

Multivariable problem with equality and inequality constraints

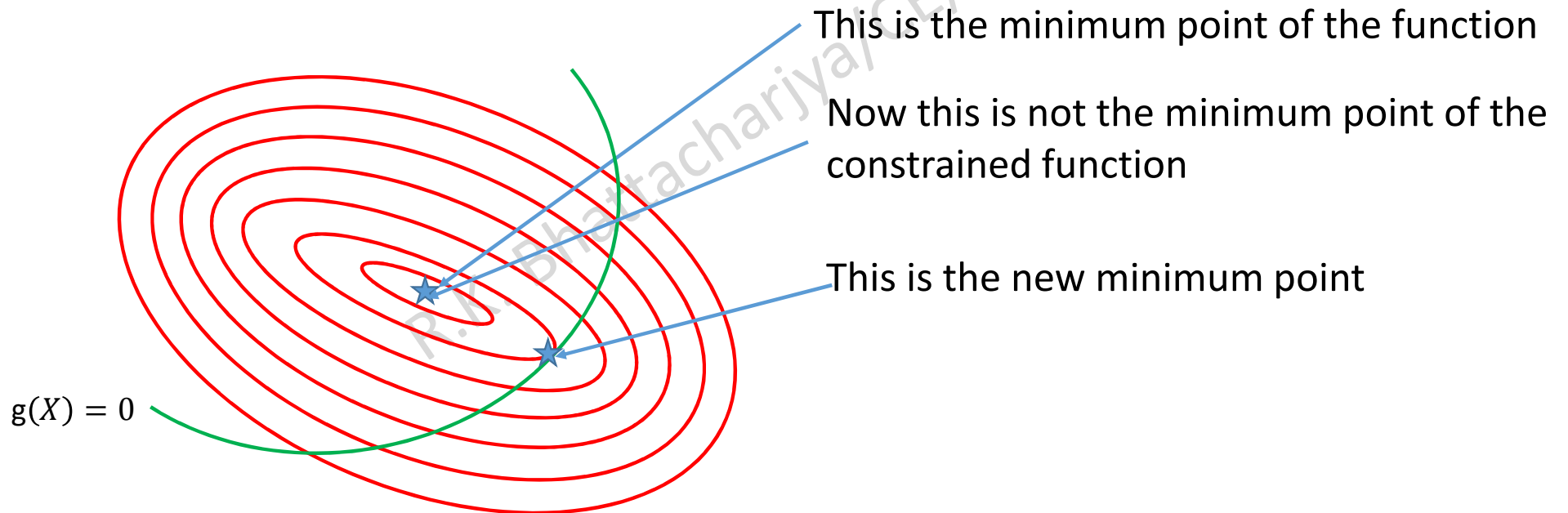
Rajib Bhattacharjya
Department of Civil Engineering
IIT Guwahati
Email: rkbc@iitg.ernet.in



General formulation

Min/Max $f(X)$ Where $X = [x_1, x_2, x_3, \dots, x_n]^T$

Subject to $g_j(X) = 0$ $j = 1, 2, 3, \dots, m$





Consider a two variable problem

Min/Max $f(x_1, x_2)$

Subject to $g(x_1, x_2) = 0$

Take total derivative of the function at (x_1, x_2)

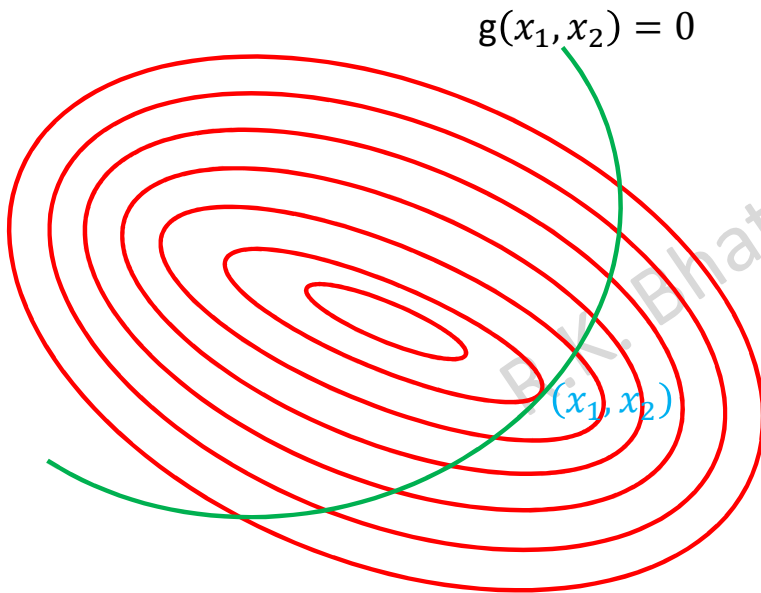
$$df = \frac{\partial f}{\partial x_1} dx_1 + \frac{\partial f}{\partial x_2} dx_2 = 0$$

