

Analyse Mathématique

Author: CatMono

Date: July, 2025

Version: 0.1

Contents

Preface		iii
Chapter	1 Preliminaries	1
1.1	Trigonometric Formulas	1
Chapter	2 Limits of Sequences and Continuity of Real Number System	3
2.1	Convergent Sequences	3
2.2	Indeterminate Form	3
2.3	Subsequences	4
2.4	Completeness of The Real Numbers	4
2.5	Iterative Sequences	5
Chapter	23 Limits and Continuity of Functions	6
3.1	Limits of Functions	6
3.2	Continuous Functions	6
3.3	Infinitesimal and Infinite Quantities	6
3.4	Continuous Functions on Closed Intervals	6
3.5	Period Three Implies Chaos	7
3.6	Functional Equations	7
Chapter	4 Differential	8
4.1	Differential and Derivative	8
4.2	Higher-Order Derivatives	8
4.3	Differential Mean Value Theorems	8
4.4	Theorems and Applications concerning Derivatives	8
4.5	Taylor Theorem	9
4.6	Applications of Taylor Theorem	9
Chapter	5 Integral	10
Chapter	6 Numerical Series	11
6.1	Convergence of Numerical Series	11
6.2	Positive Term Series	11
6.3	General Term Series	11
Chapter	7 Series of Functions	12
Chapter	8 Power Series	13
Chapter	9 Limits and Continuity in Euclidean Spaces	14
9.1	Continuous Mappings	14
Chapter	10 Multi-variable Differential Calculus	15

	CONTE	CONTENTS	
10.1	Directional Derivatives and Total Differential	15	
10.2	Differential of Vector-Valued Functions	20	
10.3	Derivatives of Composite Mappings (Chain Rule)	21	
10.4	Mean Value Theorem and Taylor's Formula	21	
10.5	Implicit Function Theorem	23	
Chapter	11 Multiple Integrals	26	

Preface

This is the preface of the book...

Chapter 1 Preliminaries

1.1 Trigonometric Formulas

Product-to-Sum Formulas:

$$\sin \alpha \cos \beta = \frac{1}{2} \left[\sin(\alpha + \beta) + \sin(\alpha - \beta) \right]$$

$$\cos \alpha \sin \beta = \frac{1}{2} \left[\sin(\alpha + \beta) - \sin(\alpha - \beta) \right]$$

$$\cos \alpha \cos \beta = \frac{1}{2} \left[\cos(\alpha + \beta) + \cos(\alpha - \beta) \right]$$

$$\sin \alpha \sin \beta = -\frac{1}{2} \left[\cos(\alpha + \beta) - \cos(\alpha - \beta) \right]$$

Sum and Difference Formulas:

$$\sin(\alpha + \beta) = \sin \alpha \cos \beta + \cos \alpha \sin \beta$$
$$\sin(\alpha - \beta) = \sin \alpha \cos \beta - \cos \alpha \sin \beta$$
$$\cos(\alpha + \beta) = \cos \alpha \cos \beta - \sin \alpha \sin \beta$$
$$\cos(\alpha - \beta) = \cos \alpha \cos \beta + \sin \alpha \sin \beta$$

Sum-to-Product Formulas:

$$\sin \alpha + \sin \beta = 2 \sin \left(\frac{\alpha + \beta}{2}\right) \cos \left(\frac{\alpha - \beta}{2}\right)$$

$$\sin \alpha - \sin \beta = 2 \sin \left(\frac{\alpha - \beta}{2}\right) \cos \left(\frac{\alpha + \beta}{2}\right)$$

$$\cos \alpha + \cos \beta = 2 \cos \left(\frac{\alpha + \beta}{2}\right) \cos \left(\frac{\alpha - \beta}{2}\right)$$

$$\cos \alpha - \cos \beta = -2 \sin \left(\frac{\alpha + \beta}{2}\right) \sin \left(\frac{\alpha - \beta}{2}\right)$$

Double Angle Formulas:

$$\sin 2\alpha = 2\sin \alpha \cos \alpha$$

$$\cos 2\alpha = \cos^2 \alpha - \sin^2 \alpha = 2\cos^2 \alpha - 1 = 1 - 2\sin^2 \alpha$$

$$\tan 2\alpha = \frac{2\tan \alpha}{1 - \tan^2 \alpha}$$

Half Angle Formulas:

$$\sin \frac{\alpha}{2} = \pm \sqrt{\frac{1 - \cos \alpha}{2}}$$

$$\cos \frac{\alpha}{2} = \pm \sqrt{\frac{1 + \cos \alpha}{2}}$$

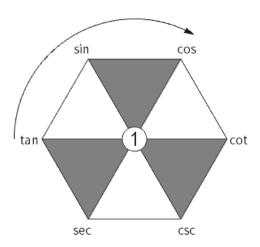
$$\tan \frac{\alpha}{2} = \frac{1 - \cos \alpha}{\sin \alpha} = \frac{\sin \alpha}{1 + \cos \alpha}$$

Power-Reducing Formulas:

$$\sin^2 \alpha = \frac{1 - \cos 2\alpha}{2}$$
$$\cos^2 \alpha = \frac{1 + \cos 2\alpha}{2}$$

Angle Decomposition Formulas:

$$\sin^2 \alpha - \sin^2 \beta = \sin(\alpha + \beta)\sin(\alpha - \beta)$$
$$\cos^2 \alpha - \sin^2 \beta = \cos(\alpha + \beta)\cos(\alpha - \beta)$$



Remark

- On the gray triangle, the sum of the squares of the two numbers above is equal to the square of the number below, for instance, $\tan^2 x + 1 = \sec^2 x$
- The three trigonometric functions in the clockwise direction have the following properties: $\tan x = \frac{\sin x}{\cos x}$, etc.

Chapter 2 Limits of Sequences and Continuity of Real Number System

2.1 Convergent Sequences

- ¶ Convergent Sequences
- ¶ Properties of Convergent Sequences
- ¶ Cauchy Proposition and Fitting Method

Proposition 2.1 (Cauchy Proposition)

Let $\lim_{n\to\infty} x_n = l$, then:

$$\lim_{n \to \infty} \frac{x_1 + x_2 + \dots + x_n}{n} = l.$$



- 1. In the proposition, l can be $+\infty$ or $-\infty$.
- 2. Let $\lim_{n\to\infty} x_n = l$, then:

$$\lim_{n \to \infty} \frac{x_1 + x_2 + \dots + x_n}{n} = \lim_{n \to \infty} \sqrt[n]{x_1 x_2 \cdots x_n} = \lim_{n \to \infty} \frac{n}{\frac{1}{x_1} + \frac{1}{x_2} + \dots + \frac{1}{x_n}} = l.$$

It can be proved directly by Stolz theorem 2.1. On top of that, it can also be proved by the **fitting method**.



Remark To prove $\lim_{n\to\infty} x_n = A$, the key is to show that $|x_n - A|$ can be arbitrarily small. For this purpose, it is generally recommended to simplify the expression of x_n as much as possible. However, in some cases, A can also be transformed into a form similar to x_n . This method is called the fitting method. The core idea behind the method of fitting is to appropriately divide into units of 1 for analysis.

