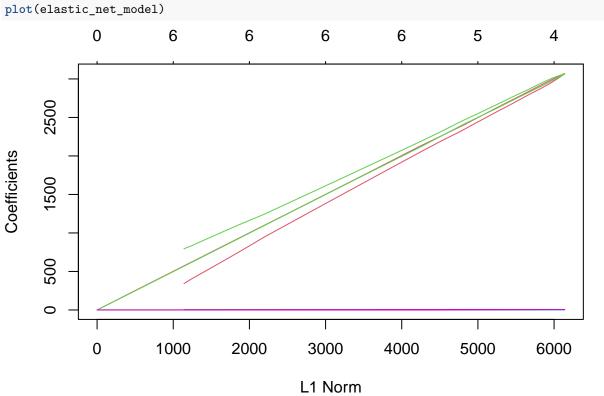
```
##
##
## Step: AIC=45945.63
  SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Qual.1 + Overall.Cond + Year.Built.1 + Year.Remod.Add +
##
       Mas.Vnr.Area + BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF +
       X1st.Flr.SF + X2nd.Flr.SF + Low.Qual.Fin.SF + Bsmt.Full.Bath +
##
##
       Bsmt.Half.Bath
##
##
## Step: AIC=45945.63
  SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
##
       Overall.Qual.1 + Overall.Cond + Year.Remod.Add + Mas.Vnr.Area +
       BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF +
##
##
       X2nd.Flr.SF + Low.Qual.Fin.SF + Bsmt.Full.Bath + Bsmt.Half.Bath
##
##
## Step: AIC=45945.63
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
       Overall.Cond + Year.Remod.Add + Mas.Vnr.Area + BsmtFin.SF.1 +
##
       BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF + X2nd.Flr.SF +
##
       Low.Qual.Fin.SF + Bsmt.Full.Bath + Bsmt.Half.Bath
##
                     Df Sum of Sq
                                          RSS
## - Bsmt.Half.Bath
                      1 2.2226e+09 2.6434e+12 45945
## <none>
                                   2.6412e+12 45946
## - Low.Qual.Fin.SF
                     1 3.3649e+09 2.6445e+12 45946
## - Lot.Frontage
                      1 8.7570e+09 2.6499e+12 45951
## - BsmtFin.SF.2
                      1 1.0367e+10 2.6515e+12 45952
## - Bsmt.Unf.SF
                      1 1.3105e+10 2.6543e+12 45955
## - Bsmt.Full.Bath
                      1 1.5556e+10 2.6567e+12 45957
## - Year.Remod.Add
                      1 2.0690e+10 2.6619e+12 45961
## - Overall.Cond
                      1 4.2075e+10 2.6832e+12 45978
## - Mas.Vnr.Area
                      1 4.4035e+10 2.6852e+12 45980
## - Lot.Area
                      1 5.1280e+10 2.6924e+12 45986
## - BsmtFin.SF.1
                      1 5.3463e+10 2.6946e+12 45988
## - Year.Built
                      1 1.1014e+11 2.7513e+12 46033
## - X1st.Flr.SF
                      1 3.4160e+11 2.9828e+12 46211
## - Overall.Qual
                      1 7.1146e+11 3.3526e+12 46467
## - X2nd.Flr.SF
                      1 7.3035e+11 3.3715e+12 46480
##
## Step: AIC=45945.48
## SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual + Year.Built +
##
       Overall.Cond + Year.Remod.Add + Mas.Vnr.Area + BsmtFin.SF.1 +
       BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF + X2nd.Flr.SF +
##
       Low.Qual.Fin.SF + Bsmt.Full.Bath
##
##
##
                         Sum of Sq
                                          RSS
                                                 AIC
## <none>
                                   2.6434e+12 45945
## + Bsmt.Half.Bath
                      1 2.2226e+09 2.6412e+12 45946
## - Low.Qual.Fin.SF
                     1 3.3661e+09 2.6468e+12 45946
## - BsmtFin.SF.2
                      1 9.0867e+09 2.6525e+12 45951
## - Lot.Frontage
                      1 9.1020e+09 2.6525e+12 45951
## - Bsmt.Unf.SF
                      1 1.3132e+10 2.6565e+12 45954
```

```
## - Year.Remod.Add
                      1 2.0769e+10 2.6642e+12 45961
