

1. MORE ON DIFFERENTIABILITY, DIFFERENTIALS AND LINEAR APPROXIMATION.

Let  $m$  and  $n$  be positive integers.

2. STANDARD BASIS VECTORS.

**Definition 2.1.** For each  $j \in \{1, \dots, n\}$  let

$$\mathbf{e}_j$$

be the vector in  $\mathbb{R}^n$  all of whose components are zero except the  $j$ -th which is one.

Thus if

$$\mathbf{x} = (x_1, \dots, x_n) \in \mathbb{R}^n$$

then

$$\mathbf{x} = \sum_{j=1}^n x_j \mathbf{e}_j.$$

3. LINEAR FUNCTIONS.

**Definition 3.1.** We say

$$L : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

is **linear** if whenever  $c \in \mathbb{R}$  and  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$  we have

- (i)  $\mathbf{L}(c\mathbf{x}) = c\mathbf{L}(\mathbf{x})$ ;
- (ii)  $\mathbf{L}(\mathbf{x} + \mathbf{y}) = \mathbf{L}(\mathbf{x}) + \mathbf{L}(\mathbf{y})$ .

Suppose  $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is linear and  $\mathbf{x} = (x_1, \dots, x_n) \in \mathbb{R}^n$ . Then

$$\begin{aligned} L(\mathbf{x}) &= L\left(\sum_{j=1}^n x_j \mathbf{e}_j\right) \\ (1) \quad &= \sum_{j=1}^n L(x_j \mathbf{e}_j) \\ &= \sum_{j=1}^n x_j L(\mathbf{e}_j). \end{aligned}$$

Thus  $L$  is determined by its values on the vectors  $\mathbf{e}_j$ ,  $j \in \{1, \dots, n\}$ . Conversely, if  $\mathbf{w}_j \in \mathbb{R}^m$ ,  $j \in \{1, \dots, n\}$ , and we define

$$L(\mathbf{x}) = \sum_{j=1}^n x_j \mathbf{w}_j \quad \text{for } \mathbf{x} \in \mathbb{R}^n$$

one easily verifies that  $L$  is linear and  $L(\mathbf{e}_j) = \mathbf{w}_j$ .

4. THE DIFFERENTIAL; THE GENERAL CASE.

Suppose  $m$  and  $n$  are positive integers,  $A \subset \mathbb{R}^n$ ,

$$f : A \rightarrow \mathbb{R}^m$$

and  $\mathbf{a} \in \text{int } A$ . (Previously  $m = 1$ .)

**Definition 4.1. (Partial derivatives.)** For each  $j \in \{1, \dots, n\}$  we let

$$\partial_j f(\mathbf{a}) = \lim_{t \rightarrow 0} \frac{f(\mathbf{a} + t\mathbf{e}_j) - f(\mathbf{a})}{t} \in \mathbb{R}^m.$$

**Remark 4.1.** Note that if  $m = 1$ ,  $n = 2$  and  $\mathbf{a} = (a, b)$  then

$$\partial_1 f(a, b) = \frac{\partial f}{\partial x}(a, b), \quad \partial_2 f(a, b) = \frac{\partial f}{\partial y}(a, b).$$

Note that if  $m = 1$ ,  $n = 3$  and  $\mathbf{a} = (a, b, c)$  then

$$\partial_1 f(a, b, c) = \frac{\partial f}{\partial x}(a, b, c), \quad \partial_2 f(a, b, c) = \frac{\partial f}{\partial y}(a, b, c), \quad \partial_3 f(a, b, c) = \frac{\partial f}{\partial z}(a, b, c).$$

**Definition 4.2.** We say  $f$  is **differentiable at  $\mathbf{a}$**  if there exists a linear map  $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$  such that

$$(2) \quad \lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{|f(\mathbf{x}) - f(\mathbf{a}) - L(\mathbf{x} - \mathbf{a})|}{|\mathbf{x} - \mathbf{a}|} = 0.$$

The linear map  $L$  is immediately seen to be unique; it is called the **differential of  $f$  at  $\mathbf{a}$**  and is written

$$\partial f(\mathbf{a}).$$

Suppose

$$\mathbf{v} = (v_1, \dots, v_n) \in \mathbb{R}^n$$

and  $f$  is differentiable at  $\mathbf{a}$ . Then

$$\partial f(\mathbf{a})(\mathbf{v}) = \sum_{j=1}^n v_j \partial_j f(\mathbf{a}).$$

