

MATH 273 - Lecture 7 (09/16/2014)

Recall: directional derivative of $f(x, y)$ at $P_0(x_0, y_0)$ in the direction $\hat{u} = u_1 \hat{i} + u_2 \hat{j}$ (\hat{u} is a unit vector, i.e., $\|\hat{u}\| = 1$) is

$$(D_{\hat{u}} f)_{P_0} = \lim_{s \rightarrow 0} \frac{f(P_0 + s\hat{u}) - f(P_0)}{s}$$

length

With $\bar{P}_0 = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix} = x_0 \hat{i} + y_0 \hat{j}$, we have $\bar{P}_0 + s\hat{u} = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix} + s \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$

$$\begin{aligned} &= (x_0 \hat{i} + y_0 \hat{j}) + s(u_1 \hat{i} + u_2 \hat{j}) \\ &= (x_0 + su_1) \hat{i} + (y_0 + su_2) \hat{j} \end{aligned}$$

$$(D_{\hat{u}} f)_{P_0} = \lim_{s \rightarrow 0} \frac{f(x_0 + su_1, y_0 + su_2) - f(x_0, y_0)}{s}$$

$x = x_0 + su_1 = x(s)$
 $y = y_0 + su_2 = y(s)$

When $\hat{u} = \hat{i}$ (i.e., the x -direction), $\hat{u} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \hat{i} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} \hat{j}$, then

$$(D_{\hat{i}} f)_{P_0} = \lim_{s \rightarrow 0} \frac{f(x_0 + s, y_0) - f(x_0, y_0)}{s} = \frac{\partial f}{\partial x}$$

Similarly, when $\hat{u} = \hat{j}$ (i.e., the y -direction), $\hat{u} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \hat{i} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} \hat{j}$, then

$$(D_{\hat{j}} f)_{P_0} = \lim_{s \rightarrow 0} \frac{f(x_0, y_0 + s) - f(x_0, y_0)}{s} = \frac{\partial f}{\partial y}.$$

How to compute $(Df)_{\hat{u}}|_{P_0}$?

We have $f(x, y)$ with x, y being functions of s . We apply chain rule to compute $f(x, y)$

$$\frac{df}{ds} \Big|_{\hat{u}, P_0} = \left(\frac{\partial f}{\partial x} \right)_{P_0} \underbrace{\frac{dx}{ds}}_{u_1} + \left(\frac{\partial f}{\partial y} \right)_{P_0} \underbrace{\frac{dy}{ds}}_{u_2}$$

$$= \left(\frac{\partial f}{\partial x} \right)_{P_0} u_1 + \left(\frac{\partial f}{\partial y} \right)_{P_0} u_2$$

$$= \underbrace{\left[\left(\frac{\partial f}{\partial x} \right)_{P_0} \hat{i} + \left(\frac{\partial f}{\partial y} \right)_{P_0} \hat{j} \right]}_{\text{gradient of } f \text{ at } P_0} \cdot \underbrace{\left[u_1 \hat{i} + u_2 \hat{j} \right]}_{\hat{u}} = \begin{bmatrix} \left(\frac{\partial f}{\partial x} \right)_{P_0} \\ \left(\frac{\partial f}{\partial y} \right)_{P_0} \end{bmatrix} \cdot \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

$$x = x_0 + s u_1$$

$$y = y_0 + s u_2$$

$$\frac{dx}{ds} = u_1, \quad \frac{dy}{ds} = u_2$$

$$\hat{u} = u_1 \hat{i} + u_2 \hat{j} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

Def The gradient vector (or, just gradient) of $f(x, y)$ at point $P_0(x_0, y_0)$ is the vector

$$\nabla f = \left(\frac{\partial f}{\partial x} \right)_{P_0} \hat{i} + \left(\frac{\partial f}{\partial y} \right)_{P_0} \hat{j}, \quad \text{where the partial derivatives are evaluated at } P_0(x_0, y_0).$$

"gradient of f "
"delf", "grad f"

Hence the directional derivative of f along \hat{u} (unit vector)

at $P_0(x_0, y_0)$ is

$$(D_{\hat{u}} f)_{P_0} = (\nabla f)_{P_0} \cdot \hat{u} \quad \text{Scalar product}$$

Prob 2 $f(x, y) = \ln(x^2 + y^2)$, $P_0(1, 1)$. Find gradient of f at P_0 , Sketch the gradient and level curve passing through P_0 .

$$\begin{aligned} (\nabla f)_{P_0} &= \left(\frac{\partial f}{\partial x}\right)_{P_0} \hat{i} + \left(\frac{\partial f}{\partial y}\right)_{P_0} \hat{j} = \left(\frac{2x+0}{x^2+y^2}\right)_{P_0} \hat{i} + \left(\frac{0+2y}{x^2+y^2}\right)_{P_0} \hat{j} \\ &= \frac{2}{2} \hat{i} + \frac{2}{2} \hat{j} = \hat{i} + \hat{j}. \end{aligned}$$

