

stepwise regression process to find best var log

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Data source and original descriptions

Data preparation

```
# Check if the caret package is installed
if (!requireNamespace("caret", quietly = TRUE)) {
  # If not installed, install it
  install.packages("caret")
}

# Load the caret package
library(caret)

## 载入需要的程辑包: ggplot2

## 载入需要的程辑包: lattice

# Load necessary libraries
library(tidyverse)

## — Attaching core tidyverse packages ————— tidyverse
rse 2.0.0 —
## ✓ dplyr      1.1.1      ✓ readr      2.1.5
## ✓ forcats   1.0.0      ✓ stringr    1.5.0
## ✓ lubridate 1.9.2      ✓ tibble     3.2.1
## ✓ purrr     1.0.1      ✓ tidyr      1.3.0

## — Conflicts ————— tidyverse_co
nflicts() —
## ✗ dplyr::filter() masks stats::filter()
## ✗ dplyr::lag()     masks stats::lag()
## ✗ purrr::lift()   masks caret::lift()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to for
ce all conflicts to become errors

# Load the data
data <- read_csv(file.choose()) # open file location

## Warning: One or more parsing issues, call `problems()` on your data
frame for details,
```

```

## e.g.:
## dat <- vroom(...)
## problems(dat)

## Rows: 34857 Columns: 21
## — Column specification —————
|-----|
## Delimiter: ","
## chr (8): Suburb, Address, Type, Method, SellerG, Date, CouncilArea,
Regionname
## dbl (13): Rooms, Price, Distance, Postcode, Bedroom2, Bathroom, Car,
Landsiz...
##
## i Use `spec()` to retrieve the full column specification for this da
ta.
## i Specify the column types or set `show_col_types = FALSE` to quiet
this message.

# Correct column names
names(data)[names(data) == "Lattitude"] <- "Latitude"
names(data)[names(data) == "Longtitude"] <- "Longitude"

# Remove unnecessary columns using dplyr's select function
data_clean <- data %>%
  dplyr::select(Suburb, Rooms, Type, Price, Distance, Bedroom2, Bathroo
m, Car, Landsize, BuildingArea, YearBuilt, CouncilArea, Latitude, Longi
tude, Propertycount, Date)

# Convert 'Date' to date type
data_clean$Date <- as.Date(data_clean$Date, format = "%d/%m/%Y")

# Calculate 'YearsAfterBuilt'
data_clean$YearsAfterBuilt <- as.numeric(format(data_clean$Date, "%Y"))
- data_clean$YearBuilt

# Calculate LogPricePerBuildingArea
data_clean$LogPricePerBuildingArea <- log(data_clean$Price / data_clean
$BuildingArea)

# Drop the "Price", "Longitude", "Latitude", "YearBuilt" and columns fr
om the dataset
# due to persistent issue of factor mis matching between training and t
esting, and since the reference models do not contain Suburb variable,
clean at the beginning
# if Date is not deleted, the result would include it
data_clean <- subset(data_clean, select = -c(Price, Longitude, Latitude
, YearBuilt, Suburb, Date))

# Remove rows with missing values
data_clean <- na.omit(data_clean)

```

```

# Convert categorical variables to factors
cat_vars <- c("Type", "CouncilArea") # Add categorical variables here
data_clean[cat_vars] <- lapply(data_clean[cat_vars], as.factor)

# Convert non-categorical variables to numeric
non_cat_vars <- setdiff(names(data_clean), c(cat_vars, "LogPricePerBuildingArea"))
data_clean[non_cat_vars] <- lapply(data_clean[non_cat_vars], as.numeric)

# Standardize non-categorical variables
data_clean[non_cat_vars] <- scale(data_clean[non_cat_vars])

# Separate predictors and target variable
predictors <- setdiff(names(data_clean), "LogPricePerBuildingArea")

# Split data into training and testing sets
set.seed(123)
indexes <- createDataPartition(data_clean$LogPricePerBuildingArea, p = 0.8, list = FALSE)
train_data <- data_clean[indexes, ]
test_data <- data_clean[-indexes, ]

problems(data_clean)

