

CHAPTER

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Topology of n-dimensional R

1.1. The Algebraic structure of \mathbb{R}^n

The space \mathbb{R}^n consists of all n -tuples $x = (x_1, x_2, \dots, x_n)$ of real numbers. That is, \mathbb{R}^n is the cartesian product of n copies of \mathbb{R} , $\mathbb{R}^n = \mathbb{R} \times \mathbb{R} \times \dots \times \mathbb{R}$ (n factors).

The members x of \mathbb{R}^n are called the points of \mathbb{R}^n , They can also be viewed as vectors, and \mathbb{R} as a field of scalars.

We define addition in \mathbb{R}^n as coordinate-wise addition, if $x = (x_1, x_2, \dots, x_n)$ and $y = (y_1, y_2, \dots, y_n)$ then $x + y = (x_1 + y_1, x_2 + y_2, \dots, x_n + y_n)$

also for $\alpha \in \mathbb{R}$, and $x \in \mathbb{R}$ we defined $\alpha x = (\alpha x_1, \alpha x_2, \dots, \alpha x_n)$

we refer to such multiplication of vector by a scalar as a scalar multiplication

With these notations of addition and scalar multiplication, \mathbb{R}^n possess the algebraic structure of vector space. We shall have need for a notion of dot product of two vectors x and y . This is defined by $x \cdot y = x_1 y_1 + x_2 y_2 + \dots + x_n y_n$. where $y = (y_1, y_2, \dots, y_n)$ and $x = (x_1, x_2, \dots, x_n)$. The dot product satisfies certain conditions that we summarize in a theorem

Theorem 1.1.1

Let $x, y, z \in \mathbb{R}^n$ and let $\alpha \in \mathbb{R}$, then

1. $x \cdot y = y \cdot x$
2. $x(y + z) = x \cdot y + x \cdot z$
3. $(\alpha x)y = \alpha(x \cdot y)$

Proof. Easy

□

Remark

Observe that $0_{\mathbb{R}^n}x = 0_{\mathbb{R}}$, but $x \cdot y$ can equal 0 without either x or y being 0. Two vectors x and y for which $x \cdot y = 0$ are said to be orthogonal vectors. Geometrically, two orthogonal vectors in \mathbb{R}^n are perpendicular.

1.2. The Metric structure of \mathbb{R}^n **Definition 1.2.1 Euclidean Norm**

Let $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$. We define the euclidean norm by:

$$\|x\|_2 = \sqrt{\sum_{i=1}^n x_i^2}$$

Note that: $\|x\|_2 \geq 0$, $\forall x \in \mathbb{R}^n$

Definition 1.2.2 Euclidean Distance

For $x, y, z \in \mathbb{R}^n$, The euclidean distance between x and y is defined by:

$$d_2(x, y) = \|x - y\|_2$$

In terms of coordinates, with $x = (x_1, x_2, \dots, x_n)$ and $y = (y_1, y_2, \dots, y_n)$, then:

$$\|x - y\|_2 = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$

Also observe that we can express the norm via the dot product: $\|x\|_2^2 = x \cdot x$

or $\|x\|_2 = \sqrt{x \cdot x}$

Theorem 1.2.1 Cauchy-Schawrtz Inequality

For $x, y \in \mathbb{R}^n$, we have :

$$|x \cdot y| \leq \|x\| \cdot \|y\|$$

Proof. Let $x, y \in \mathbb{R}^n$, for all $t \in \mathbb{R}$, we have $P(t) = (x - ty)(x - ty) \geq 0$, for all $t \in \mathbb{R}$, then we get:

$$\begin{aligned} P(t) &= x \cdot x - 2tx \cdot y + y \cdot yt^2 \\ &= \|x\|^2 - 2txy + t^2\|y\|^2 \end{aligned}$$

Then The function $P(t)$ is quadratic in t , since it is non-negative for all t , it cannot have two distinct roots, hence its discriminant cannot be positive, that is

$$4(xy)^2 - 4\|x\|^2\|y\|^2 \leq 0$$

it follows that

$$|x \cdot y| \leq \|x\| \cdot \|y\|$$

□

Theorem 1.2.2 Norm properties

Let $x, y \in \mathbb{R}^n$, and $\alpha \in \mathbb{R}$, then :

- (i) $\|x\| \geq 0$, and $\|x\| = 0$ if and only if $x = 0$
- (ii) $\|\alpha x\| = |\alpha| \|x\|$
- (iii) $\|x + y\| \leq \|x\| + \|y\|$ (triangle inequality)

Proof. • The proof of (i) is immediate.

- $\forall \alpha \in \mathbb{R}, x \in \mathbb{R}^n$, we have $\|\alpha x\| = \sqrt{\sum_{i=1}^n (\alpha x_i)^2} = \sqrt{\alpha^2 \sum_{i=1}^n x_i^2} = |\alpha| \|x\|$
- To prove (iii) we calculate:

$$\begin{aligned} \|x + y\|^2 &= (x + y)(x + y) = xx + 2xy + xy \\ &\leq \|x\|^2 + 2\|x\|\|y\| + \|y\|^2 \\ &= (\|x\| + \|y\|)^2 \end{aligned}$$

The inequality follows from the Cauchy-Schawrtz inequality

□

1.3. Elementary Topology of n-dimensional R

Let $x_0 \in \mathbb{R}^n$, and let $r > 0$, the set:

$$B(x_0, r) = \{x \in \mathbb{R}^n : \|x - x_0\| < r\}$$

is called the open ball of radius r centered at x_0 , it's also called the r -neighborhood of x_0 . For instance, for $n = 2$, and $x_0 = (x, y) \in \mathbb{R}^2$, the open ball around x_0 with radius $r > 0$ is the open disk,

$$B(x_0, r) = \left\{ (x, y) \in \mathbb{R}^2 : (x - x_0)^2 + (y - y_0)^2 \leq r^2 \right\}$$

and when $n = 1$, and $x_0 \in \mathbb{R}$, $B(x_0, r) = (x_0 - r, x_0 + r)$.

We denote by $\bar{B}(x_0, r) = \{x \in \mathbb{R}^n : \|x - x_0\| \leq r\}$, the set $\bar{B}(x_0, r)$ is called the closed ball around x_0

Definition 1.3.1 Open set

Let $E \subset \mathbb{R}^n$, The set E is called an open set in \mathbb{R}^n if each of it's points has a sufficiently small open ball around it completely contained in E , that is, for each $x \in E$ there exists $r > 0$ such that $B(x, r) \subset E$

Example *The whole space E and the empty set \emptyset are open.*

Example *The open ball is an open set*

Proof. Let $x \in B(x_0, r)$, take $\delta = r - \|x - x_0\| > 0$, we should prove $B(x, \delta) \subset B(x_0, r)$, let $y \in B(x, \delta)$, we have

$$\begin{aligned}\|y - x_0\| &= \|y - x + x - x_0\| \leq \|y - x\| + \|x - x_0\| \\ &< \delta + \|x - x_0\| = r\end{aligned}$$

That is $\|y - x_0\| < r$. Hence, $y \in B(x_0, r)$ □

Definition 1.3.2 The boundary

The boundary of a set $E \subset \mathbb{R}^n$, denoted ∂E , is the set of points $x \in \mathbb{R}^n$ such that every ball around x intersect E and the complement of E , that is for every $\varepsilon > 0$

$$\partial E = \{x \in \mathbb{R}^n : B(x, \varepsilon) \cap E \neq \emptyset, B(x, \varepsilon) \cap E^c \neq \emptyset\}$$

Clearly $\partial E = \partial(E^c)$

Example *The boundary of $B(x_0, r)$ is the sphere $S(x_0, r) = \{x \in \mathbb{R}^n : \|x - x_0\| = r\}$*

Remark *Show that an open set can not contain any of it's boundary points*

Definition 1.3.3

Let $E \subset \mathbb{R}^n$, A point $x_0 \in E$ is an interior point of E if there exist $\delta > 0$ such that $B(x_0, \delta) \subset E$. The set of all interior points of E is called the interior of E , and

denoted by E° .

Note that $E^\circ \subset E$ and E is open if and only if $E^\circ = E$

Example $\mathbb{R}^2, E = (0, 1] \times [0, 2]$

Definition 1.3.4 Closed set

A set $E \subset \mathbb{R}^n$ is said to be closed if its complement E^c in \mathbb{R}^n is open.

Theorem 1.3.1

A set E is closed in \mathbb{R}^n if and only if $\partial E \subset E$

Proof.

$$(\Rightarrow)$$

Suppose that E is closed, then E^c is open, it follows from exercise 1 that $E^c \cap \partial E = \emptyset$. Hence $\partial E \subset E$

$$(\Leftarrow)$$

Conversely, let $x \in E^c$, since $\partial E \subset E$ it follows that x is no a boundary point of E , by the definition, there exist an $r > 0$ such that $B(x, r) \cap E = \emptyset$, that is $B(x, r) \subset E^c$ and so E^c is open. \square

Definition 1.3.5

Let $E \subset \mathbb{R}^n$, we define the closure of E , denoted by \bar{E} , to be the set $\bar{E} = \partial E \cup E$

Remark

- $E \subset \bar{E}$
- E is a closed set $\iff \bar{E} = E$

Example

- by taking $E = \mathbb{R}, \bar{\mathbb{Q}} = \mathbb{R}$, and $\mathbb{Q} = \emptyset$
- by taking $E = \mathbb{Z}, \bar{\mathbb{Z}} = \mathbb{Z}$,
- by taking $E = \mathbb{R}^2, \overline{\mathbb{Q} \times \mathbb{Q}} = \mathbb{R}^2$

Example

1. Consider the set $\mathbb{Z} \subset \mathbb{R}$. we have $\mathbb{Z}^\circ = \emptyset$ and $\partial\mathbb{Z} = \overline{\mathbb{Z}} = \mathbb{Z}$. So \mathbb{Z} is closed.
2. Consider the set $\mathbb{Q} \subset \mathbb{R}$, $\mathbb{Q}^\circ = \emptyset$ and $\overline{\mathbb{Q}} = \partial\mathbb{Q} = \mathbb{R}$

Definition 1.3.6

Let $E \subset \mathbb{R}^n$ and $x_0 \in \mathbb{R}^n$, we call x_0 an accumulation point or limit point of E if every open ball around x_0 contains at least one point $x \in E$ distinct from x_0 , that is for every $r > 0$ we have $(B(x_0, r) - \{x_0\}) \cap E \neq \emptyset$.

If x_0 is not an accumulation point of E is called an isolated point of E .

The next proposition provides a useful characterisation of closeness

Theorem 1.3.2 Closeness Characterisation

A set $E \subset \mathbb{R}^n$ is closed iff E contains all its limit points.