If (x_1, x_2) is the solution of the constrained problem, then

$$g(x_1, x_2) = 0$$

Now any variation dx_1 and dx_2 is admissible only when

$$g(x_1 + dx_1, x_2 + dx_2) = 0$$





Consider a two variable problem

Min/Max $f(x_1, x_2)$

$$g(x_1 + dx_1, x_2 + dx_2) = 0$$

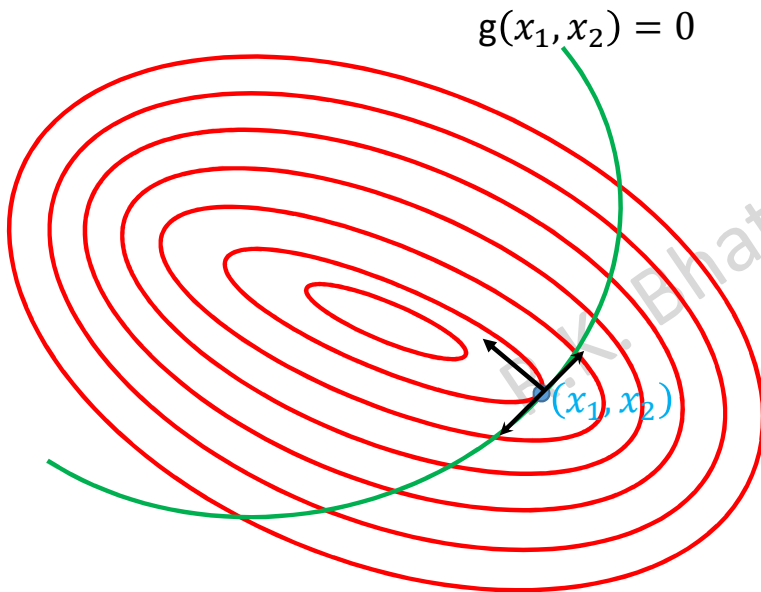
Subject to $g(x_1, x_2) = 0$

This can be expanded as

$$g(x_1 + dx_1, x_2 + dx_2) = \cancel{g(x_1, x_2)} + \frac{\partial g(x_1, x_2)}{\partial x_1} dx_1 + \frac{\partial g(x_1, x_2)}{\partial x_2} dx_2 = 0$$

$$dg = \frac{\partial g}{\partial x_1} dx_1 + \frac{\partial g}{\partial x_2} dx_2 = 0$$

$$dx_2 = -\frac{\frac{\partial g}{\partial x_1}}{\frac{\partial g}{\partial x_2}} dx_1$$





Consider a two variable problem

Min/Max $f(x_1, x_2)$

Subject to $g(x_1, x_2) = 0$

$$dx_2 = -\frac{\frac{\partial g}{\partial x_1}}{\frac{\partial g}{\partial x_2}} dx_1$$

