

# LECTURE 6

EIVIND ERIKSEN, SEP 24, 2015

GKA 6035

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MATHEMATICS

## Plan:

- ① Definiteness of quadratic forms (review)
- ② Unconstrained optimization
- ③ Convex and concave functions

## Reading:

[MEJ] 14.1-14.4, 14.8, 17.1-17.5

Reminder: Plenary Session 2  
Mon Sep 28<sup>TH</sup>

Selected problems  
Lecture 4-6

① Review: Definiteness

$Q(x_1, \dots, x_n)$  : quadr. form

$A$  : symm. matrix  
( $n \times n$ ) s.t.

$$Q(\underline{x}) = \underline{x}^T A \underline{x}$$

Ex:  $Q = \underline{-x_1^2} + 4x_1x_2 + 3x_1x_3 - \underline{5x_2^2} - \underline{6x_3^2} - \underline{x_4^2} = \underline{x}^T \cdot A \cdot \underline{x}$ ,  $\underline{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix}$

$$A = \begin{pmatrix} -1 & 2 & 3/2 & 0 \\ 2 & -5 & 0 & 0 \\ 3/2 & 0 & -6 & 0 \\ 0 & 0 & 0 & -1 \end{pmatrix}$$

$Q$  is indefinite

Leading principal minors:

$$D_1 = -1 < 0$$

$$D_2 = 1 > 0$$

$$D_3 = -6 \cdot D_2 + 3/2 \cdot 15/2 = \frac{45}{4} - \frac{24}{4}$$

$$D_4 = ?$$

$$= \frac{21}{4} > 0$$

## Criterion using principal minors

1) Compute leading principal minors  $D_1, D_2, \dots, D_n$

2) Check:

$$D_1, D_2, \dots, D_n > 0 \iff A \text{ positive definite}$$

$$D_1 < 0, D_2 > 0, D_3 < 0, \dots \iff A \text{ negative definite}$$

$$\iff (-1)^i D_i > 0 \text{ for all } i$$

3) If  $D_1, \dots, D_n$  fails both of these patterns because of a wrong sign, then  $A$  is indefinite

4) Otherwise, we need to compute all principal minors  $\Delta_1, \Delta_2, \dots, \Delta_n$ .

$$\Delta_1, \Delta_2, \dots, \Delta_n \geq 0 \iff A \text{ positive semidefinite}$$

$$\Delta_1 \leq 0, \Delta_2 \geq 0, \dots \iff A \text{ negative semidefinite}$$

$$\iff (-1)^i \Delta_i \geq 0 \text{ for all } i$$

Ex 1:  $D_1 < 0$ ,  $D_2 > 0$ ,  $D_3 > 0$ ,  $D_4 = ?$

" " " "

-1 " 2/4

If  $D_i < 0$  when  $i$  is even, then  $A$  is indefinite

If two  $D_i$ 's with  $i$  odd have opposite signs, then  $A$  is indefinite

Why this criterion:

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If  $A$  is diagonal, it is  $A = \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{pmatrix}$

Then

$$D_1 = \lambda_1$$

$$D_2 = \lambda_1 \cdot \lambda_2$$

$$D_3 = \lambda_1 \cdot \lambda_2 \cdot \lambda_3$$

⋮

$$D_n = \lambda_1 \cdot \lambda_2 \cdot \lambda_3 \cdot \dots \cdot \lambda_n$$

Ex:  $Q = x^2 + 4xy - 8xz + 4y^2 - 16yz + z^2$

$$A = \begin{pmatrix} 1 & 2 & -4 \\ 2 & 4 & -8 \\ -4 & -8 & 1 \end{pmatrix}$$

$$D_1 = 1$$

$$D_2 = 0$$

$$D_3 = 1 \cdot 0 + 8 \cdot 0 + (-4) \cdot 0 = 0$$

$$\Delta_1^{11} = 1 \quad \Delta_1^{22} = 4 \quad \Delta_1^{33} = 1$$

$$\Delta_2^{12,12} = 0 \quad \Delta_2^{13,13} = 1 - 16 = -15$$

$$\Delta_2^{23,23} = -60$$



$$\Delta_3^{123,123} = 0$$

A is indefinite

Need to compute the remaining principal minors to find out if  $A$  is } pos. semidefn.  
or } indefinite

