

$$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}^m \quad f'(a) \in \mathbb{R}^m \quad df(a) \in L(\mathbb{R}, \mathbb{R}^m)$$

3.6. Theorem Let  $A \subseteq \mathbb{R}$ , let  $a \in \text{int } A$ , and let  $f: A \rightarrow \mathbb{R}^m$ . Then the following assertions are true:

1° If  $f$  is differentiable at  $a \Rightarrow f$  is Fréchet diff. at  $a$  and

$$(1) \quad \forall h \in \mathbb{R} : df(a)(h) = h \cdot f'(a)$$

2° If  $f$  is Fréchet diff. at  $a \Rightarrow f$  is diff. at  $a$  and (1) holds.

Proof. 1° Assume that  $f$  is diff. at  $a \Rightarrow \exists \lim_{x \rightarrow a} \frac{1}{|x-a|} [f(x) - f(a)] = f'(a)$  (\*)

Consider  $\varphi: \mathbb{R} \rightarrow \mathbb{R}^m$ ,  $\varphi(h) := h f'(a) \Rightarrow \varphi \in L(\mathbb{R}, \mathbb{R}^m)$

$$\begin{aligned} \lim_{x \rightarrow a} \frac{1}{|x-a|} [f(x) - f(a) - \varphi(x-a)] &= \lim_{x \rightarrow a} \frac{1}{|x-a|} [f(x) - f(a) - (x-a) f'(a)] = \\ &= \lim_{x \rightarrow a} \frac{x-a}{|x-a|} \left[ \frac{1}{x-a} (f(x) - f(a)) - f'(a) \right] = 0_m \end{aligned}$$

$\stackrel{\text{def}}{\Rightarrow} f$  is Fréchet diff. at  $a$  and  $df(a) = \varphi \Rightarrow df(a)(h) = \varphi(h) = h f'(a) \quad \forall h \in \mathbb{R}$

2° Assume that  $f$  is Fréchet diff at  $a \Rightarrow df(a) \in L(\mathbb{R}, \mathbb{R}^m) \Rightarrow$

$\Rightarrow \exists v \in \mathbb{R}^m$  s.t.  $df(a)(h) = hv \quad \forall h \in \mathbb{R}$

$$\text{Def } \Rightarrow 0_m = \lim_{x \rightarrow a} \frac{1}{|x-a|} [f(x) - f(a) - df(a)(x-a)] = \lim_{x \rightarrow a} \frac{1}{|x-a|} [f(x) - f(a) - (x-a)v]$$

$$= \lim_{x \rightarrow a} \frac{x-a}{|x-a|} \left[ \frac{1}{x-a} (f(x) - f(a)) - v \right]$$

$$\Rightarrow \lim_{x \rightarrow a} \left\| \frac{x-a}{|x-a|} \left[ \frac{1}{x-a} (f(x) - f(a)) - v \right] \right\| = 0$$

$$\Rightarrow \lim_{x \rightarrow a} \left\| \frac{1}{x-a} (f(x) - f(a)) - v \right\| = 0$$

$$\Rightarrow \lim_{x \rightarrow a} \frac{1}{x-a} (f(x) - f(a)) = v \quad \Rightarrow f \text{ is diff at } a \text{ and } f'(a) = v \\ \Rightarrow df(a)(h) = hv = hf'(a)$$

#### 4. Directional derivatives

4.1. Definition Let  $A \subseteq \mathbb{R}^n$ ,  $a \in \text{int } A$ ,  $f: A \rightarrow \mathbb{R}^m$ ,  $v \in \mathbb{R}^n$ .

One says that  $f$  has a directional derivative at  $a$  in the direction  $v$

if  $\exists \lim_{t \rightarrow 0} \frac{1}{t} [f(a+tv) - f(a)] =: b \in \mathbb{R}^m$

4.2. Theorem Let  $A \subseteq \mathbb{R}^n$ ,  $a \in \text{int } A$ ,  $f: A \rightarrow \mathbb{R}^m$ . If  $f$  is differentiable at  $a \Rightarrow f$  has a directional derivative in every direction  $v \in \mathbb{R}^n$  and  $f'(a; v) = df(a)(v)$ .