2.2 Indeterminate Form

- ¶ Infinitely Large Quantities and Infinitesimal Quantities
- ¶ Indeterminate Forms

Theorem 2.1 (Stolz-Cesàro theorem

Type $\frac{0}{0}$ Let $\{a_n\}$, $\{b_n\}$ be two infinitesimal sequences, where $\{a_n\}$ is also a strictly monotonic decreasing sequence. If

$$\lim_{n\to\infty}\frac{b_{n+1}-b_n}{a_{n+1}-a_n}=l\ (\text{finite or }\pm\infty),$$

then

$$\lim_{n \to \infty} \frac{a_n}{b_n} = l.$$

Type $\frac{*}{\infty}$ Let $\{a_n\}$ be a strictly monotonic increasing sequence of divergent large quantities. If

$$\lim_{n\to\infty}\frac{b_{n+1}-b_n}{a_{n+1}-a_n}=l\ (\text{finite or }\pm\infty),$$

then

$$\lim_{n\to\infty}\frac{a_n}{b_n}=l.$$



Note

- 1. The inverse proposition of Stolz's Theorem does not hold.
- 2. If a_1 is an undefined infinite quantity ∞ , Stolz Theorem does not hold.

Theorem 2.2 (Silverman-Toeplitz Theorem)

Let

$$\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \\ \vdots \end{bmatrix} = \begin{bmatrix} a_{11} & 0 & \cdots & 0 \\ a_{21} & a_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \\ \vdots \end{bmatrix},$$

where the infinite triangular matrix satisfies:

- 1. $\forall j, \lim_{n\to\infty} a_{nj} = 0$. (Every column sequence converges to 0.)
- 2. $\sup_{i\in\mathbb{N}}\sum_{j=1}^{i}|a_{ij}|<\infty.$ (The absolute row sums are bounded.)

And $\lim_{n\to\infty} x_n = l$. We denote y_n as the weighted sum sequence: $y_n = \sum_{j=1}^n a_{nj}x_j$. Then the following results hold:

- 1. If l = 0, then $\lim_{n \to \infty} y_n = 0$.
- 2. If $l \neq 0$ and $\lim_{n \to \infty} \sum_{j=1}^n a_{ij} = 1$, then $\lim_{n \to \infty} y_n = l$.



2.3 Subsequences

- ¶ Subsequences
- ¶ Upper Limits and Lower Limits

2.4 Completeness of The Real Numbers

- ¶ Dedkind Completeness
- \P Least Upper Bound Property
- ¶ Monotone Convergence Theorem
- \P Bolzano-Weierstrass Theorem
- ¶ Nested Interval Theorem
- ¶ Cauchy Completeness

Definition 2.1 (Cauchy Sequence)

A sequence $\{x_n\}$ is called a **Cauchy sequence** if for any $\varepsilon > 0$, there exists a positive integer N such that when m, n > N,

$$|x_n - x_m| < \varepsilon$$
.



Theorem 2.3 (Cauchy Convergence Criterion for Sequences)

A sequence $\{x_n\}$ converges if and only if it is a Cauchy sequence.

\Diamond

\P Heine-Borel Theorem

2.5 Iterative Sequences

Formally, x_0 is a **fixed point** of the function f if $f(x_0) = x_0$.

Theorem 2.4 (Banach Fixed-Point Theorem (Contraction Mapping Theorem),

There exists a contraction mapping (in 3.2) f on an interval I, which admits a unique fixed point $x^* \in I$. Furthermore, x^* can be found as follows: start with an arbitrary point $x_0 \in I$ and define the iterative sequence $x_{n+1} = f(x_n)$ for $n = 0, 1, 2, \cdots$. Then $\lim_{n \to \infty} x_n = x^*$.

Remark The following inequalities are equivalent and describe the speed of convergence:

$$|x_n - x^*| \le \frac{L^n}{1 - L} |x_1 - x_0|,$$

 $|x_{n+1} - x^*| \le \frac{L}{1 - L} |x_{n+1} - x_n|,$
 $|x_{n+1} - x^*| \le L |x_n - x^*|.$

Any such value of L < 1 is the Lipschitz constant for f, and the smallest one is sometimes called **the best** Lipschitz constant of L.

Chapter 3 Limits and Continuity of Functions

3.1 Limits of Functions

- ¶ Definition of Limit
- ¶ Limits of Functions and Sequences

Theorem 3.1 (Heine Theorem

Let f be a function defined on a deleted neighborhood $\mathring{U}(x_0)$ of x_0 . The following two statements are equivalent:

- 1. $\lim_{x \to x_0} f(x) = A$.
- 2. For any sequence $\{x_n\} \subset \mathring{U}(x_0)$ with $\lim_{n\to\infty} x_n = x_0$, we have $\lim_{n\to\infty} f(x_n) = A$ for the sequence $\{f(x_n)\}$.

3.2 Continuous Functions

3.3 Infinitesimal and Infinite Quantities

3.4 Continuous Functions on Closed Intervals

¶ Concerning Theorems

Theorem 3.2 (The Bolzano-Cauchy Intermediate-Value Theorem)

 \Diamond

Theorem 3.3 (Zero Point Existence Theorem)

 \Diamond

 \P Uniform Continuity and Lipschitz Continuity

Definition 3.1 (Uniform Continuity)



Theorem 3.4 (Uniform Continuity Theorem



Theorem 3.5 (Cantor's Theorem



Definition 3.2 (Lipschitz Continuity)

If there exists a constant L>0 such that for any $x_1,x_2\in I$,

$$|f(x_1) - f(x_2)| \le L |x_1 - x_2|,$$

then f is called **Lipschitz continuous** on I.

Specially, if L < 1, then f is called a **contraction mapping** on I.

Remark

- If f is Lipschitz continuous on I, then f is uniformly continuous on I. ($\forall \varepsilon>0$, just let $\delta=\frac{\varepsilon}{L}$)
- $\bullet\,$ If f is uniformly continuous on I, then f is continuous on I.
- The converse of the above two statements does not hold.

3.5 Period Three Implies Chaos

3.6 Functional Equations

Chapter 4 Differential

4.1 Differential and Derivative

4.2 Higher-Order Derivatives

4.3 Differential Mean Value Theorems

Definition 4.1 (Extremum)

Let f(x) is defined on $(a,b), x_0 \in (a,b)$. If there exists $U(x_0,\delta) \subset (a,b)$ such that $f(x) \leqslant f(x_0)$ on it, then x_0 is called a local maximum point of f, and $f(x_0)$ is referred to as the corresponding local maximum value. The definition of the minimum value is analogous.

Lemma 4.1 (Fermat's Lemma)

If f is differentiable at x_0 which is a local extremum, then $f'(x_0) = 0$.

Theorem 4.1 (Rolle's Theorem

If $f \in C[a,b]$, $f \in D(a,b)$ and f(a) = f(b), then there exists $\xi \in (a,b)$ such that $f'(\xi) = 0$. Enhanced Version: If $f \in D(a,b)$ (finite or infinite interval), and $\lim_{x \to a^+} f(x) = \lim_{x \to b^-} f(x)$, then there exists $\xi \in (a,b)$ such that $f'(\xi) = 0$.

