

## Estimation of $(I_n - XDX^T)^{-1}$ with large dependence and concentration hypotheses.

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We provide in this presentation an estimation of the expectation of the matrix  $Q(D) = (I_p - XDX^T)^{-1}$  when the data matrix  $X = (x_1, \dots, x_n) \in M_{p,n}$  has independent columns (but not identically distributed) and  $D$  is random, bounded, not independent with  $X$  but satisfies some constraints on the dependence on each  $x_i$ : for any  $i$ , there exists a random diagonal matrix  $D_i$  independent of  $x_i$  and sufficiently close to  $D$ . The formula giving the estimation of  $Q$  is a classical generalization of known deterministic equivalents, the difficulty mainly lies in the proof of the convergence. It is proven under concentration of the measure hypotheses on  $X$  and it relies in particular on a formula giving the concentration of the product of such random vectors. In a sense, the study of  $Q(D)$  is a perfect example to expose the efficiency of the concentration of measure framework to prove random matrix theory inferences.

We will also provide a machine learning application of the estimation of the matrix  $Q(D)$  that concerns the prediction of the performances of robust regularized regression (Ridge regression with a general convex loss replacing the squared loss).

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