

# Linear Regression Errors Explained

## Linear Regression: Theory, Problems, and Solutions

### 1. Theory of Linear Regression

Linear regression is a statistical method used to model the relationship between a dependent variable  $Y$  and one or more independent variables  $X$ . It is commonly used for predictive modeling. The equation of a simple linear regression model is:

$$Y = mX + c + \epsilon$$

Where:

- $Y$  is the dependent variable (response)
- $X$  is the independent variable (predictor)
- $m$  (or  $\beta_1$ ) is the slope of the line
- $c$  (or  $\beta_0$ ) is the intercept
- $\epsilon$  is the error term (unexplained variance)

For multiple regression, the equation extends to:

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_n X_n + \epsilon$$

where multiple independent variables ( $X_1, X_2, \dots, X_n$ ) influence  $Y$ .

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### Error Metrics in Linear Regression

To measure how well a linear regression model fits the data, various error metrics are used:

#### 1. Mean Absolute Error (MAE):

$$MAE = \frac{1}{n} \sum_{i=1}^n |Y_i - \hat{Y}_i|$$

Measures the average absolute difference between actual and predicted values.

#### 2. Mean Squared Error (MSE):

$$MSE = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$

Measures the average squared differences between actual and predicted values.

### 3. Root Mean Squared Error (RMSE):

$$RMSE = \sqrt{MSE}$$

Provides a more interpretable measure by taking the square root of MSE.

### 4. R-Squared ( $R^2$ ):

$$R^2 = 1 - \frac{\sum(Y_i - \hat{Y}_i)^2}{\sum(Y_i - \bar{Y})^2}$$

Measures the proportion of variance in  $Y$  explained by  $X$ . A value close to 1 indicates a better fit.

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## Problem 1: Predicting House Prices Based on Size

### Given Data:

House Size (sq. ft)	Price (in \$1000s)
1400	245
1600	312
1700	279
1875	308
1100	199

### Solution:

We need to fit a linear regression model:

$$Price = m \times Size + c$$

#### Step 1: Calculate Mean Values

$$\bar{X} = \frac{1400 + 1600 + 1700 + 1875 + 1100}{5} = 1535$$

$$\bar{Y} = \frac{245 + 312 + 279 + 308 + 199}{5} = 268.6$$

#### Step 2: Calculate Slope $m$ and Intercept $c$

$$m = \frac{\sum(X_i - \bar{X})(Y_i - \bar{Y})}{\sum(X_i - \bar{X})^2}$$

$$m = \frac{(1400 - 1535)(245 - 268.6) + (1600 - 1535)(312 - 268.6) + (1700 - 1535)(279 - 268.6) + (1875 - 1535)(308 - 268.6) + (1100 - 1535)(199 - 268.6)}{(1400 - 1535)^2 + (1600 - 1535)^2 + (1700 - 1535)^2 + (1875 - 1535)^2 + (1100 - 1535)^2}$$

$$m = \frac{(-135)(-23.6) + (65)(43.4) + (165)(10.4) + (340)(39.4) + (-435)(-69.6)}{(-135)^2 + (65)^2 + (165)^2 + (340)^2 + (-435)^2}$$

$$m = \frac{3186 + 2821 + 1716 + 13396 + 30216}{18225 + 4225 + 27225 + 115600 + 189225}$$

$$m = \frac{51335}{354500} = 0.145$$

$$c = \bar{Y} - m \times \bar{X}$$

$$c = 268.6 - (0.145 \times 1535)$$

$$c = 268.6 - 222.6 = 46$$

Thus, the regression equation is:

$$Price = 0.145 \times Size + 46$$

### Step 3: Predicting and Evaluating Errors

Using this model, we predict prices and compute errors.

House Size	Actual Price	Predicted Price	Error (Actual - Predicted)	Squared Error
1400	245	$0.145(1400)+46 = 248.3$	-3.3	10.89
1600	312	$0.145(1600)+46 = 277.2$	34.8	1211.04
1700	279	$0.145(1700)+46 = 291.7$	-12.7	161.29
1875	308	$0.145(1875)+46 = 318.2$	-10.2	104.04
1100	199	$0.145(1100)+46 = 205.9$	-6.9	47.61

**MAE:**

$$MAE = \frac{|-3.3| + |34.8| + |-12.7| + |-10.2| + |-6.9|}{5}$$

$$MAE = \frac{67.9}{5} = 13.58$$

**MSE:**

$$MSE = \frac{10.89 + 1211.04 + 161.29 + 104.04 + 47.61}{5}$$

$$MSE = \frac{1534.87}{5} = 306.97$$

**RMSE:**

$$RMSE = \sqrt{306.97} = 17.52$$

**R-Squared Calculation:**

$$R^2 = 1 - \frac{\sum(Y_i - \hat{Y}_i)^2}{\sum(Y_i - \bar{Y})^2}$$

$$R^2 = 1 - \frac{1534.87}{5602.8} = 0.726$$

Thus, the model explains **72.6%** of the variance in house prices.

