

Functions of Several Variables

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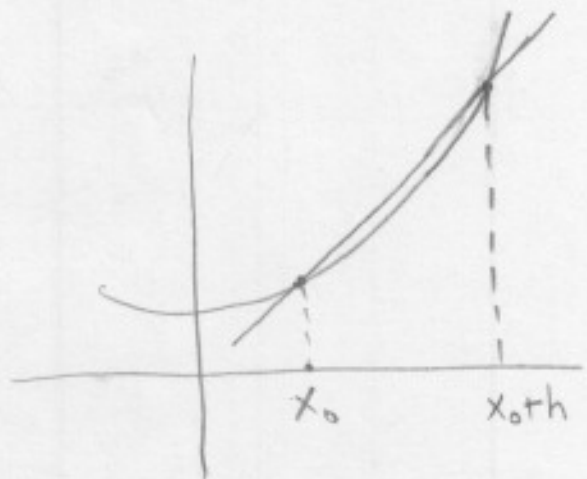
We first recall, for functions of one variable, we have:

$$f: (a, b) \rightarrow \mathbb{R}, \quad x_0 \in (a, b)$$

$$f'(x_0) = \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0}$$

or, equivalently,

$$f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h}, \quad x = x_0+h$$



We can rewrite the above limit as follows,

$$\lim_{h \rightarrow 0} \left[\frac{f(x_0+h) - f(x_0)}{h} - f'(x_0) \right] = 0$$

or

$$\lim_{h \rightarrow 0} \left| \frac{f(x_0+h) - f(x_0) - f'(x_0)h}{h} \right| = 0$$

Note that:

$$l(x) = f(x_0) + f'(x_0)h, \quad h = x - x_0$$
$$= f(x_0) + f'(x_0)(x - x_0)$$

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is the equation of the tangent line to the graph at x_0 .

We also have, with

$$r(h) := f(x_0 + h) - l(x_0 + h)$$

that:

$$\lim_{h \rightarrow 0} \frac{r(h)}{h} = 0$$

Remark: We can regard the derivative of f at x_0 , not as a real number, but as a linear operator:

$$\left. \begin{aligned} A: \mathbb{R} &\rightarrow \mathbb{R} \\ A(h) &= f'(x_0)h \end{aligned} \right\} (*)$$

From linear algebra, if X and Y are vector spaces, we define:

$$L(X, Y) = \{ A: X \rightarrow Y : A \text{ is linear} \}$$

Recall that A is linear if:

$$A(x_1 + x_2) = Ax_1 + Ax_2 \quad \forall x_1, x_2 \in X$$
$$A(cx) = cA(x), \quad \forall c \in \mathbb{R}, x \in X$$

Therefore, the linear operator in (*) belongs to $L(\mathbb{R}, \mathbb{R})$.

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Conversely, every linear function that carries \mathbb{R}^1 to \mathbb{R}^1 is multiplication by some real number.

For function of several variables we have the following:

Def 9.11: Let $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$, E open, $\vec{x}_0 \in E$.

If there exists a linear transformation $A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that:

$$\lim_{\vec{h} \rightarrow 0} \frac{|f(\vec{x}_0 + \vec{h}) - f(\vec{x}_0) - A\vec{h}|}{|\vec{h}|} = 0. \quad (1)$$

then we say that f is differentiable at \vec{x}_0 , and we write

$$f'(\vec{x}_0) = A$$

If f is differentiable at every $\vec{x} \in E$, we say that f is differentiable in E .

Theorem 9.12: Suppose E and f are as in Definition 9.11, $\vec{x} \in E$, and (1) holds with $A=A_1$ and with $A=A_2$. Then $A_1=A_2$.

