

# Chap 9 — 2

多变量函数的微分

## 9.2.1 多变量函数的偏微商

**定义** 设  $f(x,y)$  在  $(x_0, y_0)$  某邻域有定义. 仅给  $x$  以增量  $\Delta x$  相应有函数的增量(对  $x$  **偏增量**)

$$\Delta_x z = f(x_0 + \Delta x, y_0) - f(x_0, y_0)$$

函数  $f$  在点  $(x_0, y_0)$  处对  $x$  的 **偏微商**(或**偏导数**)

$$f'_x(x_0, y_0) \stackrel{\text{def}}{=} \lim_{\Delta x \rightarrow 0} \frac{\Delta_x z}{\Delta x} = \lim_{\Delta x \rightarrow 0} \frac{f(x_0 + \Delta x, y_0) - f(x_0, y_0)}{\Delta x}$$

◆ 偏微商也可记为  $\frac{\partial f}{\partial x}(x_0, y_0)$

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◆ 对变量 $y$ 的偏微商类似

◆ 可偏导: 两个偏导数都存在.

◆ 偏导(函)数:  $f'_x(x, y), f'_y(x, y)$  or  $\frac{\partial f}{\partial x}(x, y), \frac{\partial f}{\partial y}(x, y)$

例 求函数 $u = x^y$  ( $x > 0$ )的偏导数.

## ■ 连续与可偏导

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### ➤ 可偏导未必连续

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

在(0,0)的情况.

### ➤ 连续未必可偏导

例 考察  $f(x,y) = |x| + |y|$  在(0,0)的情况.

