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## ON BIASED REGULARIZING ESTIMATORS. PART II

1. Introduction

The subject of consideration is the model (see [3], [5] for details)  $NM_2 = (R^{n \times k}, S, Y = X\beta + \Xi, k_0 = k, n_0 = n, P_Y = N_Y(X\beta, \Omega))$ , where the random vector  $Y$  has probability distribution  $P_Y$  being  $n$  dimensional normal distribution  $N_Y$  with mean  $EY = X\beta$ , dispersion  $DY = \Omega, \Omega \in R^{n \times n}, X \in R^{n \times k}, \beta \in R^k$ . The purpose of this paper is a comparative analysis of statistical properties of the following three estimators derived from the functionals

$$\varphi_1 = \|\Omega^{-1}(Y - X\beta)\|^2$$

$$\varphi_2 = \varphi_1 + \|\beta\|^2, \quad \varphi_3 = \varphi_1 + \beta' \Gamma \beta$$

$$B_a = (X' \Omega^{-1} X)^{-1} X' \Omega^{-1} Y,$$

$$B_b = (X' \Omega^{-1} X + \gamma I)^{-1} X' \Omega^{-1} Y,$$

$$B_c = (X' \Omega^{-1} X + \Gamma)^{-1} X' \Omega^{-1} Y,$$

and some functions of  $B_a, B_b, B_c$ . It would be proved, among others, that (see § 2):

.) the estimators  $B_a, B_b$ , the predictors  $\hat{Y}_a, \hat{Y}_b$ , the residuals  $E_a, E_b$  are consistent and are having normal distribution;

..) the random quadratic forms  $B_a' B_a, B_b' B_b, \hat{Y}_a' \hat{Y}_a, \hat{Y}_b' \hat{Y}_b, E_a' E_a, E_b' E_b$  do not have  $\chi^2$  - distribution,

...)  $\text{cov}(B_a, S_{E_a}^2) = 0_{k \times 1}, \text{cov}(B_b, S_{E_b}^2) \neq 0_{k \times 1}$ , where  $S_{E_a}^2 = E_a' E_a / (n - k), S_{E_b}^2 = E_b' E_b / (n - k)$ .

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In § 3 there would be proved for  $B_0$  some analogous of theorems from § 2.

## 2. Properties of estimator $B_b$

From the works [6], [7] it follows that if

$$P_Y = \mathcal{N}_Y(m, \Omega), C, \Omega, A \in R^{n \times n}, L, m, d \in R^{n \times 1}, \text{ then}$$

- I)  $E(Y'AY) = \text{tr}(A\Omega) + m^T A m,$   
 II)  $\text{var}(Y'AY) = 2 \text{tr}(A\Omega)^2 + 4 m^T A \Omega A m,$   
 III)  $\text{cov}(LY, Y'AY) = 2L\Omega A m,$   
 IV)  $\text{MSE}(Y) = \text{tr} S(Y) + \text{tr ob}(Y) \text{ob}^T(Y), \text{ob}(Y) = \xi Y - m,$   
 V) if  $Y = d + CU, P_U = \mathcal{N}_U(m, \Omega),$  then  
 $P_Y = \mathcal{N}_Y(Cm + d, C\Omega C'),$   
 VI) if  $Q = Y'AY + 2a'Y + \bar{\alpha}, A \in R^{n \times n}, a \in R^{n \times 1}, \bar{\alpha} \in R,$  then  
 $P_Q = \chi^2(s, \lambda)$  iff  $\Omega A \Omega A \Omega = \Omega A \Omega,$   
 $\Omega(a + A m) = \Omega A \Omega(a + A m), \lambda = \bar{\alpha} + 2a^T m + m^T A m,$   
 $s = \text{tr} A \Omega = r z(\Omega A \Omega).$

The relations (I) and (IV) also hold for  $P_Y \neq \mathcal{N}_Y(\dots)$  with  $\xi(Y) = m, S(Y) = \Omega.$

