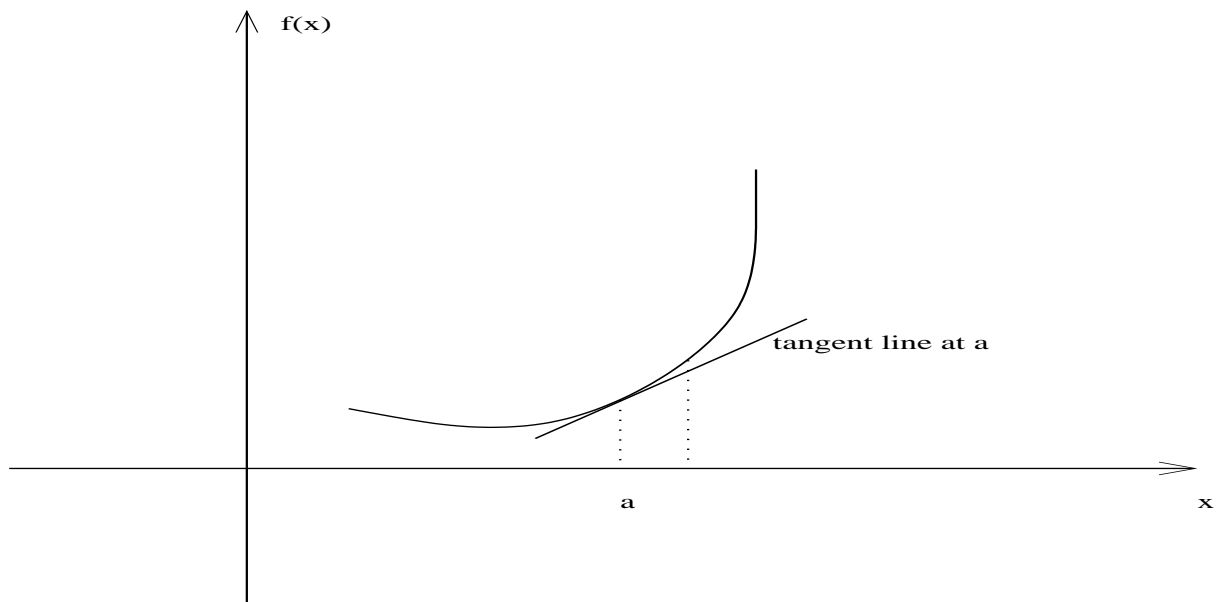


# Taylor Series

*Linear approximation:* Linear approximation is to approximate a general function using a linear function. Given a differentiable scalar function  $f(x)$  of the real variable  $x$ , the linear approximation of the function at point  $a$ , as shown in the Figure below, is obtained by

$$f(x) \approx f(a) + f'(a)(x - a)$$

where  $f'(a) = \left. \frac{df(x)}{dx} \right|_{x=a}$ . The expression on the right-hand side is just the equation for the tangent line to the graph  $f(x)$  at point  $a$ . The above is true when  $x$  is close to  $a$ .



Example: Suppose  $f(x) = x^2 - 5x + 7$  is to be linearised at  $x = 5$ .

$$f(5) = 25 - 25 + 7 = 7$$

$$f'(x) = 2x - 5$$

$$f'(5) = 10 - 5 = 5$$

Thus

$$f(x) \approx f(5) + (x - 5) * f'(5)$$

$$= 7 + (x - 5) * 5$$

$$= 5x - 18$$

Hence can also say that  $g(x) = 5x - 18$  is equation of tangent to curve  $f(x) = x^2 - 5x + 7$  at  $x = 5$ .

Linear approximation of a multivariate function can be obtained similarly with the derivative at a point replaced by the partial derivatives. Given a differentiable scalar multivariate function  $f(\mathbf{x})$ , where  $\mathbf{x} = [x_1, x_2]^T$ , the linear approximation of the function at point  $\mathbf{a} = [a_1, a_2]^T$  is

$$\begin{aligned} f(\mathbf{x}) &\approx f(\mathbf{a}) + \frac{\partial}{\partial x_1} f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}}(x_1 - a_1) \\ &\quad + \frac{\partial}{\partial x_2} f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}}(x_2 - a_2) \\ &= f(\mathbf{a}) + (\mathbf{x} - \mathbf{a})^T \nabla f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}} \end{aligned}$$

where

$$\nabla f(\mathbf{x}) = \begin{bmatrix} \frac{\partial}{\partial x_1} f(\mathbf{x}) \\ \frac{\partial}{\partial x_2} f(\mathbf{x}) \end{bmatrix}$$

