

Remark:  $D_{\vec{u}} f(\vec{a})$  can be defined for any vector  $\vec{u}$ , not necessarily  $\|\vec{u}\|=1$  and could be  $\vec{0}$ , by the same definition

$$D_{\vec{u}} f(\vec{a}) = \lim_{t \rightarrow 0} \frac{f(\vec{a} + t\vec{u}) - f(\vec{a})}{t}$$

One can show that

$$D_{\vec{u}} f(\vec{a}) = \begin{cases} \|\vec{u}\| D_{\frac{\vec{u}}{\|\vec{u}\|}} f(\vec{a}), & \text{if } \vec{u} \neq \vec{0} \\ 0, & \text{if } \vec{u} = \vec{0} \end{cases}$$

and that

$$D_{\vec{u}} f(\vec{a}) = \vec{\nabla} f(\vec{a}) \cdot \vec{u} \quad \text{if } f \text{ is differentiable at } \vec{a}$$

(not true in general, if  $f$  is not differentiable)

eg  $f(x,y) = \sqrt{|xy|}$  at  $(0,0)$

### Properties of Gradient

If  $\left\{ \begin{array}{l} \bullet f, g: \Omega \rightarrow \mathbb{R} \quad (\Omega \subset \mathbb{R}^n, \text{open}) \text{ are differentiable,} \\ \bullet c \text{ is a constant,} \end{array} \right.$

then

$$(1) \quad \vec{\nabla}(f \pm g) = \vec{\nabla}f \pm \vec{\nabla}g,$$

$$(2) \quad \vec{\nabla}(cf) = c\vec{\nabla}f$$

$$(3) \quad \vec{\nabla}(fg) = g\vec{\nabla}f + f\vec{\nabla}g$$

$$(4) \quad \vec{\nabla}\left(\frac{f}{g}\right) = \frac{g\vec{\nabla}f - f\vec{\nabla}g}{g^2} \quad \text{provided } g \neq 0$$

(Pf = Easily from properties of partial derivatives)

## Total Differential (of real-valued function)

$f: \Omega \rightarrow \mathbb{R}$  ( $\Omega \subseteq \mathbb{R}^n$ , open) differentiable at  $\vec{a} \in \Omega$ .

Then linearization at  $\vec{a}$ :

$$f(\vec{x}) = f(\vec{a}) + \sum_{\vec{i}=1}^n \frac{\partial f}{\partial x_i}(\vec{a})(x_i - a_i) + \varepsilon(\vec{x})$$

Usually denote:  $\Delta f = f(\vec{x}) - f(\vec{a})$

$$\Delta x_i = x_i - a_i$$

Then  $\Delta f \approx \sum_{\vec{i}=1}^n \frac{\partial f}{\partial x_i}(\vec{a}) \Delta x_i$  (provided  $\lim_{\vec{x} \rightarrow \vec{a}} \frac{|\varepsilon(\vec{x})|}{\|\vec{x} - \vec{a}\|} = 0$ )

Classically, this approximation is presented as

$$df = \sum_{\vec{i}=1}^n \frac{\partial f}{\partial x_i}(\vec{a}) dx_i \quad \left( \text{thinking: } \begin{array}{l} \Delta f \rightarrow df \\ \Delta x_i \rightarrow dx_i \end{array} \right)$$

Def: Let  $f: \Omega \rightarrow \mathbb{R}$ , ( $\Omega \subseteq \mathbb{R}^n$ , open)  
•  $\vec{a} \in \Omega$

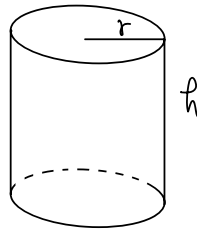
Suppose that  $f$  is differentiable on  $\Omega$ . Then the total differential of  $f$  at  $\vec{a}$  is defined to be the (formal) expression:

$$df = \sum_{\vec{i}=1}^n \frac{\partial f}{\partial x_i}(\vec{a}) dx_i$$

Remark: In the future,  $df$  and  $dx_i$  can be interpreted as a linear maps from  $\mathbb{R}^n$  to  $\mathbb{R}$ .

