

高等数学习题课

数据科学与工程学院

王婧

1. 曲线的切线
2. 曲面的法向量
3. 梯度
4. 方向导数
5. 二元函数的泰勒展开
6. 最小二乘法

空间曲面的切平面与法线方程

空间曲面方程为: $F(x,y,z)=0$

曲面 Σ 在点 $M_0(x_0, y_0, z_0)$ 处的切平面方程为

$$F'_x(x_0, y_0, z_0)(x-x_0)+F'_y(x_0, y_0, z_0)(y-y_0) \\ +F'_z(x_0, y_0, z_0)(z-z_0)=0$$

曲面 Σ 在点 M_0 处的法线方程为

$$\frac{x-x_0}{F'_x(x_0, y_0, z_0)} = \frac{y-y_0}{F'_y(x_0, y_0, z_0)} = \frac{z-z_0}{F'_z(x_0, y_0, z_0)}.$$

梯度:

$$c = \nabla f(\mathbf{x}^*) = \begin{bmatrix} \frac{\partial f(\mathbf{x}^*)}{\partial x_1} \\ \frac{\partial f(\mathbf{x}^*)}{\partial x_2} \\ \vdots \\ \frac{\partial f(\mathbf{x}^*)}{\partial x_n} \end{bmatrix} = \left[\frac{\partial f(\mathbf{x}^*)}{\partial x_1} \frac{\partial f(\mathbf{x}^*)}{\partial x_2} \dots \frac{\partial f(\mathbf{x}^*)}{\partial x_n} \right]^T$$

Hessian矩阵:

$$\frac{\partial^2 f}{\partial \mathbf{x} \partial \mathbf{x}} = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} & \cdots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

若 $f(\mathbf{X})$ 具有二阶连续偏导, 则 $\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}$,
则Hessian矩阵为对称矩阵。

方向导数的计算

若函数 $u=f(x,y,z)$ 在点 $P_0(x_0, y_0, z_0)$ 处可微,
则函数 f 在点 P_0 处沿任意方向的方向导数为,

$$\frac{\partial f}{\partial l} = \frac{\partial u}{\partial x} \cos \alpha + \frac{\partial u}{\partial y} \cos \beta + \frac{\partial u}{\partial z} \cos \gamma$$

即
$$\frac{\partial f}{\partial l} = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right) \cdot (\cos \alpha, \cos \beta, \cos \gamma)$$

L 方向的单位向量



一元函数 $y=f(x)$ 的泰勒(Taylor)公式,

$$f(x) = f(x_0) + f'(x_0)(x - x_0) + \cdots + \frac{f^{(n)}(x_0)}{n!}(x - x_0)^n \\ + \frac{f^{(n+1)}(x_0 + \theta(x - x_0))}{(n+1)!}(x - x_0)^{n+1} \quad (0 < \theta < 1)$$

可用 n 次多项式来近似表达函数 $f(x)$ ，且误差是当 $x \rightarrow x_0$ 时比 $(x - x_0)^n$ 高阶的无穷小.

下面将考虑用多个变量的多项式来近似表达一个给定的多元函数，并能具体地估算出误差的大小.

二元函数的泰勒(Taylor)公式,

定理 若函数 $z=f(x,y)$ 在点 $P_0(x_0,y_0)$ 的某一邻域 $U(P)$ 内具有直至 $n+1$ 阶的连续偏导数, 点 $(x_0+h,y_0+k) \in U(P)$ 则有

$$f(x_0+h,y_0+k) = f(x_0,y_0) + (h\frac{\partial}{\partial x} + k\frac{\partial}{\partial y})f(x_0,y_0) + \frac{1}{2!}(h\frac{\partial}{\partial x} + k\frac{\partial}{\partial y})^2 f(x_0,y_0) + \cdots + \frac{1}{n!}(h\frac{\partial}{\partial x} + k\frac{\partial}{\partial y})^n f(x_0,y_0) + R_n \quad (1)$$

$$R_n = \frac{1}{(n+1)!}(h\frac{\partial}{\partial x} + k\frac{\partial}{\partial y})^{n+1} f(x_0 + \theta h, y_0 + \theta k) \quad (0 < \theta < 1)$$

