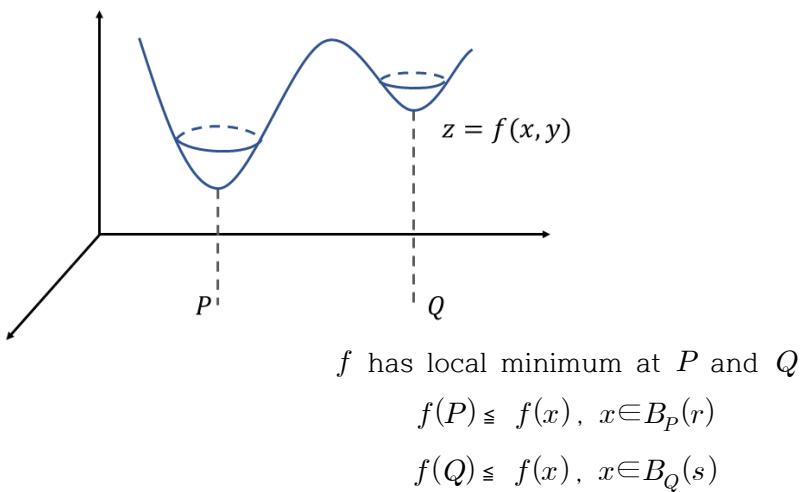
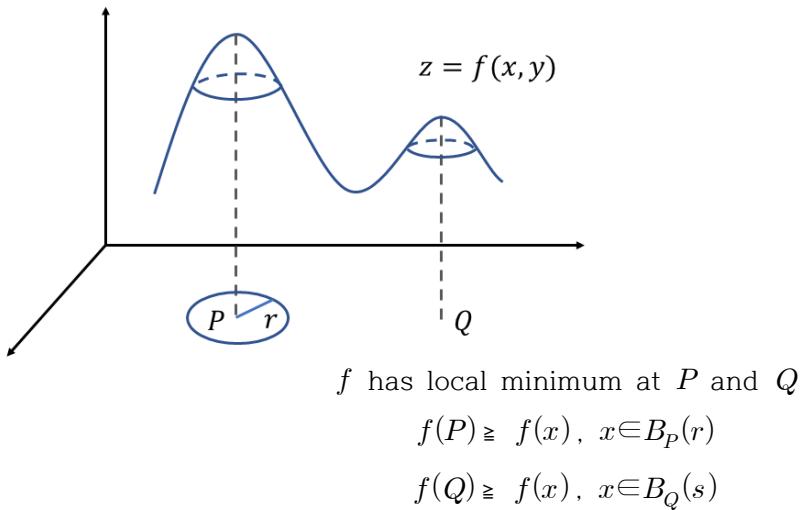


Module Optimization (several variable)



정의) P : critical point of $f \Leftrightarrow$ 다음 중 하나를 만족할 경우

- ① $f_x = f_y = 0$ at P
- ② f_x or f_y fail to exist at P

임계점 정리

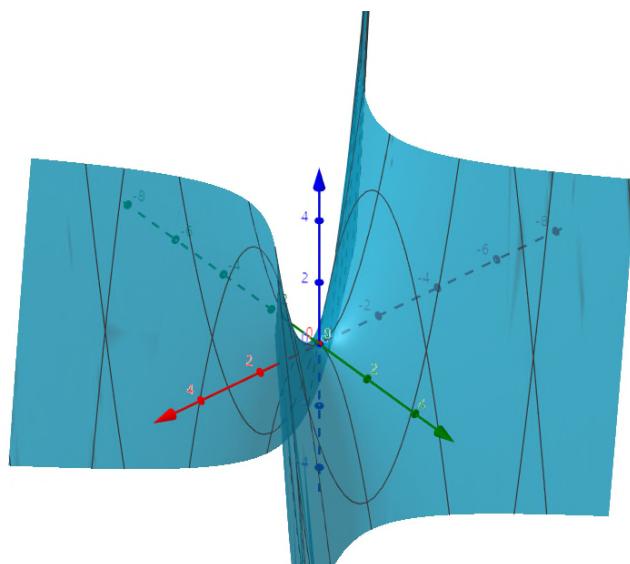
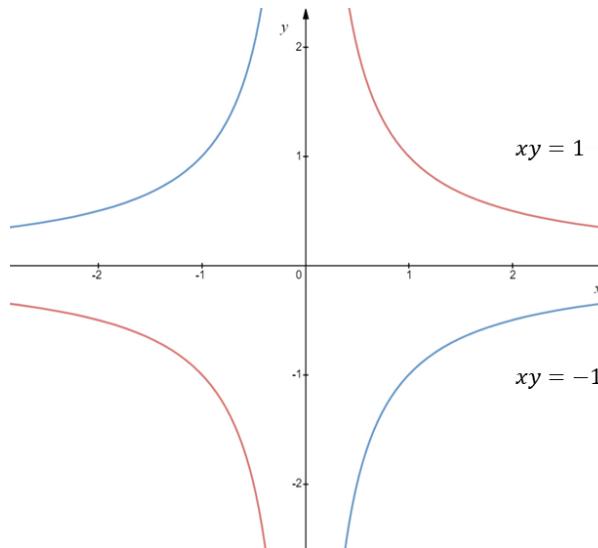
$$U \subset \mathbb{R}^2$$

