

# Contribution associée

## MAXIMUM LIKELIHOOD ESTIMATION IN PRINCIPAL COMPONENTS ANALYSIS WITH NOISE PERTURBATION : A GAUSSIAN MODEL

Jean-Louis PHILOCHE (\*), Damien WYART (\*\*)

(\*) Groupe des écoles des télécommunications,  
(\*\*) Université d'Evry

### 1 Theorem

We give a new proof of a Theorem which explicit the (M.L.E.) out of an iid sample of size  $n$  :

$(y_1, \dots, y_n)$ , each  $y_i$  follows  $\mathcal{N}_p(\mathbf{m}\Sigma)$ .

$$\text{Assumption : } \quad \mathbf{I}_1 > \mathbf{I}_2 > \dots > \mathbf{I}_q > \mathbf{I}_{q+1} = \dots = \mathbf{I}_p > 0 \quad (1)$$

The  $(\mathbf{I}_j)$  are the (unknown) eigenvalues of  $\Sigma$ . For the  $(y_i)$  given, the log-likelihood is a function of  $\mathbf{m}$  and  $\Sigma$ . The log-likelihood depends of the data only through  $\bar{y}$  and  $S$  (the sample covariance). We can limit ourselves to the case  $\mathbf{m} = 0$ , so :  $L(\Sigma) = -\ln(\det(\Sigma)) - \text{Tr}(S\Sigma^{-1})$ , ( $\Sigma \in \mathcal{U}$ ).  $\mathcal{U}$ , subset of  $\mathcal{M}^+$  (the set of the symmetric and positive definite operators (in  $\mathbb{R}^p$ ) is characterised, in  $\mathcal{M}^+$ , by the fact that every  $\Sigma$  in  $\mathcal{U}$  satisfies (1).

*Theorem (I) :* On  $\mathcal{U}$ ,  $L$  attains a maximum at an unique point  $\hat{\Sigma}$ , where the corresponding eigenvalues  $(\hat{\mathbf{I}}_j)$  are  $\hat{\mathbf{I}}_j = s_j (1 \leq j \leq q)$ , and  $\hat{\mathbf{I}}_j = \frac{1}{p-q} \sum_{k=q+1}^p s_k$ , ( $1 \leq j \leq q$ ). (Clearly, the eigenvalues of  $S$  are distinct and strictly positive (a.s), so we can suppose :  $s_1 > s_2 > \dots > 0$ ).

(II) Moreover, any maximising sequence  $(\Sigma_n)$  in  $\mathcal{U}$ , i.e. such that  $\lim_{n \rightarrow \infty} L(\Sigma_n) = L(\hat{\Sigma})$ , converges to  $\hat{\Sigma}$ .

## 2 Principal means to review to build a proof

(i) We shall use a *change of variables* :  $\Theta = \Sigma^{-1}$ ,  $\mathbf{n} = \{ \Theta \in M^+ : \Theta^{-1} \in \mathcal{U} \}$ , with  $F(\Theta) = L(\Theta^{-1}) = \ln(\det(\Theta)) - \text{Tr}(S\Theta)$ . It is very convenient to work with  $F$  which is *strictly concave*.

(ii) We also introduce the action of the group  $SO(p)$  on  $GL(p)$  : this conjugation is defined by :  $(U, \Sigma) \rightarrow U\Sigma U^{-1}$ , with  $U \in SO(p)$ ,  $S \in GL(p)$ .

As  $\mathbf{n}$  is an union of orbits of the action of  $SO(p)$ , to find the maximum of  $F$  in  $\mathbf{n}$ , we can, first, find it on any orbit, then find the maximum of the maxima.