

# Derivation of Linear Regression Loss Function Minimization

The loss function for linear regression with  $n$  data points is given by:

$$J(m, c) = \sum_{i=1}^n (y_i - mx_i - c)^2$$

## Quadratic in $m$ (treating $c$ as fixed)

To express  $J(m, c)$  as a quadratic in  $m$ , expand the squared term for each  $i$ :

$$(y_i - mx_i - c)^2 = (y_i - c - mx_i)^2 = (y_i - c)^2 - 2mx_i(y_i - c) + m^2x_i^2$$

Sum over all  $i$  from 1 to  $n$ :

$$\begin{aligned} \sum_{i=1}^n (y_i - mx_i - c)^2 &= \sum_{i=1}^n [(y_i - c)^2 - 2mx_i(y_i - c) + m^2x_i^2] \\ &= \sum_{i=1}^n (y_i - c)^2 - 2m \sum_{i=1}^n x_i(y_i - c) + m^2 \sum_{i=1}^n x_i^2 \end{aligned}$$

Expand the middle term:

$$\sum_{i=1}^n x_i(y_i - c) = \sum_{i=1}^n x_i y_i - c \sum_{i=1}^n x_i$$

Thus,

$$\begin{aligned} J(m, c) &= \sum_{i=1}^n (y_i - c)^2 - 2m \left( \sum_{i=1}^n x_i y_i - c \sum_{i=1}^n x_i \right) + m^2 \sum_{i=1}^n x_i^2 \\ &= \sum_{i=1}^n (y_i - c)^2 - 2m \sum_{i=1}^n x_i y_i + 2mc \sum_{i=1}^n x_i + m^2 \sum_{i=1}^n x_i^2 \end{aligned}$$

This is a quadratic in  $m$  of the form  $am^2 + bm + d$ , where:

$$a = \sum_{i=1}^n x_i^2, \quad b = -2 \sum_{i=1}^n x_i y_i + 2c \sum_{i=1}^n x_i, \quad d = \sum_{i=1}^n (y_i - c)^2$$

The value of  $m$  that minimizes  $J$  with respect to  $m$  occurs at  $-\frac{b}{2a}$ :

$$m = -\frac{b}{2a} = -\frac{-2 \sum_{i=1}^n x_i y_i + 2c \sum_{i=1}^n x_i}{2 \sum_{i=1}^n x_i^2} = \frac{2 \sum_{i=1}^n x_i y_i - 2c \sum_{i=1}^n x_i}{2 \sum_{i=1}^n x_i^2} = \frac{\sum_{i=1}^n x_i y_i - c \sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2}$$

## Quadratic in $c$ (treating $m$ as fixed)

To express  $J(m, c)$  as a quadratic in  $c$ , expand the squared term:

$$(y_i - mx_i - c)^2 = (y_i - mx_i)^2 - 2c(y_i - mx_i) + c^2$$

Sum over all  $i$ :

$$\sum_{i=1}^n (y_i - mx_i - c)^2 = \sum_{i=1}^n (y_i - mx_i)^2 - 2c \sum_{i=1}^n (y_i - mx_i) + \sum_{i=1}^n c^2$$

Since  $c^2$  is constant,  $\sum_{i=1}^n c^2 = nc^2$ . Expand the middle term:

$$\sum_{i=1}^n (y_i - mx_i) = \sum_{i=1}^n y_i - m \sum_{i=1}^n x_i$$

Thus,

$$\begin{aligned} J(m, c) &= \sum_{i=1}^n (y_i - mx_i)^2 - 2c \left( \sum_{i=1}^n y_i - m \sum_{i=1}^n x_i \right) + nc^2 \\ &= \sum_{i=1}^n (y_i - mx_i)^2 - 2c \sum_{i=1}^n y_i + 2cm \sum_{i=1}^n x_i + nc^2 \end{aligned}$$