eg: Let  $V(r, h) = \pi r^2 h$   
(Volume of the Cylinder)

$V$  is  $C^1 \Rightarrow$  differentiable



$$\begin{aligned} dV &= \frac{\partial V}{\partial r} dr + \frac{\partial V}{\partial h} dh \\ &= 2\pi r h dr + \pi r^2 dh \end{aligned}$$

For application:

Suppose we want to approximate the change of  $V$

when  $(r, h)$  changes  $(r, h) = (3, 12)$  to

$(3+0.08, 12-0.3)$

Then let  $dr = \Delta r = 0.08$

$dh = \Delta h = -0.3$

we have

$$\Delta V \approx dV = 2\pi r h dr + \pi r^2 dh \quad (\text{at } r=3, h=12 \\ \text{dr, dh as above})$$

$$= 2\pi \cdot 3 \cdot 12 \cdot 0.08 + \pi (3)^2 (-0.3)$$

$$= 3.06 \pi \approx 9.61$$

## Properties of Total Differential

If  $f, g: \Omega \rightarrow \mathbb{R}$  ( $\Omega \subseteq \mathbb{R}^n$ , open) are differentiable, and  
•  $c \in \mathbb{R}$  is a constant.

Then

$$(1) \quad d(f \pm g) = df \pm dg,$$

$$(2) \quad d(cf) = cdf$$

$$(3) \quad d(fg) = gdf + fdg$$

$$(4) \quad d\left(\frac{f}{g}\right) = \frac{gdf - f dg}{g^2} \quad \text{provided } g \neq 0$$

(Pf = Easily from properties of partial derivatives)

## Summary (on differentiation of a real-valued function on $\mathbb{R}^n$ )

$$f: \mathbb{R}^n \rightarrow \mathbb{R}$$

### A. Types of differentiations (derivatives)

- Directional Derivative :

$$D_{\vec{u}} f(\vec{a}) = \lim_{t \rightarrow 0} \frac{f(\vec{a} + t\vec{u}) - f(\vec{a})}{t} \quad (\|\vec{u}\| = 1)$$

- Partial derivatives :

$$\frac{\partial f}{\partial x_i}(\vec{a}) = D_{\vec{e}_i} f(\vec{a}), \quad \vec{e}_i = (0, \dots, 1, \dots, 0)$$

↑  $i$ th component.

- Gradient

$$\vec{\nabla} f(\vec{a}) = \left( \frac{\partial f}{\partial x_1}(\vec{a}), \dots, \frac{\partial f}{\partial x_n}(\vec{a}) \right)$$

- Total Differential

$$df = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\vec{a}) dx_i$$

- Higher Derivatives

$$\frac{\partial^{k_1 + \dots + k_n} f}{\partial x_1^{k_1} \dots \partial x_n^{k_n}}(\vec{a})$$

(provided  $f$  is  $C^k$ ,  $k = k_1 + \dots + k_n$ )

$\uparrow$  all partial derivatives up to order  $k$  exist and continuous.

## B. Linear approximation

- $L(\vec{x}) = f(\vec{a}) + \vec{\nabla} f(\vec{a}) \cdot (\vec{x} - \vec{a})$

- $f(\vec{x}) = L(\vec{x}) + \varepsilon(\vec{x})$   
 $\quad \quad \quad \longleftarrow$  error term

- $f$  is differentiable at  $\vec{a}$

$$\Leftrightarrow \lim_{\vec{x} \rightarrow \vec{a}} \frac{|\varepsilon(\vec{x})|}{\|\vec{x} - \vec{a}\|} = 0$$

In this case,  $df \approx \Delta f$  (by identifying  $dx_i = \Delta x_i$ )

## C. Relations among various concepts

•  $C^\infty \Rightarrow \dots \Rightarrow C^{k+1} \Rightarrow C^k \Rightarrow \dots \Rightarrow C^1 \Rightarrow C^0$  (no reverse implication)

