

Analysis 2, Chapter 5

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1 Differentiability

1.1 Definition of differentiability and the differential

Recall the definition of differentiability in $\mathbb{R} \rightarrow \mathbb{R}$:

Definition 1. Let $\Omega \subset \mathbb{R}$ *open*, then $f : \Omega \rightarrow \mathbb{R}$ is called **differentiable at a** if

$$L = \lim_{x \rightarrow a} \frac{f(x) - f(a)}{x - a}$$

exists in \mathbb{R} .

In the scalar case, L is simply a scalar. It is also a **linear map** $L : \mathbb{R} \rightarrow \mathbb{R}$ (namely, scalar multiplication with L). This concept is necessary to generalize differentiation to arbitrary **normed vector spaces**. We will discuss it for \mathbb{R}^N :

Definition 2. Let $\Omega \subset \mathbb{R}^M$ *open*, $f : \Omega \rightarrow \mathbb{R}^N$ is called **differentiable at a** if there is an $L \in \mathcal{L}(\mathbb{R}^M, \mathbb{R}^N)$ such that:

$$\lim_{x \rightarrow a} \frac{f(x) - (f(a) + L(x - a))}{|x - a|} = 0$$

Notation 1. In Landau's notation, we say $f = g + o(h)$ for functions $f, g, h : \mathbb{R} \rightarrow \mathbb{R}$, if

$$\lim_{t \rightarrow 0} \frac{f(t) - g(t)}{h(t)} = 0$$

Using this notation, **differentiability writes as**

$$f(x) = f(a) + L(x - a) + o(|x - a|)$$

Notation 2. We denote the **punctured ball** $B_r^0(x)$ around $x \in X$ with radius $r > 0$ to be the set

$$B_r^0(x) = \{y \in X : 0 < d(x, y) < r\}$$

Proposition 1. If L exists, it is **unique**.

Proof. Assume for a contradiction that there are two different linear maps, T and L . This implies that there is an $\epsilon > 0$ and $(x_n)_{n \in \mathbb{N}}$ is a sequence such that

$$\forall n \in \mathbb{N} : x_n \in B_{\frac{1}{n}}^0(a) \text{ yet } |L(x_n - a) - T(x_n - a)| \geq \epsilon$$

Clearly, $x_n \rightarrow a$ as $n \rightarrow \infty$. Moreover, it implies:

$$\forall n \in \mathbb{N} : \frac{|L(x_n - a) - T(x_n - a)|}{|x_n - a|} \geq \epsilon n$$

Now, since

$$\left| \frac{f(x_n) - f(a) - L(x_n - a)}{|x_n - a|} - \frac{f(x_n) - f(a) - T(x_n - a)}{|x_n - a|} \right| = \left| \frac{|L(x_n - a) - T(x_n - a)|}{|x_n - a|} \right|$$

It follows that this sequence is unbounded. On the other hand, by differentiability, both $\frac{f(x_n) - f(a) - L(x_n - a)}{|x_n - a|}$ and $\frac{f(x_n) - f(a) - T(x_n - a)}{|x_n - a|}$ converge to 0, so their sum sequence should converge and thus be bounded. A contradiction. \square

Definition 3. We call the unique linear map the **differential of f at a** . Notation: $D_a f = L$

Proposition 2. If $f : \Omega \rightarrow \mathbb{R}^N$ is differentiable at x , then there is an $M > 0$ and $\delta > 0$ such that

$$\forall y \in B_\delta(x) : |f(x) - f(y)| \leq M|x - y|$$

In particular, f is **Lipschitz continuous** (\implies **uniformly continuous**) in an **open neighbourhood** $B_\delta(a)$.

Proposition 3. For $\Omega \subset \mathbb{R}^N$ **open** and $a \in \Omega$, $\lambda \in \mathbb{R}$ and $f, g : \Omega \rightarrow \mathbb{R}$, **differentiable at a** , we have:

- $f + \lambda g$ is **differentiable** at a and $D_a(f + \lambda g) = D_a f + \lambda D_a g$
- $f \cdot g$ is **differentiable** at a and $D_a(f \cdot g) = f \cdot D_a g + g \cdot D_a f$

Note that here, the dimension of the target space is $N = 1$.

Proof. • These equalities follow because we can interchange limits with linear combinations and products, provided the individual terms form convergent sequences:

$$\begin{aligned} \lim_{x \rightarrow a} \frac{f(x) + \lambda g(x) - f(a) - \lambda g(a) - (D_a f)(x - a) - \lambda(D_a g)(x - a)}{|x - a|} &= \\ \lim_{x \rightarrow a} \frac{f(x) - f(a) - (D_a f)(x - a)}{|x - a|} + \lambda \cdot \lim_{x \rightarrow a} \frac{g(x) - g(a) - (D_a g)(x - a)}{|x - a|} &= \\ 0 + \lambda \cdot 0 &= 0 \end{aligned}$$