Remarks :

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(a) The relation in (1) can be rewritten in the form

$$f(\vec{x}_0 + \vec{h}) - f(\vec{x}_0) = f'(\vec{x}_0)\vec{h} + r(\vec{h}), \quad (2)$$

where the remainder $r(\vec{h})$ satisfies:

$$\lim_{\vec{h} \rightarrow 0} \frac{|r(\vec{h})|}{|\vec{h}|} = 0.$$

Hence:

$$f(\vec{x}_0 + \vec{h}) = f(\vec{x}_0) + f'(\vec{x}_0)\vec{h} + r(\vec{h})$$

$$\Rightarrow \boxed{f(\vec{x}_0 + \vec{h}) \sim f(\vec{x}_0) + f'(\vec{x}_0)\vec{h}} \quad (**)$$

So we can approximate $f(\vec{x}_0 + \vec{h})$ with the function $f(\vec{x}_0) + f'(\vec{x}_0)\vec{h}$, where $f'(\vec{x}_0): \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a linear function; i.e., $f'(\vec{x}_0) \in L(\mathbb{R}^n, \mathbb{R}^m)$

(b) Suppose $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable in E . For every $x \in E$, $f'(x) \in L(\mathbb{R}^n, \mathbb{R}^m)$. Hence, f' can be considered as a function:

$$f': E \rightarrow L(\mathbb{R}^n, \mathbb{R}^m)$$

(c) From (2) clearly follows that if f is differentiable at \vec{x} then f is continuous at \vec{x} .

Definition 9.16 . Partial derivatives

Let $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$; i.e;

$$f(\vec{x}) = (f_1(\vec{x}), \dots, f_m(\vec{x}))$$

For $\vec{x} \in E$, $1 \leq i \leq m$, $1 \leq j \leq n$, we define:

$$D_j f_i(\vec{x}) = \lim_{t \rightarrow 0} \frac{f_i(\vec{x} + te_j) - f_i(\vec{x})}{t}$$

(or $\frac{\partial f_i}{\partial x_j}$)

where $\{e_1, \dots, e_n\}$ is the standard bases of \mathbb{R}^n , i.e., $e_j = (0, 0, \dots, \underset{\uparrow}{1}, 0, \dots, 0)$
 \uparrow j component

Theorem 9.17 : Suppose $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at $\vec{x} \in E$. Then, the partial derivatives exist; i.e;

$$\frac{\partial f_i}{\partial x_j} \text{ exist, } \forall i \in \{1, \dots, m\} \\ \forall j \in \{1, \dots, n\}$$

Moreover, the linear transformation $f'(\vec{x}): \mathbb{R}^n \rightarrow \mathbb{R}^m$ is the $m \times n$ matrix:

$$f'(\vec{x}) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(\vec{x}) & \dots & \frac{\partial f_1}{\partial x_n}(\vec{x}) \\ \frac{\partial f_2}{\partial x_1}(\vec{x}) & \dots & \frac{\partial f_2}{\partial x_n}(\vec{x}) \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1}(\vec{x}) & \dots & \frac{\partial f_m}{\partial x_n}(\vec{x}) \end{pmatrix}$$

Definition: A differentiable mapping:

$$f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$$

is said to be continuously differentiable in E if the function:

$$f': E \rightarrow L(\mathbb{R}^n, \mathbb{R}^m),$$

is a continuous mapping. Recall that $L(\mathbb{R}^n, \mathbb{R}^m)$ is the set of all linear maps from \mathbb{R}^n into \mathbb{R}^m . If $A \in L(\mathbb{R}^n, \mathbb{R}^m)$, the norm of A is defined as:

$$\|A\| := \sup_{|\vec{x}| \leq 1} |A\vec{x}|$$

We define $\mathcal{C}^1(E)$ to be the set of all differentiable functions $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that $f': E \rightarrow L(\mathbb{R}^n, \mathbb{R}^m)$ is continuous.

More precisely, if $f \in \mathcal{C}^1(E)$ we have that to every $\vec{x} \in E$ and to every $\varepsilon > 0$ corresponds a $\delta > 0$ such that

$$\text{if } \vec{y} \in E, |\vec{x} - \vec{y}| < \delta \text{ then } \|f'(\vec{y}) - f'(\vec{x})\| < \varepsilon.$$

Theorem 9.21: Let $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$. Then:

$f \in \mathcal{C}^1(E)$ if and only if the partial derivatives $D_j f_i$ exist and are continuous on E , for $1 \leq i \leq m$, $1 \leq j \leq n$.

