* Non-Linear Programming:

An optimisation problem in which either the objective function and for some or all constraints are non-linear is called a non-linear programming problem (NLPP)

eg. optimise z = x1 + x2 - 5x1x2 Subject to x12 + x1x2 + 5x3 = 70

H1, 82, 7370

· Quadratic Programming Problem:-The objective function is of the type:

 $Z = a_{11} x_1^2 + a_{22} x_2^2 + - - + a_{nn} x_n^2 + a_{12} x_1 x_2 + a_{13} x_1 x_3 + - - + a_{1n} x_1 x_n +$ a23x1 x3 + a24 x2 x4+ --- + a2n x2 xn + ... + Gx1+C2x2+-.. + Cnxn

· Method to solve Quadratic Programming Problem:-

- Assume that the first of second order partial derivatives i.e. of a def exists vi,j, where f(x,,x2....xn) is the objective function we want to maximise / minimise.

- Find the Hessian matrix, which is given by,

$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} & \frac{\partial^2 f}{\partial x_2^2} \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

- we need to check whether this matrix is positive definite, negative definite or indefinite.

- Given a matrix Anxn, the principal minors are,

Given a matrix
$$A_{1}$$
 A_{1} A_{2} A_{3} A_{1} A_{2} A_{3} A_{3} A_{4} A_{4} A_{4} A_{4} A_{4} A_{5} A_{7} A_{8} A_{1} A_{1} A_{1} A_{2} A_{3} A_{4} A_{4} A_{5} A_{5} A_{6} A_{7} A_{7} A_{8} A_{8}

Let the determinant of the matrices be

 $D_1 = |A_1|$, $D_2 = |A_2|$, ---, $D_n = |A_n|$ Prof. Anushri Tambe

- If all the determinants,
 - . D., Dz ..., Dn are positive then A is positive definite
 - · D., D3, D5, ---- are negative of D2, D4, --- are positive then.

 A is negative definite.
 - · If a matrix is neither positive definite nor negative definite, then it is indefinite.
- Find the status of the Hessian matrix H at a stationary point x.
 - · If H is positive definite at a, it has a minima at no
 - . If H is negative definite of xo, it has a maxima at xo
 - . If H is indefinite at xo, it has saddle point at xo

Note: To find stationary points of a function,

Put
$$\frac{\partial f}{\partial x_i} = 0$$
, $\frac{\partial f}{\partial x_i} = 0$, $\frac{\partial f}{\partial x_i} = 0$

on solving these equations, we get stationary point $x_0 = (x_1, x_2, ..., x_n)$

* Examples :-

-> First we need to find the stationary points,

Let
$$f(x_1, x_2, x_3) = x_1^2 + x_2^2 + x_3^2 - 6x_1 - 8x_2 - 10x_3$$

$$\frac{\partial f}{\partial x} = 0 \implies 2x, -6 = 0 \implies \frac{2x}{3}$$

$$\frac{\partial f}{\partial x_2} = 0 \Rightarrow 2x_2 - 8 = 0 \Rightarrow \frac{x_2 - 4}{}$$

$$\frac{\partial f}{\partial x_3} = 0 \quad \Rightarrow \quad 2x_3 - 10 = 0 \quad \Rightarrow \quad \frac{x_3 = 5}{2x_3}$$

.. a = (3,4,5) is the stationary point.

Now we find the second order partial derivatives,

$$\frac{\partial^2 f}{\partial x_1^2} = 2 , \quad \frac{\partial^2 f}{\partial x_2^2} = 2 , \quad \frac{\partial^2 f}{\partial x_3^2} = 2$$

$$\frac{\partial^2 f}{\partial x_1 \partial x_2} = \frac{\partial^2 f}{\partial x_1 \partial x_3} = \frac{\partial^2 f}{\partial x_2 \partial x_3} = \frac{\partial^2 f}{\partial x_3 \partial x_3} = \frac{\partial^2 f}{\partial x_3 \partial x_2} = 0$$

$$\therefore \text{ The Hessian matrix is } H = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$
Let
$$A_1 = \begin{bmatrix} 2 \end{bmatrix}, A_2 = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, A_3 = H$$

