

Assignment 1: Machine learning basics

Jan Scheffczyk - 3242317
Sarah Khan - 3279206
Mahmoud Hashem - 3201329
Mohamed Saleh - 3201337

October 30, 2019

1 Linear Classifier with Logistic Function

Let

$$\begin{aligned} W & \text{ weights } \in \mathbb{R}^D \\ X & \text{ inputs } \in \mathbb{R}^{N \times D} \\ b & \text{ bias } \in \mathbb{R} \\ y & \text{ labels } \in \mathbb{B}^N \\ h(x_i, W, b) &= \sigma(Wx_i + b) \\ \sigma(a) &= \frac{1}{1 + \exp(-a)} \\ k &= Wx_i + b \end{aligned}$$

$$\begin{aligned} \frac{\partial}{\partial w} L(w, b) &= \frac{\partial}{\partial w} \sum_{i=1}^N (y_i - h(x_i, W, b))^2 \\ &= \sum_{i=1}^N 2(y_i - \sigma(k))(-) \frac{\partial}{\partial w} \sigma(k) \\ &= \sum_{i=1}^N 2(y_i - \sigma(k))(-)(\sigma(k))(1 - \sigma(k)) \frac{\partial}{\partial w} k \quad \text{using(1)} \\ &= \sum_{i=1}^N 2(y_i - \sigma(k))(-)(\sigma(k))(1 - \sigma(k)) x_i \end{aligned}$$

$$\begin{aligned}
\frac{\partial}{\partial b} L(w, b) &= \frac{\partial}{\partial b} \sum_{i=1}^N (y_i - h(x_i, W, b))^2 \\
&= \sum_{i=1}^N 2(y_i - \sigma(k)) (-) \frac{\partial}{\partial b} \sigma(k) \\
&= \sum_{i=1}^N 2(y_i - \sigma(k)) (-) (\sigma(k)) (1 - \sigma(k)) \frac{\partial}{\partial b} k \quad \text{using (1)} \\
&= \sum_{i=1}^N 2(y_i - \sigma(k)) (-) (\sigma(k)) (1 - \sigma(k)) (1)
\end{aligned}$$

$$\begin{aligned}
\frac{d}{dx} \sigma(x) &= \frac{d}{dx} \left[\frac{1}{1 + e^{-x}} \right] \\
&= \frac{d}{dx} (1 + e^{-x})^{-1} \\
&= -(1 + e^{-x})^{-2} (-e^{-x}) \\
&= \frac{e^{-x}}{(1 + e^{-x})^2} \\
&= \frac{1}{1 + e^{-x}} \cdot \frac{e^{-x}}{1 + e^{-x}} \\
&= \frac{1}{1 + e^{-x}} \cdot \frac{(1 + e^{-x}) - 1}{1 + e^{-x}} \\
&= \frac{1}{1 + e^{-x}} \cdot \left(\frac{1 + e^{-x}}{1 + e^{-x}} - \frac{1}{1 + e^{-x}} \right) \\
&= \frac{1}{1 + e^{-x}} \cdot \left(1 - \frac{1}{1 + e^{-x}} \right) \\
&= \sigma(x) \cdot (1 - \sigma(x))
\end{aligned} \tag{1}$$