Theorem 4.2 (Lagrange's Mean Value Theorem,

If $f \in C[a, b]$, $f \in D(a, b)$, then there exists $\xi \in (a, b)$ such that

$$f'(\xi) = \frac{f(b) - f(a)}{b - a}.$$

Theorem 4.3 (Cauchy's Mean Value Theorem

If $f,g\in C[a,b], f,g\in D(a,b)$ and $g'(x)\neq 0$ for all $x\in (a,b)$, then there exists $\xi\in (a,b)$ such that

$$\frac{f'(\xi)}{g'(\xi)} = \frac{f(b) - f(a)}{g(b) - g(a)}.$$

4.4 Theorems and Applications concerning Derivatives

Theorem 4.4 (Darboux's Intermediate Value Theorem for Derivatives)

If $f(x) \in D[a, b]$, and $f'_+(a) \cdot f'_-(b) < 0$, then there at least exists $\xi \in (a, b)$ such that $f'(\xi) = 0$.

Theorem 4.5 (Theorem on the Limit of Derivatives)

If $f(x) \in C(U(x_0))$, $\mathring{D}(U(x_0))$, and $\lim_{x \to x_0} f'(x) = A$, then f is differentiable at x_0 and $f'(x_0) = A$.

Remark In fact, $\lim_{x o x_0}f'(x)=A$ has already been shown to imply that $f\in \mathring{D}(U(x_0)).$

- **4.5** Taylor Theorem
- **4.6 Applications of Taylor Theorem**

Chapter 5 Integral

Chapter 6 Numerical Series

6.1 Convergence of Numerical Series

6.2 Positive Term Series

6.3 General Term Series

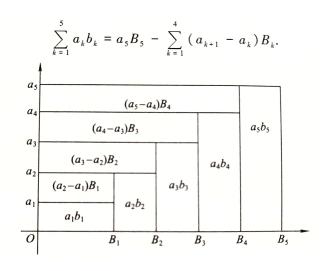
- ¶ Cauchy Convergence Criterion for Series
- ¶ Alternative Series
- ¶ Abel-Dirichlet Test

Theorem 6.1 (Abel Transform (Discrete Integration by Parts/Summation by Parts)

Let $\{a_n\}, \{b_n\}$ be two sequences, then for any $n \in \mathbb{N}^+$,

$$\sum_{k=1}^{n} a_k b_k = a_n B_n + \sum_{k=1}^{n-1} (a_{k+1} - a_k) B_k,$$

where $B_n = \sum_{k=1}^n b_k$.



Chapter 7 Series of Functions

Chapter 8 Power Series

Chapter 9 Limits and Continuity in Euclidean Spaces

9.1 Continuous Mappings

- Continuous Mappings on Compact Sets
- Continuous Mappings on Connected Sets

Definition 9.1 (Connected Set)

Let S be a set of points in \mathbb{R}^n . If a continuous mapping

$$\gamma:[0,1]\to\mathbb{R}^n$$

satisfies that the range of $\gamma([0,1])$ lies entirely within S, we call γ a path in S, where $\gamma(0)$ and $\gamma(1)$ are referred to as the starting point and ending point of the path, respectively.

If for any two points $\mathbf{x}, \mathbf{y} \in S$, there exists a path in S with \mathbf{x} as the starting point and \mathbf{y} as the ending point, Sis called path-connected, or equivalently, S is called a connected set.

A connected open set is called an (open) region. The closure of an (open) region is referred to as a closed region.

Remark Intuitively, this means that any two points in S can be connected by a curve lying entirely within S. Clearly, a connected subset of \mathbb{R} is an interval, and a connected subset of \mathbb{R} is compact if and only if it is a closed interval.

Chapter 10 Multi-variable Differential Calculus

10.1 Directional Derivatives and Total Differential

Directional Derivative ¶

Definition 10.1 (Directional Derivative)

Let $U \subset \mathbb{R}^n$ be an open set, $f: U \to \mathbb{R}^1$, **e** is a unit vector in \mathbb{R}^n , $\mathbf{x}^0 \in U$. Define

$$u(t) = f(\mathbf{x}^0 + t\mathbf{e}).$$

If the derivative of u at t = 0

$$u'(0) = \lim_{t \to 0} \frac{u(t) - u(0)}{t} = \lim_{t \to 0} \frac{f(\mathbf{x}^0 + t\mathbf{e}) - f(\mathbf{x}^0)}{t}$$

exists and is finite, it is called the directional derivative of f at \mathbf{x}^0 in the direction e, denoted by $\frac{\partial f}{\partial \mathbf{e}}(\mathbf{x}^0)$. It is the rate of change of f at \mathbf{x}^0 in the direction \mathbf{e} .

Consider the following set of unit coordinate vectors: $\mathbf{e}_1, \mathbf{e}_2, \cdots, \mathbf{e}_n$. Let $\mathbf{e}_i = (0, 0, \cdots, 0, 1, 0, \cdots, 0)$ denote the standard orthonormal basis in \mathbb{R}^n , where the 1 appears in the *i*-th position. That is,

$$\langle \mathbf{e}_i, \mathbf{e}_j \rangle = \delta_{ij} = \begin{cases} 1, & i = j, \\ 0, & i \neq j. \end{cases}$$

For a function f, the directional derivative of f at the point \mathbf{x}_0 in the direction of \mathbf{e}_i is called the ith first-order **partial derivative** of f at \mathbf{x}^0 , denoted by

$$\frac{\partial f}{\partial x_i}(\mathbf{x}^0)$$
 or $D_i f(\mathbf{x}^0)$ or $f_{x_i}(\mathbf{x}^0)$ $(i = 1, 2, \dots, n)$.

 $\mathrm{D}_i = \frac{\partial}{\partial x_i}$ is called the ith partial differential operator ($i=1,2,\cdots,n$). Let $\mathbf{e}_i = \sum_{i=0}^n \mathbf{e}_i \cos \alpha$ be a unit vector, where $\sum_{i=0}^n \cos^2 \alpha = 1$. If $\frac{\partial f}{\partial x_i}$ is continuous at \mathbf{x}^0 , then the directional derivative of f at \mathbf{x}^0 along the direction \mathbf{e} is given by:

$$\frac{\partial f}{\partial \mathbf{e}}(\mathbf{x}^0) = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x}^0) \cos \alpha_i.$$

This is the formula for expressing a directional derivative using partial derivatives.

 $ilde{Y}$ Note Let ${f e}$ be a direction, then $\|-{f e}\|=\|{f e}\|=1$, which implies that $-{f e}$ is also a direction. At this point, we have:

$$\frac{\partial f}{\partial (-\mathbf{e})}(\mathbf{x}^0) = -\frac{\partial f}{\partial \mathbf{e}}(\mathbf{x}^0).$$

Definition 10.2 (Jacobian Matrix (Gradient))

Let

$$Jf(\mathbf{x}) = (D_1 f(\mathbf{x}), D_2 f(\mathbf{x}), \dots, D_n f(\mathbf{x})),$$

which is called the **Jacobian matrix** of the function f at the point \mathbf{x} , (a 1 \times n matrix) whose counterpart is the first-order derivative of a single-variable function.

Henceforth, we represent the point x in \mathbb{R}^n and its increments h as column vectors. In this way, the differential of the function can be expressed using matrix multiplication as follows:

$$df(\mathbf{x}^0)(\mathbf{\Delta}\mathbf{x}) = Jf(\mathbf{x}^0)\mathbf{\Delta}\mathbf{x}.$$

The Jacobian matrix of the function f is also frequently denoted as grad f (or ∇f), that is,

$$\operatorname{grad} f(\mathbf{x}) = Jf(\mathbf{x}),$$

which is called the **gradient** of the scalar function f.



