

MAT124 MATHEMATICS II

Chain Rule, Gradients and Directional Derivatives

Chain Rule

Linear Approximations, Differentiability and Differentials

Gradients and Directional Derivatives

Chain Rule

The Chain Rule

$$z = f(u(t), v(t)) = g(t),$$

$$g'(t) = \lim_{h \rightarrow 0} \frac{g(t+h) - g(t)}{h} = \lim_{h \rightarrow 0} \frac{f(u(t+h), v(t+h)) - f(u(t), v(t))}{h}$$

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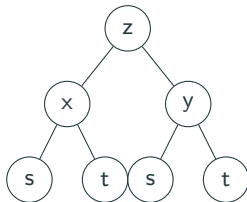
$$\frac{dz}{dt} = \frac{\partial z}{\partial x} \frac{dx}{dt} + \frac{\partial z}{\partial y} \frac{dy}{dt}.$$

The Chain Rule

If z is a function of x and y with continuous first partial derivatives, and if x and y depend on s and t , then

$$\frac{\partial z}{\partial s} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial s},$$

$$\frac{\partial z}{\partial t} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial t}.$$



The Chain Rule

EXAMPLE:

If $z = \sin(x^2y)$, where $x = st^2$ and $y = s^2 + \frac{1}{t}$, find $\frac{\partial z}{\partial s}$ and $\frac{\partial z}{\partial t}$

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$$\begin{aligned}\frac{\partial z}{\partial s} &= (2xy \cos(x^2y))t^2 + (x^2 \cos(x^2y))2s \\ &= \left(2st^2 \left(s^2 + \frac{1}{t}\right)t^2 + 2s^3t^4\right) \cos(s^4t^4 + s^2t^3) \\ &= (4s^3t^4 + 2st^3) \cos(s^4t^4 + s^2t^3)\end{aligned}$$

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$$\begin{aligned}\frac{\partial z}{\partial t} &= (2xy \cos(x^2y))2st + (x^2 \cos(x^2y)) \left(-\frac{1}{t^2}\right) \\ &= \left(2st^2 \left(s^2 + \frac{1}{t}\right) 2st + s^2t^4 \left(-\frac{1}{t^2}\right)\right) \cos(s^4t^4 + s^2t^3) \\ &= (4s^4t^3 + 3s^2t^2) \cos(s^4t^4 + s^2t^3).\end{aligned}$$

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EXAMPLE:

Find $\frac{\partial}{\partial x} f(x^2y, x + 2y)$ and $\frac{\partial}{\partial y} f(x^2y, x + 2y)$ in terms of the partial derivatives of f , assuming that these partial derivatives are continuous.

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Solution:

$$\begin{aligned} & \frac{\partial}{\partial x} f(x^2y, x + 2y) \\ &= f_1(x^2y, x + 2y) \frac{\partial}{\partial x} (x^2y) + f_2(x^2y, x + 2y) \frac{\partial}{\partial x} (x + 2y) \\ &= 2xyf_1(x^2y, x + 2y) + f_2(x^2y, x + 2y), \end{aligned}$$

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$$\begin{aligned}\frac{\partial}{\partial y} f(x^2y, x + 2y) &= f_1(x^2y, x + 2y) \frac{\partial}{\partial y} (x^2y) + f_2(x^2y, x + 2y) \frac{\partial}{\partial y} (x + 2y) \\ &= x^2 f_1(x^2y, x + 2y) + 2f_2(x^2y, x + 2y).\end{aligned}$$

The Chain Rule

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Express the partial derivatives of $z = h(s, t) = f(g(s, t))$ in terms of the derivative f' of f and the partial derivatives of g .