- Here, we not only need the lemma that we can take linear combinations of limiting sequences, but also that f and g are continuous (Proposition 2). Also, assume without loss of gne

$$\begin{aligned}
& \frac{f(x)g(x) - f(a)g(a) - g(a)(D_a f)(x-a) - f(a)(D_a g)(x-a)}{|x-a|} = \\
& \frac{f(x)g(x) + f(x)g(a) - f(x)g(a) - f(a)g(a) - f(a)(D_a f)(x-a) - g(a)(D_a g)(x-a)}{|x-a|} = \\
& \frac{f(x)g(x) + f(x)g(a) - f(x)g(a) - f(a)g(a) - f(a)(D_a f)(x-a) - g(a)(D_a g)(x-a)}{|x-a|} = \\
& \frac{f(x)g(x) - f(x)g(a) - f(x)(D_a f)(x-a)}{|x-a|} - g(a) \frac{f(x) - f(a) - (D_a f)(x-a)}{|x-a|}
\end{aligned}$$

By differentiability of f ,

$$\lim_{x \rightarrow a} \frac{f(x) - f(a) - (D_a f)(x-a)}{|x-a|} = 0$$

So it suffices to show, that for

$$\Delta(r) := \sup_{x \in B_r^0(a)} \frac{|f(x)g(x) - f(x)g(a) - f(x)(D_a f)(x-a)|}{|x-a|}, \text{ we have } \lim_{r \rightarrow 0} \Delta(r) = 0$$

With Proposition 2, we can already say that for $r < \delta$, it holds

$$|f(x) - f(a)| \leq M|x-a|, \text{ therefore}$$

$$\Delta(r) \leq \sup_{x \in B_r^0(a)} \left(\frac{|f(x)g(x) - f(x)g(a) - f(x)(D_a f)(x-a)|}{|x-a|} \right) + \sup_{x \in B_r^0(a)} (M|g(x) - g(a)|)$$

□

1.2 Chain Rule

Proposition 4. *Let $\Omega \subset \mathbb{R}^N$ and $U \subset \mathbb{R}^k$ be **open**, $f : \Omega \rightarrow \mathbb{R}^M$ **differentiable at** $a \in \Omega$ and $\varphi : U \rightarrow \mathbb{R}^k$ **differentiable at** $f(a) \in U$. Then $\varphi \circ f : \Omega \rightarrow \mathbb{R}^k$ is differentiable at a and*

$$D_a(\varphi \circ f) = D_{f(a)}\varphi \circ D_a f$$

Note that the right-hand-side of the equality is indeed a linear map, as it is the composition of two linear maps.

Remark 1. *This means that $\forall x \in \Omega$,*

$$\varphi \circ f(x) = \varphi \circ f(a) + (D_{f(a)}\varphi \circ D_a f)(x-a) = o(|x-a|)$$

1.3 Gradient, Partial Derivatives

Notation 3. For the unfamiliar reader, if V is a K -vector space we will make use the **dual vector space**, denoted as $V^* = \mathcal{L}(V, K)$.

Definition 4. For a linear map $L \in V^*$, there is a **unique** $w \in V$ such that

$$L(v) = \langle w, v \rangle$$

If the linear map is $D_x f$, where $f : \Omega \rightarrow \mathbb{R}$ is differentiable, we call this w the **gradient** and denote it as $\nabla f(x)$

Remark 2. By linearity of $D_x f$ in f , and bilinearity of the inner product, we also have this for $\nabla f(x)$ as a vector:

$$\begin{aligned} \nabla(\lambda f + g)(x) &= \lambda \nabla f(x) + \nabla g(x) \\ \nabla(f \cdot g)(x) &= f(x) \cdot \nabla g(x) + g(x) \cdot \nabla f(x) \end{aligned}$$

Definition 5. For $f : \Omega \rightarrow \mathbb{R}$, we say that f has directional derivatives at $x \in \Omega$ in the direction v if

$$\lim_{t \rightarrow 0} \frac{f(x + tv) - f(x)}{t}$$

exists and is finite. We denote it $\partial_v f(x)$

Remark 3. Like with **linear continuity**, the existence of directional derivatives is a statement about the behaviour of functions **along straight lines**. In particular, the existence of partial derivatives of f at x is not at all sufficient for differentiability at x , because f can still behave badly along arbitrary curves. Consider:

$$f : \mathbb{R}^2 \rightarrow \mathbb{R},$$

$$f(x, y) = \begin{cases} \frac{x|y|}{\sqrt{x^2+y^2}} & (x, y) \neq 0 \\ 0 & (x, y) = 0 \end{cases}$$

It has directional derivatives at the origin along every line $\{tv : t \in \mathbb{R}\}$, yet $v \mapsto \partial_v f(0, 0)$ is not linear, while this should be the case according to Proposition 9.

