



Learning subgaussian classes : Upper and minimax bounds

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We obtain sharp oracle inequalities for the empirical risk minimization procedure in the regression model under the assumption that the target Y and the model \mathcal{F} are subgaussian. The bound we obtain is sharp in the minimax sense if \mathcal{F} is convex. Moreover, under mild assumptions on \mathcal{F} , the error rate of ERM remains optimal even if the procedure is allowed to perform with constant probability. A part of our analysis is a new proof of minimax results for the gaussian regression model.

Comments: learning theory, empirical process, minimax rates

Subjects: **Statistics Theory (math.ST)**

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