FOURTH MOMENTS IN THE GENERAL LINEAR MODEL; AND THE VARIANCE OF TRANSLATION INVARIANT QUADRATIC FORMS

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Abstract

Using vec and Kronecker product operators, a detailed derivation is given of fourth moments in the general linear model and of the variance of translation invariant quadratic forms.

Introduction

We consider the general linear model

$$y = X\beta + Z_1u_1 + Z_2u_2 + \cdots + Z_nu_k$$
(1)

where X of order n × p and Z_m of order n × c_m , m = 1, ..., k are known incidence matrices, β is an unknown vector of p fixed effects, and the u_m , of order $c_m \times 1$ for m = 1, ..., k, are unknown vectors of random effects such that

- (i) the elements of u are independent having common variance σ_m^2 and kurtosis $\gamma_m,$ and
- (ii) u_{m} and u_{m} , are independent for $m \neq m'$.

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BU-630-M*

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Paper No. BU-630-M in the Biometrics Unit.

$$\operatorname{Var}(\underline{y}) = \sum_{m=1}^{k} \sigma_{m \sim m}^{2} Z_{m} Z_{m}^{\prime} = \underline{y}$$

As far as the ensuing algebra is concerned, a more convenient representation of (1) is

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}\mathbf{u} \tag{2}$$

where Z is the n X c. $(c_1 = \sum_{m=1}^{k} c_m)$ partitioned matrix $Z = [Z_1 \cdots Z_k]$, and u_1 is the c. X l vector $u'_1 = [u'_1 \cdots u'_k]$. Corresponding to model (2), the variance-covariance matrix of y is then

$$V = ZDZ'$$

where

$$D = E(uu') = \sum_{m=1}^{k} \sigma^{2}I_{m \sim c_{m}}$$
(3)

and Σ^+A , denotes the direct sum of matrices A_{i} .

Fourth Moments

The matrix of central fourth moments of the vector y is, by definition,

$$\mathbf{F} = \operatorname{Var}[(\mathbf{y} - \mathbf{X}\mathbf{\beta}) * (\mathbf{y} - \mathbf{X}\mathbf{\beta})]$$
(4)

where $A \stackrel{\text{\tiny def}}{\sim} B$ is the direct (Kronecker) product of A and B. Substituting in terms of (2),

$$\mathbf{F} = \operatorname{Var}[(\underline{Z}\underline{u}) * (\underline{Z}\underline{u})]$$

$$= \operatorname{Var}[(\underline{Z} * \underline{Z})(\underline{u} * \underline{u})]$$

$$= (\underline{Z} * \underline{Z})[\operatorname{Var}(\underline{u} * \underline{u})](\underline{Z} * \underline{Z})' . \qquad (5)$$

Defining the vec of a matrix to be the M^c matrix first introduced by Roth [1934], namely, the vector obtained from stacking the columns of the matrix one beneath the other in a single vector, and noting that vec(uu') = u * u it follows that

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$$Var(\underline{u} * \underline{u}) = E[(\underline{u} * \underline{u})(\underline{u} * \underline{u})'] - [E(\underline{u} * \underline{u})][E(\underline{u} * \underline{u})]'$$

$$= E[(\underline{u}\underline{u}') * (\underline{u}\underline{u}')] - E[vec(\underline{u}\underline{u}')]E[vec(\underline{u}\underline{u}')]'$$

$$= E[(\underline{u}\underline{u}') * (\underline{u}\underline{u}')] - vecE(\underline{u}\underline{u}')[vecE(\underline{u}\underline{u}')]'$$

$$= E[(\underline{u}\underline{u}') * (\underline{u}\underline{u}')] - vecD(vecD)', \qquad (6)$$

on using \underline{D} of (3).

To simplify (6) we define

$$c \equiv c_{1} = c_{1} + c_{2} + \cdots + c_{k}$$

$$(7)$$

and

$$w = D^{-\frac{1}{2}} w = \{w_i\} \quad i = 1, \dots, c.$$
 (8)

Then w has the properties $\widetilde{\alpha}$

$$E(\underline{w}) = 0, \quad E(\underline{ww}') = var(\underline{w}) = \underbrace{I}_{c}$$
(9)

and

$$E(w_{i}^{4}) = 3 + \dot{\gamma}_{i}$$
 for $i = 1, \dots, c$ (10)

where

$$\dot{\gamma}_{i} = i^{th} \text{ diagonal element of } \sum_{m=1}^{k} \gamma_{m \sim c_{m}}^{I}$$
 (11)