This is a quadratic in  $c$  of the form  $ac^2 + bc + d$ , where:

$$a = n, \quad b = -2 \sum_{i=1}^n y_i + 2m \sum_{i=1}^n x_i, \quad d = \sum_{i=1}^n (y_i - mx_i)^2$$

The value of  $c$  that minimizes  $J$  with respect to  $c$  occurs at  $-\frac{b}{2a}$ :

$$c = -\frac{b}{2a} = -\frac{-2 \sum_{i=1}^n y_i + 2m \sum_{i=1}^n x_i}{2n} = \frac{2 \sum_{i=1}^n y_i - 2m \sum_{i=1}^n x_i}{2n} = \frac{\sum_{i=1}^n y_i - m \sum_{i=1}^n x_i}{n}$$

## Solving for $m$ and $c$ by Substitution

To find the values of  $m$  and  $c$  that jointly minimize  $J$ , use the expressions derived:

$$\begin{aligned} c &= \frac{\sum_{i=1}^n y_i - m \sum_{i=1}^n x_i}{n} = \frac{1}{n} \sum_{i=1}^n y_i - \frac{m}{n} \sum_{i=1}^n x_i \\ m &= \frac{\sum_{i=1}^n x_i y_i - c \sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2} \end{aligned}$$

Substitute the expression for  $c$  into the expression for  $m$ :

$$m = \frac{\sum_{i=1}^n x_i y_i - \left( \frac{1}{n} \sum_{i=1}^n y_i - \frac{m}{n} \sum_{i=1}^n x_i \right) \sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2}$$

Expand the numerator:

$$\sum_{i=1}^n x_i y_i - \left( \frac{1}{n} \sum_{i=1}^n y_i \right) \left( \sum_{i=1}^n x_i \right) + \left( \frac{m}{n} \sum_{i=1}^n x_i \right) \left( \sum_{i=1}^n x_i \right)$$

$$= \sum_{i=1}^n x_i y_i - \frac{1}{n} \left( \sum_{i=1}^n y_i \right) \left( \sum_{i=1}^n x_i \right) + m \cdot \frac{1}{n} \left( \sum_{i=1}^n x_i \right) \left( \sum_{i=1}^n x_i \right)$$

So,

$$m = \frac{\sum_{i=1}^n x_i y_i - \frac{1}{n} (\sum_{i=1}^n y_i) (\sum_{i=1}^n x_i) + m \cdot \frac{1}{n} (\sum_{i=1}^n x_i) (\sum_{i=1}^n x_i)}{\sum_{i=1}^n x_i^2}$$

Multiply both sides by  $\sum_{i=1}^n x_i^2$ :

$$m \sum_{i=1}^n x_i^2 = \sum_{i=1}^n x_i y_i - \frac{1}{n} \left( \sum_{i=1}^n y_i \right) \left( \sum_{i=1}^n x_i \right) + m \cdot \frac{1}{n} \left( \sum_{i=1}^n x_i \right) \left( \sum_{i=1}^n x_i \right)$$

Move terms involving  $m$  to the left:

$$m \sum_{i=1}^n x_i^2 - m \cdot \frac{1}{n} \left( \sum_{i=1}^n x_i \right) \left( \sum_{i=1}^n x_i \right) = \sum_{i=1}^n x_i y_i - \frac{1}{n} \left( \sum_{i=1}^n y_i \right) \left( \sum_{i=1}^n x_i \right)$$

Solve for  $m$ :

$$m = \frac{\sum_{i=1}^n x_i y_i - \frac{1}{n} (\sum_{i=1}^n y_i) (\sum_{i=1}^n x_i)}{\sum_{i=1}^n x_i^2 - \frac{1}{n} (\sum_{i=1}^n x_i) (\sum_{i=1}^n x_i)}$$

The final formula for  $c$  in terms of  $m$  is:

$$c = \frac{1}{n} \sum_{i=1}^n y_i - \frac{m}{n} \sum_{i=1}^n x_i$$