

Lecture 14.6

Last time:

- The chain rule expresses the (partial) derivative(s) of a multivariable function $f(x_1, \dots, x_n)$, where x_1, \dots, x_n are each (multivariable) functions of t (or more variables), as

$$\frac{df}{dt} = \frac{\partial f}{\partial x_1} \cdot \frac{dx_1}{dt} + \dots + \frac{\partial f}{\partial x_n} \cdot \frac{dx_n}{dt}$$

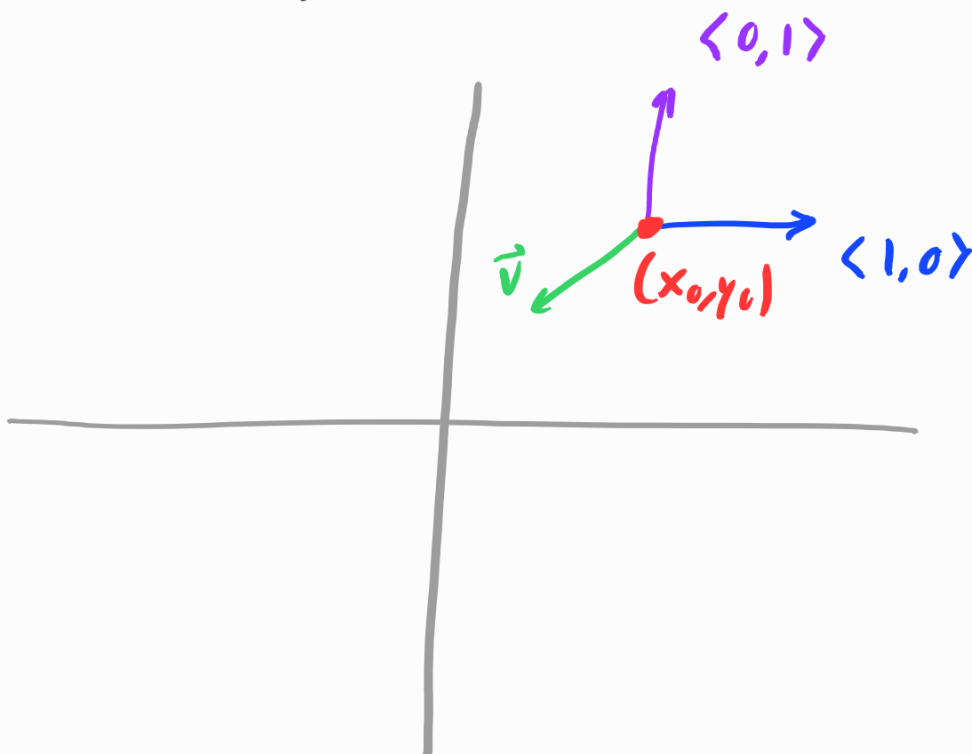
(something similar for $\frac{\partial f}{\partial t_j}$).

- For an implicit equation $F(x, y) = 0$,

$$\frac{dy}{dx} = \frac{-\frac{\partial F}{\partial x}}{\frac{\partial F}{\partial y}} = \frac{-F_x}{F_y}.$$

Directional Derivatives

Partial derivatives of $f(x,y)$ measure the rates of change of f in the x - and y -directions.



But we know there are many ways

to move around in \mathbb{R}^2 , for example
along a vector \vec{v} starting at (x_0, y_0) .

Def Let $f(x_1, \dots, x_n)$ be a function
and \vec{v} a vector in \mathbb{R}^n . Then the
directional derivative of f at a point
 $P = (a_1, \dots, a_n)$ in the direction of \vec{v} ,
written $D_{\vec{v}} f(a_1, \dots, a_n)$, is the rate
of change of f at P . Explicitly,
let $\vec{u} = \frac{\vec{v}}{|\vec{v}|}$ be the unit vector in
the direction of \vec{v} . Then

$$D_{\vec{v}} f = \lim_{h \rightarrow 0} \frac{f(P + h\vec{u}) - f(P)}{h}$$

$$= \lim_{h \rightarrow 0} \frac{f(a_1 + hu_1, \dots, a_n + hu_n) - f(a_1, \dots, a_n)}{h}$$

Note: if $\vec{u} = \frac{\vec{v}}{|\vec{v}|}$, then $D_{\vec{u}} f = D_{\vec{v}} f$

so we might as well just focus on unit vectors as inputs to $D_{\vec{v}} f$.