Ex: Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}^3$

$$f(x, y) = (xe^{2y}, yx^2).$$

Compute $f(0.9, 0.2)$ without using calculator.

We have

$$f(x, y) = (f_1(x, y), f_2(x, y))$$

$$f_1(x, y) = xe^{2y}, \quad f_2(x, y) = yx^2.$$

Since $\frac{\partial f_1}{\partial x} = e^{2y}$, $\frac{\partial f_1}{\partial y} = 2xe^{2y}$, $\frac{\partial f_2}{\partial x} = 2xy$, $\frac{\partial f_2}{\partial y} = x^2$

are all continuous functions in \mathbb{R}^2 , Theorem 9.21 gives that $f \in \mathcal{C}^1(\mathbb{R}^2)$. In particular, f is differentiable at any point of \mathbb{R}^2 . We can then use (**) to estimate:

$$f(\vec{x}_0 + \vec{h}) \sim f(\vec{x}_0) + f'(\vec{x}_0)h$$

In this example we can take $\vec{x}_0 = (1, 0)$ and $\vec{h} = (0.9, 0.2) - (1, 0) = (-0.1, 0.2)$.

$$f'(\vec{x}_0) = \begin{pmatrix} \frac{\partial f_1}{\partial x}(\vec{x}_0) & \frac{\partial f_1}{\partial y}(\vec{x}_0) \\ \frac{\partial f_2}{\partial x}(\vec{x}_0) & \frac{\partial f_2}{\partial y}(\vec{x}_0) \end{pmatrix} = \begin{pmatrix} e^{2(0)} & 2(1)e^{2(0)} \\ 2(1)(0) & (1)^2 \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix}$$

$$f(1, 0) = (1, 0)$$

Therefore:

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$$\begin{aligned} f(0.9, 0.2) &\cong \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} -0.1 \\ 0.2 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} -0.1 + 0.4 \\ 0 + 0.2 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 0.3 \\ 0.2 \end{pmatrix} \\ &= \begin{pmatrix} 1.3 \\ 0.2 \end{pmatrix} \end{aligned}$$

$$\Rightarrow f(0.9, 0.2) \cong (1.3, 0.2)$$

The true value is:

$$\begin{aligned} f(0.9, 0.2) &= (0.9 e^{0.4}, 0.2 (0.9)^2) \\ &= (1.34\dots, 0.162\dots) \end{aligned}$$

• Remark; If $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ is

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differentiable, we have seen that $f'(\vec{x})$ is a linear map, i.e., $f'(\vec{x}) \in L(\mathbb{R}^n, \mathbb{R}^m)$.

Moreover, this linear map is unique and given by the $m \times n$ matrix:

$$f'(\vec{x}) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(\vec{x}) & \dots & \frac{\partial f_1}{\partial x_n}(\vec{x}) \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1}(\vec{x}) & \dots & \frac{\partial f_m}{\partial x_n}(\vec{x}) \end{pmatrix}$$

$$\vec{x} = (x_1, \dots, x_n) \quad f(x_1, \dots, x_n) = (f_1(x_1, \dots, x_n), \dots, f_m(x_1, \dots, x_n)).$$

For the particular case:

$$m=1, \quad n>1,$$

We have $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}$. In this case, $f'(\vec{x})$ is a $1 \times n$ matrix

$$f'(\vec{x}) = \left(\frac{\partial f}{\partial x_1}(\vec{x}) \quad \dots \quad \frac{\partial f}{\partial x_n}(\vec{x}) \right)$$