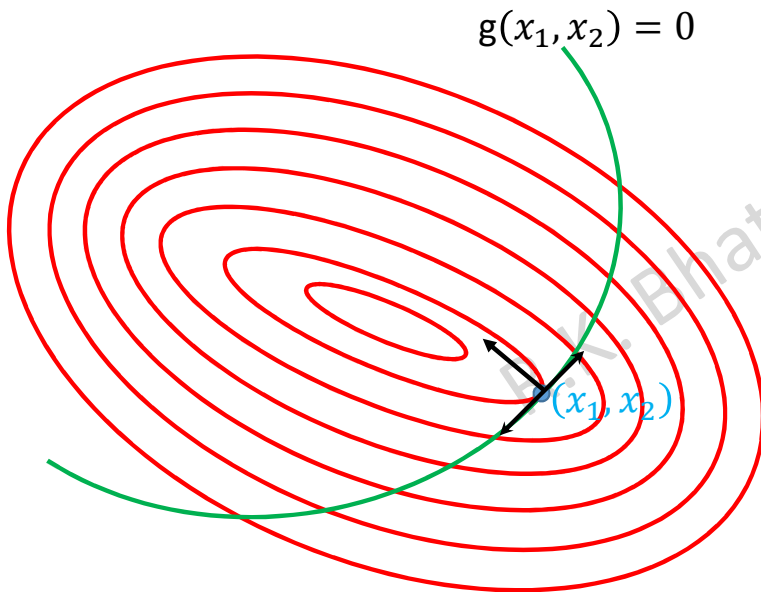
Putting in $df = \frac{\partial f}{\partial x_1} dx_1 + \frac{\partial f}{\partial x_2} dx_2 = 0$

$$df = \frac{\partial f}{\partial x_1} dx_1 - \frac{\partial f}{\partial x_2} \frac{\frac{\partial g}{\partial x_1}}{\frac{\partial g}{\partial x_2}} dx_1 = 0$$

$$\left(\frac{\partial f}{\partial x_1} \frac{\partial g}{\partial x_2} - \frac{\partial f}{\partial x_2} \frac{\partial g}{\partial x_1} \right) dx_1 = 0$$

$$\left(\frac{\partial f}{\partial x_1} \frac{\partial g}{\partial x_2} - \frac{\partial f}{\partial x_2} \frac{\partial g}{\partial x_1} \right) = 0$$

This is the necessary condition for optimality for optimization problem with equality constraints





Lagrange Multipliers

Min/Max $f(x_1, x_2)$

Subject to $g(x_1, x_2) = 0$

We have already obtained the condition that

$$\frac{\partial f}{\partial x_1} - \frac{\partial f}{\partial x_2} \frac{\frac{\partial g}{\partial x_1}}{\frac{\partial g}{\partial x_2}} = 0 \quad \rightarrow \quad \frac{\partial f}{\partial x_1} - \left(\frac{\frac{\partial f}{\partial x_2}}{\frac{\partial g}{\partial x_2}} \right) \frac{\partial g}{\partial x_1} = 0$$

By defining $\lambda = -\frac{\frac{\partial f}{\partial x_2}}{\frac{\partial g}{\partial x_2}}$

We have

$$\frac{\partial f}{\partial x_1} + \lambda \frac{\partial g}{\partial x_1} = 0$$

We can also write

$$\frac{\partial f}{\partial x_2} + \lambda \frac{\partial g}{\partial x_2} = 0$$

Also put

$$g(x_1, x_2) = 0$$

Necessary conditions for optimality



Lagrange Multipliers

Let us define

$$L(x_1, x_2, \lambda) = f(x_1, x_2) + \lambda g(x_1, x_2)$$

By applying necessary condition of optimality, we can obtain

$$\frac{\partial L}{\partial x_1} = \frac{\partial f}{\partial x_1} + \lambda \frac{\partial g}{\partial x_1} = 0$$

$$\frac{\partial L}{\partial x_2} = \frac{\partial f}{\partial x_2} + \lambda \frac{\partial g}{\partial x_2} = 0$$