Proof let  $v \in \mathbb{R}^m$  P3.3

$f$  is diff. at  $a \Rightarrow \exists \omega: A \rightarrow \mathbb{R}^m$  s.t.  $\lim_{x \rightarrow a} \omega(x) = 0_m$  and

$$\forall x \in A : f(x) = f(a) + df(a)(x-a) + \|x-a\| \cdot \omega(x) \quad (*)$$

Since  $a \in \text{int } A \Rightarrow \exists \delta > 0$  s.t.  $\forall t \in (-\delta, \delta) : a+tv \in A$

Replacing  $x$  by  $a+tv$  in  $(*)$

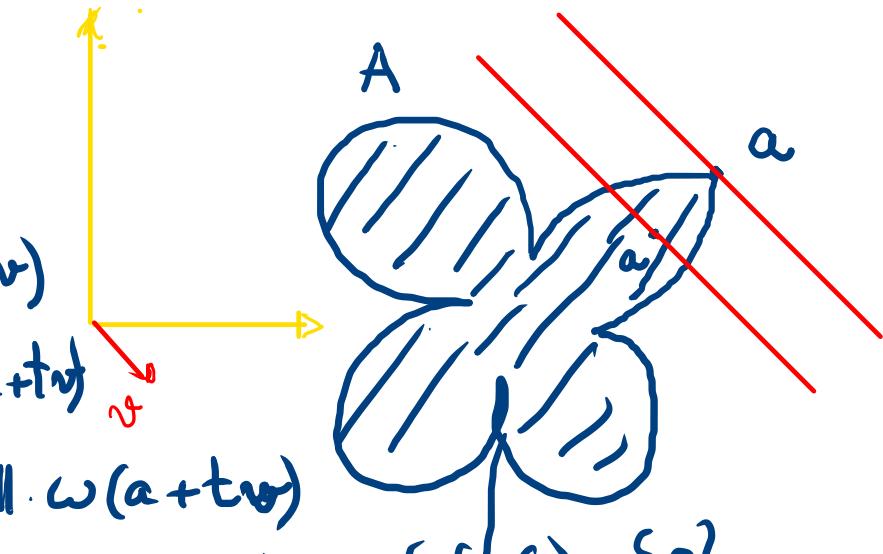
$\Rightarrow \forall t \in (-\delta, \delta) :$

$$\begin{aligned} \frac{1}{t} |f(a+tv) - f(a)| &= f(a) + df(a)(tv) + \|tv\| \omega(a+tv) \\ &= f(a) + t df(a)(v) + |t| \cdot \|v\| \cdot \omega(a+tv) \end{aligned}$$

$$\Rightarrow \frac{1}{t} [f(a+tv) - f(a)] = df(a)(v) + \frac{|t| \cdot \|v\| \cdot \omega(a+tv)}{t}$$

Letting  $t \rightarrow 0 \Rightarrow \lim_{t \rightarrow 0} \frac{1}{t} [f(a+tv) - f(a)] = df(a)(v)$   $\forall t \in (-\delta, \delta) \setminus \{0\}$

