

Двухкритериальная модификация задачи
составления расписаний с критерием
суммарного взвешенного запаздывания и
нечеткими исходными данными

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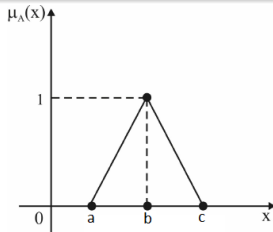
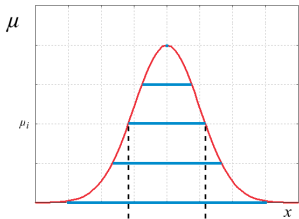
- U be a non-empty set of elements
- Fuzzy set A in U is the set of ordered pairs $\{u, \mu_A(u)\}$ for any $u \in U$
- $\mu_A(\cdot)$ is the membership function, $\mu_A(u) \in [0, 1]$
- $h(A) = \sup_{u \in U} \mu_A(u)$ is the height of a fuzzy set A
- If $h(A) = 1$, then A is called normal
- If $\mu_A(\lambda u + (1 - \lambda)w) \geq \min\{\mu_A(u), \mu_A(w)\}$ for $\forall u, w \in U$, $\lambda \in (0, 1)$, then A in linear space U is convex

Fuzzy Number

A fuzzy number is convex normal fuzzy set in the set of real numbers \mathbb{R} .

Triangular fuzzy number

$$\mu_A(x) = \begin{cases} 0 & \text{if } x < a, \\ \frac{x-a}{b-a} & \text{if } a \leq x < b, \\ \frac{c-x}{c-b} & \text{if } b \leq x < c, \\ 0 & \text{if } x \geq c, \end{cases}$$



General Approach

- Zadeh L A 1965 [Fuzzy sets](#)
- Ekel P, Pedrycz W and Schinzinger R 1998 A general approach to solving a wide class of [fuzzy optimization problems](#)
- Chanas S and Kuchta D 1998 [Discrete fuzzy optimization](#)

Applications

- Slowinski R. and Hapke, M. 2000 [Scheduling Under Fuzziness](#)
- Huang Ch.-S., Huang Y.-Ch. and Lai P.-J. 2012 Modified genetic algorithms for solving [fuzzy flow shop scheduling problems](#) and their implementation with CUDA
- Zhang Y., Hu Q., Meng Z. and Ralescu, A. 2020 [Fuzzy dynamic timetable scheduling](#) for public transit
- Pang W., Wang K.-P., Zhou Ch.-G. and Dong, L.-J. 2004 Fuzzy Discrete Particle Swarm Optimization for Solving [Traveling Salesman Problem](#)
- Tang J., Pan Z., Fung R. Y. K. and Lau H. 2009 [Vehicle routing problem](#) with fuzzy time windows
- Nucci F. 2021 [Multi-shift Single-Vehicle Routing Problem](#) Under Fuzzy Uncertainty

X is the set of feasible solutions

J is the set of parameters indices

$(p_j - \delta_j, p_j, p_j + \varepsilon_j), \varepsilon_j, \delta_j > 0, j \in J$ are the fuzzy input parameters

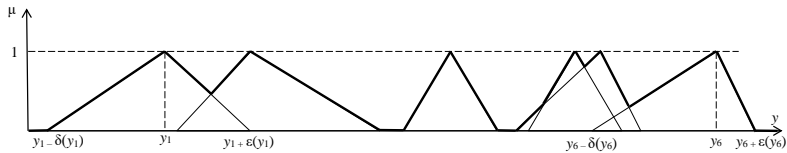
f is objective function (to be maximized, summed value of some set of input parameters values)

$Y = f(X)$ is the set of feasible outcomes

Membership function μ_f of f (principle of Zadeh)

$$\mu_f(y) = \begin{cases} \sup_{x \in f^{-1}(y)} \{\alpha \in [0, 1] \mid \mu(x) = \alpha\}, & \text{if } f^{-1}(y) \neq \emptyset, \\ 0, & \text{otherwise.} \end{cases}$$

Problem Formulation: Example of μ_f



Problem

Y is the the set of feasible alternatives

$$Y = [\min_{x \in X} f_x - \delta_{f_x}, \max_{x \in X} f_x + \varepsilon_{f_x}]$$