Then for (6)

$$E[(\underline{u}\underline{u}') * (\underline{u}\underline{u}')] = (\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}})E[(\underline{w}\underline{w}') * (\underline{w}\underline{w}')](\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}})',$$

$$= (\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}})\Sigma(\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}}) \qquad (12)$$

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on defining

$$\begin{split} \Sigma_{c}^{2} \chi_{c}^{2} &\equiv \{\Sigma_{ij}\} \quad \text{for } i, j = 1, 2, \cdots, c \\ &= \{E(w_{i}^{w} y_{j}^{ww'})\} \\ &= \{E(w_{i}^{w} y_{k}^{w} y_{k})\} \quad \text{for } i, j, k, l = 1, 2, \cdots, c . \end{split}$$
(13)

Now for i = j

$$E(w_{i}w_{i}w_{k}w_{k}) = \begin{cases} 3 + \dot{\gamma}_{i} & \text{when } i = k = l \\ 1 & \text{when } i \neq k = l \\ 0 & \text{otherwise} \end{cases}$$
(14)

and for i≠j

$$E(w_{j}w_{j}w_{k}w_{l}) = \begin{cases} 1 & \text{when } i = k, j = l \\ 1 & \text{when } i = l, j = k \\ 0 & \text{otherwise,} \end{cases}$$
(15)

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so that, on defining

$$e_{i} = i^{th} \text{ column of } I_{c}, \qquad (16)$$

(14) and (15) give the sub-matrices of (13) as

$$\sum_{\sim ii} = I + (2 + \dot{\gamma}_i) e_i e'_i$$
(17)

and

$$\Sigma_{ij} = e_i e'_j + e_j e'_j, \quad \text{for } i \neq j. \quad (18)$$

Therefore in (13), noting that $2e_{i}e'_{i} = e_{i}e'_{j} + e_{i}e'_{i}$ for i = j,

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$$\Sigma = \{\Sigma_{ij}\} = I + \{e, e'_{i} + e, e'_{i}\} \text{ for } i, j = 1, \dots, c, + \sum_{i=1}^{c} \dot{\gamma}_{i \neq i \neq i} .$$
(19)

In (19) it is important to note that for the matrix $\{e,e'_{i,j} + e,e'_{j,i}\}$ the sequence of subscripts is $j = 1, \dots, c$ within $i = 1, \dots, c$. This being so, it can be noted that

$$\{ e_{j} e_{i}^{j} \}$$
 for i, j = 1, ..., c, = $I_{(c,c)}$, (20)

the permuted identity matrix of order c^2 , as used by Tracy and Dwyer [1969] and MacRae [1974]; and

$$\{\underbrace{e}_{i \neq j}\} \text{ for } i, j = 1, \cdots, c, = \operatorname{vec}_{i \neq j}(\operatorname{vec}_{i})$$
(21)

Furthermore, using (11) in the last term of (19) gives, for $t_{m-1} = c_1 + c_2 + \cdots + c_{m-1}$

$$\overset{c}{\underset{i=1}{\Sigma^{+}}} \overset{\cdot}{\gamma}_{i \sim i \sim i} \overset{e}{\underset{m=1}{E^{+}}} \overset{t}{\gamma}_{m} \begin{pmatrix} \overset{t}{\underset{m=1}{\Sigma^{+}}} \\ \Sigma^{+} & e_{i} e_{i}' \\ i = t_{m-1} + 1 \sim i \sim i \end{pmatrix} .$$

$$(22)$$

Now for $i = 1, \dots, c$

and

$$\begin{bmatrix} e_{i}e'_{i} & 0\\ \sim i \sim i & \sim\\ 0 & e_{i}+l \sim i+1 \end{bmatrix}$$
 a diagonal matrix of order 2c with its only non-zero
= elements being 1 in the (i,i) and (c + i + 1, c + i + 1)
positions.

