

## DEEP LEARNING IN COMPUTER VISION - EXERCISE 3

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### Exercise 3.1

Let  $N, M \in \mathbb{N}$ . Further, let  $f: \mathbb{R}^N \rightarrow \mathbb{R}^M$  and  $g: \mathbb{R}^M \times \mathbb{R}^M \rightarrow \mathbb{R}$  both be differentiable. Let  $h: \mathbb{R}^N \rightarrow \mathbb{R}$  be defined by  $h(x) := g(f(x), f(x))$ .

Then,  $h$  is differentiable with

$$(Dh)(x) = (Dg)(f(x), f(x)) \cdot \begin{pmatrix} (Df)(x) \\ (Df)(x) \end{pmatrix} \in \mathbb{R}^{1 \times N}$$

for all  $x \in \mathbb{R}^N$ .

*Proof.*

For any  $x \in \mathbb{R}^N$  the matrix  $(Df)(x)$  is of size  $M \times N$  and for all  $x \in \mathbb{R}^{2M} \cong \mathbb{R}^M \times \mathbb{R}^M$  the matrix  $(Dg)(x)$  is of size  $1 \times 2M$ .

We define the functions

$$\begin{aligned} h_1(x) &:= (f(x), 0) \in \mathbb{R}^{1 \times 2M} \\ h_2(x) &:= (0, f(x)) \in \mathbb{R}^{1 \times 2M} \\ h_3(x) &:= h_1(x) + h_2(x) \in \mathbb{R}^{1 \times 2M} \end{aligned}$$

for  $x \in \mathbb{R}^N$ , such that  $h(x) = g(h_3(x))$ . Obviously,  $h_1$  and  $h_2$  (and therefore,  $h_3$  as well) are differentiable with derivatives

$$\begin{aligned} (Dh_1)(x) &= \begin{pmatrix} \nabla f_1(x)^\top \\ \vdots \\ \nabla f_M(x)^\top \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} (Df)(x) \\ 0 \end{pmatrix} \in \mathbb{R}^{2M \times N} \\ (Dh_2)(x) &= \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \nabla f_1(x)^\top \\ \vdots \\ \nabla f_M(x)^\top \end{pmatrix} = \begin{pmatrix} 0 \\ (Df)(x) \end{pmatrix} \in \mathbb{R}^{2M \times N} \end{aligned}$$

for  $x \in \mathbb{R}^N$ .

In particular,  $h = g \circ h_3$  is differentiable by the chain rule and we obtain, using the

linearity of the derivative,

$$\begin{aligned}
(Dh)(x) &= (Dg)(h_3(x)) \cdot (Dh_3)(x) \\
&= (Dg)(f(x), f(x)) \cdot [(Dh_1)(x) + (Dh_2)(x)] \\
&= (Dg)(f(x), f(x)) \cdot \begin{pmatrix} \nabla f_1(x)^\top \\ \vdots \\ \nabla f_M(x)^\top \\ \nabla f_1(x)^\top \\ \vdots \\ \nabla f_M(x)^\top \end{pmatrix} \\
&= (Dg)(f(x), f(x)) \cdot \begin{pmatrix} (Df)(x) \\ (Df)(x) \end{pmatrix} \in \mathbb{R}^{1 \times N}
\end{aligned}$$

for  $x \in \mathbb{R}^N$ . □

### Exercise 3.2

Let

$$f(x, w) := \sum_{i,j=0}^2 w_{1,i}^{2,0} w_{0,j}^{1,i} x_j$$

and

$$E(w) = \frac{1}{L} \sum_{l=1}^L (d^l - f(x^l, w))^2,$$

where  $x^l$  denotes the  $l$ -th sample and  $d^l$  denotes the  $l$ -th label.

We calculate according to the chain rule

$$\begin{aligned}
\frac{\partial}{\partial w_{1,1}^{2,0}} E(w) &= \frac{2}{L} \sum_{l=1}^L (d^l - f(x^l, w)) \frac{\partial}{\partial w_{1,1}^{2,0}} f(x^l, w) \\
&= \frac{2}{L} \sum_{l=1}^L (d^l - f(x^l, w)) \sum_{j=0}^2 w_{0,j}^{1,1} x_j^l,
\end{aligned}$$

as well as

$$\begin{aligned}
\frac{\partial}{\partial w_{0,1}^{1,0}} E(w) &= \frac{2}{L} \sum_{l=1}^L (d^l - f(x^l, w)) \frac{\partial}{\partial w_{0,1}^{1,0}} f(x^l, w) \\
&= \frac{2}{L} \sum_{l=1}^L (d^l - f(x^l, w)) w_{1,0}^{2,0} x_1^l.
\end{aligned}$$