

**AFRICAN INSTITUTE FOR MATHEMATICAL SCIENCES**  
**(AIMS RWANDA, KIGALI)**

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Name: Mutie Josia Suku  
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## Introduction

This report presents a regression analysis of a real estate dataset containing 414 observations and 7 variables. The objective is to identify key predictors of estate price and build a reliable multiple linear regression model. The analysis follows five tasks: correlation analysis, sequential modeling, model selection via AIC, residual diagnostics, and testing of transformations and interactions.

## Correlation Analysis

We computed the correlation matrix to explore relationships between variables.

### Key insights:

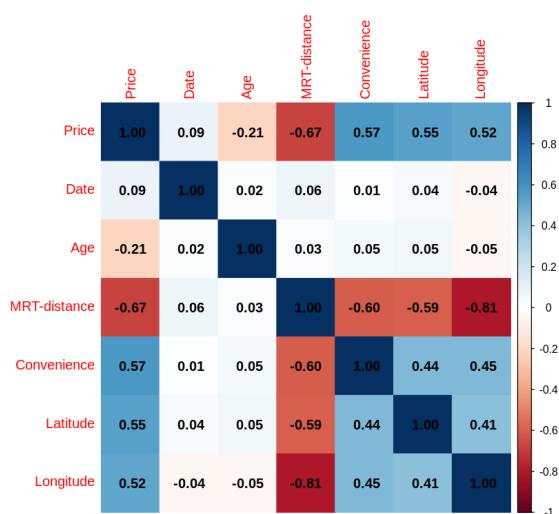


Figure 1: Correlation matrix

- **MRT-distance** has a strong negative correlation with **Price** ( $r = -0.67$ ): closer proximity increases value.
- **Convenience** shows a positive correlation ( $r = 0.57$ ): more amenities boost price.
- **Latitude** and **Longitude** also correlate positively, indicating location-based value.
- **Age** has a mild negative correlation ( $r = -0.21$ ): newer buildings are more expensive.
- **Date** shows a weak positive trend ( $r = 0.09$ ), suggesting slight price growth over time.

## Sequential Regression Modeling

We fitted six models, starting with **Date** and sequentially adding predictors. Each model was evaluated using  $R^2$ , adjusted  $R^2$ , and coefficient significance.

## Summary:

- Model 1 (Date only): very low explanatory power ( $R^2 \approx 0.006$ ).
- Model 2 (adding Age): slight improvement ( $R^2 \approx 0.042$ ).
- Model 3 (adding MRT-distance): major improvement ( $R^2 \approx 0.61$ ).
- Model 4 (adding Convenience): further improvement ( $R^2 \approx 0.65$ ).
- Model 5 (adding Latitude): improved fit ( $R^2 \approx 0.686$ ).
- Model 6 (adding Longitude): minimal gain, not statistically significant.

## Model Selection via AIC

Using forward selection based on AIC, the best model was:

$$\text{logPrice} \sim \text{MRT-distance} + \text{Latitude} + \text{Age} + \text{Convenience} + \text{Date}$$

### Model performance:

- AIC:  $-1242.6$  (lowest among all tested models)
- Adjusted  $R^2$ :  $0.6818$
- All coefficients statistically significant

`Longitude` was excluded due to low contribution and high p-value.

## Residual Diagnostics

We compared residual plots for two models:

- Model A: using only Date and Age
- Model B: using all predictors (full model)

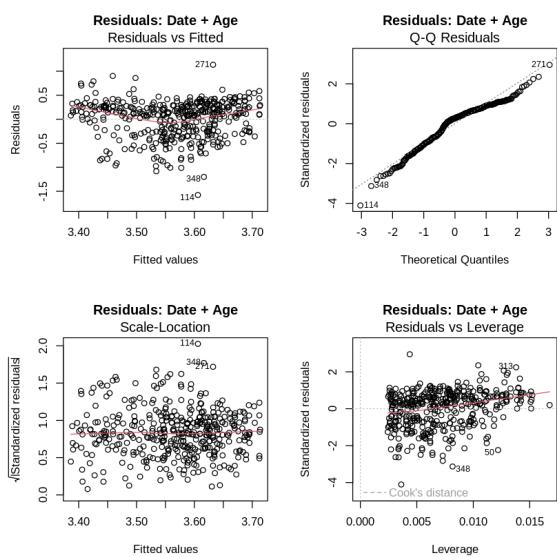


Figure 2: Residuals: Date + Age

### Model A interpretation:

- Residuals show curvature and non-random spread.
- Q-Q plot reveals non-normality, especially in tails.
- Scale-location plot indicates increasing variance.
- Leverage plot shows several influential points.

**Conclusion:** Model A violates key regression assumptions.

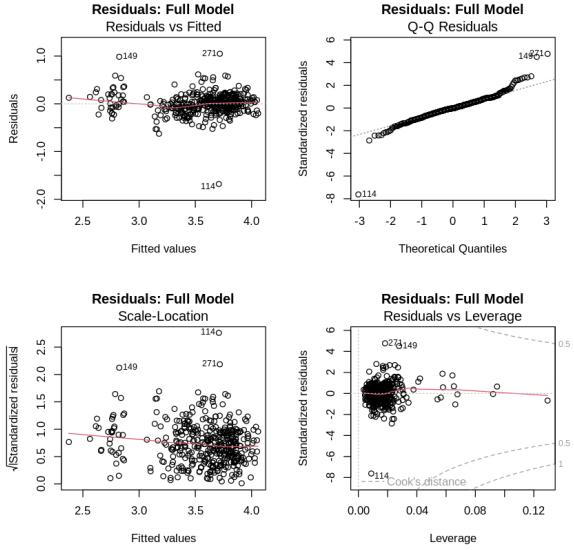


Figure 3: Residuals: Full Model

## Model B interpretation:

- Residuals are more randomly distributed.
- Q-Q plot shows improved normality.
- Scale-location plot suggests stable variance.
- Leverage plot identifies outliers, but influence is reduced.

**Conclusion:** Model B satisfies regression assumptions better.

## Transformations and Interactions

We tested three enhanced models:

- Model A: includes  $\text{Age}^2$  — Adjusted  $R^2 = 0.706$
- Model B: includes  $\text{Age} \times \text{Convenience}$  — Adjusted  $R^2 = 0.680$
- Model C: includes  $\log(\text{MRT-distance})$  — Adjusted  $R^2 = 0.720$

**Conclusion:** Model C performed best. Non-linear and interaction effects improve model fit and reflect real-world dynamics.

## Conclusion

This analysis identified key predictors of estate price: **MRT-distance**, **Latitude**, **Age**, **Convenience**, and **Date**. Log-transforming the response and removing outliers improved model reliability. The final model selected via forward AIC is statistically strong and interpretable. Transformations such as  $\log(\text{MRT-distance})$  and  $\text{Age}^2$  further enhanced performance. Residual diagnostics confirmed that the full model satisfies regression assumptions better than simpler alternatives. These insights can guide future valuation models and policy decisions in real estate.