Consider just the diagonal elements of this last matrix. Between the two l's there are c - i + i = c zeros; and this is true for all 1. Denote a row vector of c zeros by O_{c}^{i} . Then the diagonal elements of (22) are

$$\begin{bmatrix} Y_1 & 0'_1 & Y_1 & 0'_1 & \cdots & 0'_n & Y_1 & 0'_n & Y_2 & 0'_n & \cdots & Y_2 & 0'_n & \cdots & 0'_n & Y_k & 0'_n & \cdots & 0'_n & Y_k \end{bmatrix}$$
(23)

where γ_{m} occurs c_{m} times for $m = 1, 2, \dots, k$. This is a vector of c^{2} elements and by its nature is $vec \begin{pmatrix} k \\ \Sigma^{+} & \gamma_{m \sim c_{m}} \end{pmatrix}$. Hence, on using the definition

diag $x = \frac{\text{diagonal matrix with diagonal elements}}{\text{being the elements of the vector } x}$,

we have (22) as

$$\sum_{i=1}^{c} \dot{\gamma}_{i \sim i \sim i} = \operatorname{diag} \{ \operatorname{vec} (\Sigma^{+} \gamma_{m \sim c_{\pi}}) \} .$$

$$(24)$$

Substituting (20), (21) and (24) into (19) gives

$$\sum_{n=1}^{\Sigma} = \prod_{n=1}^{\Sigma} + \prod_{n=1}^{\Sigma} (c,c) + (\operatorname{vecI}_{n})(\operatorname{vecI}_{n})' + \operatorname{diag}\{\operatorname{vec}(\Sigma^{+} \gamma_{\mathrm{m}} I_{\mathrm{c}})\}.$$
(25)

Using Σ in (12) now gives

$$E[(\underline{u}\underline{u}' * \underline{u}\underline{u}')] = (\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}})[\underline{I} + \underline{I}_{(c,c)} + (\underline{vecI}_{c})(\underline{vecI}_{c})' + diag\{\underline{vec}(\Sigma^{+} \gamma_{\underline{m}\sim c_{\underline{n}}})\}](\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}}) = \underline{D} * \underline{D} + (\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}})\underline{I}_{(c,c)}(\underline{D}^{\frac{1}{2}} * \underline{D}^{\frac{1}{2}}) + \underline{z}\underline{z}' + \underline{\Gamma}$$
(26)

where we define

$$\Gamma \equiv (D^{\frac{1}{2}} * D^{\frac{1}{2}}) [\operatorname{diag} \{ \operatorname{vec} \begin{pmatrix} k \\ \Sigma^{+} & \gamma_{m \downarrow c} \\ m = 1 \end{pmatrix} \}] (D^{\frac{1}{2}} * D^{\frac{1}{2}}) .$$
(27)

Also, use is made of the results in MacRae [1974] that

$$I_{\sim}(p,p) (A_{p\times q} * B_{\sim}p\times q) I_{\sim}(q,q) = B_{p\times q} * A_{p\times q}$$
(28)

and

$$[I_{(p,p)}]^2 = I_{p^2}$$
(29)

so that

$$\mathbf{I}_{(p,p)}(\mathbf{A}_{pxq} * \mathbf{A}_{pxq}) = (\mathbf{B}_{pxq} * \mathbf{A}_{pxq})\mathbf{I}_{(q,q)} .$$
(30)

Hence for (26)

$$\left(\underbrace{D^{\frac{1}{2}}}_{\sim}^{\ast} * \underbrace{D^{\frac{1}{2}}}_{\sim}^{\ast} \right) I_{(c,c)} \left(\underbrace{D^{\frac{1}{2}}}_{\sim}^{\ast} * \underbrace{D^{\frac{1}{2}}}_{\sim}^{\ast} \right) = \left(\underbrace{D}_{\sim}^{\ast} * \underbrace{D}_{\sim}^{\ast} \right) I_{(c,c)} .$$
(31)

Furthermore, in (26)

$$z = (D^{\frac{1}{2}} * D^{\frac{1}{2}}) \operatorname{vec} I = \operatorname{vec} D$$
(32)

because, in general,

$$\operatorname{vec}(\operatorname{ABC}) = (\operatorname{C}' * \operatorname{A})\operatorname{vecB}$$
 (33)

as in Neudecker [1969].

Using (31) and (32) in (26) therefore gives

$$E[(\underline{u}\underline{u}') * (\underline{u}\underline{u}')] = (\underline{D} * \underline{D})(\underline{I} + \underline{I}_{(c,c)}) + (\underline{vec}\underline{D})(\underline{vec}\underline{D})' + \underline{\Gamma}$$
(34)

where Γ is as defined in (27), and so substitution into (6) gives

$$\operatorname{var}(\underline{u} * \underline{u}) = (\underline{D} * \underline{D})(\underline{I} + \underline{I}_{(c,c)}) + \underline{\Gamma} .$$
(35)