```

Model training and AIC process

```

# Remove rows with NA, NaN, or Inf values in the target variable
train_data <- train_data[!is.na(train_data$LogPricePerBuildingArea) & !
is.nan(train_data$LogPricePerBuildingArea) & !is.infinite(train_data$LogPricePerBuildingArea), ]

# Train stepwise regression model
model <- step(lm(LogPricePerBuildingArea ~ ., data = train_data[, c(predictors, "LogPricePerBuildingArea")]), direction = "backward")

## Start:  AIC=-15035.86
## LogPricePerBuildingArea ~ Rooms + Type + Distance + Bedroom2 +
## Bathroom + Car + Landsize + BuildingArea + CouncilArea +
## Propertycount + YearsAfterBuilt
##
##           Df Sum of Sq    RSS    AIC
## - Rooms      1     0.003  837.70 -15038
## - Propertycount  1     0.007  837.71 -15038
## <none>                837.70 -15036
## - Bedroom2      1     0.408  838.11 -15034
## - Landsize      1     0.803  838.50 -15031

```

```

## - Car          1      3.543  841.24 -15008
## - Bathroom     1      4.502  842.20 -15000
## - YearsAfterBuilt 1     28.311  866.01 -14802
## - Type         2     42.633  880.33 -14688
## - Distance     1     43.216  880.92 -14682
## - BuildingArea 1    248.853 1086.55 -13195
## - CouncilArea  32    293.991 1131.69 -12969
##
## Step: AIC=-15037.84
## LogPricePerBuildingArea ~ Type + Distance + Bedroom2 + Bathroom +
##   Car + Landsize + BuildingArea + CouncilArea + Propertycount +
##   YearsAfterBuilt
##
##           Df Sum of Sq    RSS    AIC
## - Propertycount  1      0.007  837.71 -15040
## <none>                                837.70 -15038
## - Landsize      1      0.804  838.51 -15033
## - Bedroom2     1      2.751  840.46 -15017
## - Car          1      3.551  841.26 -15010
## - Bathroom     1      4.564  842.27 -15001
## - YearsAfterBuilt 1     28.428  866.13 -14803
## - Type         2     43.024  880.73 -14687
## - Distance     1     43.213  880.92 -14684
## - BuildingArea  1    251.368 1089.07 -13181
## - CouncilArea  32    294.003 1131.71 -12971
##
## Step: AIC=-15039.78
## LogPricePerBuildingArea ~ Type + Distance + Bedroom2 + Bathroom +
##   Car + Landsize + BuildingArea + CouncilArea + YearsAfterBuilt
##
##           Df Sum of Sq    RSS    AIC
## <none>                                837.71 -15040
## - Landsize      1      0.805  838.52 -15035
## - Bedroom2     1      2.756  840.47 -15018
## - Car          1      3.549  841.26 -15012
## - Bathroom     1      4.563  842.27 -15003
## - YearsAfterBuilt 1     28.429  866.14 -14805
## - Type         2     43.252  880.96 -14687
## - Distance     1     44.233  881.94 -14677
## - BuildingArea  1    251.447 1089.16 -13182
## - CouncilArea  32    294.140 1131.85 -12972

# Make predictions on test data
predictions <- predict(model, newdata = test_data)

# Evaluate the model
rmse <- sqrt(mean((predictions - test_data$LogPricePerBuildingArea)^2))
print(paste("RMSE: ", rmse))

## [1] "RMSE: Inf"

