

DIFFERENTIABILITY

The real-valued case

Def: Let $U \subset \mathbb{R}^m$ open, $f: U \rightarrow \mathbb{R}$, $x \in U$, $v \in \mathbb{R}^m$

The **directional derivative** of f at x along v is

$$\partial_v f(x) := \lim_{t \rightarrow 0} \frac{f(x + tv) - f(x)}{t} \text{ whenever it exists}$$

Remark: since U is open, $x + tv \in U$ for t "small enough".

Remark: if $\alpha \in \mathbb{R}$ then $\partial_{\alpha v} f(x) = \alpha \partial_v f(x)$.

∴ Hence, if we know $\partial_v f(x)$ for some v , we know the directional derivatives for all the vectors with same direction.

Intuitively: by the above remark, we may assume that $\|v\| = 1$ and then $\partial_v f(x)$ is the instantaneous rate of change of f through x along the direction of v .

Ex: let $f(x, y) = x \cos(y) + y e^x$ and $v = (\cos \theta, \sin \theta)$

$$\text{then } \partial_v f(0,0) = \lim_{t \rightarrow 0} \frac{f((0,0) + t(\cos \theta, \sin \theta)) - f(0,0)}{t}$$

$$= \lim_{t \rightarrow 0} \frac{f(t \cos \theta, t \sin \theta)}{t}$$

$$= \lim_{t \rightarrow 0} \cos \theta \cos(t \sin \theta) + \sin \theta e^{t \cos \theta}$$

$$= \cos(\theta) + \sin(\theta)$$

The highest rate of change through $(0,0)$ is $\overset{\text{along}}{v} \theta = \frac{\pi}{4}$ and the lowest at $\theta = \frac{5\pi}{4}$

Homework: plot the graph on MathSageCell.

Prop: $U \subset \mathbb{R}^m$ open, $f, g: U \rightarrow \mathbb{R}$, $x \in U$, $v \in \mathbb{R}^m$, $c \in \mathbb{R}$

If $\partial_v f(x)$ and $\partial_v g(x)$ exist then $\partial_v (f+g)(x)$, $\partial_v (cf)(x)$, $\partial_v (fg)(x)$ exist and

$$\textcircled{1} \partial_v (f+g)(x) = \partial_v f(x) + \partial_v g(x)$$

$$\textcircled{2} \partial_v (cf)(x) = c \partial_v f(x)$$

$$\textcircled{3} \partial_v (fg)(x) = g(x) \cdot \partial_v f(x) + f(x) \cdot \partial_v g(x) \quad \text{"Leibniz rule"}$$

Δ $\textcircled{1}$ & $\textcircled{2}$: obvious

$$\textcircled{3}: \lim_{t \rightarrow 0} \frac{f(x+tv)g(x+tv) - f(x)g(x)}{t}$$

$$= \lim_{t \rightarrow 0} \frac{f(x+tv)g(x+tv) - f(x)g(x+tv) + f(x)g(x+tv) - f(x)g(x)}{t}$$

$$= \lim_{t \rightarrow 0} g(x+tv) \frac{f(x+tv) - f(x)}{t} + f(x) \frac{g(x+tv) - g(x)}{t}$$

$$= g(x) \partial_v f(x) + f(x) \partial_v g(x)$$

Remark: $g(x+tv) = \frac{g(x+tv) - g(x)}{t} \cdot t + g(x) \xrightarrow{t \rightarrow 0} 0 + g(x)$ \square

Prop: $U \subset \mathbb{R}^m$ open, $x \in U$, $f: U \rightarrow \mathbb{R}$, $h: I \rightarrow \mathbb{R}$, $v \in \mathbb{R}^m$

If $\partial_v f(x)$ exists and h differentiable at $f(x)$ then $\partial_v (h \circ f)(x)$ exists

$$\text{and} \quad \partial_v (h \circ f)(x) = h'(f(x)) \cdot \partial_v f(x)$$

Δ Later, but you can adapt the proof of the chain rule from MAT137 \square

Def. Let $U \subset \mathbb{R}^m$ open, $x \in U$, $i=1, \dots, m$

We define the i -th partial derivative of f at x by

$$\frac{\partial f}{\partial x_i}(x) := \partial_{e_i} f(x) = \lim_{t \rightarrow 0} \frac{f(x + te_i) - f(x)}{t}$$

whenever it exists.

