

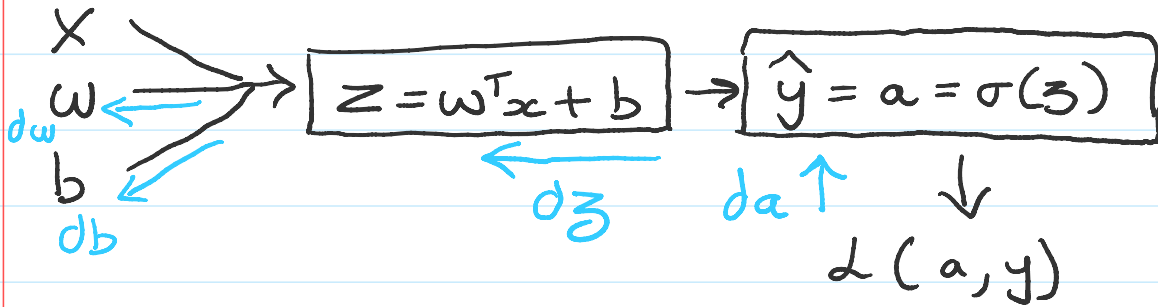
# Logistic Regression with Gradient descent

30 December 2018 02:33

$$z = w^T x + b$$

$$\hat{y} = a = \sigma(z)$$

$$\mathcal{L}(a, y) = - (y \log(a) + (1-y) \log(1-a))$$



$$da = d\mathcal{L}(a, y) / da$$

$$= d(-y \log a - (1-y) \log(1-a)) / da$$

$$= -y \frac{d}{da} \log a + (y-1) \frac{d}{da} \log(1-a)$$

$$= -y/a + (y-1) \frac{1}{(1-a)} \frac{d}{da} (1-a)$$

$$= -y/a + \frac{y-1}{(1-a)} (0-1)$$

$$= -y/a + \frac{(1-y)}{(1-a)}$$

$$= \frac{-y(1-a) + a(1-y)}{a(1-a)}$$

$$= \frac{-y + ya + a - ay}{a(1-a)}$$

$$= (a-y) / a(1-a)$$

$$dz = \partial L(a, y) / \partial z$$

$$= \frac{\partial L(a, y)}{\partial a} \times \frac{\partial a}{\partial z}$$

$$\left[ \frac{1}{u(x)} \right]' = \frac{u'(x)}{u(x)^2}$$

$$\frac{\partial a}{\partial z} = \frac{\partial (\sigma(z))}{\partial z}$$

$$x = 1 + e^{-z}$$

$$= \frac{\partial}{\partial z} \left( \frac{1}{1 + e^{-z}} \right)$$

$$= \frac{\partial}{\partial z} (1 + e^{-z})$$

$$= \frac{\frac{\partial}{\partial z} (1) + \frac{\partial}{\partial z} (e^{-z})}{(1 + e^{-z})^2}$$

$$= \frac{0 + e^{-z} \frac{\partial}{\partial z} (-z)}{(1 + e^{-z})^2}$$

$$= \frac{e^{-z} \frac{\partial}{\partial z} (-z)}{(1 + e^{-z})^2}$$

$$= \frac{e^{-z}}{(1 + e^{-z})^2}$$

$$= \frac{1}{(1 + e^{-z})} \frac{(e^{-z} + 1 - 1)}{1 + e^{-z}}$$

$$= \frac{1}{1 + e^{-z}} \left( \frac{(1 + e^{-z})}{1 + e^{-z}} - \frac{1}{1 + e^{-z}} \right)$$

$$= \sigma(z) (1 - \sigma(z))$$

$$= a (1 - a)$$

$$dz = (a - y) / \cancel{a(1-a)} \times \cancel{a(1-a)}$$

$$= a - y$$

$$= a - y$$

$$\begin{aligned} \partial w &= \frac{\partial \mathcal{L}(a, y)}{\partial w} \\ &= \frac{\partial \mathcal{L}(a, y)}{\partial z} \times \frac{\partial z}{\partial w} \end{aligned}$$

$$\begin{aligned} \frac{\partial z}{\partial w} &= \frac{\partial (w^T x + b)}{\partial w} \\ &= x \end{aligned}$$

$$\begin{aligned} \partial w &= x \frac{\partial (\mathcal{L}(a, y))}{\partial z} \\ &= x \frac{\partial z}{\partial z} \end{aligned}$$

$$\begin{aligned} \partial b &= \frac{\partial \mathcal{L}(a, y)}{\partial b} \\ &= \frac{\partial \mathcal{L}(a, y)}{\partial z} \times \frac{\partial z}{\partial b} \end{aligned}$$

$$\begin{aligned} &= \frac{\partial (w^T x + b)}{\partial b} \\ &= 1 \end{aligned}$$

$$\begin{aligned} \partial b &= \frac{\partial (\mathcal{L}(a, y))}{\partial z} \\ &= \frac{\partial z}{\partial z} \end{aligned}$$

Update weights & bias

$$w := w - \alpha \partial w$$

$$b := b - \alpha \partial b$$