Linear and Discrete Optimization (G54LDO)

Semester 1 of Academic Session 2017-2018
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Lecture 9 – Multi-objective Optimization

- •Problems With Multiple Objectives

 To explain the need for considering multiple objectives in optimization

 To explain the concept of multi-objective optimization
- •Optimality in Multi-objective Problems

 To illustrate the conflicting nature of multiple objectives

 To describe the key concepts of multi-objective optimization
- •Approaches for Multi-objective Optimization
 To explain different techniques to tackle multiple-objectives
- ·Goal Programming
 Formulate multi-objective problems using goal programming

Additional Reading

Supplement to Chapter 8 of (Hillier and Lieberman, 2015) on <u>Linear Goal Programming and its Solution Procedures</u>.

Chapter 8 of (Taha, 2007)

Chapter 8 of (Rardin, 1998)

A Review of Goal Programming and its Applications. M. Tamiz, D. Jones. Annals of Operations Research, Springer, 1995, 58, 39-53

An Integer Goal Programming Model to Allocate Offices to Staff in an Academic Institution. J. Giannikos, E. El-Darzi, P. Lees. Journal of the Operational Research Ssociety, Vol. 46, No.6, pp. 713-720, 1995.

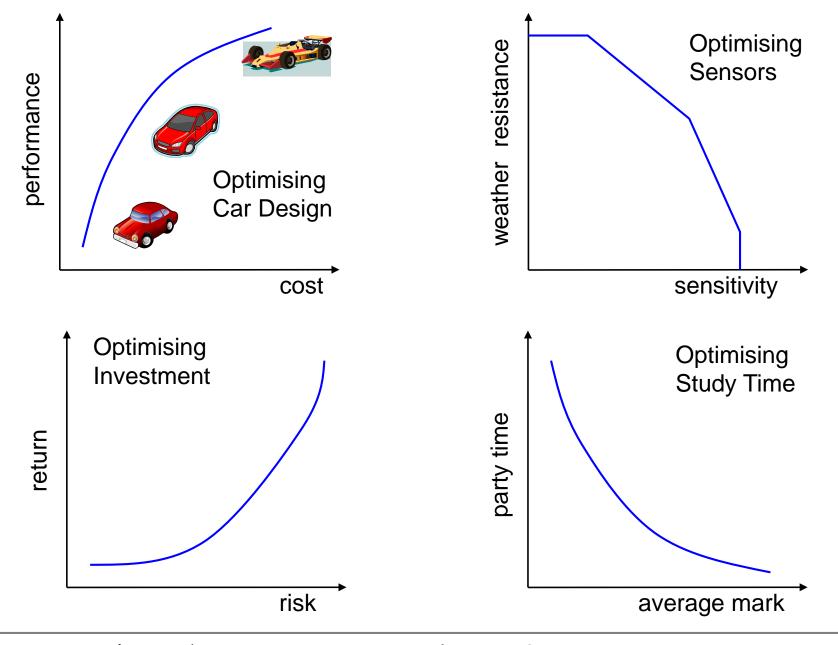
Problems with Multiple Objectives

Many real-world problems have <u>multiple objectives</u> but it is acceptable to tackle them using a <u>single-objective</u> <u>approach</u>.

However, in many scenarios various objectives, perhaps some of them conflicting, should be optimized at the same time using a <u>multi-objective approach</u>.

Moreover, the same solution might be evaluated with <u>different criteria</u> expressed by different decision-makers.

Single-objective models are preferred to multi-objective ones because <u>conflicting objectives make models less</u> tractable.



Example. Consider the following 2-objective LP model and compare the optimum with respect to each objective using the graphical method.

Maximize: $Z_1 = 3x_1 + x_2$

Minimize: $Z_2 = x_1 - x_2$

Subject to: $x_1 + x_2 \le 4$ (1)