$$\therefore D_1 = 2, D_2 = 4, D_3 = 8$$
all $D_1, D_2, D_3 > 0$

$$H \text{ is positive definite}$$

$$\therefore f(x_1, x_2, x_3) = Z = x_1^2 + x_2^2 + x_3^2 - x_1 - 8x_2 - 10x_3 \text{ has minimum.}}$$
at $x_0 (3, 4, 5)$

$$\therefore \text{ The minimum value of } Z \text{ is, } Z_{min} = 9 + 16 + 25 - 18 - 32 - 50$$

$$\therefore \overline{Z_{min}} = -50$$

$$x) \text{ Obtain the relative maximum or minimum (if any) of the function } Z = x_1 + 2x_3 + x_2x_3 - x_1^2 - x_2^2 - x_3^2$$

$$\Rightarrow \text{Let } f(x_1, x_2, x_3) = x_1 + 2x_3 + x_2x_3 - x_1^2 - x_2^2 - x_3^2$$

$$\Rightarrow \text{Let } f(x_1, x_2, x_3) = x_1 + 2x_3 + x_2x_3 - x_1^2 - x_2^2 - x_3^2$$
The stationary points are given by,
$$\frac{\partial f}{\partial x_1} = 0 \Rightarrow 1 - 2x_1 = 0 \Rightarrow \overline{|x_1 = y_2|}$$

$$\frac{\partial f}{\partial x_2} = 0 \Rightarrow x_3 - 2x_2 = 0 \qquad \text{(i)} \quad f \frac{\partial f}{\partial x_3} = 0 \Rightarrow x_2 - 2x_3 = -2 \qquad \text{(2)}$$
Solving (1) $f(x_1, x_2, x_3) = x_1 + x_2 + x_3 + x_2 + x_3 + x_2 + x_3 + x_3 + x_4 +$

The Hessian matrix is
$$H = \begin{bmatrix} -2 & 0 & 0 \\ 0 & -2 & 1 \\ 0 & 1 & -2 \end{bmatrix}$$

$$A_1 = \begin{bmatrix} -2 \end{bmatrix}, A_2 = \begin{bmatrix} -2 & 0 \\ 0 & -2 \end{bmatrix}, A_3 = H$$

$$D_1 = -2, D_2 = 4, D_3 = -6$$
Here $D_1, D_3 \ge 0$ of $D_2 \ge 0$ in H is negative definite
$$f(x_1, x_2, x_3) \text{ has } \underline{\text{maximum}} \text{ at } x_0 = \left(\frac{1}{2}, \frac{2}{3}, \frac{4}{3}\right)$$

$$Z_{max} = \frac{1}{2} + 2\left(\frac{4}{3}\right) + \left(\frac{2}{3}\right)\left(\frac{4}{3}\right) - \frac{1}{4} - \frac{4}{9} - \frac{16}{9}$$

$$= \frac{1}{2} + \frac{8}{3} + \frac{8}{9} - \frac{1}{4} - \frac{20}{9} = \frac{18 + 96 + 32 - 9 - 80}{36} = \frac{57}{36}$$

$$Z_{mag} = \frac{19}{12}$$

Practice Problems: -

Find the relative maximum or minimum of the function (i) $Z = 2x_1 + 6x_3 + 9x_2x_3 - 4x_1^2 - 9x_2^2 - 9x_3^2$ (ii) $Z = 2x_1 + 6x_3 + 9x_2x_3 - 6x_1 - 10x_2 - 14x_3 + 103$

* Optimisation with equality constraints:
A NLPP in which the objective function is non-linear but the constraints are linear.

i.e. Optimise $z = f(x_1, x_2, \dots, x_n)$ subject to $g_1(x_1, x_2, \dots, x_n) = b_1$, $g_2(x_1, x_2, \dots, x_n) = b_2$, \vdots $g_n(x_1, x_2, \dots, x_n) = b_m$ $x_1, x_2, \dots, x_n > 0$

This type of problem is solved by forming Lagrangian Function with Lagrange's multiplier >

(a) NLPP with n-variables
$$f$$
 one equality constraint:- $\frac{3}{2}$ optimise $Z = f(x_1, x_2, ..., x_n)$
Subject to $g(x_1, x_2, ..., x_n) = b$
 $x_1, x_2, ..., x_n > 0$

-first express the constraints with RHS equal to zero i.e. optimise $z = f(x_1, x_2, ..., x_n)$

