# Lecture 6: Chain rule, Mean Value Theorem, Tangent Plane

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#### Chain rule

Theorem-A: Let  $\mathbf{x} : \mathbb{R} \to \mathbb{R}^n$  be differentiable at  $t_0$  and  $f : \mathbb{R}^n \to \mathbb{R}$  be differentiable at  $\mathbf{a} := \mathbf{x}(t_0)$ . Then  $f \circ \mathbf{x}$  is differentiable at  $t_0$  and

$$\frac{\mathrm{d}}{\mathrm{d}t}f(\mathbf{x})|_{t=t_0} = \nabla f(\mathbf{a}) \bullet \mathbf{x}'(t_0) = \sum_{i=1}^n \partial_i f(\mathbf{a}) \frac{\mathrm{d}x_i(t_0)}{\mathrm{d}t}.$$

Proof: Use

 $f(\mathbf{x}(t)) = f(\mathbf{a}) + \nabla f(\mathbf{a}) \bullet (\mathbf{x}(t) - \mathbf{x}(t_0)) + e(\mathbf{x}) \|\mathbf{x} - \mathbf{a}\|$ 

and the fact that  $e(\mathbf{x}(t)) \rightarrow 0$  as  $t \rightarrow 0$ .

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### Chain rule for partial derivatives

Theorem-B: If  $\mathbf{x} : \mathbb{R}^2 \to \mathbb{R}^n$  has partial derivatives at (a, b)and  $f : \mathbb{R}^n \to \mathbb{R}$  is differentiable at  $\mathbf{p} := \mathbf{x}(a, b)$  then

$$\partial_{u}f(\mathbf{p}) = \nabla f(\mathbf{p}) \bullet \partial_{u}\mathbf{x}(a,b) = \sum_{j=1}^{n} \frac{\partial_{j}f(\mathbf{p})}{\partial x_{j}} \frac{\partial x_{j}(a,b)}{\partial u},$$
  
$$\partial_{v}f(\mathbf{p}) = \nabla f(\mathbf{p}) \bullet \partial_{v}\mathbf{x}(a,b) = \sum_{j=1}^{n} \frac{\partial_{j}f(\mathbf{p})}{\partial x_{j}} \frac{\partial x_{j}(a,b)}{\partial v}.$$

Proof: Apply Theorem-A.

Example: Find  $\partial w/\partial u$  and  $\partial w/\partial v$  when  $w = x^2 + xy$  and  $x = u^2v$ ,  $y = uv^2$ .

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### Mean Value Theorem

Theorem: Let  $f : \mathbb{R}^n \to \mathbb{R}$  be differentiable. Let  $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$ . Then there exists  $\mathbf{c} \in [\mathbf{a}, \mathbf{b}]$  such that

$$f(\mathbf{b}) - f(\mathbf{a}) = \nabla f(\mathbf{c}) \bullet (\mathbf{b} - \mathbf{a}),$$

where  $[\mathbf{a}, \mathbf{b}] := {\mathbf{a}(1-t) + t\mathbf{b} : t \in [0,1]}$  is the line segment joining  $\mathbf{a}$  and  $\mathbf{b}$ .

**Proof:** Consider  $\phi(t) := f(\mathbf{a}(1-t) + t\mathbf{b})$  for  $t \in [0, 1]$  and invoke chain rule.

### Total derivative

Let  $\mathbf{x}: \mathbb{R} \to \mathbb{R}^n$  and  $f: \mathbb{R}^n \to \mathbb{R}$  be differentiable. Then

$$\frac{\mathrm{d}}{\mathrm{d}t}f(\mathbf{x}(t)) = \nabla f(\mathbf{x}) \bullet \mathbf{x}'(t) = \sum_{i=1}^n \partial_i f(\mathbf{x}) \frac{\mathrm{d}x_i}{\mathrm{d}t}$$

is called total derivative of f.

Example: Consider 
$$f(x, y) := x^2 - y^2$$
 and  
 $(x(t), y(t)) := (\sin(t), \cos(t))$ . Then  
 $\frac{\mathrm{d}}{\mathrm{d}t}f(\mathbf{x}(t)) = \nabla f(\mathbf{x}) \bullet \mathbf{x}'(t) = (2x, -2y) \bullet (\cos t, -\sin t)$   
 $= 2\sin(t)\cos(t) + 2\cos(t)\sin(t) = 2\sin(2t)$ .

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### Differential

Let  $f : \mathbb{R}^n \to \mathbb{R}$  be differentiable. Then

$$\mathrm{d}f = \sum_{i=1}^n \partial_i f(\mathbf{a}) \mathrm{d}x_i$$

is called differential of f at  $\mathbf{a}$ .

For n = 1 this gives familiar differential df = f'(a)dx.

Question: What is df? What does it represent?