## ② Unconstrained optimization

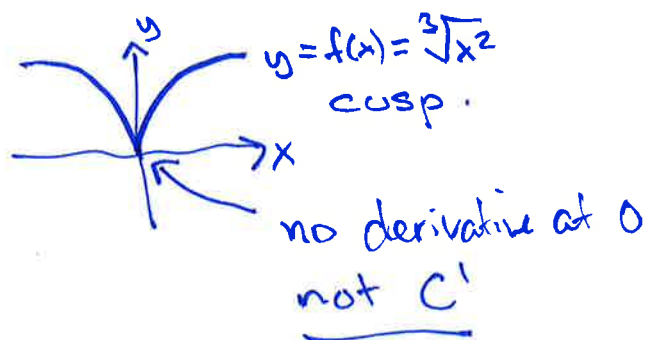
Problem:  $\max / \min f(x_1, x_2, \dots, x_n)$  when  $\underline{x} = (x_1, \dots, x_n)$   
 $f(\underline{x})$  is any point in  $\mathbb{R}^n$  (n-dimensional space)

We always assume  $f$  is  $C^2$ :

$f$  is  $C^1$  if all partial derivatives  $f'_{x_1}, f'_{x_2}, \dots, f'_{x_n}$  exist and are continuous

$f$  is  $C^2$  if all second order partial derivatives  $f''_{x_i x_j}$  exist and are continuous

Ex:  $f(x) = \sqrt[3]{x^2} = x^{2/3}$   
 $f'(x) = (2/3) \cdot x^{2/3-1} = \frac{2}{3} x^{-1/3} = \frac{2}{3 \cdot \sqrt[3]{x}}, x \neq 0$



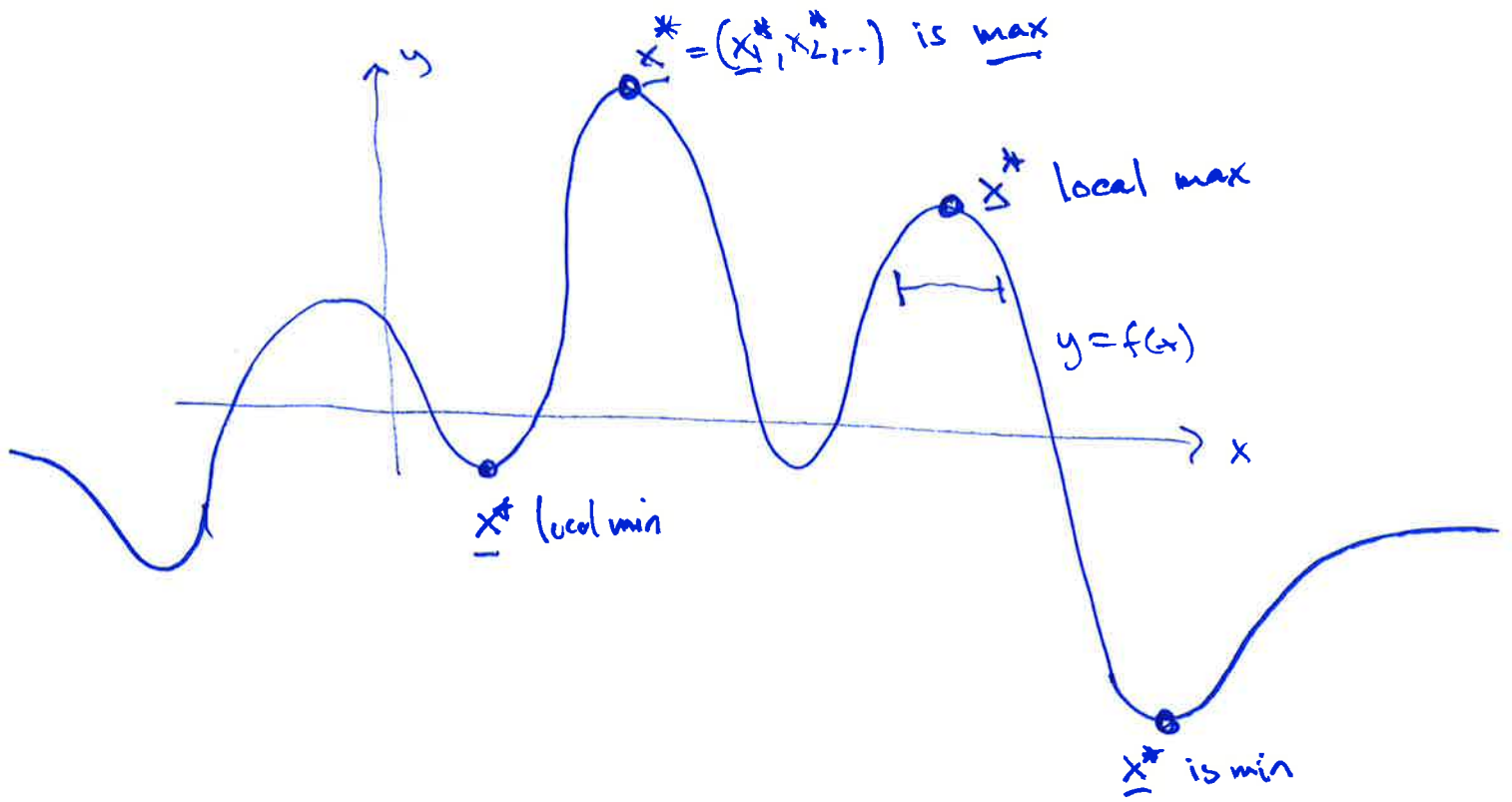
Hard: Find example that is  $C^1$  but not  $C^2$ :