- $frac{2}{3}$ Note Let $U \subset \mathbb{R}^n$ be an open set, and $\mathbf{f}: U \to \mathbb{R}^m$ be a C^k mapping:
 - k = 0, **f** is a continuous mapping;
 - $0 < k < +\infty$, f_i has continuous partial derivatives up to order k, $i = 1, 2, \dots, m$;
 - $k = +\infty$, f_i has continuous partial derivatives of all orders, $i = 1, 2, \ldots, m$;
 - $k = \omega$, f_i is really analytic, i.e., in the neighborhood of any point $\mathbf{x}^0 = (x_1^0, x_2^0, \dots, x_n^0) \in U$, f_i can be expanded into a convergent (n-dimensional) power series, $i = 1, 2, \dots, m$.

Let $C^k(U, \mathbb{R}^m)$ denote the totality of C^k mappings from U to \mathbb{R}^m .



Definition 10.3 (Total Differential)

Let $U \subset \mathbb{R}^n$ be an open set, $f: U \to \mathbb{R}^1$, $\mathbf{x}^0 \in U$, $\Delta \mathbf{x} = (\Delta x_1, \Delta x_2, \cdots, \Delta x_n) \in \mathbb{R}^n$. If

$$f(\mathbf{x}^0 + \Delta \mathbf{x}) - f(\mathbf{x}^0) = \sum_{i=1}^n A_i \Delta x_i + o(\|\Delta \mathbf{x}\|) \qquad (\|\Delta \mathbf{x}\| \to 0),$$

where A_1, A_2, \ldots, A_n are constants independent of $\Delta \mathbf{x}$, then the function f is said to be **differentiable** at the point \mathbf{x}^0 , and the linear main part $\sum_{i=1}^n A_i \Delta x_i$ is called the **total differential** of f at \mathbf{x}^0 , denoted as

$$df(\mathbf{x}^0)(\mathbf{\Delta}\mathbf{x}) = \sum_{i=1}^n A_i \Delta x_i.$$

If f is differentiable at every point in the open set U, then f is called a differentiable function on U.



Theorem 10.1 (Conditions of Differentiability)

Necessary Condition If an n-variable function f is differentiable at the point \mathbf{x}_0 , then f is continuous at \mathbf{x}^0 and possesses first-order partial derivatives $\frac{\partial f}{\partial x_i}(\mathbf{x}^0)$ at \mathbf{x}^0 for $i=1,2,\ldots,n$, and

$$\mathbf{A} = (A_1, A_2, \dots, A_n) = Jf(\mathbf{x}^0) = (D_1 f(\mathbf{x}^0), D_2 f(\mathbf{x}^0), \dots, D_n f(\mathbf{x}^0)).$$

^a However, the converse is not true.

Sufficient Condition Let $U \subset \mathbb{R}^n$ be an open set, and let $f: U \to \mathbb{R}^1$ be an n-variable function. If $Jf = (D_1 f, D_2 f, \dots, D_n f)$ is continuous at \mathbf{x}^0 (i.e., $\frac{\partial f}{\partial x_i}$ is continuous at \mathbf{x}^0 for $i = 1, 2, \dots, n$), then f is differentiable at \mathbf{x}^0 . However, the converse is not necessarily true.

^aIt is referred to as the total differential formula, and the more common form is

$$df(x_0, y_0) = \frac{\partial f}{\partial x}(x_0, y_0) dx + \frac{\partial f}{\partial y}(x_0, y_0) dy.$$



Note

- The continuity of the derivative function at \mathbf{x}^0 implies that the original function f is differentiable in some neighborhood of \mathbf{x}^0 .
- In fact, this condition can be relaxed to require that one partial derivative exists at the point, while the remaining n-1 partial derivative functions are continuous at that point.
- Proof Taking a function of three variables as an example.

Assume the 3-ary function $f: \mathbb{R}^3 \to \mathbb{R}$ meets:

- 1. There exists $f_z(x_0, y_0, z_0)$.
- 2. The partial derivative functions $f_x(x, y, z)$ and $f_y(x, y, z)$ are continuous at (x_0, y_0, z_0) , i.e. there are partial derivatives in some neighborhood of (x_0, y_0, z_0) .

Consider the total increment of f at the point (x_0, y_0, z_0) :

$$\Delta f = \underbrace{\left[f(x_0 + \Delta x, y_0 + \Delta y, z_0 + \Delta z) - f(x_0, y_0 + \Delta y, z_0 + \Delta z)\right]}_{I_1} + \underbrace{\left[f(x_0, y_0 + \Delta y, z_0 + \Delta z) - f(x_0, y_0, z_0 + \Delta z)\right]}_{I_2} + \underbrace{\left[f(x_0, y_0, z_0 + \Delta z) - f(x_0, y_0, z_0)\right]}_{I_3}.$$

For I_1,I_2 , by the Lagrange's Mean Value Theorem of unary functions, there exist $\theta_1,\theta_2\in(0,1)$ such that

$$I_{1} = f_{x}(x_{0} + \theta_{1}\Delta x, y_{0} + \Delta y, z_{0} + \Delta z)\Delta x,$$

$$I_{2} = f_{y}(x_{0}, y_{0} + \theta_{2}\Delta y, z_{0} + \Delta z)\Delta y.$$

Then by the continuity of the their partial derivatives at (x_0, y_0, z_0) , we have

$$\lim_{\Delta x, \Delta y, \Delta z \to 0} I_1 = f_x(x_0, y_0, z_0) \Delta x, \quad \lim_{\Delta x, \Delta y, \Delta z \to 0} I_2 = f_y(x_0, y_0, z_0) \Delta y.$$

They can be expressed in terms of infinitesimals($\rho = \sqrt{\Delta x^2 + \Delta y^2 + \Delta z^2}$):

$$I_1 = f_x(x_0, y_0, z_0) \Delta x + \alpha_1 \Delta x, \quad \alpha_1 \to 0 (\rho \to 0),$$

$$I_2 = f_y(x_0, y_0, z_0) \Delta y + \alpha_2 \Delta y, \quad \alpha_2 \to 0 (\rho \to 0).$$

For I_3 , by the definition of the partial derivative $f_z(x, y, z)$ at (x_0, y_0, z_0) , we have

$$I_3 = f_z(x_0, y_0, z_0)\Delta z + \alpha_3 \Delta z, \quad \alpha_3 \to 0 (\rho \to 0).$$

Accordingly,

$$\begin{split} \Delta f &= I_1 + I_2 + I_3 \\ &= \left[f_x(x_0, y_0, z_0) \Delta x + \alpha_1 \Delta x \right] + \left[f_y(x_0, y_0, z_0) \Delta y + \alpha_2 \Delta y \right] + \left[f_z(x_0, y_0, z_0) \Delta z + \alpha_3 \Delta z \right] \\ &= f_x(x_0, y_0, z_0) \Delta x + f_y(x_0, y_0, z_0) \Delta y + f_z(x_0, y_0, z_0) \Delta z + \left[\alpha_1 \Delta x + \alpha_2 \Delta y + \alpha_3 \Delta z \right]. \end{split}$$

Apparently,

$$\lim_{\rho \to 0} \frac{\alpha_1 \Delta x + \alpha_2 \Delta y + \alpha_3 \Delta z}{\rho} = 0,$$

i.e. $\alpha_1 \Delta x + \alpha_2 \Delta y + \alpha_3 \Delta z = o(\rho)$. Therefore, f(x,y,z) is differentiable at (x_0,y_0,z_0) , which completes the proof.

