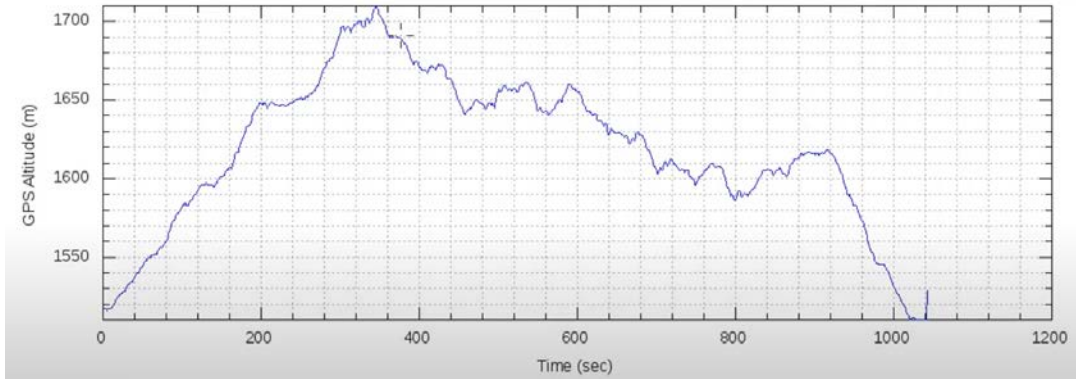


数值微分

1. 数值微分：前向和后向差分的推导



- 等间距数据
- 非等间距数据

导数的差商定义

$$f'(x) \triangleq \lim_{\Delta x \rightarrow 0} \frac{f(x + \Delta x) - f(x)}{\Delta x} \quad (1.1)$$

泰勒级数-推导前向和后向差商：

$$f(x) = f(a) + \frac{f'(a)}{1!}(x-a) + \frac{f''(a)}{2!}(x-a)^2 + \frac{f'''(a)}{3!}(x-a)^3 + \dots \quad (1.2)$$

$$f(x_{i-1}) = f(x_i) + \frac{f'(x_i)}{1!}(x_{i-1} - x_i) + \frac{f''(x_i)}{2!}(x_{i-1} - x_i)^2 + \frac{f'''(x_i)}{3!}(x_{i-1} - x_i)^3 + \dots \quad (1.3)$$

令 $h = x_i - x_{i-1} (\rightarrow x_{i-1} = x_i - h)$

$$f(x_{i-1}) = f(x_i) - \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \dots \quad (1.4)$$

$$\Rightarrow f'(x_i) = \frac{1}{h} f(x_i) - f(x_{i-1}) + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \dots \quad (1.5)$$

舍去 $\left[\frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \dots\right]$ 得到

$$\begin{aligned} f'(x_i) &= \frac{f(x_i) - f(x_{i-1})}{h} + O(h) \\ f'(x_i) &= \frac{f(x_i) - f(x_{i-h})}{h} + O(h) \end{aligned} \quad (1.6)$$

以上是向后差商近似导数。前向差商使用 x_{i+1} 和 h ，同时 $h = x_{i+1} - x_i$

$$\begin{aligned} f(x_{i+1}) &= f(x_i) + \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(x_i)}{3!}h^3 + \dots \\ \Rightarrow f'(x_i) &= \frac{1}{h} \left[-f(x_i) + f(x_{i+1}) - \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 - \dots \right] \\ f'(x_i) &= -\frac{1}{h}f(x_i) + \frac{1}{h}f(x_{i+1}) - \frac{f''(x_i)}{2!}h - \frac{f'''(x_i)}{3!}h^2 - \dots \\ f'(x_i) &= \frac{f(x_{i+1}) - f(x_i)}{h} + O(h) \end{aligned} \quad (1.7)$$

对比导数定义

$$f'(x) \triangleq \lim_{\Delta x \rightarrow 0} \frac{f(x + \Delta x) - f(x)}{\Delta x} \quad (1.8)$$

2. 数值微分：中心差分法的推导

考虑对 x_{i-1} 和 x_{i+1} 的泰勒级数

$$\begin{aligned} f(x_{i-1}) &= f(x_i) - \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \dots \\ f(x_{i+1}) &= f(x_i) + \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(x_i)}{3!}h^3 + \dots \end{aligned} \quad (1.9)$$

对 $f(x)$ 得到

$$\begin{aligned} -f(x_i) &= -f(x_{i-1}) - \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \dots \\ -f(x_i) &= -f(x_{i+1}) + \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(x_i)}{3!}h^3 + \dots \end{aligned} \quad (1.10)$$