Ex:  $f(x,y) = \begin{cases} \frac{xy(x^2-y^2)}{x^2+y^2}, & (x,y) \neq (0,0) \\ 0, & (x,y) = (0,0) \end{cases}$   $C^1$  but  $H(f)$  does not exist at  $(0,0)$  and cont.

# Local (global) max/min for $f(x)$

$x^*$  max (global max) if  $f(x^*) \geq f(x)$  for all  $x$

$x^*$  min (global min) if  $f(x^*) \leq f(x)$  — | —



$x^*$  local max if  $f(x^*) \geq f(x)$  for all  $x$  close to  $x^*$

$x^*$  local min if  $f(x^*) \leq f(x)$  — | — close to  $x^*$

Derivatives:

Ex:  $f(x,y) = x^3 + xy - y^3$

$$f'_x = 3x^2 + y$$

$$f'_y = x - 3y^2$$

first order  
partial derivatives

$$f''_{xx} = 6x \quad f''_{xy} = 1$$

$$f''_{yx} = 1 \quad f''_{yy} = -6y$$

second order partial  
derivatives

Theorem:

If  $f$  is  $C^2$ , then

$$f''_{x_i x_j} = f''_{x_j x_i}$$

Consequence:

$f$  is  $C^2 \Rightarrow H(f)$  is symmetric

Hessian matrix of  $f$ :

$$H(f) = \begin{pmatrix} f''_{xx} & f''_{xy} \\ f''_{yx} & f''_{yy} \end{pmatrix}$$

In ex:

$$H(f)(x,y) = \begin{pmatrix} 6x & 1 \\ 1 & -6y \end{pmatrix}$$

Ex:  $f(x,y) = \ln(x^2 + y^2 + 1)$

$$f'_x = \frac{1}{x^2 + y^2 + 1} \cdot 2x = \frac{2x}{x^2 + y^2 + 1}$$

$$f'_y = \frac{1}{x^2 + y^2 + 1} \cdot 2y = \frac{2y}{x^2 + y^2 + 1}$$

$$f''_{xy} = \frac{0 \cdot * - 2x \cdot 2y}{(x^2 + y^2 + 1)^2} = \frac{-4xy}{(x^2 + y^2 + 1)^2}$$

$$f''_{xx} = \frac{2 \cdot (x^2 + y^2 + 1) - 2x \cdot 2x}{(x^2 + y^2 + 1)^2}$$

$$= \frac{-2x^2 + 2y^2 + 2}{(x^2 + y^2 + 1)^2}$$

$$f''_{yy} = \frac{2x^2 - 2y^2 + 2}{(x^2 + y^2 + 1)^2}$$

$$H(f) = \begin{pmatrix} \frac{-2x^2 + 2y^2 + 2}{(x^2 + y^2 + 1)^2} & \frac{-4xy}{(x^2 + y^2 + 1)^2} \\ \frac{-4xy}{(x^2 + y^2 + 1)^2} & \frac{2x^2 - 2y^2 + 2}{(x^2 + y^2 + 1)^2} \end{pmatrix}$$

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Defn:

A stationary pt for  $f$  is a point where  $f'_{x_1} = f'_{x_2} = \dots = f'_{x_n} = 0$  (FOC)

Thm: (First order condition)

$\underline{x}^*$  local max/min  $\implies \underline{x}^*$  stationary pt.

