

Computing Gradients for the Logistic Regression

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Linear function: $f(x) = xW + b$.

Logistic Regression function: $\text{Sigmoid}(f(x))$.

Log-loss: $-(y \log f(x) + (1 - y) \log (1 - f(x)))$

Mean Log-loss: $\text{mean}(-(y \log f(x) + (1 - y) \log (1 - f(x))))$

Recall the following:

$$\text{Sigmoid}(x)' = \left(\frac{1}{1 + e^{-x}} \right)' = \text{Sigmoid}(x)(1 - \text{Sigmoid}(x)) \quad (1)$$

We now calculate partial derivatives of the loss function with respect to weights and a bias.

Let $h = \text{Sigmoid}(f(x))$.

We have:

$$(-(y \log h + (1 - y) \log (1 - h)))' = -\frac{y}{h} \times h' - \frac{1 - y}{1 - h} \times (-h') \quad (2)$$

$$= h' * \frac{(1 - y)h - y(1 - h)}{h(1 - h)} \quad (3)$$

$$= h(1 - h)f'(x) \times \frac{(1 - y)h - y(1 - h)}{h(1 - h)} \quad (4)$$

$$= f'(x) \times (h - y) \quad (5)$$

$$= (xW + b)' \times (h - y) \quad (6)$$

Then the derivative w.r.t. W is

$$\boxed{\frac{1}{n} \sum_{i=0}^n x \cdot (\text{Sigmoid}(f(x_i)) - y_i)} \quad (7)$$

and the derivative w.r.t. b is

$$\boxed{\frac{1}{n} \sum_{i=0}^n \text{Sigmoid}(f(x_i)) - y_i} \quad (8)$$