Proposition 5. Lagrange's Mean Value Theorem

Let $f : \Omega \rightarrow \mathbb{R}$ be a function and $x, y \in \Omega$. Let $v = y - x$ and assume $S = \{x + tv : t \in [0, 1]\} \subset \Omega$, and that f **has directional derivatives** (does **not** need to be **differentiable**) at all the points of S . Then:

$$f(y) - f(x) = \partial_v f(z)$$

or also

$$f(y) - f(x) = \langle \nabla f(z), y - x \rangle$$

for some $z \in S$

A more general form of the theorem requires us to define a $C^1(\Omega)$ curve:

Definition 6. A function $f : \Omega \rightarrow \mathbb{R}^M$ is called **of class** $C^r(\Omega)$ if it has partial derivatives up to and including order r in directions of all basis vectors e_1, \dots, e_N at all $x \in \Omega$ and these partial derivatives $x \mapsto \partial^\alpha f(x)$ are continuous with respect to x .

Remark 4. We then often denote $\partial_{e_i} f$ simply as ∂_i . (Like in quantum mechanics, where one often just denotes the n -th eigenstate ψ_n as $|n\rangle$).

Proposition 6. $f \in C^1(\Omega)$ and $\gamma : [0, 1] \rightarrow \Omega$ s.t. $\gamma_i \in C^1([0, 1])$. Then

$$f(y) - f(x) = \int_0^1 (D_{\gamma(t)} f)(\gamma'(t)) dt$$

As a corollary we have the estimate for **straight line segments** $S = x + [0, 1]v$:

Proposition 7.

$$|f(y) - f(x)| \leq |x - y| \sup_{z \in S} |\nabla f(z)|$$

Proposition 8. Let $\Omega \subset \mathbb{R}^M$ be open and $f : \Omega \rightarrow \mathbb{R}^M$, where we write $f = (f_1, \dots, f_M)$.

Then f is **differentiable at** $x \in \Omega$ if and only if each of its components $f_i : \Omega \rightarrow \mathbb{R}$ is differentiable in $x \in \Omega$.

Now is the time to investigate the relation between differentiability and the **existence** and **regularity** of partial derivatives. First, we discuss what differentiability implies for the partial derivatives, and then we discuss that continuous partial derivatives imply differentiability.

Proposition 9. Let $f : \Omega \rightarrow \mathbb{R}$, where $\Omega \subset \mathbb{R}^N$ is open. If f is **differentiable** at $x \in \Omega$, then

- (i) f has **directional derivatives** at x in every direction $v \in \mathbb{R}^N \setminus \{0\}$
- (ii) $\partial_v f(x) = (D_x f)(v)$, or equivalently (by definition of $\nabla f(x)$ through identification by duality)
- (iii) $\partial_v f(x) = \langle \nabla f(x), v \rangle$
- (iv) $v \mapsto \partial_v f(x)$ is **linear**.

In particular, we can conclude that if $v \mapsto \partial_v f(x)$ is **not linear** at some $x \in \Omega$, then f is **not differentiable** at x .

Remark 5. However, existence of partial derivatives in all directions v and their linearity in v is not sufficient either: consider the function f defined as:

$$f(x, y) = \begin{cases} 1 & \text{if } y = x^2 \neq 0 \\ 0 & \text{else} \end{cases}$$

Then f has directional derivatives in all directions at the origin, since

$$\frac{f(tv_1, tv_2) - f(0, 0)}{t} = \frac{0 - 0}{t} = 0, \quad \forall t \neq 0$$

And in particular, the directional derivatives are all 0, hence linear in v . Yet f is not even continuous at the origin!

In general, we have the strict inclusion

partial derivatives \supset linear partial derivatives \supset differentiable \supset of class C^1

The final \supset is called the **Total Differential theorem**

Theorem 1. Total differential theorem

Let $f : \Omega \rightarrow \mathbb{R}$ be **open** and let $f \in C^1(\Omega)$. Then, f is **differentiable** at each point of Ω and moreover, we have:

$$\forall z \in \Omega : \lim_{x \neq y \rightarrow z} \frac{f(y) - f(x) - D_z f(y - x)}{|x - y|} = 0$$

Note that we could also have used the **gradient** notation, which is equivalent by definition:

$$\forall z \in \Omega : \lim_{x \neq y \rightarrow z} \frac{f(y) - f(x) - \langle \nabla f(z), (y - x) \rangle}{|x - y|} = 0$$

Moreover, the gradient **can be directly computed** as

$$\nabla f(x) = (\partial_1 f(x), \dots, \partial_N f(x))$$

So this theorem is really about showing the following:

$$\forall z \in \Omega : \lim_{x \neq y \rightarrow z} \frac{f(y) - f(x) - \sum_{i=1}^N \partial_i f(z) \cdot (y_i - x_i)}{|x - y|} = 0$$

Definition 7. If $f : \Omega \rightarrow \mathbb{R}^M$ is differentiable at $x \in \Omega$, where $\Omega \subset \mathbb{R}^M$ is open, and we denote e_1, \dots, e_N the standard basis vectors of \mathbb{R}^N , we define the matrix isomorphism

$$\text{mat} : L(\mathbb{R}^N, \mathbb{R}^M) \rightarrow \mathbb{R}^{M \times N}$$

