Computer Science

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Systems Modelling and Analysis

Choose yourself and new technologies

L.19a. Multicriteria decision making

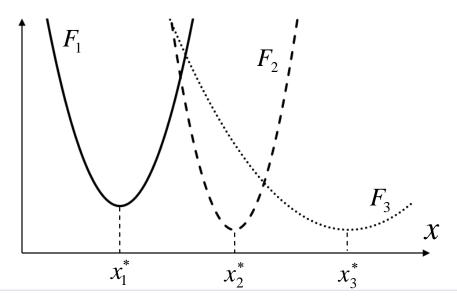






x – decision variables vector

$$F_1(x), F_2(x), \dots, F_K(x)$$
 – performance indices









Synthetic performance index

$$F(x) = H(F_1(x), F_2(x), \dots, F_K(x))$$

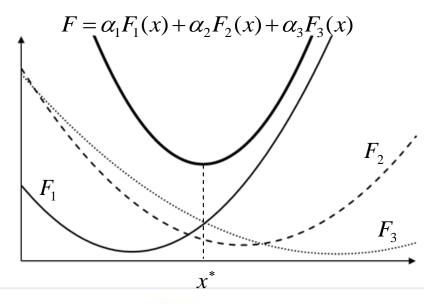
e. g.:
$$F(x) = \sum_{k=1}^{K} \alpha_k F_k(x)$$

where: $\sum_{k=1}^{K} \alpha_k = 1$, $\alpha_k > 0$, k = 1, 2, ..., K

$$F(x) = \prod_{k=1}^{K} F_k(x)$$

$$x^* \to F(x^*) = \min_{x \in D_x} F(x)$$

H(.) – monotonic for all variables









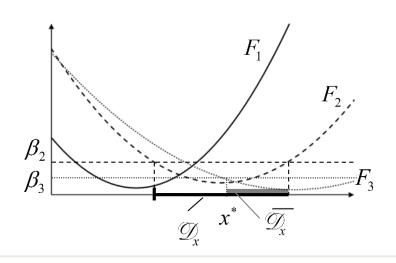
A selected performance index is optimized, Upper limits for values of another performance indices are specified.

Let $F_1(x)$ be a selected performance index

$$F_k(x) \le \beta_k, \quad k = 2, 3, \dots, K$$

Requirements for performance indices are met

$$\overline{D_x} = D_x \cap \left\{ x \in \mathbb{R} \ ^S : F_k(x) \le \beta_k, \ k = 2, \dots, K \right\}$$
$$x^* \to F_1(x^*) = \min_{x \in D_x} F_1(x)$$









Ranked/prioritized performance indices

$$F_1(x) \succ F_2(x) \succ \dots \succ F_K(x) \quad x \in \mathsf{D}_x$$

Step 1.
$$D_{x1} = D_x$$

$$x_1^* \to F_1(x_1^*) = \min_{x \in D_{x_1}} F_1(x)$$

Step 2.
$$D_{x2} = D_{x1} \cap \{x \in \mathbb{R}^{-S} : F_1(x) \le F_1(x_1^*) + \gamma_1 \}$$

$$x_2^* \to F_2(x_2^*) = \min_{x \in D_{x2}} F_2(x)$$



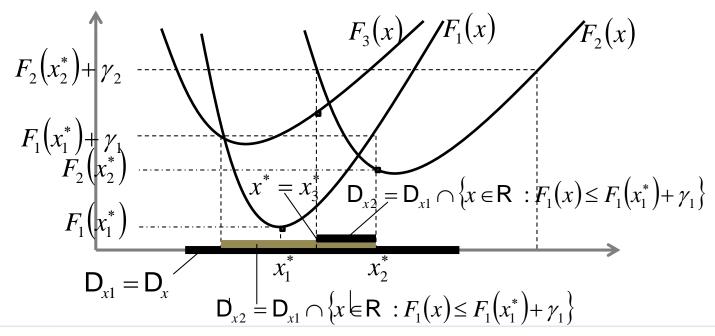






Step K.
$$D_{xK} = D_{xK-1} \cap \{x \in \mathbb{R}^{-S} : F_{K-1}(x) \leq F_1(x_{K-1}^*) + \gamma_{K-1} \}$$

 $x_K^* = x_K^* \to F_K(x_K^*) = \min_{x \in D_{xK}} F_K(x)$









Pareto-optimality

$$F_1(x), F_2(x), ..., F_K(x) \quad x \in D_x$$

 D_K – a set of non-dominated solutions

A set of non-dominated solutions $x \in D_x$ is a subset of such points $D_K \subseteq D_x$ for which performance indices has the following property: any of performance indices may decrease only if at least one of the remaining performance indices increases.