## - Bsmt.Full.Bath
                      1 2.1068e+10 2.6645e+12 45961
## - Overall.Cond
                      1 4.0931e+10 2.6843e+12 45977
## - Mas.Vnr.Area
                      1 4.3803e+10 2.6872e+12 45980
## - Lot.Area
                      1 5.0522e+10 2.6939e+12 45985
## - BsmtFin.SF.1
                      1 5.1292e+10 2.6947e+12 45986
## - Year.Built
                      1 1.0959e+11 2.7530e+12 46033
## - X1st.Flr.SF
                      1 3.4457e+11 2.9880e+12 46213
## - Overall.Qual
                      1 7.1269e+11 3.3561e+12 46468
## - X2nd.Flr.SF
                      1 7.3397e+11 3.3773e+12 46482
# Fit Lasso model
lasso_model <- glmnet(x, y, alpha = 1)</pre>
# Fit Elastic Net model
elastic_net_model <- glmnet(x, y, alpha = 0.5)</pre>
# Summarize AIC and BIC models
summary(aic_model)
##
## Call:
## lm(formula = SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual +
##
       Year.Built + Overall.Cond + Year.Remod.Add + Mas.Vnr.Area +
##
       BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF +
##
       X2nd.Flr.SF + Low.Qual.Fin.SF + Bsmt.Full.Bath, data = train_set[,
##
       selected_vars])
##
## Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
## -577788 -17002
                     -2041
                             13882
                                    256052
##
## Coefficients:
##
                     Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                   -1.281e+06 9.002e+04 -14.224 < 2e-16 ***
## Lot.Frontage
                    1.084e+02 3.957e+01
                                           2.740 0.00619 **
## Lot.Area
                    6.146e-01 9.519e-02
                                           6.456 1.32e-10 ***
                    2.102e+04 8.670e+02 24.249 < 2e-16 ***
## Overall.Qual
## Year.Built
                    3.841e+02 4.039e+01
                                           9.509 < 2e-16 ***
## Overall.Cond
                    4.610e+03 7.932e+02
                                           5.811 7.11e-09 ***
## Year.Remod.Add
                                           4.140 3.61e-05 ***
                    2.140e+02 5.169e+01
## Mas.Vnr.Area
                    2.854e+01 4.748e+00
                                           6.012 2.15e-09 ***
## BsmtFin.SF.1
                    2.318e+01 3.564e+00
                                           6.505 9.58e-11 ***
## BsmtFin.SF.2
                    1.489e+01 5.438e+00
                                           2.738 0.00623 **
## Bsmt.Unf.SF
                                           3.292 0.00101 **
                    1.059e+01 3.217e+00
## X1st.Flr.SF
                    6.098e+01 3.616e+00
                                          16.861
                                                  < 2e-16 ***
## X2nd.Flr.SF
                    5.256e+01 2.136e+00
                                          24.609 < 2e-16 ***
## Low.Qual.Fin.SF
                    2.734e+01
                              1.640e+01
                                           1.667 0.09575 .