其中记号

$$\left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y}\right) f(x_0, y_0) = hf_x(x_0, y_0) + kf_y(x_0, y_0)$$

$$\left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y}\right)^2 f(x_0, y_0) = h^2 f_{xx}(x_0, y_0) + 2hkf_{xy}(x_0, y_0) + k^2 f_{yy}(x_0, y_0)$$

一般地, 记号 $\left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y}\right)^m f(x_0, y_0)$ 表示为

$$\sum_{p=0}^m C_m^p h^p k^{m-p} \frac{\partial^m f}{\partial x^p \partial y^{m-p}} \Big|_{(x_0, y_0)}$$

利用一元函数的泰勒公式进行证明.

一元函数泰勒展开:

$$f(x) = f(x^*) + \frac{df(x^*)}{dx}(x - x^*) + \frac{1}{2} \frac{d^2 f(x^*)}{dx^2}(x - x^*)^2 + R$$

$x - x^* = d$ (a small change in the point x^*).

$$f(x^* + d) = f(x^*) + \frac{df(x^*)}{dx}d + \frac{1}{2} \frac{d^2 f(x^*)}{dx^2}d^2 + R$$

二元函数泰勒展开:

$$\begin{aligned} f(x_1, x_2) = & f(x_1^*, x_2^*) + \frac{\partial f}{\partial x_1}(x_1 - x_1^*) + \frac{\partial f}{\partial x_2}(x_2 - x_2^*) \\ & + \frac{1}{2} \left[\frac{\partial^2 f}{\partial x_1^2}(x_1 - x_1^*)^2 + 2 \frac{\partial^2 f}{\partial x_1 \partial x_2}(x_1 - x_1^*)(x_2 - x_2^*) + \frac{\partial^2 f}{\partial x_2^2}(x_2 - x_2^*)^2 \right] + R \end{aligned}$$

$$\begin{aligned} f(x_1, x_2) = & f(x_1^*, x_2^*) + \sum_{i=1}^2 \frac{\partial f}{\partial x_i}(x_i - x_i^*) \\ & + \frac{1}{2} \sum_{i=1}^2 \sum_{j=1}^2 \frac{\partial^2 f}{\partial x_i \partial x_j}(x_i - x_i^*)(x_j - x_j^*) + R \end{aligned}$$

$$f(\mathbf{x}) = f(\mathbf{x}^*) + \nabla f^T(\mathbf{x} - \mathbf{x}^*) + \frac{1}{2}(\mathbf{x} - \mathbf{x}^*)^T \mathbf{H}(\mathbf{x} - \mathbf{x}^*) + R$$

n 元函数泰勒展开:

$$f(\mathbf{x}) = f(\mathbf{x}^*) + \nabla f^T(\mathbf{x} - \mathbf{x}^*) + \frac{1}{2}(\mathbf{x} - \mathbf{x}^*)^T \mathbf{H}(\mathbf{x} - \mathbf{x}^*) + R$$

$$f(\mathbf{x}^* + \mathbf{d}) = f(\mathbf{x}^*) + \nabla f^T \mathbf{d} + \frac{1}{2} \mathbf{d}^T \mathbf{H} \mathbf{d} + R$$

