

Machine Intelligence 1 - Exercise 3: Multilayer Perceptrons and Backpropagation Algorithm

Liu, Zhiwei
387571

Moon, Chulhyun
392865

Wenzel, Daniel
365107

Ozmen, Cengizhan
388011

Pipo, Aiko
390011

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H3.1 Binary Classification (3 Points)

(a)

We can simply derivate the given function $e^{(\alpha)}$:

$$\begin{aligned}\frac{\delta e^{(\alpha)}}{\delta y(\underline{x}^{(\alpha)}; \underline{w})} &= -\frac{y_T^{(\alpha)}}{y(\underline{x}^{(\alpha)}; \underline{w})} - (1 - y_T^{(\alpha)}) \left(-\frac{1}{1 - y(\underline{x}^{(\alpha)}; \underline{w})} \right) \\ &= -\frac{y_T^{(\alpha)}}{y(\underline{x}^{(\alpha)}; \underline{w})} - \frac{1 - y_T^{(\alpha)}}{1 - y(\underline{x}^{(\alpha)}; \underline{w})} \\ &= -\frac{y_T^{(\alpha)}(1 - y(\underline{x}^{(\alpha)}; \underline{w})) - (1 - y_T^{(\alpha)})y(\underline{x}^{(\alpha)}; \underline{w})}{y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w}))} \\ &= -\frac{y_T^{(\alpha)} - y_T^{(\alpha)}y(\underline{x}^{(\alpha)}; \underline{w}) - y(\underline{x}^{(\alpha)}; \underline{w}) + y_T^{(\alpha)}y(\underline{x}^{(\alpha)}; \underline{w})}{y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w}))} \\ &= -\frac{y_T^{(\alpha)} - y(\underline{x}^{(\alpha)}; \underline{w})}{y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w}))} \\ &= \frac{y(\underline{x}^{(\alpha)}; \underline{w}) - y_T^{(\alpha)}}{y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w}))}\end{aligned}$$

(b)

Again, a simple derivation yields:

$$\begin{aligned}f'(h_1^2) &= -\frac{1}{(1 + e^{-h_1^2})^2} \cdot (-e^{-h_1^2}) \\ &= \frac{1}{1 + e^{-h_1^2}} \cdot \frac{e^{-h_1^2}}{1 + e^{-h_1^2}} \\ &= f(h_1^2) \cdot \frac{1 + e^{-h_1^2} - 1}{1 + e^{-h_1^2}} \\ &= f(h_1^2) \left(\frac{1 + e^{-h_1^2}}{1 + e^{-h_1^2}} - \frac{1}{1 + e^{-h_1^2}} \right) \\ &= f(h_1^2)(1 - f(h_1^2))\end{aligned}$$

(c)

First, we look at the derivative of $e^{(\alpha)}$ again:

$$\frac{\delta e^{(\alpha)}}{\delta w_{1j}^{21}} = \frac{y_T^{(\alpha)}}{y(\underline{x}^{(\alpha)}; \underline{w})} \left(\frac{\delta y(\underline{x}^{(\alpha)}; \underline{w})}{\delta w_{1j}^{21}} \right) - \frac{1 - y_T^{(\alpha)}}{1 - y(\underline{x}^{(\alpha)}; \underline{w})} \left(\frac{\delta y(\underline{x}^{(\alpha)}; \underline{w})}{\delta w_{1j}^{21}} \right)$$

We now look at the derivative for $y(\underline{x}^{(\alpha)}; \underline{w}) = S_1^2 = f_1^2(h_1^2)$:

$$\frac{\delta y(\underline{x}^{(\alpha)}; \underline{w})}{\delta w_{1j}^{21}} = \left(\frac{\delta}{\delta h_1^2} f_1^2(h_1^2) \right) \left(\frac{\delta}{\delta w_{1j}^{21}} h_1^2 \right)$$

With the knowledge from (b) we can simplify this to

$$\frac{\delta y(\underline{x}^{(\alpha)}; \underline{w})}{\delta w_{1j}^{21}} = f_1^2(h_1^2)(1 - f_1^2(h_1^2)) \left(\frac{\delta}{\delta w_{1j}^{21}} h_1^2 \right).$$

Transforming back and adding the explicit form for h_1^2 , we obtain

$$\frac{\delta y(\underline{x}^{(\alpha)}; \underline{w})}{\delta w_{1j}^{21}} = y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w})) \left(\frac{\delta}{\delta w_{1j}^{21}} \sum_{i=0}^N w_{1i}^{21} \cdot S_i^1 \right),$$

and thus

$$\frac{\delta y(\underline{x}^{(\alpha)}; \underline{w})}{\delta w_{1j}^{21}} = y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w})) S_j^1.$$

Now, plugging this back into our original derivation we get

$$\begin{aligned} \frac{\delta e^{(\alpha)}}{\delta w_{1j}^{21}} &= \frac{y_T^{(\alpha)}}{y(\underline{x}^{(\alpha)}; \underline{w})} y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w})) S_j^1 - \frac{1 - y_T^{(\alpha)}}{1 - y(\underline{x}^{(\alpha)}; \underline{w})} y(\underline{x}^{(\alpha)}; \underline{w})(1 - y(\underline{x}^{(\alpha)}; \underline{w})) S_j^1 \\ &= \left((1 - y_T^{(\alpha)}) y(\underline{x}^{(\alpha)}; \underline{w}) - y_T^{(\alpha)} (1 - y(\underline{x}^{(\alpha)}; \underline{w})) \right) S_j^1 \\ &= \left(y(\underline{x}^{(\alpha)}; \underline{w}) - y_T^{(\alpha)} y(\underline{x}^{(\alpha)}; \underline{w}) - y_T^{(\alpha)} + y_T^{(\alpha)} y(\underline{x}^{(\alpha)}; \underline{w}) \right) S_j^1 \\ &= \left(y(\underline{x}^{(\alpha)}; \underline{w}) - y_T^{(\alpha)} \right) S_j^1 \end{aligned}$$