



Introduction and Basic Concepts

(ii) Optimization Problem and Model Formulation



Objectives

- To study the basic components of an optimization problem.
- Formulation of design problems as mathematical programming problems.



Introduction - Preliminaries

- Basic components of an optimization problem :
 - An **objective function** expresses the main aim of the model which is either to be minimized or maximized.
 - A set of **unknowns** or **variables** which control the value of the objective function.
 - A set of **constraints** that allow the unknowns to take on certain values but exclude others.



Introduction (contd.)

- The optimization problem is then to:
 - find values of the *variables* that minimize or maximize the *objective function* while satisfying the *constraints*.



Objective Function

- As already defined the objective function is the mathematical function one wants to maximize or minimize, subject to certain constraints. Many optimization problems have a single objective function (When they don't they can often be reformulated so that they do). The two interesting exceptions are:
 - **No objective function.** The user does not particularly want to optimize anything so there is no reason to define an objective function. Usually called a *feasibility problem*.
 - **Multiple objective functions.** In practice, problems with multiple objectives are reformulated as single-objective problems by either forming a weighted combination of the different objectives or by treating some of the objectives by constraints.



Statement of an optimization problem

To find $\mathbf{X} = \begin{pmatrix} x_1 \\ x_2 \\ . \\ . \\ x_n \end{pmatrix}$ which maximizes $f(\mathbf{X})$

Subject to the constraints

$$\begin{aligned} g_i(\mathbf{X}) &\leq 0, & i &= 1, 2, \dots, m \\ l_j(\mathbf{X}) &= 0, & j &= 1, 2, \dots, p \end{aligned}$$



Statement of an optimization problem

where

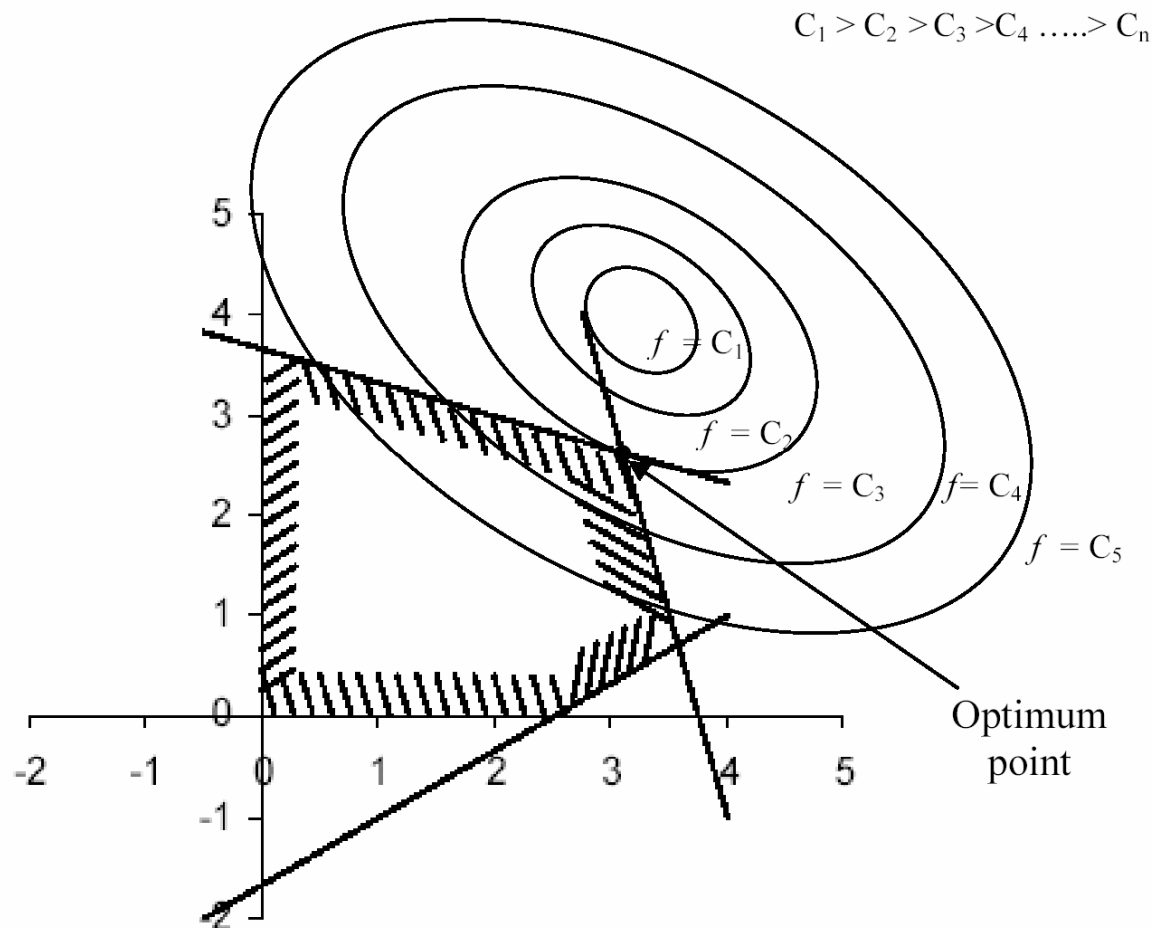
- \mathbf{X} is an n -dimensional vector called the design vector
 - $f(\mathbf{X})$ is called the *objective function*, and
 - $g_i(\mathbf{X})$ and $h_j(\mathbf{X})$ are known as inequality and equality constraints, respectively.
- This type of problem is called a *constrained optimization problem*.
 - Optimization problems can be defined without any constraints as well. Such problems are called *unconstrained optimization problems*.



