

Theorem:-

If $f(x)$ is differentiable function over the convex set $X \subseteq \mathbb{R}^n$ then $f(x)$ is convex over X iff

$$f(x^2) \geq f(x^1) + \nabla^T f(x^1) (x^2 - x^1)$$

for all $x^1, x^2 \in X$.

Proof:- Given - $f(x)$ is convex over X

For all $x^1, x^2 \in X$ and for all $\lambda \in (0, 1]$

$$f(\lambda x^2 + (1-\lambda)x^1) \leq \lambda f(x^2) + (1-\lambda)f(x^1)$$

$$\Rightarrow \frac{f[x^1 + \lambda(x^2 - x^1)] - f(x^1)}{\lambda} \leq f(x^2) - f(x^1)$$

Taking limit $\lambda \rightarrow 0$, it follows that

$$\left. \frac{df(x^1)}{d\lambda} \right|_{x^2 - x^1} \leq f(x^2) - f(x^1) \quad \text{--- (1)}$$

The directional derivative on LHS

$$\left. \frac{df(x^1)}{d\lambda} \right|_{x^2 - x^1} = \nabla^T f(x) (x^2 - x^1) \quad \text{--- (2)}$$

From (1) & (2) we have

$$\begin{aligned} \nabla^T f(x) (x^2 - x^1) &\leq f(x^2) - f(x^1) \\ \Rightarrow f(x^2) &\geq f(x^1) + \nabla^T f(x) (x^2 - x^1) \end{aligned}$$



Given :-

$$f(x^2) \geq f(x') + \nabla^T f(x') (x^2 - x')$$

$$\text{for } x = \lambda x^2 + (1-\lambda)x' \in X, \lambda \in (0,1]$$

$$\begin{aligned} f(x^2) &\geq f(x) + \nabla^T f(x) (x^2 - x) \quad] \times \lambda \\ f(x') &\geq f(x) + \nabla^T f(x) (x' - x) \quad] \times (1-\lambda) \end{aligned}$$

$$\lambda f(x^2) + (1-\lambda)f(x') \geq f(x) + \nabla^T f(x) [\lambda(x^2 - x) + (1-\lambda)(x' - x)]$$

$= 0$

$$\left(\because \lambda(x^2 - x) + (1-\lambda)(x' - x) = 0 \right)$$

$$\Rightarrow f(x) \leq \lambda f(x^2) + (1-\lambda)f(x'), \text{ Hence } f' \text{ is convex. } \square$$

Jacobian matrix

$$\left[\frac{\partial h}{\partial x} \right]_{ij} = \frac{\partial h_j}{\partial x_i} \quad \text{z.e.;} \quad \left[\frac{\partial h}{\partial x} \right] = [\nabla h_1, \nabla h_2, \dots, \nabla h_r]$$

$$\nabla h_1 = \begin{pmatrix} \partial h_1 / \partial x_1 \\ \partial h_1 / \partial x_2 \\ \vdots \\ \partial h_1 / \partial x_n \end{pmatrix}$$

$$h = [h_1, h_2, \dots, h_r] \\ \text{or } (h_1, h_2, \dots, h_r)$$

Problem:- (P)

consider equality constrained problem

$$\min f(x)$$

$$\text{such that } h_j(x) = 0, \quad j = 1, 2, \dots, r < n$$

Ex 1760 Lagrange transformed this constrained problem to an unconstrained problem

By using Lagrange Multiplier λ_j

To formulate Lagrange funⁿ for $j = 1, 2, \dots, r$

$$L(x, \lambda) = f(x) + \sum_{j=1}^r \lambda_j h_j(x) = f(x) + \lambda^T h(x)$$

The equality constrained problem (P) for f & $h_j \in C^1$ and assume that the

Jacobian matrix $\left[\frac{\partial h(x^*)}{\partial x} \right]$ is of rank 'r.'

