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ON THE DISPERSION MATRIX OF A MATRIX QUADRATIC FORM
CONNECTED WITH THE NONCENTRAL WISHART DISTRIBUTION

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Abstract

Recently Magnus and Neudecker [3] derived the dispersion matrix of $\text{vec } X'X$, when X' is a $p \times n$ random matrix ($n > p$) and $\text{vec } X'$ has the distribution $N_{np}(\text{vec } M', I_n \otimes V)$.

This note is concerned with the matrix quadratic form $X'AX$, where X' is as defined above and A is a nonrandom (not necessarily symmetric) matrix. The dispersion matrix of $\text{vec } X'AX$ will then be derived by applying results of Magnus and Neudecker [3] and Neudecker and Wansbeek [4].

It will be shown that an earlier partial and special result of Giguère and Styan [2] which assumes a symmetric A agrees with our result.

1. INTRODUCTION

Let x_i for $i=1, \dots, n$ be $p \times 1$ random vectors that are jointly independent with (normal) distribution $N_p(\mu_i, V)$.

If we define $X' := (x_1', \dots, x_n')$ and $M' := (\mu_1', \dots, \mu_n')$, then $S_I := X'X$ will have a (noncentral Wishart) distribution $W_p(n, V, M'M)$, provided $n > p$.

Magnus and Neudecker [3] derived the dispersion matrix of $\text{vec } S_I$, viz.

$$D(\text{vec } S_I) = (I_{p^2} + K_{pp})[n(V \otimes V) + V \otimes M'M + M'M \otimes V] \quad (1.1)$$

where K_{pp} is a $p^2 \times p^2$ commutation matrix.

Although V was taken to be positive definite in their derivation, it is easy to prove that the result generally holds for nonnegative definite V .

In this paper we shall consider the matrix quadratic form $S_A := X'AX$, where $X'X$ is distributed as stated above and A is a nonrandom (not necessarily symmetric matrix) and derive the dispersion matrix of $\text{vec } S_A$.

After establishing this result in an earlier version [5] of this paper, we became aware of a paper [2] by Giguère and Styan, who presented [see their (2.2.12)] an expression for the cross-covariance matrix T_{ij} , say, between the i -th and j -th columns of S_A ($i, j=1, \dots, p$) for symmetric A .

One can relatively easily derive the complete dispersion matrix $T = \{T_{ij}\}$ if one uses the commutation matrix and properties of Kronecker multiplication.

Whereas the procedure followed by Giguère and Styan is unclear and their result is a partial and special one, we shall give a full derivation of the complete dispersion matrix.

We shall apply some earlier results concerning Kronecker multiplication and the commutation matrix, viz.

$$(1) \quad \text{vec } ABC = (C' \otimes A) \text{vec } B, \text{ for compatible matrices } A, B \text{ and } C \quad (1.1)$$

$$(2) \quad K_{mn} \text{vec } A = \text{vec } A', \text{ where } A \text{ is an } m \times n \text{ matrix} \quad (1.2)$$

$$(3) \quad K_{pm} (A \otimes B) K_{nq} = B \otimes A, \text{ where } A \text{ and } B \text{ are } m \times n \text{ and } p \times q \text{ matrices} \quad (1.3)$$

$$(4) \quad (\text{vec } A)' \text{vec } B = \text{tr } A'B. \quad (1.4)$$

These results are collected in Magnus and Neudecker [3].

$$(5) \quad \text{vec}(A \otimes B) = (I_n \otimes K_{qm} \otimes I_p) (\text{vec } A \otimes \text{vec } B), \text{ where } A \text{ and } B \text{ are arbitrary } m \times n \text{ and } p \times q \text{ matrices.} \quad (1.5)$$

This is Theorem 3.1(i) of Neudecker and Wansbeek [4].

$$(6) \quad \mathcal{D}(\text{vec}(X \otimes X)) = (I_{m^2n^2} + K_{nn} \otimes K_{mm})(I_n \otimes K_{nm} \otimes I_m) \times \\ [V \otimes V + V \otimes \text{vec } M(\text{vec } M)' + \text{vec } M(\text{vec } M)' \otimes V](I_n \otimes K_{mn} \otimes I_m), \quad (1.6)$$

when X is an $m \times n$ matrix and $\text{vec } X$ has distribution $N_{mn}(\text{vec } M, V)$.

