

# Discussion # 2

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Agenda: More on FONC, SONC  
SOSC  
Optimality Conditions

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HW due today, on Gradescope  
at 11:59

OH: M 1-2, Th 4-5  
LA time

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$$f(y) = \frac{1}{2} y^T A y, A \succ 0$$

Want to show that

$f$  is strictly convex

$f$  is strictly convex if

$$f(tx + (1-t)y) < tf(x) + (1-t)f(y)$$

$$\forall x, y \in \mathcal{J}L, t \in (0, 1)$$

If  $f \in C^2$ ,  $f$  is strictly convex  
if  $\nabla^2 f > 0$ .

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$$g(x, y) = \frac{1}{2} x^T y$$

$$y = h(x) = Ax$$

$$f(x) = g(x, h(x)) = \frac{1}{2} x^T Ax$$

$$\begin{array}{ccc} & g & \\ & \swarrow \frac{\partial g}{\partial y} & \\ \frac{\partial g}{\partial x} \downarrow & | & \swarrow h \\ & x & \swarrow \frac{\partial h}{\partial x} \end{array}$$

$$\begin{aligned} \nabla f(x) &= \frac{\partial g}{\partial x} + \left( \frac{\partial h}{\partial x} \right)^T \frac{\partial g}{\partial y} \\ \parallel \\ \nabla \left( \frac{1}{2} x^T Ax \right) &= \frac{1}{2} Ax + A^T \left( \frac{1}{2} x \right) \end{aligned}$$

$$= \frac{1}{2} (A + A^T) x$$

$$= Ax$$

$\nabla^2 f = A$ , Since  $A > 0$   
we have  $f$  strictly convex.

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## Optimality Conditions

$$f: \Omega \rightarrow \mathbb{R}, \quad \Omega \subset \mathbb{R}^n$$

We want to minimize  $f$ .

FONC: If  $x^*$  is a minimizer

$$\text{then } \nabla f(x^*) = 0$$

SONC: If  $x^*$  is a minimizer,

$$\text{then } \nabla^2 f(x^*) \succeq 0$$

SOSC: If  $\nabla f(x^*) = 0$ ,

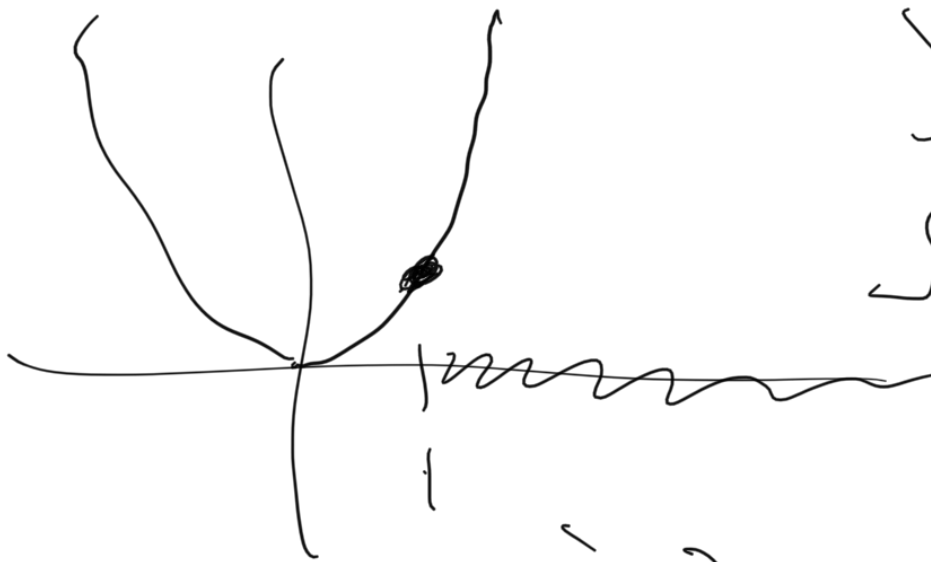
and  $\nabla^2 f(x^*) \succ 0$ , then

$x^*$

$x^*$  is a minimizer.