$$x_1, x_2 \in D_K \Leftrightarrow \forall i \in \{1, 2, ..., K\} \exists j \in \{1, 2, ..., K\}$$

$$F_i(x_1) < F_j(x_2) \Rightarrow F_j(x_1) < F_i(x_2)$$

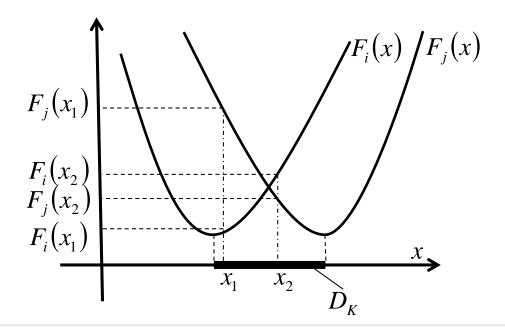






$$x_1, x_2 \in D_K \Leftrightarrow \forall j \in \{1, 2, ..., K\} \exists i \in \{1, 2, ..., K\}$$

 $F_j(x_1) > F_j(x_2) \Rightarrow F_i(x_1) < F_i(x_2)$









Analytical condition

$$\sum_{k=1}^{K} \eta_k \nabla_x F_k(x) = 0_S \quad \eta_k \ge 0 \quad k = 1, 2, ..., K$$

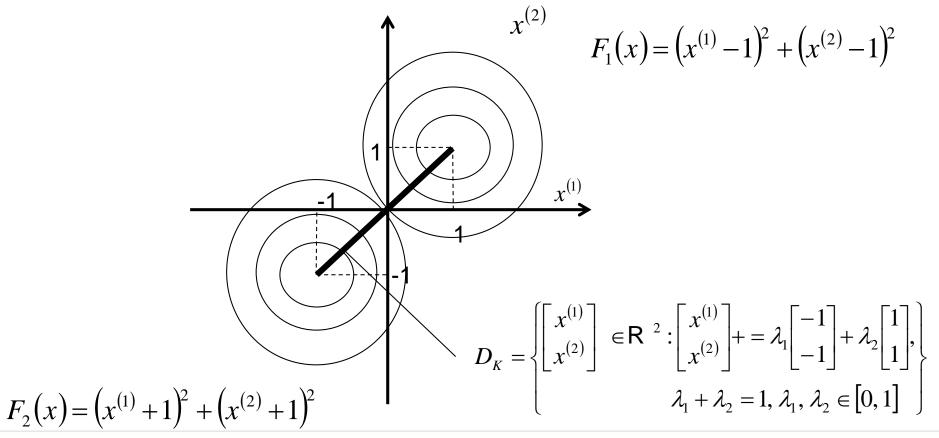
Example:

$$\begin{split} F_{1}(x) &= \left(x^{(1)} - 1\right)^{2} + \left(x^{(2)} - 1\right)^{2}, \quad F_{2}(x) = \left(x^{(1)} + 1\right)^{2} + \left(x^{(2)} + 1\right)^{2} \\ \eta_{1} \begin{bmatrix} 2(x^{(1)} - 1) \\ 2(x^{(2)} - 1) \end{bmatrix} + \eta_{2} \begin{bmatrix} 2(x^{(1)} + 1) \\ 2(x^{(2)} + 1) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad /2 \quad \eta_{1}, \, \eta_{2} \geq 0 \\ \eta_{1} \begin{bmatrix} x^{(1)} \\ x^{(2)} \end{bmatrix} + \eta_{2} \begin{bmatrix} x^{(1)} \\ x^{(2)} \end{bmatrix} = \eta_{1} \begin{bmatrix} -1 \\ -1 \end{bmatrix} + \eta_{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad /(\eta_{1} + \eta_{2}) \\ \frac{\eta_{1}}{\eta_{1} + \eta_{2}} \begin{bmatrix} x^{(1)} \\ x^{(2)} \end{bmatrix} + \frac{\eta_{2}}{\eta_{1} + \eta_{2}} \begin{bmatrix} x^{(1)} \\ x^{(2)} \end{bmatrix} = \frac{\eta_{1}}{\eta_{1} + \eta_{2}} \begin{bmatrix} -1 \\ -1 \end{bmatrix} + \frac{\eta_{2}}{\eta_{1} + \eta_{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\ \begin{bmatrix} x^{(1)} \\ x^{(2)} \end{bmatrix} + 2\lambda_{1} \begin{bmatrix} -1 \\ -1 \end{bmatrix} + \lambda_{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \, gdzie \, \lambda_{1} = \frac{\eta_{1}}{\eta_{1} + \eta_{2}}, \, \lambda_{2} = \frac{\eta_{2}}{\eta_{1} + \eta_{2}}, \, \lambda_{1} + \lambda_{2} = 1, \, \lambda_{1}, \, \lambda_{2} \in [0, 1] \end{split}$$















Thank you for attention

