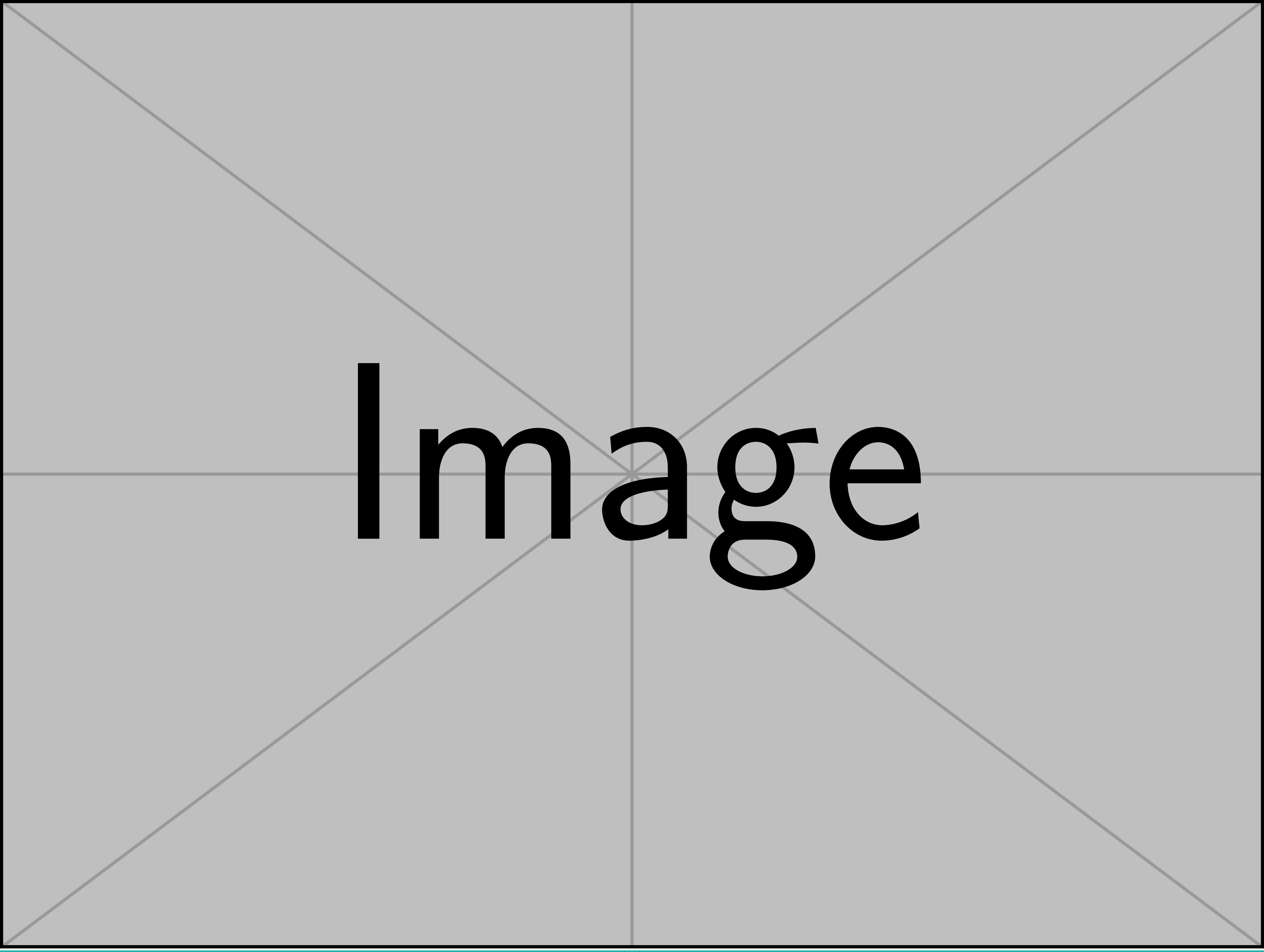

Analyse Mathématique

Author: CatMono

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Contents

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:

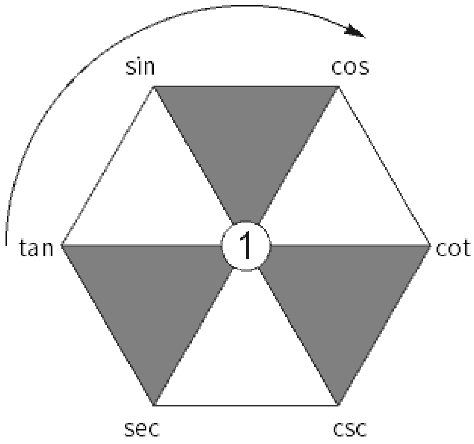
$$\sin\frac{\alpha}{2} = \pm\sqrt{\frac{1-\cos\alpha}{2}}$$
$$\cos\frac{\alpha}{2} = \pm\sqrt{\frac{1+\cos\alpha}{2}}$$
$$\tan\frac{\alpha}{2} = \frac{1-\cos\alpha}{\sin\alpha} = \frac{\sin\alpha}{1+\cos\alpha}$$