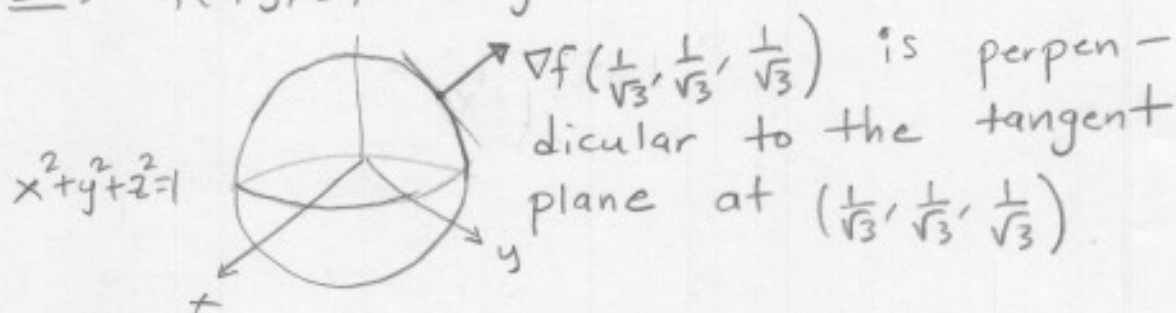
We can also consider this matrix as a vector, which is denoted as the gradient vector $\nabla f(\vec{x})$:

$$\nabla f(\vec{x}) = \left(\frac{\partial f}{\partial x_1}(\vec{x}), \frac{\partial f}{\partial x_2}(\vec{x}), \dots, \frac{\partial f}{\partial x_n}(\vec{x}) \right).$$

Besides the approximation property (**), which in this particular case means that we are approximating the function with the tangent plane, the gradient vector has two fundamental geometrical applications:

1. The gradient vector is perpendicular to the level surfaces (or level curves) of the function:

Ex: $f(x, y, z) = x^2 + y^2 + z^2$



2. The gradient vector points in the direction of maximum increase of the function.

Indeed, if \vec{u} is a unit vector in \mathbb{R}^n , recall that the directional derivative of f at \vec{x} is defined as:

$$D_{\vec{u}} f(\vec{x}) = \lim_{t \rightarrow 0} \frac{f(\vec{x} + t\vec{u}) - f(\vec{x})}{t}$$

Other notation: $\frac{\partial f}{\partial \vec{u}}(\vec{x})$.

Theorem 9.15: If $f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $g: \mathbb{R}^m \rightarrow \mathbb{R}^p$ are both differentiable at \vec{x}_0 and $f(\vec{x}_0)$ respectively, then (chain rule in several variables) the composition function:

$$h(\vec{x}) = g(f(\vec{x})), \quad h: E \subset \mathbb{R}^n \rightarrow \mathbb{R}^p$$

is differentiable at \vec{x}_0 and:

$$h'(\vec{x}_0) = g'(f(\vec{x}_0)) \cdot f'(\vec{x}_0)$$

\uparrow \uparrow \uparrow
 $p \times n$ $p \times m$ matrix $m \times n$ matrix
matrix

Using the chain rule, if f is differentiable at \vec{x} , it is easy to show that:

$$D_{\vec{u}} f(\vec{x}) = \nabla f(\vec{x}) \cdot \vec{u}, \quad f: E \subset \mathbb{R}^n \rightarrow \mathbb{R}$$

Since $\nabla f(\vec{x}) \cdot \vec{u} = |\nabla f(\vec{x})| |\vec{u}| \cos \theta$, where θ is the angle between the two vectors, it follows that $D_{\vec{u}} f$ is maximized when the unit vector \vec{u} is given by:

$$\vec{u} = \frac{\nabla f(\vec{x})}{|\nabla f(\vec{x})|}, \quad |\vec{u}| = 1.$$

Ex: $n=2, m=1$

$f: \mathbb{R}^2 \rightarrow \mathbb{R}$ differentiable

From the definition of differentiability in (1) we obtained that f is approximated by (see (**));

$$f(\vec{x}_0 + \vec{h}) \cong f(\vec{x}_0) + f'(\vec{x}_0)\vec{h}, \quad \vec{x} = \vec{x}_0 + \vec{h}$$

In this case, with $n=2$, let us write:

$$\vec{x}_0 = (x_0, y_0), \quad \vec{x} = (x, y)$$

Hence:

$$f'(\vec{x}_0) = \left(\frac{\partial f}{\partial x}(x_0, y_0) \quad \frac{\partial f}{\partial y}(x_0, y_0) \right)$$