$g = (f, \mu_f)$ is the vector criterion,

$$g(y) = (f(y), \mu_f(y)) \text{ for } y \in Y$$

The Pareto set

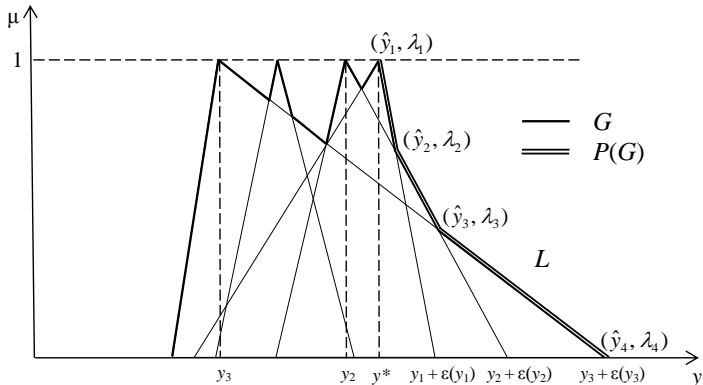
$$P_g(Y) = \{y \in Y \mid \nexists y^* \in Y : g(y^*) \geq g(y)\}$$

$$P(G) = g(P_g(Y))$$

Construction of the Pareto Set

$$G = g(Y)$$

The Pareto set $P(G)$ is the most right polygonal line L , which is constructed by the right sides of TFNs corresponding to the maximum and near-maximum values of the function $f(x)$.



The Pareto Set Reduction

Noghin V.D.: Reduction of the Pareto Set: An Axiomatic Approach (2018)

Bi-criteria choice problem $\langle Y, g, \succ \rangle$

$\langle Y, g = (g_1, g_2) \rangle +$ asymmetric binary preference relation of the decision maker (DM) \succ :

$g(y^*) \succ g(y)$ means that the DM prefers the value y^* to y .

Axioms of “reasonable” choice

- \succ is irreflexive, transitive,
- \succ is invariant with respect to a linear positive transformation,
- \succ is compatible with criteria g_1, g_2 ,
- exclusion axiom.

Elementary information quantum

We say that there is an *elementary information quantum* about the DM's preference relation \succ with given coefficient of compromise $\theta \in (0, 1)$ if vector $z' \in \mathbb{R}^2$ such that $z'_i = 1 - \theta > 0$ and $z'_j = -\theta < 0$ satisfies the expression $z' \succ 0_2$, where $0_2 = (0, 0)$. Here $i, j \in \{1, 2\}, i \neq j$.

$$\begin{pmatrix} 0.8 \\ -0.2 \end{pmatrix} \not\geq \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \begin{pmatrix} 0.8 \\ -0.2 \end{pmatrix} \not\leq \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$
$$\text{Let } \begin{pmatrix} 0.8 \\ -0.2 \end{pmatrix} \succ \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \theta = 0.2$$

The Edgeworth–Pareto Principle

If axioms of “reasonable” choice hold, then for any set of selectable outcomes $C(G) \subseteq P(G)$.

Theorem of the reduction

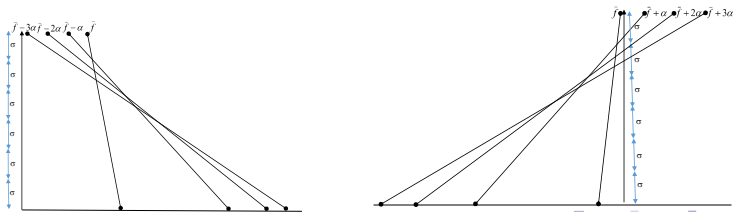
An elementary information quantum is given $f_1 \rightarrow f_2 : \theta$. Then for any set of selectable outcomes $C(G)$ the inclusions $C(G) \subseteq \hat{P}(G) \subseteq P(G)$, $\hat{P}(G) = g(P_{\hat{g}}(G))$, $\hat{g}_1 = g_1, \hat{g}_2 = \theta g_1 + (1 - \theta)g_2$.

The Pareto Set Reduction for Summed Objective and TFN

The Pareto set corresponds to a polygonal line L , which will be the southwestern border of some convex set. The polygonal line L is defined by N lines $g_2 = a^{(i)} - k^{(i)}g_1$, $i = 1, \dots, N$, so that $k^{(1)} \geq k^{(2)} \geq \dots \geq k^{(N)}$.

