

EX. 3 Find the local max and min values and saddle points of the following functions

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

1st step: Find critical points, means, find points

$$(x, y, z) \in D_f \text{ such that } \nabla f(x, y, z) = \begin{pmatrix} f_x(x, y, z) \\ f_y(x, y, z) \\ f_z(x, y, z) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\begin{cases} f_x(x, y, z) = -2x + y + 1 = 0 \\ f_y(x, y, z) = -2y + x - z + 1 = 0 \\ f_z(x, y, z) = -2z - y = 0 \end{cases} \rightarrow \begin{aligned} & y = 2x - 1 \\ & 2x - 1 = 2z \\ & \Rightarrow x = \frac{2z + 1}{2} \end{aligned}$$

Substitute $x = \frac{2z + 1}{2}$ and $y = 2z$ in the 2nd equation

$$-2 \cdot 2z + \frac{2z + 1}{2} - z + 1 = 0 \Leftrightarrow -8z + 2z + 1 - 2z + 2 = 0$$

$$\Rightarrow z = \frac{3}{8}, \quad x = \frac{2 \cdot \frac{3}{8} + 1}{2} = \frac{7}{8}, \quad y = 2 \cdot \frac{3}{8} = \frac{3}{4}$$

2nd step: to test the nature of the critical point

$$(x_0, y_0, z_0) = \left(\frac{7}{8}, \frac{3}{4}, \frac{3}{8} \right) \text{ let us calculate the Hessian in}$$

correspondence of $(x_0, y_0, z_0) \Rightarrow$

$$H(x, y, z) = \begin{pmatrix} f_{xx}(x, y, z) & f_{xy}(x, y, z) & f_{xz}(x, y, z) \\ f_{yx}(x, y, z) & f_{yy}(x, y, z) & f_{yz}(x, y, z) \\ f_{zx}(x, y, z) & f_{zy}(x, y, z) & f_{zz}(x, y, z) \end{pmatrix}$$

$$= \begin{pmatrix} -2 & 1 & 0 \\ 1 & -2 & -1 \\ 0 & -1 & -2 \end{pmatrix} \quad \text{if } H(x_0, y_0, z_0) \text{ defines a POSITIVE DEFINITE QUADRATIC FORM} \Rightarrow (x_0, y_0, z_0) \text{ IS A MINIMUM, if}$$

$H(x_0, y_0, z_0)$ defines a NEGATIVE DEFINITE QUADRATIC FORM $\Rightarrow (x_0, y_0, z_0)$ IS A MAXIMUM. We obtain this

information by analyzing the signs of the PRINCIPAL

MINORS $|f_{xx}| = -2 < 0, \quad \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix} = \begin{vmatrix} -2 & 1 \\ 1 & -2 \end{vmatrix} = 4 - 1 = 3 > 0$

NOT ABSOLUTE VALUE but Determinant

$$\begin{vmatrix} f_{xx} & f_{xy} & f_{xz} \\ f_{yx} & f_{yy} & f_{yz} \\ f_{zx} & f_{zy} & f_{zz} \end{vmatrix} = \begin{vmatrix} -2 & 1 & 0 \\ 1 & -2 & -1 \\ 0 & -1 & -2 \end{vmatrix} = \textcircled{2}$$

$$= -8 + 2 + 2 = -4 < 0$$

as these minors are ALTERNATE in signs \Rightarrow
 $H(x_0, y_0, z_0)$ defines a NEGATIVE DEFINITE
 QUADRATIC FORM $\Rightarrow (x_0, y_0, z_0) = \left(\frac{7}{8}, \frac{3}{4}, \frac{3}{8}\right)$ is
 a LOCAL MAXIMUM

EX.4 $f(x,y,z) = \ln(x^2+y^2+z^2+1)$

following the steps of ex.3 \Rightarrow :

