

EBA 1180
Spring 24
Sect. 41

Warm-up: $f(x, y) = \sqrt{x^2 + y^2} = (x^2 + y^2)^{\frac{1}{2}}$

$f'_x(x, y) = (\sqrt{u})'_x = (u^{\frac{1}{2}})'_x$

$f'_y(x, y) = \frac{y}{\sqrt{x^2 + y^2}}$

By symmetry or repeating same argument

$$\begin{aligned} &= \frac{1}{2} u^{-\frac{1}{2}} \cdot u'_x \\ &= \frac{1}{2\sqrt{u}} \cdot 2x = \frac{2x}{2\sqrt{x^2 + y^2}} \\ &= \frac{x}{\sqrt{x^2 + y^2}} \end{aligned}$$

Let $u = x^2 + y^2$

Interpretation of partial derivatives

What does $f'_x(a, b)$ and $f'_y(a, b)$ mean?

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

$$\Rightarrow f'_x(x, y) = 3x^2 - 3y, \quad f'_y(x, y) = -3x + 3y^2$$

Let $(x, y) = (2, 1)$. Then:

$$f(2, 1) = 2^3 - 3 \cdot 2 \cdot 1 + 1^3 = 3$$

$$f'_x(2, 1) = 3 \cdot 2^2 - 3 \cdot 1 = \underline{9} \quad \text{← SAME!}$$

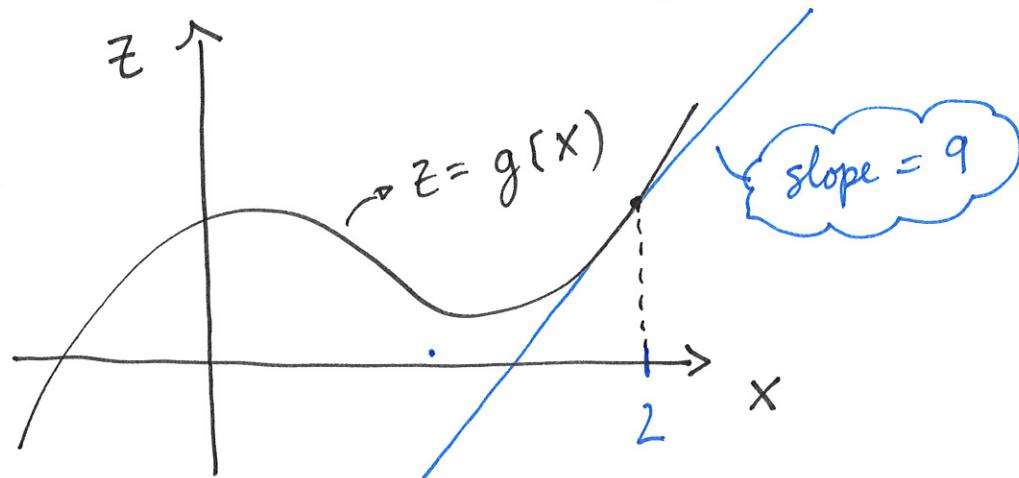
$$f'_y(2, 1) = -3 \cdot 2 + 3 \cdot 1^2 = \underline{-3}$$

In the x-direction ($y=1$):

$$f(x, 1) = x^3 - 3x \cdot 1 + 1^3 = x^3 - 3x + 1 =: g(x)$$

Defined as

$$\underbrace{g'(x)}_{= f'_x(x, 1)} = 3x^2 - 3, \quad \underbrace{g'(2)}_{= f'_x(2, 1)} = 3 \cdot 2^2 - 3 = \underline{\underline{9}}$$



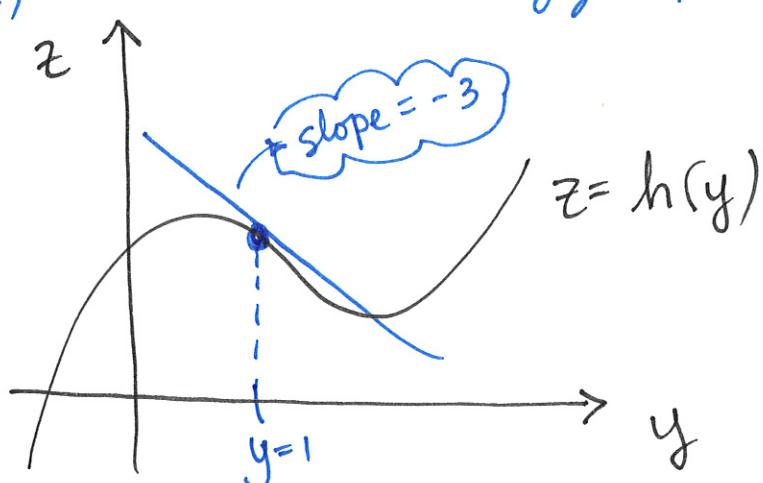
In the y-direction ($x=2$):