Example: Suppose

$$f(\mathbf{x}) = x_1^2 + 2x_2^2 - 5x_1 - 8x_2 + 7$$

is to be linearised at  $\mathbf{x} = [5 \ 4]^T$ .

$$f([5 \ 4]^T) = 25 + 2 \times 16 - 5 \times 5 - 8 \times 4 + 7 = 7$$

$$\frac{\partial}{\partial x_1} f(\mathbf{x}) = 2x_1 - 5$$

$$\frac{\partial}{\partial x_1} f(\mathbf{x})|_{\mathbf{x}=[5 \ 4]^T} = 10 - 5 = 5$$

$$\frac{\partial}{\partial x_2} f(\mathbf{x}) = 4x_2 - 8$$

$$\frac{\partial}{\partial x_2} f(\mathbf{x})|_{\mathbf{x}=[5 \ 4]^T} = 16 - 8 = 8$$

Thus

$$\begin{aligned} f(x) &\approx 7 + (x_1 - 5) * 5 + (x_2 - 4) * 8 \\ &= 5x_1 + 8x_2 - 50 \end{aligned}$$

Hence can also say that  $g(\mathbf{x}) = 5x_1 + 8x_2 - 50$  is equation of tangent to surface  $f(\mathbf{x}) = x_1^2 + 2x_2^2 - 5x_1 - 8x_2 + 7$  at  $x = [5 \ 4]^T$ .

*Taylor series:* Given a differentiable scalar function  $f(x)$  of the real variable  $x$ , its Taylor series reads

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

$$= \sum_{n=0}^{\infty} \frac{1}{n!} f^{(n)}(a)(x-a)^k$$

where  $f''(a) = \frac{d^2 f(x)}{dx^2} \big|_{x=a}$ ,  $f^{(n)}(a) = \frac{d^n f(x)}{dx^n} \big|_{x=a}$ .

- In engineering, the Taylor series of a function at a certain point up to a finite order  $n$  is used to approximate the function.
- If  $n = 1$ , this reduces to the linear approximation.

Example: let  $f(x) = x^2 e^x$ . Show that the Taylor expansion of  $f(x)$  around 0, and up to the 4th order, is

$$f(x) = x^2 + x^3 + \frac{1}{2}x^4 + \dots$$

$$f(0) = 0^2 e^0 = 0$$

$$f'(x) = 2xe^x + x^2e^x = (2x + x^2)e^x$$

$$f'(0) = 0$$

$$f''(x) = (2 + 2x)e^x + (2x + x^2)e^x$$

$$= (2 + 4x + x^2)e^x$$

$$f''(0) = 2$$

$$f^{(3)}(x) = (4 + 2x)e^x + (2 + 4x + x^2)e^x$$

$$= (6 + 6x + x^2)e^x$$

$$f^{(3)}(0) = 6$$

$$f^{(4)}(x) = (6 + 2x)e^x + (6 + 6x + x^2)e^x$$

$$= (12 + 8x + x^2)e^x$$

$$f^{(4)}(0) = 12$$

Thus

$$\begin{aligned} f(x) &= \frac{2}{2!}x^2 + \frac{6}{3!}x^3 + \frac{12}{4!}x^4 + \dots \\ &= x^2 + x^3 + \frac{1}{2}x^4 + \dots \end{aligned}$$

The Taylor series expansion can be generalized to the function of  $m$  variables as in the following.

*Taylor expansion of multivariate functions:*

Given a differentiable scalar multivariate function  $f(\mathbf{x})$ , where  $\mathbf{x} = [x_1, \dots, x_m]^T$ , the Taylor series of the function at point  $\mathbf{a} = [a_1, \dots, a_m]^T$

$$\begin{aligned} f(\mathbf{x}) &= f(\mathbf{a}) + \frac{\partial}{\partial x_1} f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}}(x_1 - a_1) + \dots \\ &\quad + \frac{\partial}{\partial x_m} f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}}(x_m - a_m) \end{aligned}$$

$$\begin{aligned}
& + \frac{1}{2} \frac{\partial^2}{\partial x_1^2} f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}} (x_1 - a_1)^2 + \dots \\
& + \frac{1}{2} \frac{\partial^2}{\partial x_m^2} f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}} (x_m - a_m)^2 + \dots \\
& = f(\mathbf{a}) + (\mathbf{x} - \mathbf{a})^T \nabla f(\mathbf{x})|_{\mathbf{x}=\mathbf{a}} \\
& \quad + \frac{1}{2} (\mathbf{x} - \mathbf{a})^T \mathbf{H} (\mathbf{x} - \mathbf{a}) + \dots
\end{aligned}$$

where

$$\nabla f(\mathbf{x}) = \begin{bmatrix} \frac{\partial}{\partial x_1} f(\mathbf{x}) \\ \frac{\partial}{\partial x_2} f(\mathbf{x}) \end{bmatrix}$$

$$H = \frac{\partial^2 f(\mathbf{x})}{\partial \mathbf{x}^2} |_{\mathbf{x}=\mathbf{a}}$$