Which identifies a **linear map** $L : \mathbb{R}^N \rightarrow \mathbb{R}^M$ with a **matrix** $M \in \mathbb{R}^{M \times N}$ where

$$M_{ij} = \langle e_i, L e_j \rangle$$

With this, we can define the **Jacobi matrix** via mat as follows:

$$J_x f = \text{mat}(D_x f)$$

Proposition 10.

$$J_x f = \begin{pmatrix} \nabla f_1(x)^T \\ \nabla f_2(x)^T \\ \dots \\ \nabla f_N(x)^T \end{pmatrix}$$

Definition 8. for $a \in \mathbb{R}^N$ and $b \in \mathbb{R}^N$, define the **tensor product** (here: simply matrix outer product) $a \otimes b$ as the $M \times N$ matrix:

$$(a \otimes b)_{ij} = a_i b_j$$

Remark 6. In general, a **rank** (r, s) **tensor** is a **multilinear form** $T : (V^*)^r \times V^s \rightarrow K$ where V is an K -vector space and V^* is its dual vector space. for S another tensor of rank (t, u) , we define its **tensor product** $T \otimes S$ as the **unique rank** $r + t, s + u$ **tensor** such that:

$$(T \otimes S)(v^1..v^r, w^1..w^t, x_1..x_s, y_1..y_u) = T(v^1..v^r, x_1..x_s)S(w^1..w^t, y_1..y_u)$$

Proposition 11. Product rule for scalar multiplication

For $\varphi : \mathbb{R}^N \rightarrow \mathbb{R}$, $f : \mathbb{R}^N \rightarrow \mathbb{R}^M$ differentiable functions at a point $x \in \mathbb{R}^N$, we have that

$$\varphi \cdot f : \mathbb{R}^N \rightarrow \mathbb{R}^M, \text{ defined as } x \mapsto \varphi(x) \cdot f(x)$$

is differentiable at x and:

$$J_x(\varphi \cdot f) = \nabla \varphi(x) \otimes f + \varphi \cdot (J_x f)$$

1.4 Leibniz' integration theorem

Proposition 12. Leibniz' Integral Rule I

For $\psi : [a, b] \times \Omega \rightarrow \mathbb{R}$ a **continuous** function where $\Omega \subset \mathbb{R}^N$ is **open**:

(i) The function $\phi : \Omega \rightarrow \mathbb{R}$ defined as

$$\psi(x) = \int_a^b \psi(t, x) dt$$

is **continuous**.

(ii) If $v \in \mathbb{R}^N \setminus \{0\}$ and $\partial_v \psi$ is continuous on $[a, b] \times \Omega$, then the **Leibniz' formula** holds:

$$\partial_v \phi(x) = \int_a^b \partial_v \psi(t, x) dt$$

Proposition 13. Leibniz' Integral Rule II

If $a, b \in C^0(\Omega; [c, d])$ and $\psi : [c, d] \times \Omega \rightarrow \mathbb{R}$ is a **continuous** function, where $\Omega \subset \mathbb{R}^N$ is **open**.

(iii) Then $\phi : \Omega \rightarrow \mathbb{R}$, defined through:

$$\phi(x) = \int_{a(x)}^{b(x)} \psi(t, x) dt$$

is **continuous**.

(iv) Moreover, if $a, b \in C^1(\Omega; [c, d])$ and for some $v \in \mathbb{R}^N \setminus \{0\}$, we have that $\partial_v \phi$ is **continuous** on $[c, d] \times \Omega$, then

$$\begin{aligned}\partial_v \phi(x) &= \int_{a(x)}^{b(x)} \partial_v \psi(t, x) dt \\ &\quad + \psi(b(x), x) \partial_v b(x) \\ &\quad - \psi(a(x), x) \partial_v a(x)\end{aligned}$$

Proof. The proof is for the **Riemann integral**. There is a similar theorem for the Lebesgue integral, but it requires a definition limiting theorems for the Lebesgue integral, which will follow in Chapter 11, 12 & 13.

(i) Let $x_n \rightarrow x$, for $(x_n)_{n \in \mathbb{N}}$ in Ω and fix $\epsilon > 0$. We need to prove there is a $N \in \mathbb{N}$ with, up to a constant factor,

$$\forall n \geq N : |\phi(x_n) - \phi(x)| < \epsilon$$

Since Ω is open, find a $R > 0$ with $\overline{B_R(x)} \subset \Omega$ and conclude that ψ is uniformly continuous on $[a, b] \times \overline{B_R(x)}$, by compactness of this set. This means that there exists a $\delta > 0$ with:

$$\forall u, v \in [a, b], \forall y, z \in \overline{B_R(x)} : |\psi(u, y) - \psi(v, z)| < \epsilon$$