•  $f$  is  $C^1$  on an open set containing  $\vec{a}$

$\Downarrow$   ~~$\Leftrightarrow$~~

$f$  is differentiable at  $\vec{a}$

$\swarrow$   ~~$\Leftrightarrow$~~

$D_{\vec{a}} f(\vec{a})$  exists

~~$\Rightarrow$~~

$\swarrow$   ~~$\Leftrightarrow$~~

$f$  is continuous

$\forall \vec{u} \in \mathbb{R}^n, \|\vec{u}\|=1$

~~$\Leftarrow$~~

at  $\vec{a}$

$\swarrow$   ~~$\Leftrightarrow$~~

~~$\Leftrightarrow$~~

$\frac{\partial f}{\partial x_i}(\vec{a})$  exists,  $\forall i=1, \dots, n$

## Counter examples:

eg 1:  $f: \mathbb{R} \rightarrow \mathbb{R}$  (in MATH2050)

$$f(x) = \begin{cases} x^2 \sin \frac{1}{x} & \text{if } x \neq 0 \\ 0 & \text{if } x = 0 \end{cases}$$

$f$  is differentiable on  $\mathbb{R}$  but

$f'(x)$  is not continuous at  $x=0$

Similarly  $g(x) = x^{2k-2} f(x)$  is  $k$ -time differentiable

but  $g^{(k)}(x)$  is not continuous at  $x=0$ .

Hence  $k$ -time differentiable  ~~$\Rightarrow$~~   $C^k$

In particular,  $C^{k-1} \not\Rightarrow C^k$   
(For multivariable:  $f(\vec{x}) = f(x_1, \dots, x_n) = g(x_1)$ .)

$$\text{eg 2} \quad f(x,y) = \begin{cases} \frac{xy^2}{x^2+y^2} & \text{if } x^2+y^2 \neq 0 \\ 0 & \text{if } x^2+y^2 = 0 \end{cases}$$

$D_{\vec{u}}f(0,0)$  exists,  $\forall$  unit vector  $\vec{u} \in \mathbb{R}^2$   
but  $f$  is not continuous at  $(0,0)$  (check!)

eg 3:  $f(x,y) = |x+y|$  is continuous on  $\mathbb{R}^2$  but  
 $f_x(0,0), f_y(0,0)$  DNE. (Check!)

eg 4:  $f(x,y) = \sqrt{|xy|}$   
 $f_x(0,0), f_y(0,0)$  exist (in fact = 0)  
but  $D_{\vec{u}}f(0,0)$  DNE for  $\vec{u} \neq \pm \vec{e}_1, \pm \vec{e}_2$ .  
(Check!)

## Review: Matrix Multiplication

$$\text{Let } A = \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \text{ be an } m \times n\text{-matrix}$$

$$= \begin{bmatrix} -\vec{a}_1- \\ \vdots \\ -\vec{a}_m- \end{bmatrix} \text{ where } \vec{a}_i = (a_{i1}, \dots, a_{in}) \in \mathbb{R}^n$$

If

$$b = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} = \begin{bmatrix} | \\ \vec{b} \\ | \end{bmatrix} \text{ be a } n \times 1\text{-matrix regarded as a column vector in } \mathbb{R}^n,$$

then (matrix multiplication)

$$\begin{aligned} Ab &= \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} = \begin{bmatrix} -\vec{a}_1- \\ \vdots \\ -\vec{a}_m- \end{bmatrix} \begin{bmatrix} | \\ \vec{b} \\ | \end{bmatrix} \\ &= \begin{bmatrix} a_{11}b_1 + \cdots + a_{1n}b_n \\ \vdots \\ a_{m1}b_1 + \cdots + a_{mn}b_n \end{bmatrix} = \begin{bmatrix} \vec{a}_1 \cdot \vec{b} \\ \vdots \\ \vec{a}_m \cdot \vec{b} \end{bmatrix} \quad \begin{array}{l} \text{(result} \\ = m \times 1\text{-matrix} \\ = \text{column } m\text{-vector)} \end{array} \end{aligned}$$