## ■ 偏导数的几何意义

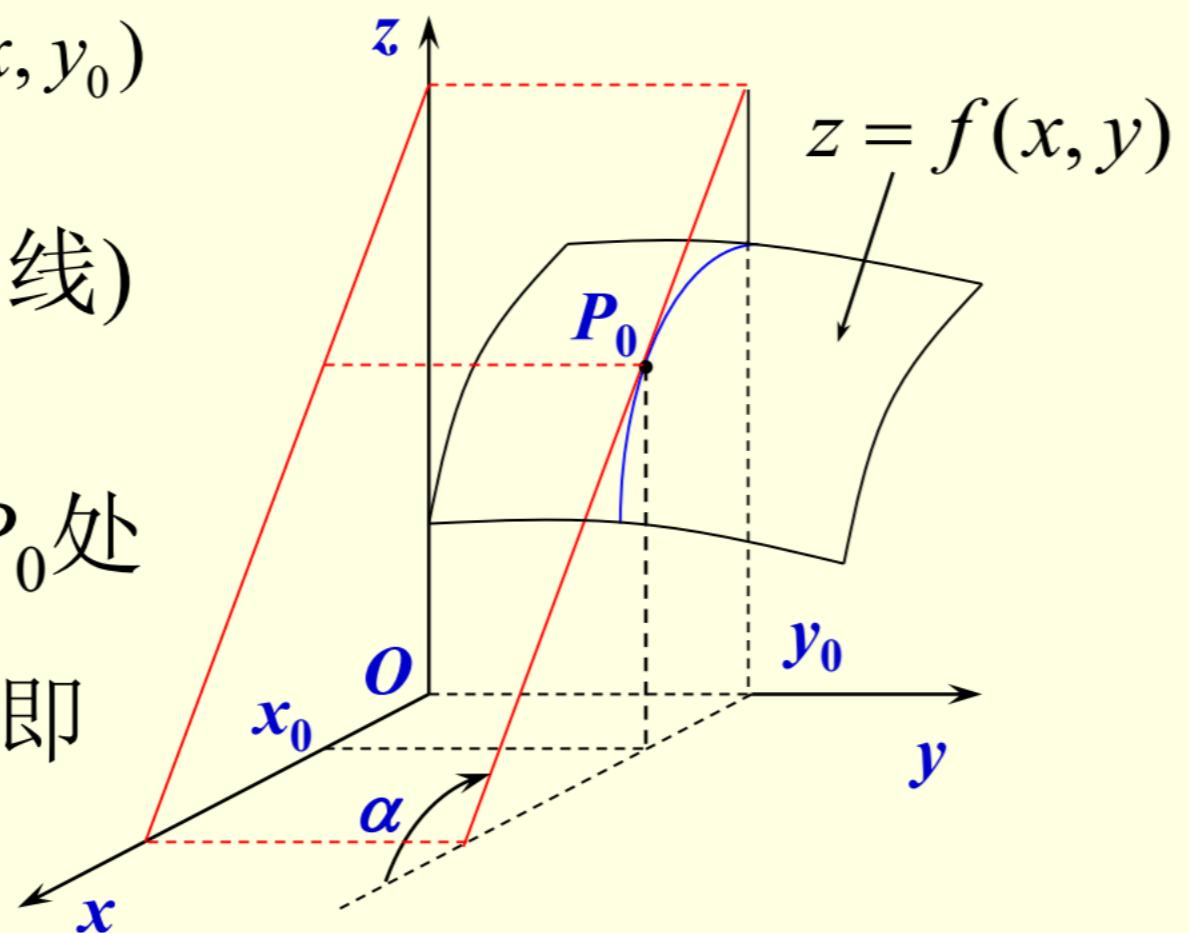
曲面  $z = f(x, y)$  与平面  $y = y_0$  的交线

$$\begin{cases} z = f(x, y) \\ y = y_0 \end{cases} \Rightarrow z = f(x, y_0)$$

(平面  $y = y_0$  上的曲线)

$f'_x(x_0, y_0)$  是该曲线在  $P_0$  处的切线关于  $x$  轴的 **斜率**. 即

$$f'_x(x_0, y_0) = \tan \alpha$$



定义  $f(x,y)$  在某邻域内的偏导数  $f'_x(x,y), f'_y(x,y)$  的偏导

数称为  $f$  的二阶偏导数. 记为

$$f''_{xx} = \frac{\partial^2 f}{\partial x^2} = \frac{\partial}{\partial x} \left( \frac{\partial f}{\partial x} \right), f''_{xy} = \frac{\partial^2 f}{\partial x \partial y} = \underbrace{\frac{\partial}{\partial y}}_{\text{Warning!}} \left( \frac{\partial f}{\partial x} \right) \quad \frac{\partial^2 f}{\partial y \partial x} \quad \text{先对 } x \text{ 再对 } y$$
$$f''_{yx} = \frac{\partial^2 f}{\partial y \partial x} = \frac{\partial}{\partial x} \left( \frac{\partial f}{\partial y} \right), f''_{yy} = \frac{\partial^2 f}{\partial y^2} = \frac{\partial}{\partial y} \left( \frac{\partial f}{\partial y} \right)$$

类似可定义三阶偏导数, 例如

$$f'''_{xxy} = \frac{\partial^3 f}{\partial x^2 \partial y} = \frac{\partial}{\partial y} \left( \frac{\partial^2 f}{\partial x^2} \right)$$

例 求函数  $z = \ln x + e^y \sin x$  的所有二阶偏导数.

## 问题: 混合偏导数是否总与求偏导次序无关?

不一定

例 设  $f(x, y) = \begin{cases} xy, & |x| \geq |y| \\ -xy, & |x| < |y| \end{cases}$ , 求  $f''_{xy}(0,0), f''_{yx}(0,0)$ .

分析  $f''_{xy}(0,0) = \lim_{y \rightarrow 0} \frac{f'_x(0,y) - f'_x(0,0)}{y} = -1$

( $y \neq 0$ )  $f'_x(0,y) = \lim_{x \rightarrow 0} \frac{f(x,y) - f(0,y)}{x} = \lim_{x \rightarrow 0} \frac{-xy - 0}{x} = -y$

$$f'_x(0,0) = \lim_{x \rightarrow 0} \frac{f(x,0) - f(0,0)}{x} = 0$$

混合偏导次序无关定理

定理 若  $f(x, y)$  的二阶混合偏导数在  $(x, y)$  连续, 则

$$f''_{xy}(x,y) = f''_{yx}(x,y)$$

$$f''_{yx}(0,0) = 1$$

证明  
方法

## 9.2.2 多变量函数的可微性

线性主部

一元情形：若  $\Delta f = a\Delta x + o(\Delta x)$ , 则称  $f$  在  $x_0$  可微，  
并把  $a\Delta x$  称为  $f$  在  $x_0$  处的微分，记为  $df|_{x=x_0} = a\Delta x$

二元情形：对函数  $z = f(x, y)$ , 若 增量 (增量)

$$\begin{aligned}\Delta z &= f(x_0 + \Delta x, y_0 + \Delta y) - f(x_0, y_0) \\ &= a\Delta x + b\Delta y + o(\rho)\end{aligned}$$

其中  $a, b$  是常数,  $\rho = \sqrt{(\Delta x)^2 + (\Delta y)^2}$ , 则称  $f$  在  $(x_0, y_0)$  可微.