만약 미분가능한 함수 $f : u \rightarrow R$ 가 점 P 에서 국소적인 최대나 최소가 되면

$$f_x(P) = f_y(P) = 0$$

Remark) 위의 정리의 역은 성립하지 않는다

생각해볼 함수 $z = xy$



saddle point

$$z|_{y=x} = x^2$$

$$z|_{y=-x} = -x^2$$

$$\frac{\partial z}{\partial x} \Big|_{(x,y)=(0,0)} = y \Big|_{(x,y)=(0,0)} = 0$$

$$\frac{\partial z}{\partial y} \Big|_{(x,y)=(0,0)} = x \Big|_{(x,y)=(0,0)} = 0$$

Def 함수 f의 Hessian determinant 또는 discriminant (해세 행렬의 행렬식)

$$H_f = \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix} = f_{xx}f_{yy} - f_{xy}^2$$

■ 극대, 극소의 판별법 P : critical point of f

$$(1) H_f(P) > 0, f_{xx}(P) > 0 \Rightarrow P : \text{local min}$$

$$(2) H_f(P) > 0, f_{yy}(P) < 0 \Rightarrow P : \text{local max}$$

$$(3) H_f(P) < 0, \Rightarrow P : \text{saddle point}$$

$$(4) H_f(P) = 0, \text{판별할 수 없다.}$$

예제])

$$f(x,y) = x^2 + y^2$$

$$f_x = 2x = 0 \quad f_y = 2y = 0 \Rightarrow (0,0)$$

$$x = 0 \quad y = 0$$

$$H_f = \begin{vmatrix} 2 & 0 \\ 0 & 2 \end{vmatrix} = 4 > 0$$

$$f_{xx} = 2 > 0 \quad (0,0) \text{은 } f \text{의 극소점}$$

예제])

$$f(x,y) = x^3 - 12xy + 8y^3$$

$$f_x = 3x^2 - 12y = 0 \quad x^2 = 4y$$

$$f_y = -12x + 24y^2 = 0 \quad 2y^2 = x$$

$$4y^4 = 4y \quad y = 0 \quad y = 1$$

$$x = 0 \quad x = 2$$

$$(0,0) \quad (2,1)$$

$$f_{xx} = 6x$$

$$f_{xy} = -12$$

$$f_{yx} = -12$$

$$f_{yy} = 48y$$

$$H_f(0,0) = \begin{vmatrix} 0 & -12 \\ -12 & 0 \end{vmatrix} = -(12)^2 < 0$$

$$(0,0) : \text{saddle point}$$

$$H_f(2,1) = \begin{vmatrix} 12 & -12 \\ -12 & 48 \end{vmatrix} = 12 \cdot 48 - 12 \cdot 12$$

$$= 12(48 - 12) > 0$$

$$f_{xx}(2,1) = 6 \cdot 2 > 0 \quad \text{극소점}$$

Second derivative test

f is continuous on a disk **centered** at (a,b)

$f_x, f_y, f_{xx}, f_{xy}, f_{yy}$ exist and **continuous** on the same disk

$$D = \begin{vmatrix} f_{xx} & f_{xy} \\ f_{xy} & f_{yy} \end{vmatrix} \quad \text{discriminant}$$