Note (At some point)

- 1. Differentiable
 - \Longrightarrow Continuous
 - \Longrightarrow Partial derivatives exist: $D_{\vec{u}} = \nabla f \cdot \vec{u}$
- 2. Directional Derivative
 - ullet All directional derivatives exist \Longrightarrow differentiable or continuous.
 - ullet All directional derivatives exist and are equal \Longrightarrow differentiable.
- 3. Partial Derivative
 - The continuity and existence of directional/partial derivatives are mutually exclusive.

¶ Higher-Order Partial Derivatives and Differential

If the first-order partial derivative of f, $\frac{\partial f}{\partial x_i}$, itself possesses partial derivatives, then the second-order partial derivative of f is defined, and is denoted as follows(the first is also called the mixed partial derivative):

$$f_{x_i x_j} = \frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial}{\partial x_j} \left(\frac{\partial f}{\partial x_i} \right), \quad f_{x_i x_i} = \frac{\partial^2 f}{\partial x_i^2} = \frac{\partial}{\partial x_i} \left(\frac{\partial f}{\partial x_i} \right), \quad i, j = 1, 2, \dots, n.$$

Similarly, higher-order partial derivatives of order $3, 4, \dots m, \dots$ can be defined.

The following theorem provides the conditions under which mixed partial derivatives are equal.

Theorem 10.2 (Conditions for Equality of Mixed Partial Derivatives

1. Let $U \subset \mathbb{R}^2$ be an open set, and $f: U \to \mathbb{R}$ be a function of two variables. If f_{xy} and f_{yx} are continuous at $(x_0, y_0) \in U$, then

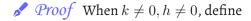
$$f_{xy}(x_0, y_0) = f_{yx}(x_0, y_0).$$

2. Let $U \subset \mathbb{R}^n$ be an open set, and $f: U \to \mathbb{R}$ be a function of n variables. If f has partial derivatives up to order k in D, and all of them are continuous at $\mathbf{x}^0 = (x_1^0, x_2^0, \dots, x_n^0) \in U$, then

$$\frac{\partial^l f}{\partial x_{i_1} \partial x_{i_2} \cdots \partial x_{i_l}}(\mathbf{x}^0) = \frac{\partial^l f}{\partial x_{i_2} \partial x_{i_1} \cdots \partial x_{i_l}}(\mathbf{x}^0) = \cdots = \frac{\partial^l f}{\partial x_{i_l} \partial x_{i_{l-1}} \cdots \partial x_{i_1}}(\mathbf{x}^0),$$

that is, the order of taking partial derivatives $l(\leq k)$ does not affect the result.

[&]quot;If the condition " f_{xy} and f_{yx} are continuous at (x_0, y_0) ", is not satisfied, then the conclusion " $f_{xy}(x_0, y_0) = f_{yx}(x_0, y_0)$ " does not necessarily hold.



$$\varphi(y) = f(x_0 + h, y) - f(x_0, y),$$

and

$$\psi(x) = f(x, y_0 + k) - f(x, y_0).$$

Applying the Lagrange Mean Value Theorem, we have

$$\begin{aligned} &[f(x_0+h,y_0+k)-f(x_0,y_0+k)]-[f(x_0+h,y_0)-f(x_0,y_0)]\\ =&\varphi(y_0+k)-\varphi(y_0)\\ =&\varphi'(y_0+\theta_1k)k\\ =&[f_y(x_0+h,y_0+\theta_1k)-f_y(x_0,y_0+\theta_1k)]k\\ =&f_{yx}(x_0+\theta_2h,y_0+\theta_1k)hk,\quad 0<\theta_1,\theta_2<1. \end{aligned}$$

On the other hand,

$$\begin{split} &[f(x_0+h,y_0+k)-f(x_0,y_0+k)]-[f(x_0+h,y_0)-f(x_0,y_0)]\\ =&[f(x_0+h,y_0+k)-f(x_0+h,y_0)]-[f(x_0,y_0+k)-f(x_0,y_0)]\\ =&\psi(x_0+h)-\psi(x_0)\\ =&\psi'(x_0+\theta_3h)h\\ =&[f_x(x_0+\theta_3h,y_0+k)-f_x(x_0+\theta_3h,y_0)]h\\ =&f_{xy}(x_0+\theta_3h,y_0+\theta_4k)hk,\quad 0<\theta_3,\theta_4<1. \end{split}$$

Therefore.

$$f_{xy}(x_0 + \theta_3 h, y_0 + \theta_4 k) = f_{yx}(x_0 + \theta_2 h, y_0 + \theta_1 k).$$

Since f_{xy} and f_{yx} are continuous at (x_0, y_0) , letting $h \to 0, k \to 0$, we obtain

$$f_{xy}(x_0, y_0) = f_{yx}(x_0, y_0).$$

By applying 10.1 and the principle of mathematical induction, one can immediately derive the following result.

Suppose z=f(x,y) has continuous partial derivatives in the domain $U\subset\mathbb{R}^2$. Then z is differentiable, and

$$\mathrm{d}z = \frac{\partial z}{\partial x} \mathrm{d}x + \frac{\partial z}{\partial y} \mathrm{d}y.$$

If z also has continuous second-order partial derivatives, then $\frac{\partial z}{\partial x}$ and $\frac{\partial z}{\partial y}$ are also differentiable, and thus $\mathrm{d}z$ is differentiable. We call the differential of $\mathrm{d}z$ the second-order differential of z, denoted as

$$d^2z = d(dz).$$

In general, based on the k-th order differential $d^k z$ of z, its (k+1)-th order differential (if it exists) is defined as

$$d^{k+1}z = d(d^kz), \quad k = 1, 2, \cdots.$$

Due to the fact that for the independent variables x and y, we always have

$$d^2x = d(dx) = 0,$$
 $d^2y = d(dy) = 0,$

the second-order differential of z = f(x, y) is given by

$$d^{2}z = d(dz) = d\left(\frac{\partial z}{\partial x}dx + \frac{\partial z}{\partial y}dy\right)$$

$$= d\left(\frac{\partial z}{\partial x}\right)dx + \frac{\partial z}{\partial x}d^{2}x + d\left(\frac{\partial z}{\partial y}\right)dy + \frac{\partial z}{\partial y}d^{2}y$$

$$= \left(\frac{\partial^{2}z}{\partial x^{2}}dx + \frac{\partial^{2}z}{\partial x\partial y}dy\right)dx + \left(\frac{\partial^{2}z}{\partial y\partial x}dx + \frac{\partial^{2}z}{\partial y^{2}}dy\right)dy$$

$$= \frac{\partial^{2}z}{\partial x^{2}}(dx)^{2} + 2\frac{\partial^{2}z}{\partial x\partial y}dxdy + \frac{\partial^{2}z}{\partial y^{2}}(dy)^{2},$$

where $(\mathrm{d}x)^2$ and $(\mathrm{d}y)^2$ denote d^2x and d^2y respectively. If we treat $\frac{\partial}{\partial x}$, $\frac{\partial}{\partial y}$ as operators for partial differentiation and define

$$\left(\frac{\partial}{\partial x}\right)^2 = \frac{\partial^2}{\partial x^2}, \quad \left(\frac{\partial}{\partial y}\right)^2 = \frac{\partial^2}{\partial y^2}, \quad \left(\frac{\partial}{\partial x}\frac{\partial}{\partial y}\right) = \frac{\partial^2}{\partial x \partial y},$$

then the formulas for the first and second differentials can be written as

$$dz = \left(dx \frac{\partial}{\partial x} + dy \frac{\partial}{\partial y}\right) z,$$
$$d^2 z = \left(dx \frac{\partial}{\partial x} + dy \frac{\partial}{\partial y}\right)^2 z.$$

