

## Efficient dimension reduction

Given a collection of high dimensional vectors  $\mathbf{Y}_1, \dots, \mathbf{Y}_n$ , identify the efficient dimension  $m^*$  and the projector  $\Pi_{m^*}$  such that  $\mathbf{Y}_i = \Pi_{m^*} \mathbf{Y}_i$  up to a measurement error for all  $i$ . The basic assumption is that the spectral gap  $\lambda_{m^*} - \lambda_{m^*+1}$  is sufficiently large (larger than  $Cn^{-1/2}$ ).

The procedure uses the idea of building a confidence set for the projector  $\Pi_m$  for each candidate efficient dimension  $m$  using the bootstrap procedure from Naumov, Spokoiny, Ulyanov (2017). The size of the confidence set is the crucial characteristics:

- for  $m = m^*$ , the eigen-subspace can be recovered with accuracy  $1/\sqrt{n}$ ;
- for  $m < m^*$ , if  $\lambda_m - \lambda_{m+1}$  is sufficiently big, again the confidence width is of order  $n^{-1/2}$ ;
- for  $m < m^*$  but  $\lambda_m - \lambda_{m+1}$  is small or zero, then  $\Pi_m$  can be recovered up to an error within the e.d.r. subspace  $\mathcal{J}$  of dimension  $m^*$ . This yields the confidence width of order  $m^*$ ;
- if  $m > m^*$ , then the confidence width is of order  $(m - m^*) * (d - m^*)$

Workpackages:

1. Efficient implementation working for large dimension  $d$  and big  $n$
2. Exploring the theoretical properties: consistency and accuracy of estimation for the efficient dimension  $m^*$  and the corresponding eigen-subspace;
3. Application financial data
4. Application bio and medical data
5. Extension to a non-Gaussian case

Literature: Naumov, Spokoiny, Ulyanov (2017) Bootstrap confidence sets for spectral projectors of sample covariance. arXiv:1703.00871.

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