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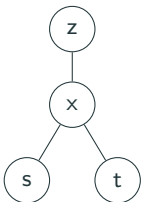
Solution: The partial derivatives of h can be calculated using the single-variable version of the Chain Rule: if $x = g(s, t)$, then $z = f(x)$ and

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$$h_1(s, t) = \frac{\partial z}{\partial s} = \frac{dz}{dx} \frac{\partial x}{\partial s} = f'(g(s, t))g_1(s, t),$$

$$h_2(s, t) = \frac{\partial z}{\partial t} = \frac{dz}{dx} \frac{\partial x}{\partial t} = f'(g(s, t))g_2(s, t).$$

The Chain Rule

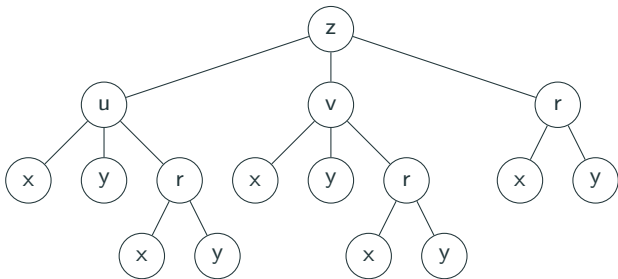
EXAMPLE:

Write the appropriate version of the Chain Rule for $\frac{\partial z}{\partial x}$, where z depends on u, v , and r ; u and v depend on x, y , and r ; and r depends on x and y .

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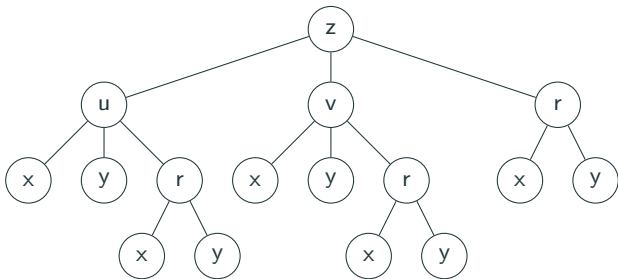
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$$\frac{\partial z}{\partial x} = \frac{\partial z}{\partial u} \frac{\partial u}{\partial x} + \frac{\partial z}{\partial u} \frac{\partial u}{\partial r} \frac{\partial r}{\partial x} + \frac{\partial z}{\partial v} \frac{\partial v}{\partial x} + \frac{\partial z}{\partial v} \frac{\partial v}{\partial r} \frac{\partial r}{\partial x} + \frac{\partial z}{\partial r} \frac{\partial r}{\partial x}.$$

The Chain Rule

Homogeneous Functions

A function $f(x_1, \dots, x_n)$ is said to be **positively homogeneous of degree k** if, for every point (x_1, x_2, \dots, x_n) in its domain and every real number $t > 0$, we have

$$f(tx_1, tx_2, \dots, tx_n) = t^k f(x_1, \dots, x_n).$$

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EXAMPLES

- $f(x, y) = x^2 + xy - y^2$ is positively homogeneous of degree 2,

The Chain Rule

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- $f(x, y) = \frac{2xy}{x^2 + y^2}$ is positively homogeneous of degree 0
- $f(x, y, z) = \frac{x - y + 5z}{yz - z^2}$ is positively homogeneous of degree -1,

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- $f(x, y) = x^2 + xy - y^2$ is positively homogeneous of degree 2,
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- $f(x, y, z) = \frac{x - y + 5z}{yz - z^2}$ is positively homogeneous of degree -1,
- $f(x, y) = x^2 + y$ is not positively homogeneous.

The Chain Rule

Homogeneous Functions

Euler's Theorem

If $f(x_1, \dots, x_n)$ has continuous first partial derivatives and is positively homogeneous of degree k , then

$$\sum_{i=1}^n x_i f_i(x_1, \dots, x_n) = k f(x_1, \dots, x_n).$$

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Proof: Differentiate the equation $f(tx_1, tx_2, \dots, tx_n) = t^k f(x_1, \dots, x_n)$ with respect to t to get

$$\begin{aligned} x_1 f_1(tx_1, \dots, tx_n) + x_2 f_2(tx_1, \dots, tx_n) + \dots + x_n f_n(tx_1, \dots, tx_n) \\ = kt^{k-1} f(x_1, \dots, x_n). \end{aligned}$$

Now substitute $t = 1$ to get the desired result.

The Chain Rule

Second order partial derivatives

EXAMPLE:

Calculate $\frac{\partial^2}{\partial x \partial y} f(x^2 - y^2, xy)$ in terms of partial derivatives of the function f . Assume that the second-order partials of f are continuous.