$g_1 \rightarrow g_2: \theta$ or $g_2 \rightarrow g_1: \theta$

For each value of $\theta \in (0, 1)$ we identify the cardinality of the Pareto set reduction. We specify the subintervals of interval $(0, 1)$, each of them gives some reduction of the Pareto set. Bounds of subintervals are formed by ratios between slope coefficients $k^{(i)}$, $i = 1, \dots, N$.



$$L = \{l^{(i)} \mid g_2 = a^{(i)} - k^{(i)}g_1, i = 1, \dots, p, k^{(1)} \geq k^{(2)} \geq \dots \geq k^{(p)}\}.$$

Theorem

Let $g_1 \rightarrow g_2: \theta$, and the Pareto set is formed by polygonal line L .

1) If $\theta < \frac{k^{(p)}}{k^{(p)}+1}$, then the reduction of the Pareto set $\widehat{P}(G)$ coincides with the Pareto set $P(G)$.

2) If $\frac{k^{(i)}}{k^{(i)}+1} \leq \theta < \frac{k^{(i-1)}}{k^{(i-1)}+1}$, then the reduction of the Pareto set $\widehat{P}(G)$ contains elements of segments $l^{(1)}, \dots, l^{(i-1)}$ and the rightmost element of L , $i = p, \dots, 2$.

3) If $\theta \geq \frac{k^{(1)}}{k^{(1)}+1}$, then the reduction of the Pareto set consists of one element.

Here elements of segment we understand in terms of $[\dots)$, i. e. including the left point and excluding the right point subject to the first coordinate.

Scheduling Problem 1 $|r_j, d_j, w_j| \sum_j w_j T_j$

Input data

n is the number of jobs;

$p_j \geq 0$ is the duration of job j ;

$r_j \geq 0$ is the release date of job j ;

$d_j \geq 0$ is the due date of job j ;

$w_j \geq 0$ is the weight of job j .

Criteria

C_j is the completion time of job j ;

$T_j = \max\{0; C_j - d_j\}$ is the tardiness of job j ;

$U_j = 1$ if $C_j > d_j$ and $U_j = 0$ otherwise;

the total weighted tardiness $\sum_j w_j T_j$;

the weighted number of tardy jobs $\sum_j w_j U_j$.

The problem is NP-hard.

$$x_{ij} = \begin{cases} 1 & \text{if job } j \text{ is placed in position } k, \\ 0 & \text{otherwise.} \end{cases}$$

$$f = \sum_{k=1}^n T_k \rightarrow \min, \quad (1)$$

$$\sum_{j=1}^n x_{jk} = 1, \quad k = 1, \dots, n, \quad (2)$$

$$\sum_{k=1}^n x_{jk} = 1, \quad j = 1, \dots, n, \quad (3)$$

$$T_k \geq \sum_{l=1}^k \sum_{j=1}^n x_{jl} p_j - \sum_{j=1}^n x_{jk} d_j, \quad k = 1, \dots, n, \quad (4)$$

$$x_{jk} \in \{0, 1\}, \quad T_k \geq 0, \quad j, k = 1, \dots, n. \quad (5)$$

$$x_{ij} = \begin{cases} 1 & \text{if job } j \text{ is placed in position } k, \\ 0 & \text{otherwise.} \end{cases}$$

$$f = \sum_{j=1}^n w_j T_j \rightarrow \min, \quad (6)$$

$$\sum_{j=1}^n x_{jk} = 1, \quad k = 1, \dots, n, \quad (7)$$

$$\sum_{k=1}^n x_{jk} = 1, \quad j = 1, \dots, n, \quad (8)$$

$$T_k \geq \sum_{l=1}^k \sum_{j=1}^n x_{jl} p_j - \sum_{j=1}^n x_{jk} d_j, \quad k = 1, \dots, n, \quad (9)$$

$$T_j \geq T_k - T_{\max}(1 - x_{jk}), \quad j, k = 1, \dots, n, \quad (10)$$

$$x_{jk} \in \{0, 1\}, \quad T_k \geq 0, \quad T_j \geq 0, \quad j, k = 1, \dots, n. \quad (11)$$

