

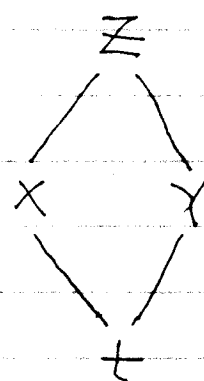
Math 17C

Kouba

Chain Rule, Gradient Vectors, and Directional Derivatives

Chain Rule: Assume $z = f(x, y)$,
 $x = g(t)$, and $y = h(t)$:
Then

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



proof: Consider the
linearization $L(x, y)$
for $f(x, y)$ at the point (x_0, y_0) :

$$L(x, y) = f(x_0, y_0) + \frac{\partial f}{\partial x}(x_0, y_0)(x - x_0) + \frac{\partial f}{\partial y}(x_0, y_0)(y - y_0).$$

Let $(x_0 + \Delta x, y_0 + \Delta y)$ be a point near
 (x_0, y_0) . The difference of the
 z -values is

$$\Delta z = f(x_0 + \Delta x, y_0 + \Delta y) - f(x_0, y_0)$$

since $\nearrow$
 $L(x, y) \approx f(x, y) \approx L(x_0 + \Delta x, y_0 + \Delta y) - f(x_0, y_0)$

$$= \cancel{f(x_0, y_0)} + \frac{\partial f}{\partial x}(x_0, y_0)((x_0 + \Delta x) - x_0) + \frac{\partial f}{\partial y}(x_0, y_0)((y_0 + \Delta y) - y_0) - \cancel{f(x_0, y_0)}$$

$$\rightarrow \Delta z \approx \frac{\partial f}{\partial x}(x_0, y_0) \cdot \Delta x + \frac{\partial f}{\partial y}(x_0, y_0) \cdot \Delta y$$

$$\rightarrow \frac{\Delta z}{\Delta t} \approx \frac{\partial f}{\partial x}(x_0, y_0) \cdot \frac{\Delta x}{\Delta t} + \frac{\partial f}{\partial y}(x_0, y_0) \cdot \frac{\Delta y}{\Delta t}$$

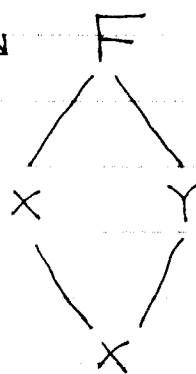
(now let $\Delta t \rightarrow 0$.)

$$\rightarrow \frac{dz}{dt} = \frac{\partial f}{\partial x} \cdot \frac{dx}{dt} + \frac{\partial f}{\partial y} \cdot \frac{dy}{dt} \quad \text{QED}$$

Note: Suppose that $F(x, y) = C$, a constant, and $y = g(x)$. Then $\rightarrow$ so that

$$\frac{dF}{dx} = F_x \cdot \frac{dx}{dx} + F_y \cdot \frac{dy}{dx} = 0$$

$$\rightarrow \boxed{\frac{dy}{dx} = -\frac{F_x}{F_y}}$$



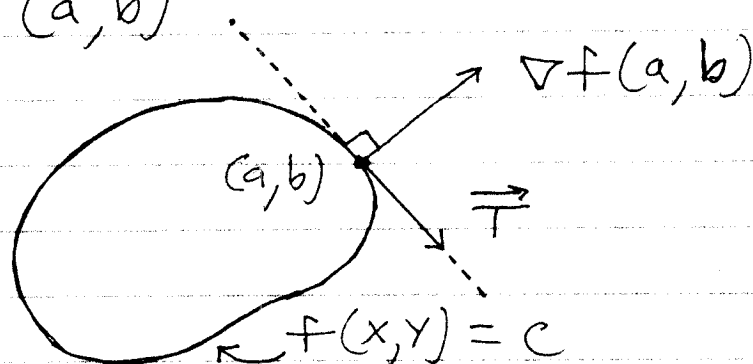
This is an alternative method to implicit differentiation.

Gradient Vector

Def: Let $z = f(x, y)$ be a function.
The gradient vector of f at point (x, y) is

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

Fact: Let $z = f(x, y)$ be a function and let $(x, y) = (a, b)$ with $f(a, b) = c$. Then the gradient vector $\nabla f(a, b)$ is perpendicular to the level curve $f(x, y) = c$ at the point (a, b) .



proof: The SLOPE of the tangent line at (a, b) is

$$\frac{dy}{dx} = -\frac{f_x}{f_y} = \frac{\text{rise}}{\text{run}} = \frac{\text{change in } y}{\text{change in } x},$$

so a vector parallel to the tangent line is

$$\vec{T} = \begin{bmatrix} f_y \\ -f_x \end{bmatrix}.$$

(Recall that vectors $\vec{v}$ and $\vec{w}$ are $\perp$ iff $\vec{v} \cdot \vec{w} = 0$.) Then

$$\begin{aligned} \vec{T} \cdot \nabla f(a, b) &= \begin{bmatrix} f_y \\ -f_x \end{bmatrix} \cdot \begin{bmatrix} f_x \\ f_y \end{bmatrix} \\ &= f_x f_y - f_x f_y = 0. \end{aligned}$$

Thus, $\nabla f(a, b)$ is $\perp$ to $f(x, y) = c$.

