

## MSE, RMSE, MAE

### MSE

Mean Squared Error

$$\frac{\sum (y_i - \hat{y}_i)^2}{n}$$

- 🎯 Squares differences
- ⚡ Heavily penalizes large errors
- 📊 Squared units

### RMSE

Root Mean Squared Error

$$\sqrt{\frac{\sum (y_i - \hat{y}_i)^2}{n}}$$

- 🎯 Square root of MSE
- ⚡ Same units as target
- 📊 More interpretable

### MAE

Mean Absolute Error

$$\frac{\sum |y_i - \hat{y}_i|}{n}$$

- 🎯 Absolute differences
- ⚡ Treats all errors equally
- 📊 Robust to outliers

### Quick Comparison

Property	MSE	RMSE	MAE
Outlier Sensitivity	High ⚠️	High ⚠️	Low ✓
Interpretability	Low	High ✓	High ✓
Units	Squared	Original ✓	Original ✓



Key Insight: Lower values = Better performance • Choose based on outlier importance