Using the definition of model  $NM_2,$  the relations (I)-(VI) and the definitions of symbols  $\xi, S, P, \text{MSE}, \text{var}, \text{cov}$  (expectation, dispersion, probability distribution, mean square error, variance, covariance) it is easy to see that:

- (1a)  $B_a = (x' \Omega^{-1} x)^{-1} x' \Omega^{-1} Y = K_a Y, K_a = (x' \Omega^{-1} x)^{-1} x' \Omega^{-1},$   
 (2a)  $E(B_a) = K_a x \beta = \beta, \text{ob}(B_a) = \xi(B_a) - \beta = 0$  (unbiasedness of  $B_a$ ),  
 (3a)  $S(B_a) = K_a \Omega K_a = (x \Omega^{-1} x)^{-1},$   
 (4a)  $P_{B_a} = \mathcal{N}_{B_a}(\beta, x' \Omega^{-1} x)^{-1},$  (normality of  $B_a$ )

- (5a)  $MSE(B_a) = \text{tr}(\mathbf{x}'\Omega^{-1}\mathbf{x})^{-1}$ ,
- (6a)  $B_a'B_a = Y'K_a'K_aY$ ,
- (7a)  $E(B_a'B_a) = \text{tr}(K_a'K_a\Omega) + \beta'\beta = \text{tr}(\mathbf{x}'\Omega^{-1}\mathbf{x})^{-1} + \beta'\beta$ ,
- (8a)  $\text{var}(B_a'B_a) = 2\text{tr}(\mathbf{x}'\Omega^{-1}\mathbf{x})^{-2} + 4\beta'(\mathbf{x}'\Omega^{-1}\mathbf{x})^{-1}\beta$ ,
- (9a)  $P_{B_a'B_a} \neq \chi^2(\dots)$  due to  $\Omega K_a'K_a\Omega K_a'K_a \neq \Omega K_a'K_a\Omega$
- (10a)  $\hat{Y}_a = \mathbf{x}B_a = \mathbf{x}K_aY$ ,
- (11a)  $E(\hat{Y}_a) = \mathbf{x}K_a\mathbf{x}\beta = \mathbf{x}\beta$ , ob  $(\hat{Y}_a) = \mathbf{x}\beta - \mathbf{x}\beta = 0$ ,
- (12a)  $S(\hat{Y}_a) = \mathbf{x}K_a\Omega K_a'\mathbf{x} = \mathbf{x}(\mathbf{x}'\Omega^{-1}\mathbf{x})^{-1}\mathbf{x}'$ ,
- (13a)  $P_{\hat{Y}_a} = \mathcal{N}_{\hat{Y}_a}(\mathbf{x}\beta, \mathbf{x}K_a\Omega K_a'\mathbf{x})$ ,
- (14a)  $MSE(\hat{Y}_a) = \text{tr}(\mathbf{x}K_a\Omega K_a'\mathbf{x})$ ,
- (15a)  $\hat{Y}_a'\hat{Y}_a = Y'K_a'\mathbf{x}'\mathbf{x}K_aY$ ,
- (16a)  $E(\hat{Y}_a'\hat{Y}_a) = \text{tr}(K_a'\mathbf{x}'\mathbf{x}K_a\Omega) + \beta'\mathbf{x}'K_a'\mathbf{x}'\mathbf{x}K_a\mathbf{x}\beta$ ,
- (17a)  $\text{var}(\hat{Y}_a'\hat{Y}_a) = 2\text{tr}(K_a'\mathbf{x}'\mathbf{x}K_a\Omega)^2 + 4\beta'\mathbf{x}'K_a'\mathbf{x}'\mathbf{x}K_a\Omega K_a'\mathbf{x}'\mathbf{x}K_a\mathbf{x}\beta$ ,
- (18a)  $P_{\hat{Y}_a'\hat{Y}_a}/\sigma^2 \neq \chi^2(\dots)$  due to  $\Omega K_a'\mathbf{x}'\mathbf{x}K_a\Omega K_a'\mathbf{x}'\mathbf{x}K_a\Omega \neq \Omega K_a'\mathbf{x}'\mathbf{x}K_a\Omega$ ,
- (19a)  $E_a = (I - \mathbf{x}K_a)Y = M_aY$ ,  $M_a = I - \mathbf{x}K_a$ ,
- (20a)  $E(E_a) = \mathbf{x}\beta - \mathbf{x}K_a\mathbf{x}\beta = 0$ , ob  $(E_a) = 0 - 0 = 0$ ,
- (21a)  $S(E_a) = M_a\Omega M_a'$ ,
- (22a)  $P_{E_a} = \mathcal{N}_{E_a}(0, M_a\Omega M_a')$ ,
- (23a)  $MSE(E_a) = \text{tr}(M_a\Omega M_a')$ ,



