

Mathematician spotlight: David Rockoff, University of Arizona

- differential item functioning - fairness of tests
- uses randomization, simulations & data

Last time: chain rule for multivariable functions

This time: vector-valued chain rule example, plus directional derivatives!

Review: Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$, $g: \mathbb{R} \rightarrow \mathbb{R}$. Then $\frac{d}{dx} f(g(x)) = \frac{d}{dx} (f \circ g)(x) = f'(g(x)) \cdot g'(x)$

New: Suppose $f: \mathbb{R}^m \rightarrow \mathbb{R}^p$, $g: \mathbb{R}^n \rightarrow \mathbb{R}^m$. function composition notation $= \frac{df}{dg} \cdot \frac{dg}{dx}$.

Then the Jacobian of $f \circ g: \mathbb{R}^n \rightarrow \mathbb{R}^p$ is

$$D(f \circ g) = \begin{bmatrix} Df \\ (p \times m) \end{bmatrix} \begin{bmatrix} Dg \\ (m \times n) \end{bmatrix} = \begin{bmatrix} Df \cdot Dg \\ (p \times n) \end{bmatrix}$$

Let's do an example!

Example. Suppose $f(x,y,z) = \begin{bmatrix} x y z \\ x + y \end{bmatrix} = \begin{bmatrix} f_1(x,y,z) \\ f_2(x,y,z) \end{bmatrix}$ and $x(s,t) = st$, $y(s,t) = st$, $z(s,t) = s^2 - t^2$, so $g(s,t) = \begin{bmatrix} st \\ st \\ s^2 - t^2 \end{bmatrix}$.

Consider $f \circ g = f(g): \mathbb{R}^2 \rightarrow \mathbb{R}^2$, because $\mathbb{R}^2 \xrightarrow{g} \mathbb{R}^3 \xrightarrow{f} \mathbb{R}^2$.

$$\text{Then } D(f \circ g) = Df \cdot Dg = \begin{bmatrix} \partial f_1 / \partial x & \partial f_1 / \partial y & \partial f_1 / \partial z \\ \partial f_2 / \partial x & \partial f_2 / \partial y & \partial f_2 / \partial z \end{bmatrix} \begin{bmatrix} \partial x / \partial s & \partial x / \partial t \\ \partial y / \partial s & \partial y / \partial t \\ \partial z / \partial s & \partial z / \partial t \end{bmatrix} = \begin{bmatrix} yz & xz & xy \\ 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} t & s \\ 1 & 1 \\ 2s & -2t \end{bmatrix}$$

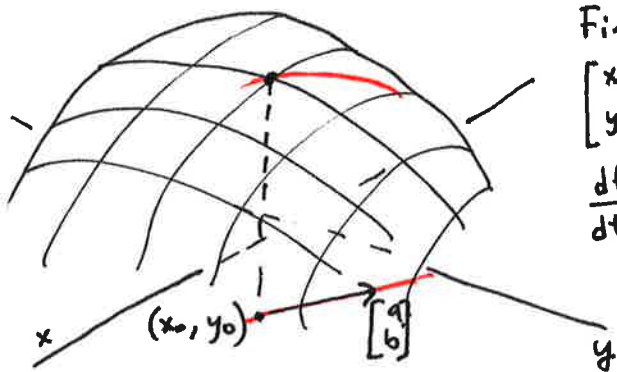
now multiply out and put everything in terms of s and t.

If you want to find this Jacobian matrix at a point $(s,t) = (1,1)$, compute $(x,y,z) = (1,2,0)$ and plug in:

$$\left. \begin{aligned} D(f \circ g)(\vec{a}) &= Df(g(\vec{a})) \cdot Dg(\vec{a}) \\ D(f \circ g)(1,1) &= Df(g(1,1)) \cdot Dg(1,1) \\ &= Df(1,2,0) \cdot Dg(1,1) \end{aligned} \right\} = \begin{pmatrix} 0 & 0 & 2 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & 1 \\ 2 & -2 \end{pmatrix} = \begin{pmatrix} 4 & -4 \\ 2 & 2 \end{pmatrix} = \begin{pmatrix} \partial f_1 / \partial s & \partial f_1 / \partial t \\ \partial f_2 / \partial s & \partial f_2 / \partial t \end{pmatrix}$$

Today: Directional derivatives! We know the "slope" of $z = f(x,y)$ in the positive x-direction is f_x , and in pos y-direction is f_y , but what about the slope in other directions?

