

Machine Learning Mathematical Notation Cheat Sheet

Basic Variables and Sets

X: Input features (independent variables)

y: Output labels (dependent variable)

m: Number of training examples

n: Number of features

i: Index for examples ($1 \leq i \leq m$)

j: Index for features ($1 \leq j \leq n$)

Vectors and Matrices

$x^{(i)}$: Feature vector for the i-th example

$y^{(i)}$: Label for the i-th example

θ (θ): Parameter vector (weights of the model)

w: Alternative notation for weights

b: Bias term (intercept)

X in $\mathbb{R}^{m \times n}$: Feature matrix (m examples, n features)

y in \mathbb{R}^m : Output vector

Functions and Operations

$h_{\theta}(x)$: Hypothesis function or model prediction

$\sigma(z)$: Sigmoid function $\rightarrow 1 / (1 + e^{-z})$

sum: Summation

d: Partial derivative

argmin: Argument of the minimum (used in optimization)

Common Equations

Linear Regression Hypothesis:

$$h_{\theta}(x) = \theta^T x + b$$

Logistic Regression Hypothesis:

$$h_{\theta}(x) = \sigma(\theta^T x + b)$$

Cost Function for Linear Regression (MSE):

$$J(\theta) = (1/2m) \sum (h_{\theta}(x^{(i)}) - y^{(i)})^2$$

Cost Function for Logistic Regression (Log Loss):

$$J(\theta) = -(1/m) \sum [y^{(i)} \log(h_{\theta}(x^{(i)})) + (1 - y^{(i)}) \log(1 - h_{\theta}(x^{(i)}))]$$

Gradient Descent Update Rule:

$$\theta_j := \theta_j - \alpha \frac{dJ(\theta)}{d\theta_j}$$

Where:

α is the learning rate

Probability and Statistics Notations

$P(y|x)$: Probability of y given x

$E[x]$: Expected value of x

$\text{Var}(x)$: Variance of x

μ (μ): Mean

σ^2 : Variance

$N(\mu, \sigma^2)$: Normal distribution with mean μ and variance σ^2

Optimization Notation

min: Minimize

max: Maximize

$\|\theta\|_1$: L1 norm (sum of absolute values of θ) [used in Lasso]

$\|\theta\|_2^2$: L2 norm squared (sum of squares of θ) [used in Ridge]