

More on surfaces:

Session 7
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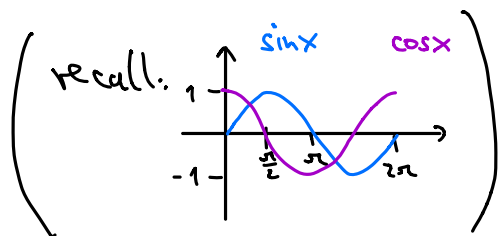
nice visualizations: geogebra.org
↳ 3D Calculator

Example from last time: $f(x, y) = x \sin(y)$

↳ tangent plane at $\vec{a} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$ is $z = a_1 \sin(a_2) + \sin(a_2)(x - a_1) + a_1 \cos(a_2)(y - a_2)$

↳ e.g., at $\vec{a} = \begin{pmatrix} 1 \\ \frac{\pi}{2} \end{pmatrix}$, we have $z = 1 \cdot \sin\left(\frac{\pi}{2}\right) + \sin\left(\frac{\pi}{2}\right)(x - 1) + 1 \cdot \cos\left(\frac{\pi}{2}\right)(y - \frac{\pi}{2})$
 $= 1 + (x - 1)$

$$= x$$



Note/recall: \exists at least two ways to represent surfaces (here, 2 dimensional)

• graph of a function: $\left\{ \begin{pmatrix} x \\ y \\ f(x, y) \end{pmatrix} \in \mathbb{R}^3 : x \in \mathbb{R}, y \in \mathbb{R} \right\}$

↳ only one z -value for each (x, y)

Ex.: upper half-sphere: $f(x, y) = \sqrt{4 - x^2 - y^2}$ for $0 \leq x^2 + y^2 \leq 4$

• solution to equation $F(x, y, z) = 0$: $\left\{ \begin{pmatrix} x \\ y \\ z \end{pmatrix} \in \mathbb{R}^3 : F(x, y, z) = 0 \right\}$

↳ several z possible given some (x, y)

Ex.: sphere: $F(x, y, z) = x^2 + y^2 + z^2 - 4 = 0$

For the special case of tangent planes, we have:

• tangent plane at $\vec{a} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$ = graph of $z(x,y) = f(\vec{a}) + \underbrace{\begin{pmatrix} \partial_x f(\vec{a}) \\ \partial_y f(\vec{a}) \end{pmatrix}}_{(\nabla f)(\vec{a})} \cdot \begin{pmatrix} x - a_1 \\ y - a_2 \end{pmatrix}$
 $= (\nabla f)(\vec{a}) \cdot (\vec{x} - \vec{a})$
 (2 dim. scalar product)

• let us write $\vec{X} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$, $\vec{A} = \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}$ with $a_3 = f(a_1, a_2)$, and $F(x, y, z) = f(x, y) - z$,

such that surface (= graph of f) is solution to $F(x, y, z) = 0$

\Rightarrow equation for tangent plane becomes $\underbrace{(\nabla F)(\vec{A})}_{\text{3-dim. scalar product}} \cdot (\vec{X} - \vec{A}) = 0$

check: $(\nabla F)(\vec{A}) = \begin{pmatrix} \partial_x f \\ \partial_y f \\ \partial_z(-z) \end{pmatrix}(\vec{A}) = \begin{pmatrix} \partial_x f \\ \partial_y f \\ -1 \end{pmatrix}(\vec{A})$, and $(\vec{X} - \vec{A}) = \begin{pmatrix} x - a_1 \\ y - a_2 \\ z - f(a_1, a_2) \end{pmatrix}$ ✓

back to differentiability ...

recall what we defined so far for $f: \mathbb{R}^n \rightarrow \mathbb{R}$

• $\frac{\partial f}{\partial x_i} = \lim_{h \rightarrow 0} \frac{f(\vec{x} + h\vec{e}_i) - f(\vec{x})}{h}$, partial derivative

• for a unit vector \vec{u} (i.e., $|\vec{u}| = 1$), $D_{\vec{u}} f := \lim_{h \rightarrow 0} \frac{f(\vec{x} + h\vec{u}) - f(\vec{x})}{h}$, directional derivative

• f (totally) differentiable at \vec{a} if $f(\vec{a} + \vec{h}) = f(\vec{a}) + \vec{m} \cdot \vec{h} + E(\vec{h})$ with $\lim_{h \rightarrow 0} \frac{E(\vec{h})}{|\vec{h}|} = 0$ for some \vec{m} .

Note: \vec{m} is called the (total) derivative of f at \vec{a} .

let us give here (without proof) the connections between these notions:

Theorem:

- If f differentiable at \vec{a} , then all partial derivatives exist, and the total derivative is the gradient (i.e., $\vec{m} = (\vec{\nabla}f)(\vec{a})$).
- f continuously differentiable at $\vec{a} \iff$ all partial derivatives at \vec{a} exist and are continuous

Note: • \exists examples where all partial derivatives exist, but the function is not differentiable,

e.g.: $f(x,y) = \begin{cases} \frac{xy^2}{x^2+y^2} & \text{for } (x,y) \neq (0,0) \\ 0 & \text{for } (x,y) = (0,0) \end{cases}$ (bonus problem on HW)

"one cannot sensibly put a tangent plane at the origin $(0,0)$ "

- If all partial derivatives exist and are continuous, f is called a C^1 -function (or "of class C^1 "). If all combinations of k partial derivatives exist and are continuous, f is called C^k -function.