- ① $D > 0, f_{xx} < 0$ f has local max
- ② $D > 0, f_{xx} > 0$ f has local min
- ③ $D < 0$ f has a saddle point
- ④ $D = 0$ test is inconclusive

Def Saddle point

The point (a,b) is **saddle point** of f

if (a,b) is critical pt of f on a disk and continuous at (a,b) . It satisfies that

there are points (x_1, y_1) and (x_2, y_2) in disk

such that $f(x_1, y_1) > f(a, b)$ and $f(x_2, y_2) < f(a, b)$

예제)

$$f(x,y) = x^2 - y^2$$

$(0,0)$ is saddle point of f

$$f(1,0) = 1 > f(0,0)$$

$$f(0,1) = -1 < f(0,0)$$

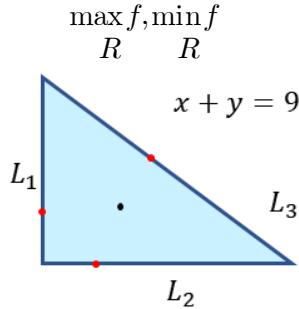
$$f_x = 2x = 0, f_y = -2y = 0$$

$\Rightarrow (0,0)$ is critical point.

예제)

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

영역 $R \equiv x=0, y=0, x+y=9$ 경계로 가는 원



Sol) 임계점, boundary에서 함숫값 조사

$$f_x = 2 - 2x = 0 \quad x = 1$$

$$f_y = 2 - 2y = 0 \quad y = 1$$

$$f_{xx} = -2 \quad f_{xy} = 0 \quad f_{yy} = -2$$

$$H_f = \begin{vmatrix} -2 & 0 \\ 0 & -2 \end{vmatrix} = 4 > 0 \quad , \quad f_{xx} < 0$$

f has a local max at (1,1)

$$f(1,1) = 2 + 2 + 2 - 1 - 1 = 4$$

$$L_1 : x=0 \quad f(0,y) = 2 + 2y - y^2 \quad 0 \leq y \leq 9$$

$$= y(2-y) + 2$$

$$f'(0,y) = 2 - 2y = 0$$

$$y = 1$$

$$\max_{L_1} f = 3, \min_{L_1} f = 9 \cdot (-7) = -63 + 2 = -61$$

$$L_2 : y=0 \quad f(x,0) = 2 - 2x - x^2$$

$$\max_{L_2} f = 3, \min_{L_2} f = -61$$

$$\begin{aligned}
L_3 : f &= 2 + 2 \cdot 9 - x^2 - (9-x)^2 \\
&= 20 - x^2 - (x^2 - 18x + 81) \\
&= -2x^2 + 18x - 61 \\
&= -2(x^2 - 9x) - 61 \\
&= -2\left(x - \frac{9}{2}\right)^2 \\
x &= \frac{9}{2} \\
f &= 20 - 2 \cdot \frac{81}{4} = \frac{-41}{2}
\end{aligned}$$

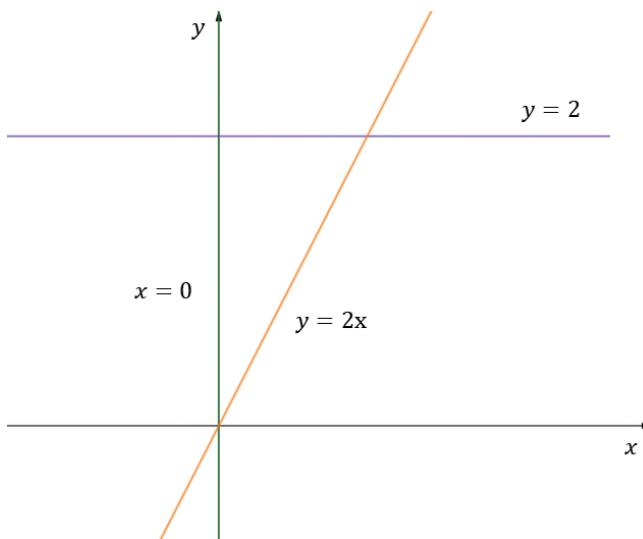
경계	내부 end points	극값
		$f(0,0) = 2$ $f(0,4) = -61$ $f(4,0) = -61$
내부	(1,1)	4