并把  $a\Delta x + b\Delta y$  称为  $f$  在  $(x_0, y_0)$  处的 微分. 记为

又称全微分

$$df|_{(x_0, y_0)} = a\Delta x + b\Delta y$$

若 $f$ 在区域 $D$ 内处处可微, 则称 $f$ 是 $D$ 内的**可微函数**.

➤ 可微必连续

➤ 可微必可偏导, 且若

$$\begin{aligned} df|_{(x_0, y_0)} &= a\Delta x + b\Delta y \\ \Rightarrow f'_x(x_0, y_0) &= a, f'_y(x_0, y_0) = b \end{aligned}$$

证: 令 $\Delta y = 0$

$$\begin{aligned} \Delta_x f &= a\Delta x + o(\Delta x) \\ f'_x(x_0, y_0) &= \lim_{\Delta x \rightarrow 0} \frac{\Delta_x f}{\Delta x} \end{aligned}$$

$$\begin{aligned} &= \lim_{\Delta x \rightarrow 0} \left( a + \frac{o(\Delta x)}{\Delta x} \right) \\ &= a \end{aligned}$$

## ■ 微分公式

$$df(x, y) = f'_x(x, y)dx + f'_y(x, y)dy$$

例 求函数 $z = x^y$ 在点 $(1, 1)$ 处的微分.

$$\frac{\partial z}{\partial x}\Big|_{(1,1)} = yx^{y-1}\Big|_{(1,1)} = 1$$

$$\frac{\partial z}{\partial y}\Big|_{(1,1)} = xy^{\ln x}\Big|_{(1,1)} = 0$$

$$dz|_{U_0(D)} = 1 \cdot dx + 0 \cdot dy = dx$$

例 求函数  $z = \arctan \frac{y}{x}$  的微分.

定理 设  $f(x, y)$  在区域  $D$  内存在偏导数, 则

- (1) 若  $f'_x(x, y), f'_y(x, y)$  在  $D$  内有界, 则  $f$  在  $D$  内连续;
- (2) 若  $f'_x(x, y), f'_y(x, y)$  在  $D$  内连续, 则  $f$  在  $D$  内可微.

### 结论

偏导数连续  $\Rightarrow$  可微  $\Rightarrow$    $\begin{cases} \text{连续} \\ \text{可偏导} \end{cases}$

$$\text{设 } \frac{\partial z}{\partial x} = \frac{-y}{1+y^2} = \frac{-y}{x^2+y^2}, \quad \frac{\partial z}{\partial y} = \frac{x}{1+\frac{y^2}{x^2}} = \frac{x}{x^2+y^2}$$

$$dz = \frac{\partial z}{\partial x} dx + \frac{\partial z}{\partial y} dy = \dots$$

$$\text{定理(1)} \Delta f = f(x+\Delta x, y+\Delta y) - f(x, y)$$

$$= [f(x+\Delta x, y+\Delta y) - f(x, y+\Delta y)] + [f(x, y+\Delta y) - f(x, y)]$$

$$\begin{aligned} & \forall \theta_1 \in [0, 1] \\ & = f'_x(x+\theta_1 \Delta x, y+\Delta y) \Delta x + f'_y(x+\Delta x, y+\theta_2 \Delta y) \Delta y \\ & \forall \theta_2 \in [0, 1] \end{aligned}$$

$x, y \in D$

$$\Delta f \leq M_1 \Delta x + M_2 \Delta y \quad (\text{有界})$$

$$\begin{array}{l} \Delta x \rightarrow 0 \\ \Delta y \rightarrow 0 \end{array} \quad \Delta f \rightarrow 0.$$

可微函数  $a, b$  不能依赖  $\Delta x, \Delta y$

$$(2) \text{ 由(1) } \Delta f = [f'_x(x, y) + \alpha] \Delta x + [f'_y(x, y) + \beta] \Delta y$$

其中  $\alpha, \beta$  为无穷小。

$$\Delta f = f'_x(x, y) \Delta x + f'_y(x, y) \Delta y + \alpha \Delta x + \beta \Delta y$$

$$\frac{|\alpha \Delta x + \beta \Delta y|}{\rho} \leq |\alpha| + |\beta| \rightarrow 0. \quad (\Delta x, \Delta y \rightarrow 0, 0)$$

### 9.2.3 方向导数与梯度

**定义** 设  $\mathbf{e} = (\cos \alpha, \cos \beta)$ , 函数  $z = f(x, y)$  在  $(x_0, y_0)$  处沿  $\mathbf{e}$  的**方向导数**定义为

$$\frac{\partial f}{\partial \mathbf{e}}(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x_0 + t \cos \alpha, y_0 + t \cos \beta) - f(x_0, y_0)}{t}$$