```

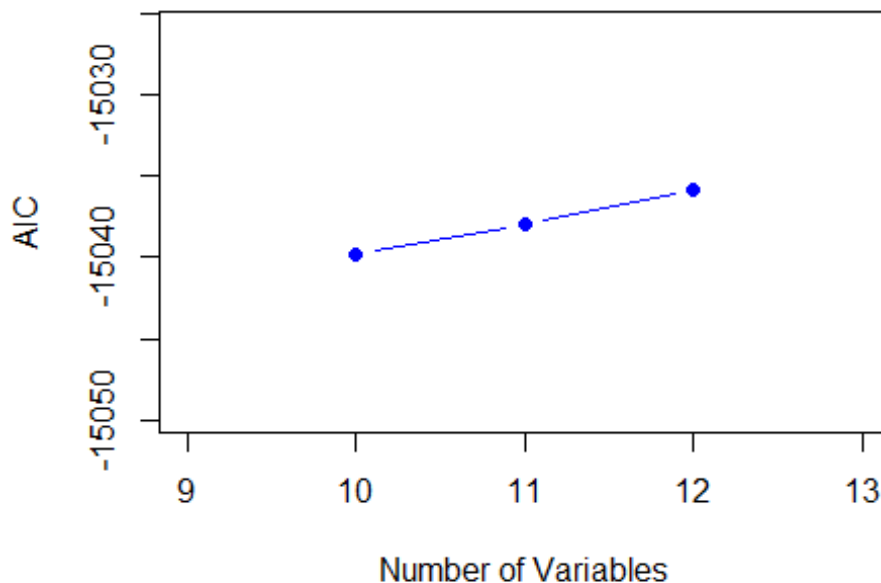
Displaying AIC value graph

```
# Update AIC values from the stepwise regression process
aic_values <- c(-15035.86, -15038, -15039.78)

# Number of variables in each step (including the intercept)
num_variables <- c(12, 11, 10) # Adjust this based on the actual steps
in model

# Plotting the updated AIC values
plot(num_variables, aic_values, type = "b",
     xlab = "Number of Variables",
     ylab = "AIC",
     main = "Stepwise Regression: AIC vs. Number of Variables",
     xlim = c(min(num_variables) - 1, max(num_variables) + 1),
     ylim = c(min(aic_values) - 10, max(aic_values) + 10),
     col = "blue",
     pch = 19)
```

Stepwise Regression: AIC vs. Number of Variable



final model summary

```
# Train the final model based on the selected predictors from the stepwise regression
final_model <- lm(LogPricePerBuildingArea ~ Type + Distance + Bedroom2
+ Bathroom + Car + Landsize + BuildingArea + CouncilArea + YearsAfterBuild, data = train_data)
```

```
# Print the summary of the final model
```

```
summary(final_model)
```

```
##
```

```
## Call:
```

```
## lm(formula = LogPricePerBuildingArea ~ Type + Distance + Bedroom2 +
```

```
## Bathroom + Car + Landsize + BuildingArea + CouncilArea +
```

```
## YearsAfterBuilt, data = train_data)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -2.4916 -0.1618 -0.0275  0.1183  5.4498
```

```
##
```

```
## Coefficients:
```

```
##                                     Estimate Std. Error t val
```

```
ue Pr(>|t|)
```

```
## (Intercept)                8.826337    0.019003 464.4
```

```
71 < 2e-16
```

```
## Typet                       -0.099754    0.017191  -5.8
```

```
03 6.82e-09
```

```
## Typeu                       -0.283054    0.014856 -19.0
```

```
53 < 2e-16
```

```
## Distance                    -0.199143    0.010327 -19.2
```

```
83 < 2e-16
```

```
## Bedroom2                    0.032268    0.006704   4.8
```

```
13 1.51e-06
```

```
## Bathroom                    0.036163    0.005839   6.1
```

```
94 6.21e-10
```

```
## Car                          0.025940    0.004749   5.4
```

```
62 4.87e-08
```

```
## Landsize                    0.010862    0.004175   2.6
```

```
02 0.009286
```

```
## BuildingArea               -0.246039    0.005352 -45.9
```

```
75 < 2e-16
```

```
## CouncilAreaBayside City Council 0.503836    0.027846  18.0
```

```
94 < 2e-16
```

```
## CouncilAreaBoroondara City Council 0.404661    0.024388  16.5
```

```
92 < 2e-16
```

```
## CouncilAreaBrimbank City Council -0.258649    0.027057  -9.5
```

```
59 < 2e-16
```

```
## CouncilAreaCardinia Shire Council 0.409020    0.116406   3.5
```

```
14 0.000445
```

```
## CouncilAreaCasey City Council  0.218291    0.072249   3.0
```

```
21 0.002525
```

```
## CouncilAreaDarebin City Council  0.024645    0.024373   1.0
```

```
11 0.311968
```

```
## CouncilAreaFrankston City Council 0.395573    0.058849   6.7
```

```
22 1.93e-11
```

```
## CouncilAreaGlen Eira City Council 0.246480    0.025588   9.6
```