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



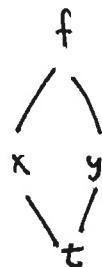
Find "slope" at (x_0, y_0) in direction $\begin{bmatrix} a \\ b \end{bmatrix}$: This line is

$$\begin{bmatrix} x(t) \\ y(t) \end{bmatrix} = \begin{bmatrix} x_0 + at \\ y_0 + bt \end{bmatrix} \text{ and we want } \frac{d}{dt} f(x(t), y(t)).$$

$$\frac{df}{dt} = \frac{\partial f}{\partial x} \cdot \frac{dx}{dt} + \frac{\partial f}{\partial y} \cdot \frac{dy}{dt} = \frac{\partial f}{\partial x} \cdot a + \frac{\partial f}{\partial y} \cdot b$$

$$= \begin{bmatrix} \partial f / \partial x & \partial f / \partial y \end{bmatrix} \cdot \begin{bmatrix} a \\ b \end{bmatrix} \Rightarrow \boxed{D_{\vec{u}} f = \nabla f \cdot \vec{u}}$$

Directional derivative of f in direction $\vec{u} = \begin{bmatrix} a \\ b \end{bmatrix}$.



Note that the "slope" should not depend on the length of the direction vector \vec{u} that we choose, so we always take \vec{u} to be a unit vector:

The directional derivative of $f(x,y)$ at (x_0, y_0) in the direction of the unit vector \vec{u} is $D_{\vec{u}} f(x_0, y_0) = \nabla f(x_0, y_0) \cdot \vec{u}$.

Example. Let $f(x,y) = x^2 - y^2$. Find the directional derivative of f at $(3,1)$ in the direction $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$.

① Find $\nabla f = \begin{bmatrix} 2x \\ -2y \end{bmatrix}$ ② Find $\nabla f(3,1) = \begin{bmatrix} 6 \\ -2 \end{bmatrix}$ ③ Find a unit direction vector: $\vec{u} = \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}$

④ Put them together! $D_{\vec{u}} f(3,1) = \nabla f(3,1) \cdot \vec{u} = \begin{bmatrix} 6 \\ -2 \end{bmatrix} \cdot \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix} = 6/\sqrt{5} - 4/\sqrt{5} = \underline{\underline{2/\sqrt{5}}}$.

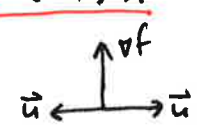
⑤ What does it mean? _____

Pondering the geometry. $D_{\vec{u}} f(x_0, y_0) = \nabla f(x_0, y_0) \cdot \vec{u} = \|\nabla f(x_0, y_0)\| \cdot \|\vec{u}\| \cdot \cos \theta$ ← where θ is the angle between $\nabla f(x_0, y_0)$ and \vec{u}
 $= \|\nabla f(x_0, y_0)\| \cdot \cos \theta$ ← since $\|\vec{u}\| = 1$

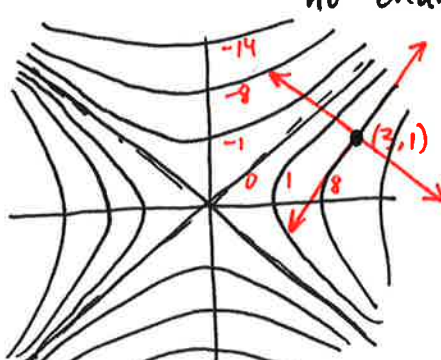
• When is $D_{\vec{u}} f$ the largest positive number? \rightarrow when $\cos \theta = 1$, $D_{\vec{u}} f$ is $\|\nabla f\|$, the largest
 \hookrightarrow happens when $\theta = 0$, i.e. ∇f and \vec{u} in same direction
 \Rightarrow gradient points in direction of steepest ascent.

• When is $D_{\vec{u}} f$ the largest negative number? \rightarrow when $\cos \theta = -1$, $D_{\vec{u}} f$ is $-\|\nabla f(x_0, y_0)\|$
 \hookrightarrow happens when $\theta = \pi$, i.e. ∇f and \vec{u} in opposite directions
 \Rightarrow direction of steepest descent is $-\nabla f(x_0, y_0)$.

• When is $D_{\vec{u}} f = 0$, i.e. which direction(s) can you go to stay on the same level? \rightarrow ① when $\cos \theta = 0$, so when $\theta = \frac{\pi}{2}$ or $\theta = \frac{3\pi}{2}$ so when ∇f and \vec{u} are perpendicular
 \rightarrow ② when you go along a level curve.



Example. For $f(x,y) = x^2 - y^2$, find directions of steepest ascent, steepest descent, and no change, and the directional derivatives (slopes) in those directions.



- steepest ascent: $\nabla f(3,1) = \begin{bmatrix} 6 \\ -2 \end{bmatrix}$ and slope is $\|\nabla f(3,1)\| = \sqrt{6^2 + (-2)^2} = \sqrt{40}$.
- steepest descent: _____
- no change: _____