                                           4.169 3.18e-05 ***
## Bsmt.Full.Bath
                    8.192e+03 1.965e+03
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 34810 on 2181 degrees of freedom
     (2 observations deleted due to missingness)
## Multiple R-squared: 0.806, Adjusted R-squared: 0.8048
```

```
## F-statistic: 647.3 on 14 and 2181 DF, p-value: < 2.2e-16
summary(bic model)
##
## Call:
## lm(formula = SalePrice ~ Lot.Frontage + Lot.Area + Overall.Qual +
##
       Year.Built + Overall.Cond + Year.Remod.Add + Mas.Vnr.Area +
       BsmtFin.SF.1 + BsmtFin.SF.2 + Bsmt.Unf.SF + X1st.Flr.SF +
##
##
      X2nd.Flr.SF + Low.Qual.Fin.SF + Bsmt.Full.Bath, data = train_set[,
##
       selected_vars])
##
## Residuals:
##
      Min
                1Q
                                3Q
                                      Max
                   Median
## -577788 -17002
                    -2041
                             13882
                                   256052
##
## Coefficients:
##
                    Estimate Std. Error t value Pr(>|t|)
                   -1.281e+06 9.002e+04 -14.224 < 2e-16 ***
## (Intercept)
## Lot.Frontage
                   1.084e+02 3.957e+01
                                          2.740 0.00619 **
## Lot.Area
                   6.146e-01 9.519e-02
                                          6.456 1.32e-10 ***
                   2.102e+04 8.670e+02 24.249 < 2e-16 ***
## Overall.Qual
## Year.Built
                   3.841e+02 4.039e+01
                                          9.509 < 2e-16 ***
## Overall.Cond
                                          5.811 7.11e-09 ***
                   4.610e+03 7.932e+02
## Year.Remod.Add
                   2.140e+02 5.169e+01
                                          4.140 3.61e-05 ***
## Mas.Vnr.Area
                   2.854e+01 4.748e+00
                                          6.012 2.15e-09 ***
## BsmtFin.SF.1
                   2.318e+01 3.564e+00
                                          6.505 9.58e-11 ***
## BsmtFin.SF.2
                   1.489e+01 5.438e+00
                                          2.738 0.00623 **
## Bsmt.Unf.SF
                                          3.292 0.00101 **
                   1.059e+01 3.217e+00
## X1st.Flr.SF
                   6.098e+01 3.616e+00
                                         16.861
                                                 < 2e-16 ***
## X2nd.Flr.SF
                                         24.609 < 2e-16 ***
                   5.256e+01 2.136e+00
## Low.Qual.Fin.SF
                  2.734e+01
                              1.640e+01
                                          1.667 0.09575 .
                   8.192e+03 1.965e+03
                                          4.169 3.18e-05 ***
## Bsmt.Full.Bath
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 34810 on 2181 degrees of freedom
     (2 observations deleted due to missingness)
## Multiple R-squared: 0.806, Adjusted R-squared: 0.8048
## F-statistic: 647.3 on 14 and 2181 DF, p-value: < 2.2e-16
# Plot Lasso and Elastic Net models
plot(lasso_model)
## Warning in plotCoef(x$beta, lambda = x$lambda, df = x$df, dev = x$dev.ratio, :
```

1 or less nonzero coefficients; glmnet plot is not meaningful





```
#Requirement 4: Calculate Mean Prediction Error
# Prepare Testing Data
x_test <- model.matrix(~ ., data = test_set[, selected_vars])
y_test <- test_set$SalePrice</pre>
```

```
# Calculate Mean Prediction Error
predict_error <- function(model, x, y) {</pre>
 predictions <- predict(model, newx = x)</pre>
  mean((predictions - y)^2)
# Calculate mean prediction error for each model
mean_error_aic <- predict_error(aic_model, x_test, y_test)</pre>
mean_error_bic <- predict_error(bic_model, x_test, y_test)</pre>
mean_error_lasso <- predict_error(lasso_model, x_test, y_test)</pre>
mean_error_elastic_net <- predict_error(elastic_net_model, x_test, y_test)</pre>
# Output the errors
c(AIC = mean_error_aic, BIC = mean_error_bic, Lasso = mean_error_lasso, ElasticNet = mean_error_elastic
##
           AIC
                        BIC
                                  Lasso ElasticNet
## 11536668354 11536668354 1040106431 1272404579
```