This is application 3 of Neudecker and Wansbeek [4].

The result will be reached in stages. First an intermediate result will be derived.

2. AN INTERMEDIATE RESULT

LEMMA.

Let the $p \times 1$ random vectors x_i be independently distributed each as $N_p(0, V)$ for $i=1, \dots, n$. Let

$$X := (x_1, \dots, x_n)' \quad \text{and} \quad M := (\mu_1, \dots, \mu_n)'.$$

Consider the quadratic form $S_A := X'AX$, where the $n \times n$ matrix A is a random (not necessarily symmetric) matrix. Then the dispersion matrix

$$\mathcal{D}(\text{vec } S_A) = (\text{tr } A'A)(I_{p^2} + K_{pp})(V \otimes V). \quad (2.1)$$

Proof. We write

$$\text{vec } S_A = \text{vec } X'AX = (X' \otimes X')\text{vec } A \quad (2.2)$$

$$= \text{vec}[I_{p^2}(X' \otimes X')\text{vec } A] = (\text{vec } A \otimes I_{p^2})' \text{vec}(X' \otimes X'), \quad (2.3)$$

by means of (1.1).

Using (1.6), (1.2), (1.3) and (1.4), we get

$$\mathcal{D}(\text{vec } S_A) = (\text{vec } A \otimes I_{p^2})'(I_{n^2p^2} + K_{nn} \otimes K_{pp})(I_n \otimes K_{np} \otimes I_p) \times \\ (I_n \otimes V \otimes I_n \otimes V)(I_n \otimes K_{pn} \otimes I_p)(\text{vec } A \otimes I_{p^2}) \quad (2.4)$$

$$= [\text{vec } A \otimes (I_{p^2} + K_{pp})]'(I_{n^2} \otimes V \otimes V)(\text{vec } A \otimes I_{p^2}) \quad (2.5)$$

$$= (\text{vec } A)' \text{vec } A \cdot (I_{p^2} + K_{pp})(V \otimes V) \quad (2.6)$$

$$= (\text{tr } A'A)(I_{p^2} + K_{pp})(V \otimes V). \quad \blacksquare \quad (2.7)$$

3. THE MAIN RESULT

THEOREM

Let the $p \times 1$ random vectors x_i be independently distributed each as $N_p(\mu_i, V)$ for $i=1, \dots, n$. Let

$$X := (x_1, \dots, x_n)' \quad \text{and} \quad M := (\mu_1, \dots, \mu_n)' .$$

Consider the matrix quadratic form $S_A := X'AX$, where the $n \times n$ matrix A is a random (not necessarily symmetric) matrix. Then the dispersion matrix

$$D(\text{vec } S_A) = (I_{p^2} + K_{pp})[(\text{tr } A'A)(V \otimes V) + M'A'AM \otimes V + V \otimes M'A'AM]. \quad (3.1)$$

Proof. We write $Y := X - M$. Hence

$$S_A = (Y + M)'A(Y + M) \quad (3.2)$$

$$= Y'AY + Y'AM + M'AY + M'AM . \quad (3.3)$$

As $\text{vec } Y'$ has distribution $N_{np}(0, I_n \otimes V)$, any third moment about the mean is zero. See Anderson [1, p.39]. Therefore

$$D(\text{vec } S_A) = D(\text{vec } Y'AY) + D\{(I_{p^2} + K_{pp})(M'A' \otimes I_p) \text{vec } Y'\} \quad (3.4)$$

$$= (\text{tr } A'A)(I_{p^2} + K_{pp})(V \otimes V) + (I_{p^2} + K_{pp})(M'A' \otimes I_p)(I \otimes V)(AM \otimes I_p)(I_{p^2} + K_{pp}) \quad (3.5)$$

$$= (I_{p^2} + K_{pp})[\frac{1}{2}(\text{tr } A'A)(V \otimes V) + M'A'AM \otimes V](I_{p^2} + K_{pp}) \quad (3.6)$$

$$= (I_{p^2} + K_{pp})[(\text{tr } A'A)(V \otimes V) + M'A'AM \otimes V + V \otimes M'A'AM], \quad (3.7)$$

by virtue of the Lemma, (1.2) and (1.3).