(3)

$$\begin{cases} f_x = \frac{2x}{x^2+y^2+z^2+1} = 0 \\ f_y = \frac{2y}{x^2+y^2+z^2+1} = 0 \\ f_z = \frac{2z}{x^2+y^2+z^2+1} = 0 \end{cases}$$

the only solution to this system is $(x_0, y_0, z_0) = (0, 0, 0)$

Let us calculate the Hessian:

$$f_{xx} = \frac{2(x^2+y^2+z^2+1) - 2x \cdot 2x}{(x^2+y^2+z^2+1)^2} = \frac{2x^2+2y^2+2z^2+2-4x^2}{(x^2+y^2+z^2+1)^2} = \frac{2(-x^2+y^2+z^2+1)}{(x^2+y^2+z^2+1)^2}$$

$$f_{xy} = \frac{4xy}{(x^2+y^2+z^2+1)^2}, \quad f_{xz} = -\frac{4xz}{(x^2+y^2+z^2+1)^2}$$

$$f_{yy} = \frac{2(x^2-y^2+z^2+1)}{(x^2+y^2+z^2+1)^2}, \quad f_{yx} = f_{xy}, \quad f_{yz} = -\frac{4yz}{(x^2+y^2+z^2+1)^2}$$

$$f_{zz} = \frac{2(x^2+y^2-z^2+1)}{(x^2+y^2+z^2+1)^2}, \quad f_{zx} = f_{xz}, \quad f_{zy} = f_{yz}$$

$$H(x,y,z) = \begin{pmatrix} f_{xx} & f_{xy} & f_{xz} \\ f_{yx} & f_{yy} & f_{yz} \\ f_{zx} & f_{zy} & f_{zz} \end{pmatrix} = \begin{pmatrix} \frac{2(-x^2+y^2+z^2+1)}{(x^2+y^2+z^2+1)^2} & \frac{4xy}{(x^2+y^2+z^2+1)^2} & -\frac{4xz}{(x^2+y^2+z^2+1)^2} \\ -\frac{4xy}{(x^2+y^2+z^2+1)^2} & \frac{2(x^2-y^2+z^2+1)}{(x^2+y^2+z^2+1)^2} & -\frac{4yz}{(x^2+y^2+z^2+1)^2} \\ -\frac{4xz}{(x^2+y^2+z^2+1)^2} & -\frac{4yz}{(x^2+y^2+z^2+1)^2} & \frac{2(x^2+y^2-z^2+1)}{(x^2+y^2+z^2+1)^2} \end{pmatrix}$$

$$H(0,0,0) = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

Analyzing the principal minors!

$$|f_{xx}| = 2 > 0 \quad \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix} = \begin{vmatrix} 2 & 0 \\ 0 & 2 \end{vmatrix} = 4 > 0 \text{ and}$$

(4)

$$\begin{vmatrix} f_{xx} & f_{xy} & f_{xz} \\ f_{yx} & f_{yy} & f_{yz} \\ f_{zx} & f_{zy} & f_{zz} \end{vmatrix} = \begin{vmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{vmatrix} = 8 > 0$$

we can declare that $(0,0,0)$ is a LOCAL MINIMUM as $H(0,0,0)$ defines a DEFINITE POSITIVE QUADRATIC FORM.

Find the extreme values of the following functions subject to the given constraints

⑤

Ex. 6 $f(x, y) = x^2 y$, $x^2 + 2y^2 = 6 \Leftrightarrow \underbrace{x^2 + 2y^2 - 6}_{g(x)} = 0$

Identify the LAGRANGIAN function

$$\mathcal{L}(x, y, \lambda) = x^2 y - \lambda(x^2 + 2y^2 - 6)$$

Find critical points for $\mathcal{L}(x, y, \lambda)$:

$$\mathcal{L}_x = 2xy - 2\lambda x = 0 \Rightarrow 2x(y - \lambda) = 0 \quad \begin{array}{l} x=0 \\ \text{or} \\ y=\lambda \end{array}$$

$$\mathcal{L}_y = x^2 - 2\lambda y = 0$$

$$\mathcal{L}_\lambda = -(x^2 + 2y^2 - 6) = 0$$

if $x=0 \Rightarrow 2\lambda y=0$ from $x^2 - 2\lambda y=0$ and hence
 $y=0$ IMPOSSIBLE for 3rd equation as we would have $-6=0$!!

