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To cite this article: Shuning Wang & Jianshe Dai (1999) Computation Of  $I_1$ , Central Estimator Of Linear Parameters, International Journal of Modelling and Simulation, 19:1, 85-88, DOI: [10.1080/02286203.1999.11759983](https://doi.org/10.1080/02286203.1999.11759983)

To link to this article: <https://doi.org/10.1080/02286203.1999.11759983>



Published online: 01 Sep 2016.



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# COMPUTATION OF $l_1$ CENTRAL ESTIMATOR OF LINEAR PARAMETERS

Shuning Wang\* and Jianshe Dai\*\*

## Abstract

The authors propose a new algorithm for computing  $l_1$  central estimators of linear parameters. A theorem is given saying that such an estimator can be computed by solving a group of linear programming problems with a common feasible region. An overall simplex algorithm for solving such a group of linear programming problems is suggested, which may avoid repeated search of any vertex of the common feasible region. These results may effectively reduce the computation of  $l_1$  central estimators of linear parameters.

## Key Words

Parameter estimation, central estimator, unknown but bounded errors, set membership uncertainty, linear programming

## 1. Introduction

Over the past several years a nonprobabilistic approach has attracted a great deal of attention in the field of parameter estimation, which is usually called the set membership uncertainty (SMU) or unknown but bounded error (UBBE) approach [1-3]. With this approach a central estimator is regarded as an optimal estimator. Its definition can be simply explained as follows. Consider a problem of parameter estimation described by the linear model:

$$y = \Phi\theta^* + \rho$$

where  $y \in R^N$  and  $\Phi \in R^{N \times n}$  are the known vector and matrix,  $\theta^* \in R^n$  is a parameter vector to be estimated, and  $\rho \in R^N$  is an unknown error vector. With UBBE approach,  $\rho$  is assumed to satisfy the constraint

$$\|\rho\| \leq \varepsilon$$

where  $\varepsilon$  is a given positive number. This inequality ensures that  $\theta^*$  belongs to the feasible parameter set:

$$\Theta = \{\theta, \theta \in R^n \mid \text{s.t. } \|y - \Phi\theta\| \leq \varepsilon\}$$

According to this assumption, the following inequality for

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(paper no. 1994-10)

any  $n$ -dimensional vector  $\hat{\theta}$  holds:

$$\|\theta^* - \hat{\theta}\| \leq \max_{\theta \in \Theta} \|\theta - \hat{\theta}\|$$

In this case, it is natural to determine an estimate of  $\theta^*$  by minimizing the right side of this inequality. This leads to the definition of a central estimator  $\hat{\theta}^c$ , that is:

$$\max_{\theta \in \Theta} \|\theta - \hat{\theta}^c\| = \min_{\hat{\theta} \in R^n} \max_{\theta \in \Theta} \|\theta - \hat{\theta}\|$$

In most literature on the UBBE approach, the norm of the model error is taken as  $l_\infty^w$ , that is:

$$\|\rho\|_\infty^w = \max_{1 \leq i \leq N} w_i |\rho_i|$$

where  $w_i$  is a component of the vector  $w \in R^N$ , which is a given positive weight coefficient,<sup>1</sup> and the norm of the estimation error is mainly taken as  $l_\infty$ , that is:

$$\|\theta - \hat{\theta}\|_\infty = \max_{1 \leq i \leq n} |\theta_i - \hat{\theta}_i|$$

Under this condition the central estimator  $\hat{\theta}^{c\infty}$  and its corresponding maximal estimation error can be computed with the formulas [2]:

$$\hat{\theta}_i^{c\infty} = 0.5 \times (\theta_i^{max} + \theta_i^{min}), \quad i = 1, \dots, n$$

$$\max_{\theta \in \Theta} \|\theta - \hat{\theta}^{c\infty}\|_\infty = \max_{1 \leq i \leq n} 0.5 \times (\theta_i^{max} - \theta_i^{min})$$

where all  $\theta_i^{max}$  and  $\theta_i^{min}$  can be obtained by solving the following  $2 \times n$  linear programming problems:

$$\theta_i^{max} = \max_{\theta \in \Theta} \theta_i, \quad \theta_i^{min} = \min_{\theta \in \Theta} \theta_i, \quad i = 1, \dots, n \quad (1)$$

Notice that all these linear programming problems have a common feasible region.

In recent years, because of the development of  $l_1$  robust control methodologies,  $l_1$  central estimation problem has been considered by some authors [4, 5], which means that the norm of the estimation error is taken as  $l_1$ , that is:

<sup>1</sup> In this work, if a symbol stands for a vector then the same symbol with a subscript will always denote a component of this vector.

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