Theorem
The necessary condⁿ for interior local minimum x^* of equality constrained problem (P) is.

x^* must coincide with the stationary point (x^*, λ^*) of Lagrange function L

O.R: there exist a λ^* such that

$$\frac{\partial L}{\partial x_i}(x^*, \lambda^*) = 0, \quad i = 1, 2, \dots, n \quad \& \quad \frac{\partial L}{\partial \lambda_j}(x^*, \lambda^*) = 0$$

for $j = 1, 2, \dots, r$

Proof: Given: (1) f & $h_j \in C^1$
 it follows an interior
 local min at $x = x^*$

(2) $\left[\frac{\partial h(x^*)}{\partial x} \right]$ is of rank r .

To prove: - for $x = x^*$, $\exists \lambda^*$ such that

$$\frac{\partial L(x^*, \lambda^*)}{\partial x_i} = 0, \quad i = 1, 2, \dots, n;$$

$$\frac{\partial L(x^*, \lambda^*)}{\partial \lambda_j} = 0, \quad j = 1, 2, \dots, r$$

As $x = x^*$ is local minimum at $x = x^*$

$$df = \nabla^T f(x^*) dx = 0 \quad \text{--- (1)}$$

($\because df \geq 0$ for dx & for $-dx$)

$$dh_j = \nabla^T h_j(x^*) dx = 0, \quad j = 1, 2, \dots, r. \quad (2)$$

consider the Lagrange funⁿ

$$L(x, \lambda) = f(x) + \sum_{j=1}^r \lambda_j h_j(x)$$

The differential of L is given by

$$dL = df + \sum_{j=1}^r \lambda_j dh_j$$

$$\Rightarrow dL = 0$$

$$\Rightarrow \frac{\partial L}{\partial x_1} dx_1 + \frac{\partial L}{\partial x_2} dx_2 + \dots + \frac{\partial L}{\partial x_n} dx_n = 0 \quad (3)$$

for all dx such that

$$h(x) = 0 \quad \text{i.e.;} \quad h(x^* + dx) = 0$$

(as the constraints
satisfied)

Choose Lagrange multipliers $\lambda_j, j=1, 2, \dots, r$ such that at x^*

$$\frac{\partial L}{\partial x_j}(x^*, \lambda) = \frac{\partial f}{\partial x_j}(x^*) + \left[\frac{\partial h}{\partial x_j}(x^*) \right]^T \lambda = 0 \quad (4)$$

$j=1, 2, \dots, r.$

The solution of this system provides the vector λ^* . Here the 'r' variables, $x_j, j=1, 2, \dots, r$ may be any appropriate set of 'r' variables from the set $x_i, i=1, 2, \dots, n$.

A unique solution for λ^* exists as it is assumed that $\left[\frac{\partial h}{\partial x}(x^*) \right]$ is of rank 'r'.

Then the equation (3) reduces to

$$dL = \frac{\partial L}{\partial x_{r+1}} (x^*, \lambda^*) dx_{r+1} + \dots + \frac{\partial L}{\partial x_n} (x^*, \lambda^*) dx_n = 0 \quad (5)$$

Again consider the constraints $h_j(x) = 0$, $j = 1, 2, \dots, r$.
 If these equations are considered as a system of r -equations in the unknowns x_1, x_2, \dots, x_r

these dependent unknowns can be solved for in terms of $x_{r+1}, x_{r+2}, \dots, x_n$.

Hence the latter $(n-r)$ variables are the independent variables.

For any choice of these independent variables the other independent variables x_1, x_2, \dots, x_r are determined by solving $h(x) = [h_1(x), \dots, h_r(x)]^T = [0, 0, \dots, 0]^T = \vec{0}$

In particular x_{r+1} to x_n may be varied one-by-one at x^* and it follows from (5) that

$$\frac{\partial L}{\partial x_j}(x^*, \lambda^*) = 0, \quad j = r+1, r+2, \dots, n$$

and, together with (4) & constraints $h(x) = 0$.

it follows that the necessary conditions for an inferior local minimum can be written as

$$\begin{cases} \frac{\partial L}{\partial x_i}(x^*, \lambda^*) = 0 & i = 1, 2, \dots, n \\ \frac{\partial L}{\partial \lambda_j}(x^*, \lambda^*) = 0 & j = 1, 2, \dots, r \end{cases}$$

$$\nabla_x L(x^*, \lambda^*) = 0 \quad \& \quad \nabla_\lambda L(x^*, \lambda^*) = 0$$