$$f(2, y) = 2^3 - 3 \cdot 2 \cdot y + y^3 = 8 - 6y + y^3 =: h(y)$$

$$\Rightarrow \underbrace{h'(y)}_{= f'_y(2, y)} = -6 + 3y^2, \text{ so } \underbrace{h'(1)}_{= f'_y(2, 1)} = -6 + 3 \cdot 1^2 = \underline{\underline{-3}}$$

$$= f'_y(2, y)$$

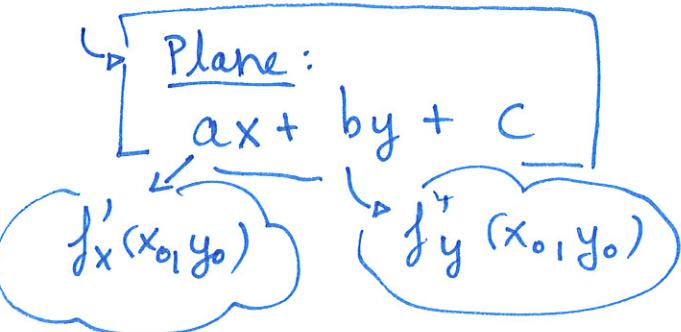
NOTE:
Same as
on pg. 1



Linear approximation of $f(x,y)$ at (x_0, y_0) :

Tangent plane of f at (x_0, y_0)

$$L(x,y) = f(x_0, y_0) + f'_x(x_0, y_0)(x - x_0) + f'_y(x_0, y_0)(y - y_0)$$



The gradient

Def (gradient): The gradient of $f(x,y)$ is

$$\nabla f = \begin{bmatrix} f'_x \\ f'_y \end{bmatrix}$$

"The gradient of f " "A vector"

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

Q: $\nabla f = \begin{bmatrix} 2x - 2 \\ 2y + 4 \end{bmatrix}$

The gradient of f in $(x,y) = (-2, 2)$:

Q: $\nabla f(-2, 2) = \begin{bmatrix} 2 \cdot (-2) - 2 \\ 2 \cdot 2 + 4 \end{bmatrix} = \begin{bmatrix} -6 \\ 8 \end{bmatrix}$

NOTE: The gradient at a point is a normal vector to the tangent line of the level curve at that point.

$$\text{Ex: } \nabla f(-2, 2) = \begin{bmatrix} -6 \\ 8 \end{bmatrix}$$

Tangent line of level curve at $(-2, 2)$:

$$y = \frac{3}{4}x + \frac{7}{2}$$

From Tuesday lecture!

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x \\ \frac{3}{4}x + \frac{7}{2} \end{bmatrix}$$

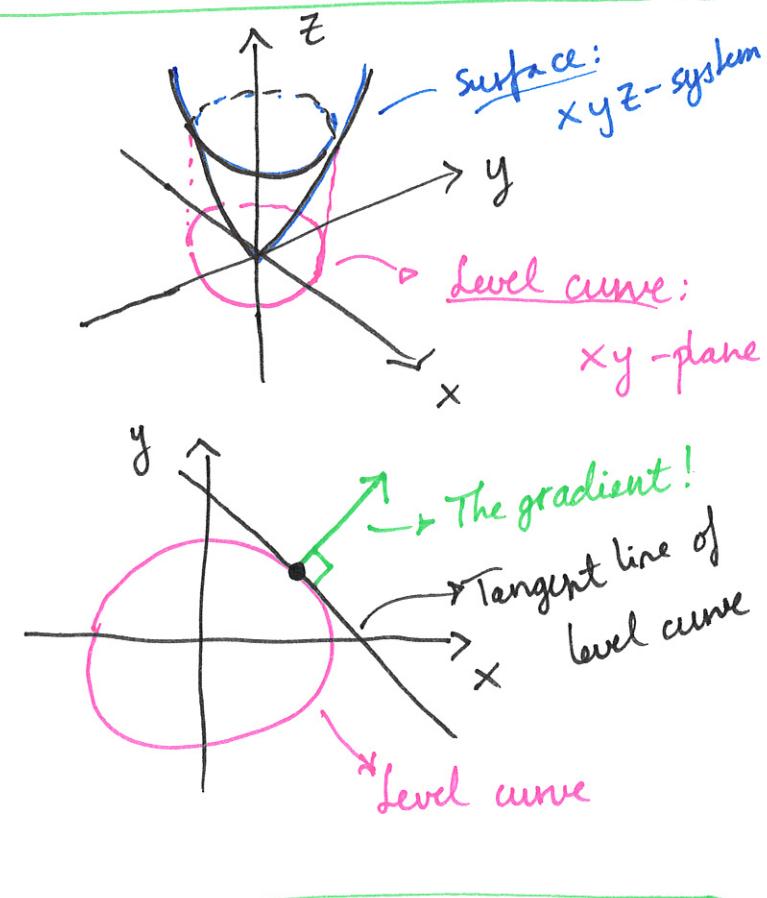
$$= \begin{bmatrix} 1 \\ \frac{3}{4} \end{bmatrix}x + \begin{bmatrix} 0 \\ \frac{7}{2} \end{bmatrix}$$

