

Loss Function Definition

A loss function $L(\hat{y}, y)$ measures the discrepancy between the predicted output \hat{y} and the true target y , guiding the network to minimize prediction errors during training.

Regression Tasks

Mean Squared Error (MSE)

$$L = (1/n) \sum (\hat{y}_i - y_i)^2$$

- **Penalizes large errors** quadratically
- **Differentiable** everywhere
- **Sensitive to outliers**

Classification Tasks

Cross-Entropy Loss

$$L = -\sum y_i \log(\hat{y}_i)$$

- **Measures probability distribution difference**
- **Works with softmax output**
- **Penalizes confident mistakes** heavily

Example Calculation

Prediction	-	Target	→	Loss
5.2	-	4.0	→	1.44

Example Calculation

Prediction	-	Target	→	Loss
0.7	log	1	→	0.36

Common Loss Functions

MSE

Task: Regression

Formula: $\sum (\hat{y} - y)^2$

Use Case: Continuous values

Cross-Entropy

Task: Classification

Formula: $-\sum y \log(\hat{y})$

Use Case: Class probabilities

MAE

Task: Regression

Formula: $\sum |\hat{y} - y|$

Use Case: Robust to outliers

🎯 Training Objective

The goal is to find parameters θ that minimize the average loss: $\theta^* = \operatorname{argmin} (1/n) \sum L(f(x_i; \theta), y_i)$