In practice: $f(x + te_i) = f(x_1, x_2, \dots, x_{i-1}, x_i + t, x_{i+1}, \dots, x_m)$

so $\frac{\partial f}{\partial x_i}(x) = g'(x_i)$ where $g(x_i) = f(x_1, \dots, x_i, \dots, x_m)$

where all the other variables are frozen.

Ex: $f(x, y) = x^2 e^{xy}$

$$\frac{\partial f}{\partial x}(x, y) = 2x e^{xy} + x^2 y e^{xy}$$

$$\frac{\partial f}{\partial x}(1, 1) = 3e$$

$$\frac{\partial f}{\partial y}(x, y) = x^3 e^{xy}$$

$$\frac{\partial f}{\partial y}(1, 1) = e$$

Ex: $f(x, y) = |x|(1+y)$

$$\frac{\partial f}{\partial x}(0, 0) \text{ DNE} : \lim_{t \rightarrow 0} \frac{f(t, 0) - f(0, 0)}{t} = \lim_{t \rightarrow 0} \frac{|t|}{t}$$

$$\frac{\partial f}{\partial y}(0, 0) = 0$$

Ex: $f(x, y) = \begin{cases} \frac{xy}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0) \end{cases}$

$$\frac{\partial f}{\partial x}(0, 0) = 0, \quad \frac{\partial f}{\partial y}(0, 0) = 0 \text{ exist}$$

$$\text{but } \partial_{(1,1)} f(0, 0) = \lim_{t \rightarrow 0} \frac{f(t, t)}{t} = \lim_{t \rightarrow 0} \frac{1}{2t} \text{ DNE}$$

\triangle ∇ : the partial derivatives exist
 ∇ ∇ : all the directional derivatives exist

Def: $U \subset \mathbb{R}^m$ open, $f: U \rightarrow \mathbb{R}$, $x \in U$, $\frac{\partial f}{\partial x_i}(x)$ exists $\forall i$

The gradient of f at x is

$$\nabla f(x) := \left(\frac{\partial f}{\partial x_1}(x), \frac{\partial f}{\partial x_2}(x), \dots, \frac{\partial f}{\partial x_m}(x) \right) \in \mathbb{R}^m$$

Ex: $f(x,y) = x \cos(y) + y e^x$

$$\nabla f(x,y) = (\cos(y) + y e^x, -x \sin(y) + e^x)$$

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

" $\nabla f(0,0)$ "

Remark: We have already seen that $(1,1)$ ($\theta = \frac{\pi}{4}$) was the direction where f has a the highest rate of change through $(0,0)$.

[We will see later that it is a general phenomenon

Def: $U \subset \mathbb{R}^m$ open, $f: U \rightarrow \mathbb{R}$, $x \in U$

We say that f is differentiable at x if there exists a linear function $d_x f: \mathbb{R}^m \rightarrow \mathbb{R}$ s.t.

$$\lim_{h \rightarrow 0} \frac{f(x+h) - f(x) - d_x f(h)}{\|h\|} = 0$$

$$\begin{aligned} d_x f(\lambda u + \mu v) &= \lambda d_x f(u) + \mu d_x f(v) \end{aligned}$$

i.e. $f(x+h) = f(x) + d_x f(h) + E(h)$ where $\frac{E(h)}{\|h\|} \rightarrow 0$

"linear approximation of f at x "

Ex 1) $f: \mathbb{R} \rightarrow \mathbb{R}$, $f(x) = x^2$, $f(x+h) = f(x) + 2xh + h^2$
so f is differentiable on \mathbb{R} and $d_x f(h) = 2xh$