$$\lambda=0 \Rightarrow 2y^2 - 6 = 0 \Rightarrow y^2 = 3 \Rightarrow y = \pm\sqrt{3}$$

\Rightarrow the first critical points are:

$$\begin{matrix} x & y & \lambda \\ (0, \sqrt{3}, 0), & (0, -\sqrt{3}, 0) \end{matrix}$$

if instead $y=\lambda \Rightarrow x^2 - 2y^2 = 0 \Rightarrow x^2 = 2y^2 \Rightarrow$

$$2y^2 + 2y^2 - 6 = 0 \Rightarrow 4y^2 - 6 = 0 \Rightarrow y^2 = \frac{3}{2} \Rightarrow y = \pm \frac{\sqrt{3}}{\sqrt{2}} = \pm \frac{\sqrt{6}}{2}$$

$$\Rightarrow x = \pm \sqrt{2}y = \pm \sqrt{3} \text{ and as } y=\lambda, \text{ we have}$$

the following critical points

$$\left(\sqrt{3}, \frac{\sqrt{6}}{2}, \frac{\sqrt{6}}{2} \right), \left(\sqrt{3}, -\frac{\sqrt{6}}{2}, -\frac{\sqrt{6}}{2} \right)$$

$$\left(-\sqrt{3}, \frac{\sqrt{6}}{2}, \frac{\sqrt{6}}{2} \right), \left(-\sqrt{3}, -\frac{\sqrt{6}}{2}, -\frac{\sqrt{6}}{2} \right)$$

Let us calculate now the BORDERED HESSIAN

$$H(x, y, \lambda) = \begin{pmatrix} 0 & g_x & g_y \\ g_x & \mathcal{L}_{xx} & \mathcal{L}_{xy} \\ g_y & \mathcal{L}_{yx} & \mathcal{L}_{yy} \end{pmatrix} = \begin{pmatrix} 0 & 2x & 4y \\ 2x & 2y - 2\lambda & 2x \\ 4y & 2x & -2\lambda \end{pmatrix}$$

$$\begin{aligned} |H(x, y, \lambda)| &= 16x^2y + 16x^2y - 32y^2(y - \lambda) + 8\lambda x^2 = \\ &= 32x^2y + 8\lambda x^2 - 32y^2(y - \lambda) \end{aligned}$$

Recall that if $|H(x_0, y_0, \lambda_0)| > 0 \Rightarrow (x_0, y_0, \lambda_0)$ is a MAX, if $|H(x_0, y_0, \lambda_0)| < 0 \Rightarrow (x_0, y_0, \lambda_0)$ is a MIN. If $|H(x_0, y_0, \lambda_0)| = 0$ this method is inconclusive. Let us analyze point by point:

① $(0, \sqrt{3}, 0)$ $|H(0, \sqrt{3}, 0)| = -32(\sqrt{3})^2 \cdot \sqrt{3} = -96\sqrt{3} < 0$
 $\Rightarrow (0, \sqrt{3}, 0)$ is a MIN

② $(0, -\sqrt{3}, 0)$ $|H(0, -\sqrt{3}, 0)| = -32 \cdot (-\sqrt{3})^2 \cdot (-\sqrt{3}) = 96\sqrt{3} > 0$
 $\Rightarrow (0, -\sqrt{3}, 0)$ is a MAX

③ $|H(\sqrt{3}, \frac{\sqrt{6}}{2}, \frac{\sqrt{6}}{2})| = 32(\sqrt{3})^2 \frac{\sqrt{6}}{2} + 8 \frac{\sqrt{6}}{2} (\sqrt{3})^2 = 0$
 $= 60\sqrt{6} > 0 \Rightarrow (\sqrt{3}, \frac{\sqrt{6}}{2}, \frac{\sqrt{6}}{2})$ is a MAX

④ $|H(\sqrt{3}, -\frac{\sqrt{6}}{2}, -\frac{\sqrt{6}}{2})| = -60\sqrt{6} < 0 \Rightarrow (\sqrt{3}, -\frac{\sqrt{6}}{2}, -\frac{\sqrt{6}}{2})$
 is a MIN

⑤ $|H(-\sqrt{3}, \frac{\sqrt{6}}{2}, \frac{\sqrt{6}}{2})| = 60\sqrt{6} > 0 \Rightarrow (-\sqrt{3}, \frac{\sqrt{6}}{2}, \frac{\sqrt{6}}{2})$
 is a MAX

⑥ $|H(-\sqrt{3}, -\frac{\sqrt{6}}{2}, -\frac{\sqrt{6}}{2})| = -60\sqrt{6} < 0$ is a MIN.