Putting (35) into (5) gives the matrix of fourth moments as

$$\mathbf{F} = (\mathbf{Z} * \mathbf{Z})[(\mathbf{D} * \mathbf{D})(\mathbf{I} + \mathbf{I}_{(c,c)}) + \mathbf{F}](\mathbf{Z}' * \mathbf{Z}')$$

and because ZDZ' = V this is:

$$\mathbf{F} = \mathbf{V} * \mathbf{V} + (\mathbf{Z}\mathbf{D} * \mathbf{Z}\mathbf{D})\mathbf{I}_{(c,c)}(\mathbf{Z'} * \mathbf{Z'}) + (\mathbf{Z} * \mathbf{Z})\mathbf{\Gamma}(\mathbf{Z'} * \mathbf{Z'}) .$$

Using (30) again leads to

$$\mathbf{F} = (\mathbf{V} \ast \mathbf{V})(\mathbf{I} + \mathbf{I}_{(n,n)}) + (\mathbf{Z} \ast \mathbf{Z})\mathbf{\Gamma}(\mathbf{Z'} \ast \mathbf{Z'})$$
(36)

and on using (27) this has the equivalent form

$$\mathbf{F} = (\mathbf{V} * \mathbf{V})(\mathbf{I} + \mathbf{I}_{(n,n)}) + (\mathbf{ZD}^{\frac{1}{2}} * \mathbf{ZD}^{\frac{1}{2}})[\operatorname{diag}\{\operatorname{vec}\left(\sum_{m=1}^{K} \gamma_{m,m} \mathbf{I}_{m,m}\right)\}](\mathbf{D}^{\frac{1}{2}} \mathbf{Z}' * \mathbf{D}^{\frac{1}{2}} \mathbf{Z}'). (37)$$

This, then, is the general expression for the matrix of fourth central moments of the vector of observations in linear model theory.

In the special case of normality assumptions, i.e., $u_m \sim N(0, \sigma_{m \sim c_m}^2)$, we have $\gamma_m = 0$ and (37) reduces to

$$\mathbf{\tilde{F}} = (\mathbf{\tilde{V}} * \mathbf{\tilde{V}})(\mathbf{\tilde{I}} + \mathbf{\tilde{I}}_{(n,n)})$$
(38)

Variance of Translation Invariant Quadratic Forms

The quadratic form y'Ay is called translation invariant when A, as well as being symmetric, satisfies AX = 0. Then the variance of the translation invariant quadratic form is

$$v(\underline{y}'\underline{A}\underline{y}) = v[(\underline{y} - \underline{X}\underline{\beta})'\underline{A}(\underline{y} - \underline{X}\underline{\beta})]$$
$$= v(\underline{u}'\underline{Z}'\underline{A}\underline{Z}\underline{u})$$
$$= v\{tr[A(\underline{Z}\underline{u}\underline{u}'\underline{Z}')]\}.$$

Now use the general result for any product PQ, that

$$tr(\underline{PQ}) = (vec\underline{P'})'vec\underline{Q}$$
(39)

and so

$$v(\underline{y}'\underline{A}\underline{y}) = v\{(\underline{vecA})'\underline{vec}(\underline{Zuu}'\underline{Z}')\}$$

$$= (\underline{vecA})'\underline{var}[\underline{vec}(\underline{Zuu}'\underline{Z}')]\underline{vecA}$$

$$= (\underline{vecA})'\underline{var}[(\underline{Zu}) * (\underline{Zu})']\underline{vecA}$$

$$= (\underline{vecA})'\underline{F}(\underline{vecA}), \text{ using } (\underline{4})$$

$$(40)$$

and on using (36) this gives

$$\mathbf{v}(\mathbf{y}'\mathbf{A}\mathbf{y}) = \mathbf{\theta}_1 + \mathbf{\theta}_2 \tag{41}$$

for

$$\theta_{1} = (\operatorname{vec} A)' (\underbrace{V}_{\sim} & \underbrace{V}_{\sim}) (\underbrace{I}_{\sim} + \underbrace{I}_{(n,n)}) \operatorname{vec} A$$

$$(42)$$

and

$$\theta_{2} = (\operatorname{vec} A)' (\operatorname{ZD}^{\frac{1}{2}} * \operatorname{ZD}^{\frac{1}{2}}) [\operatorname{diag}[\operatorname{vec}(\operatorname{\overset{k}{\Sigma^{+}}} \gamma_{m \sim c_{\pi}})] (\operatorname{\overset{1}{D}^{\frac{1}{2}}Z' * \overset{k}{D^{\frac{1}{2}}Z'}) \operatorname{vec} A . \quad (43)$$

