

Differential Vector Calculus

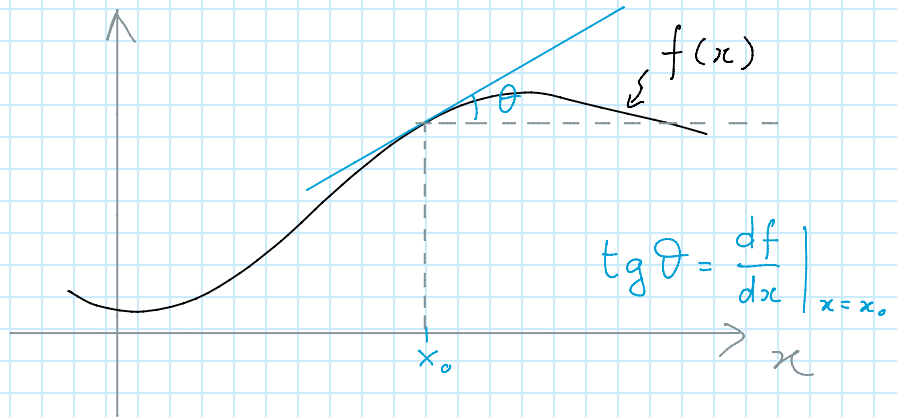
Friday, January 4, 2019 6:12 AM

Ordinary derivatives

A derivative is a measure of how rapidly a function f changes if the argument x changes by an infinitesimal amount dx

$$df = \frac{df}{dx} dx$$

The derivative at a given point is the slope of f at that point.



Gradient

Definition : $f(x, y, z)$ scalar function

$$\nabla f = \frac{\partial f}{\partial x} \hat{i} + \frac{\partial f}{\partial y} \hat{j} + \frac{\partial f}{\partial z} \hat{k}$$

$$= \left(\frac{\partial}{\partial x} \hat{i} + \frac{\partial}{\partial y} \hat{j} + \frac{\partial}{\partial z} \hat{k} \right) f(x, y, z)$$

$\equiv \nabla$ "del" operator

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The gradient is a vector.

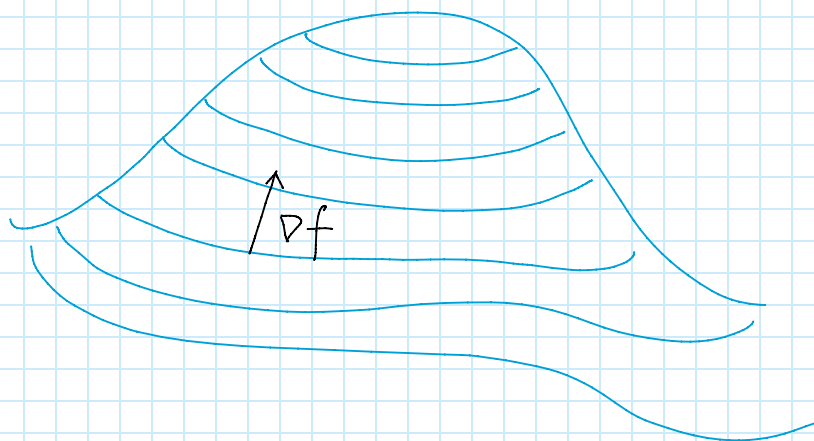
Geometrical interpretation of the gradient

Consider the temperature in a room as a function of the position in the room. The temperature is then a scalar function of three variables $T(x,y,z)$. An infinitesimal temperature change which results from an infinitesimal change in position can be written as

$$\begin{aligned} dT &= \left(\frac{\partial T}{\partial x} \right) dx + \left(\frac{\partial T}{\partial y} \right) dy + \left(\frac{\partial T}{\partial z} \right) dz \\ &= (\nabla T) \cdot d\vec{\ell} = |\nabla T| |d\vec{\ell}| \cos \theta \end{aligned}$$

If one keeps the magnitude of the gradient fixed, the largest dT is obtained when the angle θ is zero. Therefore the gradient points in the direction of maximum increase of the function T .