Similarly, for multiplication of  $(1 \times n)$  &  $(n \times k)$  matrices

$$\begin{aligned} & \begin{bmatrix} -\vec{a}- \end{bmatrix} \begin{bmatrix} | \\ \vec{b}_1 & \cdots & \vec{b}_k \\ | \end{bmatrix} \quad \begin{array}{l} (\vec{a}, \vec{b}_1, \dots, \vec{b}_k \in \mathbb{R}^n) \\ \uparrow \\ \text{row} \\ \text{vector} \end{array} \quad \begin{array}{l} \underbrace{\vec{b}_1, \dots, \vec{b}_k}_{\text{column} \\ \text{vectors}} \end{array} \\ &= [\vec{a} \cdot \vec{b}_1, \dots, \vec{a} \cdot \vec{b}_k] \\ & \text{(result} = 1 \times k\text{-matrix} = \text{row } k\text{-vector)} \end{aligned}$$



In general:  $(m \times n)$  times  $(n \times k)$

$$AB = \begin{bmatrix} -\vec{a}_1- \\ \vdots \\ -\vec{a}_m- \end{bmatrix} \begin{bmatrix} | & & | \\ \vec{b}_1 & \dots & \vec{b}_k \\ | & & | \end{bmatrix} \quad \left( \underbrace{\vec{a}_1, \dots, \vec{a}_m}_{\text{row vectors}}, \underbrace{\vec{b}_1, \dots, \vec{b}_k}_{\text{column vectors}} \in \mathbb{R}^n \right)$$

$$= \begin{bmatrix} \vec{a}_1 \cdot \vec{b}_1 & \dots & \vec{a}_1 \cdot \vec{b}_k \\ \vdots & & \vdots \\ \vec{a}_m \cdot \vec{b}_1 & \dots & \vec{a}_m \cdot \vec{b}_k \end{bmatrix}$$

$$= \begin{bmatrix} | & & | \\ A\vec{b}_1 & \dots & A\vec{b}_k \\ | & & | \end{bmatrix} \quad \left( = A \begin{bmatrix} | & & | \\ \vec{b}_1 & \dots & \vec{b}_k \\ | & & | \end{bmatrix} \right)$$

$$= \begin{bmatrix} -\vec{a}_1 B- \\ \vdots \\ -\vec{a}_m B- \end{bmatrix} \quad \left( = \begin{bmatrix} -\vec{a}_1- \\ \vdots \\ -\vec{a}_m- \end{bmatrix} B \right)$$

eg:

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 6 & 7 \\ 8 & 9 & 10 \end{bmatrix} = \begin{bmatrix} 21 & 24 & 27 \\ 47 & 54 & 61 \end{bmatrix} \quad (\text{check!})$$

A      B

$$A \begin{bmatrix} 5 \\ 8 \end{bmatrix} = \begin{bmatrix} 21 \\ 47 \end{bmatrix}, \quad A \begin{bmatrix} 6 \\ 9 \end{bmatrix} = \begin{bmatrix} 24 \\ 54 \end{bmatrix}, \quad A \begin{bmatrix} 7 \\ 10 \end{bmatrix} = \begin{bmatrix} 27 \\ 61 \end{bmatrix}$$

$$[1, 2] B = [21, 24, 27]$$

$$[3, 4] B = [47, 54, 61]$$

## Differentiability of Vector-Valued Functions

$$\vec{f}: \Omega \rightarrow \mathbb{R}^m, \quad (\Omega \subset \mathbb{R}^n, \text{open})$$

$$\vec{f}(\vec{x}) = \begin{bmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{bmatrix}$$

Suppose  $\frac{\partial f_i}{\partial x_j}(\vec{a})$  exists for each  $i=1, \dots, m$  &  $j=1, \dots, n$ .

$$f_i(\vec{x}) = f_i(\vec{a}) + \vec{\nabla} f_i(\vec{a}) \cdot (\vec{x} - \vec{a}) + \varepsilon_i(\vec{x}) \quad \text{--- } (*)_i$$