- 偏导数是方向导数的特例,

$$\frac{\partial f}{\partial x}(x_0, y_0) = \frac{\partial f}{\partial \mathbf{e}}(x_0, y_0), \quad \mathbf{e} = (1, 0)$$

方向角:  $\theta \leq x_{\text{轴}} y_{\text{轴}}$   
大正向夹角

定理  $z = f(x, y)$  在  $M_0(x_0, y_0)$  可微,  $e = (\cos \alpha, \cos \beta)$

则  $f$  在  $M_0$  点存在方向导数, 且

$$\frac{\partial f}{\partial e}(x_0, y_0) = f'_x(x_0, y_0) \cos \alpha + f'_y(x_0, y_0) \cos \beta$$

$$\begin{aligned}\Delta f &= f(x_0 + t \cos \alpha, y_0 + t \cos \beta) - f(x_0, y_0) \\ &= f'_x(x_0, y_0) t \cos \alpha + f'_y(x_0, y_0) t \cos \beta \\ &\quad + o(|t|)\end{aligned}$$

## 结论

● 可微  $\Rightarrow$  方向导数存在  $\Rightarrow$  可偏导

极限存在

连 续

例 求  $r(x, y) = \sqrt{x^2 + y^2}$  沿  $e = (\cos \alpha, \cos \beta)$  的方向导数

$(x, y) \neq (0, 0)$

(0,0) 点附近

$$\frac{\partial}{\partial e} \sqrt{x^2 + y^2} = \frac{\partial \sqrt{x^2 + y^2}}{\partial x} \cos \alpha + \frac{\partial \sqrt{x^2 + y^2}}{\partial y} \cos \beta$$

$$= \frac{x}{\sqrt{x^2 + y^2}} \cos \alpha + \frac{y}{\sqrt{x^2 + y^2}} \cos \beta$$

**定义** 函数  $f(x, y)$  在点  $M_0(x_0, y_0)$  的**梯度**定义为

$$\text{grad}f(x_0, y_0) \stackrel{\text{def}}{=} (f'_x(x_0, y_0), f'_y(x_0, y_0))$$

利用梯度符号, 得到  $\nabla f$  ( $\nabla$  (nabla) 也指梯度)

$$\frac{\partial f}{\partial \mathbf{e}}(M_0) = \text{grad } f(M_0) \cdot \mathbf{e} = |\text{grad } f(M_0)| \cos \theta$$

$\Rightarrow \theta = 0$  时, 方向导数  $\frac{\partial f}{\partial \mathbf{e}}(M_0)$  取最大值  $|\text{grad } f(M_0)|$

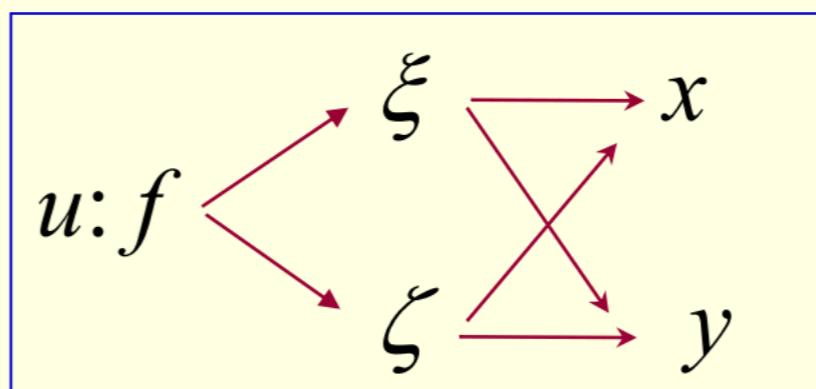
**结论** 梯度的方向是方向导数取最大值时的方向,  
其模就是方向导数的最大值.

## 9.2.4 复合函数的微分

**定理** 设  $u = f(\xi, \zeta)$  可微,  $\xi = \xi(x, y)$ ,  $\zeta = \zeta(x, y)$  可微.  
则复合函数  $u = f(\xi(x, y), \zeta(x, y))$  也可微, 且

$$\frac{\partial u}{\partial x} = \frac{\partial f}{\partial \xi} \cdot \frac{\partial \xi}{\partial x} + \frac{\partial f}{\partial \zeta} \cdot \frac{\partial \zeta}{\partial x}, \quad \frac{\partial u}{\partial y} = \frac{\partial f}{\partial \xi} \cdot \frac{\partial \xi}{\partial y} + \frac{\partial f}{\partial \zeta} \cdot \frac{\partial \zeta}{\partial y}$$

➤ 链法则



$$u = f(\xi_1, \xi_2, \dots, \xi_m)$$

$$\xi_i = \xi_i(x_1, x_2, \dots, x_n)$$

$$\frac{\partial u}{\partial x_j} = \sum_{i=1}^m \frac{\partial f}{\partial \xi_i} \frac{\partial \xi_i}{\partial x_j}$$

想一想  $m$ 个中间变量  $n$ 个自变量的链法则?