Subject to
$$h(x_1, \dots, x_n) = g(x_1, \dots, x_n) - b$$

 $x_1, x_2, \dots, x_n > 0$

- The Lagrangian function is (constructed as) $L(\varkappa_1,\varkappa_2,...,\varkappa_n,\lambda) = f(\varkappa_1,...,\varkappa_n) \lambda h(\varkappa_1,...,\varkappa_n) \ldots$ where λ is called Lagrangian multiplier.
- The necessary condition for maxima or minima subject to the constraint $h(x_1, x_2, ..., x_n) = 0$ are

$$\frac{\partial L}{\partial x_1} = 0 , \frac{\partial L}{\partial x_2} = 0 , \dots, \frac{\partial L}{\partial x_n} = 0 , \frac{\partial L}{\partial x} = 0$$
 (2)

$$(1) = \frac{\partial L}{\partial x_{1}} = \frac{\partial f}{\partial x_{1}} - \lambda \frac{\partial h}{\partial x_{1}} , \quad \frac{\partial L}{\partial x_{2}} = \frac{\partial f}{\partial x_{2}} - \lambda \frac{\partial h}{\partial x_{2}} , \quad ----, \frac{\partial L}{\partial x_{n}} = \frac{\partial f}{\partial x_{n}} - \lambda \frac{\partial h}{\partial x_{n}}$$

$$f \frac{\partial L}{\partial \lambda} = -\lambda h \qquad ---- (3)$$

Using (2), we get from (3), the following (n+1) necessary cond? $\frac{\partial f}{\partial x_1} - \lambda \frac{\partial h}{\partial x_2} = 0 , \quad \frac{\partial f}{\partial x_2} - \lambda \frac{\partial h}{\partial x_2} = 0 , \quad -\cdots, \quad \frac{\partial f}{\partial x_n} - \lambda \frac{\partial h}{\partial x_n} = 0$

solving these equations, we get x,, x2, ---, xn, \lambda. i.e. we obtain the point of maxima or minima.

- To check the point obtained above is maximal minima, consider the following determinant

$$\Delta_{n+1} = \begin{vmatrix} \frac{\partial h}{\partial x_1} & \frac{\partial^2 f}{\partial x_2} - \lambda \frac{\partial^2 h}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} - \lambda \frac{\partial^2 h}{\partial x_1 \partial x_2} - \lambda \frac{\partial^2 h}{\partial x_1 \partial x_2} - - - \frac{\partial^2 f}{\partial x_1 \partial x_n} - \lambda \frac{\partial^2 h}{\partial x_1 \partial x_n} - \lambda \frac{\partial^2 h}{\partial x_2 \partial x_n} - \lambda \frac{\partial^2 h}{\partial x_n \partial x_2} -$$

· If the principal minors b3, b4, b5, -- are alternately positive & negative i.e. 2070, 24<0, 2570,... then pt. xo is maxima . If all D3, D4, D5, --- are negative, i.e. D3 <0, D4 <0, --then to is minima

Note: ti) If z is a function of two variables only, then we get only so.

If A3 is positive then to is maxima If do is negative then to is minima

(ii) If z is a function of three variables then we get 63 f D4

If both by & by are negative then Xo is minima If Date of Duco then xo is maxima.

Examples: -

i) Using Lagrange's Multipliers, solve the following NLPP (i) Optimise Z = 4x, +8x2- x12- x2 Subject to 21+ 2= 2 21,7270

 \rightarrow NLPP is optimise $Z = 4x_1 + 8x_2 - x_1^2 - x_2^2$ subject to 21+ x2-2=0 H1, H27, 0

The Lagrangian function is L(x,, x2, x)=(4x,+8x2-x1-x2)-x(x,+x2-2)