Ex 8 $f(x, y, z) = x^2 + 2y^2 + 3z^2$, $x + y + z = 1$, $x - y + 2z = 2$

1st METHOD: write the variables x and y in terms of z by solving the system

$$\begin{cases} x + y = 1 - z \\ x - y = 2 - 2z \end{cases} \quad \text{(consider } z \text{ as a number/parameter)}$$

→ solve the system with Cramer

$$x = \frac{\begin{vmatrix} 1-z & 1 \\ 2-2z & -1 \end{vmatrix}}{\begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix}} = \frac{z-1-2+2z}{-1-1} = \frac{3z-3}{-2} = \frac{3}{2}(1-z)$$

$$y = \frac{\begin{vmatrix} 1 & 1-z \\ 1 & 2-2z \end{vmatrix}}{-2} = \frac{2-2z-1+z}{-2} = \frac{1-z}{-2} = \frac{z-1}{2}$$

substitute x and y , written in terms of z , in $f(x, y, z) = h(z) =$

$$f(x(z), y(z), z) = \frac{9}{4}(1-z)^2 + 2 \cdot \frac{1}{4}(z-1)^2 + 3z^2$$

$$= \frac{11}{4}(1-z)^2 + 3z^2 = \frac{11}{4}(1-2z+z^2) + 3z^2$$

$$= \frac{23}{4}z^2 - \frac{11}{2}z + \frac{11}{4} \quad \text{this is now a function of ONE variable to which I can}$$

apply optimization in one variable ⇒

$$h'(z) = \frac{23}{2}z - \frac{11}{2} = 0 \Leftrightarrow 23z = 11 \Rightarrow z = \frac{11}{23}$$

$$\text{and } h''(z) = \frac{23}{2} > 0 \Rightarrow z = \frac{11}{23} \text{ is a loc. min}$$

$$\text{and } x = \frac{3}{2}\left(1 - \frac{11}{23}\right) = \frac{3}{2} \cdot \frac{12}{23} = \frac{18}{23}$$

$$y = \frac{1}{2}\left(\frac{11}{23} - 1\right) = \frac{1}{2}\left(-\frac{12}{23}\right) = -\frac{6}{23}$$

2nd METHOD: LAGRANGIAN METHOD

$$\mathcal{L}(x, y, z, \lambda, \mu) = x^2 + 2y^2 + 3z^2 - \lambda(x + y + z - 1) - \mu(x - y + 2z - 2)$$

hence:

$$\begin{cases} \mathcal{L}_x = 2x - \lambda - \mu = 0 \\ \mathcal{L}_y = 4y - \lambda + \mu = 0 \\ \mathcal{L}_z = 6z - \lambda - 2\mu = 0 \\ \mathcal{L}_\lambda = -(\lambda + y + z - 1) = 0 \\ \mathcal{L}_\mu = -(x - y + 2z - 2) = 0 \end{cases}$$

I will write λ and μ in terms of x and y

$$\begin{cases} \lambda + \mu = 2x \\ \lambda - \mu = 4y \end{cases} \rightarrow \lambda = \frac{\begin{vmatrix} 2x & 1 \\ 4y & -1 \end{vmatrix}}{-2} = \frac{-2x - 4y}{-2} = x + 2y$$

$$\mu = \frac{\begin{vmatrix} 1 & 2x \\ 1 & 4y \end{vmatrix}}{-2} = \frac{4y - 2x}{-2} = x - 2y$$

Substitute λ and μ in the third equation

$6z - x - 2y - 2x + 4y = 0 \Rightarrow$ joined with the last two equations, we obtain:

$$\begin{cases} -3x + 2y + 6z = 0 \\ x + y + z = 1 \\ x - y + 2z = 2 \end{cases}$$