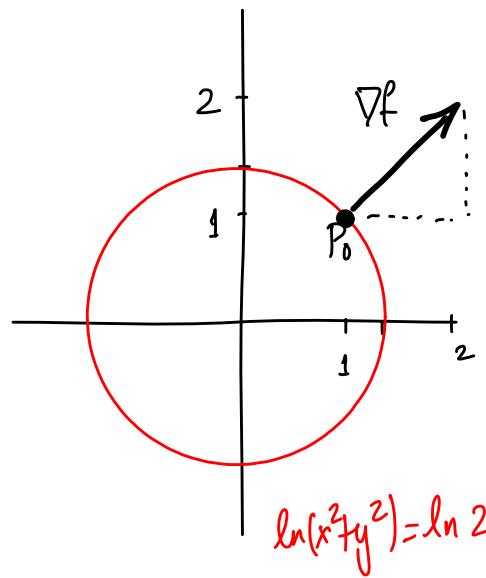
$$f(x, y) = \ln(x^2 + y^2)$$

$$\text{at } P_0(1, 1), f(1, 1) = \ln(1^2 + 1^2) = \ln 2.$$

The level curve is $f(x, y) = \ln 2$

$$\ln(x^2 + y^2) = \ln 2$$

$$\text{i.e., } x^2 + y^2 = 2$$



Definition of ∇f naturally extends to 3 (or higher) dimensions: for $f(x, y, z)$,

$$\nabla f \Big|_{P_0} = \left(\frac{\partial f}{\partial x} \right)_{P_0}^i \hat{i} + \left(\frac{\partial f}{\partial y} \right)_{P_0}^j \hat{j} + \left(\frac{\partial f}{\partial z} \right)_{P_0}^k \hat{k}$$

Prob 7 $f(x, y, z) = x^2 + y^2 - 2z^2 + z \ln x$, $P_0(1, 1, 1)$, find $(\nabla f)_{P_0}$.

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

$$= \left(2x + \frac{3}{x} \right) \hat{i} + (2y) \hat{j} + (-4z + \ln x) \hat{k} \quad \text{plug in } (1, 1, 1)$$

$$= \left(2 \cdot 1 + \frac{1}{1} \right) \hat{i} + (2 \cdot 1) \hat{j} + (-4(1) + \ln 1) \hat{k} = 3\hat{i} + 2\hat{j} - 4\hat{k}.$$

Equivalently, $\nabla f = \begin{bmatrix} 3 \\ 2 \\ -4 \end{bmatrix}$.

Prob 13 $g(x,y) = \frac{x-y}{xy+2}$, $P_0(1,-1)$. Find the directional derivative of g at P_0 in the direction of $\bar{u} = 12\hat{i} + 5\hat{j}$.

First, we find the unit vector in the direction of \bar{u} .

$$\hat{u} = \frac{\bar{u}}{\|\bar{u}\|} = \frac{12}{13}\hat{i} + \frac{5}{13}\hat{j}$$

length (or norm) of $\bar{u} = \sqrt{12^2 + 5^2} = 13$

My notation: lower case letters with a bar denote vectors
e.g., \bar{u} or \bar{a} ...

lower case letters with a hat $\hat{\cdot}$: unit vectors.
 \hat{u}, \hat{j} etc. $\|\hat{u}\|=1$.

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

$$= \frac{(xy+2)(1-0) - (x-y)(y+0)}{(xy+2)^2} \hat{i} + \frac{(xy+2)(0-1) - (x-y)(x+0)}{(xy+2)^2} \hat{j}$$

$$= \frac{(xy+2) - xy + y^2}{(xy+2)^2} \hat{i} + \frac{-(xy+2) - x^2 + xy}{(xy+2)^2} \hat{j} \quad \text{plug in } P_0(1,-1)$$

$$= \frac{y^2+2}{(xy+2)^2} \hat{i} - \frac{(x^2+2)}{(xy+2)^2} \hat{j}$$

$$= \frac{3}{1} \hat{i} + \frac{-3}{1} \hat{j} = 3\hat{i} - 3\hat{j}$$

$$\begin{aligned}
 (\nabla_{\bar{u}} g)_{P_0} &= \nabla g \cdot \hat{u} = (3\hat{i} - 3\hat{j}) \cdot \left(\frac{12}{13}\hat{i} + \frac{5}{13}\hat{j} \right) \\
 &= \frac{3 \times 12}{13} - \frac{3 \times 5}{13} = \frac{21}{13}.
 \end{aligned}$$

Notice the three steps in such problems. First, we find the unit vector \hat{u} in the direction of \bar{u} . Then we find the gradient ∇g at the given point P_0 . Third, we take the scalar product of ∇g and \hat{u} to get $(\nabla_{\bar{u}} g)_{P_0}$.