

AIMS African Masters in Machine Intelligence 2024  
**Exercises: Gradients and Jacobians**

In all these exercises,  $\langle \cdot, \cdot \rangle$  is the standard euclidean inner product  $\langle x, y \rangle = x^T y$

## 1. Gradients

Let  $x \in \mathbb{R}^n$  with  $n \in \mathbb{N}$ .

Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a differentiable function.

The *gradient*  $\nabla f : \mathbb{R}^n \rightarrow \mathbb{R}^n$  can be defined either by its components

$$[\nabla f(x)]_i = \frac{\partial f}{\partial x_i}(x),$$

or as the unique vector in  $\mathbb{R}^n$  such that for small  $h \in \mathbb{R}^n$ ,

$$f(x+h) \underset{\|h\| \rightarrow 0}{=} f(x) + \langle \nabla f(x), h \rangle + o(h).$$

In each of the questions, find  $\nabla f$ , using either one (or both) of the above definitions.

1.  $f(x) = \langle c, x \rangle$ ,  $c \in \mathbb{R}^n$
2.  $f(x) = x^T A x + c^T x + b$   $A \in \mathbb{R}^{n \times n}, c \in \mathbb{R}^n, b \in \mathbb{R}$
3.  $f(x) = \langle a(x), b(x) \rangle$ ,  $a, b : \mathbb{R}^n \rightarrow \mathbb{R}^m$  differentiable functions with Jacobians  $J_a$  and  $J_b$
4. (a)  $f(x) = \sigma(w^T x + b)$ ,  $\sigma : \mathbb{R} \rightarrow \mathbb{R}$  a differentiable function  $c \in \mathbb{R}^n$   
(b) Simplify (a) when  $\sigma$  is one of these well-known activation functions:  $\sigma(x) = x^+$ ,  
 $\sigma(x) = \frac{1}{1+e^{-x}}$
5. (a) Chain rule:  $f = a \circ b$   
 $f(x) = a(b(x))$ ,  $a : \mathbb{R} \rightarrow \mathbb{R}$ , derivative  $a'$ ,  $b : \mathbb{R}^n \rightarrow \mathbb{R}$ , gradient  $\nabla b$   
(b) what is expression (a) when  $a(x) = \log(1 + e^{-x})$

## 2. Jacobians

Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a differentiable function.

The *Jacobian*  $J_f : \mathbb{R}^n \rightarrow \mathbb{R}^{m \times n}$  can be defined in two equivalent ways:

- The matrix of partial derivatives:  $[J_f(x)]_{ij} = \frac{\partial f_i}{\partial x_j}(x)$
- The matrix such that, for  $x, h \in \mathbb{R}^n$ ,  $f(x+h) \underset{\|h\| \rightarrow 0}{=} f(x) + J_f(x)h + o(h)$

1. What is the relationship between  $\nabla f_i$  – the gradient of the  $i$ -th component of  $f$  – and the Jacobian  $J_f$  ?

In each of the following questions, find  $J_f$ , using either one (or both) of the above definitions.

2. Polar-Cartesian change of coordinates:  $f(r, \theta) = [r \cos \theta, r \sin \theta]^T$
3. Spherical-Cartesian change of coordinates.  $f(r, \theta, \varphi) = [r \cos \theta \sin \varphi, r \sin \theta \sin \varphi, r \cos \varphi]^T$
4.  $f(x) = [g(x_1), \dots, g(x_n)]^T$ ,  $g : \mathbb{R} \rightarrow \mathbb{R}$  differentiable function.
5. *Chain rule*:  $f = a \circ b$ ,  $a, b : \mathbb{R}^n \rightarrow \mathbb{R}^m$  differentiable functions with Jacobians  $J_a$  and  $J_b$

*Hint: Use the Taylor expansion of  $b$ , then the Taylor expansion of  $a$  with  $h' := J_b h$*

6. (a)  $f(x) = \sigma(Wx + b)$ ,  $W \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m, \sigma : \mathbb{R} \rightarrow \mathbb{R}$  an elementwise function with derivative  $\sigma'$   
(b) *higher dimensions*: What is the dimension of  $\nabla_W f$  ?