令 h 相同，上式得到：

$$0 = -f(x_{i+1}) + f(x_{i-1}) + 2\frac{f'(x_i)}{1!}h + 2\frac{f'''(x_i)}{3!}h^3 + \dots \quad (1.11)$$

$$\Rightarrow -f'(x_i) = \frac{1}{2h} \left[f(x_{i-1}) - f(x_{i+1}) + 2\frac{f'''(x_i)}{3!}h^3 + \dots \right] \quad (1.12)$$

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1})}{2h} - \frac{f'''(x_i)}{3!}h^2 + \dots \quad (1.13)$$

$$f'(x_i) - \frac{f(x_{i+1}) - f(x_{i-1})}{2h} + O(h^2) \quad (1.14)$$

由上述几个公式可知，步长 h 越小，计算结果越准确。但从计算角度看， h 越小， $f(x_0 + h)$ ， $f(x_0)$ 与 $f(x_0 - h)$ 越接近，直接相减会造成有效数字的严重损失，因此，需要合理选取步长，至于步长的合理选取可参阅相关参考书。

3. 数值微分：高阶微分

我们可以接着往下做：

$$\begin{aligned} f(x_{i-1}) &= f(x_i) - \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \dots \\ f(x_{i+1}) &= f(x_i) + \frac{f'(x_i)}{1!}h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(x_i)}{3!}h^3 + \dots \\ f(x_{i-2}) &= f(x_i) - \frac{f'(x_i)}{1!}2h \dots \\ f(x_{i+2}) &= f(x_i) + \frac{f'(x_i)}{1!}2h \dots \end{aligned} \quad (1.15)$$

考虑

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!}h^2 + \dots \quad (1.16)$$

$$f(x_{i+2}) = f(x_i) + f'(x_i)(2h) + \frac{f''(x_i)}{2!}(2h)^2 + \dots \quad (1.17)$$

得到

$$f(x_{i+2}) - 2f(x_{i+1}) = -f(x_i) + f''(x_i)h^2 + \dots \quad (1.18)$$

变换得到:

$$f''(x_i) = \frac{f(x_{i+2}) - 2f(x_{i+1}) + f(x_i)}{h^2} + O(h) \quad (1.19)$$

重新考虑

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2}h^2 + \dots \quad (1.20)$$

得到

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{h} - \frac{f''(x_i)}{2}h + O(h^2) \quad (1.21)$$

替换得到

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{h} - \frac{f(x_{i+2}) - 2f(x_{i+1}) + f(x_i)}{2h^2}h + O(h^2) \quad (1.22)$$

即

$$f'(x_i) = \frac{-f(x_{i+2}) + 4f(x_{i+1}) - 3f(x_i)}{2h} + O(h^2) \quad (1.23)$$

前向差分

First Derivative

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{h}$$

Error

$$O(h)$$

$$f'(x_i) = \frac{-f(x_{i+2}) + 4f(x_{i+1}) - 3f(x_i)}{2h}$$

$$O(h^2)$$

Second Derivative

$$f''(x_i) = \frac{f(x_{i+2}) - 2f(x_{i+1}) + f(x_i)}{h^2}$$

$$O(h)$$

$$f''(x_i) = \frac{-f(x_{i+3}) + 4f(x_{i+2}) - 5f(x_{i+1}) + 2f(x_i)}{h^2}$$

$$O(h^2)$$

Third Derivative

$$f'''(x_i) = \frac{f(x_{i+3}) - 3f(x_{i+2}) + 3f(x_{i+1}) - f(x_i)}{h^3}$$

$$O(h)$$

$$f'''(x_i) = \frac{-3f(x_{i+4}) + 14f(x_{i+3}) - 24f(x_{i+2}) + 18f(x_{i+1}) - 5f(x_i)}{2h^3}$$

$$O(h^2)$$

Fourth Derivative

$$f^{(4)}(x_i) = \frac{f(x_{i+4}) - 4f(x_{i+3}) + 6f(x_{i+2}) - 4f(x_{i+1}) + f(x_i)}{h^4}$$

$$O(h)$$

$$f^{(4)}(x_i) = \frac{-2f(x_{i+5}) + 11f(x_{i+4}) - 24f(x_{i+3}) + 26f(x_{i+2}) - 14f(x_{i+1}) + 3f(x_i)}{h^4}$$

