

Lecture 15

§1 Chain rule

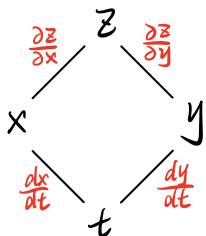
1. Special Case 1 : $f(x(t), y(t))$

令 $Z = f(x, y)$ 在 (x_0, y_0) 处 differentiable, 且 $x = x(t)$ 与 $y = y(t)$ 在 t_0 处可微, ($x_0 = x(t_0)$, $y_0 = y(t_0)$)

1^o 线式法则:

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

2^o branch diagram:



3^o 证明

- Fix $t = t_0$, 令 t 有一个改变量 Δt , 由此导致 $\Delta x, \Delta y, \Delta z$

- 因为 f 在 (x_0, y_0) 处可微,

$$\Delta z = (f_x(x_0, y_0) + \varepsilon_1) \Delta x + (f_y(x_0, y_0) + \varepsilon_2) \Delta y \quad ①$$

当 $(\Delta x, \Delta y) \rightarrow (0, 0)$ 时, $\varepsilon_1 \rightarrow 0, \varepsilon_2 \rightarrow 0$.

- 因为 x, y 在 t_0 处可导,

$$\Delta x = (x'(t_0) + \varepsilon_x) \Delta t$$

$$\Delta y = (y'(t_0) + \varepsilon_y) \Delta t$$

当 $\Delta t \rightarrow 0$ 时, $\varepsilon_x \rightarrow 0, \varepsilon_y \rightarrow 0$

- 由此, ① 式变为

$$\Delta z = (f_x(x_0, y_0) + \varepsilon_1) \cdot (x'(t_0) + \varepsilon_x) \cdot \Delta t + (f_y(x_0, y_0) + \varepsilon_2) \cdot (y'(t_0) + \varepsilon_y) \cdot \Delta t$$

因此,

$$\frac{\Delta z}{\Delta t} = (f_x(x_0, y_0) + \varepsilon_1) \cdot (x'(t_0) + \varepsilon_x) + (f_y(x_0, y_0) + \varepsilon_2) \cdot (y'(t_0) + \varepsilon_y) \quad ②$$

- 因为随着 $\Delta t \rightarrow 0$, $(\Delta x, \Delta y) \rightarrow (0, 0)$,

因此 $\varepsilon_1 \rightarrow 0, \varepsilon_2 \rightarrow 0$

- 对 ② 式两侧同时取 $\lim_{\Delta t \rightarrow 0}$ 得:

$$\frac{dz}{dt} \Big|_{t=t_0} = f_x(x_0, y_0) \cdot x'(t_0) + f_y(x_0, y_0) \cdot y'(t_0)$$

例: Suppose that $Z = x^2y + 3xy^4$, $x = \sin(2t)$ and $y = \cos t$. Find $\frac{dz}{dt}$ when $t=0$.

$$\frac{\partial z}{\partial x} = 2yx + 3y^4, \quad \frac{\partial z}{\partial y} = x^2 + 12xy^3, \quad \frac{dx}{dt} = 2\cos(2t), \quad \frac{dy}{dt} = -\sin t$$

$$\frac{dz}{dt} \Big|_{t=0} = 3 \times 2 + 0 \times 0 = 6$$

2. General Case 1: $f(x_1(t), x_2(t), \dots, x_n(t))$

令 $w = f(x_1, x_2, \dots, x_n)$, $x_1 = x_1(t)$, ..., $x_n = x_n(t)$ 可微分. 则 w 对 t 可微, 且

$$\frac{dw}{dt} = \frac{\partial w}{\partial x_1} \cdot \frac{dx_1}{dt} + \dots + \frac{\partial w}{\partial x_n} \cdot \frac{dx_n}{dt}$$