Proof.

$$(\implies)$$

Suppose E is closed and let x_0 be a limit point of E . If $x_0 \notin E$ then $x_0 \in E^c$ which is open. Therefore, there exists $\delta > 0$ such that $B(x_0, \delta) \subset E^c$ then $B(x_0, \delta) \cap E = \emptyset$ which is a contradiction with x_0 being a limit point of E

$$(\Leftarrow)$$

Suppose that E contains all of its limit points. we need to show that E is a closed set. Let $x \in E^c$, x is not a limit point of E then by the definition there exist $\delta > 0$ such that $B(x, \delta) \cap E = \emptyset$. Hence $B(x, \delta) \subset E^c$ so E^c is open which gives us that E is closed \square

1.4. Convergence of sequences of \mathbb{R}^n

We reserve the letter n for the dimension and we use letters i, j, k for the index, that is, we denote by $(x_k)_{k \in \mathbb{N}}$ a sequence of vectors in \mathbb{R}^n and the components vector $x_k = (x_{k,1}, x_{k,2}, \dots, x_{k,n})$. We now define the limit of a sequence in \mathbb{R}^n .

Definition 1.4.1 Convergent Sequence

Let (x_k) be a sequence of vectors in \mathbb{R}^n and $x \in \mathbb{R}^n$. We say that $(x_l)_{l \in \mathbb{N}}$ converges to x if $d(x_k, x) = \|x_k - x\| \rightarrow 0$ as $k \rightarrow \infty$. We write $x_k \rightarrow x$. that is:

$$x_k \rightarrow x \iff \forall \varepsilon > 0, \exists N \in \mathbb{N}, \forall k > N \implies \|x_k - x\| < \varepsilon$$

Example

- Now let $(x_k)_{k \in \mathbb{N}} = \left(\frac{1}{k}, k^2 e^{-k}, k^3\right) \subset \mathbb{R}^3$, is a sequence on \mathbb{R}^3 , this sequence is not convergent.Indeed, Suppose that there exist $x = (x_1, x_2, x_3)$ such that $x_k \rightarrow x$.

$$\|x_k - x\|^2 = \left(\frac{1}{k} - x_1\right)^2 + \left(k^2 e^{-k} - x_2\right)^2 + \left(k^3 - x_3\right)^2 \rightarrow \infty$$

we can see that the sequence is divergent.

- Now consider the sequence $(x_k)_{k \in \mathbb{N}} = \left(\frac{1}{k}, \frac{k+1}{k}\right)$ is convergent.Indeed, we have

$$\|x_k - x\|^2 = \frac{1}{k^2} + \frac{1}{k^2} = \frac{2}{k^2} \rightarrow 0 \quad \text{as } k \rightarrow \infty$$

Definition 1.4.2 Bounded Sequence

Just like in \mathbb{R} , a sequence $(x_n)_{n \in \mathbb{N}} \subset \mathbb{R}^n$ is said to be bounded if there exist $M > 0$ with $\|x_k\| \leq M \quad \forall k \geq 1$, and a sequence which is not bounded is called unbounded.In addition, every convergent sequence is bounded and any unbounded sequence diverges.

Theorem 1.4.1

Let $(x_n)_{n \in \mathbb{N}}$ and $(y_n)_{n \in \mathbb{N}}$ be two sequences in \mathbb{R}^n with $x_k \rightarrow x$ and $y_k \rightarrow y$, where $x, y \in \mathbb{R}^n$, then :

1. for all $\alpha, \beta \in \mathbb{R}$, $\alpha x_k + \beta y_k \rightarrow \alpha x + \beta y$
2. $\|x_k\| \rightarrow \|x\|$
3. $\langle x_k, y_k \rangle \rightarrow \langle x, y \rangle$
4. $x_k \rightarrow x \quad \text{if and only if} \quad x_{k_i} \rightarrow x_i \quad \forall i = 1, \dots, n$

Proof. (i)

$$\|\alpha x_k + \beta y_k - \alpha x - \beta y\| \leq |\alpha| \|x_k - x\| + |\beta| \|y_k - y\|$$

(ii)

$$\||x_k| - |x|\| \leq ||x_k - x||$$

(iii)

$$\begin{aligned} |\langle x_k, y_k \rangle - \langle x, y \rangle| &= |\langle x_k - x, y_k \rangle + \langle x, y_k - y \rangle| \\ &\leq |\langle x_k - x, y_k \rangle| + |\langle x, y_k - y \rangle| \\ &\leq \underbrace{\|x_k - x\| \|y_k\|}_{\text{Bounded}} + \|y_k - y\| \|x\| \rightarrow 0 \end{aligned}$$

(iv)

$$|x_{k_i} - x_i| \leq ||x_k - x||$$

□

By property (iv) of the above theorem, we see that to study the convergence of a sequence in \mathbb{R}^n , it's sufficient to study the convergence of it's components sequences in \mathbb{R} .

Next, we would like to restat the definition of Cuachy sequence in \mathbb{R}^n .

Definition 1.4.3

Let (X, d) be a metric space

1. A sequence $(x_k) \subset \mathbb{R}^n$ is called a Cauchy sequence if :

$$d(x_k, x_j) \rightarrow 0 \quad \text{as} \quad k, j \rightarrow \infty$$

i.e.

$$\forall \varepsilon > 0, \quad \exists N \in \mathbb{N}, \quad \forall i, j > N : \quad \|x_k - x_j\| < \varepsilon$$

2. X is called a complete metric space if every Cauchy sequence in X converges to a point in X

As in \mathbb{R} , every Cauchy sequence is bounded.

Theorem 1.4.2

1. If $x_k \rightarrow x$ then every subsequence $x_{k_n} \rightarrow x$
2. Every convergent sequence is a Cauchy sequence
3. A Cauchy sequence converges, if it has a convergent subsequence

Proof. $x_k \rightarrow x$, is equivalent to $\forall \varepsilon > 0, \exists N \in \mathbb{N}, \forall k > N : \|x_k - x\| < \varepsilon$

Let (x_{k_n}) be a subsequence. we have :

$$\begin{aligned} \forall \varepsilon > 0, \exists N_1 = \phi(N), \underbrace{\phi(k)}_{=k_m} > N_1 \implies \|x_{k_m} - x\| < \varepsilon \\ \|x_k - x\| \leq \|x_k - x_{k_m}\| + \|x_{k_m} - x\| \leq \varepsilon \end{aligned}$$

$$\begin{aligned} \forall \varepsilon > 0, \exists N_1 \in \mathbb{N}, \forall n > N_1 : \|x_k - x\| \leq \frac{\varepsilon}{2} \\ \forall \varepsilon > 0, \exists N_2 \in \mathbb{N}, \forall n > N_2 : \|x_{k_m} - x\| \leq \frac{\varepsilon}{2} \end{aligned}$$

□

Theorem 1.4.3

\mathbb{R}^n , is complete.

Proof. Since (x_k) is Cauchy, for each $i = 1, \dots, n$ we have:

$$|x_{k_i} - x_{j_i}| \leq \|x_k - x_j\| \rightarrow 0 \quad \text{as } k, j \rightarrow \infty$$

Hence each component sequence (x_{k_i}) is Cauchy in \mathbb{R} , since \mathbb{R} is complete, then there exists $x_i \in \mathbb{R}$ such that $x_{k_i} \rightarrow x_i$ for each $i = 1, \dots, n$. Taking $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$, then $x_k \rightarrow x$.

□

Theorem 1.4.4

Let S be a subset of \mathbb{R}^n , and let $x \in \mathbb{R}^n$. Then $x \in \bar{S}$ if and only if there exist a sequence of points in S that converges to x .

Proof. (\Leftarrow)

Suppose that $x \in \overline{S}$, if $x \in S$, we take $x = x_k$, if $x \notin S$, $x \in \partial S$, then for all $k \in \mathbb{N}$, there exist $x_k \in S$, such that $x_k \in B(x, e^{-k})$, it follows that $\|x_k - x\| < e^{-k}$.

(\Rightarrow)

By applying the definition, Easy! □

Definition 1.4.4

Let $S \subset \mathbb{R}^n$, the diameter $d(S)$ of S is

$$d(S) = \sup \{\|x - y\| : x, y \in S\}$$

Example

$$d(B(x, r)) = 2r$$

$$d(\mathbb{Z}^n) = +\infty$$

Definition 1.4.5

A set S in \mathbb{R}^n is called bounded if there exist $R > 0$ such that $S \subset B(0, R)$, Equivalently, if its diameter is finite.

Theorem 1.4.5 Bolzano-Weirstrass

1. Every bounded sequence in \mathbb{R}^n has a convergent subsequence.
2. Every bounded infinite set in \mathbb{R}^n has a limit point

CHAPTER

2

Functions forms \mathbb{R}^n to \mathbb{R}^n

2.1. Functions from \mathbb{R}^n to \mathbb{R}^m

A function f from a subset Ω of \mathbb{R}^n , called the domain of f , into \mathbb{R}^m is a rule that assigns to each point $x = (x_1, x_2, \dots, x_n) \in \Omega$, a unique point $y \in \mathbb{R}^m$, we write $f : \Omega \rightarrow \mathbb{R}^m$, $y = f(x)$, and we call $f(x)$ the image of x under f , the set of all images is called the range of f , and is written as:

$$\text{rang}(f) = \{f(x) : x \in \mathbb{R}\}$$

Since y is a vector in \mathbb{R}^m , then $f(x) = y = (y_1, y_2, \dots, y_m)$, The components y_j are uniquely defined by x and the function f .

Hence they define the functions $f_j : \Omega \rightarrow \mathbb{R}$, by the rule $y_j = f_j(x)$, for all $j = 1, \dots, m$. These functions are called the components functions of f , we then write

$$f(x) = (f_1(x), f_2(x), \dots, f_n(x))$$

Example

1. The function $f(x, y) = (y, 2xy, \ln(x^2 + y^2 + 1))$ is defined everywhere in \mathbb{R}^2 it is a vector-valued function $f : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ with component functions:

$$f_1(x, y) = y, \quad f_2(x, y) = 2xy, \quad f_3(x, y) = \ln(x^2 + y^2 + 1)$$

2. The function $f(x, y) = \ln(1 - x^2 - y^2)$ is real-valued function defined on the set:

$$\Omega = \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 < 1\}$$

3. The function $f(x) = (\cos x, \sin x, \alpha x)$ is a vector-valued function in one variable $x \in \mathbb{R}$ with values in \mathbb{R}^3 .

Associated with any function $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$ are two types of sets which play a special rule in the development of our subject. The graph of f and the level set of f

Definition 2.1.1 The Graph of a function

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$, the graph of f is the subset of \mathbb{R}^{n+1} (a hypersurface) defined as follows

$$\text{graph}(f) = \{(x, f(x)) : x \in \Omega\}$$

In the important case of a function of variables $(x, y) \in \Omega \subset \mathbb{R}^2$ the graph of f is the surface.