$$x_{ij} = \begin{cases} 1 & \text{if job } i \text{ immediately precedes job } j, \\ 0 & \text{otherwise.} \end{cases}$$

$$y_{ij} = \begin{cases} 1 & \text{if job } i \text{ precedes job } j, \\ 0 & \text{otherwise.} \end{cases}$$

$$f = \sum_{j=1}^n w_j T_j \rightarrow \min, \quad (12)$$

$$\sum_{j=1}^{n+1} x_{ij} = 1, \quad i = 0, \dots, n, \quad (13)$$

$$\sum_{i=0}^n x_{ij} = 1, \quad j = 1, \dots, n+1, \quad (14)$$

$$\sum_{i=1}^n y_{iq} - \sum_{i=1}^n y_{ik} + (n+2) \leq (n+1), \quad q, k = 1, \dots, n, \quad (15)$$

$$T_j \geq \sum_{i=0}^n y_{ij} p_i + p_j - d_j, \quad j = 1, \dots, n, \quad (16)$$

$$x_{ij} \in \{0, 1\}, \quad y_{ij} \in \{0, 1\}, \quad T_j \geq 0, \quad i, j = 1, \dots, n. \quad (17)$$

Fuzzy Input Data

$(w_j - \delta_j, w_j, w_j + \varepsilon_j)$ is Triangular Fuzzy Number (TFN).

$$\mu_{w_j}(w) = \begin{cases} 0 & \text{if } w < w_j - \delta_j, \\ \frac{w - w_j + \delta_j}{\delta_j} & \text{if } w_j - \delta_j \leq w < w_j, \\ \frac{w_j + \varepsilon_j - w}{\varepsilon_j} & \text{if } w_j \leq w < w_j + \varepsilon_j, \\ 0 & \text{if } w \geq w_j + \varepsilon_j, \end{cases}$$

Membership Function

$$\mu_x(y) = \begin{cases} 0 & \text{if } y < f_x - \delta_{f_x}, \\ \frac{y - f_x + \delta_{f_x}}{\delta_{f_x}} & \text{if } f_x - \delta_{f_x} \leq y < f_x, \\ \frac{f_x + \varepsilon_{f_x} - y}{\varepsilon_{f_x}} & \text{if } f_x \leq y < f_x + \varepsilon_{f_x}, \\ 0 & \text{if } y \geq f_x + \varepsilon_{f_x}, \end{cases}$$

where $f_x = \sum_{j=1}^n w_j T_j$, $\delta_{f_x} = \sum_{j=1}^n \delta_j T_j$, $\varepsilon_{f_x} = \sum_{j=1}^n \varepsilon_j T_j$.

Membership Function

$$\mu_f(y) = \sup_{x \in X} \{\alpha \in [0, 1] \mid \mu_x(y) = \alpha\}$$

Two Criteria

Objective function and its membership function represent the criteria of crisp problem, i.e. we have vector criterion $g = (f, \mu_f)$, such that $g(y) = (y, \mu_f(y))$ for all $y \in [\min_{x \in X} (f_x - \delta_{f_x}), \max_{x \in X} (f_x + \varepsilon_{f_x})]$.

Approximation of the Pareto Set

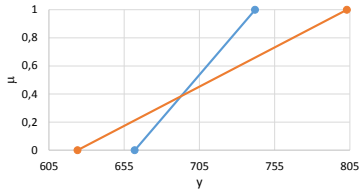
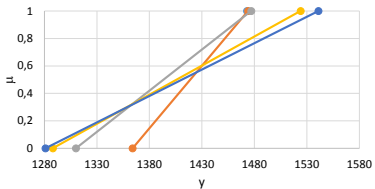
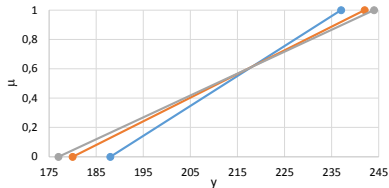
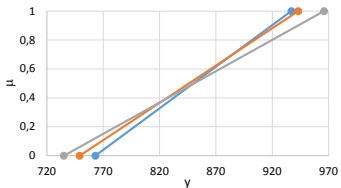
- 1: Find optimal solution of model (6)-(11) for TWTP. Let f_1^* denote the optimal objective value, and triplet $(f_1^* - \delta_1^*, f_1^*, f_1^* + \varepsilon_1^*)$ is the corresponding TFN. Put the current iteration number $iter = 1$.
- 2: Repeat until $iter \leq m_{\max}$.
 - 2.1 Put $iter = iter + 1$.
 - 2.2 Find optimal solution of model (6)-(11) for TWTP with additional constraint $f \geq f_{iter-1}^* + 1$.
 - 2.3 Let f_{iter}^* denote the obtained optimal objective value, and triplet $(f_{iter}^* - \delta_{iter}^*, f_{iter}^*, f_{iter}^* + \varepsilon_{iter}^*)$ is the corresponding TFN.
- 3: Find an approximation of the Pareto set from $\{(f_1^* - \delta_1^*, f_1^*, f_1^* + \varepsilon_1^*), \dots, (f_{m_{\max}}^* - \delta_{m_{\max}}^*, f_{m_{\max}}^*, f_{m_{\max}}^* + \varepsilon_{m_{\max}}^*)\}$.

