

# Linear Minimax Regret Estimation with Bounded Data Uncertainties

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## Abstract

We develop a new linear estimator for estimating an unknown vector of parameters  $\mathbf{x}$  in a linear model, in the presence of bounded data uncertainties. The estimator is designed to minimize the worst-case *regret* over all bounded data vectors, namely the worst-case difference between the MSE attainable using a linear estimator that does not know the true parameters  $\mathbf{x}$ , and the optimal MSE attained using a linear estimator that knows  $\mathbf{x}$ . We demonstrate through several examples that the minimax regret estimator can significantly increase the performance over the conventional least-squares estimator, as well as several other least-squares alternatives.

## 1 Introduction

The problem of estimating a vector of unknown parameters  $\mathbf{x}$  from noisy observations  $\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w}$ , where  $\mathbf{H}$  is a known matrix and  $\mathbf{w}$  is a noise vector, arises in many different fields in science and engineering, and consequently attracted much attention in the estimation literature.

If the unknown parameters  $\mathbf{x}$  are assumed to be random variables with known second-order statistics, then the linear estimator minimizing the mean-squared error (MSE) is the well-known Wiener estimator [1, 2]. However, in many problems of practical interest there is no statistical information available on  $\mathbf{x}$ , so that  $\mathbf{x}$  is treated as an unknown set of deterministic parameters. In this case, the MSE of an estimator  $\hat{\mathbf{x}}$  of  $\mathbf{x}$  depends explicitly on the unknown parameters  $\mathbf{x}$ , and therefore cannot be minimized directly.

Since the MSE between  $\hat{\mathbf{x}}$  and  $\mathbf{x}$  depends on  $\mathbf{x}$ , a common approach is to seek estimators that minimize some function of the data error  $\hat{\mathbf{y}} - \mathbf{y}$ , where  $\hat{\mathbf{y}} = \mathbf{H}\hat{\mathbf{x}}$  is the estimated data vector. The celebrated least-

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