

**Optimization**

**Fourth Class**

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# Chapter One

## Basic Concepts

### Lecture 5

## 12: Optimality Conditions for Unconstrained Optimization

### Definition (12): (Local Minimizer)

Let  $f: R^n \rightarrow R$  be a given function and  $D$  be a given set in  $R^n$ . The point  $X^* \in D$  is called **a local minimizer** of  $f$  over  $D$  if and only if

there exists  $\varepsilon > 0$  such that  $B(X^*, \varepsilon) \subset D$  and

$f(X) \geq f(X^*)$  for all  $X \in B(X^*, \varepsilon)$ , where  $B(X^*, \varepsilon)$  is the open ball with center  $X^*$  and radius  $\varepsilon$  defined by

$B(X^*, \varepsilon) = \{X \in R^n: \|X - X^*\| < \varepsilon\}$  and  $\|X - X^*\|$  means *the distance between  $X$  and  $X^*$  in the normed linear space  $(R^n, \|\cdot\|)$ .*

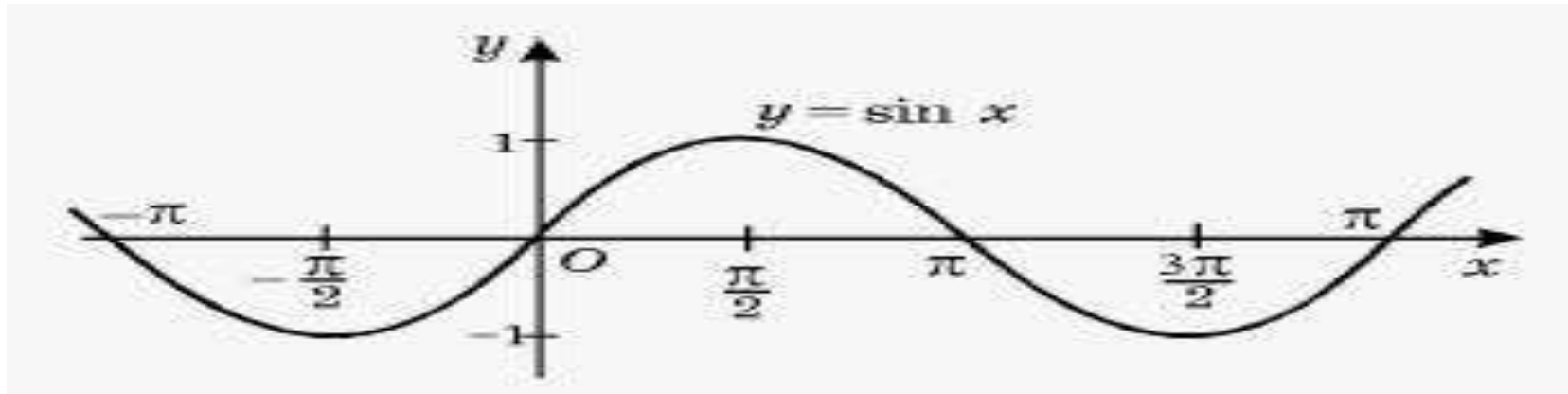
A point  $X^* \in D$  is called **a strict local minimizer** of  $f$  over  $D$  if and only if there exists  $\varepsilon > 0$  such that  $B(X^*, \varepsilon) \subset D$  and

$f(X) > f(X^*)$  for all  $X \in B(X^*, \varepsilon), X \neq X^*$ .

### Example (6):

Let  $f: \mathbb{R} \rightarrow \mathbb{R}$  be defined by  $f(x) = \sin x$ ,  $x \in \mathbb{R}$  in the normed linear space  $(\mathbb{R}, |\cdot|)$  and  $D = [0, 4\pi] \subset \mathbb{R}$ . Show that  $X^* = \frac{3\pi}{2}$  is a strict local minimizer of  $f$  over  $D$ .