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$x^* \in$  interior of  $\Omega$



$$y = x^2$$

$$\Omega = \{x \geq 1\}$$

$$f'(1) = 2$$

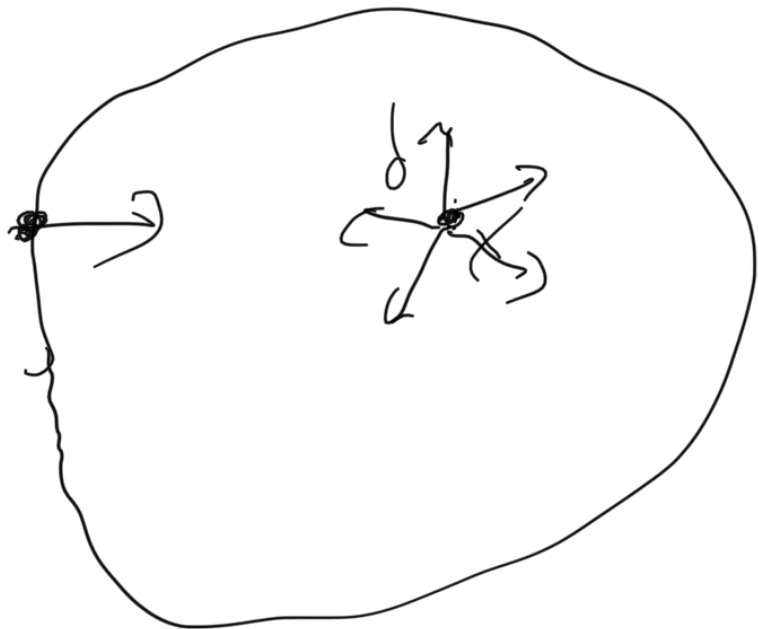
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Given  $\Omega \subset \mathbb{R}^n$ ,  $x \in \Omega$ ,

We say  $d \in \mathbb{R}^n$  is feasible if

$\exists \alpha_0 > 0$ , such that

$x + \alpha d \in \Omega$ ,  $\forall 0 < \alpha < \alpha_0$

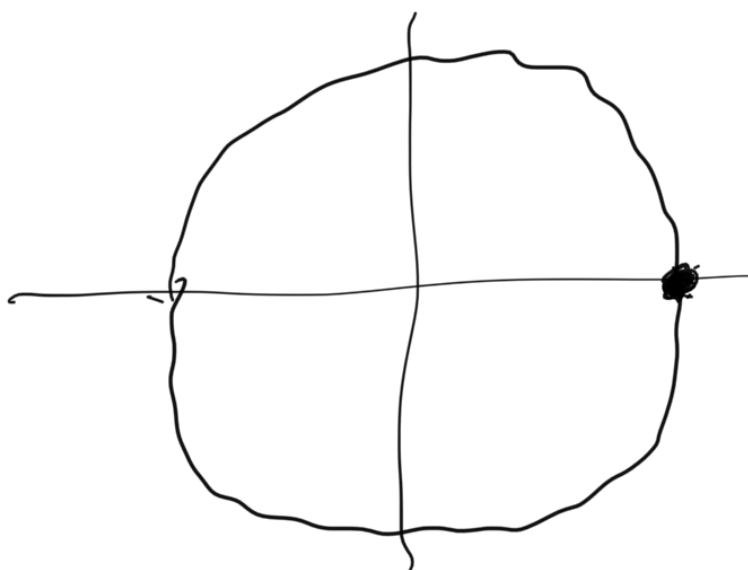


$$\Omega = \{(x, y) : x^2 + y^2 = 1\}$$

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$$f(x, y) = x$$

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$$x^* (1, 0)$$

There are no feasible directions  
at  $x^*$ .