Definition 2.1.2 The Level of a function

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$, and $\alpha \in \mathbb{R}$, the subset S_α of \mathbb{R}^n defined by:

$$S_\alpha = \{x \in \Omega : f(x) = \alpha\}$$

is called the level set of f of level α .

Note that the graph of f be regarded as the level set S of a function

$F : \Omega \times \mathbb{R} \rightarrow \mathbb{R}$, defined by $F(x, y) = f(x) - y$, that is:

$$\text{graph}(f) = S = \{(x_1, \dots, x_n, y) : F(x_1, \dots, x_n, y) = 0\} = \{(x, y) : F(x, y) = 0\}$$

Example The function $f(x, y) = e^{-(x^2+y^2)}$ its graph is:

$$\text{graph}(f) = \{(x, y, e^{-(x^2+y^2)}) : (x, y) \in \mathbb{R}^2\}$$

And the level of f of level α is:

$$S_\alpha(f) = \{(x, y) \in \mathbb{R}^2 : e^{-(x^2+y^2)} = \alpha\} = \begin{cases} \emptyset & \text{if } \alpha \leq 0 \text{ or } \alpha > 1 \\ x^2 + y^2 = -\ln(\alpha) & \end{cases}$$

Definition 2.1.3 The Algebraic Operations

Let $f, g : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$. The sum of f and g is defined by:

$$(f + g)(x) = f(x) + g(x), \quad \forall x \in \Omega$$

and the difference:

$$(f - g)(x) = f(x) - g(x), \quad \forall x \in \Omega$$

When $m = 1$ the product of f and g is defined by:

$$fg(x) = f(x)g(x), \quad \forall x \in \Omega$$

and the quotient by:

$$\left(\frac{f}{g}\right)(x) = \frac{f(x)}{g(x)}, \quad \forall x \in \Omega \text{ with } g(x) \neq 0$$

Definition 2.1.4 The Image and The Inverse Image

Given a function $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ and subsets $A \subset \mathbb{R}^n$ and $B \subset \mathbb{R}^m$, we define the image of A under f by:

$$f(A) = \{f(x) : x \in A\}$$

and the inverse image of B under f by:

$$f^{-1}(B) = \{x \in \Omega : f(x) \in B\}$$

2.2. Limits of functions

To simplify the notation, the Euclidean norm in \mathbb{R}^n and \mathbb{R}^m will be both denoted by $\|\cdot\|$.

Definition 2.2.1 Limit of a function

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ and a is an accumulation point of Ω , a point $b \in \mathbb{R}^m$ is said to be the limit of f at a written as $\lim_{x \rightarrow a} f(x) = b$ if:

$$\forall \varepsilon > 0, \quad \exists \delta > 0, \quad \forall x \in \Omega : \quad 0 < \|x - a\| < \delta \implies \|f(x) - b\| < \varepsilon$$

this is,

$$\lim_{x \rightarrow a} f(x) = b \iff \lim_{x \rightarrow a} \|f(x) - b\| = 0$$

Theorem 2.2.1 Uniqueness of the limit

If $\lim_{x \rightarrow a} f(x)$ exists then it is unique.

Proof. Assume $\lim_{x \rightarrow a} f(x)$ exists, and suppose it is not unique, so there exist $b, b' \in \mathbb{R}^m$

such that:

$$\begin{cases} \forall \varepsilon > 0, \exists \delta_1 > 0, \forall x \in \Omega : \|x - a\| < \delta_1 \implies \|f(x) - b\| < \frac{\varepsilon}{2} \\ \forall \varepsilon > 0, \exists \delta_2 > 0, \forall x \in \Omega : \|x - a\| < \delta_2 \implies \|f(x) - b'\| < \frac{\varepsilon}{2} \end{cases}$$

let $\varepsilon > 0$ and take $\delta = \min(\delta_1, \delta_2)$ then we have:

$$\|b - b'\| \leq \|b - f(x)\| + \|f(x) - b'\| < \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon$$

since ε is arbitrary, we get $\|b - b'\| = 0$ which gives us $b = b'$ contradiction. \square

Example Show that $\lim_{(x,y) \rightarrow (0,0)} f(x, y) = 0$,

$$f(x, y) = \frac{|x|^{5/2} + |y|^{5/2}}{x^2 + y^2}$$

Let $\varepsilon > 0$, we need to find $\delta > 0$ such that $\forall x \in \Omega :$

$$0 < \|x - a\| < \delta \implies \|f(x, y)\| < \varepsilon$$

We have:

$$\begin{aligned} |f(x, y)| &= \frac{|x|^{5/2} + |y|^{5/2}}{x^2 + y^2} \leq \frac{2(x^2 + y^2)^{5/4}}{x^2 + y^2} = 2(x^2 + y^2)^{1/4} = 2(\sqrt{x^2 + y^2})^{1/2} \\ &< 2\sqrt{\delta} \end{aligned}$$

Hence, by taking $\delta = (\frac{\varepsilon}{2})^2$ we get $|f(x, y)| < \varepsilon$

$$f(x, y) = \frac{\sin(e^{-\frac{1}{x^2+y^2}})}{e^{-\frac{1}{x^2+y^2}}}$$

$$\lim_{(x,y) \rightarrow (0,0)} f(x, y) = 1$$

Theorem 2.2.2

For $f, g : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ and let a be an accumulation point of Ω , we have:

$$\lim_{x \rightarrow a} (\alpha f(x) + \beta g(x)) = \alpha \lim_{x \rightarrow a} f(x) + \beta \lim_{x \rightarrow a} g(x)$$

And when $m = 1$,

$$\begin{aligned} \lim_{x \rightarrow a} f(x)g(x) &= \lim_{x \rightarrow a} f(x) \lim_{x \rightarrow a} g(x) \\ \lim_{x \rightarrow a} \frac{f(x)}{g(x)} &= \frac{\lim_{x \rightarrow a} f(x)}{\lim_{x \rightarrow a} g(x)} \end{aligned}$$

where $g(x) \neq 0, \forall x \in \Omega$ and $\lim_{x \rightarrow a} g(x) \neq 0$

Proof.

$$\lim_{x \rightarrow a} f(x) = b \iff \forall \varepsilon > 0, \exists \delta_1, \forall x \in \omega : \|x - a\| < \delta_1 \implies \|f(x) - b\| < \frac{\varepsilon}{2M}$$

$$\lim_{x \rightarrow a} g(x) = b' \iff \forall \varepsilon > 0, \exists \delta_2, \forall x \in \omega : \|x - a\| < \delta_2 \implies \|g(x) - b'\| < \frac{\varepsilon}{2\|b\|}$$

$$\lim_{x \rightarrow a} fg = bb' \iff \forall \varepsilon > 0, \exists \delta > 0, \forall x \in \omega : 0 < \|x - a\| < \delta \implies \|f(x)g(x) - bb'\|$$

$$= \|f(x)g(x) - bg(x) + bg(x) - bb'\| \leq$$

□

Theorem 2.2.3

Let $f(x) = (f_1, \dots, f_n) : \omega \rightarrow \mathbb{R}^m$ and let b be an accumulation point of Ω then:

$$\lim_{x \rightarrow a} f(x) = b \iff \lim_{x \rightarrow a} f_j(x) = b_j \quad \forall j = 1, \dots, m$$

Proof.

$$(\implies)$$

Suppose that $\lim_{x \rightarrow a} f(x) = b$, then $\forall j = 1, \dots, m$

$$|f_j(x) - b_j| \leq \|f(x) - b\|$$

and then :

$$\forall \varepsilon > 0, \exists \delta > 0, \forall x : x \in \Omega : 0 < \|x - a\| < \delta \implies |f_j(x) - b_j| < \varepsilon$$

Thus $\lim_{x \rightarrow a} f_j(x) = b_j$

$$(\iff)$$

Suppose that for all $j = 1, \dots, m$, $\lim_{x \rightarrow a} f_j(x) = b_j$.

$$\forall \varepsilon > 0, \exists \delta_j > 0, \forall x \in \Omega : 0 < \|x - a\| < \delta_j \implies \|f_j(x) - b_j\| < \frac{\varepsilon}{\sqrt{m}}$$

$$\forall \varepsilon > 0, \exists \delta = \min_{j=1, \dots, m} \delta_j : \forall x \in \Omega : 0 < \|x - a\| < \delta \implies \|f(x) - b\|$$

$$\|f(x) - b\| = \sqrt{\sum_{j=1}^m |f_j(x) - b_j|^2} \leq \left(\sum_{j=1}^m \frac{\varepsilon^2}{m} \right)^{1/2} = \varepsilon$$

then $\lim_{x \rightarrow a} f(x) = b$

□

Example

$$f(x, y) = \left(e^x, \ln(1 + |y|), \frac{x^2}{1 + y^2} \right)$$

$$\lim_{(x,y) \rightarrow (0,0)} f(x, y) = (1, 0, 0)$$

Theorem 2.2.4

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$, and let a be an accumulation point of Ω , then:

$$\lim_{x \rightarrow a} f(x) = b \iff \left(\forall (x_k) \subset \Omega : \lim_{x \rightarrow \infty} x_k = a \implies \lim_{k \rightarrow \infty} f(x_k) = b \right)$$

Proof.

$$(\implies)$$

Suppose that $\lim_{x \rightarrow a} f(x) = b$

$$\forall \varepsilon > 0, \exists \delta > 0, \forall x \in \Omega : 0 < \|x - a\| < \delta \implies \|f(x) - b\| < \varepsilon$$

Let $(x_k) \subset \Omega$ such that $\lim_{k \rightarrow \infty} x_k = a$.

$$\text{For } \delta > 0, \exists N \in \mathbb{N} : \forall k > N \quad \|x_k - a\| < \delta \implies \|f(x_k) - b\| < \varepsilon$$

$$\forall \varepsilon > 0, \exists N \in \mathbb{N} : \forall k > N \implies \|f(x_k) - b\| < \varepsilon$$

Then $\lim_{k \rightarrow \infty} f(x_k) = b$

$$(\iff)$$

Suppose that $\forall (x_k) \subset \Omega, \lim_{k \rightarrow \infty} x_k = a \implies \lim_{k \rightarrow \infty} f(x_k) = b$ and suppose that $\lim_{x \rightarrow a} f(x) \neq b$ then:

$$\exists \varepsilon > 0, \forall \delta > 0, \exists x \in \Omega : \|x - a\| < \delta \text{ and } \|f(x) - b\| > \varepsilon$$

$$\text{for } \delta = \frac{1}{k}, \exists x_k \in \Omega : \|x_k - a\| < \frac{1}{k} \text{ and } \|f(x_k) - b\| > \varepsilon$$

$$\lim_{k \rightarrow \infty} x_k = a \quad \lim_{k \rightarrow \infty} f(x_k) = b$$

contradiction! then $\lim_{x \rightarrow a} f(x) = b$

□

By the uniqueness of the limit of f at point a, b is also the limit as x approaches along of any curve passing from a , this fact will be raised to explain why certain functions fail to have a limit at a point

Example

1. Let $f(x, y) = \frac{xy+y^2}{x^2+y^2}$

- Show that the limit of f at $(0, 0)$ doesn't exist.

for $x = y$ $\lim_{x \rightarrow 0} f(x, x) = 1$

for $x = -y$ $\lim_{x \rightarrow 0} f(x, -x) = 0$

consequently, the limit doesn't exist !