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$$\frac{\partial^2}{\partial x \partial y} f(u, v), \quad \text{where } u = x^2 - y^2 \quad \text{and} \quad v = xy.$$

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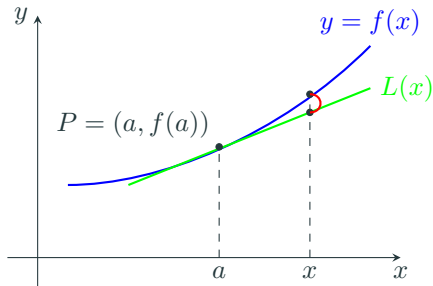
$$\frac{\partial^2}{\partial x \partial y} f(u, v), \quad \text{where } u = x^2 - y^2 \quad \text{and} \quad v = xy.$$

$$\frac{\partial}{\partial y} f(u, v) = -2yf_1(u, v) + xf_2(u, v).$$

$$\begin{aligned} \frac{\partial^2}{\partial x \partial y} f(u, v) &= -2y(2xf_{11}(u, v) + yf_{12}(u, v)) \\ &\quad + f_2(u, v) + x(2xf_{21}(u, v) + yf_{22}(u, v)) \\ &= f_2(u, v) - 4xyf_{11}(u, v) + 2(x^2 - y^2)f_{12}(u, v) + xyf_{22}(u, v). \end{aligned}$$

Linear Approximations, Differentiability and Differentials

Linear Approximations



$$f(x) \approx L(x) = f(a) + f'(a)(x - a)$$

Here, $L(x)$ is the **linearization** of f at a ; its graph is the tangent line to $y = f(x)$ at $x = a$.

Linear Approximations

Similarly, the tangent plane to the graph of $z = f(x, y)$ at (a, b) is

$$z = L(x, y) = f(a, b) + f_1(a, b)(x - a) + f_2(a, b)(y - b)$$

which is the **linearization** of f at (a, b) . We can use $L(x, y)$ to approximate values of $f(x, y)$ near (a, b) :

$$f(x, y) \approx L(x, y) = f(a, b) + f_1(a, b)(x - a) + f_2(a, b)(y - b).$$

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EXAMPLE

Find an approximate value for $f(x, y) = \sqrt{2x^2 + e^{2y}}$ at $(2.2, -0.2)$.

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Solution: It is convenient to use the linearization at $(2, 0)$, where the values of f and its partials are easily evaluated:

$$f_1(x, y) = \frac{2x}{\sqrt{2x^2 + e^{2y}}},$$

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$$f_1(2, 0) = \frac{4}{3},$$

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$$f_1(2, 0) = \frac{4}{3},$$

$$f_2(2, 0) = \frac{1}{3}.$$

Thus,

$$L(x, y) = 3 + \frac{4}{3}(x - 2) + \frac{1}{3}(y - 0)$$

and

$$f(2.2, -0.2) \approx L(2.2, -0.2) = 3 + \frac{4}{3}(2.2 - 2) + \frac{1}{3}(-0.2 - 0) = 3.2.$$

Differentiability

Definition

We say that the function $f(x, y)$ is **differentiable** at the point (a, b) if

$$\lim_{(h,k) \rightarrow (0,0)} \frac{f(a+h, b+k) - f(a, b) - hf_1(a, b) - kf_2(a, b)}{\sqrt{h^2 + k^2}} = 0.$$

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The function $f(x, y)$ is differentiable at the point (a, b) if and only if the surface $z = f(x, y)$ has a *nonvertical tangent plane* at (a, b) . This implies that $f_1(a, b)$ and $f_2(a, b)$ must exist and that f must be continuous at (a, b) .

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differentiability \Rightarrow *continuity*

Differentiability

Theorem

If f_1 and f_2 are continuous in a neighbourhood of the point (a, b) , then f is differentiable at (a, b) .