$$(24a) \quad E'_a E_a = Y' M'_a M_a Y,$$

$$(25a) \quad E(E'_a E_a) = \text{tr}(M'_a M_a \Omega) + \beta' x' M'_a M_a x \beta = \text{tr}(M'_a M_a \Omega) + 0,$$

$$(26a) \quad \text{var}(E'_a E_a) = 2\text{tr}(M'_a M_a \Omega)^2 + 4\beta' x' M'_a M_a \Omega M'_a M_a x \beta,$$

$$(27a) \quad P_{E'_a E_a} / \sigma^2 \neq \chi^2(\dots), \text{ since } \Omega M'_a M_a \Omega M'_a M_a \Omega \neq \Omega M'_a M_a \Omega,$$

$$(28a) \quad \text{cov}(B_a, S_a^2) = \text{cov}(K_a Y, \frac{1}{n_a} E'_a E_a) = \frac{2}{n_a} K_a \Omega M'_a M_a x \beta = 0,$$

$$n_a = \text{tr} M'_a M_a \Phi, \quad \Omega = \sigma^2 \Phi, \quad \text{tr}(M'_a M_a) = \text{tr}[\Phi - x' x (x' \Omega^{-1} x)^{-1}].$$

Using (I)-(VI), the assumptions of model NM and the definitions of  $\xi$ ,  $\mathcal{S}$ , var, MSE, P, cov it is easy to find out that

$$(1b) \quad B_b = K_b Y, \quad K_b = (x' \Omega^{-1} x + \gamma I)^{-1} x' \Omega^{-1},$$

$$(2b) \quad E(B_b) = K_b x \beta, \quad \text{ob}(B_b) = K_b x \beta - \beta = (K_b x - I) \beta,$$

$$(3b) \quad \mathcal{S}(B_b) = K_b \Omega K_b',$$

$$(4b) \quad P_{B_b} = \mathcal{H}_{B_b}^c(K_b x \beta, K_b \Omega K_b'),$$

$$(5b) \quad \text{MSE}(B_b) = \text{tr}(K_b \Omega K_b') + \beta'(K_b x - I)'(K_b x - I) \beta,$$

$$(6b) \quad B'_b B_b = Y' K'_b K_b Y,$$

$$(7b) \quad E(B'_b B_b) = \text{tr}(K'_b K_b \Omega) + \beta' x' K'_b K_b x \beta,$$

$$(8b) \quad \text{var}(B'_b B_b) = 2\text{tr}(K'_b K_b \Omega)^2 + 4\beta' x' K'_b K_b \Omega K'_b K_b x \beta,$$

$$(9b) \quad P_{B'_b B_b} / \sigma^2 \neq \chi^2(\dots) \text{ due to } \Omega K'_b K_b \Omega K'_b K_b \Omega \neq \Omega K'_b K_b \Omega,$$

