

Métodos Matemáticos de Bioingeniería

Grado en Ingeniería Biomédica

Lecture 6

Marius A. Marinescu

Departamento de Teoría de la Señal y Comunicaciones
Área de Estadística e Investigación Operativa
Universidad Rey Juan Carlos

15 de marzo de 2021

Outline

- 1 The Derivative - Section 2.3
 - Partial derivatives
 - Differentiability
 - Matrix notation and differentiability in \mathbb{R}^n

Definition 3.1: Partial Function

- Suppose $f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ is a scalar-valued function of n variables.
- Let $\mathbf{x} = (x_1, x_2, \dots, x_n)$ denote a point of \mathbb{R}^n .
- A **partial function F with respect to the variable x_i** is a one-variable function obtained from f by holding all variables constant except x_i .
- That is, we set x_j equal to a constant a_j for $j \neq i$.
- Then the partial function in x_i is defined by

$$F(x_i) = f(a_1, a_2, \dots, x_i, \dots, a_n)$$

Remark

- We usually do not replace the x_j 's ($j \neq i$) by constants.
- Instead, we make a **mental note**.

Definition 2.2: Analytic and Geometric interpretation

- Symbolically, we have

$$\frac{\partial f}{\partial x_j} = \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_j + h, \dots, x_n) - f(x_1, \dots, x_n)}{h}$$

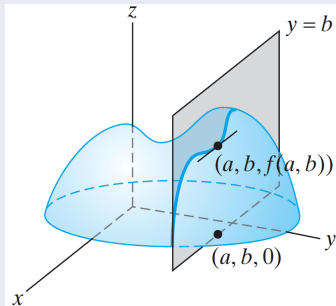
- The partial derivative is the (instantaneous) rate of change of f when all variables, except the specified one, are held fixed.
- In the case where f is a (scalar-valued) function of two variables we can consider,

$$\frac{\partial f}{\partial x}(a, b) \quad \text{and} \quad \frac{\partial f}{\partial y}(a, b)$$

Geometric Interpretation of Partial Derivatives in \mathbb{R}^2

$$\frac{\partial f}{\partial x}(a, b)$$

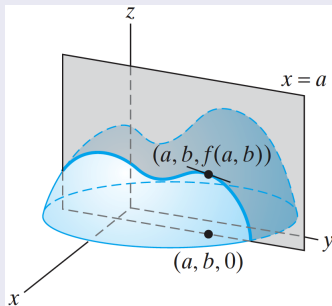
- Geometrically it is the slope at the point $(a, b, f(a, b))$ of the curve obtained by intersecting
 - The surface $z = f(x, y)$ with
 - The plane $y = b$



Geometric Interpretation of Partial Derivatives in \mathbb{R}^2

$$\frac{\partial f}{\partial y}(a, b)$$

- Geometrically it is the slope at the point $(a, b, f(a, b))$ of the curve obtained by intersecting
 - The surface $z = f(x, y)$ with
 - The plane $x = a$



Example 2a

- Let

$$f(x, y) = x^2y + \cos(x + y)$$

- Then, if we imagine y to be a constant throughout the differentiation process, we have,

$$\frac{\partial f}{\partial x} = 2xy - \sin(x + y)$$

- If we imagine x to be a constant,

$$\frac{\partial f}{\partial y} = x^2 - \sin(x + y)$$

Example 2b

- Let

$$g(x, y) = \frac{xy}{(x^2 + y^2)}$$

- Then

$$g_x(x, y) = \frac{(x^2 + y^2)y - xy(2x)}{(x^2 + y^2)^2} = \frac{y(y^2 - x^2)}{(x^2 + y^2)^2}$$

$$g_y(x, y) = \frac{(x^2 + y^2)x - xy(2y)}{(x^2 + y^2)^2} = \frac{x(x^2 - y^2)}{(x^2 + y^2)^2}$$

- Note that neither g nor its partial derivatives are defined at point $(0, 0)$.