2) $g: \mathbb{R}^m \rightarrow \mathbb{R}$, $g(x) = \|x\|^2$, $g(x+h) = \|x+h\|^2 = \|x\|^2 + 2(x \cdot h) + \|h\|^2$

so g is differentiable on \mathbb{R}^m and $d_x g(h) = 2(x \cdot h)$

Theorem: If f is differentiable at x then its differential df_x is unique

Δ Assume that f has 2 differentials at x : $l_1, l_2: \mathbb{R}^m \rightarrow \mathbb{R}$

Let $h \in \mathbb{R}^m$. Since U is open, $x+th \in U$ when $t \neq 0$ is small.

$$\begin{aligned} \text{Then: } l_1(h) - l_2(h) &= \frac{l_1(th) - l_2(th)}{t} \\ &= - \frac{f(x+th) - f(x) - l_1(th)}{t \|h\|} \|h\| \\ &\quad + \frac{f(x+th) - f(x) - l_2(th)}{t \|h\|} \|h\| \end{aligned}$$

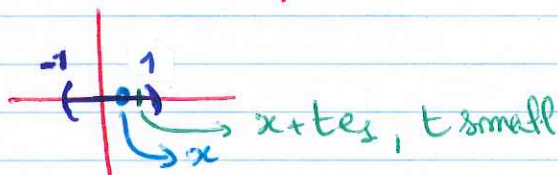
$$\xrightarrow{t \rightarrow 0^+} 0$$

Hence $l_1(h) = l_2(h)$ □

! What's why we want the domain to be open. Otherwise the differential could not be unique.

Ex: $U = (-1, 1) \times \{0\}$

We can determine $d_x f(e_1)$:



But we have no condition for $d_x f(e_2)$ since $x+te_2 \notin U \forall t \neq 0$ so we can take whatever we want.

Prop: $U \subset \mathbb{R}^m$ open, $f: U \rightarrow \mathbb{R}$, $x \in U$

If f is differentiable at x then f is continuous at x .

$$\Delta f(x) = f(x + (x-x)) = f(x) + d_x f(x-x) + E(x-x) \xrightarrow{x \rightarrow x} f(x) + 0 + 0 \quad \square$$

Prop: $f, g: U \rightarrow \mathbb{R}$ differentiable at $x \in U$, $U \subset \mathbb{R}^m$ open, then

- ① $d_x(f+g) = d_x f + d_x g$
- ② $d_x(\lambda f) = \lambda d_x f$, $\lambda \in \mathbb{R}$
- ③ $d_x(fg) = g(x) d_x f + f(x) d_x g$
- ④ $d_x(1/f) = -1/f(x)^2 \cdot d_x f$ if $f(x) \neq 0$

Theorem: $U \subset \mathbb{R}^m$ open, $f: U \rightarrow \mathbb{R}$, $x \in U$

If f is differentiable at x , then

- ① All the directional derivatives of f at x exist
 - ② $\forall v \in \mathbb{R}^m$, $\partial_v f(x) = d_x f(v)$
 - ③ $\forall h \in \mathbb{R}^m$, $d_x f(h) = \nabla f(x) \cdot h = \frac{\partial f(x)}{\partial x_1} h_1 + \dots + \frac{\partial f(x)}{\partial x_m} h_m$
 - ④ $\forall v \in \mathbb{R}^m$, $\partial_v f(x) = \nabla f(x) \cdot v$
- (that's why physicists write $df = \sum \frac{\partial f}{\partial x_i} dx_i$)

Δ ①+②:

$$\begin{aligned} \frac{f(x+tv) - f(x)}{t} &= \frac{f(x+tv) - f(x) - d_x f(tv) + d_x f(tv)}{t} \\ &= \frac{f(x+tv) - f(x) - d_x f(tv)}{t \|v\|} \|v\| + d_x f(v) \\ &\xrightarrow{t \rightarrow 0} 0 + d_x f(v) \end{aligned}$$