Objective Function Surface

- If the locus of all points satisfying $f(\mathbf{X}) = \text{a constant } \mathbf{c}$ is considered, it can form a family of surfaces in the design space called the ***objective function surfaces***.
- When drawn with the constraint surfaces as shown in the figure we can identify the optimum point (maxima).
- This is possible graphically only when the number of design variable is two.
- When we have three or more design variables because of complexity in the objective function surface we have to solve the problem as a mathematical problem and this visualization is not possible.

Objective function surfaces to find the optimum point (maxima)





Variables and Constraints

- Variables
 - These are essential. If there are no variables, we cannot define the objective function and the problem constraints.
- Constraints
 - Even though Constraints are not essential, it has been argued that almost all problems really do have constraints.
 - In many practical problems, one cannot choose the design variable arbitrarily. *Design constraints* are restrictions that must be satisfied to produce an acceptable design.



Constraints (contd.)

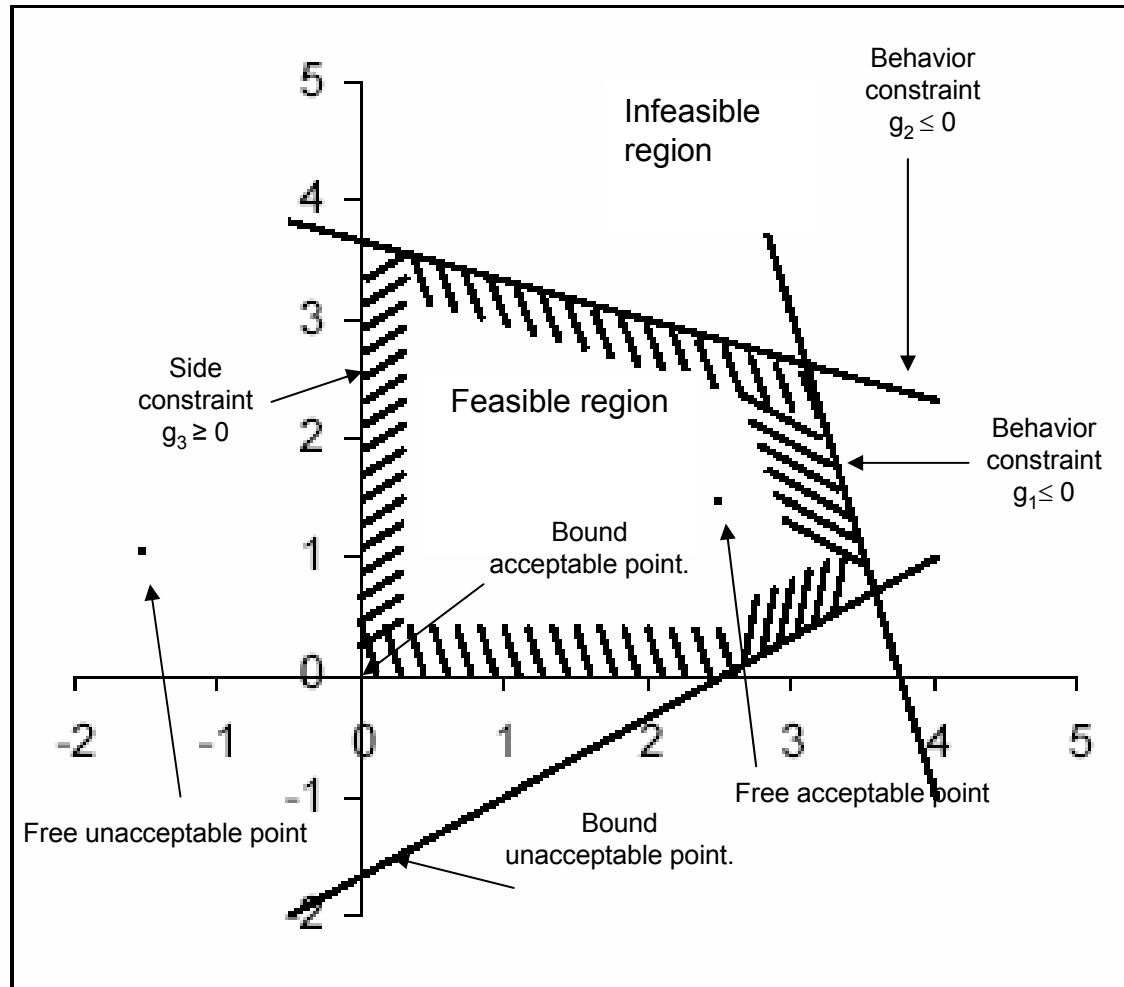
- Constraints can be broadly classified as :
 - **Behavioral or Functional constraints** : These represent limitations on the behavior and performance of the system.
 - **Geometric or Side constraints** : These represent physical limitations on design variables such as availability, fabricability, and transportability.



Constraint Surfaces