Well,  $df = Df(\mathbf{a})$ , the derivative map. The differential df is a fancy way of writing the derivative map

$$Df(\mathbf{a}): \mathbb{R}^n \to \mathbb{R}, \mathbf{h} \mapsto \nabla f(\mathbf{a}) \bullet \mathbf{h}.$$

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### Differential = Derivative map

Define coordinate projection  $dx_i : \mathbb{R}^n \to \mathbb{R}$  by  $dx_i(\mathbf{h}) = h_i$ . Then

$$Df(\mathbf{a})\mathbf{h} = \nabla f(\mathbf{a}) \bullet \mathbf{h} = \sum_{j=1}^{n} \partial_{j} f(\mathbf{a}) h_{j}$$
$$= \sum_{j=1}^{n} \partial_{j} f(\mathbf{a}) dx_{j}(\mathbf{h}) = df(\mathbf{h}).$$

Hence  $Df(\mathbf{a}) = \sum_{i=1}^{n} \partial_i f(\mathbf{a}) dx_i = df$ .

Bottomline: The differential df at **a** denotes  $Df(\mathbf{a})$  when  $Df(\mathbf{a})$  is expressed in terms of the partial derivatives  $\partial_j f(\mathbf{a})$ .

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## Differential and increament

Consider an increment  $\Delta \mathbf{x} = (\Delta x_1, \dots, \Delta x_n) \in \mathbb{R}^n$  and define

$$\Delta f(\Delta \mathbf{x}) := f(\mathbf{a} + \Delta \mathbf{x}) - f(\mathbf{a}).$$

Then

$$\Delta f(\Delta \mathbf{x}) = \mathrm{d} f(\Delta \mathbf{x}) + e(\Delta \mathbf{x}) \|\Delta \mathbf{x}\| \simeq \mathrm{d} f(\Delta \mathbf{x})$$

gives

$$\Delta f(\Delta \mathbf{x}) \simeq \mathrm{d} f(\Delta \mathbf{x}) = \sum_{j=1}^n \partial_j f(\mathbf{a}) \mathrm{d} x_j(\Delta \mathbf{x}) = \sum_{j=1}^n \partial_j f(\mathbf{a}) \Delta x_j$$

Increment of f at  $\mathbf{a} \simeq \text{sum}$  of scaled increments of the components of  $\mathbf{a}$ .

#### Level sets

Let  $f : \mathbb{R}^n \to \mathbb{R}$ . Then  $G(f) := \{(\mathbf{x}, f(\mathbf{x})) : \mathbf{x} \in \mathbb{R}^n\} \subset \mathbb{R}^{n+1}$  is the graph of f. G(f) represents a curve/surface in  $\mathbb{R}^{n+1}$ .

The set  $S(f, \alpha) := {\mathbf{x} \in \mathbb{R}^n : f(\mathbf{x}) = \alpha}$  is called a level set of f and represents a curve/surface in  $\mathbb{R}^n$ .

#### Examples:

- $f(x, y) := x^2 + y^2$ . Then G(f) is a paraboloid in  $\mathbb{R}^3$  and  $S(f, \alpha)$  with  $\alpha > 0$  is a circle in  $\mathbb{R}^2$ .
- $f(x, y) := 4x^2 + y^2$ . Then G(f) is an elliptic paraboloid in  $\mathbb{R}^3$  and  $S(f, \alpha)$  with  $\alpha > 0$  is an ellipse in  $\mathbb{R}^2$ .

#### Tangent plane to level sets

Equation of a hyperplane in  $\mathbb{R}^n$  passing through **a** and a normal vector **n** is given by  $(\mathbf{x} - \mathbf{a}) \bullet \mathbf{n} = 0$ .

Equation of a line in  $\mathbb{R}^n$  passing through **a** in the direction of **v** is given by  $\mathbf{x} = \mathbf{a} + t\mathbf{v}$ .

Let  $f : \mathbb{R}^n \to \mathbb{R}$  be differentiable at  $\mathbf{a} \in \mathbb{R}^n$ .

Let  $\mathbf{x} : (-\delta, \delta) \to S(f, \alpha)$  be differentiable at 0 and  $\mathbf{x}(0) = \mathbf{a}$ . Then

$$\frac{\mathrm{d}f}{\mathrm{d}t}(\mathbf{a}) = \nabla f(\mathbf{a}) \bullet \mathbf{x}'(0) = 0 \Rightarrow \nabla f(\mathbf{a}) \perp \mathbf{x}'(0).$$

 $\Rightarrow \mathbf{x}'(\mathbf{0})$  lies in the hyperplane  $(\mathbf{x} - \mathbf{a}) \bullet \nabla f(\mathbf{a}) = \mathbf{0}$ .

#### Tangent plane

Since the velocity  $\mathbf{x}'(0)$  is tangent to the curve  $\mathbf{x}(t)$  at  $\mathbf{a}$ , the hyperplane  $(\mathbf{x} - \mathbf{a}) \bullet \nabla f(\mathbf{a}) = 0$  is tangent to the level set  $f(\mathbf{x}) = \alpha$  at  $\mathbf{a}$ .

Hence the line  $\mathbf{x} = \mathbf{a} + t\nabla f(\mathbf{a})$  is normal to the level set  $f(\mathbf{x}) = \alpha$  at  $\mathbf{a}$ .

