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# Oracle Inequalities in Empirical Risk Minimization and Sparse Recovery Problems

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# Preface

The purpose of these lecture notes is to provide an introduction to the general theory of empirical risk minimization with an emphasis on excess risk bounds and oracle inequalities in penalized problems. In the recent years, there have been new developments in this area motivated by the study of new classes of methods in machine learning such as large margin classification methods (boosting, kernel machines). The main probabilistic tools involved in the analysis of these problems are concentration and deviation inequalities by Talagrand along with other methods of empirical processes theory (symmetrization inequalities, contraction inequality for Rademacher sums, entropy and generic chaining bounds). Sparse recovery based on  $\ell_1$ -type penalization and low rank matrix recovery based on the nuclear norm penalization are other active areas of research, where the main problems can be stated in the framework of penalized empirical risk minimization, and concentration inequalities and empirical processes tools proved to be very useful.

My interest in empirical processes started in the late 1970s and early 1980s. It was largely influenced by the work of Vapnik and Chervonenkis on Glivenko–Cantelli problem and on empirical risk minimization in pattern recognition, and, especially, by the results of Dudley on uniform central limit theorems. Talagrand’s concentration inequality proved in the 1990s was a major result with deep consequences in the theory of empirical processes and related areas of statistics, and it inspired many new approaches in analysis of empirical risk minimization problems.

Over the last years, the work of many people have had a profound impact on my own research and on my view of the subject of these notes. I was lucky to work together with several of them and to have numerous conversations and email exchanges with many others. I am especially thankful to Peter Bartlett, Lucien Birgé, Gilles Blanchard, Stephane Boucheron, Olivier Bousquet, Richard Dudley, Sara van de Geer, Evarist Giné, Gabor Lugosi, Pascal Massart, David Mason, Shahar Mendelson, Dmitry Panchenko, Alexandre Tsybakov, Aad van der Vaart, Jon Wellner and Joel Zinn.

I am thankful to the School of Mathematics, Georgia Institute of Technology and to the Department of Mathematics and Statistics, University of New Mexico where most of my work for the past years have taken place.

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I am also very thankful to Jean-Yves Audibert, Kai Ni and Jon Wellner who provided me with the lists of typos and inconsistencies in the first draft.

I am very grateful to all the students and colleagues who attended the lectures in 2008 and whose questions motivated many of the changes I have made since then.

I am especially thankful to Jean Picard for all his efforts that make The Saint-Flour School truly unique.

Atlanta  
February 2011

*Vladimir Koltchinskii*

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