For 2-variable functions $f(x, y)$, points

$P = (a, b)$ and unit vectors $\vec{u} = \langle u_1, u_2 \rangle$,

this looks like:

$$D_{\vec{u}} f = \lim_{h \rightarrow 0} \frac{f(a + hu_1, b + hu_2) - f(a, b)}{h}$$

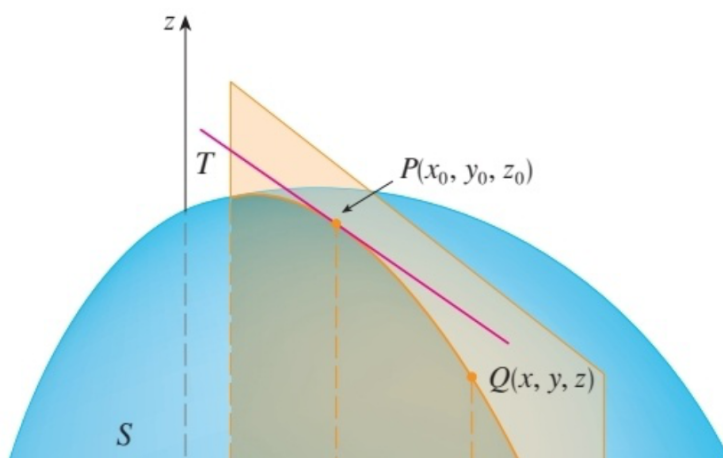
$$D_{\vec{u}} f(a, b) = \lim_{h \rightarrow 0} \frac{f(a + hu_1, b + hu_2) - f(a, b)}{h}$$

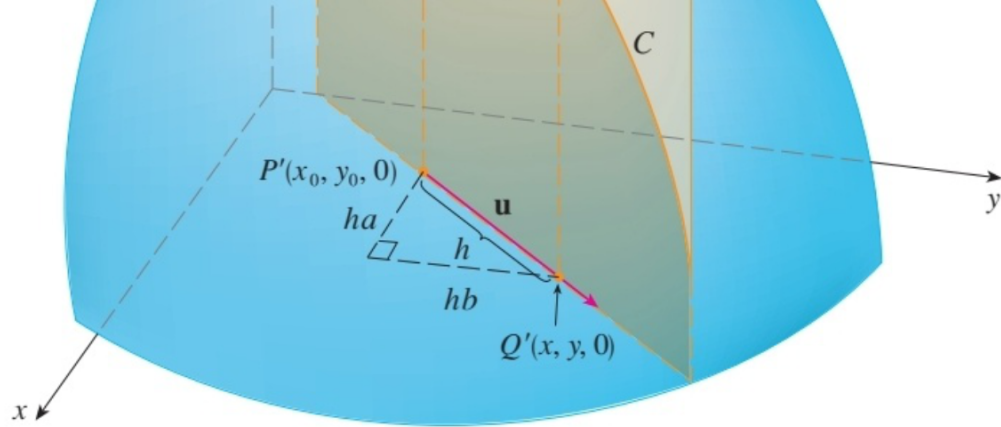
Ex For $\vec{u} = i = \langle 1, 0 \rangle$, this is just the partial derivative with respect to x :

$$D_i f(x, y) = \frac{\partial f}{\partial x}(x, y).$$

Likewise, for $\vec{u} = j = \langle 0, 1 \rangle$,

$$D_j f(x, y) = \frac{\partial f}{\partial y}(x, y).$$



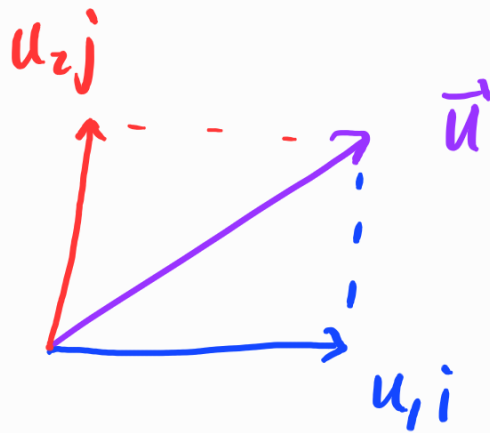


Here, you can see that the "slope" in the \vec{u} direction is the slope of a cross section of $z = f(x, y)$.

