

## LECTURE 16: THE GENERAL DERIVATIVE (pg. 180 - 192 (in 2020 notes)

**Defn:** For map  $\vec{F} = (F_1, F_2, \dots, F_m) : U \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$   
we say  $\vec{F}$  is differentiable at  $\vec{P} \in \mathbb{R}^n$  if  $\exists$  linear mapping  $\vec{L} : \mathbb{R}^n \rightarrow \mathbb{R}^m$  such that

$$\lim_{\vec{h} \rightarrow 0} \left[ \frac{\vec{F}(\vec{P} + \vec{h}) - \vec{F}(\vec{P}) - \vec{L}(\vec{h})}{\|\vec{h}\|} \right] = 0$$

We denote  $\vec{L}(\vec{h}) = d\vec{F}_{\vec{P}}(\vec{h})$  and say  $d\vec{F}_{\vec{P}}$  is the differential of  $\vec{F}$  at  $\vec{P}$ .

Essentially,  $\vec{F}(\vec{x}) \approx \vec{F}(\vec{P}) + d\vec{F}_{\vec{P}}(\vec{x} - \vec{P})$  gives linearization of  $\vec{F}$ . Since  $d\vec{F}_{\vec{P}}$  is linear transformation it has a standard matrix which is called the Jacobian Matrix. Omitting the point  $\vec{P}$ ,

$$[d\vec{F}] = J_{\vec{F}} = \begin{bmatrix} \frac{\partial \vec{F}}{\partial x_1} & \frac{\partial \vec{F}}{\partial x_2} & \cdots & \frac{\partial \vec{F}}{\partial x_n} \end{bmatrix} = \begin{bmatrix} (\nabla F_1)^T \\ (\nabla F_2)^T \\ \vdots \\ (\nabla F_m)^T \end{bmatrix}$$

$n = \#$  of domain variables

$m = \#$  of range variables

- Can view  $J_{\vec{F}}$  either in terms of columns  $\frac{\partial \vec{F}}{\partial x_i}$  or rows  $(\nabla F_i)^T$

$\overbrace{\begin{array}{l} n=1 \\ m>1 \end{array}}^{} \rightarrow \vec{F}(t) \text{ has } \frac{d\vec{F}}{dt} \text{ as } \dots J_{\vec{F}} \quad (\text{identity } x_i = t)$   
 $\vec{F} = f$

$\overbrace{\begin{array}{l} m=1 \\ n>1 \end{array}}^{} \rightarrow f(x_1, \dots, x_n) \text{ has } (\nabla f)^T = J_{\vec{F}} \quad (\text{identity } F_i = f \text{ just one component here})$