Ex:  $f = x^3 + xy - y^3$

Stationary pts:

$$f'_x = 3x^2 + y = 0$$

$$y = -3x^2: x - 3 \cdot (-3x^2)^2 = 0$$

$$f'_y = x - 3y^2 = 0$$

$$x - 3 \cdot 9x^4 = 0$$

$$x \cdot (1 - 27x^3) = 0$$

Stationary pts:

$$(x, y) = (0, 0), \left(\frac{1}{3}, -\frac{1}{3}\right)$$

$$f(0, 0) = 0$$

$$f\left(\frac{1}{3}, -\frac{1}{3}\right) = \frac{1 - 3 + 1}{27} = -\frac{1}{27}$$

$$\underline{x=0} \text{ or } 27x^3 = 1$$

$\Downarrow$

$$x^3 = \frac{1}{27}$$

$$y = 0$$

$$\underline{x = \frac{1}{3}}$$

$\Downarrow$

$$y = -3 \cdot \left(\frac{1}{3}\right)^2 = -\frac{1}{3}$$



Finding stationary pts for  $f$  gives us candidates for local max/min, and therefore candidates for max/min

Second order conditions (SOC)

Assume that  $\underline{x}^*$  is a stationary point for  $f(\underline{x})$ .

If $H(f)(\underline{x}^*)$ is <u>positive defn.</u>	then $\underline{x}^*$ is <u>local min</u>
—    —	then $\underline{x}^*$ is <u>local max</u>
—    —	then $\underline{x}^*$ is a <u>saddle point</u>
	↑ <u>point</u>

Test not conclusive if  $H(f)(\underline{x}^*)$  is pos./neg. semidefn. (but not defn.)

(i.e. a stationary pt that is neither local max nor local min)

Ex:  $f = x^3 + xy - y^3$        $H(f) = \begin{pmatrix} 6x & 1 \\ 1 & -6y \end{pmatrix}$

Stat. pts:  $(0,0)$   
 $(1/3, -1/3)$

Classification:

$H(f)(0,0) = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$        $D_1 = 0$  → indefinite  
 $D_2 = -1$        $(0,0)$  saddle pt

$H(f)(1/3, -1/3) = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$        $D_1 = 2$  positive defn.  
 $D_2 = 3$        $(1/3, -1/3)$  local min

$f(-2,0) = -8$   
no min

$f$  has no max       $f$  has local min at  $(1/3, -1/3)$  ( $f = -1/27$ )  
→ (candidate for min)



Ex:

~~$f = x^2$~~

$f = x^3$

$f = x^4$

$f = -x^4$

$f' = 3x^2$

$f' = 4x^3$

$f'_x = -4x^3$

$x = 0$

$x = 0$

St.pts:

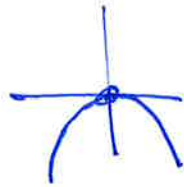
$x = 0$

$H(f)(0) = -12x^2$   
 $= 0$

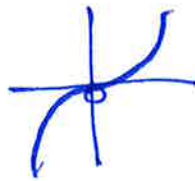
$H(f)(0) = 6x$   
 $= 0$

$H(f)(0) = 12x^2$   
 $= 0$

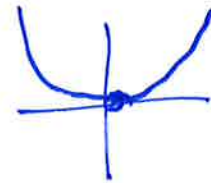
SOC:



max



saddle pt.

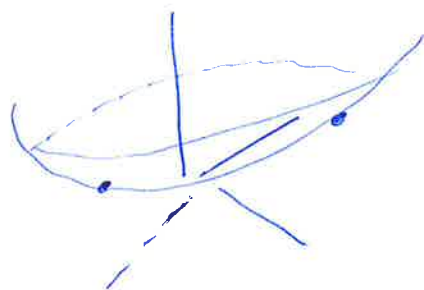


min

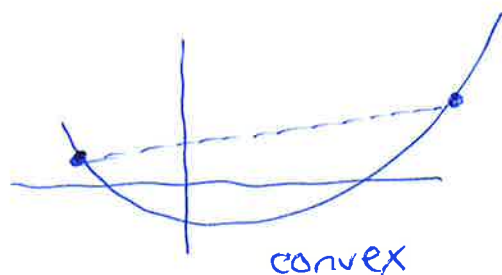
### ③ Convex and concave functions

$f(x)$  : function

Defn:

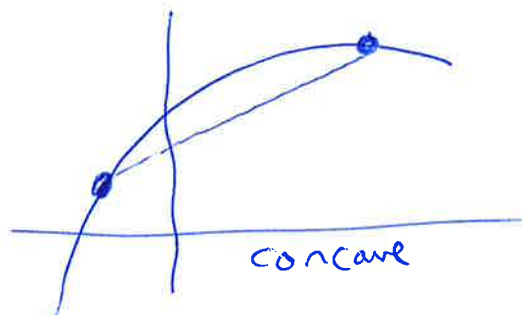


$f$  is convex if for any two points on the graph of  $f$ , the straight line between these two pts lies over (or on) the graph.



convex

$f$  is concave if for any two points on the graph of  $f$ , the straight line between these two pts lies under (or on) the graph.



concave



Neither convex  
nor concave

Note: convex / concave  
are global notions

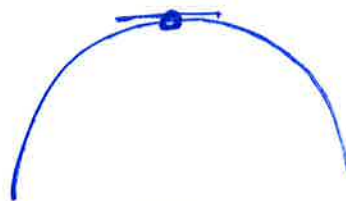
Thm:

If  $f$  is convex, then any stationary pt of  $f$  is global min.