求曲面 $x^2 + y^2 + z^2 = 2$ 上平行于平面 $x + 4y - z = 0$ 的切平面方程。

求曲面 $z = x^2 + y^2$ 在点 $(1, 1, 2)$ 处的切平面方程。

在点 $(1, 2)$ 处，计算多元函数的梯度和Hessian矩阵：

$$f(\mathbf{x}) = x_1^3 + x_2^3 + 2x_1^2 + 3x_2^2 - x_1x_2 + 2x_1 + 4x_2$$

求函数 $u = \ln(x + y^2 + z^2)$ 在点 $M_0(0, 1, 2)$ 处，沿向量 $(2, -1, -1)$ 的方向的方向导数。

求曲面 $x^2 + y^2 + z^2 = 2$ 上平行于平面 $x + 4y - z = 0$ 的切平面方程。

解： 空间曲面法向量

设函数 $F(x, y, z) = x^2 + y^2 + z^2 - 2$,

则 $n = (F_x, F_y, F_z) = (2x, 2y, 2z)$

设切点为 (x_0, y_0, z_0) , 则

$$\frac{x_0}{1} = \frac{y_0}{4} = \frac{z_0}{-1}$$
$$x_0^2 + y_0^2 + z_0^2 = 2$$

切点为 $(1/3, 4/3, -1/3), (-1/3, -4/3, 1/3)$

切面方程为：

$$(x - \frac{1}{3}) + 4(y - \frac{4}{3}) - (z + \frac{1}{3}) = 0 \text{ 或 } (x + \frac{1}{3}) + 4(y + \frac{4}{3}) - (z - \frac{1}{3}) = 0$$

$$\text{即 } x + 4y - z = 6 \text{ 或 } x + 4y - z = -6$$

求曲面 $z = x^2 + y^2$ 在点 $(1, 1, 2)$ 处的切平面方程。

解： 设函数 $F(x, y, z) = x^2 + y^2 - z$,
则 $\mathbf{n} = (F_x, F_y, F_z) = (2x, 2y, -1)$
切点为 $(1, 1, 2)$, 则
 $\mathbf{n} = (2, 2, -1)$
 $2(x - 1) + 2(y - 1) - 1(z - 2) = 0$

$$2x + 2y - z - 2 = 0$$

练习：
在点(1, 2)处，计算多元函数的梯度和Hessian矩阵：

$$f(\mathbf{x}) = x_1^3 + x_2^3 + 2x_1^2 + 3x_2^2 - x_1x_2 + 2x_1 + 4x_2$$

解：

$$\frac{\partial f}{\partial x_1} = 3x_1^2 + 4x_1 - x_2 + 2; \quad \frac{\partial f}{\partial x_2} = 3x_2^2 + 6x_2 - x_1 + 4$$

$$\mathbf{c} = (7, 27).$$

$$\frac{\partial^2 f}{\partial x_1^2} = 6x_1 + 4; \quad \frac{\partial^2 f}{\partial x_1 \partial x_2} = -1; \quad \frac{\partial^2 f}{\partial x_2 \partial x_1} = -1; \quad \frac{\partial^2 f}{\partial x_2^2} = 6x_2 + 6.$$

$$\mathbf{H}(1, 2) = \begin{bmatrix} 10 & -1 \\ -1 & 18 \end{bmatrix}$$

练习：

求函数 $u = \ln(x + y^2 + z^2)$ 在点 $M_0(0, 1, 2)$ 处，沿向量 $(2, -1, -1)$ 的方向的方向导数。

解：

所给向量的方向余弦为 $(\frac{2}{\sqrt{6}}, -\frac{1}{\sqrt{6}}, -\frac{1}{\sqrt{6}})$

点 M_0 处的梯度为 $(\frac{1}{5}, \frac{2}{5}, \frac{4}{5})$

则 M_0 处，沿所给向量的方向导数为 $-\frac{4}{5\sqrt{6}}$

1. 求函数 $f(x, y) = 2x^2 - xy - y^2 - 6x - 3y + 5$ 在点 $(1, -2)$ 的泰勒公式.

