## DISCUSSION

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As stated by the author in Section 3, his emphasis in the study of

(1) 
$$Dy = \Gamma(\theta) = \sum_{\alpha=1}^{s} \theta_{\alpha} A_{\alpha}$$

is different from mine [Anderson (1969, 1970, 1973)]. A brief exposition of this other point of view might put the present paper into a larger perspective. I shall use the notation of the present paper as well as the assumption of Ey = 0 and corresponding modifications in statements.

Anderson considered N observations from  $N(0, \Gamma(\theta))$ ; in the present paper N=1. We suppose  $A_1, \ldots, A_s$  to be symmetric matrices and that for some vector  $\theta = (\theta_1, \ldots, \theta_s)$  the matrix  $\sum_{a=1}^s \theta_a A_a$  is positive definite. Then the derivative equations for the maximum likelihood estimate of  $\theta$  are

(2) 
$$\operatorname{tr}\left(\sum_{\alpha=1}^{s}\hat{\theta}_{\alpha}A_{\alpha}\right)^{-1}A_{b} = y'\left(\sum_{\alpha=1}^{s}\hat{\theta}_{\alpha}A_{\alpha}\right)^{-1}A_{b}\left(\sum_{\alpha=1}^{s}\hat{\theta}_{\alpha}A_{\alpha}\right)^{-1}y, \quad b=1,\ldots,s.$$

If there are several vectors  $\hat{\theta}$  satisfying (2), the vector minimizing  $|\sum_{a=1}^{s} \hat{\theta}_{a} A_{a}|$  is taken. Rewriting (2) as

(3) 
$$\sum_{c=1}^{s} \operatorname{tr} \left( \sum_{a=1}^{s} \hat{\theta}_{a} A_{a} \right)^{-1} A_{b} \left( \sum_{a=1}^{s} \hat{\theta}_{a} A_{a} \right)^{-1} A_{c} \hat{\theta}_{c}$$

$$= y' \left( \sum_{a=1}^{s} \hat{\theta}_{a} A_{a} \right)^{-1} A_{b} \left( \sum_{a=1}^{s} \hat{\theta}_{a} A_{a} \right)^{-1} y, \qquad b = 1, \dots, s,$$

suggests an iterative procedure for solving (2). Let the vector  $\hat{\boldsymbol{\theta}}^{(0)}$  be an initial estimate, and define  $\Gamma(\hat{\boldsymbol{\theta}}^{(i)}) = \sum_{a=1}^{s} \hat{\boldsymbol{\theta}}_{a}^{(i)} A_{a}$ ,  $i = 0, 1, \ldots$ . The *i*th stage of the iteration is solving the linear equations

(4) 
$$\sum_{c=1}^{s} \operatorname{tr} \Gamma^{-1}(\hat{\theta}^{(i-1)}) A_b \Gamma^{-1}(\hat{\theta}^{(i-1)}) A_c \hat{\theta}_c^{(i)} \\ = y' \Gamma^{-1}(\hat{\theta}^{(i-1)}) A_b \Gamma^{-1}(\hat{\theta}^{(i-1)}) y, \qquad b = 1, \dots, s.$$

This is the method of scoring.

If the matrices  $\{A_a\}$  commute [the author's (c)], there exists an orthogonal matrix B such that  $A_a=B\Lambda_a B'$  where  $\Lambda_a$  is diagonal with diagonal elements  $\lambda_{ta},\ t=1,\ldots,n$ . Then

(5) 
$$B'\Gamma(\theta)B = \sum_{\alpha=1}^{s} \theta_{\alpha} \Lambda_{\alpha}$$

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