

Contents

1	Directional Derivative	1
2	Gradient	2
3	Properties of the Gradient	3
3.1	Gradient and Directional Derivative	3
3.2	Gradient and Level Curves	5

1 Directional Derivative

Suppose we want to calculate the rate of change in a direction which isn't x or y , but a linear combination of the two. That is the notion of a directional derivative.

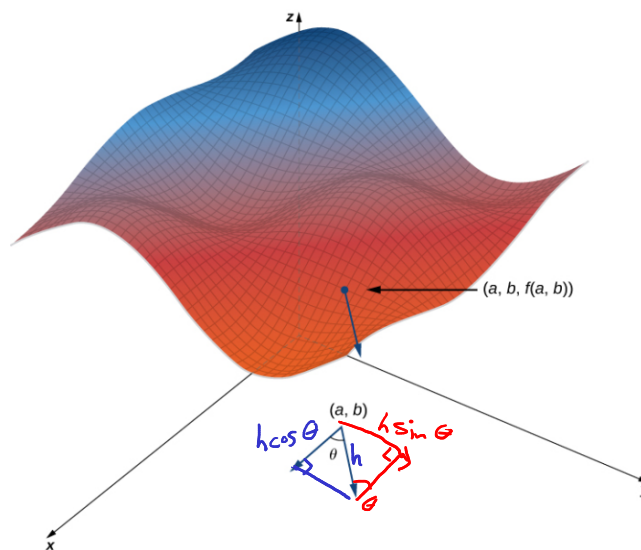


Figure 4.39 Finding the directional derivative at a point on the graph of $z = f(x, y)$. The slope of the black arrow on the graph indicates the value of the directional derivative at that point.

Definition. Given a unit vector $\mathbf{u} = \langle \cos \theta, \sin \theta \rangle$, the **directional derivative** of f in the direction $\mathbf{u}$ is given by

$$D_{\mathbf{u}}f(a, b) = \lim_{h \rightarrow 0} \frac{f(a + h \cos \theta, b + h \sin \theta) - f(a, b)}{h}$$

Another way to calculate a directional derivative involves partial derivatives:

Proposition. Let $\mathbf{u} = \langle \cos \theta, \sin \theta \rangle$ be a unit vector. The directional derivative of $f(x, y)$ in the direction of $\mathbf{u}$ is

$$\begin{aligned} D_{\mathbf{u}}f(x, y) &= f_x(x, y) \cos \theta + f_y(x, y) \sin \theta \\ &= \underbrace{\langle f_x(x, y), f_y(x, y) \rangle}_{\vec{\nabla} f} \cdot \underbrace{\langle \cos \theta, \sin \theta \rangle}_{\vec{u}} \end{aligned}$$

Proof. Uses chain rule. See the proof of Theorem 4.12 in the book. $\square$

Remark. If the given direction vector $\mathbf{v}$ is *not* a unit vector, you must normalize $\mathbf{v}$ to get the unit vector $\mathbf{u}$, before computing the directional derivative $D_{\mathbf{u}}f$.

2 Gradient

The vector $\langle f_x(x, y), f_y(x, y) \rangle$ has a name, the **gradient** of f , and is denoted ∇f . (The symbol ∇ is called “nabla” or “del”)

Definition.

$$\begin{aligned}\nabla f(x, y) &= \langle f_x(x, y), f_y(x, y) \rangle \\ \nabla f(x, y, z) &= \langle f_x(x, y, z), f_y(x, y, z), f_z(x, y, z) \rangle\end{aligned}$$

By abusing notation, we can define (in dimension 3)

$$\nabla := \left\langle \frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z} \right\rangle$$

from which we get

$$\nabla f = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right\rangle$$

(This mnemonic will really pay off when we get to divergence and curl, in Section 6.5.)

Formula for directional derivative, using the gradient

$$\begin{aligned}D_{\mathbf{u}}f &= \nabla f \cdot \vec{u} \\ D_{\mathbf{u}}f(a, b) &= \nabla f(a, b) \cdot \vec{u}\end{aligned}$$

Example (Similar to Exercise 281). Calculate $\nabla f(3, -2, 4)$, where

$$f(x, y, z) = ze^{2x+3y}$$

Solution. The gradient is

$$\nabla f = \langle 2ze^{2x+3y}, 3ze^{2x+3y}, e^{2x+3y} \rangle$$

so

$$\nabla f(3, -2, 4) = \langle 2(4)e^0, 3(4)e^0, e^0 \rangle = \langle 8, 12, 1 \rangle$$

Example (Similar to Exercise 266, 272). Let $f(x, y) = xe^y$, $P = (2, -1)$, and $\mathbf{v} = \langle 2, 3 \rangle$. Calculate the directional derivative in the direction of $\mathbf{v}$.

