

# Introduction to Nonlinear Programming

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## Introduction

A general NLP is represented as

$$\begin{aligned} & \text{Minimize } f(X) \\ & \text{Subject to } g_i(X) \leq 0, i = 1, 2 \cdots m \\ & \quad h_j(X) = 0, j = 1, 2 \cdots l \\ & \quad X^L \leq X \leq X^U \end{aligned}$$

Where,  $X = (x_1, x_2 \cdots x_n)^T$  is column vector of  $n$  real-valued design variables. And  $X^L$  and  $X^U$  represent explicit lower and upper bounds on the design variables



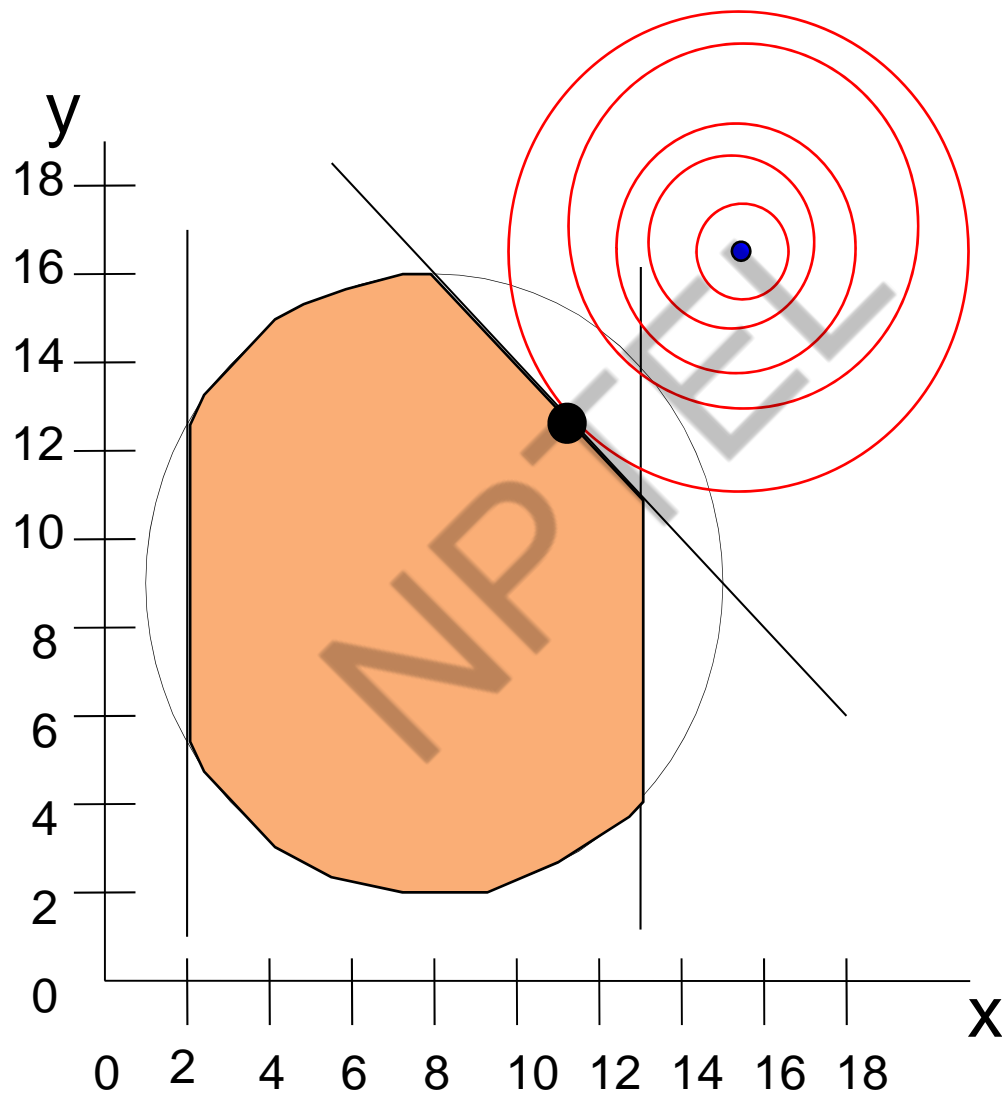
## Example

Let us take a simple example

$$\begin{aligned} & \text{Minimize } \sqrt{(x - 14)^2 + (y - 15)^2} \\ & \text{Subject to } (x - 8)^2 + (y - 9)^2 \leq 49 \\ & \quad 2 \leq x \leq 13 \\ & \quad x + y \leq 24 \\ & \quad x, y \geq 0. \end{aligned}$$