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's Proposition and Fitting Method

Proposition 2.1 (Cauchy's Proposition)

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

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

Note

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

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

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

Proof

■

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's 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's Theorem does not hold.*

2.3 Subsequences

¶ Subsequences

¶ Upper Limits and Lower Limits

2.4 Proposition on The Continuity of Real Numbers

¶ Dedkind Cut

¶ Least Supremum Property

¶ Monotone Convergence Theorem

¶ Bolzano-Weierstrass Theorem

¶ Nested Interval Theorem

¶ Cauchy Criterion for Sequences

¶ Heine-Borel Theorem

Chapter 3 Limits and Continuity of Functions

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

4.5 Taylor's Theorem

4.6 Applications of Taylor's Theorem

Chapter 5   Integral

Chapter 6 Series of Numbers

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, S is 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, $f : U \to \mathbb{R}^1$, $\mathbf{e}$ 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 $\mathbf{e}$, 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

$$\frac{\partial f}{\partial x_i}(\mathbf{x}^0) \quad \text{or} \quad \mathrm{D}_i f(\mathbf{x}^0) \quad \text{or} \quad f_{x_i}(\mathbf{x}^0) \quad (i = 1, 2, \cdots, n).$$

$\mathrm{D}_i = \frac{\partial}{\partial x_i}$ 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**.

Note *Let $\mathbf{e}$ be a direction, then $\|-\mathbf{e}\| = \|\mathbf{e}\| = 1$, which implies that $-\mathbf{e}$ 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

$$Jf(\mathbf{x}) = (\mathrm{D}_1 f(\mathbf{x}), \mathrm{D}_2 f(\mathbf{x}), \ldots, \mathrm{D}_n f(\mathbf{x})),$$

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

$$\mathrm{d}f(\mathbf{x}^0)(\boldsymbol{\Delta}\mathbf{x}) = Jf(\mathbf{x}^0)\boldsymbol{\Delta}\mathbf{x}.$$

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.

Total Differential

Definition 10.3 (Total Differential)

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

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

where $A_1, A_2, \ldots, A_n$ are constants independent of $\mathbf{\Delta x}$, 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

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

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

$$\mathbf{A} = (A_1, A_2, \ldots, A_n) = Jf(\mathbf{x}^0) = \left(\mathrm{D}_1 f(\mathbf{x}^0), \mathrm{D}_2 f(\mathbf{x}^0), \ldots, \mathrm{D}_n f(\mathbf{x}^0)\right).$$

[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 $Jf = (\mathrm{D}_1 f, \mathrm{D}_2 f, \ldots, \mathrm{D}_n f)$ is continuous at $\mathbf{x}^0$ (i.e., $\frac{\partial f}{\partial x_i}$ is continuous at $\mathbf{x}^0$ for $i = 1, 2, \ldots, n$), then f is differentiable at $\mathbf{x}^0$. However, the converse is not necessarily true.

[a] It is referred to as the total differential formula, and the more common form is

$$\mathrm{d}f(x_0, y_0) = \frac{\partial f}{\partial x}(x_0, y_0)\,\mathrm{d}x + \frac{\partial f}{\partial y}(x_0, y_0)\,\mathrm{d}y.$$

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):

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

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

$$\begin{aligned}I_1 &= f_x(x_0+\theta_1\Delta x, y_0+\Delta y, z_0+\Delta z)\Delta x, \\ I_2 &= f_y(x_0, y_0+\theta_2\Delta y, z_0+\Delta z)\Delta y.\end{aligned}$$

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

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

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

$$\begin{aligned}I_1 &= f_x(x_0,y_0,z_0)\Delta x + \alpha_1\Delta x, \quad \alpha_1 \to 0(\rho\to 0), \\ I_2 &= f_y(x_0,y_0,z_0)\Delta y + \alpha_2\Delta y, \quad \alpha_2 \to 0(\rho\to 0).\end{aligned}$$

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,

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

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:* $\mathrm{D}_{\vec{u}} = \nabla f \cdot \vec{u}$
2. *Directional Derivative*
   - *All directional derivatives exist* $\not\Longrightarrow$ *differentiable or continuous.*
   - *All directional derivatives exist and are equal* $\not\Longrightarrow$ *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):

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

Similarly, higher-order partial derivatives of order $3, 4, \cdots m, \cdots$ 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 $\mathbf{x}^0 = (x_1^0, x_2^0, \ldots, x_n^0) \in U$, 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.[a]

[a] 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.