Therefore, let $N \in \mathbb{N}$ be such that $\forall n \geq N : |x_n - x| < \delta$. Then, for all $n \geq N$:

$$\begin{aligned}|\phi(x_n) - \phi(x)| &= \left| \int_a^b [\psi(t, x_n) - \psi(t, x)] dt \right| \leq \int_a^b |\psi(t, x_n) - \psi(t, x)| dt \\ &\leq |a - b| \cdot \sup_{t \in [a, b], y \in B_\delta(x)} |\psi(t, y) - \psi(t, x)| \\ &< \delta |a - b|\end{aligned}$$

(ii) We need to show

$$\lim_{h \rightarrow 0} \frac{\phi(x) - \phi(x + hv)}{h} = \int_a^b \partial_v \psi(t, x) dt$$

By openness of Ω , for $|h|$ sufficiently small, we have $x + hv \in \Omega$, so we reason with that. On the compact neighbourhood $[a, b] \times \overline{B_r(x)}$, we have that $\partial_v \phi(t, v)$ is **uniformly continuous**, which will be useful: for all $\epsilon > 0$ there is a $\delta > 0$ with

$$\begin{aligned}\forall s, t \in [a, b], \forall h, l \text{ such that } x + hv, x + lv \in \Omega : \\ |s - t| < \delta, |h - l| < \frac{\delta}{|v|} \implies |\phi(t, x + hv) - \phi(s, x + lv)| < \epsilon\end{aligned}$$

$$\frac{\phi(x + hv) - \phi(x)}{h} = \frac{1}{h} \int_a^b [\psi(t, x + hv) - \psi(t, x)] dt$$

By Lagrange's Mean Value Theorem, applied to $l \mapsto \phi(t, x + hv)$, $[0, 1] \rightarrow \mathbb{R}$, there exists, for each $t \in [a, b]$, h as above, a $\lambda_t \in [0, 1]$ such that

$$\psi(t, x + hv) - \psi(t, x) = h \cdot (\partial_v \psi)(t, x + \lambda_t hv)$$

Therefore, we rewrite:

$$\begin{aligned} |\psi(t, x + hv) - \psi(t, x) - h \partial_v \psi(t, x)| &= |h \cdot (\partial_v \psi)(t, x + \lambda_t hv) - h \cdot \partial_v \psi(t, x)| \\ &= |h| |\partial_v \psi(t, x + \lambda_t hv) - \partial_v \psi(t, x)| \leq \epsilon |h| \end{aligned}$$

Namely, the latter equality holds for sufficiently small $|h| < \delta$. This is sufficient to derive the required approximation: for $|h| < \delta$, we have:

$$\begin{aligned} \left| \frac{\phi(x + hv) - \phi(x)}{h} - \int_a^b \partial_v \phi(t, x) dt \right| &= \left| \frac{1}{h} \int_a^b [\psi(t, x + hv) - \psi(t, x) - h \partial_v \phi(t, x)] dt \right| \\ &\leq \frac{1}{|h|} |a - b| \sup_{t \in [a, b], |h| < \delta} [\psi(t, x + hv) - \psi(t, x) - h \partial_v \phi(t, x)] \\ &\leq |a - b| \frac{|h|}{|h|} \epsilon = \epsilon |a - b| \end{aligned}$$

This was to be shown. □

1.5 Higher order differentiability and Taylor expansion

Definition 9. $N \in \mathbb{N}$. An N -**multi-index** α is an element of \mathbb{N}^N . For this element, we define:

$$|\alpha| := \sum_{i=1}^N \alpha_i \qquad \alpha! = \prod_{i=1}^k \alpha_i!$$

Moreover, for $x \in \mathbb{R}^N$, we define

$$x^\alpha := \prod_{i=1}^N x_i^{\alpha_i}$$

Definition 10. For $k \in \mathbb{N}$, $\Omega \subset \mathbb{R}^N$ open. $f \in C^k(\Omega, \mathbb{R}^N)$. $\alpha \in \mathbb{N}^N$ with $|\alpha| \leq k$. Then we define

$$\partial^\alpha f(x) := \partial_{x_1}^{\alpha_1} \dots \partial_{x_N}^{\alpha_N} f(x)$$

where we define $\partial_{x_i}^{\alpha_i} g(x) := \partial_{x_i} \dots \partial_{x_i} g(x)$, i.e. α_i times.

Theorem 2. Taylor Series with Peano Remainder

Let $k \in \mathbb{N}$, and $\Omega \subset \mathbb{R}^N$ **open**, and fix $x \in \Omega$. Assume that $\partial^\alpha f(x)$ exists for all $\alpha \in \mathbb{N}^N$ with $|\alpha| \leq k$.

Then fix $x, y \in \Omega$ and assume that the **straight line segment** $S = x + [0, 1](y - x)$ joining them is contained in Ω . Then, the following approximation holds:

$$f(y) = \sum_{\alpha \in \mathbb{N}^N: |\alpha| \leq k-1} \frac{1}{\alpha!} \partial^\alpha f(x) (y-x)^\alpha + o(|y-x|^k)$$

Theorem 3. Taylor Series with Integral Remainder

$k \in \mathbb{N}$, $\Omega \subset \mathbb{R}^N$ **open**, $x \in \Omega$ **fixed**. Assume that $\partial^\alpha f(x)$ exists for all $\alpha \in \mathbb{N}^N$ with $|\alpha| \leq k+1$.