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$$\begin{aligned} & f(x + h, y + k) - f(x, y) - f_1(x, y)h - f_2(x, y)k \\ &= (x + h)^3 + (x + h)(y + k)^2 - x^3 - xy^2 - (3x^2 + y^2)h - 2xyk \\ &= 3xh^2 + h^3 + 2yhk + hk^2 + xk^2. \end{aligned}$$

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$$\lim_{(h,k) \rightarrow (0,0)} \frac{3xh^2 + h^3 + 2yhk + hk^2 + xk^2}{\sqrt{h^2 + k^2}} = 0.$$

Differentials

If the first partial derivatives of a function $z = f(x_1, \dots, x_n)$ exist at a point, we may construct a **differential** dz or df of the function at that point in a manner similar to that used for functions of one variable:

$$\begin{aligned} dz = df &= \frac{\partial z}{\partial x_1} dx_1 + \frac{\partial z}{\partial x_2} dx_2 + \cdots + \frac{\partial z}{\partial x_n} dx_n \\ &= f_1(x_1, \dots, x_n) dx_1 + \cdots + f_n(x_1, \dots, x_n) dx_n. \end{aligned}$$

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For a *differentiable* function f , the differential df is an approximation to the change Δf in value of the function given by

$$\Delta f = f(x_1 + dx_1, \dots, x_n + dx_n) - f(x_1, \dots, x_n).$$

Differentials

If the first partial derivatives of a function $z = f(x_1, \dots, x_n)$ exist at a point, we may construct a **differential** dz or df of the function at that point in a manner similar to that used for functions of one variable:

$$\begin{aligned} dz = df &= \frac{\partial z}{\partial x_1} dx_1 + \frac{\partial z}{\partial x_2} dx_2 + \cdots + \frac{\partial z}{\partial x_n} dx_n \\ &= f_1(x_1, \dots, x_n) dx_1 + \cdots + f_n(x_1, \dots, x_n) dx_n. \end{aligned}$$

For a *differentiable* function f , the differential df is an approximation to the change Δf in value of the function given by

$$\Delta f = f(x_1 + dx_1, \dots, x_n + dx_n) - f(x_1, \dots, x_n).$$

$$\frac{\Delta f - df}{\sqrt{(dx_1)^2 + \cdots + (dx_n)^2}} \rightarrow 0 \quad \text{if all } dx_i \rightarrow 0, \quad (1 \leq i \leq n).$$

Differentials

EXAMPLE

Estimate the percentage change in the period $T = 2\pi\sqrt{\frac{L}{g}}$ of a simple pendulum if the length, L , of the pendulum increases by 2% and the acceleration of gravity, g , decreases by 0.6%.

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We are given that $dL = \frac{2}{100}L$ and $dg = -\frac{6}{1,000}g$. Thus,

$$dT = \frac{1}{100}2\pi\sqrt{\frac{L}{g}} - \left(-\frac{6}{1,000}\right)2\pi\frac{1}{2}\sqrt{\frac{L}{g}} = \frac{13}{1,000}T.$$

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Therefore, the period T of the pendulum increases by **1.3%**.

Gradients and Directional Derivatives

Gradients

Gradient

At any point (x, y) where the first partial derivatives of the function $f(x, y)$ exist, we define the **gradient vector** $\nabla f(x, y) = \mathbf{grad} f(x, y)$ by

$$\nabla f(x, y) = \mathbf{grad} f(x, y) = f_1(x, y)\mathbf{i} + f_2(x, y)\mathbf{j}.$$

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For the basis vector \mathbf{i} and \mathbf{j} in \mathbb{R}^2 , the symbol ∇ , called **del** or **nabla**, is a *vector differential operator*:

$$\nabla = \mathbf{i} \frac{\partial}{\partial x} + \mathbf{j} \frac{\partial}{\partial y}.$$

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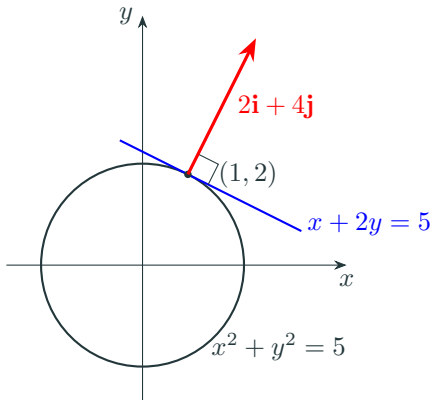
We can *apply* this operator to a function $f(x, y)$ by writing the operator to the left of the function. The result is the gradient of the function

$$\nabla f(x, y) = \left(\mathbf{i} \frac{\partial}{\partial x} + \mathbf{j} \frac{\partial}{\partial y} \right) f(x, y) = f_1(x, y)\mathbf{i} + f_2(x, y)\mathbf{j}.$$

Gradient

EXAMPLE

If $f(x, y) = x^2 + y^2$, then $\nabla f(x, y) = 2x\mathbf{i} + 2y\mathbf{j}$. In particular, $\nabla f(1, 2) = 2\mathbf{i} + 4\mathbf{j}$. Observe that this vector is perpendicular to the tangent line $x + 2y = 5$ to the circle $x^2 + y^2 = 5$ at $(1, 2)$. This circle is the level curve of f that passes through the point $(1, 2)$.



