

# Longo: Math 20C - Winter 2017

## Lecture Notes

Date: January 26, 2017

Section:

- §2.3 (part 2)

Topics Covered:

- Linear approximation
- The meaning of differentiability, and a criterion for differentiability
- The total derivative

From last time: If  $f: U \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$  is "nice" (i.e., the graph is smooth) then the equation  $z = f(a,b) + \frac{\partial f}{\partial x}(a,b)(x-a) + \frac{\partial f}{\partial y}(a,b)(y-b)$  should describe the plane tangent to the graph at  $(a,b)$ .

Example: Find the eqn of the plane tangent to the graph of  $f(x,y) = x^2 - 2xy + y^2$  at the point where  $x=1, y=2$

Sol.

- $\frac{\partial f}{\partial x} = 2x - 2y \Rightarrow \frac{\partial f}{\partial x}(1,2) = 2 - 2 \cdot 2 = 2 - 4 = -2$
- $\frac{\partial f}{\partial y} = -2x + 2y \Rightarrow \frac{\partial f}{\partial y}(1,2) = -2(1) + 2(2) = -2 + 4 = 2$
- $f(1,2) = 1^2 - 2(1)(2) + (2)^2 = 1 - 4 + 4 = 1$

$$\Rightarrow \boxed{z = 1 - 2(x-1) + 2(y-2)}$$

Differentiability ( $f: \mathbb{R}^2 \rightarrow \mathbb{R}$  case) & linear approximation:

Let  $f: U \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $(a,b)$  be a point in the domain of  $f$ . Assume  $\frac{\partial f}{\partial x}(a,b), \frac{\partial f}{\partial y}(a,b)$  exist.

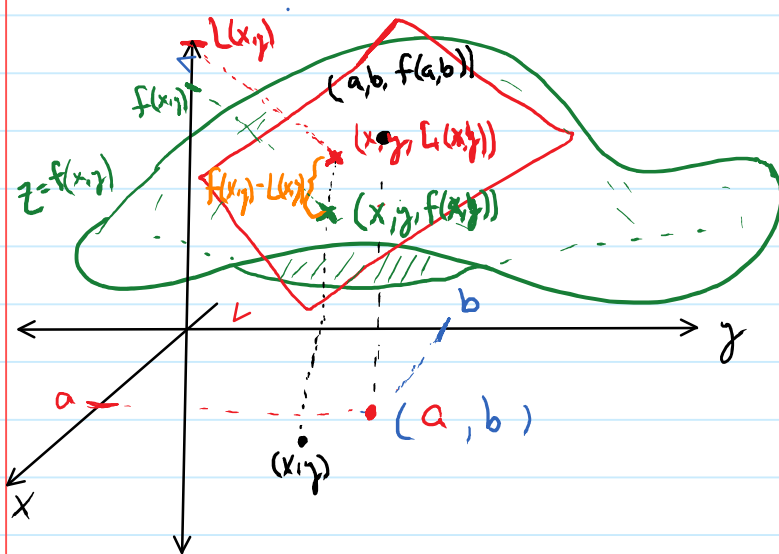
$$\text{Let } L(x,y) = f(a,b) + \frac{\partial f}{\partial x}(a,b)(x-a) + \frac{\partial f}{\partial y}(a,b)(y-b).$$

$L(x,y)$  is the fn whose graph is the plane tangent to the graph of  $f$  at  $(a,b)$ .

We saw last time that a fn of one variable,  $g(x)$ , differentiable at a point  $x=a$  iff the tangent line is a "good approximation" for  $g$ . We say  $f(x,y)$  is differentiable at  $(a,b)$  if  $f(x,y) \approx L(x,y)$  if  $(x,y)$  is close to  $(a,b)$ . More precisely:

Definition: In the above setting,  $f$  is differentiable at  $(a,b)$  if

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



Remark: We will call  $L(x,y)$  the linear approximation of  $f$  near  $(a,b)$

Problem: It is hard to check property  $(*)$  directly. Luckily, we have:

Thm: If  $\frac{\partial f}{\partial x}$ ,  $\frac{\partial f}{\partial y}$  exist and are continuous "near"  $(a,b)$ , then  $f$  is differentiable at  $(a,b)$ .

Remark: A fn that has continuous partials is called  $C^1$ . This thm says  $C^1 \Rightarrow$  differentiable.

Example: Estimate  $(0.99)^2 - 2(0.99)(2.01) + (2.01)^2$  using linear approximation.

Sol: Let  $f(x,y) = x^2 - 2xy + y^2$ . We want to estimate  $f(0.99, 2.01)$ . To do this, we use Linear approximation near  $(1, 2)$ .