Then fix $x, y \in \Omega$ and assume that the line segment $S = x + [0, 1](y - x)$ joining them is contained in Ω . The following approximation holds:

$$f(y) = \sum_{\alpha \in \mathbb{N}^N: |\alpha| \leq k} \frac{1}{\alpha!} \partial^\alpha f(x) (y-x)^\alpha + R_k(x, y)$$

Where the **remainder term** $R_k(x, y)$ satisfies:

$$R_k(x, y) = \sum_{\alpha \in \mathbb{N}^N: |\alpha|=k+1} \frac{(y-x)^\alpha}{\alpha!} \int_0^1 (k+1)(1-t)^k \partial^\alpha f(x + t(y-x)) dt$$

Remark 7. For $N = 1$, this gives the familiar **Lagrange theorem** with integral remainder:

$$f(y) = \sum_{i=0}^k \frac{1}{i!} f^{(i)}(x) (y-x)^i + R_k(x, y)$$

where

$$\begin{aligned} R_k(x, y) &= \frac{(y-x)^{k+1}}{(k+1)!} \int_0^1 (k+1)(1-t)^k f^{(k+1)}(x + t(y-x)) dt \\ &= \frac{(y-x)^{k+1}}{(k+1)!} \int_x^y (k+1) \left(1 - \frac{u-x}{y-x}\right)^k f^{(k+1)}(u) \left(\frac{u-x}{y-x}\right) du \\ &= \frac{(y-x)^{k+1}}{(k+1)!} \int_x^y (k+1) \left(\frac{y-u}{y-x}\right)^k \left(\frac{u-x}{y-x}\right) f^{(k+1)}(u) du \\ &= \frac{1}{k!} \int_x^y (y-u)^k \left(\frac{u-x}{y-x}\right) f^{(k+1)}(u) du \end{aligned}$$

Definition 11. If a function $f : \Omega \rightarrow \mathbb{R}$ with $\Omega \subset \mathbb{R}^N$ is of $C^\infty(\Omega)$, then its **Taylor series** $T(f)(a)$ at $a \in \Omega$ is defined as:

$$T(f)(a)(x) := \sum_{k=0}^{\infty} \sum_{\alpha \in \mathbb{N}^N: |\alpha|=k} \frac{1}{\alpha!} \partial^\alpha f(a) (x-a)^\alpha$$

Note: there is no natural order in which to enumerate the terms $(x - a)^\alpha$, and therefore, we adopt no ordering convention, This means that this series is only defined well if it converges **absolutely** (in that case, permutations of the terms do not change the convergence properties); moreover, not every $C^\infty(\Omega)$ -function equals its Taylor series. We call function $f \in C^\infty(\Omega)$ **analytic** at $a \in \Omega$ if that is the case:

$$T(f)(a) = f$$

Example 1. Consider $f : \mathbb{R} \rightarrow \mathbb{R}$, defined through

$$f(x) := \begin{cases} e^{-\frac{1}{x^2}} & x > 0 \\ 0 & x \leq 0 \end{cases} \implies f^{(n)}(x) = \begin{cases} 2 \cdot \left(\sum_{k=3}^{n+2} \frac{(-1)^k}{x^k} \right) e^{-\frac{1}{x^2}} & x > 0 \\ 0 & x < 0 \\ 0 & x = 0 \end{cases}$$

This writes as:

$$f^{(n)}(x) = \begin{cases} 2 \cdot \left(\sum_{k=3}^{n+2} \frac{(-1)^k}{x^k} \right) f(x) & x \neq 0 \\ 0 & x = 0 \end{cases}$$

Which is clearly continuous, also at the origin as $f \rightarrow 0$ faster than any polynomial. However, at $a = 0$, the Taylor-series is 0, which is not equal to f in any open neighbourhood of 0, let alone at \mathbb{R} .

2 Homework

Exercise 1 Consider the function $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ defined as

$$f(x, y) := \begin{cases} \frac{x|y|}{\sqrt{x^2+y^2}} & (x, y) \neq (0, 0) \\ 0 & (x, y) = (0, 0) \end{cases}$$

Prove that:

- (i) f has directional derivative in all directions at the origin;
- (ii) The map

$$v \mapsto \partial_v f(0, 0)$$

is not linear. In particular, prove that f is not differentiable at the origin.