其中 x_i 为 f 的第 i 个变量

3. Special Case 2: $z = f(x, y)$, $x = x(s, t)$, $y = y(s, t)$

令 $z = f(x, y)$, $x = x(s, t)$, $y = y(s, t)$ 可微分, 则 z 对 (s, t) 可微分, 且

$$\frac{\partial z}{\partial s} = \frac{\partial z}{\partial x} \cdot \frac{\partial x}{\partial s} + \frac{\partial z}{\partial y} \cdot \frac{\partial y}{\partial s}$$

$$\frac{\partial z}{\partial t} = \frac{\partial z}{\partial x} \cdot \frac{\partial x}{\partial t} + \frac{\partial z}{\partial y} \cdot \frac{\partial y}{\partial t}$$

例: Suppose that $z = e^x \sin y$, $x = st^2$ and $y = s^2t$. Find $\frac{\partial z}{\partial s}$

$$\frac{\partial z}{\partial x} = \sin y \cdot e^x, \quad \frac{\partial z}{\partial y} = e^x \cos y, \quad \frac{\partial x}{\partial s} = t^2, \quad \frac{\partial y}{\partial s} = 2st$$

$$\begin{aligned} \frac{\partial z}{\partial s} &= \sin y \cdot e^x \cdot t^2 + e^x \cos y \cdot 2st \\ &= \sin(s^2t) \cdot e^{st^2} \cdot t^2 + e^{st^2} \cos(s^2t) \cdot 2st \end{aligned}$$

4. General Case 2

Theorem (Chain Rule)

Suppose that $u = f(x_1, x_2, \dots, x_n)$ is a differentiable real-valued function with n variables, and suppose that for each i ,

$x_i = g_i(t_1, t_2, \dots, t_m)$ is a differentiable real-valued function with m variables. Then u is differentiable with respect to

(t_1, t_2, \dots, t_m) , and

$$\frac{\partial u}{\partial t_j} = \frac{\partial u}{\partial x_1} \frac{\partial x_1}{\partial t_j} + \frac{\partial u}{\partial x_2} \frac{\partial x_2}{\partial t_j} + \dots + \frac{\partial u}{\partial x_n} \frac{\partial x_n}{\partial t_j}.$$

§2 Implicit differentiation

1. 二元函数的隐函数求导

14.4.8

THEOREM 8—A Formula for Implicit Differentiation Suppose that $F(x, y)$ is differentiable and that the equation $F(x, y) = 0$ defines y as a differentiable function of x . Then at any point where $F_y \neq 0$,

$$\frac{dy}{dx} = -\frac{F_x}{F_y}. \quad (1)$$

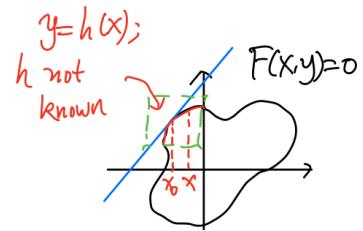
证明:

- 在 (x_0, y_0) 附近, y 可以视作 x 的隐函数 $h(x)$, given by $F(x, y) = 0$. 则在 (x_0, y_0) 附近, 曲线上的任一点均满足 $F(x, h(x)) = 0$.

- 令 $Z = F(x, y) = F(x, h(x))$, 则 $\frac{dz}{dx}|_{(x,y)=(x_0,y_0)} = 0$
- 根据 chain rule, 若 $F(x, y)$ 可微, 则

$$\frac{dz}{dx} = F_x \cdot \frac{dx}{dx} + F_y \cdot \frac{dy}{dx} = F_x + F_y \frac{dy}{dx} = 0$$

由此: $\frac{dy}{dx} = -\frac{F_x}{F_y}$ (if $F_y \neq 0$)