In case  $m = 1$

$$\partial f(\mathbf{a})(\mathbf{v}) = df(\mathbf{a})(\mathbf{v}) = \nabla f(\mathbf{a}) \bullet \mathbf{v}.$$

If we let

$$A(\mathbf{x}) = f(\mathbf{a}) + \partial f(\mathbf{a})(\mathbf{x} - \mathbf{a}) \quad \text{and} \quad \mathbf{e}(\mathbf{x}) = f(\mathbf{x}) - A(\mathbf{x})$$

for  $\mathbf{x} \in A$  we find that (2) is equivalent to

$$(3) \quad \lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{|e(\mathbf{x})|}{|\mathbf{x} - \mathbf{a}|} = 0.$$

One calls the function  $A$  the **standard affine approximation to  $f$  at  $\mathbf{a}$** . The difference  $e = f - A$  is the error in using  $A$  to approximate  $f$ .

**Theorem 4.1.** Suppose for some  $r > 0$  the function  $f$  has partial derivatives on  $A \cap \mathbf{U}(\mathbf{a}, r)$  which are continuous at  $\mathbf{a}$ . Then  $f$  is differentiable at  $\mathbf{a}$  and

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

*Proof.* Apply the corresponding result in the scalar case which we have already obtained to each component of  $f$ .  $\square$

**Theorem 4.2. (The Chain Rule.)** Suppose

- (i)  $f$  is differentiable at  $\mathbf{a}$ ;
- (ii)  $l$  is a positive integer,  $B$  is a subset of  $\mathbb{R}^m$  and  $g : B \rightarrow \mathbb{R}^l$ ;
- (iii)  $f(\mathbf{a}) \in \text{int } B$  and  $g$  is differentiable at  $f(\mathbf{a})$ .

Then  $\mathbf{a}$  is an interior point of the domain of  $g \circ f$ ,  $g \circ f$  is differentiable at  $\mathbf{a}$  and

$$(4) \quad \partial(g \circ f)(\mathbf{a}) = \partial g(f(\mathbf{a})) \circ \partial f(\mathbf{a}).$$

*Proof.* Use the affine approximation approximations of  $f$  at  $\mathbf{a}$  and  $g$  near  $f(\mathbf{a})$  to obtain an affine approximation of  $g \circ f$  near  $f(\mathbf{a})$ .  $\square$

**Remark 4.2. (What to remember about (4)).** Suppose

$$f = (f_1, \dots, f_m) \quad \text{and} \quad g = (g_1, \dots, g_l).$$

Then (4) is equivalent to

$$(5) \quad \partial_j(g_i \circ f)(\mathbf{a}) = \sum_{k=1}^m \partial_k g_i(f(\mathbf{a})) \partial_j f_k(\mathbf{a}) = \nabla g_i(f(\mathbf{a})) \bullet \partial_j f(\mathbf{a})$$

for  $i = 1, \dots, l$  and  $j = 1, \dots, n$ .

**Example 4.1.** Here is an example of the Chain Rule when  $n = 1$ ,  $m = 2$  and  $l = 1$ . Suppose  $U \subset \mathbb{R}^2$ ,  $U$  is open and  $u : U \rightarrow \mathbb{R}$  is continuously differentiable on  $u$ ;  $u$  will correspond to  $g$  above. Suppose  $P : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  is defined by

$$P(r, \theta) = (r \cos \theta, r \sin \theta), \quad (r, \theta) \in \mathbb{R}^2;$$

$P$  will correspond to  $f$  above.

We have

$$(6) \quad \begin{aligned} \frac{\partial}{\partial r} u(r \cos \theta, r \sin \theta) &= \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) \frac{\partial}{\partial r} r \cos \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) \frac{\partial}{\partial r} r \sin \theta \\ &= \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) \cos \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) \sin \theta \end{aligned}$$

as well as

$$(7) \quad \begin{aligned} \frac{\partial}{\partial \theta} u(r \cos \theta, r \sin \theta) &= \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) \frac{\partial}{\partial \theta} r \cos \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) \frac{\partial}{\partial \theta} r \sin \theta \\ &= -\frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) r \sin \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) r \cos \theta. \end{aligned}$$

