

Derivative of a vector w.r.t a scalar:

If $x(t) \in \mathbb{R}^n$ is a vector function of a scalar variable t :

$$x(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix}$$

Its derivative is the component wise derivative:

$$\frac{d}{dt} x(t) = \begin{bmatrix} \frac{dx_1}{dt} \\ \frac{dx_2}{dt} \\ \vdots \\ \frac{dx_n}{dt} \end{bmatrix}$$

A scalar $f(x)$ differentiated w.r.t a vector $x \in \mathbb{R}^n$:

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

Gradient $\nabla_x f \in \mathbb{R}^{n \times 1}$ (column vector)

$$\nabla_x f = \left(\frac{\partial f}{\partial x} \right)^T \quad \nabla_x f : \text{gradient of } f \text{ w.r.t } x$$

$$\nabla_x f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix} \in \mathbb{R}^{n \times 1}$$

Jacobian Matrix

A vector valued function

$$f(x_1, x_2, \dots, x_n) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_m(x_1, x_2, \dots, x_n) \end{bmatrix}$$

• $f(x) \in \mathbb{R}^m, x \in \mathbb{R}^n$

• $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$: f maps n -dimensional input to a m -dimensional

• output.

The Jacobian of f is the $m \times n$ matrix of first order partial derivatives:

• $J(f) = \frac{\partial f}{\partial x} =$

• (Jacobian matrix)

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \dots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}_{m \times n}$$

Chain rule:

$$z(t) = f(x(t), y(t))$$

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

Quadratic form:

$\alpha = x^T A x$, where A is a symmetric matrix ($A = A^T$) and x is $n \times 1$, A is $n \times n$, A does not depend on x , then

$$\frac{\partial \alpha}{\partial x} = 2x^T A$$