2. Let $f(x, y) = \frac{2xy^2}{x^2+y^4}$

- Show that the limit of f at $(0, 0)$ doesn't exist.

for $x = \alpha y$ $\lim_{x \rightarrow 0} f(x, \alpha x) = \infty$

for $x = y^2$ $\lim_{x \rightarrow 0} f(y^2, y) = 1$

consequently, the limit doesn't exist !

2.3. Continious functions

Definition 2.3.1 Continious functions

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$, where Ω is an open set of \mathbb{R}^n and $a \in \Omega$, we say that f is continious at a if :

$$\forall \varepsilon > 0, \exists \delta > 0, \forall x \in \Omega : \|x - a\| < \delta \implies \|f(x) - f(a)\| < \varepsilon$$

Observe that with this definition, every function f is continious at each isolated point of a , that is every point of Ω that is not a limit of Ω .

If a is an accumulation point of Ω , then the definition of continuity is given by:

$$f \text{ is continuous at } a \iff \lim_{x \rightarrow a} f(x) = f(a)$$

More over, we say that f is continuous on Ω if it's continuous at each point of Ω .

Example

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

$$\lim_{(x,y) \rightarrow (0,0)} |f(x,y)| = \lim_{(x,y) \rightarrow (0,0)} \left| \frac{x^2y}{x^2 + y^2} \right| \leq \lim_{(x,y) \rightarrow (0,0)} |y| = 0$$

since $\lim_{(x,y) \rightarrow (0,0)} f(x,y) = 0 = f(0,0)$, the function f is continuous at $(0,0)$,

as immediate consequence of the definition of continuity and the properties of the limit is the following properties

Theorem 2.3.1

1. $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$, then $f = (f_1, \dots, f_n)$ is continuous at a if and only if f_j is continuous at a for all $j = 1, \dots, n$
2. $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ then f is continuous at a if for all $(x_k) \subset \Omega$, if $\lim_{k \rightarrow \infty} x_k = a \implies \lim_{n \rightarrow \infty} f(x_k) = f(a)$
3. The sum, difference, product and quotient of two continuous functions, whenever defined, is continuous

Example Let $P_i : \mathbb{R}^n \rightarrow \mathbb{R}$ be the projection function on the i^{th} coordinate axis, that is:

$$\begin{aligned} P_i : \quad \mathbb{R}^n &\longrightarrow \mathbb{R} \\ (x_1, \dots, x_n) &\longmapsto x_i \end{aligned}$$

Then for $i = 1, \dots, n$, each P_i is continuous on \mathbb{R}^n

$$\forall x, y \in \mathbb{R}^n \quad |P_i(x) - P_i(y)| = |x_i - y_i| \leq \|x - y\| < \varepsilon$$

$$\begin{aligned} f : \quad \mathbb{R}^n &\longrightarrow \mathbb{R} \\ x &\longmapsto \|x\| \end{aligned}$$

Theorem 2.3.2

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$. Then f is continuous at $a \in \Omega$ if and only if $f(x_k) \rightarrow f(a)$ wherever (x_k) is a sequence of points of Ω such that $x_k \rightarrow a$.

Definition 2.3.2

Let $f : \Omega \rightarrow \mathbb{R}^n$ and $g : U \subset \mathbb{R}^n \rightarrow \mathbb{R}^p$ be two functions such that $f(\Omega) \subset U$.

then the composition of f and g is defined by

$$g \circ f(x) = g(f(x)) \quad g \circ f : \Omega \rightarrow \mathbb{R}^p$$

Theorem 2.3.3

Let $f : \Omega \subset \mathbb{R}^m \rightarrow \mathbb{R}^n$ and $g : U \subset \mathbb{R}^m \rightarrow \mathbb{R}^p$, such that f is continuous at a and g is continuous at $f(a)$, then $g \circ f$ is continuous at a

Proof. Suppose that f is continuous at a then for all $(x_k) \subset \Omega$, such that $\lim_{k \rightarrow \infty} x_k = a$, we get $\lim_{k \rightarrow \infty} f(x_k) = f(a)$, and g is continuous at $f(a)$, $\lim_{k \rightarrow \infty} g(f(x_k)) = g(f(a))$, then $g \circ f$, is continuous at a \square

Example Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ defined by:

$$f(x, y) = \begin{cases} e^{-\frac{1}{|x-y|}} & x \neq y \\ 0 & x = y \end{cases}$$

Show that f is continuous on \mathbb{R}^2 . Since f is a composition of continuous functions, the only points where f may not be continuous are the points on the line $y = x$, that is, at points (a, a) where $a \in \mathbb{R}$. We investigate the continuity of f at these points. Setting $t = x - y$, the function becomes:

$$f(t) = \begin{cases} e^{-\frac{1}{|t|}} & t \neq 0 \\ 0 & t = 0 \end{cases}$$

Now since:

$$\lim_{t \rightarrow 0} f(t) = f(0, 0) = \lim_{(x,y) \rightarrow (0,0)} f(x, y) = 0$$

Then f is continuous at all points on $x = y$ then f is continuous on \mathbb{R}^2

2.4. Linear Transformation

Definition 2.4.1

A function $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is called linear transformation, if $\forall x, y \in \mathbb{R}^n$, we have

$$T(x + \alpha y) = T(x) + \alpha T(y)$$

Theorem 2.4.1

A function $T : \mathbb{R}^n \rightarrow \mathbb{R}$ is a linear transformation if and only if there exist a unique $w \in \mathbb{R}^n$, such that $T(x) = w \cdot x$ for all $x \in \mathbb{R}^n$.

Proof.

[\Rightarrow]

Suppose that there exist $w \in \mathbb{R}^n$ such that $T(x) = w \cdot x$

$$\begin{aligned}\forall x, y \in \mathbb{R}^n, \forall \alpha \in \mathbb{R}, \quad T(x + \alpha y) &= w(x + \alpha y) \\ &= w \cdot x + \alpha w \cdot y \\ &= T(x) + \alpha T(y)\end{aligned}$$

(\Leftarrow)

Let (e_1, \dots, e_n) is a basis of \mathbb{R}^n , then for all $x \in \mathbb{R}^n$,

$$x = \sum_{i=1}^n x_i e_i$$

$$\begin{aligned}T(x) &= T\left(\sum_{i=1}^n x_i e_i\right) = \sum_{i=1}^n T(x_i e_i) \\ &= \sum_{i=1}^n x_i T(e_i) \\ &= x \cdot w \quad w = (T(e_1), \dots, T(e_n))\end{aligned}$$

For the unicity, suppose that there exist another w and w' where $T(x) = w'x = wx$

$$\begin{aligned}T(x) = wx = w'x &\implies wx - w'x = 0 \\ &\implies (w - w')x = 0 \quad \forall x \in \mathbb{R}^n \\ &\implies \|w - w'\|_2^2 = 0 \quad \implies w = w'\end{aligned}$$

□

Theorem 2.4.2

Any linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is continuous.

Proof. Let $x, y \in \mathbb{R}^n$, let (e_1, \dots, e_n) be some canonical bases, we have

$$\begin{aligned}\|T(x) - T(y)\| &= \|T(x - y)\| = \|T\left(\sum (x_i - y_i)e_i\right)\| \\ &= \left\| \sum (x_i - y_i)T(e_i) \right\| \\ &\leq \sum_{i=1}^n |x_i - y_i| \|T(e_i)\| \\ &\leq \sup_{i=1,n} |x_i - y_i| \underbrace{\sum_{j=1}^n \|T(e_j)\|}_{M} \leq \|x - y\| \underbrace{\sum_{i=1}^n \|T(e_i)\|}_M\end{aligned}$$

Continious functions on compact sets. \square

2.5. Compact sets

Definition 2.5.1

A set $\Omega \subset \mathbb{R}^n$ is compact if every sequence in Ω has a convergent subsequence whose limit is on Ω .

Theorem 2.5.1

A set $\Omega \subset \mathbb{R}^n$, is compact if and only if Ω is closed and bounded

Proof. Suppose that Ω is closed and bounded, let $(x_k) \subset \Omega$ then by bolzano-weirstrass (x_k) has a subsequence (x_{k_j}) convergent to a limit $x_0 \in \overline{\Omega}$, and by Ω is closed then $x_0 \in \Omega$. Suppose that Ω is compact, then Ω is closed, indeed, $(x_k) \subset \Omega$ and $x_k \rightarrow x_0$, then $x_0 \in \Omega$, Ω is unbounded.

$$\forall k \in \mathbb{N}, \exists x_k \in \Omega : \|x_k\| \geq k$$

(x_k) does not contain any convergent subsequence, contradiction with Ω is compact. \square

Theorem 2.5.2

A set $\Omega \subset \mathbb{R}^n$ is closed and bounded if and only if every infinite subset of Ω has an accumulation point that belongs to Ω .

Theorem 2.5.3

let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a continious, if Ω is compact then $f(\Omega)$, is compact.

Proof. Let $(y_k) \subset f(\Omega)$, then

$$\forall k \in \mathbb{N}, \exists (x_k) \subset \Omega : f(x_k) = y_k$$

since Ω is compact, then there exist $(x_{nk}) \subset \Omega$, such that (x_{nk}) converges to a point x_0 in Ω , since f is continuous then

$$\lim_{n \rightarrow \infty} f(x_{nk}) = f(x_0)$$

which is a subsequence of $f(x_n) = y_n$, hence converges in $f(\Omega)$, therefore $f(\Omega)$ is compact \square

Theorem 2.5.4 Weirstrass

let $f : \Omega \rightarrow \mathbb{R}$ be a continuous function and Ω is compact, then f attains its maximum and minimum value in Ω .