If you thought that was tough wait till you see what comes next! We have

$$(8) \quad \begin{aligned} \frac{\partial^2}{\partial r^2} u(r \cos \theta, r \sin \theta) &= \frac{\partial}{\partial r} \left( \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) \cos \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) \sin \theta \right) \\ &= \frac{\partial}{\partial r} \left( \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) \right) \cos \theta + \frac{\partial}{\partial r} \left( \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) \right) \sin \theta \\ &= \left( \frac{\partial^2 u}{\partial x^2}(r \cos \theta, r \sin \theta) \cos \theta + \frac{\partial^2 u}{\partial y \partial x}(r \cos \theta, r \sin \theta) \sin \theta \right) \cos \theta \\ &\quad + \left( \frac{\partial^2 u}{\partial x \partial y}(r \cos \theta, r \sin \theta) \cos \theta + \frac{\partial^2 u}{\partial y^2}(r \cos \theta, r \sin \theta) \sin \theta \right) \sin \theta \\ &= \frac{\partial^2 u}{\partial x^2}(r \cos \theta, r \sin \theta) \cos^2 \theta \\ &\quad + 2 \frac{\partial^2 u}{\partial x \partial y}(r \cos \theta, r \sin \theta) \cos \theta \sin \theta \\ &\quad + \frac{\partial^2 u}{\partial y^2}(r \cos \theta, r \sin \theta) \sin^2 \theta. \end{aligned}$$

as well as

(9)

$$\begin{aligned}
\frac{\partial^2}{\partial \theta^2} u(r \cos \theta, r \sin \theta) &= \frac{\partial}{\partial \theta} \left( -\frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) r \sin \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) r \cos \theta \right) \\
&= -\frac{\partial}{\partial \theta} \left( \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) \right) r \sin \theta - \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) \frac{\partial}{\partial \theta} r \sin \theta \\
&\quad + \frac{\partial}{\partial \theta} \left( \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) \right) r \cos \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) \frac{\partial}{\partial \theta} r \cos \theta \\
&= - \left( -\frac{\partial^2 u}{\partial x^2}(r \cos \theta, r \sin \theta) r \sin \theta + \frac{\partial^2 u}{\partial y \partial x}(r \cos \theta, r \sin \theta) r \cos \theta \right) r \sin \theta \\
&\quad - \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) r \cos \theta \\
&\quad + \left( -\frac{\partial^2 u}{\partial x \partial y}(r \cos \theta, r \sin \theta) r \sin \theta + \frac{\partial^2 u}{\partial y^2}(r \cos \theta, r \sin \theta) r \cos \theta \right) r \cos \theta \\
&\quad - \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) r \sin \theta \\
&= - \left( \frac{\partial u}{\partial x}(r \cos \theta, r \sin \theta) r \cos \theta + \frac{\partial u}{\partial y}(r \cos \theta, r \sin \theta) r \sin \theta \right) \\
&\quad + \frac{\partial^2 u}{\partial x^2}(r \cos \theta, r \sin \theta) r^2 \sin^2 \theta - 2 \frac{\partial^2 u}{\partial x \partial y}(r \cos \theta, r \sin \theta) r^2 \sin \theta \cos \theta \\
&\quad + \frac{\partial^2 u}{\partial y^2}(r \cos \theta, r \sin \theta) r^2 \cos^2 \theta.
\end{aligned}$$

Putting it all together we get

$$\left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right) (r \cos \theta, r \sin \theta) = \left( \frac{\partial^2}{\partial r^2} + \frac{1}{r} \frac{\partial}{\partial r} + \frac{\partial^2}{\partial \theta^2} \right) u(r \cos \theta, r \sin \theta).$$

So, for example, if

$$u(x, y) = \log \sqrt{x^2 + y^2} \quad \text{for } (x, y) \neq (0, 0)$$

in which case

$$u(r \cos \theta, r \sin \theta) = \log r \quad \text{for } r \neq 0$$

or if

$$u(x, y) = \arcsin \frac{y}{\sqrt{x^2 + y^2}} \quad \text{for } (x, y) \in \mathbb{R}^2 \text{ such that } x > 0 \text{ if } y = 0$$

in which case

$$u(r \cos \theta, r \sin \theta) = \theta \quad \text{if } r > 0 \text{ and } -\pi < \theta < \pi$$

then

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = 0.$$

The operator

$$\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2}$$

is called the **Laplacian**; it is, by far, the most important partial differential operator in mathematics and physics. (Typically, an *operator* is a function whose domain is a set of functions and whose range is a set of functions.) To define it the way we

did above turns out to be not so good an idea; there are much better definitions but they require a bit of machinery which we will develop later.