

# Lecture Notes: From Differentials to Jacobians

## Overview

We shall trace the passage from the simplest differential of a function of one variable all the way to the Jacobian matrix of a vector field, using the unifying language of differential forms and matrix notation:

$$\underbrace{f}_{\Omega^0} \xrightarrow{d} \underbrace{df}_{\Omega^1} \longleftrightarrow \underbrace{\nabla f}_{\substack{\text{gradient} \\ \text{vector field}}} \longrightarrow \underbrace{\mathbf{F}}_{(\Omega^0)^m} \xrightarrow{d} \underbrace{d\mathbf{F}}_{\Omega^1 \otimes \mathbb{R}^m} \longleftrightarrow \underbrace{D\mathbf{F}}_{\substack{\text{Jacobian} \\ \text{matrix}}}.$$

Each step is an instance of the exterior derivative  $d$  or its reinterpretation under the standard Euclidean metric.

## 1 Differentials of Scalar Functions

### 1.1 0-Forms and 1-Forms

**Definition 1** (Spaces of Forms). •  $\Omega^0(\mathbb{R}^n)$  is the space of smooth functions  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  (also called 0-forms).

- $\Omega^1(\mathbb{R}^n)$  is the space of smooth 1-forms  $\omega = \sum_{i=1}^n g_i(x) dx^i$ .

### 1.2 The Exterior Derivative $d$

**Definition 2** (Exterior Derivative on Functions). For  $f \in \Omega^0(\mathbb{R}^n)$ , the differential  $df$  is the 1-form

$$df = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x) dx^i \in \Omega^1(\mathbb{R}^n).$$

**Example 1** (One-Variable Case). If  $n = 1$  and  $f(x) \in \Omega^0(\mathbb{R})$ , then

$$df = f'(x) dx,$$

and the Fundamental Theorem of Calculus is  $\int_a^b df = f(b) - f(a)$ .

### 1.3 Gradient as Metric Dual

Endow  $\mathbb{R}^n$  with the standard dot-product. Then each 1-form  $\omega = \sum_i g_i dx^i$  corresponds uniquely to a vector field  $\sum_i g_i \frac{\partial}{\partial x_i}$ . In particular, the 1-form  $df$  corresponds to the *gradient* vector field

$$\nabla f(x) = \begin{pmatrix} \frac{\partial f}{\partial x_1}(x) \\ \frac{\partial f}{\partial x_2}(x) \\ \vdots \\ \frac{\partial f}{\partial x_n}(x) \end{pmatrix} \in \mathbb{R}^n.$$

Thus we have the identification

$$df \longleftrightarrow \nabla f.$$

## 2 Differentials of Vector-Valued Functions

### 2.1 Vector Fields as $(\Omega^0)^m$

A smooth vector field  $\mathbf{F} : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is an  $m$ -tuple of scalar fields,

$$\mathbf{F}(x) = \begin{pmatrix} F_1(x) \\ F_2(x) \\ \vdots \\ F_m(x) \end{pmatrix} \in (\Omega^0(\mathbb{R}^n))^m.$$

### 2.2 Applying $d$ Componentwise

**Definition 3** (Differential of a Vector Field). *The exterior derivative  $d$  acts on each component  $F_i$ , producing a column of 1-forms:*

$$d\mathbf{F} = \begin{pmatrix} dF_1 \\ dF_2 \\ \vdots \\ dF_m \end{pmatrix} \in \Omega^1(\mathbb{R}^n) \otimes \mathbb{R}^m.$$

*Explicitly,*

$$dF_i = \sum_{j=1}^n \frac{\partial F_i}{\partial x_j}(x) dx^j.$$

## 2.3 Jacobian Matrix

Choosing the basis  $\{dx^1, \dots, dx^n\}$  for  $\Omega^1(\mathbb{R}^n)$  identifies each  $dF_i$  with the row vector  $(\partial_1 F_i, \dots, \partial_n F_i)$ . Stacking these rows yields the *Jacobian matrix*:

$$D\mathbf{F}(x) = \begin{pmatrix} \frac{\partial F_1}{\partial x_1}(x) & \frac{\partial F_1}{\partial x_2}(x) & \cdots & \frac{\partial F_1}{\partial x_n}(x) \\ \frac{\partial F_2}{\partial x_1}(x) & \frac{\partial F_2}{\partial x_2}(x) & \cdots & \frac{\partial F_2}{\partial x_n}(x) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial F_m}{\partial x_1}(x) & \frac{\partial F_m}{\partial x_2}(x) & \cdots & \frac{\partial F_m}{\partial x_n}(x) \end{pmatrix}.$$