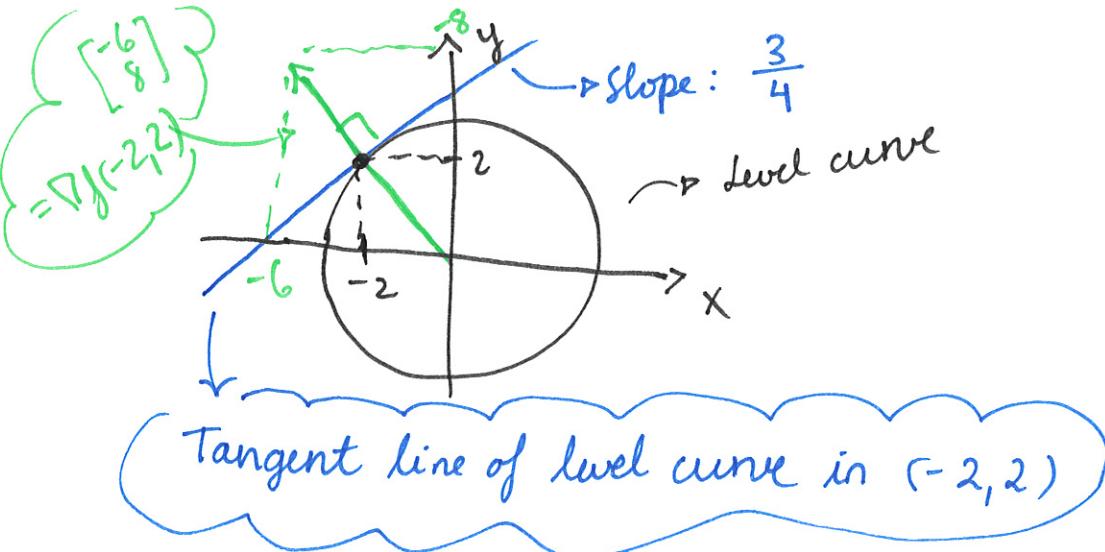
direction of the tangent line of the level curve

$$\begin{aligned} \nabla f(-2, 2) \cdot \begin{bmatrix} 1 \\ \frac{3}{4} \end{bmatrix} &= \begin{bmatrix} -6 \\ 8 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ \frac{3}{4} \end{bmatrix} = -6 + \cancel{8} \cdot \frac{3}{4} \\ &= -6 + 6 = 0 \end{aligned}$$

so:

$\nabla f(-2, 2) \perp \begin{bmatrix} 1 \\ \frac{3}{4} \end{bmatrix}$ (90° angle), i.e. the tangent line of the level curve at $(-2, 2)$.





Directional derivative

Def (Directional derivative): Let $f(x, y)$ be a function,

$\vec{a} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}$ a 2-vector. Then,

$$f'_{\vec{a}} := \vec{a} \cdot \nabla f$$

dot product

"The directional derivative of f wrt. \vec{a} "

Ex: $f(x, y)$ as before, $\vec{a} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$

$$\begin{aligned}
 f'_{\vec{a}} &= \vec{a} \cdot \nabla f = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 2x - 2 \\ 2y + 4 \end{bmatrix} \\
 &= 2(2x - 2) + 1(2y + 4) \\
 &= \dots = \underline{4x + 2y} \rightarrow \text{A number}
 \end{aligned}$$

Can insert a point, say $(1, 1)$:

$$f'_{\vec{a}}(1, 1) = 4 \cdot 1 + 2 \cdot 1 = \underline{\underline{6}}$$

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More about the second derivative test and optimization

NOTE: If (x^*, y^*) is a stationary point with

$$H(f)(x^*, y^*) = \begin{bmatrix} A & B \\ B & C \end{bmatrix}$$

f''_{xx}
 A
 B
 f''_{xy}
f''_{yx}
 B
 C
 f''_{yy}

Then,

- ① If $\underbrace{AC - B^2}_{\text{det. of Hessian}} > 0$, then $\cancel{AC} > B^2 \geq 0$
- ↓
 since square
- Local max. or local min.