Similarly, we define

$$\left(\frac{\partial}{\partial x}\right)^p \left(\frac{\partial}{\partial y}\right)^q = \frac{\partial^{p+q}}{\partial x^p \partial y^q} = \frac{\partial^q}{\partial y^q} \left(\frac{\partial}{\partial x}\right)^p, \quad (p, q = 1, 2, \dots)$$

It is easy to use mathematical induction to prove the formula for higher-order differentials:

$$\mathrm{d}^k z = \left(\mathrm{d}x \frac{\partial}{\partial x} + \mathrm{d}y \frac{\partial}{\partial y}\right)^k z, \quad k = 1, 2, \cdots.$$

For an n-variable function $u=f(x_1,x_2,\ldots,x_n)$, higher-order differentials can be similarly defined, and the

following holds:

$$d^{k}u = \left(dx_{1}\frac{\partial}{\partial x_{1}} + dx_{2}\frac{\partial}{\partial x_{2}} + \dots + dx_{n}\frac{\partial}{\partial x_{n}}\right)^{k}u, \quad k = 1, 2, \dots$$

10.2 Differential of Vector-Valued Functions

Consider an *n*-dimensional vector-valued function defined on a domain $U \subset \mathbb{R}^n$:

$$f: U \to \mathbb{R}^m,$$

 $\mathbf{x} \mapsto \mathbf{v} = f(\mathbf{x})$

Expressed in coordinate vector form:

$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{pmatrix} = \begin{pmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_m(x_1, x_2, \dots, x_n) \end{pmatrix}, \qquad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \in U$$

1. If each component function $f_i(x_1, x_2, \dots, x_n)$ ($i = 1, 2, \dots, m$) is partially differentiable at \mathbf{x}^0 , then the vector-valued function \mathbf{f} is differentiable at \mathbf{x}^0 , and we define the matrix

$$\left(\frac{\partial f}{\partial x_j}(\mathbf{x}^0)\right)_{m \times n} = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(\mathbf{x}^0) & \frac{\partial f_1}{\partial x_2}(\mathbf{x}^0) & \cdots & \frac{\partial f_1}{\partial x_n}(\mathbf{x}^0) \\ \frac{\partial f_2}{\partial x_1}(\mathbf{x}^0) & \frac{\partial f_2}{\partial x_2}(\mathbf{x}^0) & \cdots & \frac{\partial f_2}{\partial x_n}(\mathbf{x}^0) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(\mathbf{x}^0) & \frac{\partial f_m}{\partial x_2}(\mathbf{x}^0) & \cdots & \frac{\partial f_m}{\partial x_n}(\mathbf{x}^0) \end{pmatrix}$$

This matrix is called the Jacobian matrix of \mathbf{f} at \mathbf{x}^0 , denoted by $f'(\mathbf{x}^0)$ (or $\mathrm{D}f(\mathbf{x}^0)$, $J_f(\mathbf{x}^0)$).

For the special case m=1, i.e., n-variable scalar function $z=f(x_1,x_2,\ldots,x_n)$, the derivative at \mathbf{x}^0 is

$$f'(\mathbf{x}^0) = \left(\frac{\partial f}{\partial x_1}(\mathbf{x}^0), \frac{\partial f}{\partial x_2}(\mathbf{x}^0), \cdots, \frac{\partial f}{\partial x_n}(\mathbf{x}^0)\right)$$

If the vector-valued function \mathbf{f} is differentiable at every point in U, then \mathbf{f} is said to be differentiable on U, and the corresponding relationship is

$$\mathbf{x} \in U \mapsto f'(\mathbf{x}) = J_f(\mathbf{x})$$

where $f'(\mathbf{x})$ (or $Df(\mathbf{x})$, $J_f(\mathbf{x})$) denotes the derivative of \mathbf{f} at \mathbf{x} in U.

2. If every component function $f_i(x_1, x_2, ..., x_n)$ (i = 1, 2, ..., m) of \mathbf{f} has continuous partial derivatives at \mathbf{x}^0 , then every element of the Jacobian matrix of \mathbf{f} is continuous at \mathbf{x}^0 . In this case, \mathbf{f} is said to have a continuous derivative at \mathbf{x}^0 as a vector-valued function.

If the derivative of a vector-valued function \mathbf{f} is continuous at every point in U, then \mathbf{f} is said to have a continuous derivative on U.

3. If there exists an $m \times n$ matrix A that depends only on \mathbf{x}^0 (and not on $\Delta \mathbf{x}$), such that in the neighborhood of \mathbf{x}^0 ,

$$\Delta \mathbf{y} = f(\mathbf{x}^0 + \Delta \mathbf{x}) - f(\mathbf{x}^0) = A\Delta \mathbf{x} + o(\|\Delta \mathbf{x}\|)$$

(where $\Delta \mathbf{x} = (\Delta x_1, \Delta x_2, \dots, \Delta x_n)^T$ is a column vector and $\|\Delta \mathbf{x}\|$ denotes its norm), then f is said to be differentiable at \mathbf{x}^0 as a vector-valued function, and $A\Delta \mathbf{x}$ is called the differential of f at \mathbf{x}^0 , denoted as $d\mathbf{y}$. If we denote $\Delta \mathbf{x}$ by $d\mathbf{x}$ ($d\mathbf{x} = (dx_1, dx_2, \dots, dx_n)^T$), then

$$d\mathbf{v} = A d\mathbf{x}$$
.

If the vector-valued function f is differentiable at every point in U, then f is said to be differentiable on U.

Combining the above three points, we obtain the following unified statement:

A vector-valued function \mathbf{f} is continuous, differentiable, and has derivatives if and only if each of its coordinate component functions $f_i(x_1, x_2, \dots, x_n)$ ($i = 1, 2, \dots, m$) is continuous, differentiable, and has derivatives.