.. the partial derivatives are $\frac{\partial L}{\partial x_1} = 4 - 2x_1 - \lambda$, $\frac{\partial L}{\partial x_2} = 8 - 2x_2 - \lambda$, $\frac{\partial L}{\partial \lambda} = -(x_1 + x_2 - 2)$ solving the equations, $\frac{\partial L}{\partial x_i} = 0$, $\frac{\partial L}{\partial x_i} = 0$, $\frac{\partial L}{\partial \lambda} = 0$ we get (2)+(3) = $12-2(x_1+x_2)-2\lambda=0$ $4-2x, -\lambda = 0$ — (2) = 12-2(2)=2 x from (3) 8-2×2-1 =0 -(3) $-(x_1+x_2-2)=0$ — (4) : (1) => 2x, = 4-4 =0 => [x, =0] (2) => $2x_2 = 8-4=4$ => $x_2 = 2$.. $x_0 = (0,2)$ Now, h(x1, x2) = x1 + x2 - 2 = 0 - dh = 1, dh = 1 & all other partial derivatives are zero. $f(x_1, x_2) = 4x_1 + 8x_2 - x_1^2 - x_2^2$ $\frac{\partial f}{\partial x} = 4 - 2x, \quad , \quad \frac{\partial^2 f}{\partial x_0} = 0 \quad , \quad \frac{\partial^2 f}{\partial x_0^2} = -2$ $\frac{\partial f}{\partial x_2} = 8 - 2x_2$, $\frac{\partial^2 f}{\partial x_1 \partial x_2} = 0$, $\frac{\partial^2 f}{\partial x_2^2} = -2$ $\Delta_3 = \begin{vmatrix} \frac{\partial h}{\partial x_1} & \frac{\partial^2 h}{\partial x_2} \\ \frac{\partial h}{\partial x_1} & \frac{\partial^2 h}{\partial x_1^2} - \lambda \frac{\partial^2 h}{\partial x_1^2} \\ \frac{\partial h}{\partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_1} - \lambda \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial h}{\partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_1} - \lambda \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_1} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial x_2 \partial x_2} \\ \frac{\partial^2 h}{\partial x_2 \partial x_2} & \frac{\partial^2 h}{\partial$ = 0-1(-2)+1(2) = 4

Here D3 = 470 - Xo is maxima

(ii) Optimise
$$z = 12 \times 1 + 8 \times 2 + 6 \times 3 - 21^2 -$$

Where $f(x_1, x_2, x_3) = 12x_1 + 8x_2 + 6x_3 - x_1^2 - x_2^2 - x_3^2 - 23$ $f(x_1, x_2, x_3) = x_1 + x_2 + x_3 - 10$

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

$$21 = 5 \qquad \boxed{21 = 3} \qquad \boxed{21 = 2}$$

* Practice Problems:

Using Lagrange's multipliers, solve the following MLPP

(i) Optimise
$$Z = 6\pi_1^2 + 5\pi_2^2$$

subject to $3\pi_1 + 5\pi_2 = 3$

(ii) Optimise
$$z = 2\pi_1 + 6\chi_2 - \chi_1^2 - \chi_2^2 + 14$$

Subject to $\pi_1 + 3\chi_2 = 4$
 $\pi_1, 3\chi_2 > 0$

(iii) aptimise
$$z = 231, + 32 + 333 + 1031,$$

 $+832 + 633 - 100$
54bject to $31 + 32 + 33 = 20$
 $31132, 31370$

(iv) optimise
$$z = 3x_1^2 + x_2^2 + x_3^2$$

subject to $x_1 + x_2 + x_3 = 2$
 x_1, x_2, x_3, x_3

(b) NLPP with n variables of more than one (m) equality constraints
$$(m \times n)$$
:-

Optimise $Z = f(x_1, x_2, ..., x_n)$

Subject to $h_1(x_1, ..., x_n) = 0$
 $h_2(x_1, ..., x_n) = 0$
 $h_1(x_1, ..., x_n) = 0$

The Aagrangian function with m multipliers $\lambda_1, ..., \lambda_n$ is

 $L(x_1, ..., x_n, \lambda) = f(x_1, ..., x_n) - \lambda_1 h_1(x_1, ..., x_n) - ... - \lambda_m h_m(x_1, ..., x_n)$