# Graph

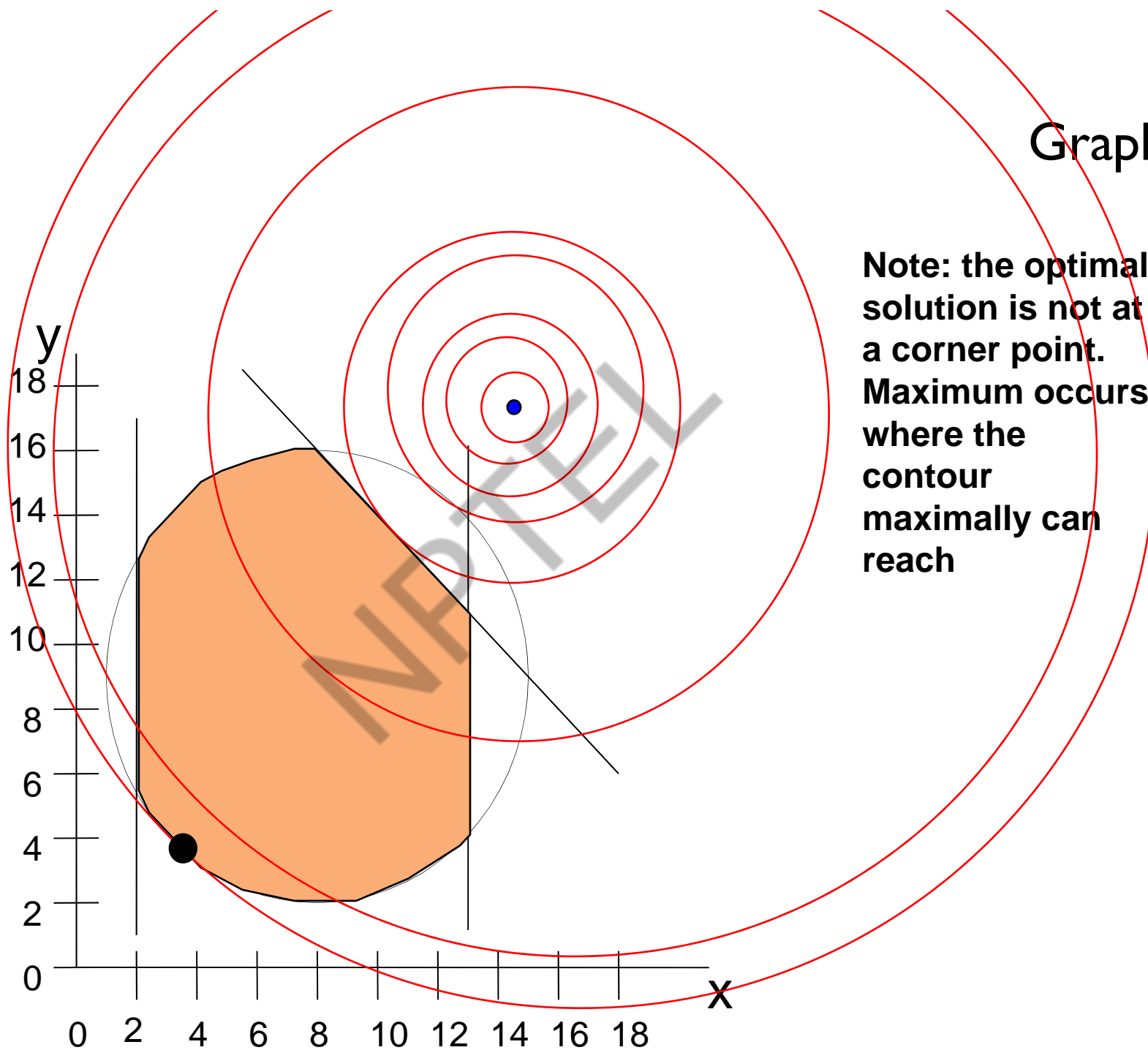


**Note: the optimal solution is not at a corner point. It is where the contour first hits the feasible region**



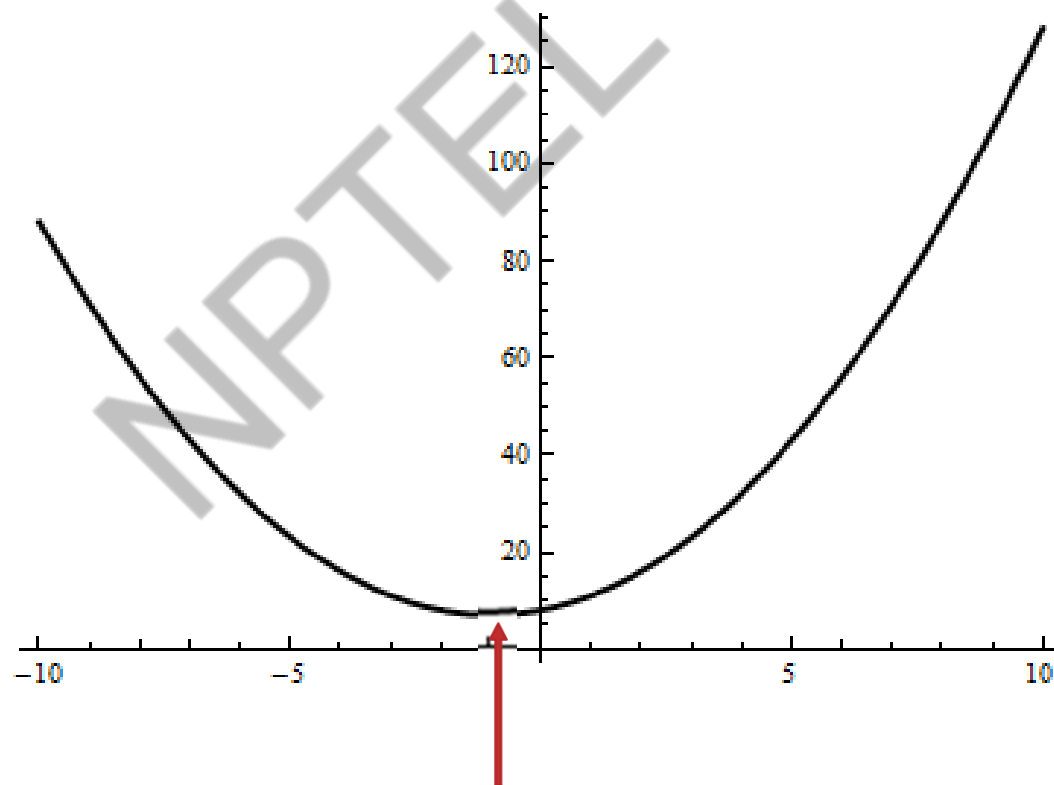
## Graph

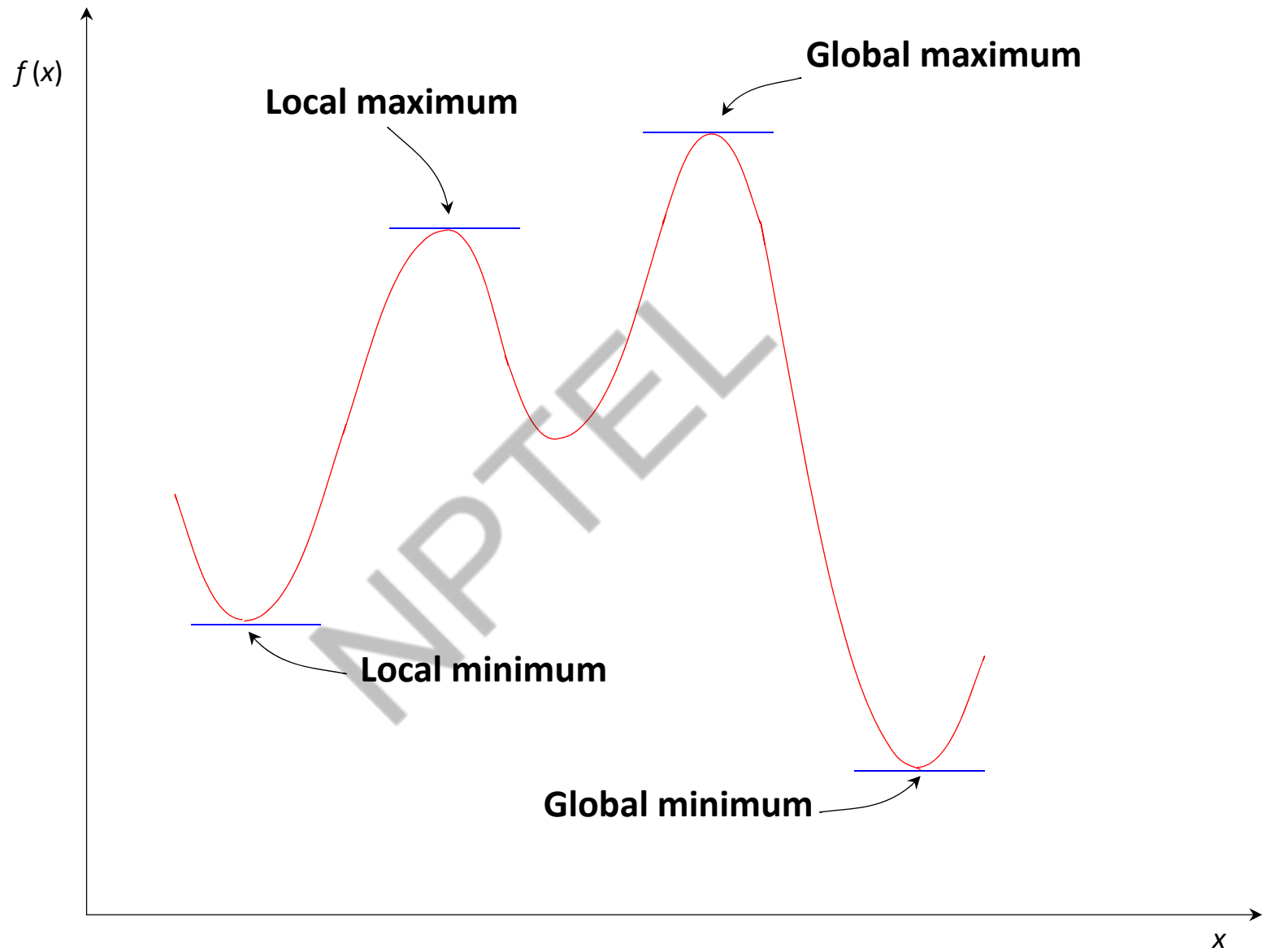
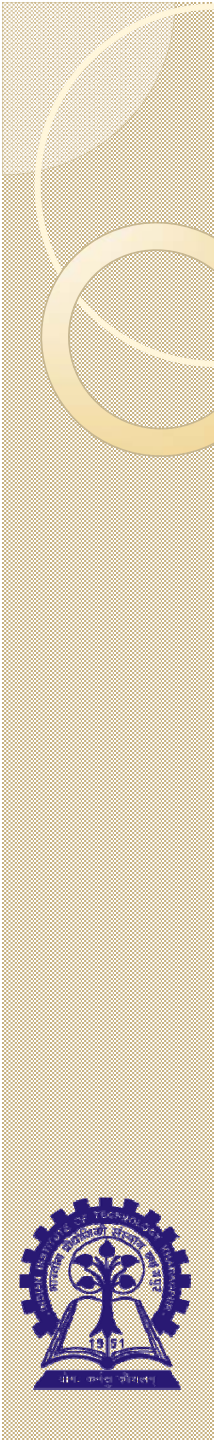
**Note: the optimal solution is not at a corner point. Maximum occurs where the contour maximally can reach**