例: Find the slope of the tangent line to the curve $x^3 + y^3 = bxy$ at the point $(3, 3)$

$$F(x, y) = x^3 + y^3 - bxy$$

$$F_x = 3x^2 - by, \quad F_y = 3y^2 - bx$$

$$\frac{dy}{dx}|_{(x,y)=(3,3)} = -\frac{F_x}{F_y} = -1$$

2. 三元函数的偏导数求导

$$\frac{\partial z}{\partial x} = -\frac{F_x}{F_z} \text{ and } \frac{\partial z}{\partial y} = -\frac{F_y}{F_z}, \text{ whenever } F_z \neq 0.$$

例: Find $\frac{\partial z}{\partial x}$ and $\frac{\partial z}{\partial y}$ if $x^3 + y^3 + z^3 + bxy = 1$

$$F_x = 3x^2 + by$$

$$F_y = 3y^2 + bx$$

$$F_z = 3z^2$$

$$\frac{\partial z}{\partial x} = -\frac{F_x}{F_z} = -\frac{x^2+by}{z^2}$$

$$\frac{\partial z}{\partial y} = -\frac{F_y}{F_z} = -\frac{y^2+bx}{z^2}$$

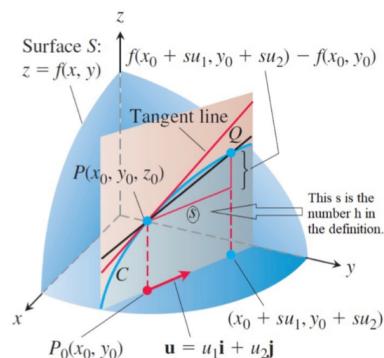
3.3 Directional Derivative and Gradient Vectors

1. Directional derivative (方向导数) 的定义

Definition

Let (x_0, y_0) be an interior point of the domain of a two-variable function f , and let \vec{u} be a unit vector in \mathbb{R}^2 . The (directional) derivative of f at (x_0, y_0) in the direction of \vec{u} , denoted by $D_{\vec{u}}f(x_0, y_0)$, is defined by

$$D_{\vec{u}}f(x_0, y_0) := \lim_{h \rightarrow 0} \frac{f(x_0 + hu_1, y_0 + hu_2) - f(x_0, y_0)}{h}.$$



注: 1° $\vec{u} = u_1 \vec{i} + u_2 \vec{j}$ 可以视作方向余弦

2° 数值上偏导可以视作方向导数的 special case:

$$f_x = D_{\vec{i}}f, \quad \vec{i} = \langle 1, 0 \rangle$$

$$f_y = D_{\vec{j}}f, \quad \vec{j} = \langle 0, 1 \rangle$$

3° Vector notation (generalize to n-variable functions)

$$D_{\vec{u}}f(\vec{x}_0) = \lim_{h \rightarrow 0} \frac{f(\vec{x}_0 + h\vec{u}) - f(\vec{x}_0)}{h}$$

2. 方向导数的性质与梯度

Theorem

If f is a differentiable two-variable function, then

$$D_{\vec{u}}f(x_0, y_0) = f_x(x_0, y_0)u_1 + f_y(x_0, y_0)u_2.$$

In other words,

$$D_{\vec{u}}f(x_0, y_0) = \nabla f(x_0, y_0) \cdot \vec{u},$$

where $\nabla f := \langle f_x, f_y \rangle$.

- 注: 1° 向量 ∇f 被称为 gradient (梯度) 或 gradient vector of f
2° 对于 n 元函数 $f(x_1, \dots, x_n)$, $\nabla f := \langle f_{x_1}, \dots, f_{x_n} \rangle$.
若 f 在一点 $\vec{x}_0 \in \mathbb{R}^n$ 处可微, 则有 $D_{\vec{u}}f(\vec{x}_0) = \nabla f(\vec{x}_0) \cdot \vec{u}$

证明:

- 定义 $g(t) := f(x_0 + tu_1, y_0 + tu_2)$. 则 $g(0) = f(x_0, y_0)$, 且
$$D_{\vec{u}}f(x_0, y_0) = \lim_{h \rightarrow 0} \frac{g(h) - g(0)}{h} = g'(0)$$

- 根据链式法则,

$$\begin{aligned} g'(0) &= f_x(x(0), y(0)) \cdot x'(0) + f_y(x(0), y(0)) \cdot y'(0) \\ &= f_x(x_0, y_0) \cdot u_1 + f_y(x_0, y_0) \cdot u_2 \\ &= \nabla f(x_0, y_0) \cdot \vec{u} \end{aligned}$$

where $\nabla f := \langle f_x, f_y \rangle$ is the gradient vector.