$\left( \begin{array}{ccccccc} (1 \times 1) & (1 \times 1) & (1 \times n) & \cdot & (n \times 1) & (1 \times 1) & \text{matrix} \\ & & \uparrow & & \uparrow & & \\ & & \text{row} & & \text{column} & & \end{array} \right)$

Put all  $(*)_i$ , we have

$$\begin{bmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{bmatrix} = \begin{bmatrix} f_1(\vec{a}) \\ \vdots \\ f_m(\vec{a}) \end{bmatrix} + \underbrace{\begin{bmatrix} -\vec{\nabla} f_1(\vec{a})- \\ \vdots \\ -\vec{\nabla} f_m(\vec{a})- \end{bmatrix}}_{\substack{m \times n \text{ matrix} \\ \text{of } \left[ \frac{\partial f_i}{\partial x_j} \right]_{\substack{i=1, \dots, m \\ j=1, \dots, n}}}} \begin{bmatrix} x_1 - a_1 \\ \vdots \\ x_n - a_n \end{bmatrix} + \underbrace{\begin{bmatrix} \varepsilon_1(\vec{x}) \\ \vdots \\ \varepsilon_m(\vec{x}) \end{bmatrix}}_{\text{Errors}}$$

$\vec{L}(\vec{x})$

In the following definitions,

- $\vec{f}: \Omega \rightarrow \mathbb{R}^m$  ( $\Omega \subset \mathbb{R}^n$ , open)
- $\vec{f}(\vec{x}) = \begin{bmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{bmatrix}$  (in component form)
- $\vec{a} = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \in \Omega$
- $\vec{x} - \vec{a} = \begin{bmatrix} x_1 - a_1 \\ \vdots \\ x_n - a_n \end{bmatrix}$

Def Jacobian Matrix of  $\vec{f}$  at  $\vec{a}$  is defined to be

$$D\vec{f}(\vec{a}) = \begin{bmatrix} -\vec{\nabla} f_1(\vec{a}) - \\ \vdots \\ -\vec{\nabla} f_m(\vec{a}) - \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(\vec{a}) & \dots & \frac{\partial f_1}{\partial x_n}(\vec{a}) \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1}(\vec{a}) & \dots & \frac{\partial f_m}{\partial x_n}(\vec{a}) \end{bmatrix}$$

(a  $m \times n$ -matrix)

Def Linearization of  $\vec{f}$  at  $\vec{a}$  is defined to be

$$\vec{L}(\vec{x}) = \vec{f}(\vec{a}) + D\vec{f}(\vec{a})(\vec{x} - \vec{a})$$

↑ matrix multiplication

Def:  $\vec{f}$  is said to be differentiable at  $\vec{a} \in \Omega$ ,

- if
- $\frac{\partial f_i}{\partial x_j}(\vec{a})$  exists  $\forall i=1, \dots, m$  &  $j=1, \dots, n$
  - Error term of the linear approximation

$$\vec{E}(\vec{x}) = \vec{f}(\vec{x}) - \vec{L}(\vec{x})$$

satisfies

$$\lim_{\vec{x} \rightarrow \vec{a}} \frac{\|\vec{E}(\vec{x})\|}{\|\vec{x} - \vec{a}\|} = 0$$

Remarks (1)  $[D\vec{f}(\vec{a})]_{ij}$  (ij-entry of  $D\vec{f}(\vec{a})$ )

$$= \frac{\partial f_i}{\partial x_j}(\vec{a})$$

$$(2) \quad \vec{f}(\vec{x}) = \vec{f}(\vec{a}) + D\vec{f}(\vec{a})(\vec{x} - \vec{a}) + \vec{E}(\vec{x})$$