If  $f$  is concave, then any stationary pt of  $f$  is global max.

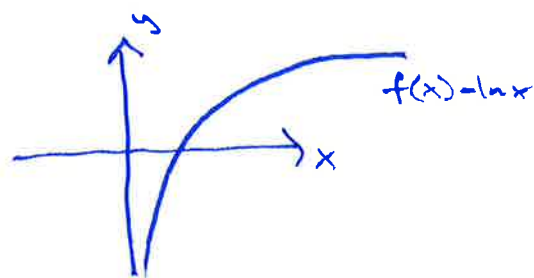


Convex



Concave

Ex:  $f(x) = \ln x, x > 0$   
 $f'(x) = 1/x$   
 $f''(x) = -1/x^2 < 0$   
 $f$  concave



no stationary pts  
no global max

Criterion for convexity / concavity:

If  $H(f)(x)$  is positive semidefinite for all  $x$ ,  
 then  $f$  is convex

If  $H(f)(x)$  is negative semidefinite for all  $x$ ,  
 then  $f$  is concave

$f$  convex  $\iff H(f)(\pm)$  pos. semidefn.  
for all  $\pm$

$f$  concave  $\iff H(f)(\pm)$  neg. semidefn.  
for all  $\pm$ .

Ex:  $f = x^3 + xy - y^3$   $H(f) = \begin{pmatrix} 6x & 1 \\ 1 & -6y \end{pmatrix}$

$$D_1 = 6x \quad \Delta_1 = -6y$$

$$D_2 = -36xy - 1$$

convex:  $D_1 \geq 0 \quad \Delta_1 \geq 0$   
 $D_2 \geq 0$

$$\begin{aligned} 6x \geq 0 \quad -6y \geq 0 \\ -36xy - 1 \geq 0 \\ \text{for all } x, y \end{aligned}$$

$D_1 = 6x$  can be pos. or neg.  $\implies$  no not convex

concave:  $D_1 \leq 0 \quad 6x \leq 0$  for all  $x, y$   
no not concave

Ex:  $f = e^{2x-y}$   $f'_x = e^{2x-y} \cdot 2$   $f'_y = e^{2x-y} \cdot (-1)$

$$D_1 = 4e^{2x-y} > 0$$

$$D_2 = 4 \cdot (e^{-y})^2 - 4(e^{-y})^2 = 0$$

$$\Delta_1 = e^{2x-y} > 0$$

$$H(f) = \begin{pmatrix} 4e^{2x-y} & -2e^{2x-y} \\ -2e^{2x-y} & e^{2x-y} \end{pmatrix}$$

pos. semidefn.

$\iff$

$f$  convex

## An Example

Let  $f(x,y) = x^2y^3 + y^2 - 2y$ . Then  $f$  has just one stationary pt, which is a local min. However,  $f$  has no global min.

$$f'_x = 2xy^3 = 0$$

$$f'_y = 3x^2y^2 + 2y - 2 = 0$$

FOC: ~~\*~~  $x=0$  or  $y=0$

If  $x=0$ , then (2) gives

$$2y - 2 = 0 \Rightarrow \underline{y=1}$$

If  $y=0$ , then (2) gives

$$-2 = 0 \quad \text{no solution}$$

$$H(f) = \begin{pmatrix} 2y^3 & 6xy^2 \\ 6xy^2 & 6x^2y + 2 \end{pmatrix}$$

$$H(f)(0,1) = \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix} \quad D_1 = 2 > 0$$

$$D_2 = 4 > 0$$

pos. defn.  $\Rightarrow (0,1)$  local min

Conclusion:  $(0,1)$  is the only stationary pt, with  $\underline{f(0,1) = -1}$

But  $(x,y) = (0,1)$  with  $f(0,1) = -1$  is not global min  
For example,

$$f(3,-1) = -9 + 1 + 2 = -6 < -1$$

Conclusion:  $f$  has no global min even if  $(0,1)$  is local min and this is the only stationary point.