$$\Rightarrow f'(a; v) = df(a)(v).$$

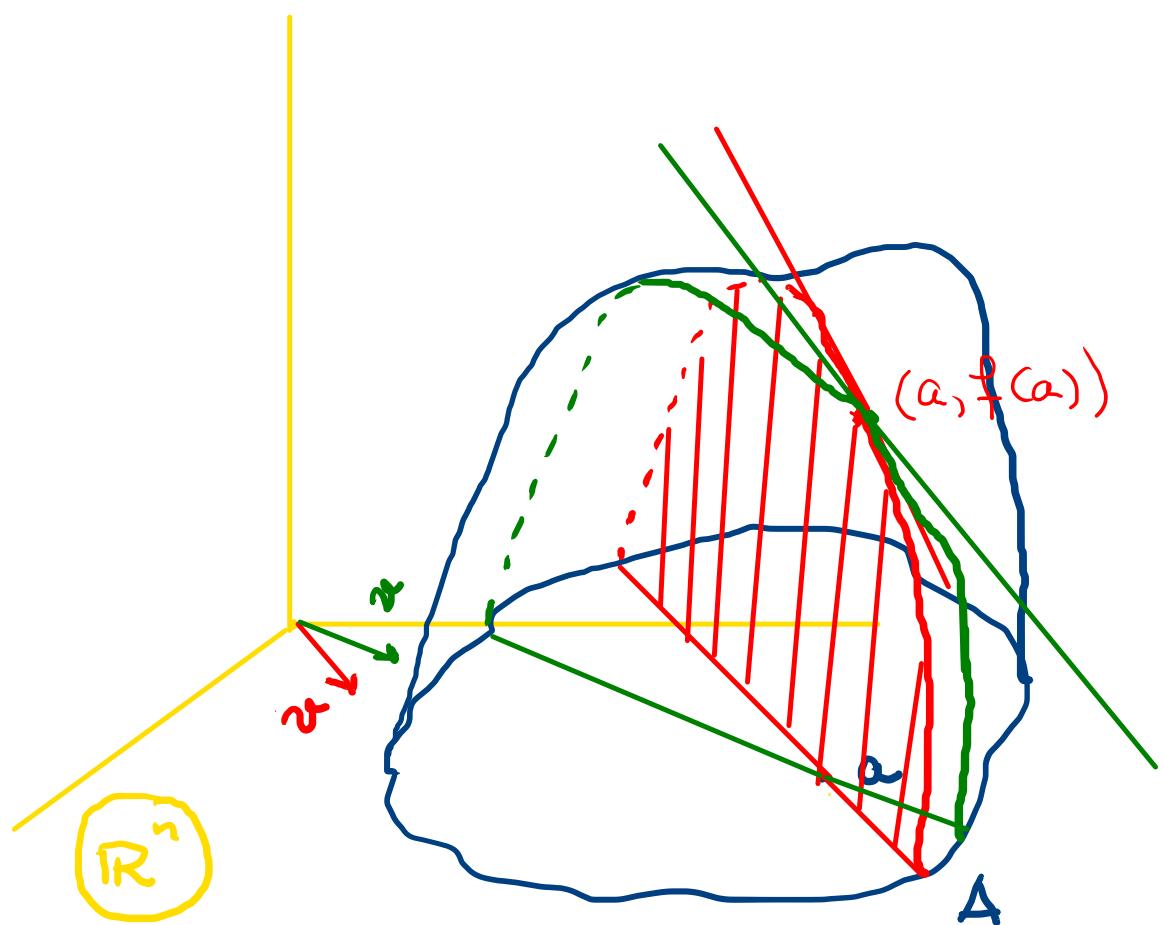


The vector  $b$  is called the directional derivative of  $f$  at  $a$  in the direction  $v$  and it is denoted by  $f'(a; v)$  or by  $f'_v(a)$

$$m=1 \quad \|v\|=1$$

$f'(a; v) = \text{instantaneous rate of change of } f \text{ at the point } a \text{ in the direction } v$

$$(x, y) \in A \mapsto z = f(x, y)$$



## 5. Partial derivatives

5.1. Definition. Let  $A \subseteq \mathbb{R}^n$ ,  $a \in \text{int } A$ ,  $f: A \rightarrow \mathbb{R}^m$ ,  $j \in \{1, \dots, n\}$ . If  $f$  has a directional derivative at  $a$  in the direction  $e_j$ , then  $f$  is said to be partially differentiable at  $a$  wrt to the variable  $x_j$ .

The directional derivative  $f'(a; e_j)$  is called the partial derivative of  $f$  at  $a$  wrt the variable  $x_j$  and it is denoted by

$$\frac{\partial f}{\partial x_j}(a) \quad (\text{Jacobi's notation}) \quad \text{or by} \quad f'_{x_j}(a).$$

If  $f$  is partially diff. wrt all variables  $x_1, \dots, x_n$ , then we say that  $f$  is partially diff. at  $a$ .

$$\frac{\partial f}{\partial x_j}(a) = f'(a; e_j) = \lim_{t \rightarrow 0} \frac{1}{t} [f(a + te_j) - f(a)]$$

$$= \lim_{t \rightarrow 0} \frac{1}{t} [f(a_1, \dots, a_{j-1}, a_j + t, a_{j+1}, \dots, a_n) - f(a_1, \dots, a_{j-1}, a_j, a_{j+1}, \dots, a_n)]$$

$$x_j := a_j + t$$

$$\frac{\partial f}{\partial x_j}(a) = \lim_{x_j \rightarrow a_j} \frac{1}{x_j - a_j} [f(a_1, \dots, a_{j-1}, x_j, a_{j+1}, \dots, a_n) - f(a_1, \dots, a_{j-1}, a_j, a_{j+1}, \dots, a_n)]$$

