

Methodology

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Introduction

In this project, we study [JTMW99].

The Weighted Quadratic Approach

In the weighted quadratic approach a quadratic function of the objective functions

$$\begin{aligned} & \text{minimize} && \mathbf{f}(\mathbf{x})^\top \mathbf{Q} \mathbf{f}(\mathbf{x}) + \mathbf{q}^\top \mathbf{f}(\mathbf{x}) \\ & \text{subject to} && \mathbf{x} \in X \end{aligned} \tag{1}$$

where \mathbf{Q} is a $p \times p$ matrix and where \mathbf{q} is a vector in \mathbb{R}^p .

Theorem 0.1. *Under conditions of a quadratic Lagrangian duality, if $\hat{\mathbf{x}} \in X$ is efficient, then there exist a symmetric $p \times p$ matrix \mathbf{Q} and a vector $\mathbf{q} \in \mathbb{R}^p$ such that $\hat{\mathbf{x}}$ is an optimal solution to (1)*

MOP Formulation

Let $\mathbf{f}: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a vector-valued objective function. Thus we have

$$\mathbf{f}(\mathbf{x}) = (f_1(\mathbf{x}), \dots, f_m(\mathbf{x}))^\top$$

for all $\mathbf{x} \in \mathbb{R}^n$, where $f_k: \mathbb{R}^n \rightarrow \mathbb{R}$ are the component objective functions of \mathbf{f} for $1 \leq k \leq m$. Next, let $\mathbf{h}: \mathbb{R}^n \rightarrow \mathbb{R}^p$ be a vector-valued constraint function. Thus we have

$$\mathbf{h}(\mathbf{x}) = (h_1(\mathbf{x}), \dots, h_p(\mathbf{x}))^\top$$

for all $\mathbf{x} \in \mathbb{R}^n$, where $h_j: \mathbb{R}^n \rightarrow \mathbb{R}$ are the component constraint functions of \mathbf{h} for $1 \leq j \leq p$. Finally, let

$$X = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{h}(\mathbf{x}) = 0\}.$$

We consider the following multi objective program (MOP) below:

$$\begin{aligned} & \text{maximize} && \mathbf{f}(\mathbf{x}) \\ & \text{subject to} && \mathbf{x} \in X \end{aligned} \tag{2}$$

A point $\mathbf{x}^0 \in X$ is called an efficient solution of this MOP if there is no other point $\mathbf{x} \in X$ such that $\mathbf{f}(\mathbf{x}) \geq \mathbf{f}(\mathbf{x}^0)$ with strict inequality holding for at least one component (i.e. $f_k(\mathbf{x}) > f_k(\mathbf{x}^0)$ for some $1 \leq k \leq m$).

Tchebycheff Approach

We first use the Tchebycheff approach to find the efficient solutions to (2). Let $\boldsymbol{\lambda} \in \mathbb{R}^m$ be a weight vector and let $\mathbf{z}^* \in \mathbb{R}^m$ be the ideal point whose i -th component is given by $z_i^* = \max_{\mathbf{x} \in X} f_i(\mathbf{x})$. Now consider the problem

$$\min_{\mathbf{x} \in X} \max_{1 \leq i \leq m} \{\lambda_i (z_i^* - f_i(\mathbf{x}))\} \tag{3}$$

where $\lambda_i \geq 0$ and $\sum_{i=1}^m \lambda_i = 1$. All efficient solutions of (2) can be found as optimal solutions of (3) by adjusting the λ -values.

Primal Problem

We can rewrite (5) as

$$\begin{aligned} &\text{minimize} && \alpha \\ &\text{subject to} && \mathbf{g}(\mathbf{x}, \alpha) = 0 \\ &&& \mathbf{x} \in X \end{aligned} \tag{4}$$

where we set

$$\mathbf{g}(\mathbf{x}, \alpha) = \boldsymbol{\lambda}^\top (\mathbf{z}^* - \mathbf{f}(\mathbf{x})) - \alpha.$$

Next, let A be the $m \times m$ diagonal matrix whose diagonal entries are all equal to a , and define the augmented Lagrange function:

$$L_Q(\mathbf{x}, \alpha, a, \mathbf{y}) = \alpha + \mathbf{g}(\mathbf{x}, \alpha)^\top A \mathbf{g}(\mathbf{x}, \alpha) + \mathbf{y}^\top \mathbf{g}(\mathbf{x}, \alpha),$$

and the following dual program:

$$\max_{a>0, \mathbf{y}} \min_{\mathbf{x} \in X, \alpha} L_Q(\mathbf{x}, \alpha, a, \mathbf{y}). \tag{5}$$

Then subject to certain conditions, program (??) has an optimal solution (\mathbf{x}, α) if and only if its dual (??) has an optimal solution (a, \mathbf{y}) , and in this case the objective values of both programs are equal. Furthermore, $L_Q(\mathbf{x}, \alpha, a, \mathbf{y})$ has a saddle point in the primal variables (\mathbf{x}, α) and the dual variables (a, \mathbf{y}) .

Main Result

Theorem 0.2.

Conclusion

References

References

[JTMW99] JØRGEN TIND1 and MARGARET M. WIECEK. “Augmented Lagrangian and Tchebycheff Approaches in Multiple Objective Programming”. In: Journal of Global Optimization 14: 251–266, 1999