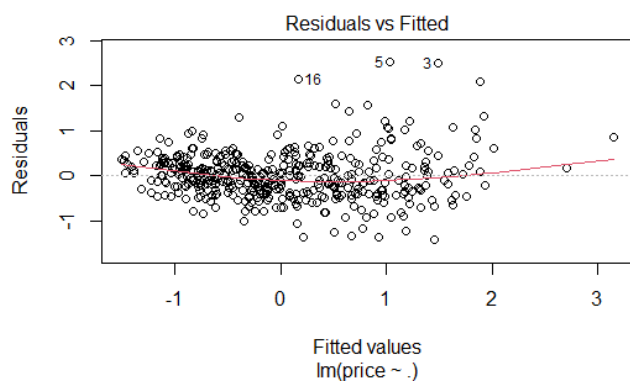


MICRO PROJECT



```
> summary(lm_model)
```

Call:

```
lm(formula = price ~ ., data = train_data)
```

Residuals:

```
      Min       1Q   Median       3Q      Max
-1.43425 -0.33007 -0.03193  0.27224  2.52018
```

Coefficients: (1 not defined because of singularities)

| | Estimate | Std. Error | t value | Pr(> t) | |
|----------------------------------|----------|------------|---------|----------|-----|
| (Intercept) | -0.24018 | 0.06367 | -3.773 | 0.000185 | *** |
| area | 0.26829 | 0.02910 | 9.219 | < 2e-16 | *** |
| bedrooms | 0.02672 | 0.02965 | 0.901 | 0.368120 | |
| bathrooms | 0.26896 | 0.02890 | 9.307 | < 2e-16 | *** |
| stories | 0.23637 | 0.03118 | 7.582 | 2.19e-13 | *** |
| parking | 0.13578 | 0.02808 | 4.836 | 1.86e-06 | *** |
| mainroadno | -0.22699 | 0.07846 | -2.893 | 0.004014 | ** |
| mainroadyes | NA | NA | NA | NA | |
| guestroomyes | 0.18028 | 0.07141 | 2.525 | 0.011945 | * |
| basementyes | 0.19691 | 0.06085 | 3.236 | 0.001307 | ** |
| hotwaterheatingyes | 0.59405 | 0.12093 | 4.912 | 1.29e-06 | *** |
| airconditioningyes | 0.47084 | 0.06111 | 7.705 | 9.46e-14 | *** |
| prefareayes | 0.32824 | 0.06500 | 5.050 | 6.59e-07 | *** |
| `furnishingstatussemi-furnished` | -0.03091 | 0.06528 | -0.474 | 0.636086 | |
| furnishingstatusunfurnished | -0.21678 | 0.07081 | -3.062 | 0.002342 | ** |

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.5368 on 422 degrees of freedom
Multiple R-squared: 0.7177, Adjusted R-squared: 0.709
F-statistic: 82.53 on 13 and 422 DF, p-value: < 2.2e-16

```
>
> # Make predictions on the test set
> predictions <- predict(lm_model, newdata = test_data)
>
> # Evaluate regression metrics
> cat("Metrics:\n")
Metrics:
> cat("MAE:", mae(predictions, test_data$price), "\n")
MAE: 0.5093771
> cat("RMSE:", rmse(predictions, test_data$price), "\n")
RMSE: 0.694099
```

```

# Load necessary libraries
library(dplyr)
library(tidyr)
library(Metrics)

# Load the data
data <- read.csv("C:\\Users\\online\\Documents\\ML LAB\\Micro Project\\housing.csv")

# Preprocessing
# 4.1: Handling missing values
data <- na.omit(data)
# 4.2: Removing duplicate rows
data <- distinct(data)
# 4.3: Scaling numeric variables
numeric_columns <- c("price", "area", "bedrooms", "bathrooms", "stories", "parking")
data[numeric_columns] <- scale(data[numeric_columns])
# 4.4: Encoding categorical variables
categorical_columns <- c("mainroad", "guestroom",
  "basement", "hotwaterheating", "airconditioning", "prefarea", "furnishingstatus")
data <- model.matrix(~. - 1, data = data[, c(numeric_columns, categorical_columns)])
# Split the data into training and testing sets
set.seed(123) # for reproducibility
train_indices <- sample(seq_len(nrow(data)), size = 0.8 * nrow(data))
train_data <- data[train_indices, ]
test_data <- data[-train_indices, ]
# Convert train_data matrix to data frame
train_data <- as.data.frame(train_data)
test_data <- as.data.frame(test_data)
# Fit Multiple Linear Regression model
lm_model <- lm(price ~ ., data = train_data)
# Print the Multiple Linear Regression model summary
summary(lm_model)
# Make predictions on the test set
predictions <- predict(lm_model, newdata = test_data)
# Evaluate regression metrics
cat("Metrics:\n")
cat("MAE:", mae(predictions, test_data$price), "\n")
cat("RMSE:", rmse(predictions, test_data$price), "\n")
plot(lm_model, which = 1)

```