33	< 2e-16			
##	CouncilAreaGreater Dandenong City Council	0.113331	0.058574	1.9
35	0.053050			
##	CouncilAreaHobsons Bay City Council	0.024723	0.029806	0.8
29	0.406859			
##	CouncilAreaHume City Council	-0.235406	0.029167	-8.0
71	8.13e-16			
##	CouncilAreaKingston City Council	0.267839	0.034269	7.8
16	6.26e-15			
##	CouncilAreaKnox City Council	0.177618	0.044424	3.9
98	6.45e-05			
##	CouncilAreaMacedon Ranges Shire Council	0.492425	0.132234	3.7
24	0.000198			
##	CouncilAreaManningham City Council	0.203025	0.030926	6.5
65	5.58e-11			
##	CouncilAreaMaribyrnong City Council	-0.068159	0.026653	-2.5
57	0.010572			
##	CouncilAreaMaroondah City Council	0.204202	0.045207	4.5
17	6.37e-06			
##	CouncilAreaMelbourne City Council	0.166427	0.029536	5.6
35	1.82e-08			
##	CouncilAreaMelton City Council	-0.261441	0.045606	-5.7
33	1.03e-08			
##	CouncilAreaMitchell Shire Council	0.301946	0.180071	1.6
77	0.093623			
##	CouncilAreaMonash City Council	0.297117	0.029246	10.1
59	< 2e-16			
##	CouncilAreaMoonee Valley City Council	0.023695	0.025656	0.9
24	0.355731			
##	CouncilAreaMoorabool Shire Council	-0.019687	0.347555	-0.0
57	0.954830			
##	CouncilAreaMoreland City Council	-0.061928	0.024916	-2.4
85	0.012961			
##	CouncilAreaNillumbik Shire Council	0.004139	0.076393	0.0
54	0.956791			
##	CouncilAreaPort Phillip City Council	0.301432	0.030442	9.9
02	< 2e-16			
##	CouncilAreaStonnington City Council	0.408323	0.031168	13.1
01	< 2e-16			
##	CouncilAreaWhitehorse City Council	0.312729	0.040145	7.7
90	7.67e-15			
##	CouncilAreaWhittlesea City Council	-0.153068	0.033574	-4.5
59	5.22e-06			
##	CouncilAreaWyndham City Council	-0.382601	0.036906	-10.3
67	< 2e-16			
##	CouncilAreaYarra City Council	0.149312	0.031848	4.6
88	2.81e-06			
##	CouncilAreaYarra Ranges Shire Council	0.220659	0.091819	2.4
03	0.016279			
##	YearsAfterBuilt	0.080845	0.005230	15.4

```

59 < 2e-16
##
## (Intercept) ***
## Typet ***
## Typeu ***
## Distance ***
## Bedroom2 ***
## Bathroom ***
## Car ***
## Landsize **
## BuildingArea ***
## CouncilAreaBayside City Council ***
## CouncilAreaBoroondara City Council ***
## CouncilAreaBrimbank City Council ***
## CouncilAreaCardinia Shire Council ***
## CouncilAreaCasey City Council **
## CouncilAreaDarebin City Council
## CouncilAreaFrankston City Council ***
## CouncilAreaGlen Eira City Council ***
## CouncilAreaGreater Dandenong City Council .
## CouncilAreaHobsons Bay City Council
## CouncilAreaHume City Council ***
## CouncilAreaKingston City Council ***
## CouncilAreaKnox City Council ***
## CouncilAreaMacedon Ranges Shire Council ***
## CouncilAreaManningham City Council ***
## CouncilAreaMaribyrnong City Council *
## CouncilAreaMaroondah City Council ***
## CouncilAreaMelbourne City Council ***
## CouncilAreaMelton City Council ***
## CouncilAreaMitchell Shire Council .
## CouncilAreaMonash City Council ***
## CouncilAreaMoonee Valley City Council
## CouncilAreaMoorabool Shire Council
## CouncilAreaMoreland City Council *
## CouncilAreaNillumbik Shire Council
## CouncilAreaPort Phillip City Council ***
## CouncilAreaStonnington City Council ***
## CouncilAreaWhitehorse City Council ***
## CouncilAreaWhittlesea City Council ***
## CouncilAreaWyndham City Council ***
## CouncilAreaYarra City Council ***
## CouncilAreaYarra Ranges Shire Council *
## YearsAfterBuilt ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3449 on 7042 degrees of freedom
## Multiple R-squared:  0.5565, Adjusted R-squared:  0.5539
## F-statistic: 215.5 on 41 and 7042 DF, p-value: < 2.2e-16

