

# Method of Lagrangian Multipliers.

## Inequality constraints.

Prob<sup>o</sup>: Minimize  $f(x)$

subject to  $g_j(x) \leq 0 ; j=1,2,\dots,m$

where  $x = (x_1, \dots, x_n)^T \in \mathbb{R}^n \leftarrow$  decision vector.

$\swarrow \searrow$   
decision variables.

After introducing 'm' - positive slack variables

we get,  $g_j(x) + \delta_j^2 = 0 ; j=1,2,\dots,m$

Consider the Lagrangian  $f_L$ ,

$$L(x_1, \dots, x_n, \delta_1, \dots, \delta_m, \lambda_1, \dots, \lambda_m) =$$

$$f(x) + \sum_{j=1}^m \lambda_j (g_j(x) + \delta_j^2) \text{ where}$$

$\lambda_1, \lambda_2, \dots, \lambda_m$  are Lagrange multipliers.

Necessary Conditions for its local/relative minima

Karush-Kuhn-Tucker Conditions  $\uparrow$

$$\frac{\partial L}{\partial x_i} = 0 \Rightarrow \frac{\partial f}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j(x)}{\partial x_i} = 0 ; i=1,2,\dots,n \quad (\text{Optimality})$$
$$\frac{\partial L}{\partial \lambda_j} = 0 \Rightarrow g_j(x) + \delta_j^2 = 0 ; j=1,2,\dots,m$$
$$\Rightarrow g_j(x) \leq 0 ; j=1,\dots,m \quad (\text{Feasibility})$$
$$\frac{\partial L}{\partial \delta_j} = 0 \Rightarrow 2\lambda_j \delta_j = 0 \Rightarrow \lambda_j \delta_j = 0 ; j=1,2,\dots,m \quad (\text{Complementary slackness})$$
$$\lambda_j \geq 0 \quad (\text{Non-negativity})$$

At a pt.  $x^*$  (Local minima)

Complementary slackness condition

$$\Rightarrow \lambda_j s_j = 0 \begin{cases} \lambda_j = 0, s_j \neq 0 \text{ (Case-I)} \\ \lambda_j \neq 0, s_j = 0 \text{ (Case-II)} \end{cases}$$

(Case I)

$\lambda_j = 0, s_j \neq 0$  & feasibility condition;

$$\odot \quad g_j(x^*) + s_j^2 = 0 \Rightarrow \underbrace{g_j(x^*)}_{\text{inactive constraints}} \leq 0$$

(Case II)

$\lambda_j \neq 0, s_j = 0$  & feasibility condition;

$$g_j(x^*) + s_j^2 = 0 \Rightarrow \underbrace{g_j(x^*)}_{\text{active constraints}} = 0$$

For active constraints  $\lambda_j > 0$  & for inactive constraints  $\lambda_j = 0$ . (For a minimization problem)  
Opposite will be for a maximization problem.

~~Conclusion~~

So KKT conditions are the necessary conditions.

## Sufficient conditions (Minimization)

→ The KKT conditions are sufficient conditions if  $f(x)$  is convex, and the feasible space is convex.

In general,

$$L = f(x) + \sum_{j=1}^m \lambda_j (g_j(x) + \delta_j^2)$$

$\lambda_j \geq 0$  &  $x^*$  is local minimum  
if  $\nabla^2 L|_{x^*}$  is positive definite.

### Exercise:

1). Minimize  $f(x) = x_1^2 - 4x_1 + x_2^2 - 6x_2$   
subject to  $x_1 + x_2 \leq 3$   
 $-2x_1 + x_2 \leq 2$ .

2). Maximize  $f(x) = 3x + 4y$   
subject to  $x^2 + y^2 \leq 4$   
 $x \geq 1$ .

3). Minimize  $f(x) = x_1^2 + x_2$   
subject to  $x_1^2 + x_2^2 \leq 9$   
 $x_1 + x_2 \leq 1$ .

4). Minimize  $f(x)$   
subject to  $x_1 + x_2 = 90$

$$f(x) = 6x_1^2 + 12x_2^2.$$

show that the pt.  $(1, 0)$  doesn't satisfy the KKT conditions.

- Use Newton's method to minimize the Powell fn. :

$$f(x_1, x_2, x_3, x_4) = (x_1 + 10x_2)^2 + 5(x_3 - x_4)^2 + (x_2 - 2x_3)^4 + 10(x_1 - x_4)^4$$

Use the starting point  $x^{(0)} = (3, -1, 0, 1)^T$ .  
 Compute for three iterations. You can take help of softwares for computing matrix inverses & matrix multiplications.

- Find the maximum value of  $f(x) = 2 \sin x - \frac{x^2}{10}$  with an initial guess of  $x^{(0)} = 2.5$ . (Use Newton's method)

- How many basic sol<sup>ns</sup> are there in the following linearly independent set of eq<sup>ns</sup>? Find all of them.

$$2x_1 - x_2 + 3x_3 + x_4 = 6 \quad ; \quad x_1, x_2, x_3, x_4 \geq 0.$$

$$4x_1 - 2x_2 - x_3 + 2x_4 = 10$$

- Solve graphically the LPP

$$\text{Min } z = -2x_1 + x_2$$

$$\text{subject to } x_1 + x_2 \geq 6, 3x_1 + 2x_2 \geq 16, x_2 \leq 9; x_1, x_2 \geq 0.$$

- Solve graphically the LPP

$$\text{Max } z = 4x_1 + 7x_2$$

$$\text{subject to } 12x_1 + 7x_2 \leq 42$$

$$5x_1 + 4x_2 \leq 20$$

$$2x_1 + 3x_2 \geq 6 \quad ; \quad x_1, x_2 \geq 0.$$