In θ_1 of (42), the elements of the $[(i-1)n+j]^{th}$ row of $I_{(n,n)}$ are all zero except for a 1 in the $[(j-1)n+i]^{th}$ column (and vice versa); and also, because A is symmetric, the $[(i-1)n+j]^{th}$ and $[(j-1)n+i]^{th}$ elements of vecA are the same. Hence

$$I_{\sim(n,n)}(\text{vecA}) = \text{vecA} . \tag{44}$$

Also using (33) and (39)

$$(\operatorname{vecA})'(\underline{V} * \underline{V})\operatorname{vecA} = (\operatorname{vecA})'\operatorname{vec}(\underline{VAV}) = \operatorname{tr}(\underline{AV})^2$$
, (45)

so that

$$\theta_1 = 2 \operatorname{tr}(AV)^2 . \tag{46}$$

Simplification of θ_2 in (43) starts with using (33) to get

$$\theta_{2} = \left[\operatorname{vec}\left(\underbrace{\mathbf{D}^{\frac{1}{2}}}_{m=1}^{2}, \operatorname{AZD}^{\frac{1}{2}}\right) \right] \left[\operatorname{diag}\left\{ \operatorname{vec}\left(\underbrace{\mathbf{D}^{\frac{1}{2}}}_{m=1}^{k}, \operatorname{V}_{m \sim c_{m}}^{\mathbf{I}}\right) \right\} \right] \operatorname{vec}\left(\underbrace{\mathbf{D}^{\frac{1}{2}}}_{m \sim m}^{2}, \operatorname{AZD}^{\frac{1}{2}}\right) \right]$$

which is of the form

$$\theta_2 = (\text{vecH})'[\text{diag}\{\text{vecL}\}] \text{vecH}$$
(47)

for

$$H = D^{\frac{1}{2}} AZD^{\frac{1}{2}} \text{ and } L = \sum_{m=1}^{k} \gamma_{m} I.$$
(48)

The nature of the vec and diag operators means that (47) is

$$\theta_2 = \sum_{ij} \sum_{ij} \ell_{ij}$$
(49)

for $\underline{H} = {h_{ij}}$ and $\underline{L} = {l_{ij}}$ of (48). But with this L, the only non-zero l_{ij} 's are the diagonal ones, $l_{tt} = \gamma_m$ for $t = 1, \dots, c_m$ and $m = 1, \dots, k$. Furthermore, as in (23), these diagonal elements have c zeros between them in vecL so that the use of (48) in (49) gives

$$\theta_{2} = \sum_{m=1}^{k} \gamma_{m} \sum_{t=1}^{c_{m}} h_{t=1}^{2}$$

for

$$h_{tt}^2 = t^{th}$$
 diagonal element of the mth diagonal sub-matrix of $\sum_{n=1}^{\frac{1}{2}} AZD^{\frac{1}{2}}$
= $\sigma_m^2(t^{th}$ diagonal element of the mth diagonal sub-matrix of $Z'AZ$).

Therefore

$$\theta_{2} = \sum_{m=1}^{k} \gamma \sigma_{mm}^{2} (\text{sum of squares of diagonal elements of } Z'AZ_{mmm}) .$$
(50)

Substituting (46) and (50) into (41) gives

$$v(\underline{y}'\underline{A}\underline{y}) = 2tr(\underline{A}\underline{V})^2 + \sum_{m=1}^{K} \gamma_m \sigma_m^2 (\text{sum of squares of diagonal elements of } \underline{Z}'\underline{A}\underline{Z}).$$
(51)

This is the variance, under non-normality, of a translation invariant (AX = 0)quadratic form y'Ay. Under normality, $\gamma_m = 0$ for all m and (51) reduces to the familiar form

$$\mathbf{v}(\mathbf{y}'\mathbf{A}\mathbf{y}) = 2\mathbf{tr}(\mathbf{A}\mathbf{V})^2 . \tag{52}$$

Equation (51) is, of course, equivalent to the result given by Rao [1971] where he writes A_{1} for D and $A_{2} = \sum_{m=1}^{k} \gamma_{m} \sigma^{4}I$ and $\tilde{B} = \text{diag}\{\text{diagonal elements of }B\}$ and so gets

$$v(\underline{y}'\underline{A}\underline{y}) = 2tr(\underline{B}\underline{\Delta}_{1})^{2} + tr(\underline{\tilde{B}}\underline{\Delta}_{2}\underline{\tilde{B}})$$
,

for B = Z'AZ. With Y being ZDZ' this is readily seen to be the same as (51).

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