Important computational tool: rather than computing $D_{\vec{u}}f$ by a limit everytime, let's decompose \vec{u} into its coordinate directions. Algebraically,

$$\vec{u} = \langle u_1, u_2 \rangle = u_1 i + u_2 j.$$

Visually,



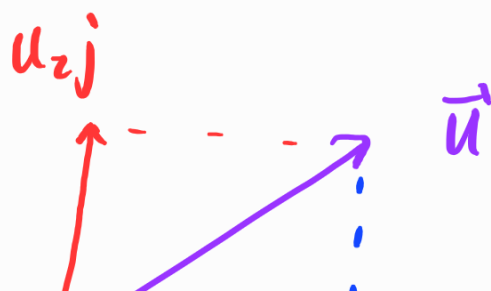
Theorem

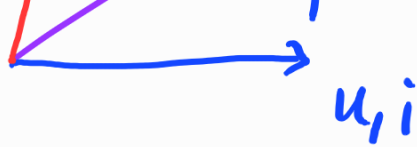
For any function $f(x, y)$ and

unit vector $\vec{u} = \langle u_1, u_2 \rangle$,

$$D_{\vec{u}} f(x, y) = \langle f_x, f_y \rangle \cdot \vec{u}$$

$$= f_x u_1 + f_y u_2.$$





$$\text{change in } f \text{ in } \vec{u} \text{ direction} = \text{change in } f \text{ in } x\text{-direction} + \text{change in } f \text{ in } y\text{-direction}$$

This vector $\langle f_x, f_y \rangle$ has a special name, which we state more generally now.

Def The gradient of $f(x_1, \dots, x_n)$ is the vector-valued function

$$\nabla f = \langle f_{x_1}, f_{x_2}, \dots, f_{x_n} \rangle.$$

↳ "nabla"

Theorem For any $f(x_1, \dots, x_n)$ and unit

vector \vec{u} in \mathbb{R}^n ,

$$D_{\vec{u}} f = \nabla f \cdot \vec{u},$$

Ex 1 For $f(x, y) = x^2 + y^2$,

$$f_x = 2x \quad \text{and} \quad f_y = 2y.$$

In the direction of $\vec{v} = \langle 1, 1 \rangle$,

a unit vector is $\vec{u} = \langle \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \rangle$