Solution. First note that $\mathbf{v}$ is not a unit vector. So we replace it with the vector

$$\mathbf{u} = \frac{\mathbf{v}}{\|\mathbf{v}\|} = \frac{\langle 2, 3 \rangle}{\sqrt{13}} = \left\langle \frac{2}{\sqrt{13}}, \frac{3}{\sqrt{13}} \right\rangle$$

Then

$$\begin{aligned} D_{\mathbf{u}}f(P) &= \nabla f(2, -1) \cdot \mathbf{u} \\ &= \left. \langle e^y, xe^y \rangle \right|_{(2, -1)} \cdot \mathbf{u} \\ &= \langle e^{-1}, 2e^{-1} \rangle \cdot \left\langle \frac{2}{\sqrt{13}}, \frac{3}{\sqrt{13}} \right\rangle = \boxed{\frac{8}{e\sqrt{13}}} \end{aligned}$$

3 Properties of the Gradient

3.1 Gradient and Directional Derivative

Using the dot product-cosine angle formula $\mathbf{a} \cdot \mathbf{b} = \|\mathbf{a}\| \|\mathbf{b}\| \cos \theta$, we can get

$$D_{\mathbf{u}}f(x_0, y_0) = \nabla f(x_0, y_0) \cdot \mathbf{u} = \|\nabla f(x_0, y_0)\| \|\mathbf{u}\| \cos \theta = \|\nabla f(x_0, y_0)\| \cos \theta$$

(θ is the angle between ∇f and $\mathbf{u}$). We can conclude that

- The directional derivative at a point (x_0, y_0) is maximized when $\mathbf{u}$ is pointing in the *same* direction as $\nabla f(x_0, y_0)$.
 - The maximum value of $D_{\mathbf{u}}f(x_0, y_0)$ is $\|\nabla f(x_0, y_0)\|$
 - Another way to phrase this: The gradient vector $\nabla f(P)$ points in the direction of steepest ascent. This maximum rate of ascent is $\|\nabla f(P)\|$
- The directional derivative at a point (x_0, y_0) is minimized when $\mathbf{u}$ is pointing in the *opposite* direction as $\nabla f(x_0, y_0)$.
 - The minimum value of $D_{\mathbf{u}}f(x_0, y_0)$ is $-\|\nabla f(x_0, y_0)\|$

$$-1 \leq \cos \theta \leq 1$$



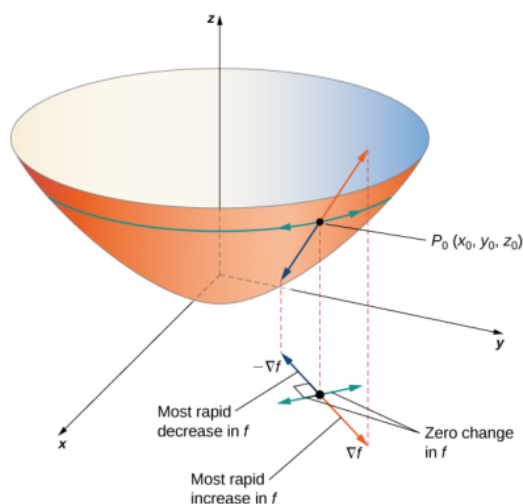


Figure 4.41 The gradient indicates the maximum and minimum values of the directional derivative at a point.

Example (Similar to Exercise 295). Let $f(x, y) = x^4y^{-2}$ and $P = (2, 1)$. Find the unit vector that points in the direction of maximum rate of increase at P and determine that maximum rate.

Solution. The gradient points in the direction of maximum rate of increase, so we evaluate the gradient at P :

$$\vec{\nabla} f = \langle 4x^3y^{-2}, -2x^4y^{-3} \rangle, \quad \vec{\nabla} f(2, 1) = \langle 32, -32 \rangle$$

The unit vector in this direction is

$$\mathbf{u} = \frac{\langle 32, -32 \rangle}{\|\langle 32, -32 \rangle\|} = \frac{\langle 32, -32 \rangle}{32\sqrt{2}} = \left\langle \frac{\sqrt{2}}{2}, -\frac{\sqrt{2}}{2} \right\rangle$$

The maximum rate, which is the rate in this direction, is given by

$$\|\vec{\nabla} f(2, 1)\| = \sqrt{32^2 + (-32)^2} = \boxed{32\sqrt{2}}$$

3.2 Gradient and Level Curves

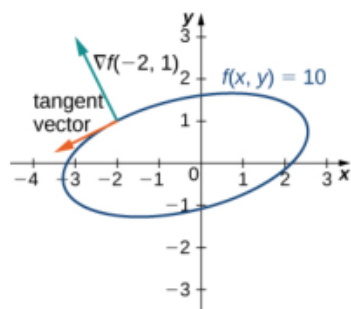


Figure 4.45 Tangent and normal vectors to $2x^2 - 3xy + 8y^2 + 2x - 4y + 4 = 18$ at point $(-2, 1)$.

The gradient is normal to level curves.

This property is the genesis for the Lagrange Multiplier method (Section 4.8).