The necessary conditions for maxima / minima is,

 $\frac{\partial L}{\partial x_1} = 0$, $i = b^2, ..., n$, $\frac{\partial L}{\partial \lambda_1} = 0$, $j = 1, ..., m$

solving these equations, we get stationary point To decide whether the point is maxima/minima, consider bordered Hessian matrix denoted by HB

$$H^{B} = \begin{bmatrix} 0 & i & P \\ -p^{i} & i & Q \end{bmatrix}_{(m+n) \times (m+n)}$$

where
$$0 = \begin{bmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix} \quad \neq \quad P = \begin{bmatrix} \frac{\partial h_1}{\partial x_1} & \frac{\partial h_1}{\partial x_2} & \cdots & \frac{\partial h_1}{\partial x_n} \\ \frac{\partial h_2}{\partial x_1} & \frac{\partial h_2}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_1} & \cdots & \frac{\partial h_n}{\partial x_n} & \cdots \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial h_n}{\partial x_n} \\ \frac{\partial h_n}{\partial x_n} & \cdots & \frac{\partial$$

$$Q = \begin{bmatrix} \frac{\partial^2 L}{\partial x_1^2} & \frac{\partial^2 L}{\partial x_1 \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_1 \partial x_n} \\ \frac{\partial^2 L}{\partial x_2 \partial x_1} & \frac{\partial^2 L}{\partial x_2^2} & \cdots & \frac{\partial^2 L}{\partial x_2 \partial x_n} \\ \frac{\partial^2 L}{\partial x_2 \partial x_1} & \frac{\partial^2 L}{\partial x_2 \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n \partial x_2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_1} & \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} \\ \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2 L}{\partial x_n^2} & \cdots \\ \frac{\partial^2 L}{\partial x_n^2} & \cdots & \frac{\partial^2$$

& P' is transpose of P

- · Condition for maxima & minima:

 The nature of the function at to is determined by the signs of (n-m) principal minors of the matrix HB

 (i) start with the principal minor of order (2m+1) & check the signs of (n-m) principal minors. If these signs are alternately positive & negative, starting with common then the to maxima.
- (ii) If the signs of these minors are (-1)^m, then χ_0 is minimal Note: (i) If there are two unknowns of two linear constraints then bordered Hessian matrix H^0 is not useful. So find the signs of two principal minors of $\begin{bmatrix} \frac{\partial z}{\partial x_1^2} & \frac{\partial z}{\partial x_1 \partial x_2} \\ \frac{\partial z}{\partial x_2 \partial x_1} & \frac{\partial z}{\partial x_2} \end{bmatrix}$ Find determinant of both principal minors $\begin{bmatrix} \frac{\partial z}{\partial x_1} & \frac{\partial z}{\partial x_2} \\ \frac{\partial z}{\partial x_2 \partial x_1} & \frac{\partial z}{\partial x_2} \end{bmatrix}$
 - (i) If both A, & Az are positive then xo is minima.
 - (ii) If A is negative of Az is positive then xo is maxima
 - (2) Geometrically, the constraints are two straight lines of they intersect in only one point at the most.
 - (3) If there are three unknowns of two constraints, then simplify the bordered matrix HB

 If its sign is negative, then xo is maxima

 of it its sign is positive, then xo is minima
 - (4) Geometrically the constraints are the two planes of they intersect in a line.

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Examples: Using Lagrangian multipliers, solve the following NLPP
1) Optimise z = 42, +922 - 21 - 22
   subject to 4x1 + 3x2 = 15
                    3x1 + 5x2 = 14
                     21, 21270
-> Let f(x1, x2) = 4x1 +9x2 - x12 - x22
              h, (x,, x2) = 4x, +3x2-15
               h_2(x_1, x_2) = 3x_1 + 5x_2 - 14
        Lagrangian function is
  L(x_1, x_2, \lambda_1, \lambda_2) = f(x_1, x_2) - \lambda_1 h_1(x_1, x_2) - \lambda_2 h_2(x_1, x_2)
: L(x, x2, \1, \2) = 4x, + 9x2 - x, - x2 - \1 (4x, + 3x2-15) - \2 (3x, +5x2-14)
\frac{\partial L}{\partial x_1} = 4 - 2\lambda_1 - 4\lambda_1 - 3\lambda_2 \qquad \frac{\partial L}{\partial x_2} = 9 - 2\lambda_2 - 3\lambda_1 - 5\lambda_2
  \frac{\partial L}{\partial \lambda} = -(4x_1 + 3x_2 - 15), \frac{\partial L}{\partial \lambda} = -(3x_1 + 5x_2 - 14)
 Solving, \frac{\partial L}{\partial x_1} = 0, \frac{\partial L}{\partial x_2} = 0, \frac{\partial L}{\partial \lambda_1} = 0, \frac{\partial L}{\partial \lambda_2} = 0
 321+522-14=0 -- (4)
       4x1 +3x2 -15 =0 - (3)
4 \times (1) + 3 \times (2) = 16 - 8 \times (-16 \lambda_1 - 12 \lambda_2 = 6
                          + 27 - 6 x2 - 921 - 1521 = 0
( we wont eq
                            43 - 8 \times 1 - 6 \times 2 - 25 \times 1 - 27 \times 2 = 0
  in he form of
                          =) 43 - 2(4\pi, +3\pi_2) - 25\lambda, -27\lambda_2 = 0
   (5)
                          =) 43-2(15)-25/1-27/2=0 -- From (3)
                           = 25\lambda_1 + 27\lambda_2 = 13 --- (5)
NOW 3x(1) + 5x(2) => 12-621-1221-92=0
+45-10x2-1521-252=0
  ( in the form
                          57-6x,-10×2-272,-342=0
        03 (41)
                       => 57-2 (3×1+5×2)-27×1-34×2=0
                      => 57-2(14)-27h1-34h2 =0
                       =) 27\lambda_1 + 34\lambda_2 = 29 — (6) Prof. Anushri Tambe
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$$27 \times (5) - 25 \times (6) \Rightarrow -121 \lambda_{2} = -374 \rightarrow \lambda_{2} = \frac{374}{121}$$