5.2. Definition (Jacobi matrix) Let  $A \subseteq \mathbb{R}^n$ ,  $a \in \text{int } A$ ,  $f: A \rightarrow \mathbb{R}^m$  s.t.  $f$  is partially diff. at  $a$ . Then we define

$$J(f)(a) = J_f(a) := \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(a) & \frac{\partial f_1}{\partial x_2}(a) & \cdots & \frac{\partial f_1}{\partial x_n}(a) \\ \frac{\partial f_2}{\partial x_1}(a) & \frac{\partial f_2}{\partial x_2}(a) & \cdots & \frac{\partial f_2}{\partial x_n}(a) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \frac{\partial f_m}{\partial x_2}(a) & \cdots & \frac{\partial f_m}{\partial x_n}(a) \end{pmatrix} \in M_{m \times n}(\mathbb{R})$$

→ the Jacobi matrix of  $f$  at  $a$

$f$  is partially diff. w.r.t  $x_j$  at  $a \Leftrightarrow f_1, \dots, f_m$  are partially diff. w.r.t  $x_j$  at  $a$

$$\frac{\partial f}{\partial x_j}(a) = \left( \frac{\partial f_1}{\partial x_j}(a), \dots, \frac{\partial f_m}{\partial x_j}(a) \right)$$

$$m=1 \Rightarrow J(f)(a) = \begin{pmatrix} \frac{\partial f}{\partial x_1}(a) & \frac{\partial f}{\partial x_2}(a) & \dots & \frac{\partial f}{\partial x_n}(a) \end{pmatrix} \in M_{1 \times n}(\mathbb{R})$$

Consider the vector  $\nabla f(a) := \left( \frac{\partial f}{\partial x_1}(a), \frac{\partial f}{\partial x_2}(a), \dots, \frac{\partial f}{\partial x_n}(a) \right) \in \mathbb{R}^n$   
 $\nabla = \text{matrix}$   
 $\hookrightarrow$  the gradient of  $f$  at  $a$

$$J(f)(a) = \nabla f(a)^t$$

The geometrical meaning of partial derivatives:

$f: A \rightarrow \mathbb{R}$        $a \in A \cap A'$ ,  $f$  is differentiable at  $a$   
 $A \subseteq \mathbb{R}$

The tangent line to  $G_f$  at  $(a, f(a))$  has the equation

$$y - f(a) = f'(a)(x - a)$$

$$f'(a) \cdot x - y - af'(a) + f(a) = 0$$

$\Rightarrow n = (f'(a), -1)$  is a normal vector to  $G_f$

$$\alpha x + \beta y + \delta = 0$$

$$\Rightarrow n = (\alpha, \beta) =$$

$$= \alpha \vec{i} + \beta \vec{j}$$

is a normal vector  
to the tangent line

$z = f(x, y)$  is the equation of a surface

$$f: A \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$(x_0, y_0) \in \text{int } A, \quad z_0 = f(x_0, y_0)$$

$$\Rightarrow n = (f'_x(x_0, y_0), f'_y(x_0, y_0), -1)$$

Take for instance an elliptic paraboloid

$$z = \frac{x^2}{a^2} + \frac{y^2}{b^2} = f(x, y)$$

The tangent plane of the paraboloid at  $(x_0, y_0, z_0)$

$$\frac{z+z_0}{2} = \frac{xx_0}{a^2} + \frac{yy_0}{b^2} \quad | \cdot 2$$

$$\frac{2x_0}{a^2} \cdot x + \frac{2y_0}{b^2} \cdot y - z - z_0 = 0$$

$$\Rightarrow n = \left( \underbrace{\frac{2x_0}{a^2}}_{f'_x(x_0, y_0)}, \underbrace{\frac{2y_0}{b^2}}_{f'_y(x_0, y_0)}, -1 \right) \text{ is the normal to the tangent plane}$$

is a normal vector to the surface  
(i.e. a normal vector to the tangent plane of the surface  
at  $(x_0, y_0, z_0)$ )