差增量  $\Delta x \Delta y$

$$\Delta \xi = \frac{\partial \xi}{\partial x} \Delta x + \frac{\partial \xi}{\partial y} \Delta y + o(\rho)$$

where  $\rho = \sqrt{\Delta x^2 + \Delta y^2}$

$$\stackrel{\text{def}}{=} a_1 \Delta x + a_2 \Delta y + o(\rho)$$

$$\Delta \zeta = \frac{\partial \zeta}{\partial x} \Delta x + \frac{\partial \zeta}{\partial y} \Delta y + o(\rho) \stackrel{\text{def}}{=} b_1 \Delta x + b_2 \Delta y + o(\rho)$$

$$\Delta u = \frac{\partial f}{\partial \xi} \Delta \xi + \frac{\partial f}{\partial \zeta} \Delta \zeta + o(\tilde{\rho}) \quad \text{where } \tilde{\rho} = \sqrt{\Delta \xi^2 + \Delta \zeta^2}$$

$$= \frac{\partial f}{\partial \xi} \left( \frac{\partial \xi}{\partial x} \Delta x + \frac{\partial \xi}{\partial y} \Delta y + o(\rho) \right) + \frac{\partial f}{\partial \zeta} \left( \frac{\partial \zeta}{\partial x} \Delta x + \frac{\partial \zeta}{\partial y} \Delta y + o(\rho) \right) + o(\rho)$$

$$= \left( \frac{\partial f}{\partial \xi} \frac{\partial \xi}{\partial x} + \frac{\partial f}{\partial \zeta} \frac{\partial \zeta}{\partial x} \right) \Delta x + \left( \frac{\partial f}{\partial \xi} \frac{\partial \xi}{\partial y} + \frac{\partial f}{\partial \zeta} \frac{\partial \zeta}{\partial y} \right) \Delta y + o(\rho) + o(\tilde{\rho})$$

$$\frac{o(\tilde{\rho})}{\rho} = \frac{o(\tilde{\rho})}{\tilde{\rho}} \frac{\tilde{\rho}}{\rho} \quad \text{只需说明 } \frac{\tilde{\rho}}{\rho} \text{ 有界}$$

$$\left( \frac{\tilde{\rho}}{\rho} \right)^2 = \frac{\Delta \xi^2 + \Delta \zeta^2}{\Delta x^2 + \Delta y^2} = \frac{1}{\rho^2} \left[ (a_1 \Delta x + a_2 \Delta y + o(\rho))^2 + (b_1 \Delta x + b_2 \Delta y + o(\rho))^2 \right]$$

$$\geq \left( \frac{a_1 \Delta x}{\rho} + \frac{a_2 \Delta y}{\rho} + \frac{o(\rho)}{\rho} \right)^2 + \dots$$

$$\leq (a_1 + a_2 + M_1)^2 + (b_1 + b_2 + M_2)^2$$

有界

$$\therefore \Delta u = (\dots) \Delta x + (\dots) \Delta y + o(\rho)$$

## ■ 一阶微分形式的不变性

函数  $u = f(\xi, \zeta)$  的微分

$$du = \frac{\partial f}{\partial \xi} d\xi + \frac{\partial f}{\partial \zeta} d\zeta$$

若  $\xi, \zeta$  又是  $x, y$  的可微函数  $\xi = \xi(x, y), \zeta = \zeta(x, y)$ , 则

复合函数  $u = f(\xi(x, y), \zeta(x, y))$  的微分

$$du = \frac{\partial u}{\partial x} dx + \frac{\partial u}{\partial y} dy$$

$$= \left( \frac{\partial f}{\partial \xi} \frac{\partial \xi}{\partial x} + \frac{\partial f}{\partial \zeta} \frac{\partial \zeta}{\partial x} \right) dx + \left( \frac{\partial f}{\partial \xi} \frac{\partial \xi}{\partial y} + \frac{\partial f}{\partial \zeta} \frac{\partial \zeta}{\partial y} \right) dy$$

$$= \frac{\partial f}{\partial \xi} \left( \frac{\partial \xi}{\partial x} dx + \frac{\partial \xi}{\partial y} dy \right) + \frac{\partial f}{\partial \zeta} \left( \frac{\partial \zeta}{\partial x} dx + \frac{\partial \zeta}{\partial y} dy \right)$$