Example 3

- Occasionally, it is necessary to appeal explicitly to limits to evaluate partial derivatives.
- Suppose $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ is defined by

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

- For $(x, y) \neq (0, 0)$, we have

$$\begin{aligned} \frac{\partial f}{\partial x} &= \frac{8xy^3}{(x^2 + y^2)^2} \\ \frac{\partial f}{\partial y} &= \frac{3x^4 - 6x^2y^2 - y^4}{(x^2 + y^2)^2} \end{aligned}$$

Example 3

- Occasionally, it is necessary to appeal explicitly to limits to evaluate partial derivatives:
- Suppose $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ is defined by

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

2. For $(x, y) = (0, 0)$, we return to [Definition 3.2](#)

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

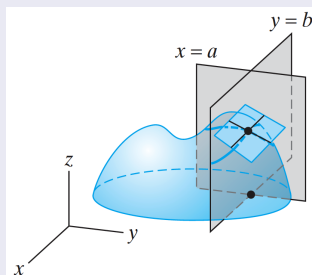
$$\frac{\partial f}{\partial y}(0, 0) = \lim_{h \rightarrow 0} \frac{f(0, 0 + h) - f(0, 0)}{h} = \lim_{h \rightarrow 0} \frac{-h - 0}{h} = \lim_{h \rightarrow 0} -1 = -1$$

Outline

- 1 The Derivative - Section 2.3
 - Partial derivatives
 - Differentiability
 - Matrix notation and differentiability in \mathbb{R}^n

Tangency for scalar-valued functions of two variables

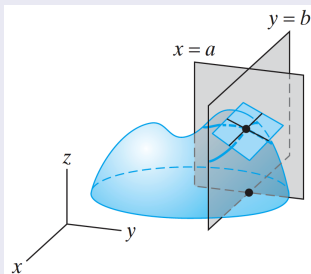
1. The partial derivative $f_x(a, b)$ is the slope of the line tangent at the point $(a, b, f(a, b))$ to the curve obtained by intersecting the surface $z = f(x, y)$ with the plane $y = b$



- If we travel along this tangent line, then for every unit change in the positive x -direction, there is a change of $f_x(a, b)$ units in the z -direction.

Tangency for scalar-valued functions of two variables

- The partial derivative $f_x(a, b)$ is the slope of the line tangent at the point $(a, b, f(a, b))$ to the curve obtained by intersecting the surface $z = f(x, y)$ with the plane $y = b$

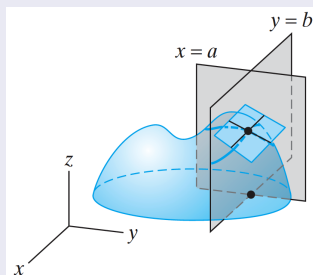


- Thus, a vector parallel to this tangent line is

$$\mathbf{u} = \mathbf{i} + f_x(a, b)\mathbf{k}$$

Tangency for scalar-valued functions of two variables

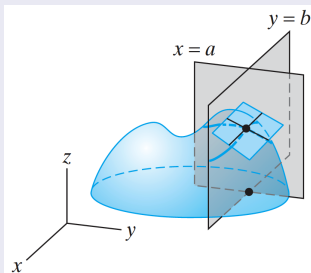
- Analogously, the partial derivative $f_y(a, b)$ is the slope of the line tangent at the point $(a, b, f(a, b))$ to the curve obtained by intersecting the surface $z = f(x, y)$ with the plane $x = a$.



- If we travel along this tangent line, then for every unit change in the positive y -direction, there is a change of $f_y(a, b)$ units in the z -direction.

Tangency for scalar-valued functions of two variables

- The partial derivative $f_y(a, b)$ is the slope of the line tangent at the point $(a, b, f(a, b))$ to the curve obtained by intersecting the surface $z = f(x, y)$ with the plane $x = a$.

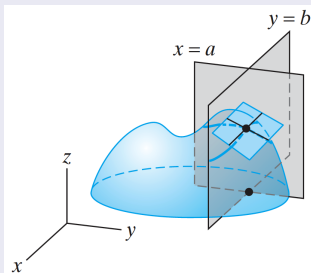


- The tangent line is given in vector parametric form as:

$$\mathbf{l}_2(t) = (a, b, f(a, b)) + t(0, 1, f_y(a, b))$$

Tangency for scalar-valued functions of two variables

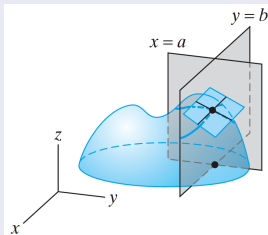
2. The partial derivative $f_y(a, b)$ is the slope of the line tangent at the point $(a, b, f(a, b))$ to the curve obtained by intersecting the surface $z = f(x, y)$ with the plane $x = a$