- $(x-a)f_x(a,b) + (y-b)f_y(a,b) = 0$  is the equation of the tangent line to  $f(x,y) = \alpha$  at (a,b).
- $(x-a)f_x(a, b, c) + (y-b)f_y(a, b, c) + (z-c)f_z(a, b, c) = 0$ is the equation of the tangent plane to  $f(x, y, z) = \alpha$  at (a, b, c).

#### Tangent to the graph

Let  $f : \mathbb{R}^n \to \mathbb{R}$  be differentiable at  $\mathbf{a} \in \mathbb{R}^n$ .

Define 
$$g : \mathbb{R}^n \times \mathbb{R} \to \mathbb{R}$$
 by  $g(\mathbf{x}, z) = f(\mathbf{x}) - z$ . Then  $S(g, 0) = G(f)$  and  $\nabla g = (\nabla f, -1)$ .

The tangent plane to S(g,0) = G(f) at  $(\mathbf{a}, f(\mathbf{a}))$  is given by

$$(\mathbf{x} - \mathbf{a}, z - f(\mathbf{a})) \bullet (\nabla f(\mathbf{a}), -1) = 0$$

which gives  $z = f(\mathbf{a}) + \nabla f(\mathbf{a}) \bullet (\mathbf{x} - \mathbf{a})$ .

• 
$$y = f(a) + f'(a)(x - a)$$
 is tangent to  $y = f(x)$  at  $(a, f(a))$ .

• 
$$z = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b)$$
 is tangent  
to  $z = f(x, y)$  at  $(a, b, f(a, b))$ .

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## Differentiability of $f: U \subset \mathbb{R}^n \to \mathbb{R}^m$

**Definition:** Let  $U \subset \mathbb{R}^n$  be open. Then  $f : U \subset \mathbb{R}^n \to \mathbb{R}^m$  is differentiable at  $\mathbf{a} \in U$  if there exists a linear map  $L : \mathbb{R}^n \to \mathbb{R}^m$  such that

$$\lim_{\mathbf{h}\to 0}\frac{\|f(\mathbf{a}+\mathbf{h})-f(\mathbf{a})-\mathcal{L}(\mathbf{h})\|}{\|\mathbf{h}\|}=0.$$

The linear map L is called the derivative of f at **a** and is denoted by  $Df(\mathbf{a})$ , that is,  $L = Df(\mathbf{a})$ .

Other notations:  $f'(a), \frac{df}{dx}(a)$ .

### Characterization of differentiability

Theorem: Consider  $f : \mathbb{R}^n \to \mathbb{R}^m$  with  $f(\mathbf{x}) = (f_1(\mathbf{x}), \ldots, f_m(\mathbf{x}))$ , where  $f_i : \mathbb{R}^n \to \mathbb{R}$ . Then f is differentiable at  $\mathbf{a} \in \mathbb{R}^n \iff f_i$  is differentiable at  $\mathbf{a}$  for  $i = 1, 2, \ldots, m$ . Further

$$Df(\mathbf{a})\mathbf{h} = (\nabla f_1(\mathbf{a}) \bullet \mathbf{h}, \dots, \nabla f_m(\mathbf{a}) \bullet \mathbf{h}).$$

Proof: Blackboard.

The matrix of  $Df(\mathbf{a})$  is called the Jacobian matrix of f at  $\mathbf{a}$ . Jacobian matrix is obtained by writing  $\mathbf{x}$  and  $f(\mathbf{x})$  as column vectors.

Jacobian matrix of  $f : \mathbb{R}^n \to \mathbb{R}^m$ 

The jacobian matrix 
$$Df(\mathbf{a}) = \begin{bmatrix} -\nabla f_1(\mathbf{a})^\top - \\ \vdots \\ -\nabla f_m(\mathbf{a})^\top - \end{bmatrix}_{m \times n}$$
.

• 
$$f(x,y) = (f_1(x,y), f_2(x,y), f_3(x,y))$$
  

$$Df(a,b) = \begin{bmatrix} \partial_x f_1(a,b) & \partial_y f_1(a,b) \\ \partial_x f_2(a,b) & \partial_y f_2(a,b) \\ \partial_x f_3(a,b) & \partial_y f_3(a,b) \end{bmatrix}$$

• 
$$f(x, y, z) = (f_1(x, y), f_2(x, y))$$
  

$$Df(a, b, c) = \begin{bmatrix} \partial_x f_1(a, b, c) & \partial_y f_1(a, b, c) & \partial_z f_1(a, b, c) \\ \partial_x f_2(a, b, c) & \partial_y f_2(a, b, c) & \partial_z f_2(a, b, c) \end{bmatrix}$$

## Examples

• If 
$$f(x, y) = (xy, e^x y, \sin y)$$
 then

$$Df(x,y) = \begin{bmatrix} y & x \\ e^{x}y & e^{x} \\ 0 & \cos y \end{bmatrix}$$

• If 
$$f(x, y, z) = (x + y + z, xyz)$$
 then  

$$Df(x, y, z) = \begin{bmatrix} 1 & 1 & 1 \\ yz & xz & xy \end{bmatrix}$$

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