Proof

- (i) let $v = (v_x, v_y) \in \mathbb{R}^2 \setminus \{0\}$. Then

$$\frac{f((0, 0) + t(v_x, v_y)) - f(0, 0)}{t} = \frac{t|v_x v_y|}{t|t|\sqrt{v_x^2 + v_y^2}}$$

And this is a constant for any t , hence goes to $\frac{v_x|v_y|}{\sqrt{v_x^2+v_y^2}}$ as $t \rightarrow 0$. So partial derivatives exist in all directions and are calculated to be:

$$\partial_v f(0,0) = \frac{v_x|v_y|}{\sqrt{v_x^2+v_y^2}}$$

(ii) The map

$$v \mapsto \partial_v f(0,0) = \frac{v_x|v_y|}{\sqrt{v_x^2+v_y^2}}$$

is clearly not linear: $\partial_{(0,1)}f(0,0) = 0$, $\partial_{(1,0)}f(0,0) = 0$ but $(1,1) = (0,1) + (1,0)$, yet $\partial_{(1,1)}f(0,0)1/\sqrt{2} \neq 0+0 = \partial_{(0,1)}f(0,0) + \partial_{(1,0)}f(0,0)$. In particular, we see f is not differentiable at the origin, since then it follows $\partial_v f(0,0) = \langle \nabla f(0,0), v \rangle$, which would be linear in v .

Exercise 2 In this exercise we want to compute the derivative of the determinant. We will do that by seeing the determinant as a **multilinear map of the columns**. Since, recall, the determinant is the unique mapping $L : \mathbb{R}^{N \times N} \rightarrow \mathbb{R}$ that is **alternating**, **N -linear on the columns (rows)**, and has $LId = 1$. We will only make use of the second property, and maybe the third.

(i) Consider an N -linear map $L : (\mathbb{R}^N)^N \rightarrow \mathbb{R}$. Namely, a map such that for all $i \in \{1, \dots, N\}$, all $V = (v_1, \dots, v_N) \in (\mathbb{R}^N)^N$, all $w \in \mathbb{R}^N$, and all $\lambda \in \mathbb{R}$ it holds

$$L(v_1, \dots, v_i + \lambda w, \dots, v_N) = L(v_1, \dots, v_i, \dots, v_N) + \lambda L(v_1, \dots, w, \dots, v_N)$$

. Fix $V = (v_1, \dots, v_N)$ and $W = (w_1, \dots, w_N)$, where $v_i, w_i \in \mathbb{R}^N$ for each $i = 1, \dots, N$. Compute the **directional derivative of L at V in the direction W** . Namely, compute

$$\lim_{t \rightarrow 0} \frac{L(V + tW) - L(V)}{t} = \lim_{t \rightarrow 0} \frac{L(v_1 + tw_1, \dots, v_N + tw_N) - L(v_1, \dots, v_N)}{t}$$

Solution

(i) We will denote the matrix arising when the i -th column of V is replaced by the i -th column of W , i.e. $(v_1, \dots, v_{i-1}, w_i, v_{i+1}, \dots, v_N)$, with $V^i W$. With this, we prove by induction that

$$\partial_W L(V) = \sum_{i=1}^N L(V^i W)$$

Induction Basis: for $N = 1$, L is just a linear map. L not only has partial derivatives, but even a total derivative, as can be seen from the fact that

$$\lim_{x \rightarrow v} \frac{L(x) - L(v) - T(x-v)}{|x-v|} = 0 \text{ trivially holds for } T = L$$

From this, we can conclude $D_v L = L$, and therefore $\partial_w L(v) = Lw = L(V^1 W)$ for $V = (v)$, $W = (w)$.

Induction Hypothesis: suppose that for $n-1$ -linear maps $L : (\mathbb{R}^N)^{n-1} \rightarrow \mathbb{R}$ we have the result that for any two sequences of column vectors $V = (v_1, \dots, v_{n-1})$, $W = (w_1, \dots, w_n)$, the partial derivative $\partial_W L(V)$ exists and equals $\sum_{i=1}^n L(V^i W)$.

Induction Step Then, let $L : (\mathbb{R}^N)^n \rightarrow \mathbb{R}$. Consider, for $t \neq 0$,

$$\frac{L(v_1 + tw_1, \dots, v_n + tw_n) - L(v_1, \dots, v_n)}{t}$$

We use linearity in the N -th component to rewrite this as

$$\begin{aligned} & \frac{L(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}, v_n) + tL(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}, w_n) - L(v_1, \dots, v_n)}{t} \\ &= \frac{L(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}, v_n) - L(v_1, \dots, v_n)}{t} + L(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}, w_n) \end{aligned}$$

Next, define $\tilde{L} : (\mathbb{R}^N)^{n-1} \rightarrow \mathbb{R}$ as $\tilde{L}(x_1, \dots, x_{n-1}) := L(x_1, \dots, x_{n-1}, v_n)$. This is a $n-1$ -linear map, so the induction hypothesis holds for this map; what is more, we can substitute it everywhere where L is given v_n as its n -th argument:

$$\begin{aligned} & \frac{L(v_1 + tw_1, \dots, v_n + tw_n) - L(v_1, \dots, v_n)}{t} \\ &= \frac{\tilde{L}(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}) - \tilde{L}(v_1, \dots, v_{n-1})}{t} + L(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}, w_n) \end{aligned}$$

Now note that in the lower expression, we can take two limits separately: by the induction hypothesis,

$$\lim_{t \rightarrow 0} \frac{\tilde{L}(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}) - \tilde{L}(v_1, \dots, v_{n-1})}{t} = \sum_{i=1}^{n-1} L(V^i W)$$