10.3 Derivatives of Composite Mappings (Chain Rule)

Let $U \subset \mathbb{R}^l$ and $V \subset \mathbb{R}^n$ be open sets, and let

$$\mathbf{g}: U \to V$$
 and $\mathbf{f}: V \to \mathbb{R}^m$

be mappings. If \mathbf{g} is derivative at $\mathbf{u}^0 \in U$ and \mathbf{f} is differentiable at $\mathbf{x}^0 = \mathbf{g}(\mathbf{u}^0)$, then the composite mapping $\mathbf{f} \circ \mathbf{g}$ is differentiable at \mathbf{u}^0 , and:

$$J(\mathbf{f} \circ \mathbf{g})(\mathbf{u}^0) = J\mathbf{f}(\mathbf{x}^0)J\mathbf{g}(\mathbf{u}^0).$$



- 1. outer differentiable + inner derivative = total derivative
- 2. outer differentiable + inner differentiable = total differentiable

3

Specially, define $z=f(x,y), (x,y)\subset D_f\subset \mathbb{R}^2$, $\mathbf{g}:D_g\to \mathbb{R}^2, (u,v)\mapsto (x(u,v),y(u,v))$, and $g(D_g)\subset D_f$, then we have composite function

$$z = f \circ \mathbf{g} = f[x(u, v), y(u, v)], \quad (u, v) \in D_g.$$

$$\mathbb{R}^2 \xrightarrow{\mathbf{g}: \text{derivative}} \mathbb{R}^2 \xrightarrow{f: \text{differentiable}} \mathbb{R}$$

If g is derivative at $(u_0, v_0) \in D_g$, and f is differentiable at $(x_0, y_0) = \mathbf{g}(u_0, v_0)$, then $z = f \circ \mathbf{g}$ is differentiable at (u_0, v_0) , and at the point,

$$\begin{bmatrix} \frac{\partial z}{\partial u} & \frac{\partial z}{\partial v} \end{bmatrix} = \begin{bmatrix} \frac{\partial z}{\partial x} & \frac{\partial z}{\partial y} \end{bmatrix} \begin{bmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{bmatrix}$$

A Proof

10.4 Mean Value Theorem and Taylor's Formula

Definition 10.4 (Convex Region)

Let $D \subseteq \mathbb{R}^n$ be a region. If every line segment connecting any two points $\mathbf{x}_0, \mathbf{x}_1 \in D$ (denoted by $\overline{\mathbf{x}_0}\overline{\mathbf{x}_1}$) is entirely contained in D, i.e., for any $\lambda \in [0, 1]$, we have

$$\mathbf{x}_0 + \lambda(\mathbf{x}_1 - \mathbf{x}_0) \in D$$
,

then D is called a convex region.



 \Diamond

Theorem 10.3 (Lagrange's Mean Value Theorem

Let f be <u>differentiable</u> on <u>a convex region</u> $D \subseteq \mathbb{R}^n$. For any two points $\mathbf{a}, \mathbf{b} \in D$, there exists a point $\xi \in \overline{\mathbf{ab}}$ such that:

$$f(\mathbf{b}) - f(\mathbf{a}) = Jf(\xi)(\mathbf{b} - \mathbf{a}).$$



Theorem 10.4

Let D be a region in \mathbb{R}^n . If for any $\mathbf{x} \in D$, we have

$$Jf(\mathbf{x}) = 0$$
,

then f is constant on D.

Proof

Theorem 10.5 (Taylor's Formula)

Lagrange's Remainder Let $D \subseteq \mathbb{R}^n$ be a convex region, and let $f: D \to \mathbb{R}$ have m+1 continuous partial derivatives. For $\mathbf{x}^0 = (x_1^0, x_2^0, \dots, x_n^0) \in D$ and $\mathbf{x} = (x_1, x_2, \dots, x_n) \in D$, there exists $\xi \in \overline{\mathbf{x}^0 \mathbf{x}}$ such that:

$$f(\mathbf{x}) = f(\mathbf{x}^0) + \sum_{k=1}^{m} \frac{1}{k!} \left(\sum_{i=1}^{n} (x_i - x_i^0) \frac{\partial}{\partial x_i} \right)^k f(\mathbf{x}^0) + \frac{1}{(m+1)!} \left(\sum_{i=1}^{n} (x_i - x_i^0) \frac{\partial}{\partial x_i} \right)^{m+1} f(\xi).$$

Peano's Remainder Let $D \subseteq \mathbb{R}^n$ be a convex region, and let $f: D \to \mathbb{R}$ have m continuous partial derivatives. Then:

$$f(\mathbf{x}) = f(\mathbf{x}^0) + \sum_{k=1}^m \frac{1}{k!} \sum_{i_1, i_2, \dots, i_k = 1}^n \frac{\partial^k f}{\partial x_{i_1} \partial x_{i_2} \dots \partial x_{i_k}} (\mathbf{x}^0) \prod_{j=1}^k (x_{i_j} - x_{i_j}^0) + R_m(\mathbf{x} - \mathbf{x}^0),$$
 where $R_m(\mathbf{x} - \mathbf{x}^0) = O(\|\mathbf{x} - \mathbf{x}^0\|^{m+1})$ or $o(\|\mathbf{x} - \mathbf{x}^0\|^m)$, as $\|\mathbf{x} - \mathbf{x}^0\| \to 0$.

In applications, particularly important is the expression of the first three terms in Taylor's formula, which is given as (let $x_1 - x_1^0$ be denoted by Δx_1 , and similarly for other variables; $\Delta \mathbf{x} = (\Delta x_1, \Delta x_2, \dots, \Delta x_n)$):

$$f(\mathbf{x}) = f(\mathbf{x}^0) + Jf(\mathbf{x}^0)(\Delta \mathbf{x}) + \frac{1}{2!}(\Delta \mathbf{x})Hf(\mathbf{x}^0)(\Delta \mathbf{x})^{\mathrm{T}} + \cdots,$$

where the matrix

$$Hf(\mathbf{x}^{0}) = \begin{bmatrix} \frac{\partial^{2} f}{\partial x_{1}^{2}} & \frac{\partial^{2} f}{\partial x_{1} \partial x_{2}} & \cdots & \frac{\partial^{2} f}{\partial x_{1} \partial x_{n}} \\ \frac{\partial^{2} f}{\partial x_{2} \partial x_{1}} & \frac{\partial^{2} f}{\partial x_{2}^{2}} & \cdots & \frac{\partial^{2} f}{\partial x_{2} \partial x_{n}} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^{2} f}{\partial x_{n} \partial x_{1}} & \frac{\partial^{2} f}{\partial x_{n} \partial x_{2}} & \cdots & \frac{\partial^{2} f}{\partial x_{n}^{2}} \end{bmatrix}_{\mathbf{x}^{0}}$$

is called the **Hessian matrix** of the function f. It is an $n \times n$ symmetric matrix.

10.5 Implicit Function Theorem

Theorem 10.6 (Implicit Function Theorem,

Let $U\subset\mathbb{R}^{n+1}$ be an open set, and $F:U\to\mathbb{R}$ be an n+1-variable function. If:

- 1. $F \in C^k(U, \mathbb{R})$, where $1 \le k \le +\infty$;
- 2. $F(\mathbf{x}^0, y^0) = 0$, where $\mathbf{x}^0 = (x_1^0, x_2^0, \dots, x_n^0) \in \mathbb{R}^n$, $y^0 \in \mathbb{R}$, and $(\mathbf{x}^0, y^0) \in U$ (i.e., the equation $F(\mathbf{x}, y) = 0$ has a solution (\mathbf{x}^0, y^0));
- 3. $F_y'(\mathbf{x}^0, y^0) \neq 0$.