### Solution:



Let  $\varepsilon > 0$  such that

$$B(X^*, \varepsilon) = \{X \in \mathbb{R} : |X - X^*| < \varepsilon\} = (X^* - \varepsilon, X^* + \varepsilon) \subset D.$$

Take  $0 < \varepsilon < \frac{3\pi}{2}$ . Then  $f(X) > f\left(\frac{3\pi}{2}\right)$  for all  $X$  in  $(X^* - \varepsilon, X^* + \varepsilon)$ ,  $X \neq \frac{3\pi}{2}$ .

Hence  $X^* = \frac{3\pi}{2}$  is a strict local minimizer of  $f$  over  $D$ .

**Example (7):**

Let  $f: \mathbb{R} \rightarrow \mathbb{R}$  be defined by  $f(x) = \begin{cases} 1 & , x \leq \frac{1}{2} \\ 0 & , \frac{1}{2} < x < \frac{3}{2} \\ 1 & , \frac{3}{2} \leq x \leq \frac{5}{2} \\ 0 & , x > \frac{5}{2} \end{cases}$ . Let  $D = (\frac{1}{2}, \frac{3}{2}) \subset \mathbb{R}$ .

Show that any point  $X^*$  in  $(\frac{1}{2}, \frac{3}{2})$  is a local minimizer of  $f$  over  $(\frac{1}{2}, \frac{3}{2})$ .

**Solution:**

Let  $\varepsilon > 0$  such that

$$B(X^*, \varepsilon) = \{X \in \mathbb{R} : |X - X^*| < \varepsilon\} = (X^* - \varepsilon, X^* + \varepsilon) \subset (\frac{1}{2}, \frac{3}{2}).$$

Then  $f(X) = f(X^*)$  for all values of  $X$  in  $(X^* - \varepsilon, X^* + \varepsilon)$ .

Hence  $X^*$  is a local minimizer of  $f$  over  $(\frac{1}{2}, \frac{3}{2})$ .

### Definition (13): (Global Minimizer)

Let  $f: R^n \rightarrow R$  be a given function and  $D$  be a given *set in*  $R^n$ . The point  $X^* \in D$  is called **a global minimizer** of  $f$  over  $D$  if and only if  $f(X) \geq f(X^*)$  for all  $X \in D$ .

The point  $X^* \in D$  is called **a strict global minimizer** of  $f$  over  $D$  if and only if  $f(X) \geq f(X^*)$  for all  $X \in D, X \neq X^*$ .

### Note (13):

The norm  $\|X\|_E$  is called **Euclidean norm** and defined as:

$$\|X\|_E = (\sum_{i=1}^n x_i^2)^{\frac{1}{2}}, X = [x_1, x_2, x_3, \dots, x_n]^T.$$

### Example (8):

Let  $f: R^2 \rightarrow R$  defined by  $f(X) = x_1^2 + x_2^2, X \in R^2$ .

Let  $D = B(0, 1) = \{X \in R^2: \|X\|_E < 1\}$ . Show that  $X^* = 0$  is a strict global minimizer of  $f$  over  $D$ .

### Solution:

Notice that, geometrically, the set  $D$  consists of all points  $X$  in the interior of the circular disk with *center 0 and radius 1*. Clearly  $f(X) > f(0)$  for all  $X \in D, X \neq 0$ . So,  $X^* = 0$  is a strict global minimizer of  $f$  over  $D$ .

**Example (9):**

Let  $f: R^2 \rightarrow R$  defined by  $f(X) = (x_2 - x_1^2)^2 + (1 - x_1)^2$ . Show that  $X^* = [1, 1]^T$  is a strict global minimizer of  $f$  over  $R^2$ .