A two dimensional example is easier to visualize



$$\nabla f = 0 \rightarrow \text{stationary point} \begin{cases} \text{max (summit)} \\ \text{min (valley)} \\ \text{saddle (pass)} \\ \text{point} \end{cases}$$

Divergence

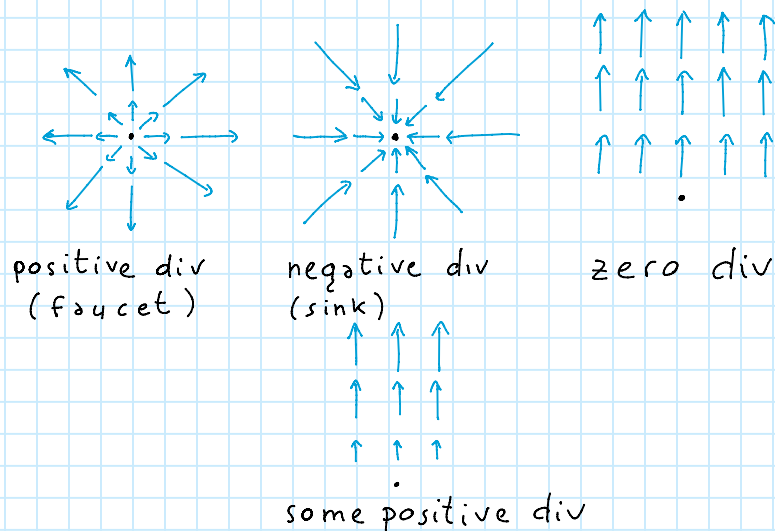
The divergence is a scalar obtained from the scalar product of a del operator with a vector

$$\begin{aligned} \nabla \cdot \vec{v} &= \left(\hat{i} \frac{\partial}{\partial x} + \hat{j} \frac{\partial}{\partial y} + \hat{k} \frac{\partial}{\partial z} \right) \cdot (v_x \hat{i} + v_y \hat{j} + v_z \hat{k}) \\ &= \frac{\partial v_x}{\partial x} + \frac{\partial v_y}{\partial y} + \frac{\partial v_z}{\partial z} \end{aligned}$$

Geometrical interpretation of a divergence

The divergence is a measure of how much a vector spreads out (i.e. diverges) from a given point.

Two dimensional examples



Curl

The curl is the cross product of the del operator with a vector function. The curl is a vector

$$\nabla \times \vec{v} = \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ \partial_x & \partial_y & \partial_z \\ v_x & v_y & v_z \end{vmatrix}$$

$$= \hat{i} (\partial_y v_z - \partial_z v_y) - \hat{j} (\partial_x v_z - \partial_z v_x) + \hat{k} (\partial_x v_y - \partial_y v_x)$$

In Einstein's notation

$$[\nabla \times \vec{v}]_i = \epsilon_{ijk} \partial_j v_k$$

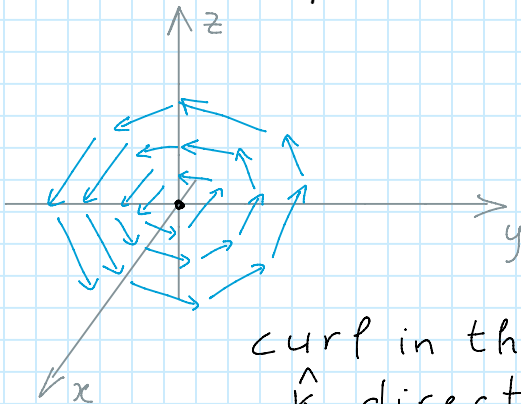
$$i, j, k \in \{1, 2, 3\}$$

rem there is a summation over repeated indices

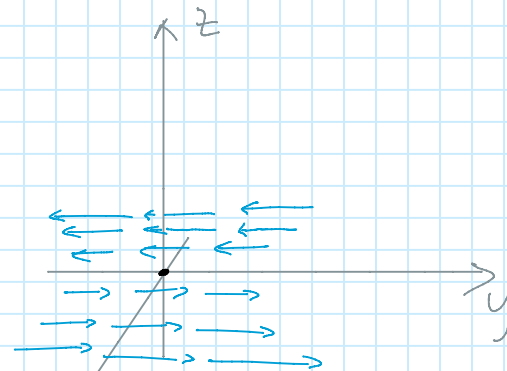
Geometrical interpretation of the curl

The curl is a measure of how much the vector v swirls around the point in which the curl is calculated.