I will solve this system through Cramer

$$x = \frac{\begin{vmatrix} 0 & 2 & 6 \\ 1 & 1 & 1 \\ 2 & -1 & 2 \end{vmatrix}}{\begin{vmatrix} -3 & 2 & 6 \\ 1 & 1 & 1 \\ 1 & -1 & 2 \end{vmatrix}}} = \frac{18}{23}$$

$$\lambda = x + 2y = \frac{18}{23} - \frac{12}{23} = \frac{6}{23}$$

$$y = \frac{\begin{vmatrix} -3 & 0 & 6 \\ 1 & 1 & 1 \\ 1 & 2 & 2 \end{vmatrix}}{-23} = -\frac{6}{23}$$

$$\mu = x - 2y = \frac{18}{23} + \frac{12}{23} = \frac{30}{23}$$

$$z = \frac{\begin{vmatrix} -3 & 2 & 0 \\ 1 & 1 & 1 \\ 1 & -1 & 2 \end{vmatrix}}{-23} = \frac{11}{23}$$

We will test the nature of this critical point

$$\left(\frac{18}{23}, -\frac{6}{23}, \frac{11}{23}, \frac{6}{23}, \frac{30}{23} \right) \text{ through the analysis}$$

of the ~~Bordered~~ Hessian. Let us briefly

Summarize some results for testing nature of critical points.

Consider the general problem of maximizing or minimizing $f: \mathbb{R}^n \rightarrow \mathbb{R}$ over the set $(g_1(x))$

$$\tilde{D} = \bigcup \{x \mid g(x) = 0\} \rightarrow g(x) = \begin{pmatrix} g_1(x) \\ \vdots \\ g_k(x) \end{pmatrix}$$

where $g: \mathbb{R}^n \rightarrow \mathbb{R}^k$ (these are the constraints, if you have 1 constraint $\rightarrow g: \mathbb{R}^n \rightarrow \mathbb{R}$, if you have two constraints $g: \mathbb{R}^n \rightarrow \mathbb{R}^2, \dots$)

We will assume that f and g are both C^2 functions (continuous functions with continuous first and second derivative). We form the Lagrangian

$$\mathcal{L}(x, \lambda) = f(x) + \sum_{i=1}^k \lambda_i g_i(x)$$

$\lambda \in \mathbb{R}^k \Rightarrow \lambda = \begin{pmatrix} \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_k \end{pmatrix}$

Let us consider the Hessian matrix of $\mathcal{L}(x, \lambda)$:

$$H = \begin{pmatrix} \frac{\partial^2 \mathcal{L}}{\partial \lambda_1^2} & \frac{\partial^2 \mathcal{L}}{\partial \lambda_1 \partial \lambda_2} & \dots & \frac{\partial^2 \mathcal{L}}{\partial \lambda_1 \partial \lambda_k} & \frac{\partial^2 \mathcal{L}}{\partial x_1 \partial \lambda_1} & \dots & \frac{\partial^2 \mathcal{L}}{\partial x_n \partial \lambda_1} \\ \vdots & \vdots & & \vdots & \vdots & & \vdots \\ \frac{\partial^2 \mathcal{L}}{\partial \lambda_k \partial \lambda_1} & \frac{\partial^2 \mathcal{L}}{\partial \lambda_k \partial \lambda_2} & \dots & \frac{\partial^2 \mathcal{L}}{\partial \lambda_k^2} & \frac{\partial^2 \mathcal{L}}{\partial x_1 \partial \lambda_k} & \dots & \frac{\partial^2 \mathcal{L}}{\partial x_n \partial \lambda_k} \\ \frac{\partial^2 \mathcal{L}}{\partial \lambda_1 \partial x_1} & \dots & \frac{\partial^2 \mathcal{L}}{\partial \lambda_k \partial x_n} & \frac{\partial^2 \mathcal{L}}{\partial x_1^2} & \dots & \frac{\partial^2 \mathcal{L}}{\partial x_n \partial x_1} \\ \vdots & & \vdots & \vdots & & & \vdots \\ \frac{\partial^2 \mathcal{L}}{\partial \lambda_1 \partial x_n} & \dots & \frac{\partial^2 \mathcal{L}}{\partial \lambda_k \partial x_n} & \frac{\partial^2 \mathcal{L}}{\partial x_1 \partial x_n} & \dots & \frac{\partial^2 \mathcal{L}}{\partial x_n^2} \end{pmatrix}$$

