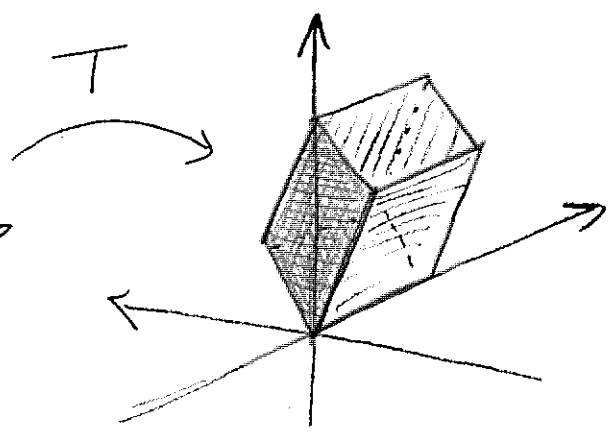
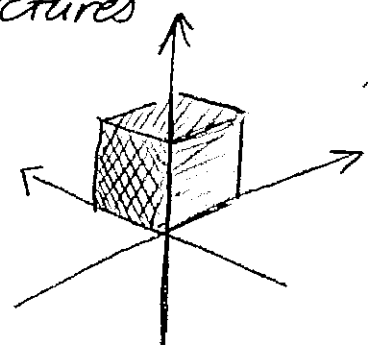
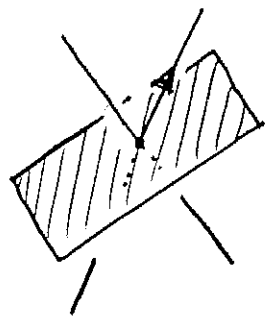
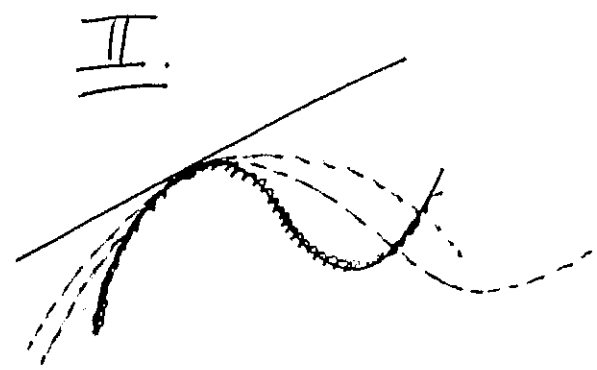
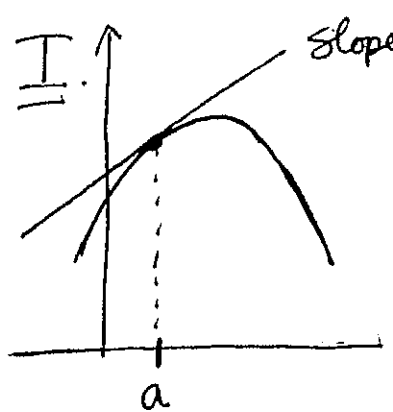


Lecture 40: Linear approximation, diagonalization, and the 2<sup>nd</sup> derivative test. ①

Linear algebra in pictures



(Differential) Calculus in pictures.



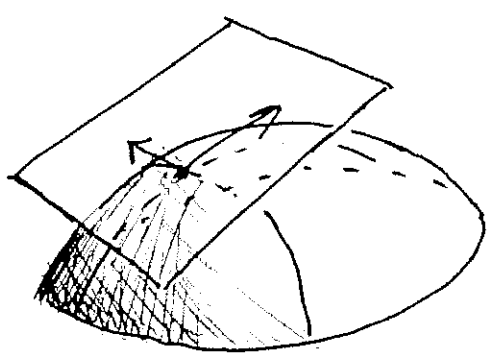
$f: \mathbb{R} \rightarrow \mathbb{R}$

III.  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$

$$f(a+h) \approx f(a) + f'(a)h$$

$$f(a+h) \approx f(a) + f'(a)h + \frac{f''(a)}{2}h^2 + \frac{f^{(3)}(a)}{3!}h^3 + \dots$$

$$\frac{\partial f}{\partial x}, \quad \frac{\partial f}{\partial y}$$



What about  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ ? Thinking of  $f$  as  $(f_1, \dots, f_m)$  with  $f_i: \mathbb{R}^n \rightarrow \mathbb{R}$  have  $\frac{\partial f_i}{\partial x_j}: \mathbb{R}^n \rightarrow \mathbb{R}$  but is there more?

(2)

Def: A function  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is differentiable at  $a \in \mathbb{R}^n$  when there exists a linear transformation

$D_a f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  such that

$$f(a+h) = f(a) + D_a f(h) + E(h)$$

where  $\lim_{h \rightarrow 0} \frac{E(h)}{\|h\|} = 0$ .

Moral:  $f$  is differentiable at  $a$  if it can be well-approximated by a linear transformation (+ const) near  $a$ .

called the  
 $\downarrow$  derivative of  
 $f$  at  $a$ .

Thm: If  $f$  is differentiable at  $a$ , then  $D_a f$  above is unique and

$$[D_a f]_{\text{Std}(\mathbb{R}^n)}^{\text{Std}(\mathbb{R}^m)} = \left( \begin{array}{ccc} \frac{\partial f_1}{\partial x_1}(a) & \dots & \frac{\partial f_1}{\partial x_n}(a) \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \dots & \frac{\partial f_m}{\partial x_n}(a) \end{array} \right) \left. \vphantom{\begin{array}{ccc} \frac{\partial f_1}{\partial x_1}(a) & \dots & \frac{\partial f_1}{\partial x_n}(a) \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \dots & \frac{\partial f_m}{\partial x_n}(a) \end{array}} \right\} \begin{array}{l} \text{Jacobian} \\ \text{matrix} \end{array}$$

Thm: If all  $\frac{\partial f_i}{\partial x_{j_1} \dots \partial x_{j_k}}$  exist on all of  $\mathbb{R}^n$ , then  $f$  is differentiable at all  $a \in \mathbb{R}^n$ .