$$x_1 \le 3 \tag{2}$$

$$x_2 \le 3 \tag{3}$$

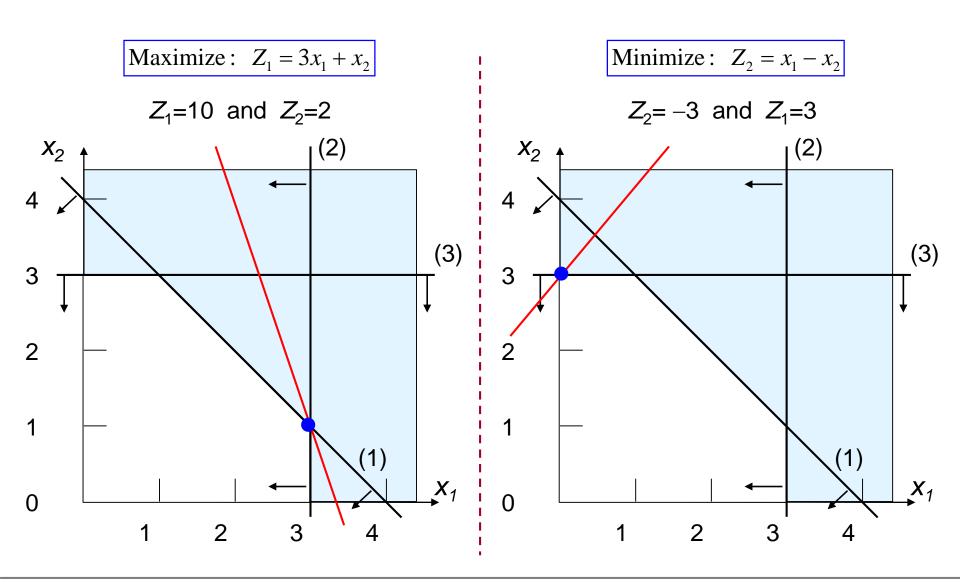
$$x_1, x_2 \ge 0 \tag{4}$$

Objective Z₁ could be profit.

Objective Z_2 could be waste.

First, solve for Z_1 while ignoring Z_2 and then solve for Z_2 while ignoring Z_1 .

Example (cont.)



Optimality in Multi-objective Problems

The objectives in multi-objective problems might be:

- Conflicting
- Incommensurable

As a result, the concept of optimal solution becomes less clear in multi-objective problems because each solution represents a <u>trade-off or compromise</u> between objectives.

Given a k-objective optimization problem:

Efficient solution. A solution S is efficient if there is no other solution that is at least as good as S is all k objectives and strictly better than S in one objective.

Efficient frontier. The collection of all efficient points corresponding to efficient solutions in a multi-objective optimisation problem.

An efficient solution is also called <u>Pareto optimal</u> solution or <u>non-dominated</u> solution.

The whole set of efficient solutions is also called the Pareto optimal set of the non-dominated set.

The efficient frontier is also called <u>Pareto optimal front</u> or the non-dominated front.

Other terms used in multi-objective optimization are <u>trade-off</u> and <u>compromise</u>.

Efficient solutions and efficient points

Maximize: $Z_1 = 3x_1 + x_2$

Minimize: $Z_2 = x_1 - x_2$

Subject to: $x_1 + x_2 \le 4$ (1)

$$x_1 \le 3 \tag{2}$$

$$x_2 \le 3 \tag{3}$$

$$x_1, x_2 \ge 0 \tag{4}$$

Efficient solutions

Efficient points

$$\mathbf{x}_1 = 0$$

$$Z_1 = 3$$

$$x_2 = 3$$

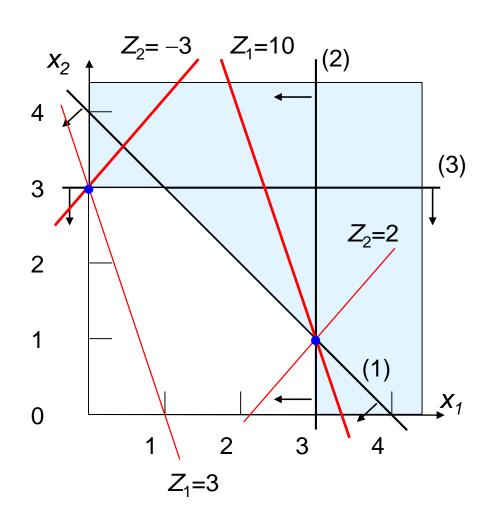
$$Z_2 = -3$$

$$x_1 = 3$$

$$Z_1 = 10$$

$$x_2 = 1$$

$$Z_2 = 2$$



Efficient solutions and efficient points

Maximize: $Z_1 = 3x_1 + x_2$

Minimize: $Z_2 = x_1 - x_2$

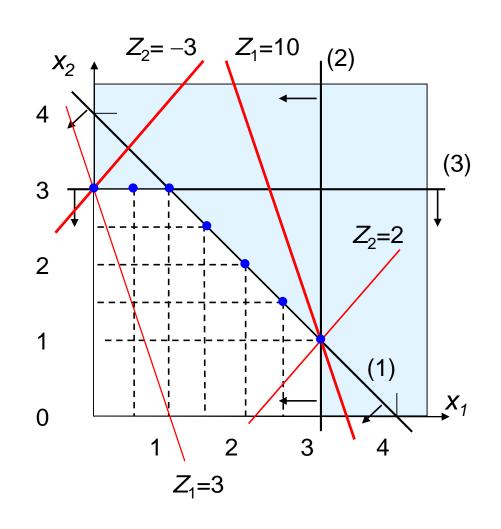
Subject to: $x_1 + x_2 \le 4$ (1)