$$\frac{\partial L}{\partial \lambda} = g(x_1, x_2) = 0$$

Necessary conditions for optimality



Lagrange Multipliers

Sufficient condition for optimality of the Lagrange function can be written as

$$L(x_1, x_2, \lambda) = f(x_1, x_2) + \lambda g(x_1, x_2)$$

$$H = \begin{bmatrix} \frac{\partial^2 L}{\partial x_1 \partial x_1} & \frac{\partial^2 L}{\partial x_1 \partial x_2} & \frac{\partial^2 L}{\partial x_1 \partial \lambda} \\ \frac{\partial^2 L}{\partial x_2 \partial x_1} & \frac{\partial^2 L}{\partial x_2 \partial x_2} & \frac{\partial^2 L}{\partial x_2 \partial \lambda} \\ \frac{\partial^2 L}{\partial \lambda \partial x_1} & \frac{\partial^2 L}{\partial \lambda \partial x_2} & \frac{\partial^2 L}{\partial \lambda \partial \lambda} \end{bmatrix} \rightarrow H = \begin{bmatrix} \frac{\partial^2 L}{\partial x_1 \partial x_1} & \frac{\partial^2 L}{\partial x_1 \partial x_2} & \frac{\partial g}{\partial x_1} \\ \frac{\partial^2 L}{\partial x_2 \partial x_1} & \frac{\partial^2 L}{\partial x_2 \partial x_2} & \frac{\partial g}{\partial x_2} \\ \frac{\partial g}{\partial x_1} & \frac{\partial g}{\partial x_2} & 0 \end{bmatrix}$$

If H is positive definite, the optimal solution is a minimum point

If H is negative definite, the optimal solution is a maximum point

Else it is neither minima nor maxima



Lagrange Multipliers

Necessary conditions for general problem

Min/Max $f(X)$ Where $X = [x_1, x_2, x_3, \dots, x_n]^T$

Subject to $g_j(X) = 0$ $j = 1, 2, 3, \dots, m$

$$L(x_1, x_2, \dots, x_n, \lambda_1, \lambda_2, \lambda_3, \dots, \lambda_m) = f(X) + \lambda_1 g_1(X) + \lambda_2 g_2(X), \dots, \lambda_m g_m(X)$$

Necessary conditions

$$\frac{\partial L}{\partial x_i} = \frac{\partial f}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j}{\partial x_i} = 0$$

$$\frac{\partial L}{\partial \lambda_j} = g_j(X) = 0$$



Lagrange Multipliers

Sufficient condition for general problem

The hessian matrix is

Where,

$$H = \begin{bmatrix} L_{11} & L_{12} & L_{13} & \dots & L_{1n} & g_{11} & g_{21} & \dots & g_{m1} \\ L_{21} & L_{22} & L_{23} & \dots & L_{2n} & g_{12} & g_{22} & \dots & g_{m2} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ L_{n1} & L_{n2} & L_{n3} & \dots & L_{nn} & g_{1n} & g_{2n} & \dots & g_{mn} \\ g_{11} & g_{12} & g_{13} & \dots & g_{1n} & 0 & 0 & \dots & 0 \\ g_{21} & g_{22} & g_{23} & \vdots & g_{2n} & \vdots & \vdots & \vdots & \vdots \\ g_{31} & g_{32} & g_{33} & \dots & g_{3n} & 0 & \dots & \dots & \dots \\ \vdots & \vdots & \vdots & \dots & \vdots & \vdots & \dots & \dots & \dots \\ g_{m1} & g_{m2} & g_{m3} & \dots & g_{2n} & 0 & \dots & \dots & 0 \end{bmatrix}$$

$$L_{ij} = \frac{\partial^2 L}{\partial x_i \partial x_j}$$

$$g_{ij} = \frac{\partial g_i}{\partial x_j}$$



Lagrange Multipliers

Min/Max $f(X)$ Where $X = [x_1, x_2, x_3, \dots, x_n]^T$

Subject to $g(X) = b$ Or, $b - g(X) = 0$

Further $db - dg = 0$

$$db = dg = \sum_{i=1}^n \frac{\partial g}{\partial x_i} dx_i$$

Applying necessary conditions

$$\frac{\partial f}{\partial x_i} - \lambda \frac{\partial g}{\partial x_i} = 0 \quad \text{Where, } i = 1, 2, 3, \dots, n$$