- Thus, a vector parallel to this tangent line is:

$$\mathbf{v} = \mathbf{j} + f_y(a, b)\mathbf{k}$$

Tangency for scalar-valued functions of two variables

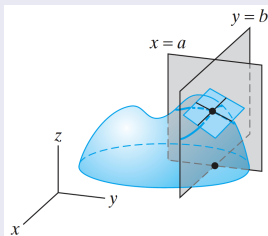


$$\mathbf{l}_1(t) = (a, b, f(a, b)) + t(1, 0, f_x(a, b)), \quad \mathbf{u} = \mathbf{i} + f_x(a, b)\mathbf{k}$$

$$\mathbf{l}_2(t) = (a, b, f(a, b)) + t(0, 1, f_y(a, b)), \quad \mathbf{v} = \mathbf{j} + f_y(a, b)\mathbf{k}$$

- Both of the tangent lines must be contained in the plane tangent to $z = f(x, y)$ at $(a, b, f(a, b))$, if one exists.
- A vector \mathbf{n} normal to the tangent plane must be perpendicular to both \mathbf{u} and \mathbf{v} .

Tangency for scalar-valued functions of two variables



$$\mathbf{l}_1(t) = (a, b, f(a, b)) + t(1, 0, f_x(a, b)), \quad \mathbf{u} = \mathbf{i} + f_x(a, b)\mathbf{k}$$

$$\mathbf{l}_2(t) = (a, b, f(a, b)) + t(0, 1, f_y(a, b)), \quad \mathbf{v} = \mathbf{j} + f_y(a, b)\mathbf{k}$$

- A vector \mathbf{n} normal to the tangent plane must be perpendicular to both \mathbf{u} and \mathbf{v}
- Therefore, we may take \mathbf{n} to be:

$$\mathbf{n} = \mathbf{u} \times \mathbf{v} = -f_x(a, b)\mathbf{i} - f_y(a, b)\mathbf{j} + \mathbf{k}$$

Tangency for scalar-valued functions of two variables

We have the normal vector and a point of the tangent plane.

$$\mathbf{n} = \mathbf{u} \times \mathbf{v} = -f_x(a, b)\mathbf{i} - f_y(a, b)\mathbf{j} + \mathbf{k}; \quad P = (a, b, f(a, b))$$

- So, the **equation for the tangent plane** through $(a, b, f(a, b))$ with normal \mathbf{n} is

$$(-f_x(a, b), -f_y(a, b), 1) \cdot (x - a, y - b, z - f(a, b)) = 0$$

or

$$-f_x(a, b)(x - a) - f_y(a, b)(y - b) + z - f(a, b) = 0$$

Theorem 3.3

- If the graph of $z = f(x, y)$ has a tangent plane at $(a, b, f(a, b))$ then that tangent plane has equation

$$z = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b)$$

- Let define the function $h(x, y)$ to be

$$h(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b)$$

- Then h has the following properties

1. $h(a, b) = f(a, b)$

The values of h and f
are the same at (a, b)

2. $\frac{\partial h}{\partial x}(a, b) = \frac{\partial f}{\partial x}(a, b)$ and $\frac{\partial h}{\partial y}(a, b) = \frac{\partial f}{\partial y}(a, b)$

Partial derivatives of h and f
are the same at (a, b)

Definition 3.4: Differentiability

- Let X be open in \mathbb{R}^2 .
- Let $f : X \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$ be a scalar-valued function of two variables.
- We say that f is **differentiable** at $(a, b) \in X$ if
 - The partial derivatives $f_x(a, b)$ and $f_y(a, b)$ exist, and
 - The function

$$h(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b)$$

is a good linear approximation to f near (a, b) :

$$\lim_{(x,y) \rightarrow (a,b)} \frac{f(x, y) - h(x, y)}{\|(x, y) - (a, b)\|} = 0$$

Definition 3.4: Differentiability

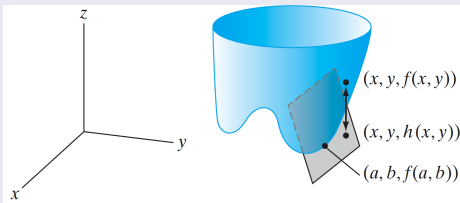
- Mathematically is not necessary to suppose the partial derivatives exists. It is enough to say that f is differentiable if exists a **linear function** h that:

$$\lim_{(x,y) \rightarrow (a,b)} \frac{f(x,y) - h(x,y)}{\|(x,y) - (a,b)\|} = 0$$

- If f is differentiable at (a, b) , then the equation $z = h(x, y)$ defines the **tangent plane** to the graph of f at the point $(a, b, f(a, b))$.