Then there exists an open interval $I \times J$ containing (\mathbf{x}^0, y^0) (I being an open interval in \mathbb{R}^n containing \mathbf{x}^0 , and J being an open interval in \mathbb{R} containing y^0), as shown in Fig. 10.1, such that:

- 1. $\forall x \in I$, the equation $F(\mathbf{x}, y) = 0$ has a unique solution $y = f(\mathbf{x})$, where $f : I \to J$ is an n-variable function (called the **implicit function** f, hidden within the equation $F(\mathbf{x}, f(\mathbf{x})) = 0$, though not necessarily explicitly expressed);
- 2. $y^0 = f(\mathbf{x}^0);$
- 3. $f \in C^k(I, \mathbb{R})$;
- 4. When $x \in I$, $\frac{\partial f}{\partial x_i} = \frac{\partial y}{\partial x_i} = -\frac{F_x(\mathbf{x}, y)}{F_y(\mathbf{x}, y)}$, $i = 1, 2, \dots, n$, where y = f(x).

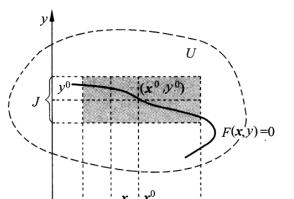


Figure 10.1: Implicit Function

Proof Only the single-variable implicit function theorem is proved; the multi-variable case can be derived using mathematical induction.

Without loss of generality, assume $F_y(x^0, y^0) > 0$.

First, prove the existence of the implicit function. From the continuity of $F_y(x^0, y^0) > 0$ and $F_y(x, y)$, it is known that there exist closed rectangle:

$$D^* = \{(x, y) \mid |x - x_0| \le \alpha, |y - y_0| \le \beta\} \subset U,$$

where the following holds:

$$F_y(x,y) > 0.$$

Thus, for fixed x_0 , the function $F(x^0, y)$ is strictly monotonically increasing within $[y^0 - \beta, y^0 + \beta]$. Furthermore, since:

$$F(x^0, y^0) = 0,$$

it follows that:

$$F(x^0, y^0 - \beta) < 0, \quad F(x^0, y^0 + \beta) > 0.$$

Due to the continuity of F(x, y) within D^* , there exists $\rho > 0$ such that along the line segment:

$$x = x^{0} + \rho, y = y^{0} + \beta,$$

we have F(x, y) > 0, and along the line segment:

$$x = x^{0} + \rho, y = y^{0} - \beta,$$

we have F(x,y) < 0. Therefore, for any point $\bar{x} \in (x^0 - \rho, x^0 + \rho)$, treat F(x,y) as a single-variable function of y. Within $[y^0 - \beta, y^0 + \beta]$, this function is continuous. From the previous discussion, we know:

$$F(\bar{x}, y^0 - \beta) < 0, \quad F(\bar{x}, y^0 + \beta) > 0.$$

According to the zero point existence theorem 3.3, there must exist a unique $\bar{y} \in [y^0 - \beta, y^0 + \beta]$ such that $F(\bar{x}, \bar{y}) = 0$. Furthermore, because $F_y(x, y) > 0$ within D^* , this \bar{y} is unique. Denote the corresponding relationship as $\bar{y} = f(\bar{x})$, then the function y = f(x) is defined within $(x^0 - \rho, x^0 + \rho)$, satisfying F(x, f(x)) = 0, and clearly:

$$y^0 = f(x^0).$$

Further proving the continuity of the implicit function y=f(x) on $(x^0-\rho,x^0+\rho)$: Let $\bar x\in(x^0-\rho,x^0+\rho)$ be any point. For any given $\varepsilon>0$ (ε being sufficiently small), since $F(\bar x,\bar y)=0$ ($\bar y=f(\bar x)$), from the previous discussion we know:

$$F(\bar{x}, \bar{y} - \varepsilon) < 0, \quad F(\bar{x}, \bar{y} + \varepsilon) > 0.$$

Furthermore, due to the continuity of F(x, y) on D^* , there exists $\delta > 0$ such that:

$$F(x, \bar{y} - \varepsilon) < 0$$
, $F(x, \bar{y} + \varepsilon) > 0$, when $x \in O(x^0, \delta)$.

By reasoning similar to the previous discussion, it can be obtained that when $x \in O(x^0, \delta)$, the corresponding implicit function value must satisfy $f(x) \in (\bar{y} - \varepsilon, \bar{y} + \varepsilon)$, i.e.,

$$\left| f(x) - f(x^0) \right| < \varepsilon.$$

This implies that y = f(x) is continuous on $(x^0 - \rho, x^0 + \rho)$.

Finally, prove the <u>differentiability</u> of y=f(x) on $(x^0-\rho,x^0+\rho)$: Let $\bar{x}\in(x^0-\rho,x^0+\rho)$ be any point. Take Δx sufficiently small such that $\bar{x}=x+\Delta x\in(x^0-\rho,x^0+\rho)$. Denote $\bar{y}=f(\bar{x})$ and $\bar{y}+\Delta y=f(\bar{x})$. Clearly,

$$F(\bar{x}, \bar{y}) = 0$$
 and $F(\bar{x}, \bar{y} + \Delta y) = 0$.

Using the multi-variable function's mean value theorem 10.3, we obtain:

$$0 = F(\bar{x}, \bar{y} + \Delta y) - F(\bar{x}, \bar{y})$$

= $F_x(\bar{x} + \theta \Delta x, \bar{y} + \theta \Delta y) \Delta x + F_y(\bar{x} + \theta \Delta x, \bar{y} + \theta \Delta y) \Delta y$,

where $0 < \theta < 1$. Note that $F_y \neq 0$ on D^* , hence:

$$\frac{\Delta y}{\Delta x} = -\frac{F_x(\bar{x} + \theta \Delta x, \bar{y} + \theta \Delta y)}{F_y(\bar{x} + \theta \Delta x, \bar{y} + \theta \Delta y)}.$$

Let $\Delta x \to 0$. Considering the continuity of F_x and F_y , we obtain:

$$\frac{dy}{dx}\Big|_{x=\bar{x}} = -\frac{F_x(\bar{x},\bar{y})}{F_y(\bar{x},\bar{y})}.$$

Thus:

$$f'(\bar{x}) = -\frac{F_x(\bar{x}, \bar{y})}{F_y(\bar{x}, \bar{y})}.$$

The proof is complete.

Chapter 11 Multiple Integrals

Bibliography

- [1] 徐森林, 薛春华. 数学分析. 第一版. 清华大学出版社, 2005.
- [2] 陈纪修, 於崇华. 数学分析. 第三版. 高等教育出版社, 2019.
- [3] 常庚哲, 史济怀. 数学分析教程. 第三版. 中国科学技术大学出版社, 2012.
- [4] 裴礼文. 数学分析中的典型问题与方法. 第三版. 高等教育出版社, 2021.
- [5] 汪林. 数学分析中的问题与反例. 第一版. 高等教育出版社, 2015.
- [6] 谢惠民, 恽自求, 易法槐, 钱定边. 数学分析习题课讲义. 第二版. 高等教育出版社, 2019.
- [7] Walter Rudin. Principles of Mathematical Analysis. Third Edition. McGraw-Hill, 1976.
- [8] 菲赫金哥尔茨. 微积分学教程. 第八版. 高等教育出版社, 2006.
- [9] Wikipedia.https://en.wikipedia.org/wiki/.