

## Prodigy ML Internship – Task 1

### Linear Regression for House Price Prediction

#### 1. Objective

Implement a linear regression model to predict house sale prices using:

- GrLivArea (above-grade living area)
- BedroomAbvGr (bedrooms above grade)
- TotalBath = FullBath + 0.5×HalfBath

#### 2. Dataset

Kaggle “House Prices – Advanced Regression Techniques”

- Train: 1,460 rows, 81 columns
- No missing values in selected features

#### 3. Methodology

- a) Load train.csv → rename to dataset.csv
- b) Engineer TotalBath
- c) Train-test split (80/20, random\_state=42)
- d) Fit sklearn.LinearRegression
- e) Evaluate with  $R^2$ , RMSE, MAE

#### 4. Results (example run)

$R^2$  : 0.6398

RMSE : \$70,124

MAE : \$49,832

Coefficients:

GrLivArea  $\approx$  \$109.6 per sq ft

BedroomAbvGr  $\approx$  \$15,200 per bedroom

TotalBath  $\approx$  \$39,800 per bath

Intercept  $\approx$  -\$48,300

#### 5. Visualizations

- Actual vs Predicted scatter (model\_plots.png)
- Residual plot (random scatter around zero)
- Correlation heatmap

#### 6. Interpretation

- Square footage is the dominant driver (correlation 0.71 with SalePrice).
- Adding bedrooms & bathrooms improves  $R^2$  from ~0.50 to 0.64.
- Model is simple, interpretable, and meets the task requirement.

#### 7. Possible Extensions

- Log-transform SalePrice → optimize RMSLE (Kaggle metric)
- Include basement baths, garage area, year built

- Try Ridge/Lasso for regularization

#### 8. Disclaimer

“This project is based on an open-source example from GitHub, modified to include additional features and improvements.”

Author: Tsehay Araya Hailemariam

Date: October 2025