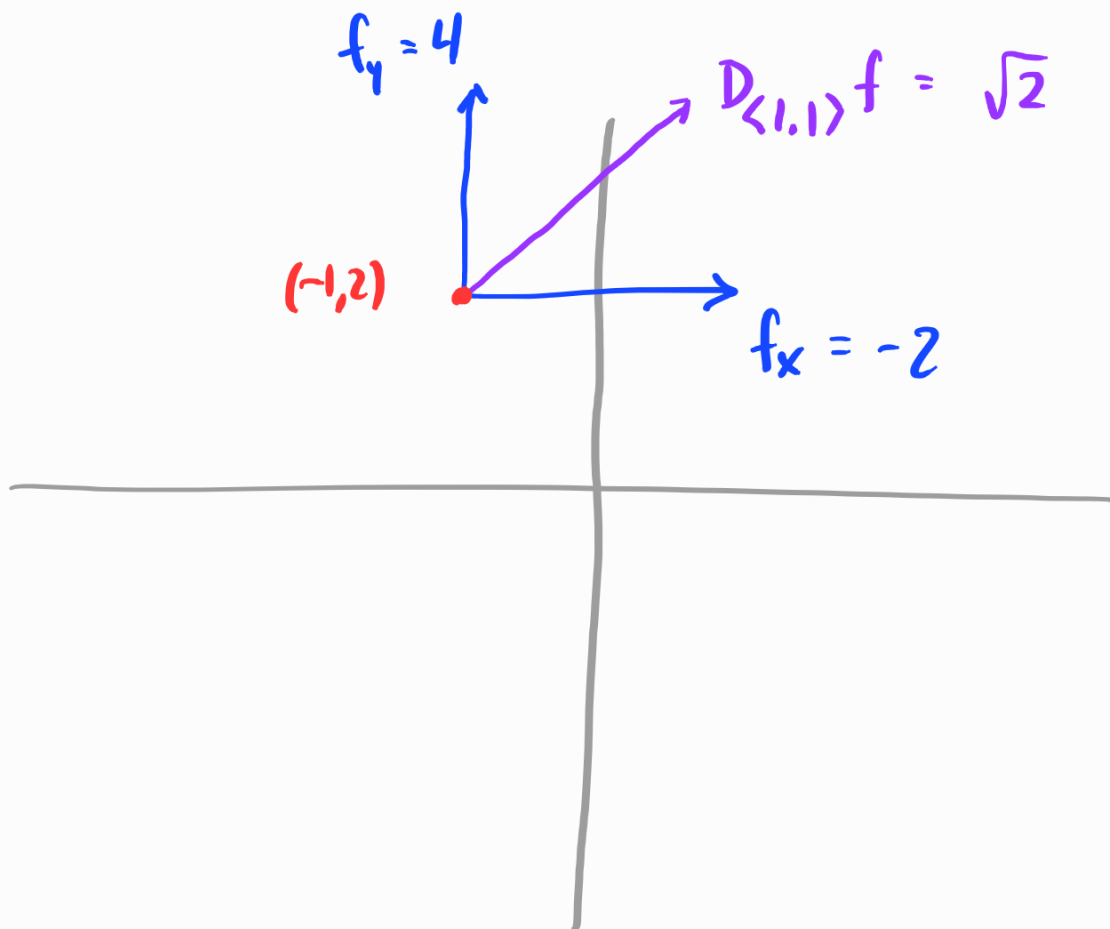
and the directional derivative is

$$D_{\vec{u}} f = \nabla f \cdot \vec{u}$$

$$= \langle 2x, 2y \rangle \cdot \langle \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \rangle$$

$$= x\sqrt{2} + y\sqrt{2}.$$

Thus at $P = (-1, 2)$ for example,



Exercise 1: Compute the gradients of each function.

(a) $f = \sqrt{x^2 + y^4}$

$$(b) f = x^2 \sec(3x) - \frac{x^2}{y^3}$$

$$(c) f = \ln(xy + z^2)$$

Exercise 2: Compute $D_{\vec{v}}f$ for each of the following functions and vectors and evaluate at the point, if given.

$$(a) f = x \cos y, \quad \vec{v} = \langle 2, 1 \rangle, \quad (a, b) = (0, \pi)$$

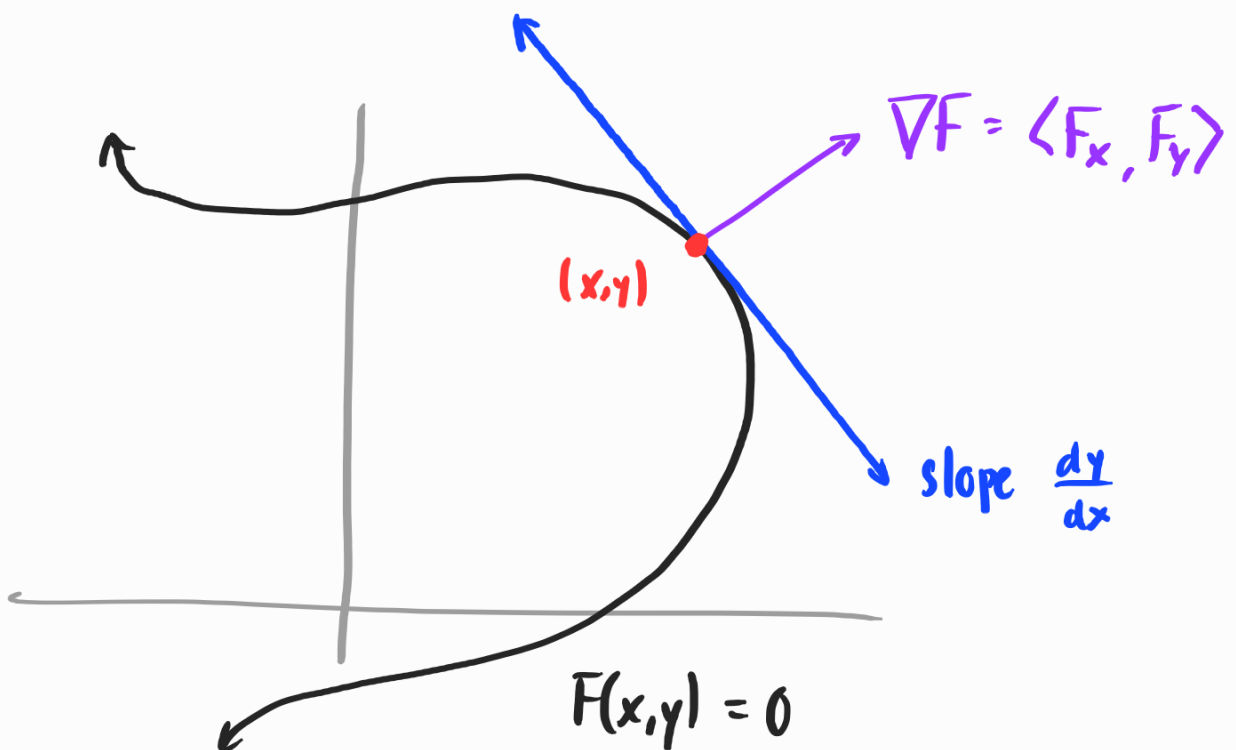
$$(b) f = x^3 - 3xy + 4y^2, \quad \vec{u} = \text{unit vector with angle } \theta = \pi/6, \quad (a, b) = (1, 2)$$