```


variable correlation check

```
if (!requireNamespace("GGally", quietly = TRUE)) {
  install.packages("GGally")
}

## Registered S3 method overwritten by 'GGally':
##   method from
##   +.gg   ggplot2

# Load the GGally package
library(GGally)

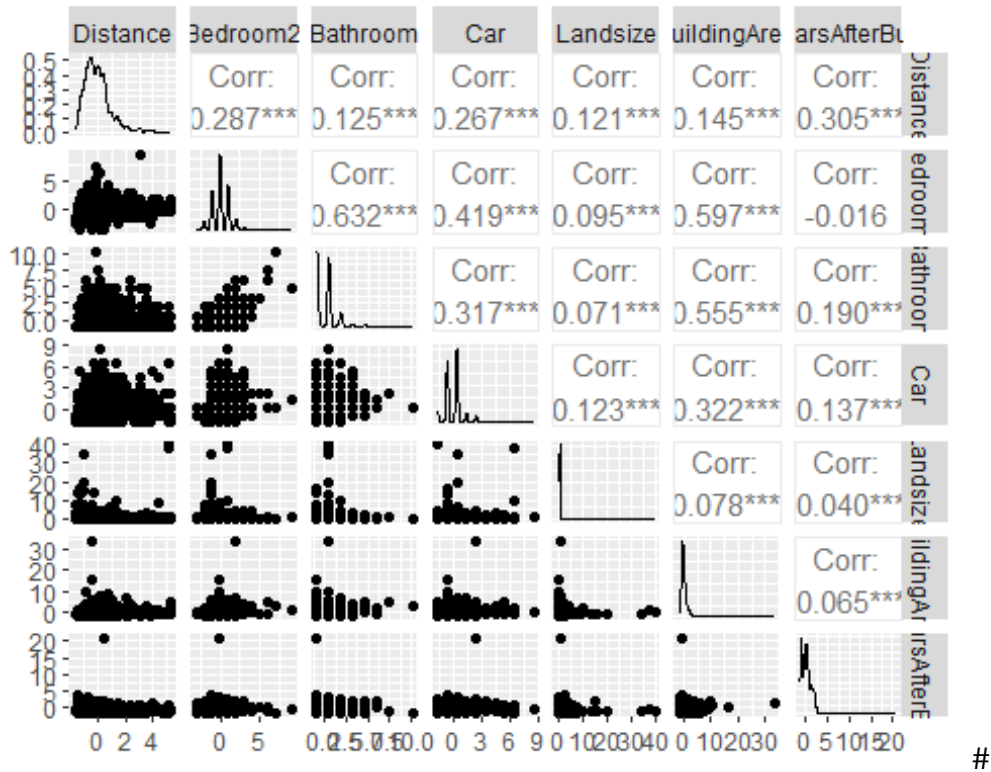
#lm(formula = LogPricePerBuildingArea ~ Type + Distance + Bedroom2 +
#   Bathroom + Car + Landsize + BuildingArea + CouncilArea +
#   YearsAfterBuilt, data = train_data)
# Select predictors for correlation analysis based on the final model (
# excluding non-numeric variables if they are not numerically encoded)
cor_data <- train_data[, c("Distance", "Bedroom2", "Bathroom", "Car", "
Landsize", "BuildingArea", "YearsAfterBuilt")]

# Compute pairwise correlations
correlation_matrix <- cor(cor_data)

# Print pairwise correlations
print(correlation_matrix)
```