$\max = f = 4, \min = -61$

예제)

$$f(x,y) = 2x^2 - 4x + y^2 - 4y + 1$$

triangular region enclosed by lines $x=0, y=2, y=2x$



내부:

$$f_x = 4x - 4 = 0 \quad x = 1$$

$$f_y = 2y - 4 = 0 \quad y = 2$$

$$f_{xx} = 4, \quad f_{xy} = 0, \quad f_{yy} = 2$$

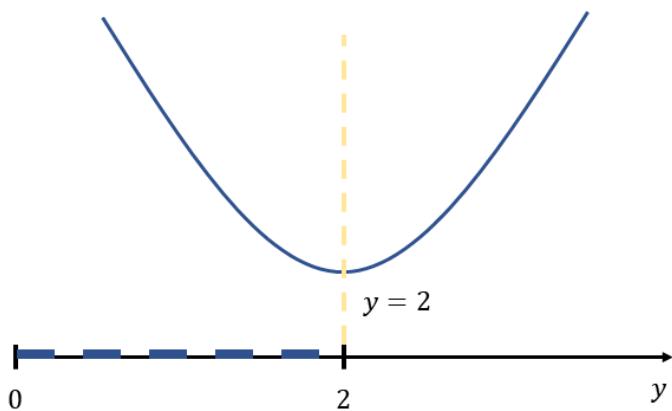
$$D = 8 - 0^2 > 0, \quad f_{xx} > 0$$

local min

$$f(1, 2) = 2 - 4 + 4 - 8 + 1 = -5$$

$$x = 0 \quad 0 < y < 2$$

$$f(0, y) = y^2 - 4y + 1 = (y - 2)^2 - 3$$



$$y = 2 \quad f(x, 2) = 2x^2 - 4x + 4 - 8 + 1$$

$$= 2(x^2 - 2x + 1) - 5$$

$$= 2(x - 1)^2 - 5$$

선 내부에 없음

$$y = 2x \quad f = 2x^2 - 4x + 4x^2 - 8x + 1$$

$$0 < x < 1 \quad = 6x^2 - 12x + 1$$

$$= 6(x^2 - 2x + 1) - 5$$

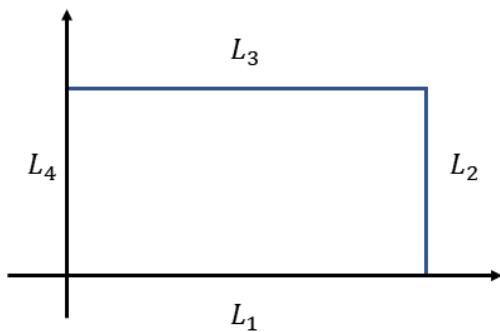
$$= 6(x - 1)^2 - 5$$

corners : $(0, 0), (0, 2), (1, 2)$

예제)

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

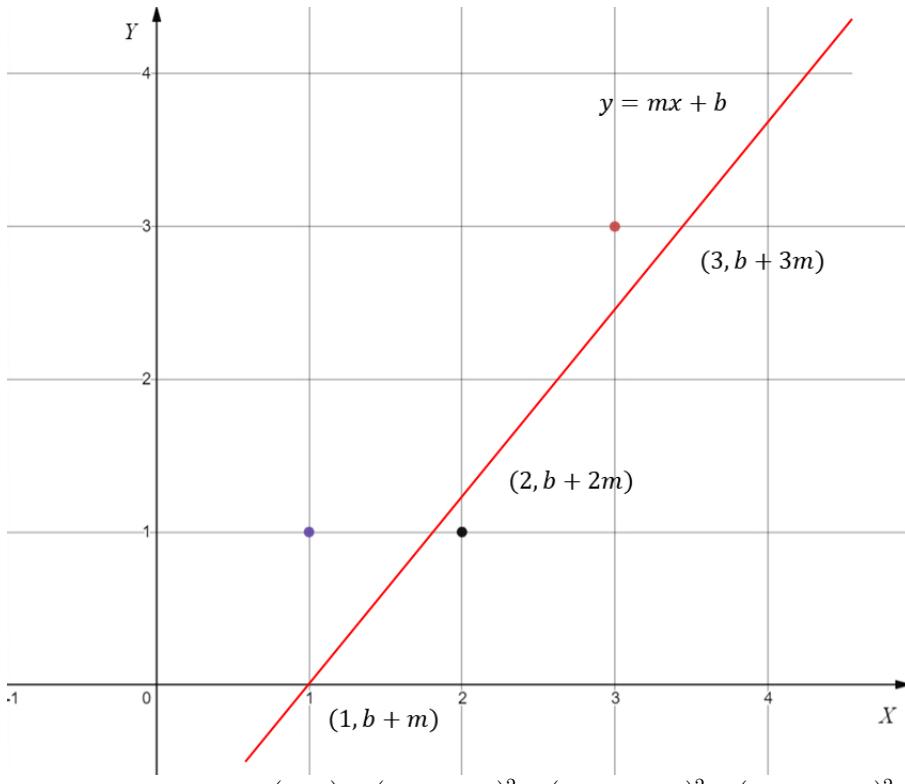
$$R = \{(x, y) \mid 0 \leq x \leq 3, 0 \leq y \leq 2\}$$