If f is differentiable at all points of its domain,
then we say that f is **differentiable**

Definition 3.4: Differentiability

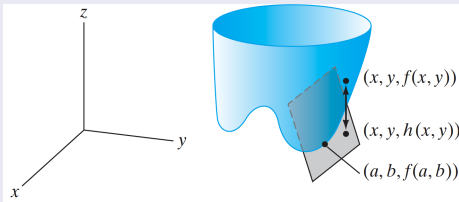


$$h(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b)$$

$$\lim_{(x,y) \rightarrow (a,b)} \frac{f(x, y) - h(x, y)}{\|(x, y) - (a, b)\|} = 0$$

- To say that $z = f(x, y)$ has a tangent plane at $(a, b, f(a, b))$ is to say that f is differentiable at (a, b) .

Definition 3.4: Differentiability



$$h(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b)$$

$$\lim_{(x, y) \rightarrow (a, b)} \frac{f(x, y) - h(x, y)}{\|(x, y) - (a, b)\|} = 0$$

- The **vertical distance** between the graph of f and the tangent plane $z = h(x, y)$ **must approach zero faster** than the point (x, y) approaches (a, b) .

The limit condition can be difficult to apply in practice. Hence, this theorem could be useful.

Theorem 3.5

- Suppose X is open in \mathbb{R}^2 .
- If $f : X \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$ has continuous partial derivatives in a neighbourhood of (a, b) in X , then f is differentiable at (a, b) .

Example 6

- Let $f(x, y) = x^2 + 2y^2$
- Then

$$\frac{\partial f}{\partial x} = 2x \quad \text{and} \quad \frac{\partial f}{\partial y} = 4y$$

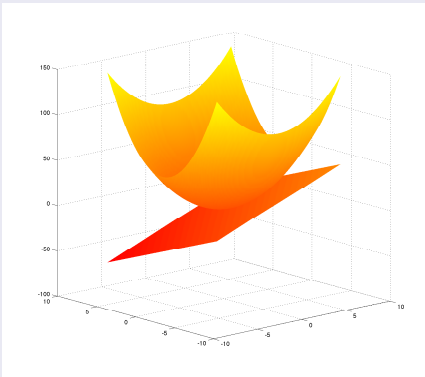
Example 6

- Let $f(x, y) = x^2 + 2y^2$
- Thus, [Theorem 3.5](#) implies that f is differentiable everywhere.
- The surface $z = x^2 + 2y^2$ must have a tangent plane at every point ,
- At the point $(2, -1)$, for example, this tangent plane is given by the equation:

$$z = 6 + 4(x - 2) - 4(y + 1) \quad \text{or} \quad 4x - 4y - z = 6$$

Example 6

- Let $f(x, y) = x^2 + 2y^2$
- At the point $(2, -1)$, for example, this tangent plane is
$$z = 6 + 4(x - 2) - 4(y + 1) \quad \text{or} \quad 4x - 4y - z = 6$$



Theorem 3.6

- If $f : X \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$ is differentiable at (a, b) , then it is continuous at (a, b) .

Example 7

- Let the function $f : X \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$ be defined by

$$f(x, y) = \begin{cases} \frac{x^2 y^2}{x^4 + y^4} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0) \end{cases}$$

- The function f is not continuous at the origin, since

$$\lim_{(x,y) \rightarrow (0,0)} f(x, y) \text{ does not exist}$$

- However, f is continuous everywhere else in \mathbb{R}^2 .
- By [Theorem 3.6](#), f cannot be differentiable at the origin.

Theorem 3.6

- If $f : X \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$ is differentiable at (a, b) , then it is continuous at (a, b)

Example 7

$$f(x, y) = \begin{cases} \frac{x^2 y^2}{x^4 + y^4} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0) \end{cases}$$

- Nonetheless, the partial derivatives of f do exist at the origin since the partial functions are constant

$$f(x, 0) = \frac{0}{x^4 + 0} \equiv 0 \Rightarrow \frac{\partial f}{\partial x}(0, 0) = 0$$

$$f(0, y) = \frac{0}{0 + y^4} \equiv 0 \Rightarrow \frac{\partial f}{\partial y}(0, 0) = 0$$

The existence of partial derivatives alone is not enough

Differentiation Remarks

We have the following hierarchy:

*Continuous partials \Rightarrow Differentiable \Rightarrow Continuous function and
Partials exist (but not necessary continuous)*

The inverse implications doesn't follow. To prove it you can take this example for the first:

$$f(x) = x^2 \sin(1/x), \quad f(0) = 0.$$

And for the second:

$$f(x, y) = xy/\sqrt{x^2 + y^2}, \quad f(0, 0) = 0$$

(continuous function).