Proof. The set $f(\Omega)$ is a compact subset of \mathbb{R} . Hence it is bounded and closed, Let $[c, d]$ be the smallest closed interval containing $f(\Omega)$. Such an interval exists since $f(\Omega)$ is bounded. Since $f(\Omega)$ is closed $c \in f(\Omega)$ and $d \in f(\Omega)$ then $\max f(\Omega) = d$ and $\min f(\Omega) = c$. \square

2.6. Uniform continuity

Definition 2.6.1

Let $\Omega \subset \mathbb{R}^n$ and let $f : \Omega \rightarrow \mathbb{R}^m$, we say f is uniformly continuous on Ω if for every $\varepsilon > 0$, there exist $\delta > 0$ such that if $x, y \in \Omega$ and $\|x - y\| < \delta$, then $\|f(x) - f(y)\| < \varepsilon$

Theorem 2.6.1

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a continuous function if Ω is compact then f is uniformly continuous.

Proof. suppose that f is continuous and Ω is compact with f not uniformly continuous, we have that f is not uniformly continuous on Ω , then :

$$\exists \varepsilon > 0, \forall n \in \mathbb{N}, \exists (x_n)_{n \in \mathbb{N}}, (y_n)_{n \in \mathbb{N}} \subset \Omega : \|x_n - y_n\| < \frac{1}{n} \text{ and } \|f(x_n) - f(y_n)\| \geq \varepsilon$$

x_n and $y_n \in \Omega$, and Ω is compact set by definition x_n has a convergent subsequence $(x_{nk})_{k \in \mathbb{N}}$ and the sequence $(y_{nk})_{k \in \mathbb{N}} \subset \Omega$ then (y_{nk}) has a convergent subsequence $(y_{k'n})$

converges on Ω , then the subsequences $(x_{k'n})$ and $(y_{k'n})$ are convergent to x and y in Ω respectively and by the above

$$\|x_{k'n} - y_{k'n}\| \leq \frac{1}{k'n}$$

we deduce that $x = y$, then

$$\|f(x_{k'n}) - f(y_{k'n})\| \geq \varepsilon$$

$0 \geq \varepsilon > 0$ contradiction then f is uniformly continuous. \square

2.7. Connectedness

Definition 2.7.1

A subset $\Omega \subset \mathbb{R}^n$ is said to be connected if its not possible to write $\Omega = (A \cap \Omega) \cup (B \cap \Omega)$, with A and B are open sets of \mathbb{R}^n and that $\Omega \cap A$ and $\Omega \cap B$ both non empty and

$$A \cap B \cap \Omega = \emptyset$$

Theorem 2.7.1

let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a continuous function and Ω is connected, then $f(\Omega)$ is connected

Proof. assume the contrary, that $f(\Omega)$ is not connected, then $f(\Omega) = A' \cup B'$, are open sets in $f(\Omega)$ such that $A' \cap B' = \emptyset$, let $A = \Omega \cap f^{-1}(A')$, $B = \Omega \cap f^{-1}(B')$, A and B are open sets of Ω , and $A \cap B = \emptyset$ and $\Omega = A \cup B$, contradiction with Ω is connected set hence $f(\Omega)$ is connected \square

Theorem 2.7.2

let Ω be a connected set of \mathbb{R}^n and $f : \Omega \rightarrow \mathbb{R}$ is continuous, and let $c \in \mathbb{R}$ such that

$$f(x_1) \leq c \leq f(x_2)$$

then by definition of an interval there exist $x_0 \in \Omega$ such that $f(x_0) = c$

Proof. we have $f(\Omega)$ is connected set on \mathbb{R} , then $f(\Omega)$ is an interval, and

$$f(x_1) \leq c \leq f(x_2)$$

then $\exists x_0 \in \Omega$ such that $f(x_0) = c$, where $c \in [f(x_1), f(x_2)]$. \square

2.8. Path-connectedness

Definition 2.8.1

let x, y be points of metric space X , a path in X from x to y is a continuous function
 $\gamma : [a, b] \rightarrow X$, where $\gamma(a) = x$ and $\gamma(b) = y$

Definition 2.8.2

A metric space X is called pathwise connected if any two points in X can be joined by a path in X

Corollary 2.8.1 *Any pathwise connected metric space is connected*

Theorem 2.8.1

Continuous images of pathwise connected sets are pairwise connected

Theorem 2.8.2

Let Ω be an open set connected in \mathbb{R}^n , then Ω is pathwise connected

Proof. Let x_0 be fixed point of X , and consider the sets A and B defined by :

$$A = \{x \in \Omega : x \text{ can be joined with } x_0 \text{ by a path in } \Omega\}$$

$$B = \{z \in \Omega : z \notin A\}$$

we have that $A \cup B = \Omega$, now we show that A and B are open sets of Ω , let $x \in A \subset \Omega$ then there exist $r > 0$, such that $B(x, r) \subset \Omega$, let $y \in B(x, r)$ and since $B(x, r)$ is convex then we can join x with y by a segment and x joined by x_0 because $x \in A$, then y is joined with x_0 in Ω then $y \in A$, hence A is an open set of Ω .

Secondly we show that B is an open set, let $z \in B \subset \Omega$, there exist $\delta > 0$ such that $B(z, \delta) \subset \Omega$, suppose that there exist $x \in B(z, \delta)$ and $x \notin B$ then $x \in A$ and we can join z to x (convexity of the open ball), and we can join x to x_0 , so $z \in A$ which is a contradiction since we took $z \in B$, therefore $x \in B$ hence B is an open set, we have that Ω is a connected set of X , therefore either $A = \emptyset$ or $B = \emptyset$, since A contains x_0 then $B = \emptyset$, so $A = \Omega$. \square

Proposition 2.8.1 Let Ω be an open set in \mathbb{R}^n . Then Ω is connected if and only if for any two points in Ω , they can be joined a polynomial path.

2.9. Convex Sets

Definition 2.9.1

A non empty set \mathcal{C} in \mathbb{R}^n is said to be convex if the line segment joining any two points x and y of \mathcal{C} is contained in \mathcal{C} , that is

$$x, y \in \mathcal{C} \implies \gamma(t) = tx + (1 - t)y \in \mathcal{C} \quad \forall t \in [0, 1]$$

Example Any open or closed ball of X is a convex set, let $x, y \in B(r, a)$ and $t \in [0, 1]$, then

$$\begin{aligned} \|tx + (1 - t)y - a\| &= \|t(x - a) + (1 - t)(y - a)\| \leq t\|x - a\| + (1 - t)\|y - a\| \\ &\leq tr + (1 - t)r \\ &\leq r \end{aligned}$$

then we have

$$tx + (1 - t)y \in B(r, a)$$

CHAPTER

3

Differentiable functions from \mathbb{R}^n to \mathbb{R}^m

Definition 3.0.1

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a function defined on an open set Ω of \mathbb{R}^n , we say that f is differentiable at a if there is a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that

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

if f is differentiable at every point of Ω , we say that f is differentiable on Ω

Remark Setting $h = x - a \neq 0$, then $x \rightarrow a$ in \mathbb{R}^n is equivalent to $h \rightarrow 0$, and so the limit of the above definition is equivalent to

$$\lim_{h \rightarrow 0} \frac{\|f(a + h) - f(a) - T(h)\|_m}{\|h\|_n} = 0$$

Theorem 3.0.1

If $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at a if and only if there exist a function $\varepsilon(x)$ so that for $x \in \Omega$, we have

$$f(x) = f(a) + T(x - a) + \varepsilon(x)\|x - a\|$$

with $\varepsilon(x) \rightarrow 0$ as $x \rightarrow a$

Proof. set

$$\varepsilon(x) = \frac{f(x) - f(a) - T(x - a)}{\|x - a\|}$$

now if f is differentiable at a , then $\lim_{x \rightarrow a} \varepsilon(x) = 0$, conversely if

$$f(x) = f(a) + T(x - a) + \varepsilon(x)\|x - a\|$$

the result also holds, since $x \neq a$, we have

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

therefore f is differentiable at a □

Proposition 3.0.1 *Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a function defined on a open subset $\Omega \subset \mathbb{R}^n$ and $a \in \Omega$. If f is differentiable at a , then T is unique.*

Proof. Suppose T, S are linear transformations satisfying

$$f(a + h) = f(a) + T(h) + \varepsilon_1(h)\|h\|$$

$$f(a + h) = f(a) + S(h) + \varepsilon_2(h)\|h\|$$

with

$$\varepsilon_1(h) \rightarrow 0$$

$$\varepsilon_2(h) \rightarrow 0$$

as $h \rightarrow 0$, now by subtracting the two equations we get

$$(T - S)(h) = (\varepsilon_1 - \varepsilon_2)\|h\|$$

setting $L = T - S$ and dividing by $\|h\|$, it follows that

$$\frac{\|L(h)\|}{\|h\|} = \|\varepsilon_1 - \varepsilon_2\| \rightarrow 0 \quad \text{as } h \rightarrow 0$$

now, let $x \in \mathbb{R}^n$ be any non zero vector and let $t \in \mathbb{R}$, take $h = tx$, then $h \rightarrow 0$ is equivalent to $t \rightarrow 0$, and since L is linear, we get the following

$$0 = \lim_{h \rightarrow 0} \frac{\|L(h)\|}{\|h\|} = \lim_{t \rightarrow 0} \frac{\|L(tx)\|}{\|tx\|} = \lim_{t \rightarrow 0} \frac{\|L(x)\|}{\|x\|} = \frac{\|L(x)\|}{\|x\|}$$

thus $L(x) = 0$, for all $x \in \mathbb{R}^n$ which means $L \equiv 0$. Hence $S = T$ □

Remark As we show that T is unique we denoted T by $d_f^{(a)}$ or $D_f^{(a)}$, the differentiable (derivative) of f at the point a .