$= D^2 \mathcal{L}(x, \lambda)$

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$$\begin{pmatrix}
 0 & \dots & 0 & \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_1}{\partial x_n} \\
 \vdots & & \vdots & \vdots & & \vdots \\
 0 & \dots & 0 & \frac{\partial g_k}{\partial x_1} & \dots & \frac{\partial g_k}{\partial x_n} \\
 \hline
 \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_k}{\partial x_1} & \frac{\partial^2 \mathcal{L}}{\partial x_1^2} & \dots & \frac{\partial^2 \mathcal{L}}{\partial x_n \partial x_1} \\
 \vdots & & \vdots & \vdots & & \vdots \\
 \frac{\partial g_1}{\partial x_n} & \dots & \frac{\partial g_k}{\partial x_n} & \frac{\partial^2 \mathcal{L}}{\partial x_1 \partial x_n} & \dots & \frac{\partial^2 \mathcal{L}}{\partial x_n^2}
 \end{pmatrix}
 = D_{(\lambda, x)}^2 \mathcal{L}(x, \lambda) = \bar{H}$$

called **BORDERED HESSIAN**: this matrix

plays the role of normal Hessian matrix in constrained problems for second order conditions

Hence, suppose there exists a point $x^* \in \tilde{D}$ (recall this is the set of points that satisfy the constraints) and $\lambda^* \in \mathbb{R}^k$ such that $\text{rank } Dg(x^*) = k$

(this is the **JACOBIAN** of the constraint functions, i.e. the matrix of the first ^{partial} derivatives of the constraints)

$$Dg(x) = \begin{pmatrix} \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_k}{\partial x_1} \\ \vdots & & \vdots \\ \frac{\partial g_1}{\partial x_n} & \dots & \frac{\partial g_k}{\partial x_n} \end{pmatrix}$$

and $Df(x^*) + \sum_{i=1}^k \lambda_i^* Dg_i(x^*) = 0$ (FIRST ORDER CONDITIONS,

this is what we do when we calculate the partial first derivatives of the Lagrangian, and we set them to 0). Consider the bordered Hessian given above \bar{H} . Let \bar{H}_r denote the r -th order leading principal submatrix of $\bar{H} \Rightarrow$:

- 1) if $(-1)^{r-k} |\bar{H}_r| > 0$ for all $r = 2k+1, \dots, n+k$ then x^* is a STRICT LOCAL MAXIMIZER of f on \bar{D}
- 2) if $(-1)^k |\bar{H}_r| > 0$ for all $r = 2k+1, \dots, n+k$ then x^* is a STRICT LOCAL MINIMIZER of f on \bar{D}
- 3) if either of the above conditions is violated by nonzero leading minors, then x^* is neither a local maximizer nor a local minimizer

Let us go back to our exercise. We have to test the nature of $(\frac{18}{23}, -\frac{6}{23}, \frac{11}{23}, \frac{6}{23}, \frac{30}{23})$

and the bordered Hessian is

$$\bar{H} = \begin{pmatrix} 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & -1 & 2 \\ 1 & 1 & 2 & 0 & 0 \\ 1 & -1 & 0 & 4 & 0 \\ 1 & 2 & 0 & 0 & 6 \end{pmatrix} \quad \begin{array}{l} \text{in our case:} \\ \rightarrow (x, y, z) \\ f: \mathbb{R}^3 \rightarrow \mathbb{R} \\ n = 3 \\ g(x, y, z) = \begin{pmatrix} g_1(x, y, z) \\ g_2(x, y, z) \end{pmatrix} \end{array}$$

$$= \begin{pmatrix} x+y+z-1 \\ x-y+2z-2 \end{pmatrix} \quad \text{hence } g(x, y, z): \mathbb{R}^3 \rightarrow \mathbb{R}^2 \rightarrow K=2$$