Gradient

Theorem

If $f(x, y)$ is differentiable at the point (a, b) and $\nabla f(a, b) \neq 0$, then $\nabla f(a, b)$ is a normal vector to the level curve of f that passes through (a, b) .

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Proof: Let $\mathbf{r} = \mathbf{r}(t) = x(t)\mathbf{i} + y(t)\mathbf{j}$ be a parametrization of the level curve of f such that $x(0) = a$ and $y(0) = b$. Then for all t near 0, $f(x(t), y(t)) = f(a, b)$. Differentiating this equation with respect to t using the Chain Rule, we obtain

$$f_1(x(t), y(t)) \frac{dx}{dt} + f_2(x(t), y(t)) \frac{dy}{dt} = 0.$$

At $t = 0$ this says that

$$\nabla f(a, b) \cdot \left. \frac{d\mathbf{r}}{dt} \right|_{t=0} = 0;$$

that is, ∇f is perpendicular to the tangent vector $d\mathbf{r}/dt$ to the level curve at (a, b) .

Directional Derivative

Directional Derivative

Let $\mathbf{u} = u\mathbf{i} + v\mathbf{j}$ be a unit vector, so that $u^2 + v^2 = 1$. The **directional derivative** of $f(x, y)$ at (a, b) in the direction of \mathbf{u} is the rate of change of $f(x, y)$ with respect to distance measured at (a, b) along a ray in the direction of \mathbf{u} in the xy -plane. This directional derivative is given by

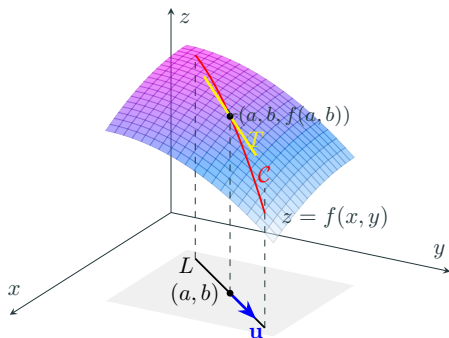
$$D_{\mathbf{u}}f(a, b) = \lim_{h \rightarrow 0^+} \frac{f(a + hu, b + hv) - f(a, b)}{h}.$$

It is also given by

$$D_{\mathbf{u}}f(a, b) = \left. \frac{d}{dt} f(a + tu, b + tv) \right|_{t=0}$$

if the derivative on the right-hand side exists.

Directional Derivative



Unit vector \mathbf{u} determines a line L through (a, b) in the domain of f . The vertical plane containing L intersects the graph of f in a curve C whose tangent T at $(a, b, f(a, b))$ has slope $D_{\mathbf{u}}f(a, b)$.

Directional Derivative

Theorem - Using the gradient to find directional derivatives

If f is differentiable at (a, b) and $\mathbf{u} = u\mathbf{i} + v\mathbf{j}$ is a unit vector, then the directional derivative of f at (a, b) in the direction of \mathbf{u} is given by

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

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Proof: By the Chain Rule:

$$\begin{aligned} D_{\mathbf{u}}f(a, b) &= \left. \frac{d}{dt} f(a + tu, b + tv) \right|_{t=0} \\ &= uf_1(a, b) + vf_2(a, b) = \mathbf{u} \cdot \nabla f(a, b). \end{aligned}$$

Directional Derivative

Given any nonzero vector \mathbf{v} , we can always obtain a unit vector in the same direction by dividing \mathbf{v} by its length. The directional derivative of f at (a, b) in the direction of \mathbf{v} is therefore given by

$$D_{\mathbf{v}/|\mathbf{v}|}f(a, b) = \frac{\mathbf{v}}{|\mathbf{v}|} \cdot \nabla f(a, b).$$

Directional Derivative Examples

EXAMPLE

Find the rate of change of $f(x, y) = y^4 + 2xy^3 + x^2y^2$ at $(0, 1)$ measured in each of the following directions:

(a) $\mathbf{i} + 2\mathbf{j}$, (b) $\mathbf{j} - 2\mathbf{i}$, (c) $3\mathbf{i}$, (d) $\mathbf{i} + \mathbf{j}$.