Proposition 3.0.2 *If f is differentiable at a , then it is continuous at a*

Proof. Since f is differentiable at a , we have

$$f(x) = f(a) + d_f^{(a)}(x - a) + \varepsilon(x)\|x - a\|$$

by taking limit on both sides

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

since $(\lim_{x \rightarrow a} d_f^{(a)}(x - a) = 0)$, due to continuity of d_f . \square

Theorem 3.0.2

let f and g be function from an open set Ω in \mathbb{R}^n to \mathbb{R}^m differentiable at $a \in \Omega$ and let $c \in \mathbb{R}$, then :

- (1) $f + g$ is differentiable at a and $d_{f+g}(a) = d_f(a) + d_g(a)$
- (2) cf is differentiable at a and $d_{cf}(a) = cd_f(a)$
- (3) if $m = 1$, then $f \cdot g(x) = f(x) \cdot g(x)$ is differentiable at a and $d_{f \cdot g}(a) = f(a) \cdot d_g(a) + d_f(a)g(a)$
- (4) if $m = 1$, and $g(a) \neq 0$, then $\left(\frac{f}{g}(a)\right)$ is differentiable at a and $d_{\frac{f}{g}} = \frac{g(a)d_f^{(a)} - f(a)d_g^{(a)}}{g(a)^2}$

Proof. we have f and g are differentiable at a , i.e.

$$\begin{aligned} f(a+h) &= f(a) + d_f^{(a)}(h) + \varepsilon_1(h)\|h\| \quad \varepsilon_1(h) \rightarrow 0 \quad \text{as } h \rightarrow 0 \\ g(a+h) &= g(a) + d_g^{(a)}(h) + \varepsilon_2(h)\|h\| \quad \varepsilon_2(h) \rightarrow 0 \quad \text{as } h \rightarrow 0 \end{aligned}$$

we get

$$\begin{aligned} fg(a+h) - fg(a) &= f(a+h)g(a+h) - f(a)g(a) \\ &= (f(a+h) - f(a))g(a+h) + f(a)(g(a+h) - g(a)) \\ &= (d_f^{(a)}(h) + \varepsilon_1(h)\|h\|)g(a+h) + f(a)(d_g^{(a)}(h) + \varepsilon_2(h)\|h\|) \\ &= d_f^{(a)}(h)g(a+h) + d_g^{(a)}(h)f(a) + (\varepsilon_1(h)g(a+h) + \varepsilon_2(h)f(a))\|h\| \end{aligned}$$

on the other side we have

$$\begin{aligned} &\frac{fg(a+h) - fg(a) - g(a)d_f^{(a)} - f(a)d_g^{(a)}(h)f(a)}{\|h\|} \\ &= \frac{(g(h+a) - g(a))d_f^{(a)}(h) + (\varepsilon_1(h)g(a+h) + \varepsilon_2(h)f(a))\|h\|}{\|h\|} \end{aligned}$$

the RHS tends to zero as $h \rightarrow 0$, therefore we get the equality. \square

Theorem 3.0.3

Let Ω be an open set in \mathbb{R}^n and $f : \Omega \rightarrow \mathbb{R}^m$ and $g : U \rightarrow \mathbb{R}^p$ where U is an open set in \mathbb{R}^m with $f(\Omega) \subset U$, if f is differentiable at $a \in \Omega$ and g is differentiable at $f(a)$ then $g \circ f$ is differentiable at a and $d_{g \circ f}^{(a)} = d_g^{(f(a))} \cdot d_f^{(a)}$

Proof. f and g are differentiable at a , then let $y = f(x)$ and $b = f(a)$, we have :

$$\begin{aligned} f(x) &= f(a) + d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\| \\ g(y) &= g(b) + d_g^{(b)}(y - b) + \varepsilon_2(y)\|y - b\| \end{aligned}$$

Substituting the first equation into the second we get

$$\begin{aligned} g(f(x)) &= g(f(a)) + d_g^{(f(a))} \left(d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\| \right) + \varepsilon_2(f(x)) \|d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\|\| \\ &= g(f(a)) + d_g^{(f(a))} d_f^{(a)}(x - a) + \\ &\quad \underbrace{\left[d_g^{(f(a))} (\varepsilon_1(x)\|x - a\|) + \varepsilon_2(f(x)) \|d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\|\| \right]}_{\|x - a\|\varepsilon(x)} \end{aligned}$$

the proof is complete if we show that $\lim_{x \rightarrow a} \varepsilon(x) = 0$ where

$$\varepsilon(x) = \frac{d_g^{(f(a))} (\varepsilon_1(x)\|x - a\|) + \varepsilon_2(f(x)) \|d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\|\|}{\|x - a\|}$$

we get

$$\varepsilon(x) = d_g^{(f(a))} (\varepsilon_1(x)) + \frac{1}{\|x - a\|} (\varepsilon_2(f(x)) \|d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\|\|)$$

clearly $\lim_{x \rightarrow a} d_g^{(f(a))} (\varepsilon_1(x)) = 0$, we must show the rest tends to zero

$$\begin{aligned} \left\| \frac{1}{\|x - a\|} (\varepsilon_2(f(x)) \|d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\|\|) \right\| &\leq \|\varepsilon_2(f(x))\| c \frac{\|x - a\|}{\|x - a\|} + \|\varepsilon_1(x)\| \\ &= c \|\varepsilon_2(f(x))\| + \varepsilon_1(x) \rightarrow 0 \quad \text{as } x \rightarrow a \end{aligned}$$

□

Proposition 3.0.3 Let Ω be an open set in \mathbb{R}^n and $f : \Omega \rightarrow \mathbb{R}^m$ with component functions $f = (f_1, \dots, f_n)$, then f is differentiable at a if f_j is differentiable at a , for all $j = 1, \dots, m$, Moreover

$$d_f^{(a)} = (d_{f_1}^{(a)}, d_{f_2}^{(a)}, \dots, d_{f_n}^{(a)})$$

Proof. let $f = (f_1, f_2, \dots, f_n)$ and $d_f^{(a)} = (\lambda_1, \dots, \lambda_n)$, we have the vectorial equality from

$$f(x) = f(a) + d_f^{(a)}(x - a) + \varepsilon_1(x)\|x - a\|$$

then

$$f_j(x) = f_j(a) + \lambda_j(x - a) + \varepsilon_j(x)\|x - a\| \quad \forall j = 1, \dots, m$$

then for all $j = 1, \dots, m$:

$$\sup_{j=1,\dots,m} \|\varepsilon_j(x)\| \leq \|\varepsilon(x)\| \leq m \sup_{j=1,\dots,m} \|\varepsilon_j(x)\|$$

suppose that f is differentiable at a , then $\varepsilon(x) \rightarrow 0$ as $x \rightarrow a$, to show that $f_j(x)$ is differentiable we have to show that $e_j(x)$ goes to 0 as x tends to a , from the above inequality, we get

$$\forall j = 1, \dots, n \quad \lim_{x \rightarrow a} e_j(x) = 0$$

now suppose that $f_j(x)$ is differentiable at a for all $j = 1, \dots, n$, again from the above inequality, by taking the limit we get the inverse implication. \square

One should notice by the above proposition, we see that to study the differentiability of a function $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$, it suffices to study the differentiability of its component functions $f_j : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$, for $j = \overline{1, m}$. Hence, we turn to real valued functions of several variables.

Example

1. let

$$\begin{aligned} f : \Omega \subset \mathbb{R}^n &\longrightarrow \mathbb{R}^n \\ x &\longmapsto x_0 \end{aligned}$$

f is differentiable at any point $a \in \mathbb{R}^n$ and $d_f^{(a)} = 0$

$$\frac{\|f(x) - f(a) - d_f^{(a)}(x - a)\|_n}{\|x - a\|_m} = 0 \rightarrow 0 \text{ as } x \rightarrow a$$

2. let

$$\begin{aligned} f : \mathbb{R}^n &\longrightarrow \mathbb{R}^m \\ x &\longmapsto T(x) \end{aligned}$$

where T is a linear transformation, f is differentiable at each point $a \in \mathbb{R}^n$, and we have $d_f^{(a)}(h) = T(h)$

$$\frac{\|f(x) - f(a) - T(x - a)\|_n}{\|x - a\|} = 0 \rightarrow 0 \text{ as } x \rightarrow a$$

3. define the function

$$\begin{aligned} f : \mathbb{R}^n &\longrightarrow \mathbb{R} \\ x &\longmapsto \|x\|^2 \end{aligned}$$

we have that f is differentiable at any point of \mathbb{R}^n

$$\begin{aligned} f(a+h) - f(a) &= \|x+h\|^2 - \|a\|^2 = \langle a+h, a+h \rangle - \|a\|^2 \\ &= \|a\|^2 + 2\langle a, h \rangle + \|h\|^2 - \|a\|^2 \\ &= 2\langle a, h \rangle + \|h\|^2 \end{aligned}$$

hence

$$f(a+h) - f(a) - 2\langle a, h \rangle = \underbrace{\|h\| \|h\|}_{\varepsilon(h)}$$

4. Consider the following function

$$\begin{aligned} f : \mathbb{R}^2 &\longrightarrow \mathbb{R}^2 \\ (x, y) &\longmapsto (x+y, x-y) \end{aligned}$$

we have :

$$d_f^{(a)} = \begin{pmatrix} a_1 + a_2 & a_1 - a_2 \end{pmatrix}$$

and

$$d_f^{(a)} h = \begin{pmatrix} a_1 + a_2 & a_1 - a_2 \end{pmatrix} \begin{pmatrix} h_1 \\ h_2 \end{pmatrix}$$

5. Consider the following

$$\begin{aligned} f : \mathbb{R}^2 &\longrightarrow \mathbb{R} \\ (x, y) &\longmapsto xy \end{aligned}$$

$$\begin{aligned} f(a_1 + h_1, a_2 + h_2) - f(a_1, a_2) &= (a_1 + h_1)(a_2 + h_2) - a_1 a_2 \\ &= a_1 h_2 + a_2 h_1 + \|h\| \frac{h_1 h_2}{\|h\|} \\ &\quad \underbrace{}_{\varepsilon(h)} \end{aligned}$$

therefore we can see that

$$d_f^{(a_1, a_2)}(h_1, h_2) = a_1 h_1 + a_2 h_2$$

6. We will leave this as an exercise for the reader

$$\begin{aligned} f : \mathbb{R}^n &\longrightarrow \mathbb{R} \\ (x_1, \dots, x_n) &\longmapsto |x_1| + \dots + |x_n| \end{aligned}$$

3.1. Partial and directional derivatives

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$ where Ω is an open set of \mathbb{R}^n and $a \in \mathbb{R}^n$, if f is differentiable at a then there exist a linear function $\lambda : \mathbb{R}^n \rightarrow \mathbb{R}$ such that

$$f(x) = f(a) + \lambda^{(a)}(x - a) + \varepsilon(x)\|x - a\|$$

with $\lim_{x \rightarrow a} \varepsilon(x) = 0$, we know that $\lambda(x)$ is an inner product product by some fixed vector $w \in \mathbb{R}^n$, that is

$$\lambda(x) = \langle w, x \rangle \quad w \in \mathbb{R}^n$$

this vector which depends on a , is called the gradient of f at a and is denoted by $\nabla f(a)$, we can write

$$f(x) = f(a) + \langle \nabla f(a), x - a \rangle + \varepsilon(x)\|x - a\|$$

we want to get a convenient and explicit form of the gradient

Definition 3.1.1

Let $f : \Omega \rightarrow \mathbb{R}$, and let Ω be an open set of \mathbb{R}^n , and let $a \in \Omega$. the directional derivative of f at a in direction of a nonzero vector $u \in \mathbb{R}^n$, denoted by $d_{u,f}^{(a)}$ is defined by

$$d_{u,f}^{(a)} = \lim_{t \rightarrow 0} \frac{f(a + tu) - f(a)}{t}$$

wherever the limit exist

Theorem 3.1.1

If $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable at $a \in \Omega$, then for every $u \neq 0$, where $u \in \mathbb{R}^n$, $d_{u,f}^{(a)}$ exist and

$$d_{u,f}^{(a)} = \langle \nabla f(a), u \rangle$$

Proof. Since f is differentiable at a we have

$$f(x) = f(a) + \langle \nabla f(a), x - a \rangle + \varepsilon(x)\|x - a\|$$

by setting $x = a + tu$, this yields

$$f(a + tu) - f(a) = \langle \nabla f(a), tu \rangle + \varepsilon(a + tu)\|tu\|$$

so we get

$$\begin{aligned} d_{u,f}^{(a)} &= \lim_{t \rightarrow 0} \frac{f(a + tu) - f(a)}{t} = \lim_{t \rightarrow 0} \left(\langle \nabla f(a), u \rangle + \varepsilon(a + tu) \|u\| \frac{|t|}{t} \right) \\ &= \langle \nabla f(a), u \rangle \end{aligned}$$