Two dimensional examples



curl in the \hat{k} direction



(right hand rule)

Product rules

Given two objects, which can be either scalar or a vector, we have two ways to form a scalar and two ways to form a vector

$$\begin{array}{ll} \text{scalars} & fg, \quad \bar{A} \cdot \bar{B} \\ \text{vectors} & f\bar{A}, \quad \bar{A} \times \bar{B} \end{array}$$

One can then calculate the gradient of the two scalars

$$\nabla(fg) = f\nabla g + g\nabla f \quad (i)$$

$$\nabla(\bar{A} \cdot \bar{B}) = \bar{A} \times (\nabla \times \bar{B}) + \bar{B} \times (\nabla \times \bar{A}) + (\bar{A} \cdot \nabla)\bar{B} + (\bar{B} \cdot \nabla)\bar{A} \quad (ii)$$

One can also calculate the divergence of the two vectors

$$\nabla \cdot (f\bar{A}) = f(\nabla \cdot \bar{A}) + \bar{A} \cdot (\nabla f) \quad (iii)$$

$$\nabla \cdot (\bar{A} \times \bar{B}) = \bar{B} \cdot (\nabla \times \bar{A}) - \bar{A} \cdot (\nabla \times \bar{B}) \quad (iv)$$

Or the curl of the two vectors

$$\nabla \times (f\bar{A}) = f(\nabla \times \bar{A}) - \bar{A} \times (\nabla f) \quad (v)$$

$$\nabla \times (\bar{A} \times \bar{B}) = (\bar{B} \cdot \nabla)\bar{A} - (\bar{A} \cdot \nabla)\bar{B} + \bar{A}(\nabla \cdot \bar{B}) - \bar{B}(\nabla \cdot \bar{A}) \quad (vi)$$

Homework: Check the six relations above by using components and your preferred computer algebra system. Hint: It is often convenient to use components notation.

Ex

$$\nabla \cdot (f\bar{A}) = \partial_i (fA_i) = (\partial_i f)A_i + f\partial_i A_i = \nabla f \cdot \bar{A} + f\nabla \cdot \bar{A}$$

Second derivatives

By combining gradients, divergence and curl one can build five different kinds of second derivatives

- 1) Divergence of gradient $\nabla \cdot (\nabla f)$
- 2) Curl of a gradient $\nabla \times (\nabla f)$
- 3) Gradient of a divergence $\nabla (\nabla \cdot \vec{v})$
- 4) Divergence of a curl $\nabla \cdot (\nabla \times \vec{v})$
- 5) Curl of a curl $\nabla \times (\nabla \times \vec{v})$

Case 1)

$$\begin{aligned}\nabla \cdot (\nabla f) &= \left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} + \frac{\partial^2}{\partial z^2} \right) f && \text{LAPLACIAN} \\ & && \text{OPERATOR} \\ &\equiv \nabla^2 f = \Delta f\end{aligned}$$

The Laplacian can also be applied to a vector, with the understanding that the operator must be applied to each component of the vector

$$\Delta \vec{v} = (\Delta v_x) \hat{i} + (\Delta v_y) \hat{j} + (\Delta v_z) \hat{k}$$

Case 2)

$$\nabla \times (\nabla f) = 0 \quad \text{ALWAYS!}$$

To prove this just calculate

$$\nabla \times (\nabla f) = \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ \partial_x & \partial_y & \partial_z \\ \partial_x f & \partial_y f & \partial_z f \end{vmatrix}$$

Case 3) $\nabla (\nabla \cdot \vec{v})$

Seldom used, it is not the same as the Laplacian

Case 4)

$$\nabla \cdot (\nabla \times \vec{v}) = 0 \quad \text{ALWAYS!}$$

The relation above is easily proven by writing it out in components.

Case 5)

$$\nabla \times (\nabla \times \vec{v}) = \nabla (\nabla \cdot \vec{v}) - \Delta \vec{v}$$

in fact

$$\nabla \times (\nabla \times \vec{v}) = \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ \partial_x & \partial_y & \partial_z \\ \partial_y v_z - \partial_z v_y & \partial_z v_x - \partial_x v_z & \partial_x v_y - \partial_y v_x \end{vmatrix}$$

Consider now the x component

$$\begin{aligned} & \partial_y (\partial_x v_y - \partial_y v_x) - \partial_z (\partial_z v_x - \partial_x v_z) \\ = & \partial_y \partial_x v_y - \partial_y^2 v_x - \partial_z^2 v_x + \partial_z \partial_x v_z - \underbrace{\partial_x^2 v_x + \partial_x^2 v_x}_{\text{add and subtract}} \\ = & \partial_x (\partial_x v_x + \partial_y v_y + \partial_z v_z) - (\partial_x^2 + \partial_y^2 + \partial_z^2) v_x \\ = & \partial_x (\nabla \cdot \vec{v}) - \Delta v_x \end{aligned}$$

The other components follow a similar pattern