응용문제

Least squares lines (Regressive line)

Find least square line for data points $(1,1), (2,1), (3,3)$



$$f(b, m) = (b + m - 1)^2 + (b + 2m - 1)^2 + (b + 3m - 3)^2$$

$$f_b = 2(b + m - 1) + 2(b + 2m - 1) + 2(b + 3m - 3)$$

$$= 6b + (2 + 4 + 6)m - 2 - 2 - 6$$

$$= 6b + 12m - 10$$

$$f_m = -24 + 12b + 28m$$

$$f_k = 0, \quad f_m = 0, \quad b = -\frac{1}{3}, \quad m = 1, \quad f_{bb} = 6 > 0$$

$$D=24$$

$$= f_{bb}f_{mm} - f_{bm}^2$$

$$\textcolor{red}{y=1-\frac{1}{3}}$$

We want to fit the best line to some data in the plane. We measure the distance **from** a line to the data points.

The smallest this sum of squares is the better the line fits the data **general**.

$$\begin{aligned}
f &= \sum (mx_j + b - y_j)^2 \\
&= \sum m^2x_j^2 + b^2 + y_j^2 + 2mbx_j - 2by_j - 2mx_jy_j \\
&= m^2 \sum x_j^2 + 2mb \sum x_j - 2m \sum x_j y_j - 2b \sum y_j + nb^2 + \sum y_j^2 \\
f_m &= 2m \sum x_j^2 + 2b \sum x_j - 2 \sum x_j y_j = 0 \\
f_b &= 2m \sum x_j - 2 \sum y_j + 2nb = 0 \\
\begin{bmatrix} A & B \\ B & n \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} &= \begin{bmatrix} \sum x_j y_j \\ \sum y_j \end{bmatrix} \\
\begin{bmatrix} m \\ b \end{bmatrix} &= \frac{1}{An - B^2} \begin{bmatrix} n & -B \\ -B & A \end{bmatrix} \begin{bmatrix} \sum x_j y_j \\ \sum y_j \end{bmatrix} \\
m &= \frac{n \sum x_j y_j - \sum x_j \sum y_j}{n \sum x_j^2 - (\sum x_j)^2} \\
&= \frac{\sum x_j \sum y_j - n \sum x_j y_j}{(\sum x_j)^2 - n \sum x_j^2} \\
m \sum x_j^2 + b \sum x_j &= \sum x_j y_j \\
m \sum x_j + nb &= \sum y_j \\
b &= \frac{1}{n} (\sum y_j - m \sum x_j) \\
&= \bar{y} - m \bar{x} \\
\sigma^2 &= \sum (x_j - \bar{x})^2 = \sum x_j^2 - 2\bar{x} \sum x_j + n\bar{x}^2 \\
&= \sum x_j^2 - 2n\bar{x}^2 + n\bar{x}^2 \\
&= \sum x_j^2 - nx^2
\end{aligned}$$

$$(\sum x_j)^2 - n \sum x_j^2 = n^2 \overline{x^2} - n \sum x_j^2$$

$$= -n\sigma^2$$

$$\frac{n\sum x_jy_j - \sum x_j\sum y_j}{n\sigma^2} = \frac{\frac{\sum x_jy_j}{n} - \frac{\sum x_j}{n}\frac{\sum y_j}{n}}{\frac{\sigma^2}{n}}$$

$$= \frac{\overline{(x \cdot y)} - \overline{x}\overline{y}}{\frac{\sigma^2}{n}}$$