注意到  $d\xi = \frac{\partial \xi}{\partial x} dx + \frac{\partial \xi}{\partial y} dy$ ,  $d\zeta = \frac{\partial \zeta}{\partial x} dx + \frac{\partial \zeta}{\partial y} dy$

从而

$$du = \frac{\partial f}{\partial \xi} d\xi + \frac{\partial f}{\partial \zeta} d\zeta$$

因此, 对于函数  $u = f(\xi, \zeta)$ , 无论  $\xi, \zeta$  是自变量  
还是函数, 都有

$$du = \frac{\partial f}{\partial \xi} d\xi + \frac{\partial f}{\partial \zeta} d\zeta$$

想一想  $n$  元函数的一阶微分形式不变性?

$$\begin{pmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \end{pmatrix} = \begin{pmatrix} \frac{\partial f}{\partial \xi} & \frac{\partial f}{\partial \eta} \end{pmatrix} \begin{pmatrix} \frac{\partial \xi}{\partial x} & \frac{\partial \eta}{\partial y} \\ \frac{\partial \xi}{\partial y} & \frac{\partial \eta}{\partial y} \end{pmatrix}$$

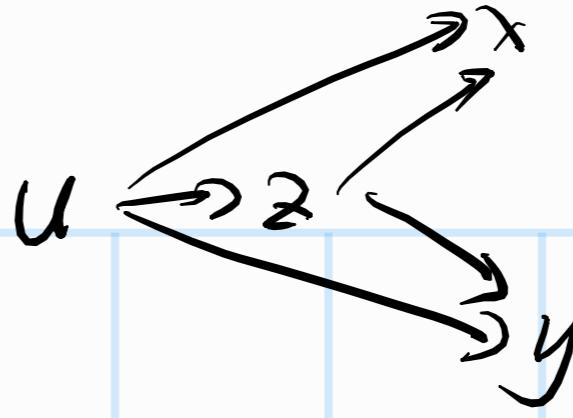
**例** 设函数  $u = f(x, y, z)$  可微, 而  $z = z(x, y)$  可偏导.

求复合函数  $u = f(x, y, z(x, y))$  对  $x$  的偏导数.

**例** 原点处电荷  $q$  产生电势  $u = q/r$ , 其中  $r$  是点  $\mathbf{r} = (x, y, z)$  到原点的距离. 当  $r \neq 0$  时, 求  $u$  在  $(x, y, z)$  处的梯度及沿方向  $\mathbf{r}$  的变化率, 并证明  $u$  满足方程

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} + \frac{\partial^2 u}{\partial z^2} = 0 \quad \text{———— Laplace 方程}$$

$$\textcircled{1} \quad \frac{\partial u}{\partial x} = \frac{\partial f}{\partial x} + \frac{\partial f}{\partial z} \frac{\partial z}{\partial x}$$



$$\textcircled{2} \quad u = \frac{q}{\sqrt{x^2+y^2+z^2}} \quad \frac{\partial u}{\partial x} = -q(x^2+y^2+z^2)^{-\frac{3}{2}}x$$

$$\frac{\partial u}{\partial y} = -q(x^2+y^2+z^2)^{-\frac{3}{2}}y$$

$$\frac{\partial u}{\partial z} = -q(x^2+y^2+z^2)^{-\frac{3}{2}}z$$

梯度  $-q(x^2+y^2+z^2)^{-\frac{3}{2}}(x, y, z)$

$$\frac{\partial u}{\partial r} = \text{负梯度的模}$$

$$\frac{\partial u}{\partial x} = (\frac{q}{r})' \frac{\partial r}{\partial x}$$

$$= -\frac{q}{r^2} \frac{x}{r}$$

$$= -\frac{q x}{r^3}$$

$$\nabla u = -\frac{q}{r^3}(x, y, z)$$

$$= -\frac{q}{r^3} \vec{r}$$

$$\frac{\partial u}{\partial r} = \nabla u \cdot \frac{\vec{r}}{r}$$

$$= -\frac{q}{r^3} \vec{r} \cdot \frac{\vec{r}}{r}$$

$$= -\frac{q}{r^2}$$

$$\frac{\partial^2 u}{\partial x^2} = -\frac{q}{r^3} + \frac{3q}{r^4} \frac{x}{r} x$$