□

certain directions are special, namely those of the standard bases elements $(e_i)_{i=1,n}$
the direction of the coordinate axes

Corollary 3.1.1 *For nonzero vectors $u, v \in \mathbb{R}^n$ and $c, d \in \mathbb{R}$, such that $cu + dv \neq 0$, we have*

$$d_{cu+dv,f}^{(a)} = cd_{u,f}^{(a)} + dd_{v,f}^{(a)}$$

Corollary 3.1.2 $|d_{u,f}^{(a)}|$ is maximized in the direction of $u = \nabla f(a)$

Proof. By the Cauchy-Schwartz inequality

$$|d_{u,f}^{(a)}| = |\langle \nabla f(a), u \rangle| \leq \|\nabla f(a)\| \|u\|$$

with equality only if u is a scalar multiple of $\nabla f(a)$

□

Definition 3.1.2

Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$, the directional derivative of f at a in the direction of e_i is denoted by

$$\frac{\partial f}{\partial x_i}(a) = d_{e_i,f}^{(a)} = \langle \nabla f(a), e_i \rangle = \lim_{t \rightarrow 0} \frac{f(a_1, \dots, a_i + t, \dots, a_n) - f(a_1, \dots, a_n)}{t}$$

is simply the ordinary derivative of f considered as a function of x_i solely, and considering the other components as fixed, since

$$\lambda(x) = \langle w, x \rangle \quad i = \overline{1, n}$$

we have

$$\nabla f(a) = \left(\frac{\partial f}{\partial x_1}(a), \frac{\partial f}{\partial x_2}(a), \dots, \frac{\partial f}{\partial x_n}(a) \right)$$

and so

$$f(x) = f(a) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(a)(x_i - a_i) + \varepsilon(x) \|x - a\|$$

Example Consider this function

$$f(x, y) = \begin{cases} 0 & x = 0 \text{ or } y = 0 \\ 1 & \text{otherwise} \end{cases}$$

$$\frac{\partial f}{\partial x}(0, 0) = \lim_{t \rightarrow 0} \frac{f(t, 0) - f(0, 0)}{t} = 0$$

$$\frac{\partial f}{\partial y}(0, 0) = \lim_{t \rightarrow 0} \frac{f(t, 0) - f(0, 0)}{t} = 0$$

but when

$$\begin{aligned} \frac{\partial f}{\partial x}(x, y) &= \lim_{t \rightarrow 0} \frac{f(x + t, y) - f(x, y)}{t} \\ &= \lim_{t \rightarrow 0} \frac{f(t, y) - f(0, y)}{t} \\ &= \lim_{t \rightarrow 0} \frac{1}{t} \rightarrow \infty \end{aligned}$$

partial derivatives at $(0, 0)$ exist but the function is not even differentiable, but the reverse is true i.e. if a function is differentiable then the partial derivatives exist

Example Let $f(x, y, z) = 2x^2 + 3y^2 + z^2$, find the directional derivative of f at $a = (2, 1, 3)$ in the direction $v = (1, 0, -2)$, we have

$$\nabla f(x, y, z) = (4x, 6y, 2z)$$

$$\nabla f(2, 1, 3) = (8, 6, 6)$$

$$d_{v,f}^{(a)} = \langle (8, 6, 6), (1, 0, -2) \rangle = -4$$

in the other hand

$$\lim_{t \rightarrow 0} \frac{f((2, 1, 3) + (t, 0, -2t)) - f(2, 1, 3)}{t} = -4$$

Next we show how matrices arrive in connection with derivations, we shall see that if $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at $a \in \Omega$, then the partial derivative of its component functions $\frac{\partial f_j}{\partial x_i}(a)$ exist and determine the linear transformation $d_f^{(a)}$ completely

Theorem 3.1.2

Let Ω be an open set in \mathbb{R}^n and $f : \Omega \rightarrow \mathbb{R}^m$ be differentiable at $a \in \Omega$, then $\frac{\partial f_j}{\partial x_i}(a)$ exist and the standard matrix representation of $d_f^{(a)}$ is the $m \times n$ matrix, given by :

$$d_f^{(a)} = \left(\frac{\partial f_j}{\partial x_i} \right)_{1 \leq j \leq m} \quad 1 \leq i \leq n$$

Proof. Let $d_f^{(a)} = (d_{f_1}^{(a)}, \dots, d_{f_m}^{(a)}) = T$, and let $\{e_1, \dots, e_n\}$ and $\{u_1, \dots, u_m\}$ be the standard bases of \mathbb{R}^n and \mathbb{R}^m respectively, by the definition of a linear map the j^{th} entry of the standard of the standard matrix of T , say c_{ji} is given by the j^{th} component of the vector

$$T(e_i) = \sum_{j=1}^m c_{ji} u_j$$

since f is differentiable at a , then f_j are differentiable for all $j = 1, \dots, m$, hence each $\frac{\partial f_j}{\partial x_i}(a)$ exist and

$$\frac{\partial f_j}{\partial x_i}(a) = \langle \nabla f_j(a), e_i \rangle = d_{f_j}^{(a)}(e_i)$$

so we have

$$T(e_i) = \sum_{j=1}^m d_{f_j}^{(a)}(e_i) u_j = \sum_{j=1}^m \frac{\partial f_j}{\partial x_i}(a) u_j$$

then $c_{ji} = \frac{\partial f_j}{\partial x_i}(a)$, therefore we can write our matrix :

$$\begin{pmatrix} \frac{\partial f_1}{\partial x_1}(a) & \dots & \frac{\partial f_1}{\partial x_n}(a) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \dots & \frac{\partial f_m}{\partial x_n}(a) \end{pmatrix} = \begin{pmatrix} d_{f_1}^{(a)} \\ \vdots \\ d_{f_n}^{(a)} \end{pmatrix}$$

□

Definition 3.1.3

The standard matrix $d_f^{(a)}$ is called the Jacobian matrix of f at a .

$$d_f^{(a)} = \begin{pmatrix} \frac{\partial f}{\partial x_1}(a) & \dots & \frac{\partial f}{\partial x_n}(a) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \dots & \frac{\partial f_m}{\partial x_n}(a) \end{pmatrix} = \begin{pmatrix} d_{f_1}^{(a)} \\ \vdots \\ d_{f_n}^{(a)} \end{pmatrix}$$

and reduces the problem of computing the differential of differentiable function f to that of computing the partial derivatives of it's components f_1, \dots, f_n

Definition 3.1.4

The matrix $d_f^{(a)}$ is called the Jacobian matrix of f at a , where $m = n$, the determinant of $d_f^{(a)}$ is called the Jacobian of f at a and is denoted $\mathcal{J}_f^{(a)}$, where

$$\mathcal{J}_f^{(a)} = \det(d_f^{(a)})$$

Example Let

$$\begin{aligned} f : \quad \mathbb{R}^3 &\longrightarrow \mathbb{R}^2 \\ (x, y, z) &\longmapsto \left(\frac{x^2+y^2-z^2}{2}, xy \right) \end{aligned}$$

what is $d_f^{(3,1,0)}$? let's find it out!

$$d_f^{(x,y,z)} = \begin{pmatrix} x & y & -z \\ y & x & 0 \end{pmatrix} \implies d_f^{(3,1,0)} = \begin{pmatrix} 3 & 1 & 0 \\ 1 & 3 & 0 \end{pmatrix}$$

Corollary 3.1.3(Chain Rule) let $f : \mathbb{R}^n \longrightarrow \mathbb{R}^m$ and let $g : \mathbb{R}^m \longrightarrow \mathbb{R}^p$

$$d_{g \circ f}^{(a)} = d_g^{(f(a))} \cdot d_f^{(a)} = \underbrace{\begin{pmatrix} \frac{\partial g_1}{\partial y_1}(f(a)) & \dots & \frac{\partial g_1}{\partial y_m}(f(a)) \\ \vdots & \ddots & \vdots \\ \frac{\partial g_p}{\partial y_1}(f(a)) & \dots & \frac{\partial g_p}{\partial y_m}(f(a)) \end{pmatrix}}_{(p \times m)} \underbrace{\begin{pmatrix} \frac{\partial f_1}{\partial x_1}(a) & \dots & \frac{\partial f_1}{\partial x_n}(a) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \dots & \frac{\partial f_m}{\partial x_n}(a) \end{pmatrix}}_{(m \times n)}$$

Example Let $f(x, y) = (e^{x+y}, e^{x-y})$ and $\phi : \mathbb{R} \longrightarrow \mathbb{R}^2$ a curve in \mathbb{R}^2 with $\phi(0) = (0, 1)$

and $\phi'(0) = (1, 1)$, find the tangent vector to the image of the curve $\phi(t)$ under f at $t = 0$

let

$$\begin{aligned} \Psi : \quad \mathbb{R} &\longrightarrow \mathbb{R}^2 \\ x &\longmapsto f \circ \phi(t) \end{aligned}$$

we clearly have

$$\Psi'(t) = d_f^{(\phi(t))} \cdot \phi'(t)$$

and

$$\Psi'(0) = d_f^{(\phi(0))} \cdot \phi'(0) = d_f^{(0,1)} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

calculating $d_f^{(0,1)}$ separately

$$d_f^{(0,1)} = \begin{pmatrix} e & e \\ e^{-1} & -e^{-1} \end{pmatrix}$$

hence we find :

$$\Psi'(0) = \begin{pmatrix} e & e \\ e^{-1} & -e^{-1} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 2e \\ 0 \end{pmatrix}$$

Theorem 3.1.3 Differentiability Criterion

Let $f : \Omega \rightarrow \mathbb{R}^n$ where Ω is an open set in \mathbb{R}^n , if all partial derivative $\frac{\partial f_j}{\partial x_i}$ for $j = \overline{1, m}, i = \overline{1, n}$ exist in a neighborhood of $a \in \Omega$ and are continuous, then f is differentiable at a