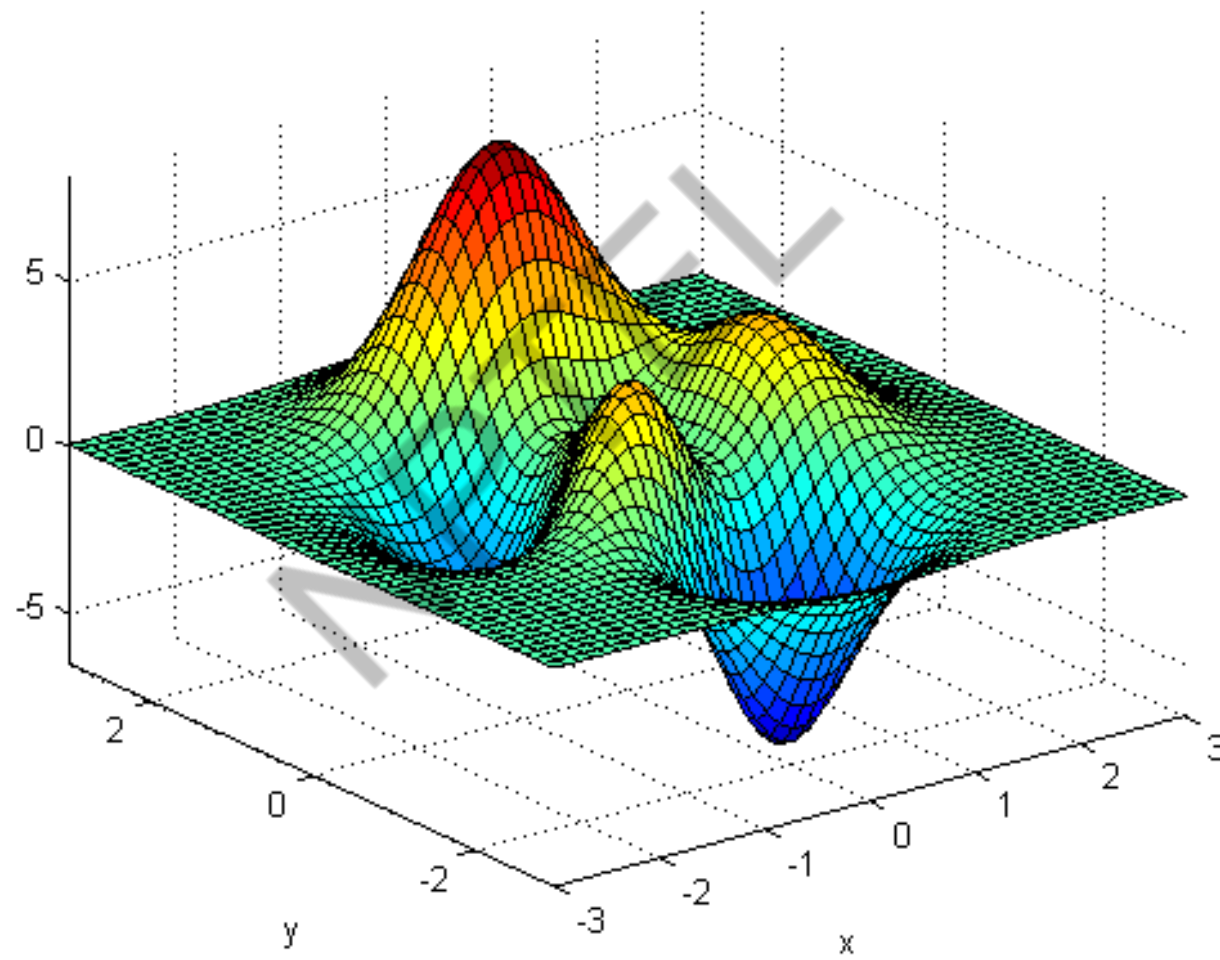
## Unconstrained problem

$$\text{Minimize } x^2 + 2x + 8$$





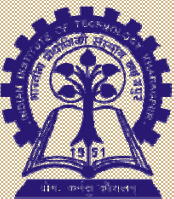
Peaks





## Optimality Criteria

- In finding optimal solution, two questions generally must be addressed:
  1. **Static Question.** How can one determine whether a given point  $x^*$  is the optimal solution?
  2. **Dynamic Question.** If  $x^*$  is not the optimal point, then how does one go about finding a solution that is optimal?



# What is a Function?

- Monotonic and unimodal functions

Monotonic:

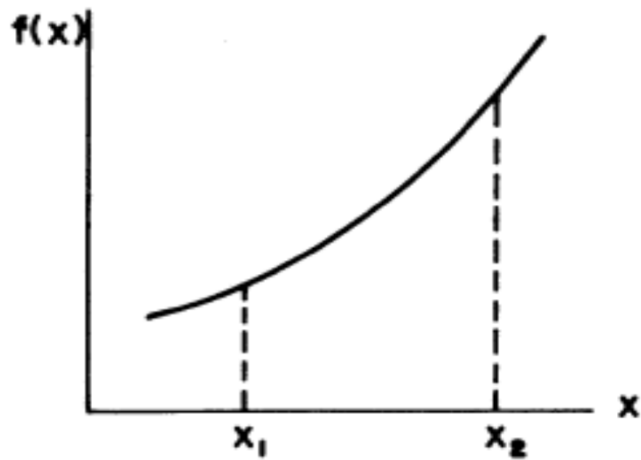
*for any two points  $x_1$  and  $x_2$ , where  $x_1 \leq x_2$   
if  $f(x_1) \leq f(x_2)$  (monotonically increasing)  
if  $f(x_1) \geq f(x_2)$  (monotonically decreasing)*

Unimodal:

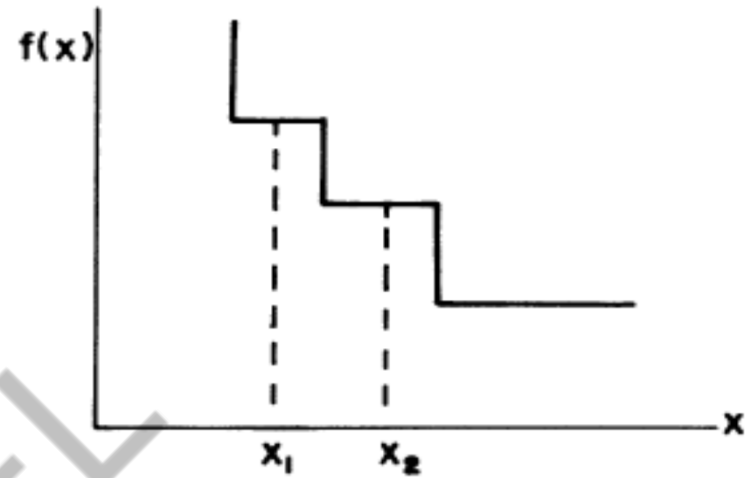
*$f(x)$  is unimodal on the interval  $a \leq x \leq b$  if and only if it is monotonic on either of the single optimal point  $x^*$  in the interval*

