

Different Quotient \rightarrow Derivative \rightarrow Partial Derivative \rightarrow Chainrule
 Gradient

(\rightarrow Jacobian/Hessian Matrix
 \rightarrow Multivariate Taylor Series

Chapter 5 Vector Calculus

Difference Quotient $\frac{\Delta y}{\Delta x}$

$$\frac{\Delta y}{\Delta x} := \frac{f(x+\Delta x) - f(x)}{\Delta x}$$

Derivative differentiation $\frac{dy}{dx}$

$$\frac{df}{dx} := \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

Partial Derivative $\frac{\partial f}{\partial x_i}$

$$\frac{\partial f}{\partial x_1} = \lim_{h \rightarrow 0} \frac{f(x_1+h, x_2, \dots, x_n) - f(x)}{h}$$

$$\nabla_x f = \text{grad} f = \frac{df}{dx}$$

$$= \left[\frac{\partial f(x)}{\partial x_1} \quad \frac{\partial f(x)}{\partial x_2} \quad \dots \quad \frac{\partial f(x)}{\partial x_n} \right] \in \mathbb{R}^{1 \times n}$$

column vector $\in \mathbb{R}^n$

Chain Rule

$$\frac{df}{dt} = \left[\frac{\partial f}{\partial x_1} \quad \frac{\partial f}{\partial x_2} \right] \begin{bmatrix} \frac{\partial x_1(t)}{\partial t} \\ \frac{\partial x_2(t)}{\partial t} \end{bmatrix}$$

$$= \frac{\partial f}{\partial x_1} \frac{\partial x_1}{\partial t} + \frac{\partial f}{\partial x_2} \frac{\partial x_2}{\partial t}$$

Jacobian Matrix Jacobian 행렬

$$J = \nabla_x f = \frac{df(x)}{dx}$$

$$= \begin{bmatrix} \frac{\partial f_1(x)}{\partial x_1} & \dots & \frac{\partial f_1(x)}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial f_m(x)}{\partial x_1} & \dots & \frac{\partial f_m(x)}{\partial x_n} \end{bmatrix}$$

$$J(i, j) = \frac{\partial f_i}{\partial x_j}$$

Gradient $\frac{df}{dx}$ of a Vector-Valued Function

$$\frac{df}{dx} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1} & \dots & \frac{\partial f_m}{\partial x_n} \end{bmatrix} = \begin{bmatrix} A_{11} & \dots & A_{1n} \\ \vdots & & \vdots \\ A_{m1} & \dots & A_{mn} \end{bmatrix}$$

$$= A \in \mathbb{R}^{m \times n}$$

Higher-Order Derivatives

- Taylor Polynomial

$$T_n(x) := \sum_{k=0}^n \frac{f^{(k)}(x_0)}{k!} (x-x_0)^k$$

- Taylor Series

$$T_\infty(x) = \sum_{k=0}^{\infty} \frac{f^{(k)}(x_0)}{k!} (x-x_0)^k$$

- Multivariate Taylor Series

$$f(x) = \sum_{k=0}^{\infty} \frac{D_x^k f(x_0)}{k!} (x-x_0)^k$$

- Taylor Polynomial

$$T_n(x) = \sum_{k=0}^n \frac{D_x^k f(x_0)}{k!} (x-x_0)^k$$

Hessian Matrix Hessian 행렬

$$H_f = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

$$(H_f)_{ij} = \frac{\partial^2 f}{\partial x_i \partial x_j}$$

$$= \frac{\partial}{\partial x_i} \left(\frac{\partial f}{\partial x_j} \right)$$