Since $\vec{h} = \vec{x} - \vec{x}_0 = (x, y) - (x_0, y_0)$ we have:

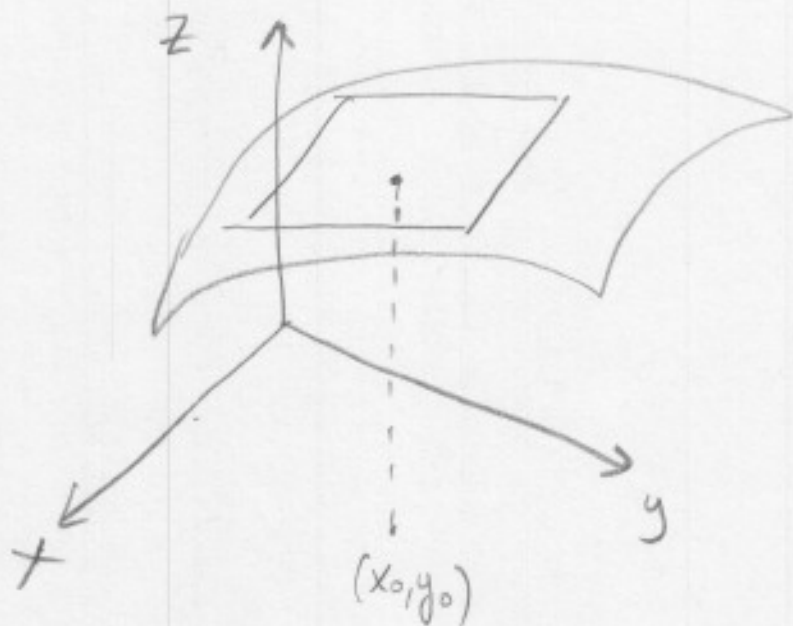
$$\begin{aligned} f'(\vec{x}_0)\vec{h} &= \left(\frac{\partial f}{\partial x}(x_0, y_0) \quad \frac{\partial f}{\partial y}(x_0, y_0) \right) \begin{pmatrix} x - x_0 \\ y - y_0 \end{pmatrix} \\ &= \frac{\partial f}{\partial x}(x_0, y_0)(x - x_0) + \frac{\partial f}{\partial y}(x_0, y_0)(y - y_0) \end{aligned}$$

Hence:

$$f(x, y) \cong f(x_0, y_0) + \frac{\partial f}{\partial x}(x_0, y_0)(x - x_0) + \frac{\partial f}{\partial y}(x_0, y_0)(y - y_0)$$

which is the equation of the tangent plane.

Ex: If $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is differentiable at (x_0, y_0) then there exists a unique tangent plane that approximates f in a neighborhood of (x_0, y_0) .



However, even if all the directional derivatives $D_{\vec{u}} f(x_0, y_0)$ exist (recall that $\vec{u} = (1, 0)$ and $\vec{u} = (0, 1)$ correspond to $\frac{\partial f}{\partial x}(x_0, y_0)$ and $\frac{\partial f}{\partial y}(x_0, y_0)$ respectively), the function might not be differentiable at (x_0, y_0) and hence there is no tangent plane that approximates f .

Ex: Recall the geometrical meaning of $D_{\vec{u}} f(x_0, y_0)$, $\vec{u} = (u_1, u_2)$ a unit vector.

$$D_{\vec{u}} f(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x_0 + tu_1, y_0 + tu_2) - f(x_0, y_0)}{t}$$

$$\vec{u} = (1, 0) \quad D_{\vec{u}} f(x_0, y_0) = \frac{\partial f}{\partial x}(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x_0 + t, y_0) - f(x_0, y_0)}{t}$$

$$\vec{u} = (0, 1) \quad D_{\vec{u}} f(x_0, y_0) = \frac{\partial f}{\partial y}(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x_0, y_0 + t) - f(x_0, y_0)}{t}$$