**Solution:**

Let  $\delta = [\delta_1, \delta_2]^T \in R^2$ .

$$\begin{aligned} \therefore f(X^* + \delta) &= f(1 + \delta_1, 1 + \delta_2) = [1 + \delta_2 - (1 + \delta_1)^2]^2 + [1 - (1 + \delta_1)]^2 \\ &= [1 + \delta_2 - (1 + 2\delta_1 + \delta_1^2)]^2 + [1 - 1 - \delta_1]^2 \\ &= [1 + \delta_2 - 1 - 2\delta_1 - \delta_1^2]^2 + \delta_1^2 = [\delta_2 - 2\delta_1 - \delta_1^2]^2 + \delta_1^2. \end{aligned}$$

Since  $f(X^*) = f(1, 1) = 0$ .

$$\therefore f(X^* + \delta) = 0 + [\delta_2 - 2\delta_1 - \delta_1^2]^2 + \delta_1^2 = f(X^*) + [\delta_2 - 2\delta_1 - \delta_1^2]^2 + \delta_1^2.$$

Since  $[\delta_2 - 2\delta_1 - \delta_1^2]^2 > 0$  and  $\delta_1^2 > 0$ , for all  $\delta \neq 0$ .

$\therefore f(X^* + \delta) > f(X^*)$ , for all  $\delta \neq 0$ .

Hence  $X^*$  is a strong global minimizer of  $f$  over  $R^2$ .

**Example (10):**

Let  $f: R \rightarrow R$  be defined by

$$f(x) = \begin{cases} (x - 1)^2, & x \text{ not in the interval } [0, 2] \\ 1 & , x \text{ in the interval } [0, 2] \end{cases}.$$

Show that any point in  $[0, 2]$  is a global minimizer of  $f$  over  $R^2$ .

**Solution:**

Let  $x^*$  be any point *in the interval*  $[0, 2]$ , then  $f(x^*) = 1$   
and  $f(x^*) > 1$  *for all*  $x^*$  *not in the interval*  $[0, 2]$ .

Hence  $x^*$  is a global minimizer of  $f$  over  $R^2$ .



### Definition (14): (Maximizer)

Let  $f: R^n \rightarrow R$  be a given function and  $D$  be a given *set in  $R^n$* . The point  $X^* \in D$  is called **a maximizer** of  $f$  over  $D$  if and only if  $X^*$  is a minimizer of  $-f$  over  $D$ .

### Definition (15): (Critical Point)

Let  $f: R^n \rightarrow R$  have *first partial derivatives in  $D \subset R^n$* . The point  $X^* \in D$  is called a **critical point**  $f$  in  $D$  if and only if  $g(X^*) = 0$ , where  $g$  is the gradient vector of  $f$ .

**Example (11):**

Let  $f: R^2 \rightarrow R$  be defined by  $f(X) = x_1^3 - 2x_1^2x_2 + x_2^2$ ,  $X \in R^2$ . Find the critical points of  $f$ .

**Solution:**

First, we find the gradient of  $f$  which is  $g(X) = \left[ \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2} \right]^T$ .

$$\frac{\partial f}{\partial x_1} = 3x_1^2 - 4x_1x_2, \quad \frac{\partial f}{\partial x_2} = -2x_1^2 + 2x_2.$$

$$\therefore g(X) = [3x_1^2 - 4x_1x_2, -2x_1^2 + 2x_2]^T.$$

Second, put  $g(X) = 0$ .

$$\therefore \quad 3x_1^2 - 4x_1x_2 = 0 \dots\dots\dots (1)$$

$$\text{and} \quad -2x_1^2 + 2x_2 = 0 \dots\dots\dots (2)$$

$$\text{From (2) we get } x_2 = x_1^2 \dots\dots\dots (3)$$

From (3) and (1) we get  $3x_1^2 - 4x_1^3 = 0 \rightarrow x_1^2(3 - 4x_1) = 0 \rightarrow$

$$x_1 = 0 \text{ or } x_1 = \frac{3}{4}. \text{ From (3) we get } x_2 = 0 \text{ or } x_2 = \frac{9}{16}.$$

$\therefore$  The critical points are  $[0, 0]^T$  and  $[\frac{3}{4}, \frac{9}{16}]^T$ .