解: $f(1, -2) = 5, f'_x(1, -2) = (4x - y - 6) \Big|_{(1, -2)} = 0, f'_y(1, -2) = (-x - 2y - 3) \Big|_{(1, -2)} = 0,$

$$f''_{xx}(1, -2) = 4, f''_{xy} = -1, f''_{yy}(1, -2) = -2.$$

又阶数为 3 的各偏导数为零, 则

$$\begin{aligned} f(x, y) &= f[1 + (x-1), -2 + (y+2)] \\ &= f(1, -2) + (x-1)f'_x(1, -2) + (y+2)f'_y(1, -2) + \\ &\quad \frac{1}{2!}[(x-1)^2 f''_{xx}(1, -2) + 2(x-1)(y+2)f''_{xy}(1, -2) + (y+2)^2 f''_{yy}(1, -2)] \\ &= 5 + \frac{1}{2!}[4(x-1)^2 - 2(x-1)(y+2) - 2(y+2)^2] \\ &= 5 + 2(x-1)^2 - (x-1)(y+2) - (y+2)^2. \end{aligned}$$



1. 某种合金的含铅量百分比(%)为 p , 其溶解温度($^{\circ}\text{C}$)为 θ , 由实验测得 p 与 θ 的数据如下表:

$p/\%$	36.9	46.7	63.7	77.8	84.0	87.5
$\theta/^{\circ}\text{C}$	181	197	235	270	283	292

试用最小二乘法建立 θ 与 p 之间的经验公式 $\theta = ap + b$.

解: 由方程组
$$\begin{cases} a \sum_{i=1}^6 p_i^2 + b \sum_{i=1}^6 p_i = \sum_{i=1}^6 \theta_i p_i, \\ a \sum_{i=1}^6 p_i + 6b = \sum_{i=1}^6 \theta_i, \end{cases}$$
 确定先验公式中的 a, b , 先算出

$$\begin{cases} \sum_{i=1}^6 p_i^2 = 28\,365.28, & \sum_{i=1}^6 p_i = 396.6, \\ \sum_{i=1}^6 \theta_i p_i = 101\,176.3, & \sum_{i=1}^6 \theta_i = 1\,458, \end{cases}$$

代入方程组, 得
$$\begin{cases} 28\,365.28a + 396.6b = 101\,176.3, \\ 396.6a + 6b = 1\,458. \end{cases}$$
 解此方程组, 得 $a = 2.234, b = 95.33$.

则经验公式 $\theta = 2.234p + 95.33$.



2. 已知一组实验数据为 $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$. 现若假定经验公式是 $y = ax^2 + bx + c$. 试按最小二乘法建立 a, b, c 应满足的三元一次方程组.

解: 设 M 是每个数据差的平方和: $M = \sum_{i=1}^n [y_i - (ax_i^2 + bx_i + c)]^2 = M(a, b, c)$, 令

$$\begin{cases} \frac{\partial M}{\partial a} = -2 \sum_{i=1}^n [y_i - (ax_i^2 + bx_i + c)](x_i)^2 = 0, \\ \frac{\partial M}{\partial b} = -2 \sum_{i=1}^n [y_i - (ax_i^2 + bx_i + c)](x_i) = 0, \\ \frac{\partial M}{\partial c} = -2 \sum_{i=1}^n [y_i - (ax_i^2 + bx_i + c)] = 0, \end{cases}$$

$$\text{即} \begin{cases} \sum_{i=1}^n (y_i x_i^2 - ax_i^4 - bx_i^3 - cx_i^2) = 0, \\ \sum_{i=1}^n (y_i x_i - ax_i^3 - bx_i^2 - cx_i) = 0, \\ \sum_{i=1}^n (y_i - ax_i^2 - bx_i - c) = 0, \end{cases} \text{ 即} \begin{cases} a \sum_{i=1}^n x_i^4 + b \sum_{i=1}^n x_i^3 + c \sum_{i=1}^n x_i^2 = \sum_{i=1}^n x_i^2 y_i, \\ a \sum_{i=1}^n x_i^3 + b \sum_{i=1}^n x_i^2 + c \sum_{i=1}^n x_i = \sum_{i=1}^n x_i y_i, \\ a \sum_{i=1}^n x_i^2 + b \sum_{i=1}^n x_i + nc = \sum_{i=1}^n y_i. \end{cases}$$