So $\partial_v f(x)$ exists and $\partial_v f(x) = d_x f(v)$

③ $d_x f(e_i) = \partial_{e_i} f(x) = \frac{\partial f}{\partial x_i}(x)$

Hence $d_x f(h_1, \dots, h_m) = d_x f(\sum h_i e_i) = \sum h_i d_x f(e_i) = \sum h_i \frac{\partial f}{\partial x_i}(x) = \nabla f(x) \cdot h$

④ $\partial_v f(x) = d_x f(v) = \nabla f(x) \cdot v$ □

Remark: if f is differentiable at x then $d_x f(h) = \nabla f(x) \cdot h$, there is no other possibility

Ex: $f(x,y) = \begin{cases} \frac{xy}{x^2+y^2} & \text{if } (x,y) \neq (0,0) \\ 0 & \text{otherwise} \end{cases}$ is not differentiable at $(0,0)$ since $\partial_{(1,0)} f(0,0)$ DNE

⚠ The converse of the above theorem is false:

$f(x,y) = \begin{cases} \frac{x^3}{x^2+y^2} & \text{if } (x,y) \neq (0,0) \\ 0 & \text{otherwise} \end{cases}$ all the directional derivatives exist at $(0,0)$ but f is not differentiable at $(0,0)$

• $\partial_v f(0,0) = \lim_{t \rightarrow 0} \frac{t^3 v_1^3}{t^3 v_1^2 + t^3 v_2^2} = \frac{v_1^3}{v_1^2 + v_2^2}$

• Assume by contradiction that f is differentiable at 0 , then $df(1,1) = df(1,0) + df(0,1) = \partial_{(1,0)} f(0,0) + \partial_{(0,1)} f(0,0) = 1 + 0 = 1 = \partial_{(1,1)} f(0,0) = 1/2$

Remark: If $f: U \rightarrow \mathbb{R}$ is differentiable at x and $\nabla f(x) \neq \vec{0}$ then the direction of $\nabla f(x)$ is the direction of fastest increase at x and the magnitude of $\nabla f(x)$ is the instantaneous rate of change in that direction.

Δ Let $v \in \mathbb{R}^m$ be a unit vector then

$$D_v f(x) = d_x f(v) = \nabla f(x) \cdot v = |v| \cdot |\nabla f(x)| \cdot \cos \theta = |\nabla f(x)| \cdot \cos \theta$$

the max is when $\cos \theta = 1$, i.e. $v = \frac{\nabla f(x)}{|\nabla f(x)|}$ and then $D_v f(x) = |\nabla f(x)|$ \square

(Compare with the examples about $f(x,y) = x \cos(y) + y e^x$ above.)

Theorem: $U \subset \mathbb{R}^m$ open, $f: U \rightarrow \mathbb{R}$, $x \in U$ at least in an open ball around x .

If the partial derivatives of f exist on U and are continuous at x then f is differentiable at x

Δ ~~Idia~~: we apply the MVT to $t \mapsto f(t, x_2, \dots, x_m)$ on $[x_1, x_1 + h_1]$

$$\text{then } \exists \theta_1 \text{ s.t. } f(x_1 + h_1, x_2, \dots, x_m) - f(x_1, x_2, \dots, x_m) = h_1 \frac{\partial f}{\partial x_1}(x_1 + \theta_1 h_1, x_2, \dots, x_m)$$

By MVT to $t \mapsto f(x_1 + h_1, t, x_3, \dots, x_m)$ on $[x_2, x_2 + h_2]$

$$\exists \theta_2 \text{ s.t. } f(x_1 + h_1, x_2 + h_2, x_3, \dots, x_m) - f(x_1 + h_1, x_2, x_3, \dots, x_m) = h_2 \frac{\partial f}{\partial x_2}(x_1 + h_1, x_2 + \theta_2 h_2, x_3, \dots, x_m)$$

and so on for x_3, \dots, x_m .

$$\text{Then: } f(x+h) - f(x) - \sum h_i \frac{\partial f}{\partial x_i}(x) = \sum h_i \left(\frac{\partial f}{\partial x_i}(a_i) - \frac{\partial f}{\partial x_i}(x) \right) \xrightarrow{h \rightarrow 0} 0 \text{ by continuity of } \frac{\partial f}{\partial x_i}$$