$$\begin{aligned} x &\rightarrow \frac{x+x_0}{2} & x^2 &\rightarrow xx_0 \\ y &\rightarrow \frac{y+y_0}{2} & y^2 &\rightarrow yy_0 \\ z &\rightarrow \frac{z+z_0}{2} & z^2 &\rightarrow zz_0 \end{aligned}$$

$$\pi: \alpha x + \beta y + \gamma z + \delta = 0$$

$n = (\alpha, \beta, \gamma)$  is the normal to  $\pi$

$F(x, y) = 0$  is the implicit equation of a curve

$$F(x_0, y_0) = 0 \Rightarrow (x_0, y_0) \in \text{curve}$$

$$\nabla F(x_0, y_0) = \left( \frac{\partial F}{\partial x}(x_0, y_0), \frac{\partial F}{\partial y}(x_0, y_0) \right)$$

$$(x_0, y_0) \text{ belonging to } \frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$$

The tangent line to the ellipse at  $(x_0, y_0)$

$$\frac{x_0}{a^2} + \frac{y_0}{b^2} = 1 \Leftrightarrow \frac{x_0}{a^2}x + \frac{y_0}{b^2}y - 1 = 0$$

$\Rightarrow n = \left( \frac{x_0}{a^2}, \frac{y_0}{b^2} \right)$  is the normal vector of the ellipse at  $(x_0, y_0)$

$$\nabla F(x_0, y_0) = \left( \frac{2x_0}{a^2}, \frac{2y_0}{b^2} \right) = 2n$$

$\cdot F(x, y, z) = 0$  the implicit equation of a surface

$$F(x, y, z) = \frac{x^2}{a^2} + \frac{y^2}{b^2} - \frac{z^2}{c^2} - 1$$

$\nabla F(x_0, y_0, z_0)$  is a normal vector to the curve at  $(x_0, y_0)$  (i.e. a normal vector to the tangent of the curve at  $(x_0, y_0)$ )

$\nabla F(x_0, y_0, z_0)$  is a normal vector of the surface at  $(x_0, y_0, z_0)$

$$F(x_0, y_0, z_0) = 0$$

$$F(x, y, z) = \frac{x^2}{a^2} + \frac{y^2}{b^2} - \frac{z^2}{c^2} - 1$$

5.3. Theorem (Fréchet differential vs partial derivatives) Let  $A \subseteq \mathbb{R}^n$ ,  $a \in \text{int } A$ ,  $f: A \rightarrow \mathbb{R}^m$  s.t.  $f$  is diff. at  $a$ . Then:

1°  $f$  is partially diff. at  $a$ , and  $[df(a)] = J(f)(a)$

2°  $\forall h = (h_1, \dots, h_n) \in \mathbb{R}^n : df(a)(h) = h_1 \frac{\partial f}{\partial x_1}(a) + \dots + h_n \frac{\partial f}{\partial x_n}(a)$

Proof 1° Since  $f$  is differentiable at  $a \Rightarrow f$  has directional derivatives in every direction at  $a$ , and  $f'(a; v) = df(a)(v) \quad \forall v \in \mathbb{R}^n$   
 $\Rightarrow f$  is partially diff. at  $a$  and  $\frac{\partial f}{\partial x_j}(a) = f'(a; e_j) = df(a)(e_j), j=1, n$

But  $\frac{\partial f}{\partial x_j}(a) = j^{\text{th}} \text{ column in } J(f)(a)$

$df(a)(e_j) = j^{\text{th}} \text{ column in } [df(a)]$

$$[df(a)] = J(f)(a)$$

2° If  $h = (h_1, \dots, h_n) \in \mathbb{R}^n \Rightarrow h = h_1 e_1 + \dots + h_n e_n$

$$\Rightarrow df(a)(h) = h_1 df(a)(e_1) + \dots + h_n df(a)(e_n)$$

$$= h_1 \frac{\partial f}{\partial x_1}(a) + \dots + h_n \frac{\partial f}{\partial x_n}(a).$$

5.4 Corollary Let  $A \subseteq \mathbb{R}^n$ ,  $a \in \text{int } A$ ,  $f: A \rightarrow \mathbb{R}$  s.t.  $f$  is diff. at  $a$ . Then  $\forall h \in \mathbb{R}^n : df(a)(h) = \langle h, \nabla f(a) \rangle$

