

PARCIMONIE ET GRANDE DIMENSION

SPARE AND LARGE

Lecturers: Marie-Christophette BLANCHET, Alexandre SAIDI, Céline HARTWEG-| Lecturers : 22 | TC : 4 | PW : 0.0 | Autonomy : 0.0 | Study : 4 | Project : 0.0 | Language : FR

Objectives

Sparsity and convexity are ubiquitous notions in Machine Learning and Statistics. In this course, we study the mathematical foundations of some powerful methods based on convex relaxation: L1-regularisation techniques in Statistics and Signal Processing; Nuclear Norm minimization in Matrix Completion. These approaches turned to be Semi-Definite representable (SDP) and hence tractable in practice. The theoretical part of the course will focus on the guarantees of these algorithms under the sparsity assumption. The practical part of this course will present the standard solvers of these learning problems.

Keywords: L1-regularization; Matrix Completion; Semi-Definite Programming; Proximal methods;

Programme

Learning outcomes

Independent study Objectifs : This activity is not concerned with framed autonomy activities outside personal work.

Méhodes : This activity is not concerned with framed autonomy activities outside personal work.

Core texts

Christophe Giraud, INTRODUCTION TO HIGH-DIMENSIONAL STATISTICS, Chapman and Hall/CRC Martin J. Wainwright HIGH-DIMENSIONAL STATISTICS: A NON-ASYMPTOTIC VIEWPOINT, Cambridge University Press Simon Foucart and Holger RauhutA MATHEMATICAL INTRODUCTION TO COMPRESSIVE SENSING

Assessment