$$O(h^2)$$

后向差分

First Derivative	Error
$f'(x_i) = \frac{f(x_i) - f(x_{i-1}))}{h}$	$O(h)$
$f'(x_i) = \frac{3f(x_i) - 4f(x_{i-1})) + f(x_{i-2}))}{2h}$	$O(h^2)$
Second Derivative	
$f''(x_i) = \frac{f(x_i) - 2f(x_{i-1})) + f(x_{i-2}))}{h^2}$	$O(h)$
$f''(x_i) = \frac{2f(x_i) - 5f(x_{i-1})) + 4f(x_{i-2})) - f(x_{i-3}))}{h^2}$	$O(h^2)$
Third Derivative	
$f'''(x_i) = \frac{f(x_i) - 3f(x_{i-1})) + 3f(x_{i-2})) - f(x_{i-3}))}{h^3}$	$O(h)$
$f'''(x_i) = \frac{5f(x_i) - 18f(x_{i-1})) + 24f(x_{i-2})) - 14f(x_{i-3})) + 3f(x_{i-4}))}{2h^3}$	$O(h^2)$
Fourth Derivative	
$f^{(4)}(x_i) = \frac{f(x_i) - 4f(x_{i-1})) + 6f(x_{i-2})) - 4f(x_{i-3})) + f(x_{i-4}))}{h^4}$	$O(h)$
$f^{(4)}(x_i) = \frac{3f(x_i) - 14f(x_{i-1})) + 26f(x_{i-2})) - 24f(x_{i-3})) + 11f(x_{i-4})) - 2f(x_{i-5}))}{h^4}$	$O(h^2)$

中心差分

First Derivative	Error
$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1}))}{2h}$	$O(h^2)$
$f'(x_i) = \frac{-f(x_{i+2}) + 8f(x_{i+1}) - 8f(x_{i-1}) + f(x_{i-2}))}{12h}$	$O(h^4)$
Second Derivative	
$f''(x_i) = \frac{f(x_{i+1}) - 2f(x_i) + f(x_{i-1}))}{h^2}$	$O(h^2)$
$f''(x_i) = \frac{-f(x_{i+2}) + 16f(x_{i+1}) - 30f(x_i) + 16f(x_{i-1}) - f(x_{i-2}))}{12h^2}$	$O(h^4)$
Third Derivative	
$f'''(x_i) = \frac{f(x_{i+2}) - 2f(x_{i+1}) + 2f(x_{i-1}) - f(x_{i-2}))}{2h^3}$	$O(h^2)$
$f'''(x_i) = \frac{-f(x_{i+3}) + 8f(x_{i+2}) - 13f(x_{i+1}) + 13f(x_{i-1}) - 8f(x_{i-2}) + f(x_{i-3}))}{8h^3}$	$O(h^4)$
Fourth Derivative	
$f^{(4)}(x_i) = \frac{f(x_{i+2}) - 4f(x_{i+1}) + 6f(x_i) - 4f(x_{i-1}) + f(x_{i-2}))}{h^4}$	$O(h^2)$
$f^{(4)}(x_i) = \frac{-f(x_{i+3}) + 12f(x_{i+2}) - 39f(x_{i+1}) + 56f(x_i) - 39f(x_{i-1}) + 12f(x_{i-2}) - f(x_{i-3}))}{6h^4}$	$O(h^4)$

切线、法线、梯度和法向量

考虑一元函数 $f(x)$ ，导数为 $f'(x)$ ，则对应的切线、法线、梯度和法向量为：

$$f_t(x) = f'(x_0)(x - x_0) + f(x_0) \quad (1.24)$$

$$f_n(x) = \frac{(x - x_0)}{-f'(x_0)} + f(x_0) \quad (1.25)$$

$$\text{gradient}(f(x)) \text{ at } x_0 = f'(x_0)\hat{x} \quad (1.26)$$

$$\text{normal}(f(x)) \text{ at } x_0 = \begin{pmatrix} f'(x_0) \\ -1 \end{pmatrix} \quad (1.27)$$

考虑二元函数 $f(x, y)$ ，对 x 和对 y 的偏导数分别为 $f_x(x, y) = \frac{\partial}{\partial x} f(x, y)$ 和 $f_y(x, y) = \frac{\partial}{\partial y} f(x, y)$ ，则对应的切平面、法线、梯度和法向量为：

$$f_t(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0) \quad (1.28)$$

$$\ell_n(t) = \begin{cases} x = x_0 + at \\ y = y_0 + bt \\ z = f(x_0, y_0) - t \end{cases} \quad (1.29)$$

$$a = f_x(x_0, y_0) \quad \text{and} \quad b = f_y(x_0, y_0)$$

$$\text{gradient}(f(x, y)) \text{ at } (x_0, y_0) = f_x(x_0, y_0)\hat{x} + f_y(x_0, y_0)\hat{y} \quad (1.30)$$

$$\text{normal}(f(x, y)) \text{ at } (x_0, y_0) = \begin{pmatrix} f_x(x, y) \\ f_y(x, y) \\ -1 \end{pmatrix} \quad (1.31)$$