***Unimodality is an extremely important functional property used in optimization.***

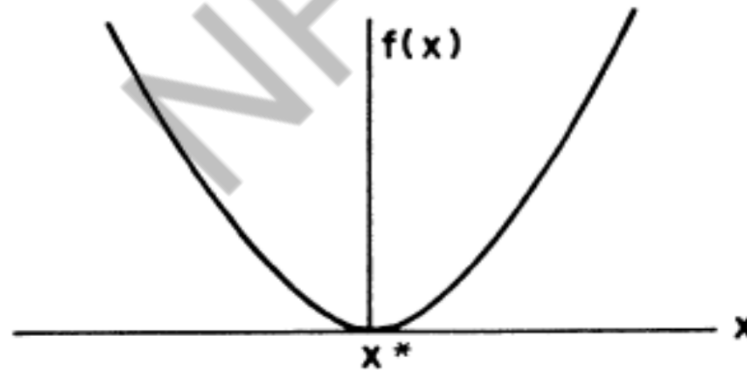




A monotonic increasing function



A monotonic decreasing function



An unimodal function



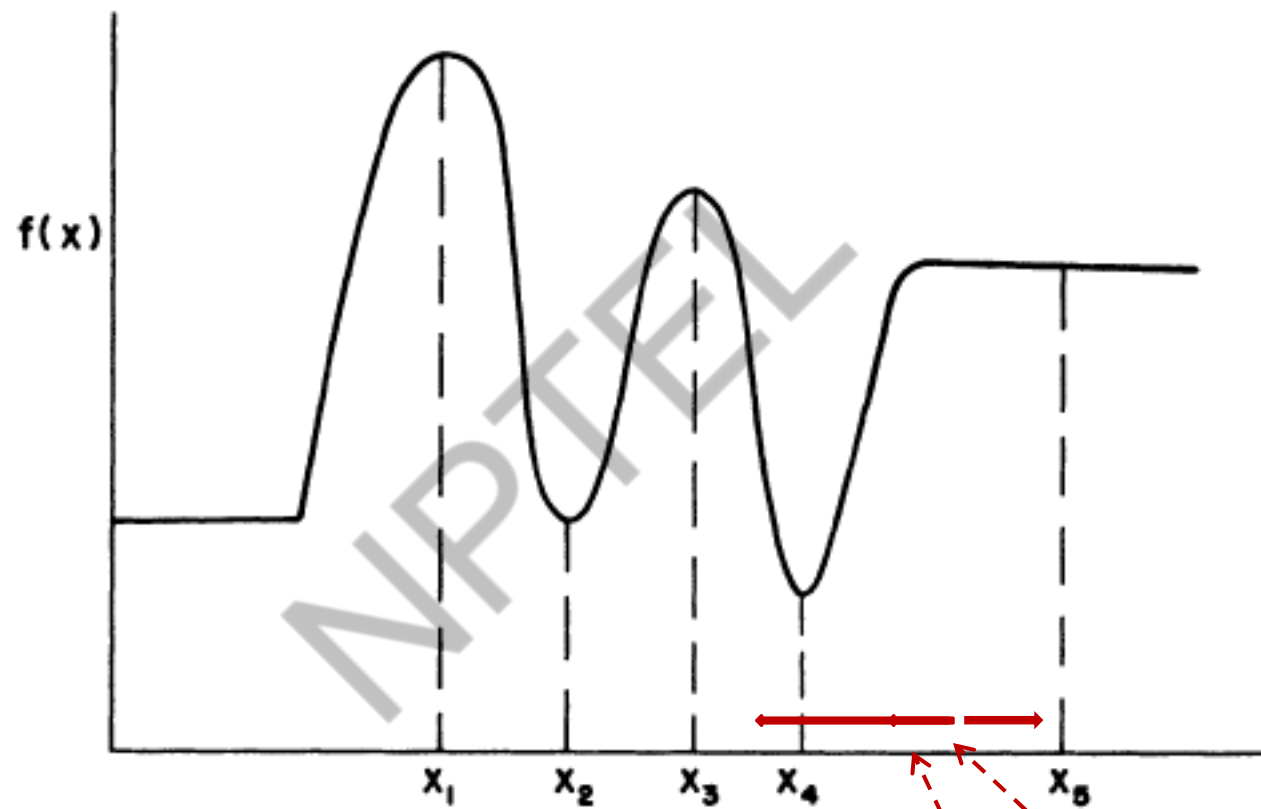


Figure 2.7. Local and global optima.



## Basic Philosophy for solving NLP

To produce a sequence of improved approximations to the optimum according to the following scheme

1. Start with an initial trial point  $X_i$
2. Find a suitable direction  $S_i$
3. Find an appropriate step length  $\lambda_i$
4. Obtain a new approximation

$$X_{i+1} = X_i + \lambda_i S_i$$

5. Test whether  $X_{i+1}$  is optimum



## Issues to be addressed

- Nature of the functions : convex/concave
- Modality
- Gradient of functions involved
- Optima are not restricted to extreme points
- Distinguish between local and global optimum
- If the feasible region is disconnected or combination of discrete spaces



## Issues to be addressed

- Different starting point leads to different solution
- Difficult to find the feasible starting point
- It is not possible to identify whether the model is infeasible / unbounded
- There are numerous algorithm to solve NLP
- How will you know the function is convex or concave in the region of interest
- You need to know how to use the available different solver