$$325 \lambda_{1} = 13 - 27 \left(\frac{374}{121}\right) = -\frac{6521}{121} \Rightarrow \lambda_{1} = -\frac{341}{121}$$
From (1), $23_{1} = 4 - 4\lambda_{1} - 3\lambda_{2} = 4 - 4\left(\frac{-341}{121}\right) - 3\left(\frac{374}{121}\right) = \frac{726}{121} = 6$

$$\frac{37}{12} \times \frac{3}{12} = \frac{3}{12$$

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:. L(x_1, x_2, x_3, \lambda_1, \lambda_2) = x_1^2 + x_2^2 + x_3^2 - \lambda_1(x_1 + x_2 + x_3 - 13) -
                                                    12 (3x1+ x2+ x3-27)
 \frac{\partial L}{\partial x_1} = 2 x_1 - \lambda_1 - 3 \lambda_2 \qquad , \quad \frac{\partial L}{\partial x_2} = -2 x_2 - \lambda_1 - \lambda_2
 3L = 2x3-1,-12, 3L = -(x,+x2+x3-13), 3L = -(3x1+x2+x3-27)
 Solving, \frac{\partial L}{\partial x_1} = 0, \frac{\partial L}{\partial x_2} = 0, \frac{\partial L}{\partial x_3} = 0, \frac{\partial L}{\partial \lambda_1} = 0, \frac{\partial L}{\partial \lambda_2} = 0
 = 2 \times 1 - \lambda_1 - 3 \lambda_2 = 0  (1)
                                                     2 + \lambda_1 - \lambda_2 = 0 — (2)
       -2×3-11-12=0 - (3) 21+2+2=13 - (4) 3×1+2+2=27
(Now to find values of x,, x, x, x, x, \, \, \, \)
(1)+(2)+(3) \Rightarrow 2x_1+2x_2+2x_3-\lambda_1-\lambda_1-\lambda_1-3\lambda_2-\lambda_2-\lambda_2=0
                 =) 2(31 + 31 + 31) - 3\lambda_1 - 5\lambda_2 = 0
                  =) 2(13) = 3\lambda_1 - 5\lambda_2 = 0 -- from (4)
                 \Rightarrow 3\lambda_1 + 5\lambda_2 = 26 — (6)
 3x(1)+(2)+(3)=) 6x_1+2x_2+2x_3-3x_1-x_1-x_1-9x_2-x_2-x_2=0
                    => 2 (3×1+×2+×3) - 5×1-11×2=0
                     = 2 (27) - 5\lambda, -11\lambda_2 = 0
                     \Rightarrow 5\lambda_1 + 11\lambda_2 = 54 --- (7)
 5 \times (6) - 3 \times (7) = ) 15 \lambda_1 + 25 \lambda_2 = 130
                             \frac{15\lambda_1 + 55\lambda_2 = 162}{-8\lambda_2 = -32} \Rightarrow \lambda_2 = 4
 (6) =) 3\lambda_1 = 26 - 5(4) = 6 \Rightarrow \overline{\lambda_1 = 2}
 (1) =) 2x_1 = \lambda_1 + 3\lambda_2 = 2 + 12 = 14 =) [21, =7]
  (2) =) 2 \times 2 = \lambda_1 + \lambda_2 = 2 + 4 = 6 \Rightarrow [3(2 - 3)]
  (3) =) 2 \times 3 = \lambda_1 + \lambda_2 = 2 + 4 = 6 \Rightarrow \sqrt{3} = 3
                                                                         Prof. Anushri Tambe
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Now,
$$\frac{\partial h_1}{\partial x_1} = 1$$
, $\frac{\partial h_1}{\partial x_2} = 1$, $\frac{\partial h_2}{\partial x_3} = 1$