Note that the second term is an inner product of a gradient of  $f(\mathbf{x})$  with the vector  $\mathbf{x} - \mathbf{a}$  and the third term is a quadratic form with the Hessian matrix  $\frac{\partial^2 f}{\partial \mathbf{x}^2}$ .



Example: let  $f(\mathbf{x}) = (x_1^2 + 2x_2^2 - 5x_1 - 8x_2 + 7)e^{x_1}$ .  
Show that the Taylor expansion of  $f(\mathbf{x})$  around  $\mathbf{0}$ , and up to the 2nd order, is

$$f(\mathbf{x}) = 7 + 2x_1 - 8x_2 - \frac{1}{2}x_1^2 - 8x_1x_2 + 2x_2^2 + \dots$$

$$f([0 \ 0]^T) = 7$$

$$\frac{\partial}{\partial x_1} f(\mathbf{x}) = (2x_1 - 5)e^{x_1} + (x_1^2 + 2x_2^2 - 5x_1 - 8x_2 + 7)e^{x_1}$$

$$= (x_1^2 + 2x_2^2 - 3x_1 - 8x_2 + 2)e^{x_1}$$

$$\frac{\partial}{\partial x_1} f(\mathbf{x})|_{\mathbf{x}=[0 \ 0]^T} = 2$$

$$\frac{\partial}{\partial x_2} f(\mathbf{x}) = (4x_2 - 8)e^{x_1}$$

$$\frac{\partial}{\partial x_2} f(\mathbf{x})|_{\mathbf{x}=[0 \ 0]^T} = -8$$

$$\frac{\partial^2}{\partial x_1^2} f(\mathbf{x}) = (2x_1 - 3)e^{x_1} + (x_1^2 + 2x_2^2 - 3x_1 - 8x_2 + 2)e^{x_1}$$

$$= (x_1^2 + 2x_2^2 - x_1 - 8x_2 - 1)e^{x_1}$$

$$\frac{\partial^2}{\partial x_1^2} f(\mathbf{x})|_{\mathbf{x}=[0 \ 0]^T} = -1$$

$$\frac{\partial^2}{\partial x_1 \partial x_2} f(\mathbf{x}) = (4x_2 - 8)e^{x_1}$$

$$\frac{\partial^2}{\partial x_1 \partial x_2} f(\mathbf{x})|_{\mathbf{x}=[0 \ 0]^T} = -8$$

$$\frac{\partial^2}{\partial x_2 \partial x_1} f(\mathbf{x}) = (4x_2 - 8)e^{x_1}$$

$$\frac{\partial^2}{\partial x_2 \partial x_1} f(\mathbf{x})|_{\mathbf{x}=[0 \ 0]^T} = -8$$

$$\frac{\partial^2}{\partial x_2^2} f(\mathbf{x}) = 4e^{x_1}$$

$$\frac{\partial^2}{\partial x_2^2} f(\mathbf{x})_{\mathbf{x}=[0 \ 0]^T} = 4$$

Thus

$$\begin{aligned} f(x) &\approx 7 + [x_1 \ x_2] \times [2 \ -8]^T \\ &+ \frac{1}{2} [x_1 \ x_2] \times \begin{pmatrix} -1 & -8 \\ -8 & 4 \end{pmatrix} \times [x_1 \ x_2]^T \end{aligned}$$

Similarly for a scalar function of a matrix variable:

$$f(\mathbf{X}) = f(\mathbf{A}) + \text{trace}\left[\left(\frac{\partial f}{\partial \mathbf{X}}\right)^T (\mathbf{X} - \mathbf{A})\right] + \dots$$

The above formula shows 2 terms of Taylor expansion. It uses the extension of definition of an inner product to matrices

$$\mathbf{A}\mathbf{B} = \sum_{i=1}^m \sum_{j=1}^m a_{ij}b_{ij}$$

but

$$\text{trace}(\mathbf{A}^T \mathbf{B}) = \sum_{i=1}^m (\mathbf{A}^T \mathbf{B})_{ii} = \sum_{i=1}^m \sum_{j=1}^m a_{ij}b_{ij}$$