##	Distance	Bedroom2	Bathroom	Car	Landsize
## Distance 72475	1.0000000	0.28744646	0.12531840	0.2668261	0.12072475
## Bedroom2 91182	0.2874465	1.0000000	0.63180352	0.4185207	0.09491182
## Bathroom 18188	0.1253184	0.63180352	1.0000000	0.3171892	0.0718188
## Car 62196	0.2668261	0.41852066	0.31718916	1.0000000	0.12262196
## Landsize 00000	0.1207248	0.09491182	0.07118188	0.1226220	1.0000000
## BuildingArea 93123	0.1452202	0.59721723	0.55463622	0.3219861	0.07793123
## YearsAfterBuilt 93164	-0.3050505	-0.01601471	-0.18987369	-0.1371532	-0.03993164
##	BuildingArea	YearsAfterBuilt			
## Distance	0.14522024	-0.30505052			
## Bedroom2	0.59721723	-0.01601471			
## Bathroom	0.55463622	-0.18987369			
## Car	0.32198611	-0.13715316			
## Landsize	0.07793123	-0.03993164			
## BuildingArea	1.00000000	-0.06529981			
## YearsAfterBuilt	-0.06529981	1.00000000			

```
# Create a histogram grid for visualization
ggpairs(cor_data)
```



correlation graph display

```
# Load necessary Libraries
library(corrplot)

## corrplot 0.92 loaded

# Convert non-numeric columns to numeric
cor_data_numeric <- as.data.frame(sapply(cor_data, as.numeric))

# Compute pairwise correlations
correlation_matrix <- cor(cor_data_numeric)

# Print pairwise correlations
print(correlation_matrix)

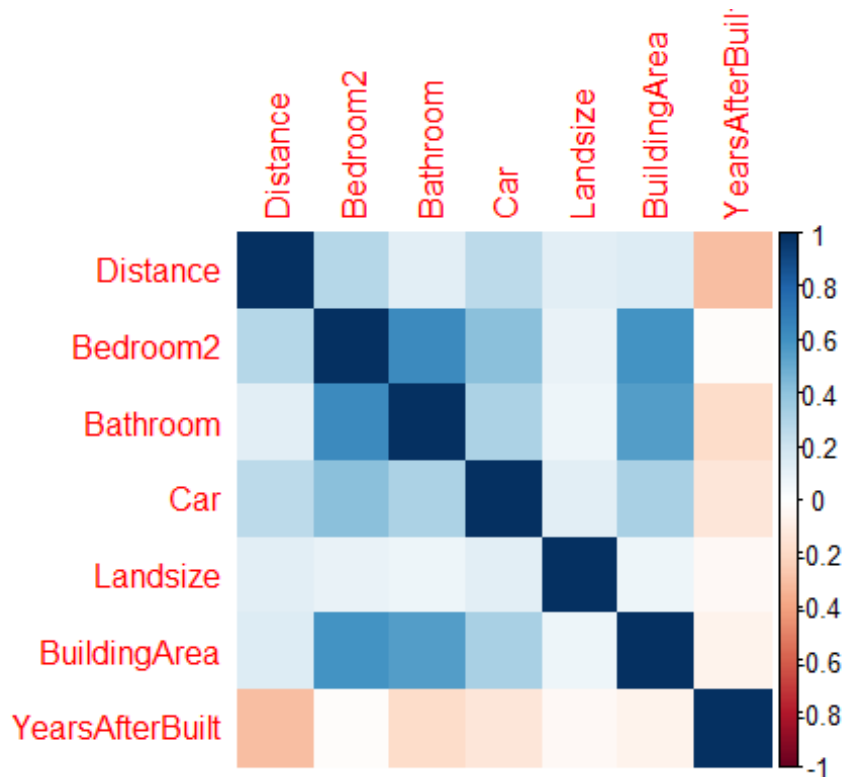
##           Distance    Bedroom2    Bathroom    Car    Lan
dsize
## Distance    1.0000000  0.28744646  0.12531840  0.2668261  0.120
72475
## Bedroom2    0.2874465  1.00000000  0.63180352  0.4185207  0.094
91182
## Bathroom    0.1253184  0.63180352  1.00000000  0.3171892  0.071
18188
## Car         0.2668261  0.41852066  0.31718916  1.0000000  0.122
```

```

62196
## Landsize      0.1207248  0.09491182  0.07118188  0.1226220  1.000
00000
## BuildingArea  0.1452202  0.59721723  0.55463622  0.3219861  0.077
93123
## YearsAfterBuilt -0.3050505 -0.01601471 -0.18987369 -0.1371532 -0.039
93164
##              BuildingArea YearsAfterBuilt
## Distance      0.14522024      -0.30505052
## Bedroom2      0.59721723      -0.01601471
## Bathroom      0.55463622      -0.18987369
## Car           0.32198611      -0.13715316
## Landsize     0.07793123      -0.03993164
## BuildingArea  1.00000000      -0.06529981
## YearsAfterBuilt -0.06529981      1.00000000

# Create a correlation plot with color
corrplot(correlation_matrix, method = "color")