$\Rightarrow AC > 0$. But then,

i) $A, C > 0$, so $\underbrace{A+C > 0}_{\text{trace}}$

local min. from
2nd derivative
test

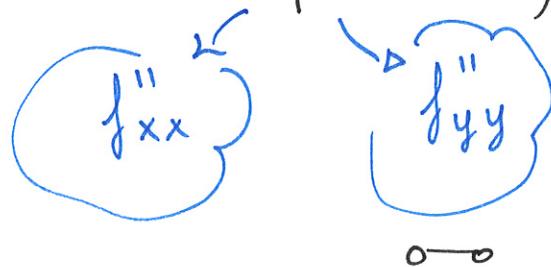
ii) $A, C < 0$, so $A+C < 0$

local max. from
2nd. derivative
test

Hence, the possible cases are:

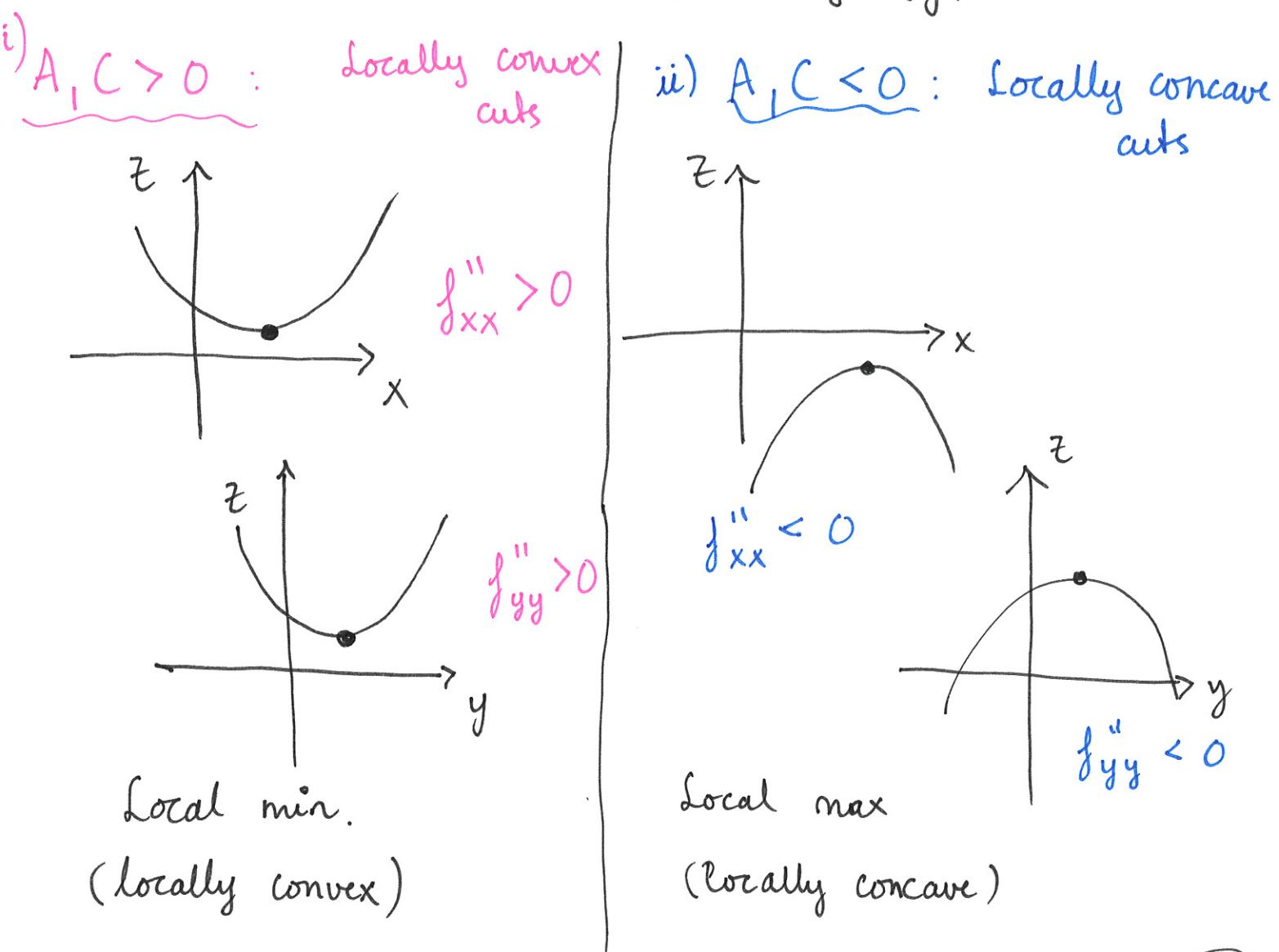
i) $A, C > 0$; local min.