- Consider the optimization problem presented earlier with only inequality constraints $g_i(\mathbf{X})$. The set of values of \mathbf{X} that satisfy the equation $g_i(\mathbf{X}) = 0$ forms a boundary surface in the design space called a *constraint surface*.
- The constraint surface divides the design space into two regions: one with $g_i(\mathbf{X}) < 0$ (feasible region) and the other in which $g_i(\mathbf{X}) > 0$ (infeasible region). The points lying on the hyper surface will satisfy $g_i(\mathbf{X}) = 0$.

The figure shows a hypothetical two-dimensional design space where the feasible region is denoted by hatched lines.





Formulation of design problems as mathematical programming problems

- The following steps summarize the procedure used to formulate and solve mathematical programming problems.
 1. Analyze the process to identify the process variables and specific characteristics of interest i.e. make a list of all variables.
 2. Determine the criterion for optimization and specify the objective function in terms of the above variables together with coefficients.

3. Develop via mathematical expressions a valid process model that relates the input-output variables of the process and associated coefficients.
 - a) Include both equality and inequality constraints
 - b) Use well known physical principles
 - c) Identify the independent and dependent variables to get the number of degrees of freedom
4. If the problem formulation is too large in scope:
 - a) break it up into manageable parts/ or
 - b) simplify the objective function and the model
5. Apply a suitable optimization technique for mathematical statement of the problem.
6. Examine the sensitivity of the result to changes in the coefficients in the problem and the assumptions.



Thank You