 $x_1 \le 3 \tag{2}$

 $x_2 \le 3 \tag{3}$

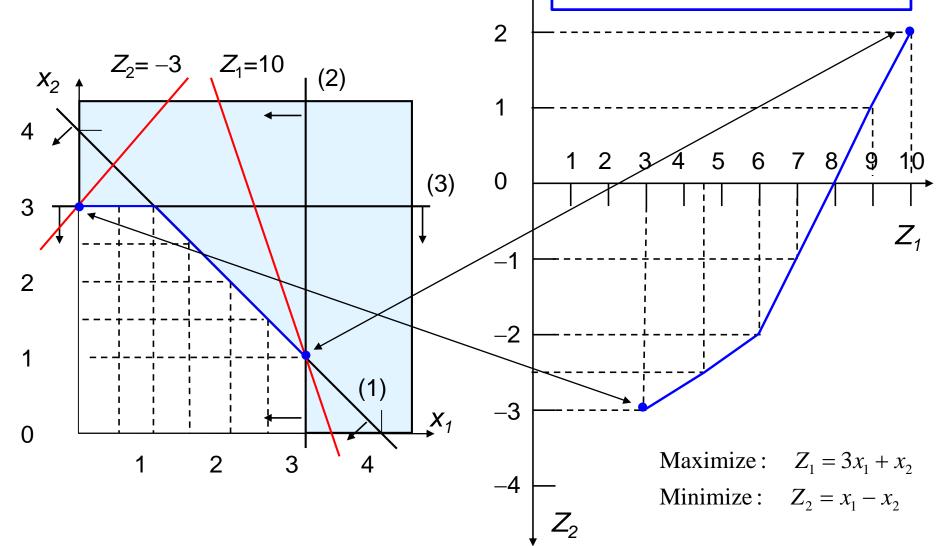
$$x_1, x_2 \ge 0 \tag{4}$$

$ES(x_1,x_2)$		$\text{EP}\left(\mathbf{Z}_{1},\mathbf{Z}_{2}\right)$	
0	3	3	-3
0.5	3	4.5	-2.5
1	3	6	-2
1.5	2.5	7	-1
2	2	8	0
2.5	1.5	9	1
3	1	10	2



Decision Space and Objective Space

Efficient frontier formed by a set of efficient points corresponding to a set of efficient solutions.



Approaches for MOO

A multi-objective optimization (MOO) problem can be expressed as:

Optimize: $F(x) = (f_1(x), f_2(x), ..., f_k(x))$

Subject to: $x \in X$

where F(x) is the objective vector

 $f_i(x)$ is the value of the i^{th} objective

x is the decison vector (set of decision variables)

X is the set of all feasible solutions

 x^* is an efficient solution iff x^* is non - dominated with respect to X

 x^* is also called Pareto - optimal solution

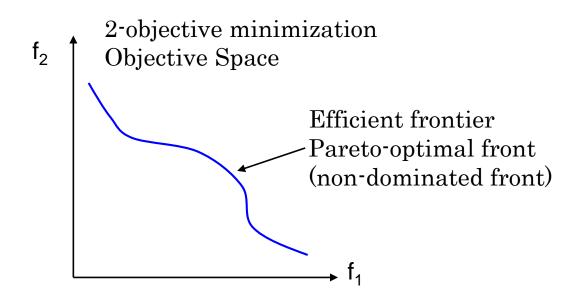
The Pareto optimal set or Efficient frontier is the set of all x^*

How to Combine Decision-Making and Search?

<u>A priori</u> – establish priority between objectives and then search for trade-off solution.

<u>A posteriori</u> – search for a set of efficient solutions and then select trade-off solution.

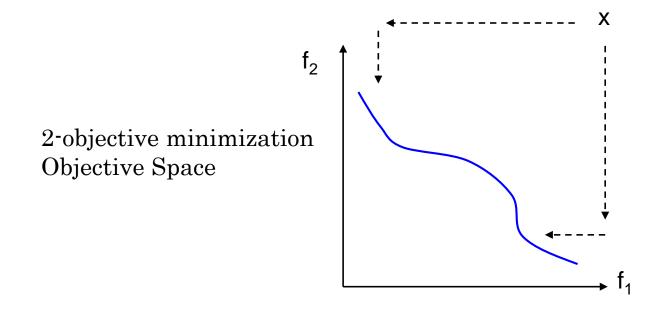
<u>Interactive</u> – decision-making and search are carried out simultaneously.