$$= -\frac{q}{r^3} + \frac{3q x^2}{r^5}$$

$$\frac{\partial u^2}{\partial x^2} + \frac{\partial u^2}{\partial y^2} + \frac{\partial u^2}{\partial z^2} = -\frac{3q}{r^3} + \frac{3q^2}{r^5} (x^2+y^2+z^2)$$

$$= -\frac{3q}{r^3} + \frac{3q}{r^3} = 0$$

## 9.2.5 向量值函数的微商和微分

设有单变量向量值函数(参数曲线)

$$t \mapsto \mathbf{r}(t), \quad t \in [\alpha, \beta]$$

它在 $t_0$ 处的**微商**定义为

$$\mathbf{r}'(t_0) = \frac{d\mathbf{r}}{dt}(t_0) = \lim_{\Delta t \rightarrow 0} \frac{\mathbf{r}(t_0 + \Delta t) - \mathbf{r}(t_0)}{\Delta t}$$

- 几何意义  $\mathbf{r}'(t_0)$ 是曲线在参数为的 $t_0$ 点处**切向量**,  
且指向参数增加方向.

➤ 坐标形式 若

$$\mathbf{r}(t) = x(t)\mathbf{i} + y(t)\mathbf{j} + z(t)\mathbf{k}$$

则有

$$\mathbf{r}'(t) = \underbrace{x'(t)\mathbf{i} + y'(t)\mathbf{j} + z'(t)\mathbf{k}}$$

$$\frac{d^2 \mathbf{r}}{dt^2}(t) = \frac{d^2 x}{dt^2}(t)\mathbf{i} + \frac{d^2 y}{dt^2}(t)\mathbf{j} + \frac{d^2 z}{dt^2}(t)\mathbf{k}$$

其**微分**定义为  $d\mathbf{r}(t) = \frac{d\mathbf{r}}{dt} dt$  , 即

$$d\mathbf{r} = dx\mathbf{i} + dy\mathbf{j} + dz\mathbf{k}$$

## ■ 运算性质 设 $\mathbf{a}(t), \mathbf{b}(t)$ 是向量函数, $f(t)$ 是数量函数, 则

$$1^\circ \quad \frac{d}{dt}(f\mathbf{a}) = f \frac{d\mathbf{a}}{dt} + \frac{df}{dt} \mathbf{a} \quad \text{乘积求导}$$

$$2^\circ \quad \frac{d}{dt}(\mathbf{a} \cdot \mathbf{b}) = \frac{d\mathbf{a}}{dt} \cdot \mathbf{b} + \mathbf{a} \cdot \frac{d\mathbf{b}}{dt}$$

$$3^\circ \quad \frac{d}{dt}(\mathbf{a} \times \mathbf{b}) = \frac{d\mathbf{a}}{dt} \times \mathbf{b} + \mathbf{a} \times \frac{d\mathbf{b}}{dt}$$

例 设  $\mathbf{a}(t)$  是向量函数, 且  $|\mathbf{a}(t)| = c$  (常数). 证明

$$\mathbf{a} \cdot \frac{d\mathbf{a}}{dt} = 0$$

$$\frac{d(\vec{a} \times \vec{b})}{dt} = \frac{d}{dt} \begin{vmatrix} \vec{i} & \vec{j} & \vec{k} \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{vmatrix}$$

$$= \begin{vmatrix} \vec{i} & \vec{j} & \vec{k} \\ a'_1 & a'_2 & a'_3 \\ b'_1 & b'_2 & b'_3 \end{vmatrix} + \begin{vmatrix} \vec{i} & \vec{j} & \vec{k} \\ a_1 & a_2 & a_3 \\ b'_1 & b'_2 & b'_3 \end{vmatrix}$$

$$= \frac{d\vec{a}}{dt} \times \vec{b} + \vec{a} \times \frac{d\vec{b}}{dt}$$

$$\text{例 } \vec{a} \cdot \vec{a} = C^2$$

$$\text{左端 } \frac{d(\vec{a} \cdot \vec{a})}{dt} = 2 \frac{d\vec{a}}{dt} \cdot \vec{a}$$

设有两变量向量值函数(参数曲面)

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$$\mathbf{r}(u, v) = x(u, v)\mathbf{i} + y(u, v)\mathbf{j} + z(u, v)\mathbf{k} \quad (u, v) \in D$$

它的**偏微商**定义为

$$\frac{\partial \mathbf{r}}{\partial u} = \frac{\partial x}{\partial u} \mathbf{i} + \frac{\partial y}{\partial u} \mathbf{j} + \frac{\partial z}{\partial u} \mathbf{k}$$

$$\frac{\partial \mathbf{r}}{\partial v} = \frac{\partial x}{\partial v} \mathbf{i} + \frac{\partial y}{\partial v} \mathbf{j} + \frac{\partial z}{\partial v} \mathbf{k}$$

