

Lecture: Dr. Michael Stiglmayr**Exercises:** David WillemsDownload this sheet at <http://uni-ko-ld.de/io>**Exercise 5.1**

Consider finding a compromise solution by maximizing the distance to the nadir point.

- a) Let
- $\|\cdot\|$
- be a norm. Show that an optimal solution of the problem

$$\begin{aligned} \max_{x \in X} \quad & \|f(x) - y^N\| \\ \text{s. t.} \quad & f_k(x) \leq y_k^N \text{ for } k = 1, \dots, p \end{aligned} \quad (1)$$

is weakly efficient. Give a condition under which an optimal solution of (1) is efficient.

- b) Another probability is to solve

$$\begin{aligned} \max_{x \in X} \quad & \min_{k=1, \dots, p} |f_k(x) - y_k^N| \\ \text{s. t.} \quad & f_k(x) \leq y_k^N \text{ for } k = 1, \dots, p \end{aligned} \quad (2)$$

Prove that an optimal solution of (2) is weakly efficient.

Exercise 5.2

Consider three points in the Euclidean plane, $x^1 = (1, 1)^T$, $x^2 = (1, 4)^T$ and $x^3 = (4, 4)^T$. The l_2^2 -location problem is to find a point $x = (x_1, x_2) \in \mathbb{R}^2$ such that the sum of weighted squared distances from x to the three points x^i , $i = 1, 2, 3$ is minimal. We consider a bicriterion l_2^2 -location problem, i.e. two weights for each of the points x^i are given through two weight vectors $w^1 = (1, 1, 1)$ and $w^2 = (2, 1, 4)$. The two objectives measuring weighted distances are given by

$$f_k(x) = \sum_{i=1}^3 w_i^k \left((x_1^i - x_1)^2 + (x_2^i - x_2)^2 \right) \quad k = 1, 2$$

Use Definition 1 and Theorem 2 (you do not need to prove correctness of Theorem 2!) to check the points $x^1 = (2, 2)$ and $x^2 = (2, 3)$ for Pareto optimality.

Definition 1: Let $X \subseteq \mathbb{R}^n$, $f: X \rightarrow \mathbb{R}$ and $\hat{x} \in X$.

$$\mathcal{L}_{\leq}(f(\hat{x})) := \{x \in X: f(x) \leq f(\hat{x})\}$$

is called the *level set* of f at $\hat{x}$ and

$$\mathcal{L}_{=}(f(\hat{x})) := \{x \in X: f(x) = f(\hat{x})\}$$

is called the *level curve* of f at $\hat{x}$.

Theorem 2: Let $\hat{x} \in X$ be a feasible solution of a multiobjective optimization problem with $p \geq 2$ objective functions and define $\hat{y}_k := f_k(\hat{x})$, $k = 1, \dots, p$. Then $\hat{x}$ is efficient if and only if

$$\bigcap_{k=1}^p \mathcal{L}_{\leq}(\hat{y}_k) = \bigcap_{k=1}^p \mathcal{L}_{=}(\hat{y}_k).$$

Exercise 5.3

Let

$$X := \left\{ (x_1, x_2) \in \mathbb{R}^2 : -x_1 \leq 0, \quad -x_2 \leq 0, \quad (x_1 - 1)^3 + x_2 \leq 0 \right\},$$

$$f_1(x) = -3x_1 - 2x_2 + 3 \text{ and}$$

$$f_2(x) = -x_1 - 3x_2 + 1.$$

Graph X and $Y = f(X)$. Show that $\hat{x} = (1, 0)$ is properly efficient.

Exercise 5.4

In optimization, totally unimodular matrices (see Definition 3 below) are of great interest, since they give a quick way to verify that a linear program is integral (has an integral optimum, when any optimum exists). Specifically, if A is TU and the right hand side b is integral, then linear programs of forms like $\{\min c^T x \mid Ax \leq b, x \geq 0\}$ or $\{\max c^T x \mid Ax \leq b\}$ have integral optima, for any $c \in \mathbb{R}^n$. Hence if A is totally unimodular and b is integral, every extreme point of the feasible region (e.g. $\{x \in \mathbb{R}^n \mid Ax \leq b\}$) is integral and thus the feasible region is an integral polyhedron.

Definition 3 (Total unimodularity): A matrix $A \in \mathbb{R}^{m \times n}$ is *totally unimodular* (TU), if every square submatrix of A has a determinant of 0, +1 or -1.

Consider the bicriterion linear integer problem

$$\begin{aligned} \min \quad & \begin{matrix} c^T x \\ e^T x \end{matrix} \\ \text{s. t.} \quad & Ax \leq b \end{aligned} \tag{3}$$

with a TU matrix A given in (5) and $e = (1, 1, \dots, 1) \in \mathbb{R}^n$ the vector of all ones. Show that the constraint matrix $\tilde{A}$ of the scalarized problem (4)

$$\begin{aligned} \min \quad & c^T x \\ \text{s. t.} \quad & ex \leq \varepsilon \\ & Ax \leq b \end{aligned} \tag{4}$$

is not TU.

Use the following totally unimodular matrix for your thoughts

$$A = \begin{pmatrix} -1 & -1 & 0 & 0 & 0 & 1 \\ 1 & 0 & -1 & -1 & 0 & 0 \\ 0 & 1 & 1 & 0 & -1 & 0 \\ 0 & 0 & 0 & 1 & 1 & -1 \end{pmatrix}. \tag{5}$$

Exercise 5.5 – Presentation Only

Consider the biobjective nonlinear optimization problem

$$\min_{x \in \mathbb{R}^2} \begin{pmatrix} x_1^2 + x_2^3 - 3x_1x_2 + x_1 \\ x_1^3 - 3x_1x_2 + \frac{1}{2}x_2^2 \end{pmatrix}.$$

Determine the ideal and nadir point.

Hint: To determine the (local) minima of a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$, $\nabla f = 0$ (the gradient of f is zero) is a necessary condition and the positive definiteness of the Hessian matrix H_f in the critical points is a sufficient condition.