Lexicographic ordering

- 1. Order the objectives according to priority i=1,2,...,k
- 2. For each objective i in the priority order
 - 2.1 Solve the problem for the objective i
 - 2.2 Add constraint to restrict detriment in objective i

If each stage gives a single-objective optimum then the final solution is an efficient solution.



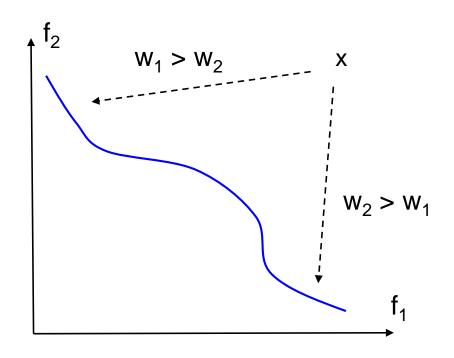
Weighted aggregation

Problem is treated as a single-optimisation problem

$$F(x) = w_1 f_1(x) + w_2 f_2(x) + \dots + w_k f_k(x)$$

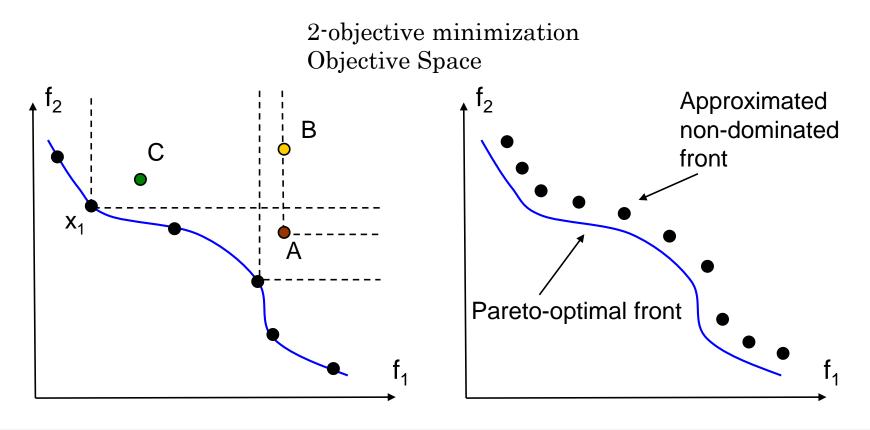
If the weighted aggregation approach gives an optimum then this solution is an efficient solution.

2-objective minimization Objective Space



Pareto optimization

Search for a set of non-dominated solutions (hopefully Pareto optimal). A solution x_1 is non-dominated with respect to a set S if there is no other solution in S that dominates x_1 . The aim is to obtain only efficient solutions.



Example. Use the lexicographic approach to solve the following 2-objective optimization problem.

Maximize:
$$Z_1 = 3x_1 + x_2$$

Minimize:
$$Z_2 = x_1 - x_2$$

Subject to:
$$x_1 + x_2 \le 4$$
 (1)

$$x_1 \le 3 \tag{2}$$

$$x_2 \le 3 \tag{3}$$

$$x_1, x_2 \ge 0 \tag{4}$$

Assuming priority order: Z_1 (higher priority) Z_2 (lower priority)

- •First solve for Z_1 obtaining $x_1 = 3$, $x_2 = 1$, $Z_1 = 10$
- •Then add the constraint $3x_1 + x_2 = 10$
- ·Now solve for Z_2 obtaining $x_1 = 3$, $x_2 = 1$, $Z_2 = 2$
- •Then the solution vector (3,1) is an efficient solution and the point (10,2) is an efficient point
- Reversing the priority and following the same procedure gives the efficient solution (0,3) and efficient point (3,-3)

Example. Use the weighted approach to solve the following 2-objective optimization problem and obtain a Pareto optimal front with 3 solutions.