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Solution: We calculate

$$\nabla f(x, y) = (2y^3 + 2xy^2)\mathbf{i} + (4y^3 + 6xy^2 + 2x^2y)\mathbf{j},$$

$$\nabla f(0, 1) = 2\mathbf{i} + 4\mathbf{j}.$$

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(a) The unit vector in the direction of $\mathbf{i} + 2\mathbf{j}$ is $\frac{\mathbf{i} + 2\mathbf{j}}{\sqrt{5}}$. Thus, the directional derivative of f at $(0, 1)$ in that direction is

$$\frac{\mathbf{i} + 2\mathbf{j}}{\sqrt{5}} \cdot (2\mathbf{i} + 4\mathbf{j}) = \frac{2 + 8}{\sqrt{5}} = \frac{10}{\sqrt{5}} = 2\sqrt{5}.$$

Observe that $\mathbf{i} + 2\mathbf{j}$ points in the same direction as $\nabla f(0, 1)$ so the directional derivative is positive and equal to the length of $\nabla f(0, 1)$.

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Solution: (b) The unit vector in the direction of $\mathbf{j} - 2\mathbf{i}$ is $\frac{\mathbf{j} - 2\mathbf{i}}{\sqrt{5}}$. Thus, the directional derivative of f at $(0, 1)$ in that direction is

$$\frac{-2\mathbf{i} + \mathbf{j}}{\sqrt{5}} \cdot (2\mathbf{i} + 4\mathbf{j}) = \frac{-4 + 4}{\sqrt{5}} = 0.$$

Since $\mathbf{j} - 2\mathbf{i}$ is perpendicular to $\nabla f(0, 1)$, it is tangent to the level curve of f through $(0, 1)$, so the directional derivative in that direction is zero.

Directional Derivative Examples

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(a) $\mathbf{i} + 2\mathbf{j}$, (b) $\mathbf{j} - 2\mathbf{i}$, (c) $3\mathbf{i}$, (d) $\mathbf{i} + \mathbf{j}$.

Solution: (c) The unit vector in the direction of $3\mathbf{i}$ is just \mathbf{i} , so the directional derivative of f at $(0, 1)$ in that direction is

$$\mathbf{i} \cdot (2\mathbf{i} + 4\mathbf{j}) = 2.$$

As noted previously, the directional derivative of f in the direction of the positive x -axis is just $f_1(0, 1)$.

Directional Derivative Examples

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Find the rate of change of $f(x, y) = y^4 + 2xy^3 + x^2y^2$ at $(0, 1)$ measured in each of the following directions:

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Solution: (d) The unit vector in the direction of $\mathbf{i} + \mathbf{j}$ is $\frac{\mathbf{i} + \mathbf{j}}{\sqrt{2}}$, so the directional derivative of f at $(0, 1)$ in that direction is

$$\frac{\mathbf{i} + \mathbf{j}}{\sqrt{2}} \cdot (2\mathbf{i} + 4\mathbf{j}) = \frac{2 + 4}{\sqrt{2}} = 3\sqrt{2}.$$

If we move along the surface $z = f(x, y)$ through the point $(0, 1, 1)$ in a direction making horizontal angles of 45° with the positive directions of the x - and y -axes, we would be rising at a rate of $3\sqrt{2}$ vertical units per horizontal unit moved.

Directional Derivative

For any unit vector \mathbf{u} we have

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

where θ is the angle between the vectors \mathbf{u} and $\nabla f(a, b)$.

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- The directional derivative is zero in the direction $\theta = \pi/2$; this is the direction of the (tangent line to the) level curve of f through (a, b) .

Geometric Properties of the Gradient Vector

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- At (a, b) , $f(x, y)$ increases most rapidly in the direction of the gradient vector $\nabla f(a, b)$. The maximum rate of increase is $|\nabla f(a, b)|$.