$$g = 0$$

$$\frac{\partial g}{\partial x_i} = \frac{\partial f}{\lambda}$$

There may be three conditions

$$\lambda^* > 0$$

$$\lambda^* < 0$$

$$\lambda^* = 0$$

$$db = \sum_{i=1}^n \frac{1}{\lambda} \frac{\partial f}{\partial x_i} dx_i$$

$$db = \frac{df}{\lambda}$$

$$\lambda = \frac{df}{db}$$

$$df = \lambda db$$



Multivariable problem with inequality constraints

Minimize $f(X)$ Where $X = [x_1, x_2, x_3, \dots, x_n]^T$

Subject to $g_j(X) \leq 0$ $j = 1, 2, 3, \dots, m$

We can write $g_j(X) + y_j^2 = 0$

Thus the problem can be written as

Minimize $f(X)$

Subject to $G_j(X, Y) = g_j(X) + y_j^2 = 0$ $j = 1, 2, 3, \dots, m$

Where $Y = [y_1, y_2, y_3, \dots, y_m]^T$



Multivariable problem with inequality constraints

Minimize $f(X)$ Where $X = [x_1, x_2, x_3, \dots, x_n]^T$

Subject to $G_j(X, Y) = g_j(X) + y_j^2 = 0 \quad j = 1, 2, 3, \dots, m$

The Lagrange function can be written as

$$L(X, Y, \lambda) = f(X) + \sum_{j=1}^m \lambda_j G_j(X, Y)$$

The necessary conditions of optimality can be written as

$$\frac{\partial L(X, Y, \lambda)}{\partial x_i} = \frac{\partial f(X)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$

$$\frac{\partial L(X, Y, \lambda)}{\partial \lambda_j} = G_j(X, Y) = g_j(X) + y_j^2 = 0 \quad j = 1, 2, 3, \dots, m$$

$$\frac{\partial L(X, Y, \lambda)}{\partial y_j} = 2\lambda_j y_j = 0 \quad j = 1, 2, 3, \dots, m$$



Multivariable problem with inequality constraints

From equation $\frac{\partial L(X,Y,\lambda)}{\partial y_j} = 2\lambda_j y_j = 0$

Either $\lambda_j = 0$ Or, $y_j = 0$

If $\lambda_j = 0$, the constraint is not active, hence can be ignored

If $y_j = 0$, the constraint is active, hence have to consider

Now, consider all the active constraints,

Say set J_1 is the active constraints

And set J_2 is the active constraints

The optimality condition can be written as

$$\frac{\partial f(X)}{\partial x_i} + \sum_{j \in J_1} \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$

$$g_j(X) = 0 \quad j \in J_1$$

$$g_j(X) + y_j^2 = 0 \quad j \in J_2$$



Multivariable problem with inequality constraints

$$-\frac{\partial f}{\partial x_i} = \lambda_1 \frac{\partial g_1}{\partial x_i} + \lambda_2 \frac{\partial g_2}{\partial x_i} + \lambda_3 \frac{\partial g_3}{\partial x_i} + \dots + \lambda_p \frac{\partial g_p}{\partial x_i} \quad i = 1, 2, 3, \dots, n$$

$$-\nabla f = \lambda_1 \nabla g_1 + \lambda_2 \nabla g_2 + \lambda_3 \nabla g_3 + \dots + \lambda_m \nabla g_m$$

This indicates that negative of the gradient of the objective function can be expressed as a linear combination of the gradients of the active constraints at optimal point.

$$\nabla f = \begin{Bmatrix} \partial f / \partial x_1 \\ \partial f / \partial x_2 \\ \vdots \\ \partial f / \partial x_n \end{Bmatrix}$$

$$\nabla g_j = \begin{Bmatrix} \partial g_j / \partial x_1 \\ \partial g_j / \partial x_2 \\ \vdots \\ \partial g_j / \partial x_n \end{Bmatrix}$$