# Problem Formulations & Graphical Solution

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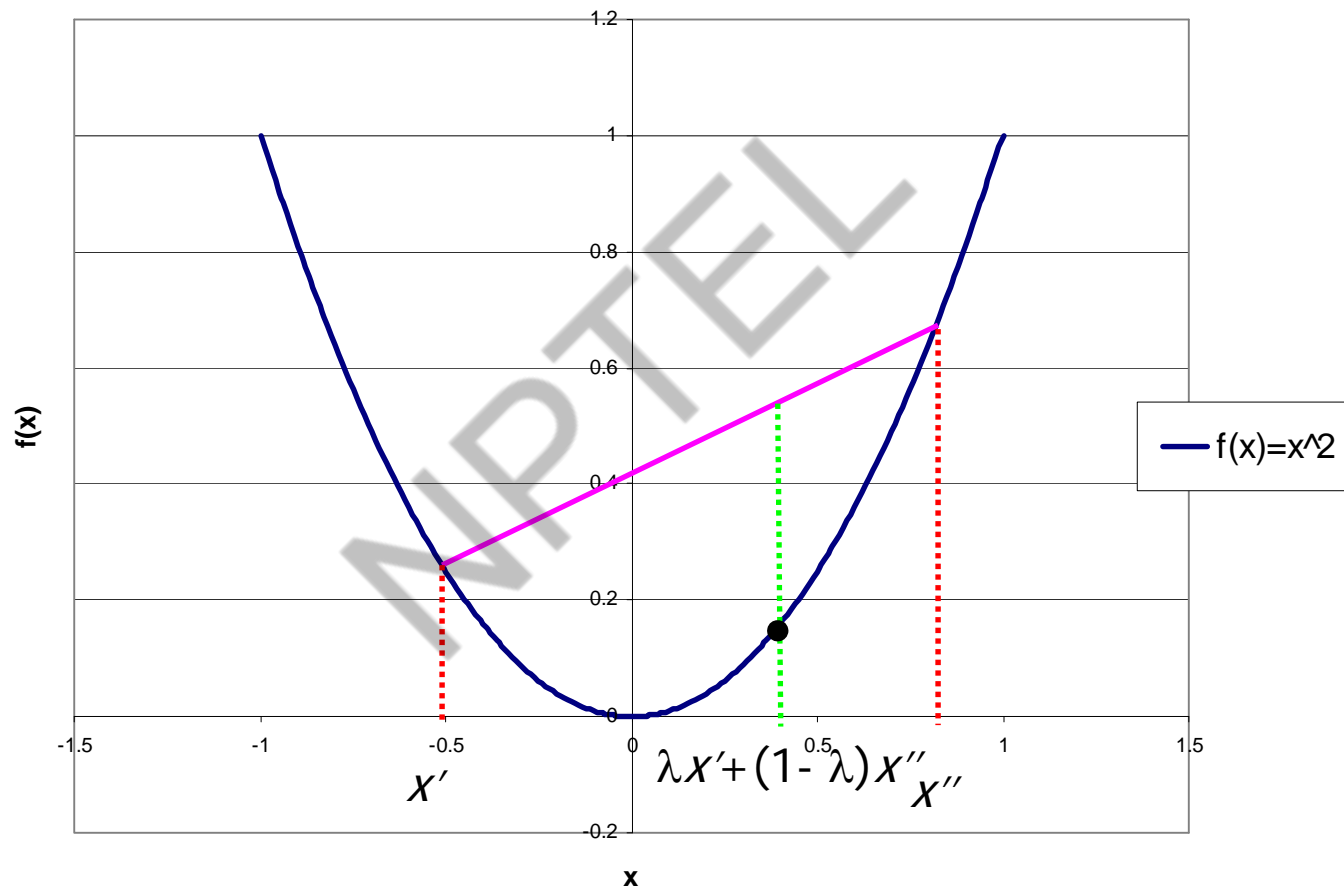
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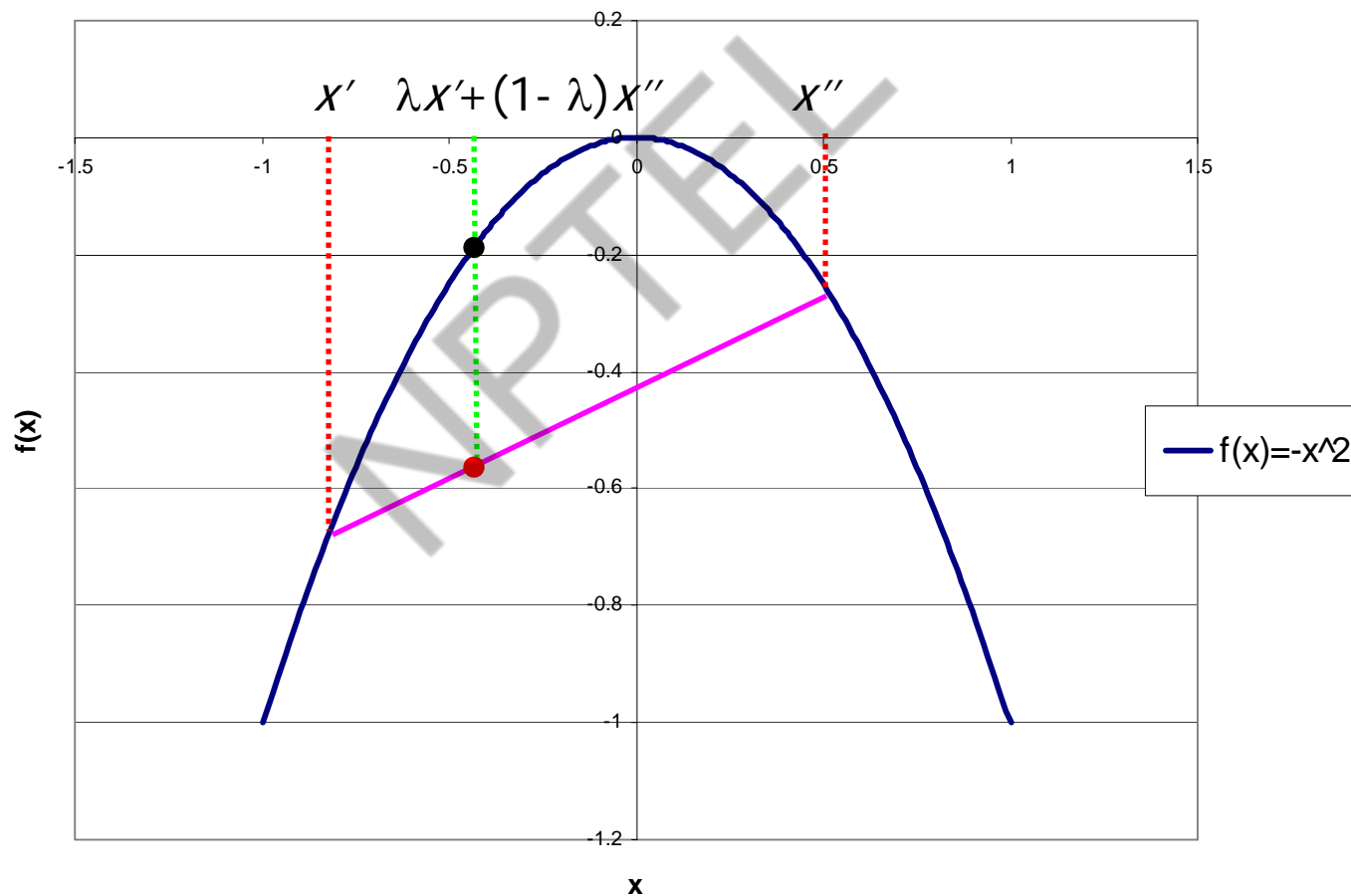
# Concave Upward or Convex downward or Convex