Proof. It suffies to prove the result for a real-valued function f , we must show that :

$$\lim_{x \rightarrow a} \frac{|f(x) - f(a) - \langle \nabla f(a), x - a \rangle|}{\|x - a\|} = 0$$

we let

$$g_i(t) = f(x_1, \dots, x_{i-1}, t, a_{i+1}, \dots, a_m)$$

and let

$$v_i = \begin{cases} v_0 = a & i = 0 \\ (x_1, \dots, x_i, a_{i+1}, \dots, a_m) & i = \overline{1, n} \end{cases}$$

if $x \in B(a, r)$ for all $i = \overline{1, n}$, then we have

$$v_i \in B(a, r)$$

the function g_i maps the interval between a_i and x_i to \mathbb{R} and

$$g'_i(t) = \frac{\partial f}{\partial x_i}(x_1, \dots, x_{i+1}, t, a_{i+1}, \dots, a_m)$$

by the one variable Mean Value Theorem, there is ζ_i in the interval between x_i and a_i such that

$$g_i(x_i) - g_i(a_i) = (x_i - a_i) g'_i(\zeta_i) = (x_i - a_i) \frac{\partial f}{\partial x_i}(x_1, \dots, \zeta_i, \dots, a_m)$$

$$g_i(x_i) - g_i(a_i) = f(v_i) - f(v_{i-1})$$

we have

$$\begin{aligned} f(x) - f(a) - \langle \nabla f(a), x - a \rangle &= \sum_{i=1}^n (f(v_i) - f(v_{i-1})) - \sum_{i=1}^n \frac{\partial f}{\partial x_i}(a)(x_i - a_i) \\ &= \sum_{i=1}^n (x_i - a_i) \left(\frac{\partial f}{\partial x_i}(x_1, \dots, \zeta_i, \dots, a_m) - \frac{\partial f}{\partial x_i}(a) \right) \rightarrow 0 \end{aligned}$$

and

$$\begin{aligned} |f(x) - f(a) - \langle \nabla f(a), x - a \rangle| &\leq \sum_{i=1}^n |x_i - a_i| \left| \frac{\partial f}{\partial x_i}(\zeta_i) - \frac{\partial f}{\partial x_i}(a) \right| \\ &\leq \|x - a\| \sum_{i=1}^n \left| \frac{\partial f}{\partial x_i}(\zeta_i) - \frac{\partial f}{\partial x_i}(a) \right| \end{aligned}$$

then

$$0 \leq \frac{|f(x) - f(a) - \langle \nabla f(a), x - a \rangle|}{\|x - a\|} \leq \sum_{i=1}^n \left| \frac{\partial f}{\partial x_i}(x_1, \dots, x_{i-1}, \zeta_i, a_{i+1}, \dots, a_m) - \frac{\partial f}{\partial x_i}(a) \right|$$

by the continuity of $\frac{\partial f}{\partial x_i}$ for all $i = 1, \dots, n$ at a we get

$$\lim_{x \rightarrow a} \sum_{i=1}^n \left| \frac{\partial f}{\partial x_i}(x_1, \dots, \zeta_i, \dots, a_m) - \frac{\partial f}{\partial x_i}(a) \right| = 0$$

hence we get the result

$$\lim_{x \rightarrow a} \frac{\partial f}{\partial x_i}(x_1, \dots, x_{i-1}, \zeta_i, \dots, a_m) = \frac{\partial f}{\partial x_i}(a)$$

□

Example Let

$$f(x, y) = \begin{cases} (x^2 + y^2) \sin\left(\frac{1}{\sqrt{x^2+y^2}}\right) & (x, y) \neq (0, 0) \\ 0 & (x, y) = (0, 0) \end{cases}$$

f is differentiable at each point $(x, y) \in \mathbb{R}^2 \setminus (0, 0)$

$$\begin{aligned} \forall x, y \in \mathbb{R}^2 \setminus \{(0, 0)\} \quad \frac{\partial f}{\partial x}(x, y) &= 2y \sin\left(\frac{1}{\sqrt{x^2+y^2}}\right) - \frac{y}{\sqrt{x^2+y^2}} \cos\left(\frac{1}{\sqrt{x^2+y^2}}\right) \\ \frac{\partial f}{\partial x}(0, 0) &= \lim_{t \rightarrow 0} \frac{f(t, 0) - f(0, 0)}{t} = 0 \\ \frac{\partial f}{\partial y}(0, 0) &= \lim_{t \rightarrow 0} \frac{f(0, t) - f(0, 0)}{t} = 0 \end{aligned}$$

hence we get

$$\frac{\partial f}{\partial x}(x, y) = \begin{cases} 2x \sin\left(\frac{1}{\sqrt{x^2+y^2}}\right) - \frac{x}{\sqrt{x^2+y^2}} & (x, y) \neq (0, 0) \\ 0 & (x, y) = (0, 0) \end{cases}$$

we have $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$ are not continuous at $(0, 0)$, however

$$\begin{aligned} \frac{|f(x, y) - f(0, 0) - \langle \nabla f(0, 0), (x, y) \rangle|}{\|(x, y)\|} &= \frac{(x^2 + y^2) \sin\left(\frac{1}{\sqrt{x^2+y^2}}\right)}{\|(x, y)\|} \\ &= \frac{(x^2 + y^2)}{\sqrt{x^2 + y^2}} \sin\left(\frac{1}{\sqrt{x^2 + y^2}}\right) \\ &= \sqrt{x^2 + y^2} \sin\left(\frac{1}{\sqrt{x^2 + y^2}}\right) \rightarrow 0 \text{ as } (x, y) \rightarrow 0 \end{aligned}$$

then the function f is differentiable at $(0, 0)$

Theorem 3.1.4 The Mean Value Theorem

Suppose $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$ is a differentiable function on the convex set Ω , let $a, b \in \Omega$ and $\zeta(t) = ta + (1-t)b$, then there exist $c \in \zeta((0, 1))$ such that

$$f(b) - f(a) = \langle \nabla f(c), a - b \rangle$$

Proof. Let

$$\phi(t) = f \circ \zeta(t)$$

we have for all t

$$\begin{aligned} \phi'(t) &= d_f^{(\zeta(t))} \zeta'(t) = \langle \nabla f(\zeta(t)), \zeta'(t) \rangle \\ &= \langle \nabla f(\zeta(t)), a - b \rangle \end{aligned}$$

by the one variable mean value, we have

$$\begin{aligned} \phi(1) - \phi(0) &= \phi'(\mu) \quad \mu \in (0, 1) \\ \phi(1) - \phi(0) &= f(\zeta(1)) - f(\zeta(0)) \\ &= f(a) - f(b) = \langle \nabla f(\zeta(\mu)), a - b \rangle \end{aligned}$$

take $\zeta(\mu) = c$, we get

$$f(a) - f(b) = \langle \nabla f(c), a - b \rangle$$

□

Corollary 3.1.4 Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$ be differentiable function on a convex subset K of Ω , if

$$\|\nabla f(x)\| \leq M \quad \forall x \in K$$

then

$$|f(x) - f(y)| \leq M\|x - y\| \quad \forall x, y \in K$$

Proof. Let $x, y \in K$, and K is convex subset of Ω , then there is $c_{x,y}$ such that

$$f(x) - f(y) = \langle \nabla f(c_{x,y}), x - y \rangle$$

by Cauchy-Schwartz inequality we get

$$|f(x) - f(y)| \leq \|\nabla f(c_{x,y})\| \|x - y\| \leq M\|x - y\| \quad \forall x, y \in K$$

□

Corollary 3.1.5 Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$ with Ω is a convex set in \mathbb{R}^n , if

$$\nabla f(x) = 0 \quad \forall x \in \Omega$$

then f is constant

Proof. let $a \in \Omega$, by the Mean Value Theorem for all $x \in \Omega$ there is c_x such that

$$f(x) - f(a) = \langle \nabla f(c_x), x - a \rangle = 0 \implies f(x) = f(a) \quad \forall x \in \Omega$$

□

Corollary 3.1.6 let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$, with Ω an open connected set in \mathbb{R}^n , if

$$\nabla f(x) = 0 \quad \forall x \in \Omega$$

then f is constant in Ω

Proof. Let $a \in \Omega$, we define

$$A = \{x \in \Omega : f(x) = f(a)\}$$

it suffies to show that A is closed and open at the same time (to conclude that $A = \Omega$), it's clear that A is closed since

$$f^{-1}(\{f(a)\}) = A$$

and $\{f(a)\}$ is closed, and f is continuous, we show that A is open set, let $x \in A \subset \Omega$, there exist $\delta > 0$ then there exist $B(x, \delta) \subset \Omega$, we let $y \in B(x, \delta)$, since $B(x, \delta)$ is convex, then by the above corollary f is constant in $B(x, \delta)$, then $f(y) = f(x) = f(a)$, then $y \in A$, therefore we get

$$B(x, \delta) \subset A \implies A \text{ is open set}$$

then $A = \Omega$, because $a \in A$ i.e $(A \neq \emptyset)$ \square

Corollary 3.1.7 *Let $f : \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ where Ω is open connected set in \mathbb{R}^n if*

$$d_f^{(x)} = 0 \quad \forall x \in \Omega$$

then f is constant in Ω

Corollary 3.1.8 *Let Ω be an open bounded convex set in \mathbb{R}^n and $f : \overline{\Omega} \rightarrow \mathbb{R}^m$ with $f \in \mathcal{C}^1(\overline{\Omega})$, then there exist a constant $M > 0$ such that for all $x, y \in \overline{\Omega}$*

$$\|f(x) - f(y)\| \leq M\|x - y\|$$

Proof. Let $f = (f_1, \dots, f_n)$, since Ω is convex then for all $x, y \in \Omega$, we apply Mean Value Theorem to each component and by cauchy schwartz inequality we have

$$\begin{aligned} \|f(x) - f(y)\|^2 &= \sum_{j=1}^m |f_j(x) - f_j(y)|^2 = \sum_{j=1}^m \left| \langle \nabla f_j(c_j), x - y \rangle \right| \\ &\leq \sum_{j=1}^m \|\nabla f_j(c_j)\|^2 \|x - y\|^2 \\ &\leq \|x - y\|^2 \sum_{j=1}^m \|\nabla f_j(c_j)\|^2 \end{aligned}$$

for all $x \in \Omega$,

$$\nabla f_j(c_j) \leq M_j$$

$\overline{\Omega}$ is compact set, then for all $x, y \in \Omega$

$$\|f(x) - f(y)\| \leq M\|x - y\| \quad \text{with} \quad M = \sqrt{\sum_{j=1}^m M_j}$$

\square