ESTABLISHING THE POSITIVE DEFINITENESS OF THE SAMPLE COVARIANCE MATRIX

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When dealing with a nonsingular multivariate normal distribution, one makes repeated use of the fact that the sample covariance matrix is positive definite with probability one if the sample size N is larger than the dimension of the random vectors. A direct proof of this fact which does not depend upon other results is unknown to me; and I would here like to suggest such a proof.

THEOREM. Suppose X_1, \dots, X_N is a random sample from a p-variate normal distribution whose covariance matrix $\sum_{i=1}^{p \times p} is$ of full rank. Then the sample covariance matrix $A = \sum_{i=1}^{N} (X_i - \overline{X})(X_i - \overline{X})'$ is positive definite with probability one if and only if N > p.

PROOF. By a suitable linear transformation, we may represent $A^{p \times p}$ as $A^{p \times p} = BB'$ where $B^{p \times N-1} = (\mathbf{Z}_1, \dots, \mathbf{Z}_{N-1})$ and the \mathbf{Z}_i are mutually independent $N(\mathbf{0}, \sum_{i=1}^{p \times p})$ variables. By [1] (Theorem 7, page 399) it will suffice to show $B^{p \times N-1}$ has rank p with probability one if and only if N > p. It is clear that adding more columns cannot diminish the rank of B. Equally clear is the fact that rank B < p if $N \le p$. Thus it will suffice to show that B has rank p with probability one when N = p + 1.

For any set $\{\mathbf{a}_1, \cdots, \mathbf{a}_{p-1}\}$ of vectors in R^p , let $S\{\mathbf{a}_i; i=1, \cdots, p-1\}$ be the subspace spanned by $\mathbf{a}_1, \cdots, \mathbf{a}_{p-1}$. Note that $P(\mathbf{Z}_i \in S\{\mathbf{a}_i; i=1, \cdots, p-1\}) = 0$ for any given set of p-dimensional vectors $\mathbf{a}_1, \cdots, \mathbf{a}_{p-1}$ by the fact that $\sum_{p \neq p} \mathbf{a}_p \mathbf{a}_p$

$$P (\text{rank } B < p)$$

$$= P[\mathbf{Z}_{1}, \dots, \mathbf{Z}_{p} \text{ are linearly dependent}]$$

$$\leq \sum_{i=1}^{p} P[\mathbf{Z}_{i} \in S\{\mathbf{Z}_{1}, \dots, \mathbf{Z}_{i-1}, \mathbf{Z}_{i+1}, \dots, \mathbf{Z}_{p}\}]$$

$$= p \cdot P[\mathbf{Z}_{1} \in S\{\mathbf{Z}_{2}, \dots, \mathbf{Z}_{p}\}]$$

$$= p \cdot E(P[\mathbf{Z}_{1} \in S\{\mathbf{Z}_{2}, \dots, \mathbf{Z}_{p}\} \mid \mathbf{Z}_{2}, \dots, \mathbf{Z}_{p}])$$

$$= p \cdot \int_{R^{p(p-1)}} P[\mathbf{Z}_{1} \in S\{\mathbf{z}_{2}, \dots, \mathbf{z}_{p}\} \mid \mathbf{Z}_{2} = \mathbf{z}_{2}, \dots, \mathbf{z}_{p} = \mathbf{z}_{p}] dF(\mathbf{z}_{2}, \dots, \mathbf{z}_{p})$$

$$= p \cdot \int_{R^{p(p-1)}} P(\mathbf{Z}_{1} \in S\{\mathbf{z}_{2}, \dots, \mathbf{z}_{p}\}) dF$$

$$= p \cdot \int_{R^{p(p-1)}} 0 dF$$

$$= 0.$$

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It can be shown that questions of measurability present no problem.

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REFERENCE

[1] Scheffé, Henry (1959). The Analysis of Variance. Wiley, New York.