



Sharp oracle inequalities for low-complexity priors

Tung Duy Luu¹ · Jalal Fadili¹ · Christophe Chesneau²

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Abstract

In this paper, we consider a high-dimensional statistical estimation problem in which the number of parameters is comparable or larger than the sample size. We present a unified analysis of the performance guarantees of exponential weighted aggregation and penalized estimators with a general class of data losses and priors which encourage objects which conform to some notion of simplicity/complexity. More precisely, we show that these two estimators satisfy sharp oracle inequalities for prediction ensuring their good theoretical performances. We also highlight the differences between them. When the noise is random, we provide oracle inequalities in probability using concentration inequalities. These results are then applied to several instances including the Lasso, the group Lasso, their analysis-type counterparts, the ℓ_∞ and the nuclear norm penalties. All our estimators can be efficiently implemented using proximal splitting algorithms.

Keywords High-dimensional estimation · Exponential weighted aggregation · Penalized estimation · Oracle inequality · Low-complexity models

✉ Jalal Fadili
Jalal.Fadili@ensicaen.fr

Tung Duy Luu
duy-tung.luu@ensicaen.fr

Christophe Chesneau
christophe.chesneau@unicaen.fr

¹ GREYC, Department of computer science, ENSICAEN, CNRS, Normandie Univ, 6, Bd du Maréchal Juin, 14050 Caen, France

² Department of mathematics, LMNO, CNRS, Normandie Univ, Université de Caen Normandie, BP 5186, 14032 Caen, France