$\begin{array}{cccccc} \uparrow & \uparrow & \uparrow & \uparrow & \uparrow & \\ \text{column} & \text{column} & \text{m} \times \text{n} & \text{column} & \text{column} & \\ \text{m-vector} & \text{n-vector} & \text{matrix} & \text{n-vector} & \text{m-vector} & \\ \uparrow & \uparrow & \underbrace{\phantom{\uparrow}} & \uparrow & \uparrow & \\ \text{m} \times 1 & \text{m} \times 1 & (\text{m} \times \text{n}) \cdot (\text{n} \times 1) & \text{m} \times 1 & \text{m} \times 1 & (\text{matrix}) \end{array}$

(3) If  $f$  is real-valued ( $m=1$ ), then

$$Df(\vec{a}) = \vec{\nabla} f(\vec{a}) \quad ((1 \times n)\text{-matrix})$$

(4)  $\|\vec{E}(\vec{x})\|$  &  $\|\vec{x} - \vec{a}\|$  are length in  $\mathbb{R}^m$  &  $\mathbb{R}^n$  respectively.

$$(5) \lim_{\vec{x} \rightarrow \vec{a}} \frac{\|\vec{\epsilon}(\vec{x})\|}{\|\vec{x} - \vec{a}\|} = 0 \iff \lim_{\vec{x} \rightarrow \vec{a}} \frac{|\epsilon_i(\vec{x})|}{\|\vec{x} - \vec{a}\|} = 0$$

Hence

$\vec{f}$  is differentiable at  $\vec{a} \iff f_i$  is differentiable at  $\vec{a}, \forall i=1, \dots, m$

Approximation:

$$\vec{f}(\vec{x}) \approx \vec{L}(\vec{x}) = \vec{f}(\vec{a}) + D\vec{f}(\vec{a})(\vec{x} - \vec{a})$$

$$\Rightarrow \underbrace{\vec{f}(\vec{x}) - \vec{f}(\vec{a})}_{\Delta \vec{f} = \text{change of } \vec{f}} \approx \underbrace{D\vec{f}(\vec{a})}_{\substack{\uparrow \\ \text{Jacobian} \\ \text{matrix}}} \underbrace{(\vec{x} - \vec{a})}_{\Delta \vec{x} = \text{change in } \vec{x}}$$

Notation:  $d\vec{f} = D\vec{f}(\vec{a})(\vec{x} - \vec{a})$  approximated change of  $f$   
 (total differential)  
 i.e.  $\Delta \vec{f} \approx d\vec{f}$

eg:  $\vec{f}(x,y) = ((y+1)\ln x, x^2 - \sin y + 1)$

$$= \begin{pmatrix} (y+1)\ln x \\ x^2 - \sin y + 1 \end{pmatrix} = \begin{pmatrix} f_1(x,y) \\ f_2(x,y) \end{pmatrix} \quad \left( \begin{array}{l} \text{Rewrite as} \\ \text{column vector} \end{array} \right)$$

(1) Find  $D\vec{f}(1,0)$

(2) Approximate  $\vec{f}(0.9, 0.1)$

Solu:  $D\vec{f}(x,y) = \begin{bmatrix} -\vec{\nabla}f_1 \\ -\vec{\nabla}f_2 \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} \end{bmatrix}$

$$= \begin{bmatrix} \frac{y+1}{x} & \ln x \\ 2x & -\cos y \end{bmatrix}$$

$$\therefore D\vec{f}(1,0) = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix}$$

$$(2) \vec{L}(x,y) = \vec{f}(1,0) + D\vec{f}(1,0) \cdot \begin{bmatrix} x-1 \\ y-0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 2 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} x-1 \\ y \end{bmatrix}$$

$$\vec{f}(0.9, 0.1) \simeq \vec{L}(0.9, 0.1)$$

$$= \begin{bmatrix} 0 \\ 2 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 0.9-1 \\ 0.1 \end{bmatrix}$$

$$= \begin{bmatrix} -0.1 \\ 1.7 \end{bmatrix} \quad \underbrace{\begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 0.9-1 \\ 0.1 \end{bmatrix}}_{d\vec{f}} \quad \leftarrow \Delta\vec{x} = d\vec{x}$$