QED

Directional Derivative

Consider a point (a,b) and a unit vector (length = 1) $\vec{u} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$. The

line through (a,b) in the direction of $\vec{u}$ is given parametrically by

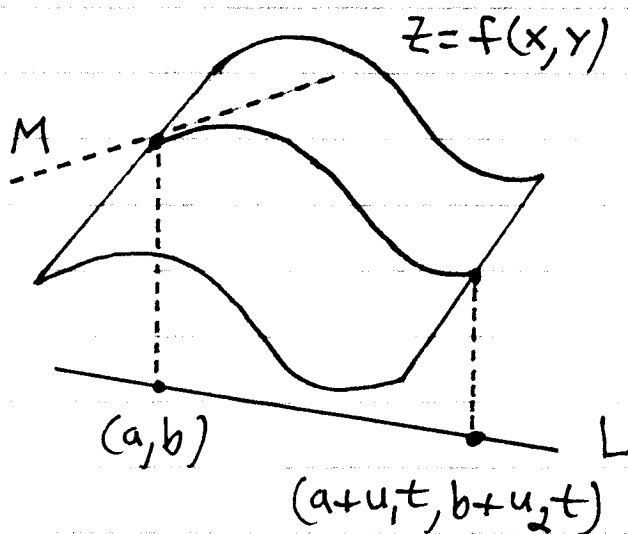
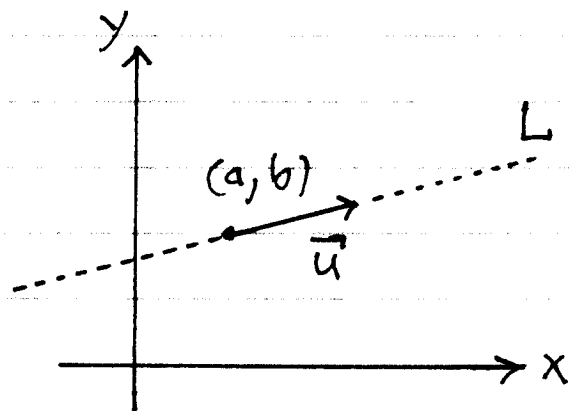
$$L: \begin{cases} x = a + u_1 t \\ y = b + u_2 t \end{cases} \quad \text{for } -\infty < t < \infty.$$

Let $z = f(x,y)$ be a surface defined on points (x,y) on line L .

We want the slope of the line M , which is tangent to surface $z = f(x,y)$ at the point (a,b) in the direction $\vec{u}$.

Note: Along line L we have

$$\begin{aligned} z = f(x,y) &= f(a + u_1 t, b + u_2 t) \\ &= z(t), \text{ a function of } t \text{ only!} \end{aligned}$$

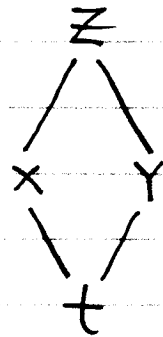


Def: The slope of the line tangent to the surface $z = f(x, y)$ at point (a, b) in the direction of unit vector $\vec{u}$ is called the directional derivative of f and is written

$$D_{\vec{u}} f(a, b) .$$

Q: How do we compute the value of $D_{\vec{u}} f(a, b)$?

$$D_{\vec{u}} f(x, y) = \frac{d}{dt} z(t)$$



$$= \frac{d}{dt} f(a+u_1 t, b+u_2 t)$$

Chain

Rule $\searrow$

$$= f_x(a+u_1 t, b+u_2 t) \cdot \frac{dx}{dt}$$

$$+ f_y(a+u_1 t, b+u_2 t) \cdot \frac{dy}{dt}$$

$$= f_x(a+u_1 t, b+u_2 t) \cdot u_1$$

$$+ f_y(a+u_1 t, b+u_2 t) \cdot u_2 .$$

now let $t=0$ so that $(x, y) = (a, b) \rightarrow$

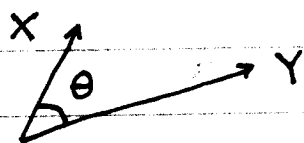
$$D_{\vec{u}} f(a, b) = f_x(a, b) \cdot u_1 + f_y(a, b) \cdot u_2$$

$$= \begin{bmatrix} f_x(a, b) \\ f_y(a, b) \end{bmatrix} \cdot \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} ,$$

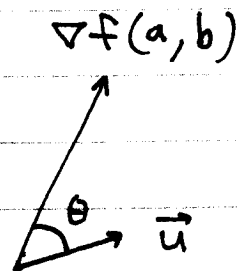
i.e.,

$$D_{\vec{u}} f(a, b) = \nabla f(a, b) \cdot \vec{u}$$

Recall: $X \cdot Y = |X||Y| \cos \theta$



$$\begin{aligned} \text{Then } D_{\vec{u}} f(a,b) &= \nabla f(a,b) \cdot \vec{u} \\ &= |\nabla f(a,b)| |\vec{u}| \cos \theta \\ &= |\nabla f(a,b)| \cdot \cos \theta \end{aligned}$$



This leads to the following facts:

Facts: 1.) $D_{\vec{u}} f(a,b)$ is largest (f is $\uparrow$ most rapidly) when $\theta = 0$ and $\vec{u}$ points in the direction of the gradient vector $\nabla f(a,b)$.

2.) $D_{\vec{u}} f(a,b)$ is smallest (f is $\downarrow$ most rapidly) when $\theta = \pi$ and $\vec{u}$ points in the opposite direction of the gradient vector $\nabla f(a,b)$.