其**微分**定义为

$$d\vec{x}(u,v) = \frac{\partial \vec{x}}{\partial u} du \vec{i} + \frac{\partial \vec{x}}{\partial v} dv \vec{j}$$

$$dr(u,v) = dx(u,v)\vec{i} + dy(u,v)\vec{j} + dz(u,v)\vec{k}$$

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$$= \frac{\partial \vec{r}}{\partial u} du + \frac{\partial \vec{r}}{\partial v} dv$$

## ■ 向量值函数的微分

设有向量函数  $f: \mathbf{R}^n \rightarrow \mathbf{R}^m$ , 分量形式为  
 $m$  维  $n$  元

$$\mathbf{y} = (y_1, y_2, \dots, y_m) = (f_1(\mathbf{x}), f_2(\mathbf{x}), \dots, f_m(\mathbf{x}))$$

其中  $\mathbf{x} = (x_1, x_2, \dots, x_n)$ ,  $f$  的微分定义为

$$df(\mathbf{x}) = (df_1(\mathbf{x}), df_2(\mathbf{x}), \dots, df_m(\mathbf{x}))$$

由于

$$df_j = \sum_{i=1}^n \frac{\partial f_j}{\partial x_i} dx_i \quad df_j = \left( \frac{\partial f_j}{\partial x_1}, \frac{\partial f_j}{\partial x_2}, \dots \right) \begin{pmatrix} dx_1 \\ dx_2 \\ \vdots \\ dx_n \end{pmatrix}$$

导出

$$\begin{pmatrix} df_1 \\ \vdots \\ df_m \end{pmatrix} = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \dots & \frac{\partial f_m}{\partial x_n} \end{pmatrix} \begin{pmatrix} dx_1 \\ \vdots \\ dx_n \end{pmatrix}$$

或

$$df(x) = \frac{\partial f}{\partial x_1} dx_1 + \frac{\partial f}{\partial x_2} dx_2 + \dots + \frac{\partial f}{\partial x_n} dx_n$$

记  $f$  的**Jacobi**矩阵为

$$J_x(f) = \begin{pmatrix} \frac{\partial y_1}{\partial x_1} & \dots & \frac{\partial y_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial y_m}{\partial x_1} & \dots & \frac{\partial y_m}{\partial x_n} \end{pmatrix}$$

当  $m = n$  时, 该方阵的行列式

$$\det J_x(f) = \frac{\partial(y_1, y_2, \dots, y_n)}{\partial(x_1, x_2, \dots, x_n)}$$

称为  $f$  的**Jacobi**行列式.

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$$\begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} \sin t + \cos s \\ st \\ e^{st} \end{pmatrix} \quad y = f(s, t)$$

$$J(f) = \begin{pmatrix} -\sin s, \cos t \\ t, s \\ e^{st}, e^s \end{pmatrix}$$

## ➤ 向量值复合函数链法则

$$\mathbf{f}(\xi, \zeta) = \begin{pmatrix} f_1(\xi, \zeta) \\ f_2(\xi, \zeta) \end{pmatrix}, \quad \mathbf{g}(x, y) = \begin{pmatrix} \xi(x, y) \\ \zeta(x, y) \end{pmatrix}$$

可微, 则  $\mathbf{f} \circ \mathbf{g} = \begin{pmatrix} u_1(x, y) \\ u_2(x, y) \end{pmatrix}$  的**Jacobi矩阵**

$$J(\mathbf{f} \circ \mathbf{g}) = \begin{pmatrix} \frac{\partial u_1}{\partial x} & \frac{\partial u_1}{\partial y} \\ \frac{\partial u_2}{\partial x} & \frac{\partial u_2}{\partial y} \end{pmatrix} = \begin{pmatrix} \frac{\partial f_1}{\partial \xi} & \frac{\partial f_1}{\partial \zeta} \\ \frac{\partial f_2}{\partial \xi} & \frac{\partial f_2}{\partial \zeta} \end{pmatrix} \begin{pmatrix} \frac{\partial \xi}{\partial x} & \frac{\partial \xi}{\partial y} \\ \frac{\partial \zeta}{\partial x} & \frac{\partial \zeta}{\partial y} \end{pmatrix} = J(\mathbf{f}) \cdot J(\mathbf{g})$$



➤ 想一想  $f$  为  $m$  维  $k$  元,  $\mathbf{g}$  为  $k$  维  $n$  元向量值函数的情形