$$(c) f = xe^{xy} + yz, \quad \vec{u} = \langle -1, 2, 2 \rangle$$

Recall that if $F(x,y) = 0$, then

$$\frac{dy}{dx} = -\frac{F_x}{F_y}.$$

In a picture,



Notice that ∇F is orthogonal to the

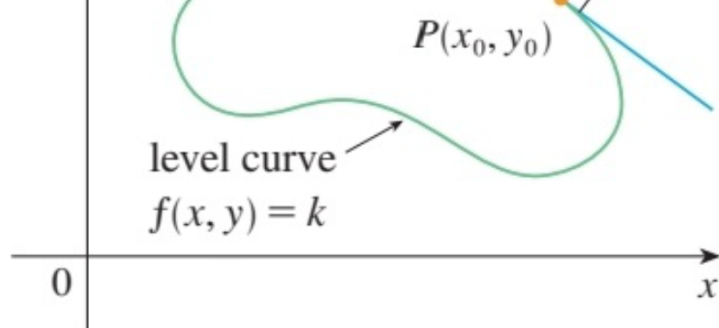
tangent line:

$$\begin{aligned}\nabla F \cdot \left\langle 1, \frac{dy}{dx} \right\rangle &= \langle F_x, F_y \rangle \cdot \left\langle 1, -\frac{F_x}{F_y} \right\rangle \\ &= F_x - F_y \frac{F_x}{F_y} = 0.\end{aligned}$$

This is true for any level curve:

Theorem For any function $f(x, y)$,
the gradient $\nabla f(x_0, y_0)$ is orthogonal
to the level curve $f(x, y) = k$
containing (x_0, y_0) (namely, $k = f(x_0, y_0)$).





What does this mean in terms of rates of change of f ?

Well, $D_{\vec{u}} f = \nabla f \cdot \vec{u}$ can be rewritten

$$D_{\vec{u}} f = |\nabla f| \cos(\theta)$$

where θ is the angle between ∇f and \vec{u} .

This is **maximized** when $\theta = 0$, i.e.

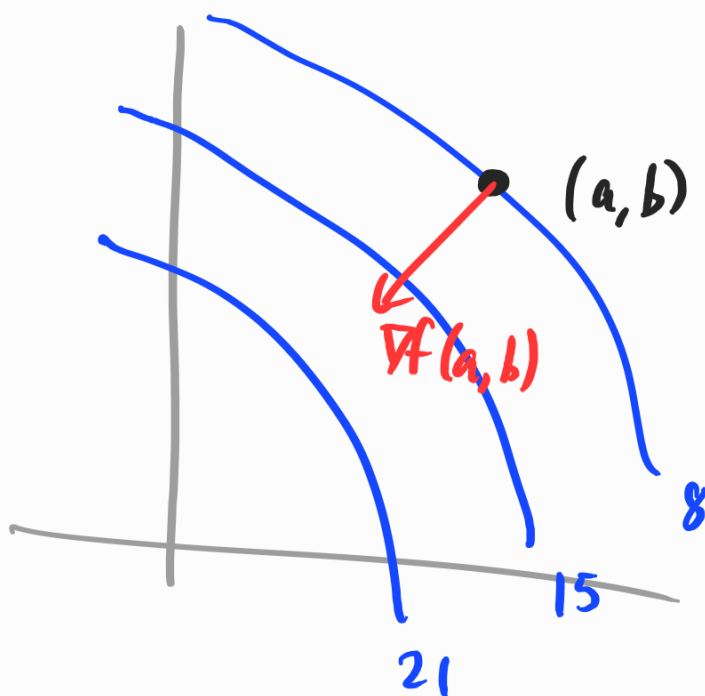
when \vec{u} is in the same direction

as ∇f .

Similarly, $D_{\vec{u}} f$ is minimized when $\theta = \pi$,

i.e. when \vec{u} is in the opposite direction as ∇f .

Visually, "the fastest rate of change is at a right angle to the level curves":



Exercise 3: Find the maximum rate of change of $f(x,y) = \sqrt{x^2 + y^4}$ at $(-2, 3)$. Which direction does this maximum rate occur in?

Exercise 4: The elevation of a hill is given by the height function

$$h(x,y) = 1000 - \frac{1}{100}x^2 - \frac{1}{50}y^2.$$

standing at the point $(60, 100)$,

(a) in what direction is the elevation changing fastest?

(b) what is this elevation change?

Next time: optimization 