$$f(x) = x^2$$



# Concave Downward or Convex Upward or Concave

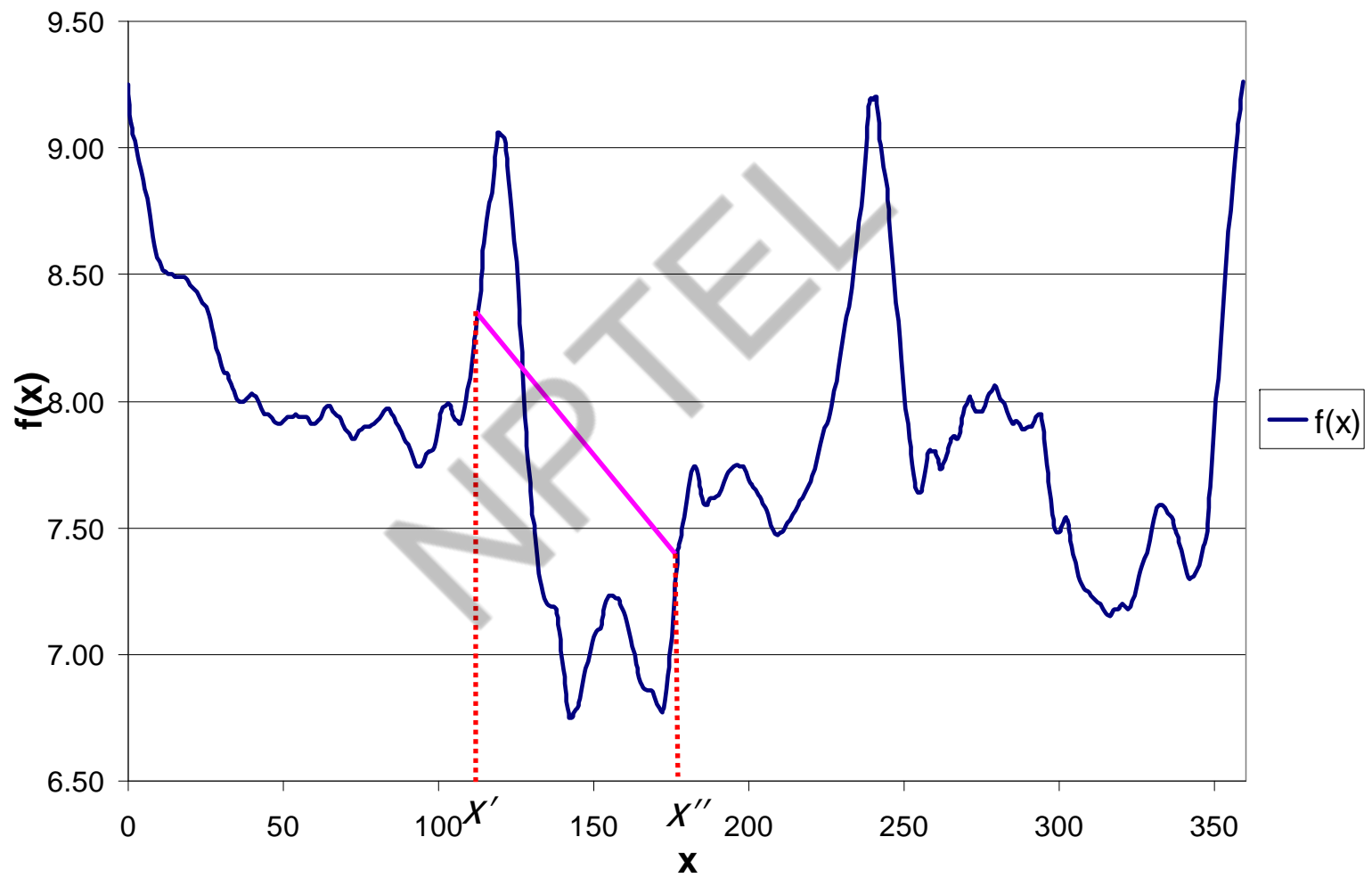
$$f(x) = -x^2$$



# Convex and Concave Function

- The function  $f$  is a *convex* function if
$$f(\lambda x' + (1 - \lambda)x'') \leq \lambda f(x') + (1 - \lambda)f(x'')$$
- The function  $f$  is a *concave* function if
$$f(\lambda x' + (1 - \lambda)x'') \geq \lambda f(x') + (1 - \lambda)f(x'')$$





# Important Fact

- ✚ Minimization of a Convex Function over Convex Sets any local minimum is a global minimum.
- ✚ Maximization of a Concave Function over Convex Sets any local maximum is a global maximum.



# Local Optimality

A function of one variable is said to have a *relative or local minimum* at  $x = x^*$  if  $f(x^*) \leq f(x^* + h)$  for all sufficiently small positive and negative value of  $h$ .

A function of one variable is said to have a *relative or local maximum* at  $x = x^*$  if  $f(x^*) \geq f(x^* + h)$  for all sufficiently small positive and negative value of  $h$ .





# Global Optimality

A function of one variable is said to have a *global minimum* at  $x = x^*$  if  $f(x^*) \leq f(x)$  for all  $x$  in the domain of the function.

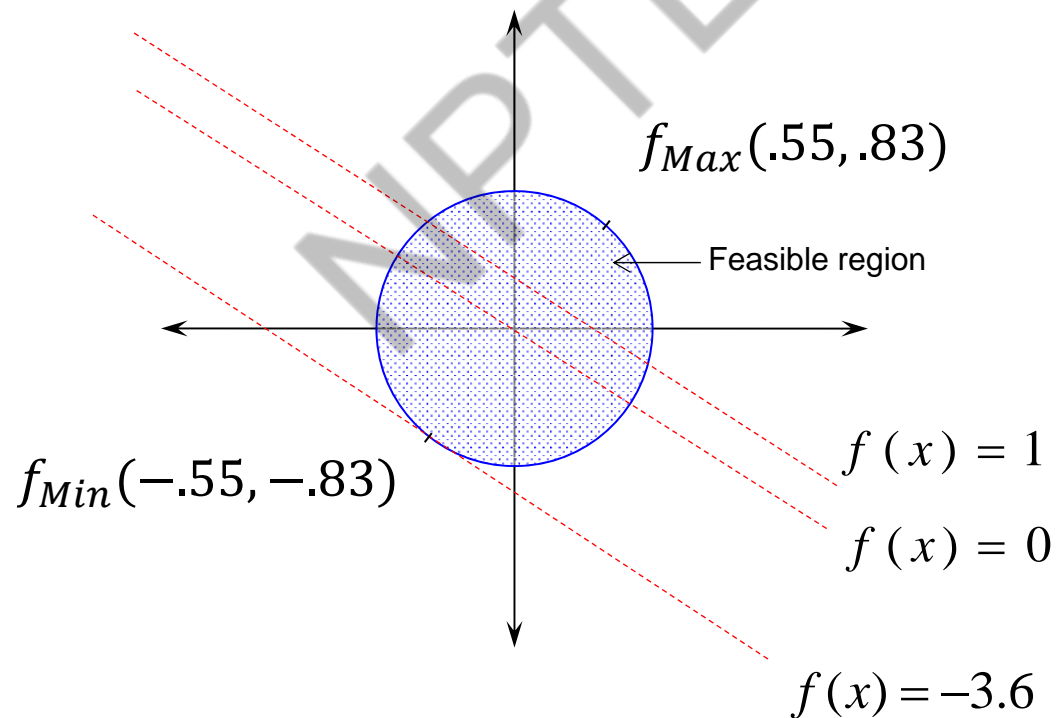
A function of one variable is said to have a *global maximum* at  $x = x^*$  if  $f(x^*) \geq f(x)$  for all  $x$  in the domain of the function.

If the function is increasing (or decreasing) on either side of the point  $x^*$ , then  $x^*$  is the *point of inflection*.