```



Actual vs predicted graph

```

# Calculate predicted values
predicted_values <- predict(final_model, train_data)

# Extract actual values
actual_values <- train_data$LogPricePerBuildingArea

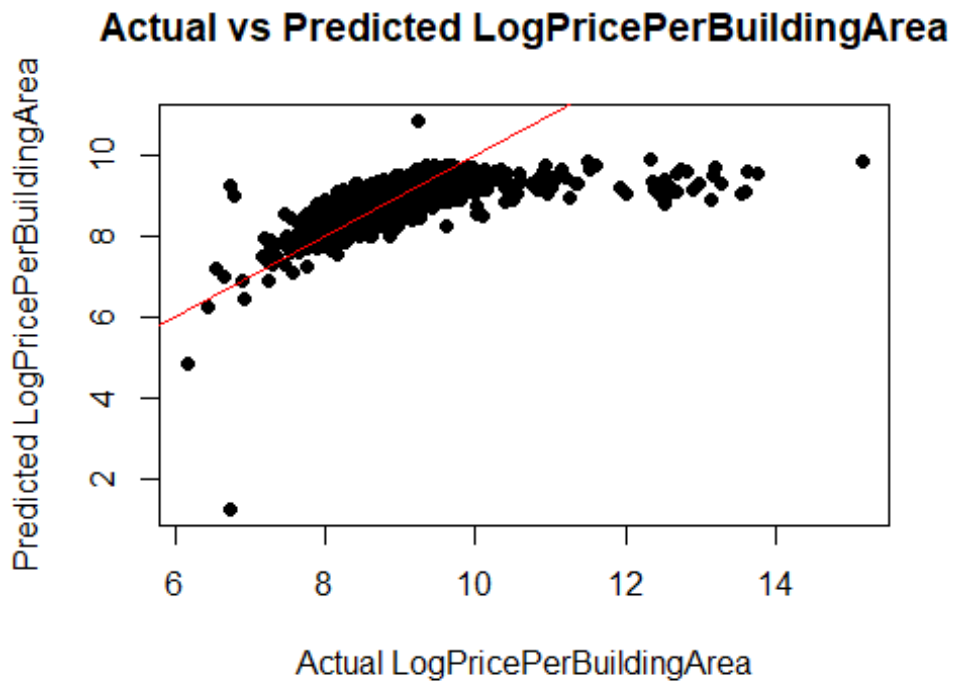
```

```

# Create a scatter plot of actual vs predicted values
plot(actual_values, predicted_values,
      main = "Actual vs Predicted LogPricePerBuildingArea",
      xlab = "Actual LogPricePerBuildingArea",
      ylab = "Predicted LogPricePerBuildingArea",
      pch = 19) # pch = 19 makes the points solid

# Add a line of perfect fit for reference
abline(a = 0, b = 1, col = "red")