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- The rate of change of $f(x, y)$ at (a, b) is zero in directions tangent to the level curve of f that passes through (a, b) .

Directional Derivative Example

EXAMPLE

The temperature at position (x, y) in a region of the xy -plane is T °C, where

$$T(x, y) = x^2 e^{-y}.$$

In what direction at the point $(2, 1)$ does the temperature increase most rapidly? What is the rate of increase of f in that direction?

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Solution: We have

$$\begin{aligned}\nabla T(x, y) &= 2xe^{-y}\mathbf{i} - x^2e^{-y}\mathbf{j}, \\ \nabla T(2, 1) &= \frac{4}{e}\mathbf{i} - \frac{4}{e}\mathbf{j} = \frac{4}{e}(\mathbf{i} - \mathbf{j}).\end{aligned}$$

At $(2, 1)$, $T(x, y)$ increases most rapidly in the direction of the vector $\mathbf{i} - \mathbf{j}$. The rate of increase in this direction is

$$|\nabla T(2, 1)| = 4\sqrt{2}/e \text{ °C/unit distance.}$$

Directional Derivative Example

EXAMPLE

A hiker is standing beside a stream on the side of a mountain, examining her map of the region. The height of land (in metres) at any point (x, y) is given by the function

$$h(x, y) = \frac{20,000}{3 + x^2 + 2y^2},$$

where x and y (in kilometres) denote the coordinates of the point on the hiker's map. The hiker is at the point $(3, 2)$.

- What is the direction of flow of the stream at $(3, 2)$ on the hiker's map? How fast is the stream descending at her location?
- Find the equation of the path of the stream on the hiker's map.
- At what angle to the path of the stream (on the map) should the hiker set out if she wishes to climb at a 15° inclination to the horizontal?
- Make a sketch of the hiker's map, showing some curves of constant elevation, and showing the stream.

Directional Derivative Example

Solution: (a) We begin by calculating the gradient of h and its length at $(3, 2)$:

$$\nabla h(x, y) = -\frac{20,000}{(3 + x^2 + 2y^2)^2} (2x\mathbf{i} + 4y\mathbf{j}),$$

$$\nabla h(3, 2) = -100(3\mathbf{i} + 4\mathbf{j}),$$

$$|\nabla h(3, 2)| = 500.$$

The stream is flowing in the direction whose horizontal projection at $(3, 2)$ is $-\nabla h(3, 2)$, that is, in the horizontal direction of the vector $3\mathbf{i} + 4\mathbf{j}$. The stream is descending at a rate of 500 m/km, that is, 0.5 m per horizontal metre travelled.

Directional Derivative Example

Solution: (b) If the vector $d\mathbf{r} = dx\mathbf{i} + dy\mathbf{j}$ is tangent to the path of the stream at point (x, y) on the map, then $d\mathbf{r}$ is parallel to $\nabla h(x, y)$. Hence, the components of these two vectors are proportional:

$$\frac{dx}{2x} = \frac{dy}{4y} \quad \text{or} \quad \frac{dy}{y} = \frac{2dx}{x}.$$

Integrating both sides of this equation, we get $\ln y = 2 \ln x + \ln C$, or $y = Cx^2$. Since the path of the stream passes through $(3, 2)$, we have $C = 2/9$ and the equation is $9y = 2x^2$.

Directional Derivative Example

Solution: (c) Suppose the hiker moves away from $(3, 2)$ in the direction of the unit vector \mathbf{u} . She will be ascending at an inclination of 15° if the directional derivative of h in the direction of \mathbf{u} is $1,000 \tan 15^\circ \approx 268$. (The 1,000 compensates for the fact that the vertical units are metres while the horizontal units are kilometres.) If θ is the angle between \mathbf{u} and the upstream direction, then

$$500 \cos \theta = |\nabla h(3, 2)| \cos \theta = D_{\mathbf{u}} h(3, 2) \approx 268.$$

Hence, $\cos \theta \approx 0.536$ and $\theta \approx 57.6^\circ$. She should set out in a direction making a horizontal angle of about 58° with the upstream direction.

Directional Derivative Example

Solution: (d)