ii) $A, C < 0$; local max.



Recall: Second order derivative ≥ 0 ; Convex
Second order derivative ≤ 0 ; Concave

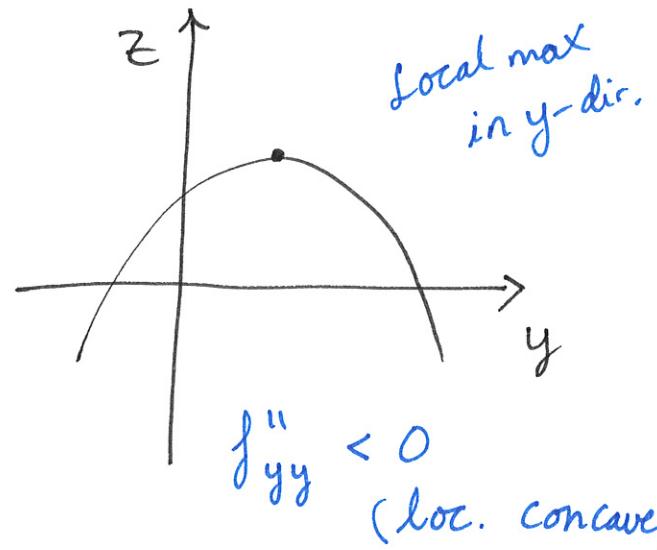
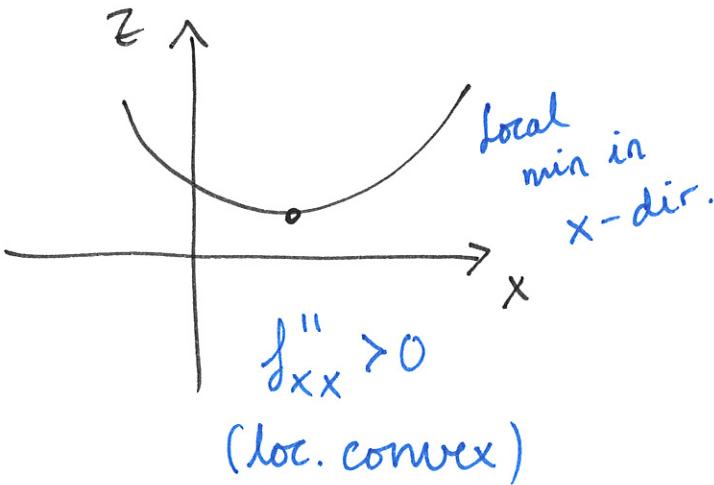
Graphically: Cuts of the graph $z = f(x, y)$:



② If $\underbrace{AC - B^2}_{\det H} < 0$, a typical case is

$$\begin{array}{l} A > 0, \quad C < 0 \\ f''_{xx} > 0 \quad f''_{yy} < 0 \end{array}$$

Saddle point



Ex: $f(x,y) = \sqrt{x^2 + y^2}$, $D_f = \mathbb{R}^2$



Stationary pts?

$$\begin{aligned} \frac{x}{\sqrt{x^2+y^2}} &= f'_x = 0 \Rightarrow x = 0 \\ \frac{y}{\sqrt{x^2+y^2}} &= f'_y = 0 \Rightarrow y = 0 \end{aligned}$$

No boundary points

Is $(0,0)$ a stationary point? NO! Because $f'_x(0,0)$ and $f'_y(0,0)$ are not defined \rightarrow divide by 0

\Rightarrow No stationary points for f . But $(0,0)$ is a critical point, and hence a candidate point.

$$f(0,0) = \sqrt{0^2 + 0^2} = 0,$$

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which is the smallest value f can attain.

Hence, $(0, 0)$ is a global minimum for f .

↳ minimum point