$$(10b) \quad \hat{Y}_b = x B_b = x K_b Y,$$

$$(11b) \quad E(\hat{Y}_b) = x K_b x \beta, \quad \text{ob}(\hat{Y}_b) = (x K_b - I) x \beta,$$

$$(12b) \quad \mathcal{S}(\hat{Y}_b) = x K_b \Omega K'_b x',$$

$$(13b) \quad P_{\hat{Y}_b} = \sigma^2_{\hat{Y}_b} (\mathbf{x}K_b\mathbf{x}\beta, \mathbf{x}K_b\Omega K_b'\mathbf{x}')$$

$$(14b) \quad \text{MSE}(\hat{Y}_b) = \text{tr}(\mathbf{x}K_b\Omega K_b'\mathbf{x}') + \beta'\mathbf{x}'(\mathbf{x}K_b - \mathbf{I})'(\mathbf{x}K_b - \mathbf{I})\mathbf{x}\beta,$$

$$(15b) \quad \hat{Y}_b'\hat{Y}_b = \mathbf{Y}'K_b'\mathbf{x}'\mathbf{x}K_b\mathbf{Y},$$

$$(16b) \quad \mathcal{E}(\hat{Y}_b'\hat{Y}_b) = \text{tr}(K_b'\mathbf{x}'\mathbf{x}K_b\Omega) + \beta'\mathbf{x}'K_b'\mathbf{x}'\mathbf{x}K_b\mathbf{x}\beta,$$

$$(17b) \quad \text{var}(\hat{Y}_b'\hat{Y}_b) = 2\text{tr}(K_b'\mathbf{x}'\mathbf{x}K_b)^2 + 4\beta'\mathbf{x}'K_b'\mathbf{x}'\mathbf{x}K_b\Omega K_b'\mathbf{x}'\mathbf{x}K_b\mathbf{x}\beta,$$

$$(18b) \quad P_{\hat{Y}_b'\hat{Y}_b/\sigma^2} \neq \chi^2(\dots) \text{ since } \Omega K_b'\mathbf{x}'\mathbf{x}K_b\Omega K_b'\mathbf{x}'\mathbf{x}K_b\Omega \neq \Omega K_b'\mathbf{x}'\mathbf{x}K_b\Omega$$

$$(19b) \quad \mathbf{E}_b = (\mathbf{I} - \mathbf{x}K_b)\mathbf{Y} = \mathbf{M}_b\mathbf{Y}, \quad \mathbf{M}_b = \mathbf{I} - \mathbf{x}K_b,$$

$$(20b) \quad \mathcal{E}(\mathbf{E}_b) = \mathbf{M}_b\mathbf{x}\beta \neq \mathbf{0}, \quad \text{ob}(\mathbf{E}_b) = \mathbf{M}_b\mathbf{x}\beta \neq \mathbf{0},$$

$$(21b) \quad \mathcal{S}(\mathbf{E}_b) = \mathbf{M}_b\Omega\mathbf{M}_b',$$

$$(22b) \quad P_{\mathbf{E}_b} = \sigma^2_{\mathbf{E}_b} (\mathbf{M}_b\mathbf{x}\beta, \mathbf{M}_b\Omega\mathbf{M}_b'),$$

$$(23b) \quad \text{MSE}(\mathbf{E}_b) = \text{tr}(\mathbf{M}_b\Omega\mathbf{M}_b') + \beta'\mathbf{x}'\mathbf{M}_b'\mathbf{M}_b\mathbf{x}\beta,$$

$$(24b) \quad \mathbf{E}_b'\mathbf{E}_b = \mathbf{Y}'\mathbf{M}_b'\mathbf{M}_b\mathbf{Y},$$

$$(25b) \quad \mathcal{E}(\mathbf{E}_b'\mathbf{E}_b) = \text{tr}(\mathbf{M}_b'\mathbf{M}_b\Omega) + \beta'\mathbf{x}'\mathbf{M}_b'\mathbf{M}_b\mathbf{x}\beta,$$