(3)

Ex:  $f(x, y, z) = (x^2 + 3y + \sin(z), 3x + z^2, y + \cos(x + z^2))$   
is a differentiable function  $\mathbb{R}^3 \rightarrow \mathbb{R}^3$ .

Chain Rule: Suppose  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is differentiable at  $a$  and  $g: \mathbb{R}^m \rightarrow \mathbb{R}^k$  is differentiable at  $f(a)$ .  
Then  $g \circ f: \mathbb{R}^n \rightarrow \mathbb{R}^k$  is diff at  $a$  with

$$D_a(g \circ f) = (D_{f(a)}g) \circ (D_a f)$$

[Compare: Converting to Jacobian matrices, this gives the special cases of the chain rule you learned about in Calculus III.]

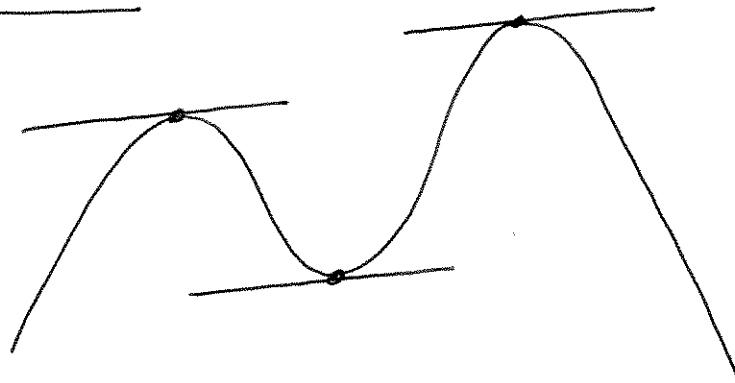
[Other connections: Change of variables formula in multivariable integration...]

## 2<sup>nd</sup> derivative test:

(4)

1-variable

$$f: \mathbb{R} \rightarrow \mathbb{R}$$



Critical points:  $f'(a) = 0$

Test:

- $f''(a) > 0$  : local min.
- $f''(a) < 0$  : local max.
- $f''(a) = 0$  : inconclusive.

2-variables:  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$

Critical points:  $\nabla f(a) = \left( \frac{\partial f}{\partial x}(a), \frac{\partial f}{\partial y}(a) \right) = 0$

Test: Set  $M = \begin{pmatrix} \frac{\partial^2 f}{\partial x^2}(a) & \frac{\partial^2 f}{\partial x \partial y}(a) \\ \frac{\partial^2 f}{\partial y \partial x}(a) & \frac{\partial^2 f}{\partial y^2}(a) \end{pmatrix} = [D^2 f]_{\text{std}}$

Then  $\det(M) > 0$  and  $\frac{\partial^2 f}{\partial x^2}(a) > 0$  : local min.  
\_\_\_\_\_ and  $\frac{\partial^2 f}{\partial x^2}(a) < 0$  : local max

$\det(M) < 0$  : saddle       $\det(M) = 0$   
inconclusive.

n-variables:  $f: \mathbb{R}^n \rightarrow \mathbb{R}$

(5)

Critical Points:  $\nabla f(a) = 0 \iff D_a f = 0$ . For a crit

pt  $a$ , <sup>define</sup>  $M \in M_{n \times n}(\mathbb{R})$  by  $M_{ij} = \frac{\partial^2 f}{\partial x_i \partial x_j}(a)$

Key: If  $f$  is reasonably smooth at  $a$ , then

$M^t = M$ . By Thm,  $M$  is diagonalizable via an orthonormal  $Q$ :  $Q^t M Q = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}$

Test: If all  $\lambda_i > 0$ , then  $a$  is a local min.

If all  $\lambda_i < 0$ , then  $a$  is a local max.

If some  $\lambda_i > 0$  but other  $\lambda_j < 0$  then  $a$  is a saddle. If some  $\lambda_i = 0$ , who knows?

Exercise: Check that the test in 2<sup>d</sup> follows from this.

⑥

Proof idea: Suppose  $a$  is a critical point of  $f$ .

Then it turns out that right Taylor series is

$$f(a+h) = f(a) + \frac{1}{2}h^t M h + E(h)$$

where  $\lim_{h \rightarrow 0} \frac{E(h)}{\|h\|^2} = 0$ . Changing coordinates

via  $h' = Q^t h \Rightarrow h = Q h'$  and so

$$\begin{aligned} f(a+h) &= f(a) + \frac{1}{2}(Q h')^t M (Q h') + E'(h') \\ &= f(a) + \frac{1}{2}(h')^t \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix} h' + E'(h') \\ &= f(a) + \frac{1}{2}(\lambda_1 x_1^2 + \dots + \lambda_n x_n^2) + \underbrace{E'(h')}_{\text{small}} \end{aligned}$$

if  $h' = (x_1, \dots, x_n)$ .

Thus if all  $\lambda_i > 0$  we have that  $a$  is a local min and similarly for the other cases. ▣

Note: We're actually using here that  $Q$  is orthogonal to understand how the mixed partials transform under this change of coordinates...