

Partial Differentiation

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

$$y = x^2 + t^2 \leftarrow \text{Variable}$$

↑
variable

Denoted by $\frac{\partial f}{\partial x}$ \nearrow function
 \searrow variable

$$\frac{\partial y}{\partial x} = 2x \quad \bigg| \quad \frac{\partial y}{\partial t} = 2t$$

Homogeneous functions:-

Any function $f(x, y)$ which can be expressed in the form $x^n \phi(y/x)$ is called homogeneous f^n of degree n

Euler's Theorem:-

If $u(x, y)$ is a homogeneous f^n of degree n

in x & y then

$$x \frac{\partial u}{\partial x} + y \frac{\partial u}{\partial y} = nu$$

Total differentiation :-

$$\frac{du}{dt} = \frac{\partial u}{\partial x} \cdot \frac{dx}{dt} + \frac{\partial u}{\partial y} \cdot \frac{dy}{dt} + \frac{\partial u}{\partial z} \cdot \frac{dz}{dt}$$

Differentiation of implicit f^n

$$0 = \frac{df}{dx} = \frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} \cdot \frac{dy}{dx}$$

$$\text{where } \Rightarrow \frac{dy}{dx} = -\frac{\partial f / \partial x}{\partial f / \partial y}$$

Jacobian:-

If u, v are f^n of the two independent variables

x, y then determinant
$$\begin{vmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \end{vmatrix}$$

is called Jacobian denoted by $\frac{\partial(u,v)}{\partial(x,y)} = J\left(\frac{u,v}{x,y}\right)$

If u, v, w are f^n of the three independent

variables x, y, z then determinant

$$\begin{vmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} & \frac{\partial u}{\partial z} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} & \frac{\partial v}{\partial z} \\ \frac{\partial w}{\partial x} & \frac{\partial w}{\partial y} & \frac{\partial w}{\partial z} \end{vmatrix} \quad \begin{array}{l} \text{is called Jacobian} \\ \text{denoted by} \\ \frac{\partial(u,v,w)}{\partial(x,y,z)} = J\left(\frac{u,v,w}{x,y,z}\right) \end{array}$$

Properties of Jacobian:-

$$\textcircled{1} \frac{\partial(u,v)}{\partial(x,y)} \times \frac{\partial(x,y)}{\partial(u,v)} = JJ^{-1} = 1$$

\nearrow u, v are functions of x, y

$$\frac{\partial(u,v,w)}{\partial(x,y,z)} \times \frac{\partial(x,y,z)}{\partial(u,v,w)} = JJ^{-1} = 1$$

\nearrow u, v are functions of x, y, z

$\textcircled{2}$ If u, v are f^n of x, y where x, y are f^n of x, y then

$$\frac{\partial(u,v)}{\partial(x,y)} = \frac{\partial(u,v)}{\partial(x,y)} \times \frac{\partial(x,y)}{\partial(x,y)} \quad \left. \vphantom{\frac{\partial(u,v)}{\partial(x,y)}} \right\} \text{Similar to chain rule}$$

$\textcircled{3}$ If f^n u, v, w of three independent variables x, y, z are not independent then

$$\frac{\partial(u,v,w)}{\partial(x,y,z)} = 0$$

i.e., if $J\left(\frac{u,v,w}{x,y,z}\right) = 0 \rightarrow \text{Dependent}$

$$J\left(\frac{u,v,w}{x,y,z}\right) \neq 0 \rightarrow \text{Independent}$$

Jacobian of implicit functions:-

If variables x, y, u, v are connected by implicit f^n

$$f_1(x, y, u, v) = 0 \quad \& \quad f_2(x, y, u, v) = 0$$

then

$$\frac{\partial(u, v)}{\partial(x, y)} = (-1)^n \frac{\frac{\partial(f_1, f_2)}{\partial(x, y)}}{\frac{\partial(f_1, f_2)}{\partial(u, v)}}$$

If it is x, y connected with u, v then $n=2$
& if x, y, z are connected with u, v, w then $n=3$

Vector Diff:-

The ∇ operator

$$\nabla = \hat{i} \frac{\partial}{\partial x} + \hat{j} \frac{\partial}{\partial y} + \hat{k} \frac{\partial}{\partial z}$$

If f is a f^n there are 2 cases

- i) f is scalar
- ii) f is vector

If f is a scalar:-

$$\nabla f = \left(\hat{i} \frac{\partial}{\partial x} + \hat{j} \frac{\partial}{\partial y} + \hat{k} \frac{\partial}{\partial z} \right) f$$

$$\nabla f = \frac{\partial f}{\partial x} \hat{i} + \frac{\partial f}{\partial y} \hat{j} + \frac{\partial f}{\partial z} \hat{k}$$

Gradient of f
(grad f)

If f is a vector

$$\vec{f} = f_1 \hat{i} + f_2 \hat{j} + f_3 \hat{k}$$

$$\nabla \cdot \vec{f}$$

dot product

cross product

Divergence of f
(div f)

$$\nabla \times \vec{f}$$

Curl of f
(curl f)

Normal Vector & unit Normal Vector:-

$$\vec{N} = \text{grad } f \rightarrow \text{Normal vector}$$

$$\hat{N} = \frac{\text{grad } f}{|\text{grad } f|} = \frac{\vec{N}}{|\vec{N}|} \rightarrow \text{Unit normal vector}$$

Directional Derivative:-

$$\text{Directional derivative in } \vec{A} \text{ direction} = (\text{grad } f) \cdot \hat{A}$$

Unit vector of \vec{A}

Angle b/w two scalar f :- \rightarrow at point (a, b, c)

$$\cos \theta = \frac{(\text{grad } f)_{(a,b,c)} \cdot (\text{grad } g)_{(a,b,c)}}{|\text{grad } f| |\text{grad } g|}$$

Substituting the points in both the gradients

Tangent Vector:-

$$\text{Suppose } \vec{r} = f(t)\hat{i} + g(t)\hat{j} + h(t)\hat{k}$$

$$\text{then tangent vector } \vec{T} = \frac{d\vec{r}}{dt}$$