hence we have to analyze the sign of $|\bar{H}_r|$ for $r = 2k+1, \dots, n+k = 2 \cdot 2 + 1, \dots, 3+2 = 5$

this means we have to see the sign of $|\bar{H}|$, only case

- 1) if $(-1)^{r-k} |\bar{H}_r| > 0$ $r = 2k+1, \dots, n+k \Rightarrow x^*$ is a Max, and in this case $r = 5, k = 2 \Rightarrow 5 - 2 = 3$
 $\Rightarrow (-1)^3 = -1 \Rightarrow |\bar{H}_5| = |\bar{H}| < 0$ if this condition has to verify \Rightarrow if $|\bar{H}| < 0 \Rightarrow x^*$ is Max

2) if instead $(-1)^k |\bar{H}_r| > 0$ $r=2k+1, \dots, n+k$
 $\Rightarrow x^*$ is a MIN, \Rightarrow here $k=2$ $\xrightarrow{r=5} (-1)^2 = 1$
 \Rightarrow for this condition to verify I want (12)
 $|\bar{H}_5| = |\bar{H}| > 0 \Rightarrow$ here if $|\bar{H}| > 0 \Rightarrow x^*$ is a MIN

Hence let us check the sign of $|\bar{H}|$:

$$\begin{vmatrix} 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & -1 & 2 \\ 1 & 1 & 2 & 0 & 0 \\ 1 & -1 & 0 & 4 & 0 \\ 1 & 2 & 0 & 0 & 6 \end{vmatrix} = (-1)^{5+1} \cdot 1 \cdot \begin{vmatrix} 0 & 1 & 1 & 1 \\ 0 & 1 & -1 & 2 \\ 1 & 2 & 0 & 0 \\ -1 & 0 & 4 & 0 \end{vmatrix}$$

Laplace according to the last row first

$$+ (-1)^{5+2} \cdot 2 \cdot \begin{vmatrix} 0 & 1 & 1 & 1 \\ 0 & 1 & -1 & 2 \\ 1 & 2 & 0 & 0 \\ 1 & 0 & 4 & 0 \end{vmatrix} + (-1)^{5+5} \cdot 6 \cdot \begin{vmatrix} 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & -1 \\ 1 & 1 & 2 & 0 \\ 1 & -1 & 0 & 4 \end{vmatrix}$$

$$= (-1)^{3+1} \cdot 1 \cdot \begin{vmatrix} 1 & 1 & 1 \\ 1 & -1 & 2 \\ 0 & 4 & 0 \end{vmatrix} + (-1)^{4+1} \cdot (-1) \cdot \begin{vmatrix} 1 & 1 & 1 \\ 1 & -1 & 2 \\ 2 & 0 & 0 \end{vmatrix} - 2 \cdot (-1)^{3+1} \cdot 1 \cdot \begin{vmatrix} 1 & 1 & 1 \\ 1 & -1 & 2 \\ 0 & 4 & 0 \end{vmatrix}$$

determinant through Laplace according to the first column

$$+ (-1)^{4+1} \cdot 1 \cdot \begin{vmatrix} 1 & 1 & 1 \\ 1 & -1 & 2 \\ 2 & 0 & 0 \end{vmatrix} + 6 \cdot (-1)^{3+1} \cdot 1 \cdot \begin{vmatrix} 0 & 1 & 1 \\ 0 & 1 & -1 \\ -1 & 0 & 4 \end{vmatrix} + (-1)^{4+1} \cdot 1 \cdot \begin{vmatrix} 0 & 1 & 1 \\ 0 & 1 & -1 \\ 1 & 2 & 0 \end{vmatrix}$$

$$= -4 + 6 - 2(-4 - 6) + 6(2 + 2) = +2 - 2(-10) + 24$$

$$= 46 > 0 \Rightarrow \text{the point } \left(\frac{18}{23}, -\frac{6}{23}, \frac{11}{23} \right) \text{ with}$$

LAGRANGE MULTIPLIERS $(\lambda, \mu) = \left(\frac{6}{23}, \frac{30}{23} \right)$

is a CONSTRAINED LOCAL MINIMIZER.