**Remark 1** (Linear Approximation). *For a small increment  $h \in \mathbb{R}^n$ ,*

$$\mathbf{F}(x + h) = \mathbf{F}(x) + D\mathbf{F}(x)h + o(\|h\|),$$

*so  $D\mathbf{F}(x)$  is the total derivative (the best linear approximation) of  $\mathbf{F}$  at  $x$ .*

## Putting It All Together

We summarize the full chain of ideas:

$$\underbrace{f}_{\Omega^0} \xrightarrow{d} \underbrace{df}_{\Omega^1} \longleftrightarrow \underbrace{\nabla f}_{\substack{\text{gradient} \\ \text{vector field}}} \longrightarrow \underbrace{\mathbf{F}}_{(\Omega^0)^m} \xrightarrow{d} \underbrace{d\mathbf{F}}_{\Omega^1 \otimes \mathbb{R}^m} \longleftrightarrow \underbrace{D\mathbf{F}}_{\substack{\text{Jacobian} \\ \text{matrix}}}.$$

**Key Takeaway:** The single operator  $d$  applied to 0-forms produces 1-forms. Under the Euclidean metric we identify those 1-forms with gradients (vector fields). When  $d$  is applied to each component of a vector field, it produces exactly the rows of the Jacobian matrix, which encodes the full linearization of the vector-valued function.

# When Is a Vector Field a Gradient?

Let  $U \subset \mathbb{R}^n$  be a region (ideally simply-connected), and let

$$\mathbf{F} : U \longrightarrow \mathbb{R}^n, \quad \mathbf{F}(x) = (F_1(x), \dots, F_n(x)).$$

Define the associated 1-form

$$\alpha = F_1 dx^1 + \dots + F_n dx^n \in \Omega^1(U).$$

We seek a scalar potential  $f$  with

$$\mathbf{F} = \nabla f \iff \alpha = df.$$

## Closed vs. Exact Forms

**Definition 4.** A 1-form  $\alpha$  is closed if  $d\alpha = 0$ , and exact if there exists  $f$  with  $\alpha = df$ .

On any domain  $U \subset \mathbb{R}^n$ ,

$$d\alpha = 0 \iff \frac{\partial F_i}{\partial x_j} = \frac{\partial F_j}{\partial x_i} \quad \text{for all } i, j.$$

In vector-calculus language (for  $n = 3$ ),  $d\alpha = 0$  is exactly  $\nabla \times \mathbf{F} = 0$ .

## Poincaré Lemma (Simply-Connected Case)

If  $U$  is simply connected, then

$$(d\alpha = 0) \implies (\alpha \text{ is exact}).$$

Hence on such a  $U$ , the condition

$$\frac{\partial F_i}{\partial x_j} = \frac{\partial F_j}{\partial x_i} \quad \forall i, j$$

is *necessary and sufficient* for the existence of a scalar potential  $f$ . In that case one recovers

$$f(x) = \int_{x_0}^x \alpha = \int_{x_0}^x \mathbf{F} \cdot d\mathbf{r},$$

and indeed  $\nabla f = \mathbf{F}$ .

## Summary

$$\mathbf{F} = \nabla f \iff d\alpha = 0 \quad (\text{on a simply-connected domain}).$$

Equivalently, in coordinates,

$$\frac{\partial F_i}{\partial x_j} = \frac{\partial F_j}{\partial x_i} \quad \forall i, j.$$

This integrability condition ensures the tight connection between the differential  $df$  and the vector field  $\mathbf{F}$ .