```



qq and

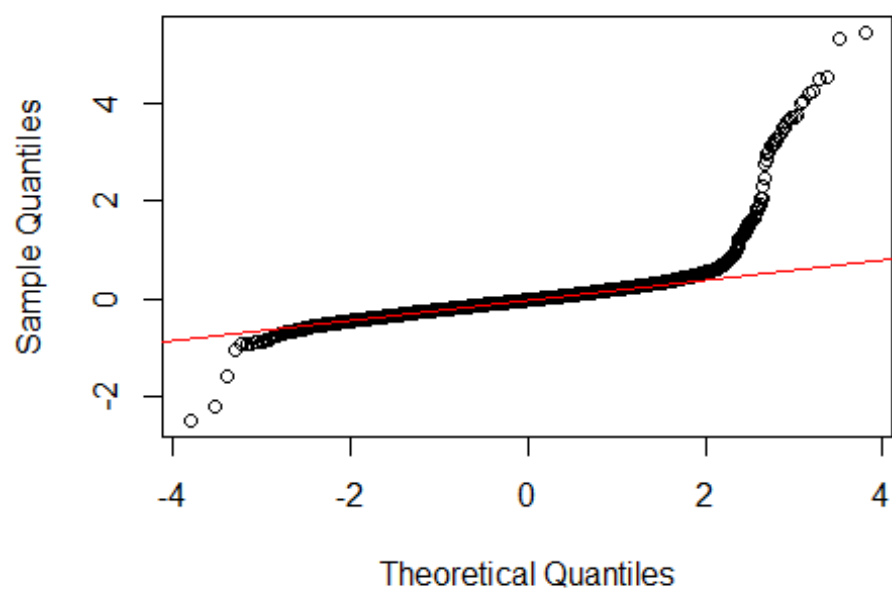
residual plots

```

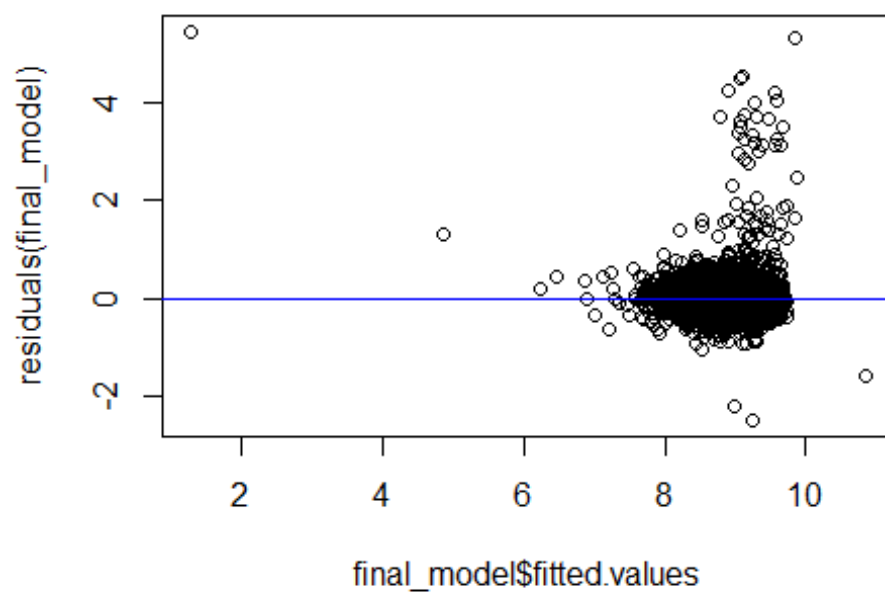
# QQ plot for the first model
qqnorm(residuals(final_model))
qqline(residuals(final_model), col = "red")

```

Normal Q-Q Plot



```
# Residual plot for the first model  
plot(final_model$fitted.values, residuals(final_model))  
abline(h = 0, col = "blue")
```



Comment of results

Comment of business implications