# Graphical Solution

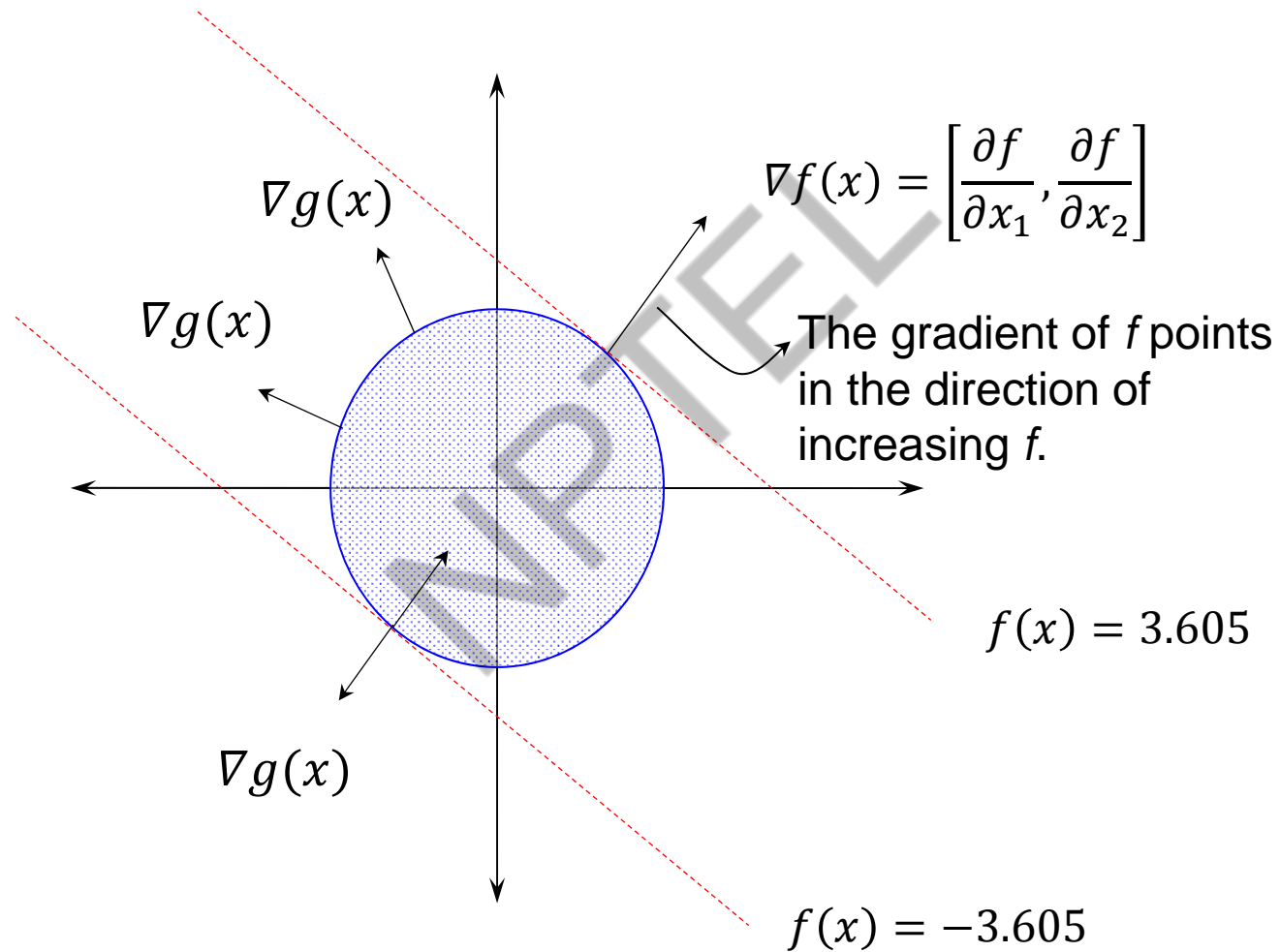
*Minimize  $2x + 3y$*   
*Subject to  $x^2 + y^2 = 1, x, y \geq 0$*



# Graphical Solution

$$f(x) = 3.605$$

$$f(x) = -3.605$$



## Problem 2

$$\text{Minimize } \frac{1}{2}x^2 + y$$

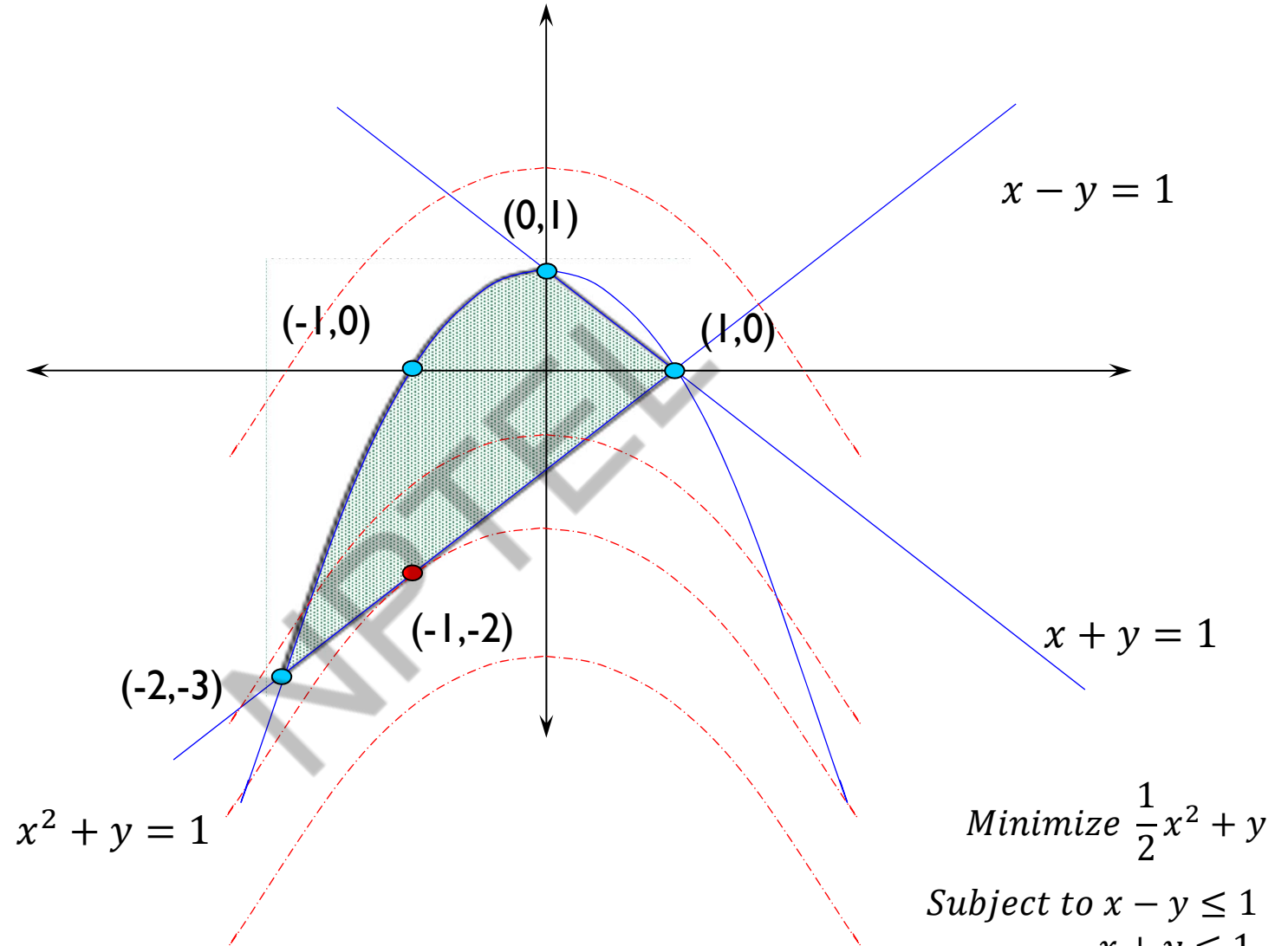
$$\text{Subject to } x - y \leq 1$$

$$x + y \leq 1$$

$$x^2 + y \leq 1$$

$$x, y \geq 0$$





# Classification based on :

- Constraints
- Nature of Design Variable
- Nature of Equations involved
- Permissible Values of Design Variables
- Randomness involved in Design Variables
- Separability of Functions
- Number of Objective Functions





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# Types of Optimization Problems

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# Classification of Optimization Problems

Classification can be based on:

- **Constraints**
- **Nature of the design variables**
- **Physical structure of the problem**
- **Nature of the equations involved**
- **Permissible values of the design variables**
- **Deterministic nature of the variables**
- **Separability of the functions**





# Classification of Optimization Problems

- **Constraints**
  - ✓ Constrained optimization problem
  - ✓ Unconstrained optimization problem
- **Nature of the design variables**
  - ✓ Static optimization problems
  - ✓ Dynamic optimization problems



# Classification of Optimization Problems

- **Physical structure of the problem**
  - ✓ Optimal control problems
  - ✓ Non-optimal control problems
- **Nature of the equations involved**
  - ✓ Nonlinear programming problem
  - ✓ Geometric programming problem
  - ✓ Quadratic programming problem
  - ✓ Linear programming problem



# Classification of Optimization Problems

- **Permissible values of the design variables**
  - ✓ Integer programming problems
  - ✓ Real valued programming problems
- **Deterministic nature of the variables**
  - ✓ Stochastic programming problem
  - ✓ Deterministic programming problem



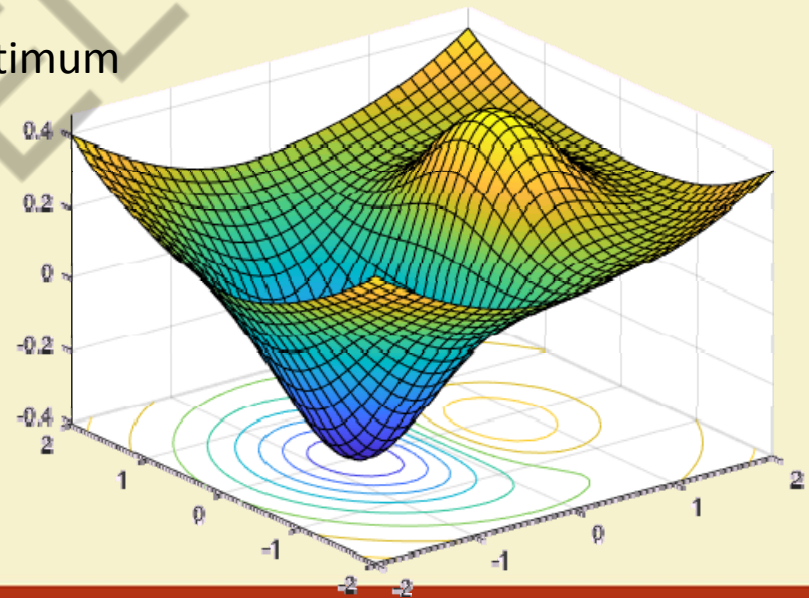
# Classification of Optimization Problems

- **Separability of the functions**
  - ✓ Separable programming problems
  - ✓ Non-separable programming problems
- **Number of the objective functions**
  - ✓ Single objective programming problem
  - ✓ Multiobjective programming problem



# Unconstrained General Optimization Problem

- ✓ **Objective:** Find minimum of  $F(x)$  where  $x$  is a vector of design variables
- ✓ We may know lower and upper bounds for optimum
- ✓ No constraints involved



# Constrained General Optimization Problem

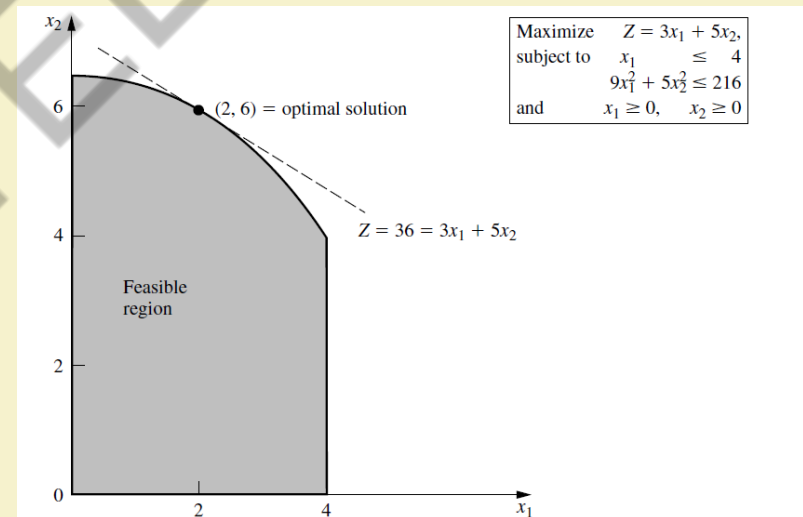
- ✓ **Objective:** Find minimum of  $F(x)$  where  $x$  is a vector of design variables subject to a set of constraints

- ✓ General format for Minimization problem:

$$\text{Minimize } F(x)$$

$$\text{subject to } G_i(x) \geq 0$$

$$x \geq 0, \quad i = 1, \dots, n$$



# Quadratic Programming Problem

- ✓ A **quadratic programming problem** is a nonlinear programming problem with a quadratic objective function and linear constraints. It is usually formulated as follows:

subject to

$$\begin{aligned} F(\mathbf{X}) &= c + q^T X + \frac{1}{2} X^T Q X \\ &= c + \sum_{i=1}^n q_i x_i + \sum_{i=1}^n \sum_{j=1}^n Q_{ij} x_i x_j \\ \sum_{i=1}^n a_{ij} x_i &= b_j, \quad j = 1, 2, \dots, m \\ x_i &\geq 0, \quad i = 1, 2, \dots, n \end{aligned}$$

where  $c$ ,  $q_i$ ,  $Q_{ij}$ ,  $a_{ij}$ , and  $b_j$  are constants.



$$\begin{aligned} & \text{Minimize } 2x_1^2 + 2x_2^2 - 2x_1x_2 - 4x_1 - 6x_2 + 8 \\ & \text{Subject to } x_1 + x_2 \leq 2 \\ & \quad x_1 + 5x_2 \leq 5, -x_1 \leq 0, -x_2 \leq 0 \end{aligned}$$





# Integer Programming Problem

- ✓ If some or all of the design variables  $x_1, x_2, \dots, x_n$  of an optimization problem are restricted to take on only integer (or discrete) values, the problem is called an ***integer programming problem***.
- ✓ **General form:** maximize  $c^T x$   
subject to  $Ax \leq b$   
 $x \geq 0, x \in \mathbb{Z}^n$

where  $c, b$  are vectors and  $A$  is a matrix whose all entries are integers.



# Separable Programming Problem

- ✓ A function  $f(x)$  is said to be **separable** if it can be expressed as the sum of  $n$  single variable functions,  $f_1(x_1), f_2(x_2), \dots, f_n(x_n)$ , that is,

$$f(\mathbf{X}) = \sum_{i=1}^n f_i(x_i)$$

- ✓ A **separable programming problem** is one in which the objective function and the constraints are separable.

Find  $\mathbf{X}$  which minimizes  $f(\mathbf{X}) = \sum_{i=1}^n f_i(x_i)$   
subject to

$$g_j(\mathbf{X}) = \sum_{i=1}^n g_{ij}(x_i) \leq b_j, \quad j = 1, 2, \dots, m$$

where  $b_j$  is constant



Example

$$\begin{array}{ll}\text{Minimize} & x_1^2 + x_2^2 + x_3^2 \\ \text{Subject to} & x_1 + x_2 + x_3 \geq 15, \quad x_1, x_2, x_3 \geq 0\end{array}$$

Example

$$\begin{array}{ll}\text{Maximize} & x_1 x_2 x_3 \\ \text{Subject to} & x_1 + x_2 + x_3 = 5, \quad x_1, x_2, x_3 \geq 0\end{array}$$