While, by continuity of linear maps,

$$\lim_{t \rightarrow 0} L(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}, w_n) = L(v_1, \dots, v_{n-1}, w_n) = L(V^n W)$$

Since sums of limits equal limits of sums, we discover that

$$\lim_{t \rightarrow 0} \frac{L(v_1 + tw_1, \dots, v_n + tw_n) - L(v_1, \dots, v_n)}{t} =$$

indeed exists, and equals

$$\begin{aligned} & \lim_{t \rightarrow 0} \frac{\tilde{L}(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}) - \tilde{L}(v_1, \dots, v_{n-1})}{t} \\ &+ \lim_{t \rightarrow 0} L(v_1 + tw_1, \dots, v_{n-1} + tw_{n-1}, w_n) = \\ & \sum_{i=1}^{n-1} L(V^i W) + L(V^n W) = \sum_{i=1}^n L(V^i W) \end{aligned}$$

This completes the induction step and thereby the proof. To repeat, the partial derivative $\partial_W L(V)$ exists and equals $\sum_{i=1}^n L(V^i W)$, for L an N -linear map where $N \geq 1$.

- (ii) We now use that the determinant on $N \times N$ -matrices is N -linear on the columns. We use this to compute $\partial_B \det(A) = \lim_{t \rightarrow 0} \frac{\det(A+tB) - \det(A)}{t}$. This equals, by (i), $\sum_{i=1}^N \det(A^i B)$. This is the first part of the proof.

Next, taking $A = I$, we can write:

$$\lim_{t \rightarrow 0} \frac{\det(I + tB) - \det(I)}{t} = \sum_{i=1}^N \det(I^i B) = \lim_{t \rightarrow 0} \frac{t \sum_{i=1}^N \det(I^i B)}{t},$$

therefore, subtracting the right limit from the left (using additivity of limits) and using $\det(I) = 1$:

$$\lim_{t \rightarrow 0} \frac{\det(I + tB) - 1 - t \sum_{i=1}^N \det(I^i B)}{t} = 0, \text{ therefore, by definition of } o :$$

$$\det(I + tB) = 1 + t \sum_{i=1}^N \det(I^i B) + o(t)$$

So if we can show $\sum_{i=1}^N \det(I^i B) = \text{trace}(B)$, we are done.

Let's calculate $\det(I^i B)$ by using the **Laplace expansion** to the i -th **row**: $\det(M) = \sum_{j=1}^N (-1)^{i+j} M_{ij} \det(\tilde{M}_{ij})$, where \tilde{M}_{ij} is the $(N-1) \times (N-1)$ -matrix arising from M by deletion of the i -th row and j -th column. If we apply this to row i of $I^i B$, we notice that all elements $(I^i B)_{ij} \neq 0$ if and only if $i = j$, in which case the element is $(I^i B)_{ii} = B_{ii}$, and deletion of column i and row i will leave the $(N-1) \times (N-1)$ identity matrix, which has determinant 1.

Hence,

$$\begin{aligned} \det(I^i B) &= \det(M) = \sum_{j=1}^N (-1)^{i+j} (I^i B)_{ij} \det((I^i B)_{ij}) \\ &= (-1)^{i+i} B_{ii} \det(I_{N-1}) = B_{ii} \end{aligned}$$

. Therefore,

$$\sum_{i=1}^N \det(I^i B) = \sum_{i=1}^N B_{ii} = \text{trace}(B)$$

, and we are done.

Exercise 3 Let $f : \Omega \rightarrow \mathbb{R}$ be a function that is differentiable at all $x \in \Omega$. Assume that Ω is an open connected set. Namely, for each $x, y \in \Omega$ it is possible to find a curve

$$\gamma : [0, 1] \rightarrow \Omega$$

with $\gamma \in C^1([0, 1]; \mathbb{R}^N)$, such that $\gamma(0) = x$, and $\gamma(1) = y$. Assume that $df(x) = 0$ for all $x \in \Omega$. Prove that f is constant

Proof Fix arbitrary $x, y \in \Omega$. It suffices to show that $f(x) = f(y)$. To this end, consider a $\gamma \in C^1([0, 1]; \mathbb{R}^N)$ with $\gamma(0) = x$, $\gamma(1) = y$. By Lagrange's Mean Value Theorem, we have:

$$f(y) - f(x) = \int_0^1 \langle \nabla f(\gamma(t)), \gamma'(t) \rangle dt$$

Since $\gamma(t) \in \Omega$ for all $t \in [0, 1]$, and $df(x) = 0$ for all $x \in \Omega$, it follows $\nabla f(\gamma(t)) = 0$ for all $t \in [0, 1]$. Therefore:

$$f(y) - f(x) = \int_0^1 \langle 0, \gamma'(t) \rangle dt = \int_0^1 0 dt = 0$$

Hence $f(y) = f(x)$, as was to be shown.