

# Discussion #1

Math 164 - Optimization

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OH: Monday -  
Thursday -

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This week, Friday 3-4 PM

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Simple Function:

$$f(x) = x^2 - 4x + 3$$

Minimize  $f$ ! (31A)

1. Set  $f' = 0$

$$f'(x) = 2x - 4 = 0$$

$$\Rightarrow x = 2$$

2. Check  $f''(x) > 0$

$$f''(x) = 2 > 0$$

# 1. First Order Necessary Condition (FONC)

If  $f'(x) \neq 0$ ,  $x$  is not a minimizer of  $f$ .

# 2. Second Order Sufficient Condition (SOSC)

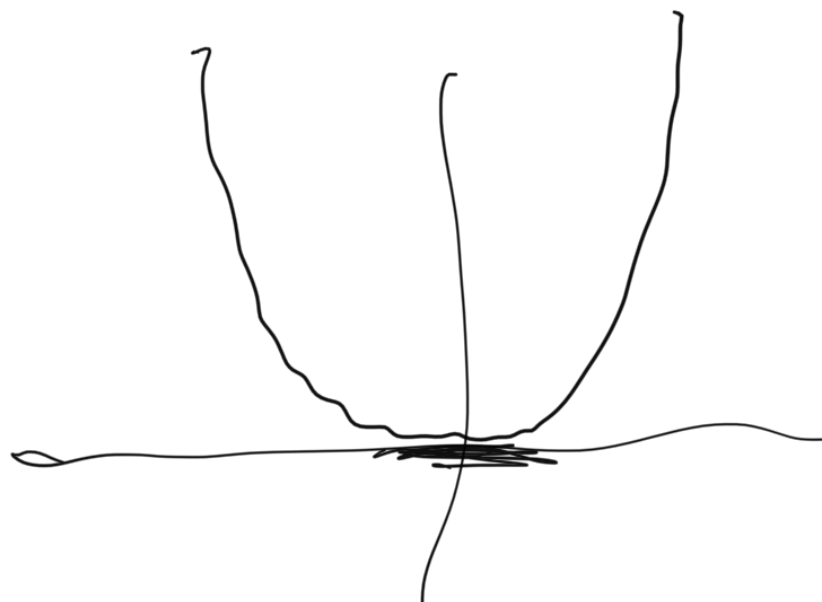
If  $f'(x) = 0$ , and  $f''(x) > 0$ , then  $x$  is a strict local minimizer of  $f$ .

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It is not necessary for SOSC to hold for  $x$  to be a minimizer,

Can have  $f''(x) = 0$ , and  $x$  is a minimizer.

$$f(x) = x^4$$



Generalize to  $f: \mathbb{R}^n \rightarrow \mathbb{R}$

1. FONC:  $\nabla f(x) = 0$

2. SOSC:  $\nabla^2 f(x) > 0$

Hessian

$f(x, y)$ ,

$$\nabla f(x, y) = \begin{pmatrix} f_x \\ f_y \end{pmatrix}$$

$$\nabla^2 f(x, y) = \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix}$$

$\nabla^2 f(x) > 0$  means

this matrix is positive definite.

We say a symmetric matrix is positive definite if

A is positive definite if any of the following conditions hold:

$$1. \forall x \neq 0, \underbrace{x^T A x}_{\approx ax^2} > 0$$

2. A has only positive eigenvalues

3. The principal minors of A are all positive.

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 4 & 5 \end{pmatrix}$$

1st minor:  $\det [1] = 1$

2nd minor:  $\det \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix} = 3 - 4 = -1$

3rd minor:  $\det \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 4 & 5 \end{bmatrix}$

$$f(x, y) = x^2 - 4xy + y^2$$

$$\nabla f(x, y) = (2x - 4y, -4x + 2y)$$

$$1. \text{FOC: } \nabla f(x, y) = \begin{pmatrix} 2x - 4y \\ 2y - 4x \end{pmatrix}$$

$$2. \text{SOC: } \nabla^2 f(x, y) = \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix}$$

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

$$1. \nabla f = 0, \begin{cases} 2x - 4y = 0 \\ 2y - 4x = 0 \end{cases}$$

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$2. \det [2] = 2 > 0$$

$$\det \begin{bmatrix} 2 & -4 \\ -4 & 2 \end{bmatrix} = 4 - (16) = -12 < 0$$

$\Rightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix}$  does not satisfy  
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not necessarily a  
minimizer

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Special Class of Functions:

Convex Functions:

Def:  $f \in C^2$

↗  
2x cont. diff.,

We say  $f$  is convex if

$$\nabla^2 f(x) \geq 0 \quad (\text{semi positive definite})$$

1.  $x^T \nabla^2 f x \geq 0$
2. nonnegative eigenvalues
3. principal minors nonnegative

$f$  is strictly convex if

$$\nabla^2 f > 0$$

For strictly convex functions,

Only need to check F-DNC,  
because SOSOC are automatically  
satisfied.