例: Find the directional derivative of $f(x, y) := xe^y + \cos(xy)$ at the point $(2, 0)$ in the direction of $\vec{v} = \langle 3, -4 \rangle$.

- $\vec{u} = \left\langle \frac{3}{5}, -\frac{4}{5} \right\rangle$
- $f_x = e^y - y \sin(xy)$
- $f_y = xe^y - x \sin(xy)$
- $\nabla f(2, 0) = \langle 1, 2 \rangle$
- $D_{\vec{u}}f(2, 0) = \frac{3}{5} - \frac{8}{5} = -1$

3. 方向导数与可微

existence of all directional derivatives \rightarrow differentiability

e.g. Consider $f(x, y) := \begin{cases} xy^2/(x^2+y^4) & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0) \end{cases}$

- Let \vec{u} be any unit vector $\langle u_1, u_2 \rangle$

$$\begin{aligned}
 D_{\vec{u}} f(0,0) &= \lim_{h \rightarrow 0} \frac{f(hu_1, hu_2) - f(0,0)}{h} \\
 &= \lim_{h \rightarrow 0} \frac{h^3 u_1 u_2^3 / (h^2 u_1^2 + h^4 u_2^4)}{h} \\
 &= \lim_{h \rightarrow 0} \frac{u_1 u_2^2}{u_1^2 + h^2 u_2^4} \\
 &= \begin{cases} u_2^2 / u_1, & \text{if } u_1 \neq 0 \\ 0, & \text{if } u_1 = 0 \end{cases}
 \end{aligned}$$

- 因此， f 在 $(0,0)$ 处的所有方向导数均存在。
但 f 在 $(0,0)$ 处不连续，因此不可微

4. 梯度向量的几何意义

1° 意义1：指向 *fastest increase* 的方向

Theorem

Let f be a differentiable function (with multiple variables), and let \vec{x}_0 represent an interior point of the domain of f . Then, among all directional derivatives $D_{\vec{u}} f(\vec{x}_0)$:

- The maximum value of $D_{\vec{u}} f(\vec{x}_0)$ is $|\nabla f(\vec{x}_0)|$, and it occurs when \vec{u} is the direction of $\nabla f(\vec{x}_0)$.
- The minimum value of $D_{\vec{u}} f(\vec{x}_0)$ is $-|\nabla f(\vec{x}_0)|$, and it occurs when \vec{u} is the direction of $-\nabla f(\vec{x}_0)$.

If $\nabla f(\vec{x}_0) \neq \vec{0}$ and \vec{u} is orthogonal to $\nabla f(\vec{x}_0)$, then $D_{\vec{u}} f(\vec{x}_0) = 0$.

注：上述定理表明，在任意一点，一个可微函数沿着 **梯度方向增长最快**，
沿着 **-梯度方向增长最慢**，增长率为 **$|\nabla f|$** （梯度的模）

例 Exercise

Suppose that the temperature at a point (x, y, z) in space is given by $T(x, y, z) = 80/(1+x^2+2y^2+3z^2)$, where temperature is measured in degrees Celsius and distance is measured in meters. In which direction does the temperature increase fastest at the point $(1, 1, -2)$? What is the maximum rate of increase?