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FONC (boundary):

If  $x^*$  is a minimizer, for all  
feasible directions  $d$  at  $x^*$ ,

$$d^T \nabla f(x^*) \geq 0$$

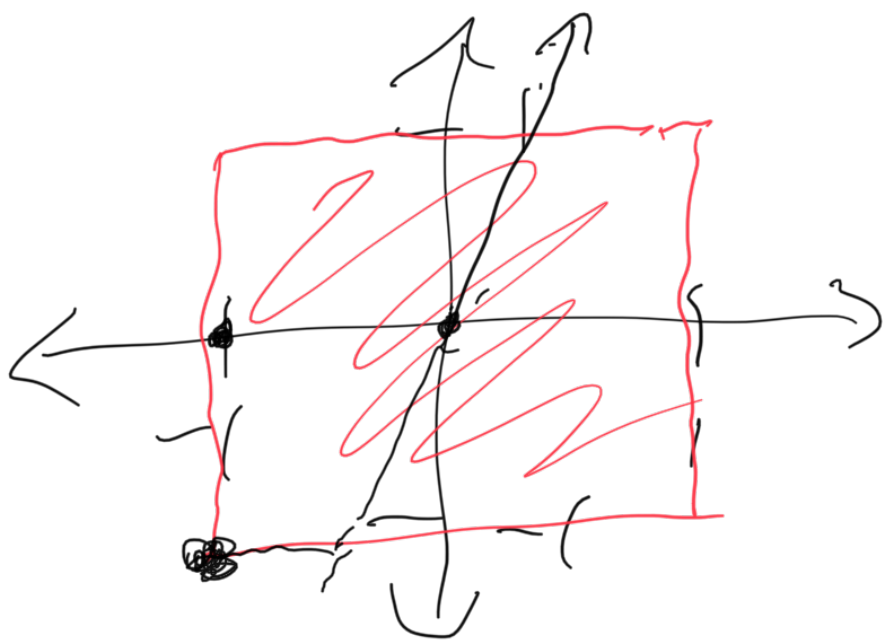
directional  
derivative in direction  $d$ .

SONC. If  $x^*$  is a minimizer  
for all feasible directions  $d$  at  $x^*$

$$d^T \nabla^2 f(x^*) d \geq 0$$

$$f(x, y) = \underline{x + 2y}$$

$$\text{s.t. } (x, y) \in \Omega, \quad \Omega = \begin{cases} -1 \leq x \leq 1 \\ -1 \leq y \leq 1 \end{cases}$$



1. Is any point in the interior  
a minimizer?

$$\nabla f = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

No, since  $\nabla f \neq 0$  anywhere

$\Rightarrow$  ...