Maximize: $Z_1 = 3x_1 + x_2$

Minimize: $Z_2 = x_1 - x_2$

Subject to: $x_1 + x_2 \le 4$ (1)

$$x_1 \le 3 \tag{2}$$

$$x_2 \le 3 \tag{3}$$

$$x_1, x_2 \ge 0 \tag{4}$$

Replace the two objective functions by:

 $Z = w_1(3x_1+x_2) - w_2(x_1-x_2)$

Then solve for 3 combinations of weights, for example:

 $(w_1=10, w_2=1), (w_1=4, w_2=10), (w_1=1, w_2=10)$

This gives the efficient solutions and efficient points:

Solution: (3,1) Point: (10,2)

Solution: (1,3) Point: (6,-2)

Solution: (0,3) Point: (3,-3)

Goal Programming

Given a MOO problem, the goal programming approach consists in transforming the multi-objective formulation to a single-objective model by <u>setting goals for each objective and minimizing the deviation</u> from these goals.

- 1. Establish a goal or target value for each $f_i(x)$ of the k objectives in the problem
- 2. Convert the objective expressions to constraints using the goal for each objective
- 3. Introduce a deficiency variables d_i for each of the k objectives to model the deviation from the goal
- 4. Minimize the sum of deviation variables and solve as a single-objective problem

Example. Goal programming for a MOO problem.

Maximize: $Z_1 = 3x_1 + x_2$

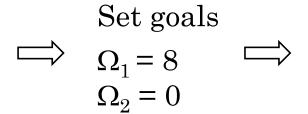
Minimize: $Z_2 = x_1 - x_2$

Subject to: $x_1 + x_2 \le 4$ (1)

$$x_1 \le 3 \tag{2}$$

$$x_2 \le 3 \tag{3}$$

$$x_1, x_2 \ge 0 \tag{4}$$



Note: the deficiency variables are introduced with positive or negative sign (or both) depending on the inequality or equality in the constraint.

Minimize: $Z = d_1 + d_2$

Subject to:

$$3x_1 + x_2 + d_1 \ge 8$$
 (0a)

$$x_1 - x_2 - d_2 \le 0$$
 (0b)

$$x_1 + x_2 \le 4 \tag{1}$$

$$x_1 \le 3 \tag{2}$$

$$x_2 \le 3 \tag{3}$$

$$x_1, x_2, d_1, d_2 \ge 0$$
 (4)

Subject to:

$$3x_1 + x_2 \ge 8$$
 (0a)

$$x_1 - x_2 \le 0 \quad (0b)$$

$$x_1 + x_2 \le 4$$
 (1)

$$x_1 \le 3 \tag{2}$$

$$x_2 \le 3 \tag{3}$$

$$x_1, x_2 \ge 0 \tag{4}$$



Introduce deficiency variables d_1 and d_2

Example. Formulate the following MOO problem as a Goal Programming model. Assume that the decision maker has set these goals: at least Ω_1 for Z_1 , at most Ω_2 for Z_2 , at most Ω_3 for Z_3 , Z_4 equal to Ω_4 .

Maximize:
$$Z_1 = 4x_2 + 4.5x_3 + 5.5x_4 + 7x_5$$

Minimize:
$$Z_2 = 5x_1 + 4x_2 + 7.5x_3 + 5x_4 + x_5$$

Minimize:
$$Z_3 = (x_1 + x_3 + x_5)/10$$

Maximize:
$$Z_4 = 10x_1 + 2x_4 + 3x_5$$

Subject to:
$$x_1 + x_2 + x_3 = 140$$
 (1)

$$x_2 + 9x_3 + 6x_4 - x_5 = 0 (2)$$

$$x_1 - 0.5x_4 + 0.7x_5 \le 30 \tag{3}$$

$$x_1, x_2, x_3, x_4, x_5 \ge 0$$
 (4)

Example (cont.)

Minimize:
$$d_1 + d_2 + d_3 + d_4 + d_5$$

Subject to:
$$4x_2 + 4.5x_3 + 5.5x_4 + 7x_5 + d_1 \ge \Omega_1$$
 (0a)

$$5x_1 + 4x_2 + 7.5x_3 + 5x_4 + x_5 - d_2 \le \Omega_2$$
 (0b)

$$(x_1 + x_3 + x_5)/10 - d_3 \le \Omega_3 \tag{0c}$$

$$10x_1 + 2x_4 + 3x_5 + d_4 - d_5 = \Omega_4 \tag{0d}$$

$$x_1 + x_2 + x_3 = 140 \tag{1}$$

$$x_2 + 9x_3 + 6x_4 - x_5 = 0 (2)$$

$$x_1 - 0.5x_4 + 0.7x_5 \le 30 \tag{3}$$

$$x_1, x_2, x_3, x_4, x_5 \ge 0 \tag{4}$$

$$d_1, d_2, d_3, d_4, d_5 \ge 0 \tag{5}$$