Remark.  $f'(a; v) = df(a)(v) = \langle v, \nabla f(a) \rangle$

$$\|v\|=1 \quad |f'(a; v)| = |\langle v, \nabla f(a) \rangle| \leq \|v\| \cdot \|\nabla f(a)\| = \|\nabla f(a)\|$$

↑  
Cauchy-Schwarz

$$\Rightarrow -\|\nabla f(a)\| \leq f'(a; v) \leq \|\nabla f(a)\|$$

we have equality  $\iff$

$$\Leftrightarrow v = -\frac{\nabla f(a)}{\|\nabla f(a)\|}$$

$\hookrightarrow$  we have equality  $\Leftrightarrow v = \frac{\nabla f(a)}{\|\nabla f(a)\|}$

Conclusion :  $\nabla f(a)$  gives the direction of steepest ascent of  $f$   
 $-\nabla f(a)$  gives the direction of steepest descent of  $f$

Example 1. The elevation of a portion of a hill is given by  $z = f(x, y) = 600 - 2x^2 - 3y$

Consider the point  $P$  lying above  $(1, 2) \Rightarrow P(1, 2, f(1, 2))$

In which direction will water fall from  $P$ ?

Answer:  $-\nabla f(1, 2) = (4, 3)$

$$\nabla f(x, y) = (-4x, -3)$$

Example. Captain Gigi is in trouble near the sunny side of Mercury.

The temperature of ship's hull at  $(x, y, z)$  is given by

$$T = T(x, y, z) = \frac{T_0}{1 + x^2 + 2y^2 + 3z^2}$$

In what direction should he proceed in order to decrease the temperature most rapidly?

$(x_0, y_0, z_0) \rightarrow$  the direction of steepest descent of temperature  
is  $-\nabla T(x_0, y_0, z_0) = (\dots)$

## Summarizing

$f$  is differentiable  
at  $a$



$f$  has directional  
derivatives in every  
direction at  $a$



$f$  is partially  
diff. at  $a$



$f$  is continuous  
at  $a$

5.4. Theorem Let  $A \subseteq \mathbb{R}^n$ ,  $a \in \text{int } A$ ,  $f: A \rightarrow \mathbb{R}^m$ . Suppose that

$\exists r > 0$  s.t.  $B(a, r) \subseteq A$  and

- $f$  is partially diff. at each point of  $B(a, r)$

- $\forall j = \overline{1, n}$  the function  $\forall x \in B(a, r) \mapsto \frac{\partial f}{\partial x_j}(x) \in \mathbb{R}^m$   
is continuous at  $a$ .

Then  $f$  is differentiable at  $a$ .

5.5. Corollary If  $A \subseteq \mathbb{R}^n$  is an open set, and  $f: A \rightarrow \mathbb{R}^m$  is partially differentiable at each point of  $A$ , with all partial derivatives  $\frac{\partial f}{\partial x_j}$  ( $j=1, \dots, n$ ) continuous on  $A$ , then  $f$  is diff. on  $A$ .

In order to study the differentiability of a scalar function  $f: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  at some point  $a \in \text{int } A$ , one can use the following algorithm:

- Step I Check the existence of all partial derivatives of  $f$  at  $a$
- if at least one partial derivative  $\frac{\partial f}{\partial x_j}(a)$  does not exist  
 $\Rightarrow f$  is not diff at  $a$  STOP
  - otherwise compute  $\frac{\partial f}{\partial x_j}(a)$ ,  $j=1, \dots, n$  and pass to Step 2

Step II Check the existence of the limit

$$l = \lim_{x \rightarrow a} \frac{f(x) - f(a) - \sum_{j=1}^n (x_j - a_j) \frac{\partial f}{\partial x_j}(a)}{\|x-a\|} \quad h = x-a$$

$$= \lim_{\substack{h_1 \rightarrow 0 \\ \vdots \\ h_n \rightarrow 0}} \frac{f(a_1+h_1, \dots, a_n+h_n) - f(a_1, \dots, a_n) - \sum_{j=1}^n h_j \frac{\partial f}{\partial x_j}(a)}{\sqrt{h_1^2 + \dots + h_n^2}}$$

- if  $l=0 \Rightarrow f$  is differentiable at  $a$
- otherwise  $\Rightarrow f$  is not differentiable at  $a$