Theorem: If f differentiable at \vec{a} , then the directional derivatives $(D_{\vec{u}}f)(\vec{a})$ exist for all $\vec{u} \in \mathbb{R}^n$, $|\vec{u}| = 1$, and

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

Proof: f differentiable at \vec{a} means $\frac{f(\vec{a} + \vec{h}) - f(\vec{a}) - (\vec{\nabla}f)(\vec{a}) \cdot \vec{h}}{|\vec{h}|} \xrightarrow{h \rightarrow 0} 0$

now let $\vec{h} = t\vec{u}$ ($t > 0$) \implies $\frac{f(\vec{a} + t\vec{u}) - f(\vec{a}) - (\vec{\nabla}f)(\vec{a}) \cdot t\vec{u}}{t|\vec{u}|} = \frac{f(\vec{a} + t\vec{u}) - f(\vec{a})}{t} - (\vec{\nabla}f)(\vec{a}) \cdot \vec{u}$
 $\xrightarrow{t \rightarrow 0} (D_{\vec{u}}f)(\vec{a}) \quad \square$

Note: $f(x,y) = \begin{cases} \frac{xy^2}{x^2+y^2} & \text{for } (x,y) \neq (0,0) \\ 0 & \text{for } (x,y) = (0,0) \end{cases}$

is an example where all directional derivatives at $(0,0)$ exist, but f is not differentiable at $(0,0)$.

Geometric interpretation for $(\vec{\nabla} f)(\vec{a}) \neq 0$:

↳ recall that $(\vec{\nabla} f)(\vec{a}) \cdot \vec{u} = |(\vec{\nabla} f)(\vec{a})| \underbrace{|\vec{u}|}_{=1} \cos \theta$ ($\theta = \text{angle between } (\vec{\nabla} f)(\vec{a}) \text{ and } \vec{u}$)

\Rightarrow directional derivative maximal if \vec{u} points in same direction as $(\vec{\nabla} f)(\vec{a})$

$\Rightarrow (\vec{\nabla} f)(\vec{a})$ points in direction of largest directional derivative

Example from last time: $f(x,y) = x^2 y + 5x$, $\vec{u} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ (s.t. $|\vec{u}| = 1$)

$$\Rightarrow (D_{\vec{u}} f)(x,y) = \begin{pmatrix} \partial_x f \\ \partial_y f \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} \frac{1}{\sqrt{2}}$$

$$= \frac{1}{\sqrt{2}} \left(\partial_x f + \partial_y f \right) = \frac{1}{\sqrt{2}} \left(2xy + 5 + x^2 \right), \text{ same as last time}$$

Next: chain rule

consider $x_1(t), \dots, x_n(t)$, and a function $f(x_1(t), \dots, x_n(t))$

heuristic computation: the differential is $df = \frac{\partial f}{\partial x_1} dx_1 + \dots + \frac{\partial f}{\partial x_n} dx_n$

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

Theorem: let f be differentiable at $\vec{x} = \vec{b}$ and $x_1(t), \dots, x_n(t)$ be differentiable at $t = a$,

$$\vec{b} = \vec{x}(a), \vec{x}(t) = \begin{pmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{pmatrix}. \text{ Then } \frac{df}{dt}(a) = (\vec{\nabla} f)(\vec{b}) \cdot \left(\frac{d\vec{x}}{dt} \right)(a)$$

$$= \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\vec{b}) \frac{dx_i}{dt}(a).$$

Note: $\vec{x}(t)$ could, e.g., be the trajectory of a particle

$\Rightarrow \frac{d\vec{x}}{dt}$ is the velocity

Ex.: $f(x_1, x_2) = x_1 e^{-x_2}$, $x_1(t) = t^2$, $x_2(t) = t^3$

$$\begin{aligned}\Rightarrow \frac{df}{dt} &= \frac{\partial f}{\partial x_1} \frac{dx_1}{dt} + \frac{\partial f}{\partial x_2} \frac{dx_2}{dt} = e^{-x_2(t)} 2t - x_1(t) e^{-x_2(t)} 3t^2 \\ &= (2t - 3t^4) e^{-t^3}\end{aligned}$$

Note: $\frac{df}{dt}$ is here the usual one-dimensional derivative, so in this simple example we also could have computed explicitly

$$\frac{d}{dt} f(x_1(t), x_2(t)) = \frac{d}{dt} (t^2 e^{-t^3}) = 2t e^{-t^3} + t^2 (-3t^2) e^{-t^3} = (2t - 3t^4) e^{-t^3}$$