2) Check the boundary

Check  $x^* = (-1, -1)$

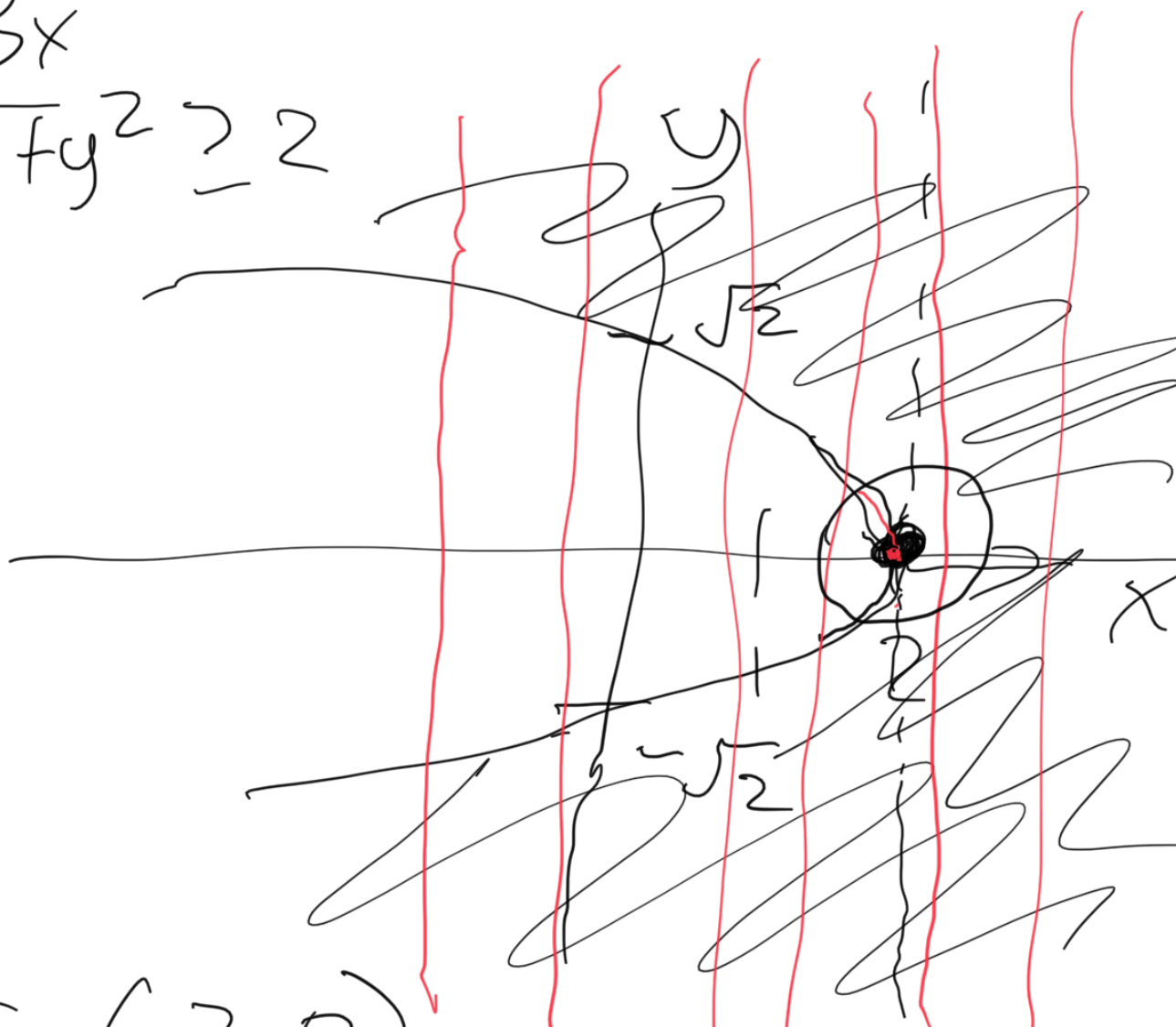
Observe  $d = \begin{pmatrix} d_1 \\ d_2 \end{pmatrix}$  is feasible

if  $\underline{d_1, d_2} \geq 0$

$$d^T \nabla f(x^*) = d_1 + 2d_2 \geq 0$$

Therefore  $x^*$  satisfies FONC

$$\begin{array}{l} \min 3x \\ \text{s.t. } x + y^2 \geq 2 \end{array}$$



... for (x, y)

Consider  $(2, 0)$

at this point,  $d$  is feasible  
if it points to the right,

$$d = \begin{pmatrix} d_1 \\ d_2 \end{pmatrix}, d_1 \geq 0$$

$$\text{FONC: } \nabla f = \begin{pmatrix} 3 \\ 0 \end{pmatrix}$$

$$d^T \nabla f(2, 0) = 3d_1$$

$x^*$  satisfies FONC,  $\geq 0$

$$\text{SONC: } \nabla^2 f = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$d^T \nabla^2 f d = 0 \geq 0$$

so  $x^*$  satisfies SONC

$x^*$  is not a local minimizer.