$$\frac{\partial h_2}{\partial x_1} = 3$$
, $\frac{\partial h_2}{\partial x_2} = 1$, $\frac{\partial h_2}{\partial x_3} = 1$

$$\frac{\partial L}{\partial x_1^2} = 2$$
, $\frac{\partial^2 L}{\partial x_1 \partial x_2} = 0$, $\frac{\partial^2 L}{\partial x_1 \partial x_3} = 0$

$$\frac{\partial^2 L}{\partial x_2^2} = 2$$
, $\frac{\partial^2 L}{\partial x_3 \partial x_1} = 0$, $\frac{\partial^2 L}{\partial x_2 \partial x_3} = 0$

$$\frac{\partial^2 L}{\partial x_1^2} = 2$$
, $\frac{\partial^2 L}{\partial x_3 \partial x_1} = 0$, $\frac{\partial^2 L}{\partial x_2 \partial x_3} = 0$

$$\therefore P = \begin{bmatrix} \frac{\partial h_1}{\partial x_1} & \frac{\partial h_1}{\partial x_2} & \frac{\partial h_2}{\partial x_3} \\ \frac{\partial h_2}{\partial x_1} & \frac{\partial h_2}{\partial x_2} & \frac{\partial h_2}{\partial x_3} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 3 & 1 & 1 \end{bmatrix}$$
, $P' = \begin{bmatrix} 1 & 3 \\ 1 & 1 \\ 1 & 1 \end{bmatrix}$

$$Q = \begin{bmatrix} \frac{\partial^2 L}{\partial x_1^2} & \frac{\partial^2 L}{\partial x_1 \partial x_2} & \frac{\partial^2 L}{\partial x_1 \partial x_3} \\ \frac{\partial^2 L}{\partial x_2 \partial x_1} & \frac{\partial^2 L}{\partial x_2 \partial x_2} & \frac{\partial^2 L}{\partial x_3 \partial x_3} \end{bmatrix} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

$$\therefore H_0^S = \begin{bmatrix} 0 & 1 & P \\ P' & Q \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 3 & 1 & 1 \\ 1 & 1 & 0 & 2 & 0 \\ 1 & 1 & 0 & 2 & 0 \\ 1 & 1 & 0 & 0 & 2 \end{bmatrix}$$

Here n= 3 (number of variables)
m=2 (number of constraints)

Use Laplace method to find determinant of Ho

as HoB is positive, Xo = (7,3,3) is a minima

$$4 Z_{min} = (7)^2 + (3)^2 + (3)^2 = 67$$
 $Z_{min} = 67$

Practice Problems: Using Lagrangian multiplier solve the following NLPP.

1) Maximise
$$z = 6\pi_1 + 8\pi_2 - 34_1^2 - 34_2^2$$

Subject to $4\pi_1 + 3\pi_2 = 16$
 $3\pi_1 + 5\pi_2 = 15$
 $3\pi_1, 3\pi_2 = 16$

1) Optimise
$$Z = 23^2 + 38^2 + 38^2$$

Subject to $31 + 32 + 239 = 13$
 $2x_1 + 32 + 33 = 10$
 $31, 32, 33 = 3$

(Laplace Method: consider Ho from eg. (2). Multiply the determinant 13 1 in which these elements lie i.e. by 1:30 . The sign is determined by (-1)3+4+1 as the elements of 13 1 lie in the third of forth column, similarly take the product of all other determinants with proper sign] Prof. Anushri Tambe