Sol: $T_x(x, y, z) = -\frac{80 \cdot 2x}{(1+x^2+2y^2+3z^2)^2}$

$$T_y(x, y, z) = -\frac{80 \cdot 4y}{(1+x^2+2y^2+3z^2)^2}$$

$$T_z(x, y, z) = -\frac{80 \cdot 6z}{(1+x^2+2y^2+3z^2)^2}$$

$$\nabla T(1, 1, -2) = -\frac{5}{8} \langle 1, 2, -6 \rangle$$

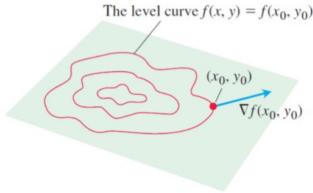
- increases fastest in the direction $\vec{u} = \nabla T(1, 1, -2) / |\nabla T(1, 1, -2)| = -\frac{1}{\sqrt{41}} \langle 1, 2, 6 \rangle$
- Max $D_{\vec{u}} T(1, 1, -2) = |\nabla T(1, 1, -2)| = \frac{5}{8}\sqrt{41}$ ($^{\circ}\text{C}/\text{m}$)

2^o 意义2：与 level curves 的联系.

平面上一曲线(等位线)上 (x_0, y_0) 处切线为： $f_x(x_0, y_0) \cdot (x - x_0) + f_y(x_0, y_0) \cdot (y - y_0) = 0$

Theorem

At any point (x_0, y_0) in the domain of a differentiable function f , if C is the level curve through (x_0, y_0) , then the gradient of f at (x_0, y_0) is perpendicular to the tangent vector to C at (x_0, y_0) .



This geometric fact allows us to describe the tangent line to the level curve of f at a point (x_0, y_0) using the equation

$$f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) = 0.$$

证明：

- 令 $f(x, y)$ 在 (x_0, y_0) 处可微.

令一条曲线 C 在定义域平面内经过 (x_0, y_0) , 其参数方程为 $\vec{r}(t) = \langle x(t), y(t) \rangle$

考虑沿曲线 C 的 "height" : $Z = f(x, y)$

$$Z = f(x(t), y(t)) = f(\vec{r}(t))$$

- By chain rule .

$$\frac{dZ}{dt}|_{t=t_0} = f_x(x(t_0), y(t_0)) x'(t_0) + f_y(x(t_0), y(t_0)) y'(t_0),$$

$$\text{i.e. } (f \circ \vec{r})'(t_0) = \nabla f(\vec{r}(t_0)) \cdot \vec{r}'(t_0)$$

- 若令 $f(x, y) = K$, 则对于 C 上任一点 (x_0, y_0) , 有 $\frac{dZ}{dt} = 0$, 即
$$\nabla f(x_0, y_0) \cdot \vec{r}'|_{(x_0, y_0)} = 0$$

例: Find an equation for the tangent line to the ellipse $\frac{x^2}{8} + \frac{y^2}{2} = 1$ at point $(-2, 1)$

$$f(x, y) = \frac{x^2}{8} + \frac{y^2}{2}$$

$$f_x = \frac{x}{4}, \quad f_y = y$$

$$f_x(-2, 1) = -\frac{1}{2}, \quad f_y(-2, 1) = 1$$

$$\text{Tangent is } -\frac{1}{2}(x+2) + 1 \cdot (y-1) = 0 \Rightarrow y = \frac{1}{2}x + 2$$

5. 梯度的性质

Algebra Rules for Gradients

1. Sum Rule: $\nabla(f + g) = \nabla f + \nabla g$

2. Difference Rule: $\nabla(f - g) = \nabla f - \nabla g$

3. Constant Multiple Rule: $\nabla(kf) = k\nabla f$ (any number k)

4. Product Rule: $\nabla(fg) = f\nabla g + g\nabla f$ Scalar multipliers on left of gradients

5. Quotient Rule: $\nabla\left(\frac{f}{g}\right) = \frac{g\nabla f - f\nabla g}{g^2}$