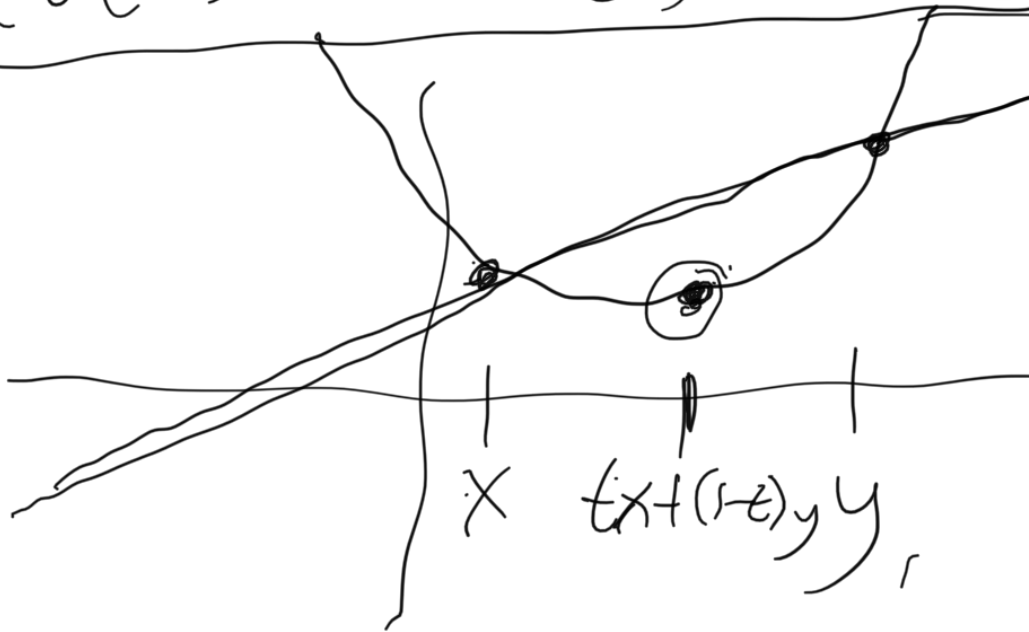
For arbitrary  $F: \Omega \rightarrow \mathbb{R}$ ,

$F$  is convex if

for all  $x, y \in \Omega$ ,  $t \in (0, 1)$ ,

$$F(t(x) + (1-t)y) \leq t f(x) + (1-t)f(y)$$

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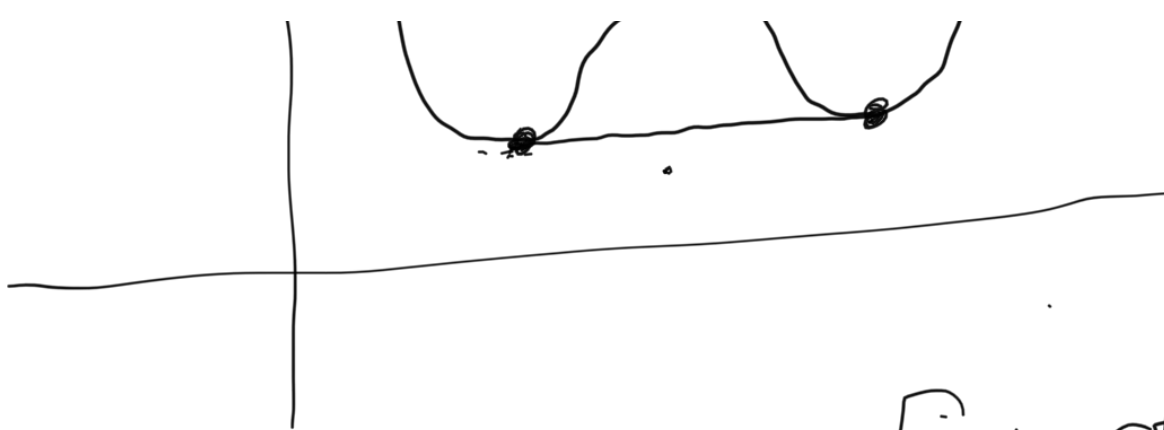


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$f$  is strictly convex if  
the  $\leq$  is a  $<$

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Thm: A strictly convex  
function has at most 1 minimizer



Strictly convex Functions  
have unique minimizer

$\Rightarrow$  If we minimize a strictly  
convex function, we are guaranteed  
to find the unique global minimizer

If  $x$  is a local minimizer  
of a convex function,  
it is a global minimizer.

Prove  $f(x, y) = e^{x+y}$  is convex

Show  $\nabla^2 f \geq 0$

$$\nabla f = \begin{pmatrix} e^{x+y} \\ e^{x+y} \end{pmatrix}$$

$$\nabla^2 f(x, y) = \begin{pmatrix} e^{x+y} & e^{x+y} \\ e^{x+y} & e^{x+y} \end{pmatrix}$$



$$\det [e^{x+y}] = e^{x+y} > 0$$

$$\det \begin{bmatrix} e^{x+y} & e^{x+y} \\ e^{x+y} & e^{x+y} \end{bmatrix} = 0 \geq 0$$

$\Rightarrow$   $e^{x+y}$  is convex.

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