$$-\nabla f = \lambda_1 \nabla g_1 + \lambda_2 \nabla g_2$$

Let S be a feasible direction, then we can write

$$-S^T \nabla f = \lambda_1 S^T \nabla g_1 + \lambda_2 S^T \nabla g_2$$

Since S is a feasible direction

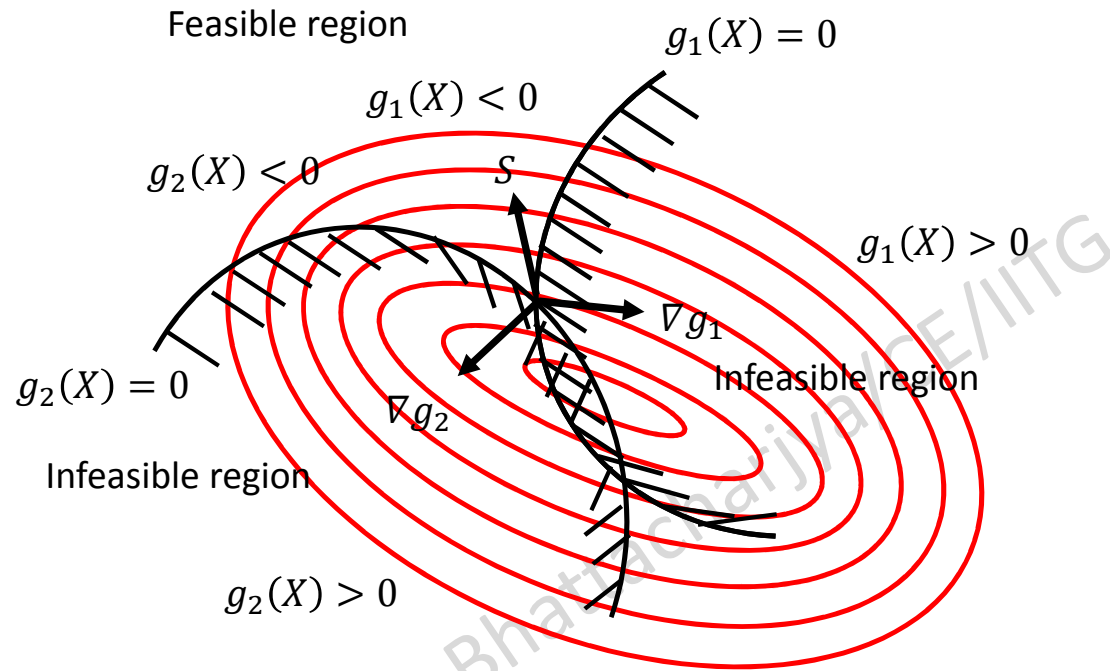
$$S^T \nabla g_1 < 0 \quad \text{and} \quad S^T \nabla g_2 < 0$$

If $\lambda_1, \lambda_2 > 0$

Then the term $S^T \nabla f$ is +ve

This indicates that S is a direction of increasing function value

Thus we can conclude that if $\lambda_1, \lambda_2 > 0$, we will not get any better solution than the current solution





Multivariable problem with inequality constraints

The necessary conditions to be satisfied at constrained minimum points X^* are

$$\frac{\partial f(X)}{\partial x_i} + \sum_{j \in J_1} \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$
$$\lambda_j \geq 0 \quad j \in J_1$$

These conditions are called **Kuhn-Tucker conditions**, the necessary conditions to be satisfied at a relative minimum of $f(X)$.

These conditions are in general not sufficient to ensure a relative minimum, However, in case of a convex problem, these conditions are the necessary and sufficient conditions for global minimum.



Multivariable problem with inequality constraints

If the set of active constraints are not known, the Kuhn-Tucker conditions can be stated as

$$\frac{\partial f(X)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$

$$\left. \begin{array}{l} \lambda_j g_j = 0 \\ g_j \leq 0 \\ \lambda_j \geq 0 \end{array} \right\} j = 1, 2, 3, \dots, m$$



Multivariable problem with equality and inequality constraints

For the problem

$$\begin{array}{lll} \text{Minimize} & f(X) & \text{Where } X = [x_1, x_2, x_3, \dots, x_n]^T \\ \text{Subject to} & g_j(X) = 0 & j = 1, 2, 3, \dots, m \\ & h_k(X) = 0 & k = 1, 2, 3, \dots, p \end{array}$$

The Kuhn-Tucker conditions can be written as

$$\begin{aligned} \frac{\partial f(X)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j(X)}{\partial x_i} + \sum_{k=1}^p \beta_k \frac{\partial h_k(X)}{\partial x_i} &= 0 & i = 1, 2, 3, \dots, n \\ \lambda_j g_j &= 0 & j = 1, 2, 3, \dots, m \\ g_j &\leq 0 & j = 1, 2, 3, \dots, m \\ h_k &= 0 & k = 1, 2, 3, \dots, p \\ \lambda_j &\geq 0 & j = 1, 2, 3, \dots, m \end{aligned}$$



Thanks for your attention

R.K. Bhattacharjya/CE/IITG