4. THE RESULT OF GIGUÈRE AND STYAN

Giguère and Styan presented the cross-covariance matrix T_{ij} between the i -th and j -th columns of S_A for symmetric A , viz.

$$T_{ij} = (\text{tr } A^2)(v_j v_i' + v_{ij} V) + h_j v_i' + v_j h_i' + v_{ij} H + h_{ij} V, \quad (4.1)$$

where

$$H = \{h_{ij}\} = (h_1, \dots, h_p) := M'A^2M$$

and

$$V = \{v_{ij}\} = (v_1, \dots, v_p) .$$

Clearly the (i,j) th submatrix of $V \otimes H$ is $v_{ij} H$.

Further

$$\begin{aligned} K_{pp}(V \otimes H) &= K_{pp}(v_1 \otimes H, \dots, v_p \otimes H) = [K_{pp}(v_1 \otimes H) \dots K_{pp}(v_p \otimes H)] \\ &= (H \otimes v_1 \dots H \otimes v_p) . \end{aligned}$$

Its (i,j) th submatrix is $h'_i \otimes v_j = v_j \otimes h'_i = v_j h'_i$, etcetera.

From this follows

$$T = \{T_{ij}\} = (I_{p^2} + K_{pp})[(\text{tr } A^2)(V \otimes V) + H \otimes V + V \otimes H]. \quad (4.2)$$

5. COMMENT I

When A is symmetric idempotent of sufficient rank, r say, then S_A will have a (noncentral) Wishart distribution.

In this case we get

$$\mathcal{D}(\text{vec } S_A) = (I_{p^2} + K_{pp})[r(V \otimes V) + M'AM \otimes V + V \otimes M'AM]. \quad (5.1)$$

The result of Magnus and Neudecker [3] is a special case of (5.1) for $A = I_n$. Another special case arises for $A = N$, $N := I_n - \frac{1}{n} s_n s'_n$, where $s'_n := (1, \dots, 1)$ is the n -dimensional summation vector. We then find

$$\mathcal{D}(\text{vec } S_N) = (I_{p^2} + K_{pp})[(n-1)(V \otimes V) + M'NM \otimes V + V \otimes M'NM]. \quad (5.2)$$

6. COMMENT II

The dispersion matrix for the symmetric case can, of course, be found very easily.

Let $R'AR = \Lambda$ be the diagonal representation of A . Let further $Y' := X'R$, and $Y' = (y'_1, \dots, y'_n)$. Then $E(Y') = M'R$ and $\mathcal{D}(\text{vec } Y') = I_n \otimes V$. Hence the vectors y_i are jointly independent with (normal) distribution $N_p(M'R_{\cdot i}, V)$, where $R_{\cdot i}$ is the i -th column of R . Then $X'AX = X'RAR'X = Y'\Lambda Y = \sum_i \lambda_i y_i y'_i$.

Further $\mathcal{D}(\text{vec } X'AX) = \mathcal{D}(\sum_i \lambda_i y_i y'_i)$

$$= \sum_i \lambda_i^2 \mathcal{D}(y_i y'_i) = \sum_i \lambda_i^2 (I_{p^2} + K_{pp})(V \otimes V + V \otimes M'R_{\cdot i}(R_{\cdot i})'M$$

$$+ M'R_{\cdot i}(R_{\cdot i})'M \otimes V) = (I_{p^2} + K_{pp})[(\text{tr } A^2)(V \otimes V) + V \otimes M'A^2M$$

$$+ M'A^2M \otimes V], \text{ by virtue of Theorem 4.3(iv) of Magnus and Neudecker [3].}$$

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