TWTP instances from OR-Library ($n = 40$, $n = 50$).

Values $\delta_j = \varepsilon_j$ are generated randomly with uniform distribution from interval $[\frac{w_{aver}}{8}, \frac{w_{aver}}{4})$, where w_{aver} is the average weight.

Series	n	Approximation		
		min	aver	max
Weighted tardiness	40	3	5	10
Weighted tardiness	50	4	6	12

TFNs giving the Pareto set approximation



The Pareto set consists of N TFNs:

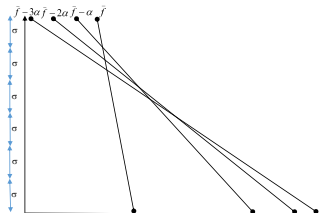
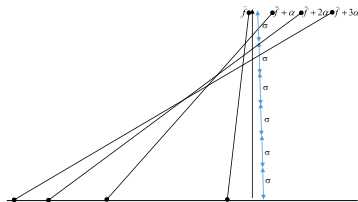
$$(\bar{f} - \Delta\bar{f}, \bar{f}, \bar{f} + \Delta\bar{f}),$$

$$\left((\bar{f} + i\theta) - \left(\Delta\bar{f} + \frac{\theta_{i-1}}{i\sigma} + \frac{\theta}{\sigma} \right), \bar{f} + i\theta, (\bar{f} + i\theta) + \left(\Delta\bar{f} + \frac{\theta_{i-1}}{i\sigma} + \frac{\theta}{\sigma} \right) \right),$$

$$i = 1, \dots, N - 1,$$

where $N = \lceil \frac{1}{\sigma} \rceil$, $0 < \sigma < 1$,

$\theta_0 = 0$, $\theta_i = \theta_{i-1} \left(1 + \frac{1}{i} \right) + \theta$, $i = 1, \dots, N - 1$.



We select σ such that $N = \lceil \frac{1}{\sigma} \rceil = n$.

p_j are arbitrary crisp processing times

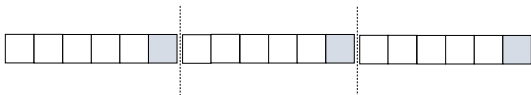
$d_j = d = \sum_{i=1}^n p_i - \min_{i=1, \dots, n} p_i$ is the common crisp due date

\tilde{w}_j are fuzzy numbers represented weights of the form

$(w_j - \delta_j, w_j, w_j + \delta_j)$, where

$w_j = \bar{f} + j\theta$ for $j = 0, \dots, N - 1$.

$\delta_j = \Delta \bar{f} + \frac{\theta}{\sigma} + \frac{\theta_{j-1}}{j\sigma}$, for $j = 1, \dots, N - 1$, $\delta_0 = \Delta \bar{f}$.



- Fuzzy version of the single-machine scheduling problem is formulated as bi-objective one, where the total cost (main objective) and its membership function are optimized.
- An approximation of the Pareto set is constructed by the greedy algorithm using MIP-solver.
- Experimental evaluation on instances from TSPLIB library shows that the approximate set of the non-dominated solutions has a very low cardinality, which is appealing from the practical point of view.
- A series of instances with special structure of the Pareto set is constructed.

Thank you for your attention!