Outline

- 1 The Derivative - Section 2.3
 - Partial derivatives
 - Differentiability
 - Matrix notation and differentiability in \mathbb{R}^n

Generalisation of scalar-valued function in \mathbb{R}^n

- Let X be open in \mathbb{R}^n .
- Let $f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ be a scalar-valued function.
- Let $\mathbf{a} = (a_1, a_2, \dots, a_n) \in X$.
- We say that f is **differentiable** at \mathbf{a} if
 - All the partial derivatives $f_{x_i}(\mathbf{a})$, $i = 1, \dots, n$, exist, and
 - The function $h : \mathbb{R}^n \rightarrow \mathbb{R}$ defined by

$$h(\mathbf{x}) = f(\mathbf{a}) + f_{x_1}(\mathbf{a})(x_1 - a_1) + f_{x_2}(\mathbf{a})(x_2 - a_2) + \dots + f_{x_n}(\mathbf{a})(x_n - a_n)$$

is a good linear approximation to f near \mathbf{a}

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

Matrix Notation and Gradient

- Let $f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ be a scalar-valued function.
- We define the **gradient** of f to be the vector ,

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

- Consequently,

$$\nabla f(\mathbf{a}) = (f_{x_1}(\mathbf{a}), f_{x_2}(\mathbf{a}), \dots, f_{x_n}(\mathbf{a}))$$

- Alternatively, we can use matrix notation and define the **derivative** of f at \mathbf{a} .

The **derivative** of f at \mathbf{a} , $Df(\mathbf{a})$, is the row matrix whose entries are the components of $\nabla f(\mathbf{a})$

Matrix of partial derivatives for vector-valued functions

- Let X be open in \mathbb{R}^n .
- Let $\mathbf{f} : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a vector-valued function of n variables.
- We define the **matrix of partial derivatives** of \mathbf{f} , denoted $D\mathbf{f}$ and called the **Jacobian Matrix**. This is the $m \times n$ matrix whose ij th entry is:

$$\frac{\partial f_i}{\partial x_j}$$

where $f_i : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ is the i th component function of \mathbf{f} .

$$D\mathbf{f}(x_1, x_2, \dots, x_n) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \cdots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

Matrix of partial derivatives for vector-valued functions

$$D\mathbf{f}(x_1, x_2, \dots, x_n) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \cdots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

- The i th row of $D\mathbf{f}$ is nothing more than Df_i .
- The entries of Df_i are precisely the components of the gradient vector ∇f_i .
- In the case where $m = 1$, $\nabla \mathbf{f}$ and $D\mathbf{f}$ mean exactly the **same thing**.

Definition 3.8: Grand Definition of Differentiability

- Let X be open in \mathbb{R}^n and let $\mathbf{a} \in X$
- Let $\mathbf{f} : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$
- We say that \mathbf{f} is **differentiable** at \mathbf{a} if
 - $D\mathbf{f}(\mathbf{a})$ exists, and
 - The function $\mathbf{h} : \mathbb{R}^n \rightarrow \mathbb{R}^m$ defined by

$$\mathbf{h}(\mathbf{x}) = \mathbf{f}(\mathbf{a}) + D\mathbf{f}(\mathbf{a})(\mathbf{x} - \mathbf{a})$$

is a good linear approximation to \mathbf{f} near \mathbf{a}

$$\lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{\|\mathbf{f}(\mathbf{x}) - \mathbf{h}(\mathbf{x})\|}{\|\mathbf{x} - \mathbf{a}\|} = \lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{\|\mathbf{f}(\mathbf{x}) - [\mathbf{f}(\mathbf{a}) + D\mathbf{f}(\mathbf{a}) \cdot (\mathbf{x} - \mathbf{a})]\|}{\|\mathbf{x} - \mathbf{a}\|} = 0$$

- The term $D\mathbf{f}(\mathbf{a})(\mathbf{x} - \mathbf{a})$ should be interpreted as the product of the $m \times n$ matrix $D\mathbf{f}(\mathbf{a})$ and the $n \times 1$ column matrix

$$\begin{bmatrix} x_1 - a_1 & x_2 - a_2 & \cdots & x_n - a_n \end{bmatrix}^T$$

Theorem 3.11

A function $\mathbf{f} : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at $\mathbf{a} \in X$

(in the sense of [Definition 3.8](#))

if and only if

Each of its component functions $f_i : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}, i = 1, \dots, m$, is differentiable at \mathbf{a}

(in the sense of [Definition 3.7](#))