Problem 2: Predicting Student Scores Based on Study Hours

Given Data:

Study Hours (X)	Exam Score (Y)
1.5	55
3.0	60
4.5	68
5.0	72
6.0	80

We will fit a linear regression model:

$$Y = mX + c$$

Step 1: Calculate Mean Values

$$\bar{X} = \frac{1.5 + 3.0 + 4.5 + 5.0 + 6.0}{5} = 4.0$$

$$\bar{Y} = \frac{55 + 60 + 68 + 72 + 80}{5} = 67.0$$

Step 2: Calculate Slope *m* and Intercept *c*

The formula for slope:

$$m = \frac{\sum(X_i - \bar{X})(Y_i - \bar{Y})}{\sum(X_i - \bar{X})^2}$$
$$m = \frac{(1.5 - 4.0)(55 - 67) + (3.0 - 4.0)(60 - 67) + (4.5 - 4.0)(68 - 67) + (5.0 - 4.0)(72 - 67) + (6.0 - 4.0)(80 - 67)}{(1.5 - 4.0)^2 + (3.0 - 4.0)^2 + (4.5 - 4.0)^2 + (5.0 - 4.0)^2 + (6.0 - 4.0)^2}$$
$$m = \frac{(-2.5)(-12) + (-1.0)(-7) + (0.5)(1) + (1.0)(5) + (2.0)(13)}{(-2.5)^2 + (-1.0)^2 + (0.5)^2 + (1.0)^2 + (2.0)^2}$$

$$m = \frac{30 + 7 + 0.5 + 5 + 26}{6.25 + 1.0 + 0.25 + 1.0 + 4.0}$$

$$m = \frac{68.5}{12.5} = 5.48$$

Now, calculate the intercept:

$$c = \bar{Y} - m \times \bar{X}$$

$$c = 67 - (5.48 \times 4)$$

$$c = 67 - 21.92 = 45.08$$

Thus, the regression equation is:

$$\text{Score} = 5.48 \times \text{Study Hours} + 45.08$$

### Step 3: Predicting Scores and Errors

Study Hours	Actual Score	Predicted Score $5.48X + 45.08$	Error (Actual - Predicted)	Squared Error
1.5	55	$5.48(1.5) + 45.08 = 53.4$	$55 - 53.4 = 1.6$	2.56
3.0	60	$5.48(3.0) + 45.08 = 61.5$	$60 - 61.5 = -1.5$	2.25
4.5	68	$5.48(4.5) + 45.08 = 69.6$	$68 - 69.6 = -1.6$	2.56
5.0	72	$5.48(5.0) + 45.08 = 72.5$	$72 - 72.5 = -0.5$	0.25
6.0	80	$5.48(6.0) + 45.08 = 77.9$	$80 - 77.9 = 2.1$	4.41

### Step 4: Compute Error Metrics

Mean Absolute Error (MAE):

$$MAE = \frac{|1.6| + |1.5| + |1.6| + |0.5| + |2.1|}{5}$$

$$MAE = \frac{7.3}{5} = 1.46$$

Mean Squared Error (MSE):

$$MSE = \frac{2.56 + 2.25 + 2.56 + 0.25 + 4.41}{5}$$

$$MSE = \frac{12.03}{5} = 2.41$$

Root Mean Squared Error (RMSE):

$$RMSE = \sqrt{MSE} = \sqrt{2.41} = 1.55$$

R-Squared ( $R^2$ ) Calculation:

$$R^2 = 1 - \frac{\sum(Y_i - \hat{Y}_i)^2}{\sum(Y_i - \bar{Y})^2}$$

Total variation:

$$\begin{aligned}\sum(Y_i - \bar{Y})^2 &= (55 - 67)^2 + (60 - 67)^2 + (68 - 67)^2 + (72 - 67)^2 + (80 - 67)^2 \\ &= (-12)^2 + (-7)^2 + (1)^2 + (5)^2 + (13)^2 \\ &= 144 + 49 + 1 + 25 + 169 = 388\end{aligned}$$

Residual variation:

$$\begin{aligned}\sum(Y_i - \hat{Y}_i)^2 &= 12.03 \\ R^2 &= 1 - \frac{12.03}{388} = 0.969\end{aligned}$$

Thus, the model explains **96.9%** of the variance in exam scores.

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## Final Summary of Results:

### Problem 1: Predicting House Prices

- **Regression Equation:**  $Y = 0.145X + 46$
- **MAE:** 13.58
- **MSE:** 306.97
- **RMSE:** 17.52
- $R^2$ : 72.6%

### Problem 2: Predicting Exam Scores

- **Regression Equation:**  $Y = 5.48X + 45.08$
  - **MAE:** 1.46
  - **MSE:** 2.41
  - **RMSE:** 1.55
  - $R^2$ : 96.9%
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## Conclusion

- The **house price model** has an  $R^2$  of **72.6%**, meaning it explains 72.6% of price variations.
- The **exam score model** has an  $R^2$  of **96.9%**, indicating a **very strong correlation**.
- Both models perform well, but the exam score model fits better due to lower error values and higher  $R^2$ .