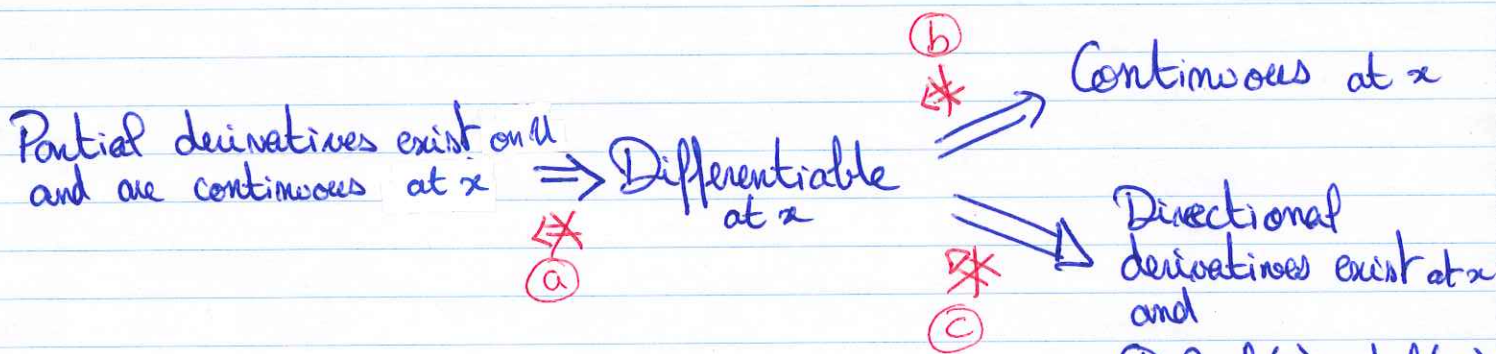
$$\text{where } a_i = (x_1 + h_1, \dots, x_i + h_i \theta_i, x_{i+1}, \dots, x_m) \xrightarrow{h \rightarrow 0} x$$

The converse is false: $f(x) = \begin{cases} x^2 \sin(1/x) & \text{if } x \neq 0 \\ 0 & \text{otherwise} \end{cases}$

is differentiable on \mathbb{R} but f' is not continuous at 0

Summary: $f: U \rightarrow \mathbb{R}$, $U \subset \mathbb{R}^m$ open

name	nature	notation
Directional derivative at $x \in U$ along $v \in \mathbb{R}^m$	Real number	$\partial_v f(x)$
Partial derivative at x	Real number	$\frac{\partial f}{\partial x_i}(x)$
Gradient at x	Vector of \mathbb{R}^m	$\nabla f(x)$
Differential at x	Linear function $\mathbb{R}^m \rightarrow \mathbb{R}$	$dx f(h)$



Directional derivatives exist at x and

① $\partial_v f(x) = dx f(v)$

② $dx f(h) = \nabla f(x) \cdot h = \sum_{i=1}^m \frac{\partial f}{\partial x_i}(x) h_i$

③ $\partial_v f(x) = \nabla f(x) \cdot v$

Counter-examples:

(a) $f: \mathbb{R} \rightarrow \mathbb{R}$, $f(x) = \begin{cases} x^2 \sin(\frac{1}{x}) & \text{if } x \neq 0 \\ 0 & \text{if } x = 0 \end{cases}$

(b) $f(x) = |x|$

(c) $f(x,y) = \begin{cases} \frac{x^3}{x^2+y^2} & \text{if } (x,y) \neq (0,0) \\ 0 & \text{otherwise} \end{cases}$ on \mathbb{R}^2

useful \uparrow to prove that a function is not differentiable. See (c)

* Partial derivatives exist ~~*~~ Directional derivatives exist

Ex: $f(x,y) = \begin{cases} \frac{xy}{x^2+y^2} & \text{if } (x,y) \neq (0,0) \\ 0 & \text{otherwise} \end{cases}$