Proof When $k \neq 0, h \neq 0$, define

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

and

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

Applying the Lagrange Mean Value Theorem, we have

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

On the other hand,

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

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

$$\mathrm{d}^2 z = \mathrm{d}(\mathrm{d}z).$$

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

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

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

$$\mathrm{d}^2 x = \mathrm{d}(\mathrm{d}x) = 0, \qquad \mathrm{d}^2 y = \mathrm{d}(\mathrm{d}y) = 0,$$

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

$$\begin{aligned}
\mathrm{d}^2 z = \mathrm{d}(\mathrm{d}z) &= \mathrm{d}\left(\frac{\partial z}{\partial x}\mathrm{d}x + \frac{\partial z}{\partial y}\mathrm{d}y\right) \\
&= \mathrm{d}\left(\frac{\partial z}{\partial x}\right)\mathrm{d}x + \frac{\partial z}{\partial x}\mathrm{d}^2 x + \mathrm{d}\left(\frac{\partial z}{\partial y}\right)\mathrm{d}y + \frac{\partial z}{\partial y}\mathrm{d}^2 y \\
&= \left(\frac{\partial^2 z}{\partial x^2}\mathrm{d}x + \frac{\partial^2 z}{\partial x \partial y}\mathrm{d}y\right)\mathrm{d}x + \left(\frac{\partial^2 z}{\partial y \partial x}\mathrm{d}x + \frac{\partial^2 z}{\partial y^2}\mathrm{d}y\right)\mathrm{d}y \\
&= \frac{\partial^2 z}{\partial x^2}(\mathrm{d}x)^2 + 2\frac{\partial^2 z}{\partial x \partial y}\mathrm{d}x\mathrm{d}y + \frac{\partial^2 z}{\partial y^2}(\mathrm{d}y)^2,
\end{aligned}$$

where $(\mathrm{d}x)^2$ and $(\mathrm{d}y)^2$ denote $\mathrm{d}^2 x$ and $\mathrm{d}^2 y$ 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

$$\begin{aligned}
\mathrm{d}z &= \left(\mathrm{d}x\frac{\partial}{\partial x} + \mathrm{d}y\frac{\partial}{\partial y}\right)z, \\
\mathrm{d}^2 z &= \left(\mathrm{d}x\frac{\partial}{\partial x} + \mathrm{d}y\frac{\partial}{\partial y}\right)^2 z.
\end{aligned}$$

Similarly, we define

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

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:

$$\mathrm{d}^k u = \left(\mathrm{d}x_1\frac{\partial}{\partial x_1} + \mathrm{d}x_2\frac{\partial}{\partial x_2} + \cdots + \mathrm{d}x_n\frac{\partial}{\partial x_n}\right)^k u, \quad k = 1, 2, \ldots .$$

10.2 Differential of Vector-Valued Functions

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

$$f : U \to \mathbb{R}^m,$$
$$\mathbf{x} \mapsto \mathbf{y} = f(\mathbf{x})$$

Expressed in coordinate vector form:

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

1. If each component function $f_i(x_1, x_2, \ldots, x_n)$ $(i = 1, 2, \ldots, m)$ 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 $\mathrm{D}f(\mathbf{x})$, $J_f(\mathbf{x})$) denotes the derivative of $\mathbf{f}$ at $\mathbf{x}$ in U.
2. If every component function $f_i(x_1, x_2, \ldots, x_n)$ $(i = 1, 2, \ldots, m)$ 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 $\Delta\mathbf{x} = (\Delta x_1, \Delta x_2, \ldots, \Delta x_n)^T$ 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 $\mathrm{d}\mathbf{y}$. If we denote $\Delta\mathbf{x}$ by $\mathrm{d}\mathbf{x}$ ($\mathrm{d}\mathbf{x} = (\mathrm{d}x_1, \mathrm{d}x_2, \ldots, \mathrm{d}x_n)^T$), then
$$\mathrm{d}\mathbf{y} = A\,\mathrm{d}\mathbf{x}.$$
If the vector-valued function $\mathbf{f}$ is differentiable at every point in U, then $\mathbf{f}$ 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 $f_i(x_1, x_2, \ldots, x_n)$ $(i = 1, 2, \ldots, m)$ 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

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

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:

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

Note

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

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

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

If $\mathbf{g}$ 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,

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

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 $\overline{\mathbf{x}_0\mathbf{x}_1}$) 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.

Theorem 10.3 (Lagrange's Mean Value Theorem)

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

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

Proof

■

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 $\mathbf{x}^0 = (x_1^0, x_2^0, \ldots, x_n^0) \in D$ and $\mathbf{x} = (x_1, x_2, \ldots, x_n) \in D$, 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:

$$f(\mathbf{x}) = f(\mathbf{x}^0) + \sum_{k=1}^{m} \frac{1}{k!} \sum_{i_1, i_2, \ldots, i_k = 1}^{n} \frac{\partial^k f}{\partial x_{i_1} \partial x_{i_2} \ldots \partial x_{i_k}} (\mathbf{x}^0) \prod_{j=1}^{k} (x_{i_j} - x_{i_j}^0) + R_m(\mathbf{x} - \mathbf{x}^0),$$

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; $\Delta\mathbf{x} = (\Delta x_1, \Delta x_2, \ldots, \Delta x_n)$):

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

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.

Chapter 11 Multiple Integrals

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