We saw earlier:  $\frac{\partial f}{\partial x} = 2x - 2y$ ,  $\frac{\partial f}{\partial y} = -2x + 2y$ , which are continuous everywhere. By the Thm,  $f$  is diff'ble, so  $f(x,y) \approx L(x,y) = f(1,2) + \frac{\partial f}{\partial x}(1,2)(x-1) + \frac{\partial f}{\partial y}(1,2)(y-2)$

near  $(1,2)$ . We calculated earlier:

$$L(x,y) = 1 - 2(x-1) + 2(y-2).$$

Since  $(0.99, 2.01)$  is close to  $(1,2)$

$$f(0.99, 2.01) \approx L(0.99, 2.01) = 1 - 2(-0.01) + 2(0.01) = 1 + 0.02 + 0.02$$

$$= 1.04$$


Hint: Existence of partials is not good enough for differentiability.  
 for ex.  $f(x,y) = \begin{cases} \frac{x^2 y}{\sqrt{x^2 + y^2}} & \text{if } (x,y) \neq (0,0) \\ 0 & \text{if } (x,y) = (0,0) \end{cases}$  has partials everywhere,

but is not diff'ble at  $(0,0)$  (see book pg. 114 for picture).

### Differentiability (general case) and the total derivative (a survey):

Let  $f(x,y)$  be diff'ble at  $(a,b)$ ,  $L(x,y) = f(a,b) + \frac{\partial f}{\partial x}(a,b)(x-a) + \frac{\partial f}{\partial y}(a,b)(y-b)$  be the linear approximation near  $(a,b)$ .

Let  $\nabla f(a,b) = \begin{bmatrix} \frac{\partial f}{\partial x}(a,b) & \frac{\partial f}{\partial y}(a,b) \end{bmatrix}$  be the  $1 \times 2$  matrix of partial derivatives. Then,  $L(x,y) = f(a,b) + \nabla f(a,b) \cdot (x-a, y-b)$

dot product 

This should look familiar: If  $g(x)$  is a one variable diff'ble fun,  $y = g(a) + g'(a)(x-a)$  describes the tangent line at  $x=a$ . In  $(*)$ ,  $\nabla f(a,b)$  is playing the same roll as the derivative in the single variable case!

$\nabla f = \left( \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right)$  is called the gradient of  $f$ .

Rmk: Intuitively, in order to fully encapsulate how  $f$  changes w.r.t. to a deviation from  $(a,b)$ , we need to know how  $f$  changes w.r.t.  $x$  and  $y$ . So we need a matrix to hold multiple pieces of information.

More generally: Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ . Then if  $\vec{x} = (x_1, x_2, \dots, x_n)$  is an arbitrary point in the domain, we write

$$f(\vec{x}) = (f_1(\vec{x}), f_2(\vec{x}), \dots, f_m(\vec{x}))$$

*each of these is a function of  $n$ -variables*

Example:  $f: \mathbb{R}^2 \rightarrow \mathbb{R}^3$ ,  $f(x,y) = \left( \underline{2x+y}, \underline{3xy}, \underline{x^2-y^2} \right)$   
 $f_1 \quad f_2 \quad f_3$

Assume each of the  $f_i$ 's has continuous partial derivatives for each input variable. Then the theorem discussed before says  $f$  is differentiable. What does this mean?

Let  $\vec{a} = (a_1, \dots, a_n)$  be a vector in  $\mathbb{R}^n$  (the domain of  $f$ ), and let

$$[Df](\vec{a}) = \begin{bmatrix} \frac{\partial f_1(\vec{a})}{\partial x_1} & \frac{\partial f_1(\vec{a})}{\partial x_2} & \dots & \frac{\partial f_1(\vec{a})}{\partial x_n} \\ \frac{\partial f_2(\vec{a})}{\partial x_1} & \dots & \dots & \frac{\partial f_2(\vec{a})}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_m(\vec{a})}{\partial x_1} & \dots & \dots & \frac{\partial f_m(\vec{a})}{\partial x_n} \end{bmatrix}$$

Then for all points  $\vec{x}$  "close to"  $\vec{a}$  (i.e.,  $\|\vec{x} - \vec{a}\|$  is small),

$$(**) \quad f(x) \approx f(\vec{a}) + [Df](\vec{a}) \cdot (\vec{x} - \vec{a})$$



In this equation, we view these as column vectors so that matrix multiplication makes sense.

Here, the matrix of partial derivatives  $[Df](\vec{a})$  is called the (total) derivative or the differential of  $f$  at  $\vec{a}$ .

In the special case  $m=1$ , and we have  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ ,

$[Df](\vec{x})$  is just the gradient:  $\nabla f$ .

- Remk:
- ① We will use the total derivative when we talk about the multivariable chain rule.
  - ② We will only really consider fns  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  where  $n, m \leq 3$ .

Example from before:

Let  $f: \mathbb{R}^2 \rightarrow \mathbb{R}^3$ ,  $f(x, y) = (2x+y, 3xy, x^2-y^2)$ .

Then  $[Df]_{(x,y)} = \begin{pmatrix} 2 & 1 \\ 3y & 3x \\ 2x & -2y \end{pmatrix}$

(\*\*) More precisely:

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

Note that: ① the numerator is a vector in  $\mathbb{R}^m$ , so

" $f(\vec{x}) \approx f(\vec{a}) + [Df](\vec{a}) \cdot (\vec{x} - \vec{a})$ " means the difference vector has small magnitude.

② This limit criterion is the official condition for a function to be differentiable.