$$(26b) \quad \text{var}(\mathbf{E}_b'\mathbf{E}_b) = 2\text{tr}(\mathbf{M}_b'\mathbf{M}_b\Omega)^2 + 4\beta'\mathbf{x}'\mathbf{M}_b'\mathbf{M}_b\Omega\mathbf{M}_b'\mathbf{M}_b\mathbf{x}\beta,$$

$$(27b) \quad P_{\mathbf{E}_b'\mathbf{E}_b/\sigma^2} \neq \chi^2(\dots) \text{ since } \Omega\mathbf{M}_b'\mathbf{M}_b\Omega\mathbf{M}_b'\mathbf{M}_b\Omega \neq \Omega\mathbf{M}_b'\mathbf{M}_b\Omega,$$

$$(28b) \quad \text{cov}(\mathbf{E}_b, \mathbf{S}_{\mathbf{E}_b}^2) = \text{cov}(K_b\mathbf{Y}, \frac{1}{n_b}\mathbf{E}_b'\mathbf{E}_b) = \frac{2}{n_b}K_b\Omega\mathbf{M}_b'\mathbf{M}_b\mathbf{x}\beta \neq \mathbf{0},$$

Under the assumptions

$$(29) \quad \text{plim} \frac{1}{\sqrt{n}} (\mathbf{X}'_{(n)}\Omega_{(n)}^{-1}\mathbf{X}_{(n)})^{-1} = \mathbf{Q}_a^{-1} \neq \mathbf{0},$$

$$(\text{or plim} \frac{1}{\sqrt{n}} \mathbf{X}'_{(n)}\Omega_{(n)}\mathbf{X} = \mathbf{Q}_a \text{ by nonsingularity of } \mathbf{Q}_a^{-1}),$$

$$\infty > |x_{i,1,(n)}|, \quad |w_{i,1,(n)}| < \infty, \quad \forall i, 1, n \in N = \{1, \dots, \}$$

$$(30) \quad \text{plim} \frac{1}{\sqrt{n}} ((X'_{(n)} \Omega_{(n)}^{-1} X_{(n)} + \gamma I)^{-1} = Q_b^{-1} \neq 0,$$

$$(31) \quad \text{plim} \frac{1}{\sqrt{n}} (X'_{(n)} \Omega_{(n)}^{-1} X_{(n)} + \Gamma)^{-1} = Q_c^{-1} \neq 0,$$

$$(32) \quad \text{plim} \frac{1}{\sqrt{n}} (X'_{(n)} \Omega_{(n)}^{-1} E_{(n)}) = 0,$$

by the same arguments as in Anderson, Taylor work [1], Milo's works [4] the estimators  $B_a, B_b$  are consistent, that is,  $\text{plim} B_a = \text{plim} B_b = \beta$

The relations (1a) - (28a) prove.

Theorem 1. Let the assumptions of model  $NM_1$  and (29), (32) hold. Then

- $B_a$  is unbiased, consistent, efficient and normally distributed; the quadratic form  $B'_a B_a$  does not have  $\chi^2$  - distribution;
- $\hat{Y}_a$  is unbiased, consistent, normally distributed predictor; the quadratic form  $\hat{Y}'_a \hat{Y}_a$  does not have  $\chi^2$  - distribution;
- $E_a$  is unbiased, consistent, normally distributed residual vector; the quadratic form  $E'_a E_a$  does not have  $\chi^2$  - distribution;
- $\text{cov}(B_a, S_{E_a}^2) = 0$ .  $\blacklozenge$

[Note: consistency of  $\hat{Y}_a$  follows from consistency of  $B_a$  and  $\hat{Y} = xB_a$ ; consistency of  $E_a$  follows from consistency of  $B_a$  and the fact that  $E_a = M_a Y$ ; efficiency of  $B_a$  follows from the fact that for each  $K_a^*$ ,  $K_a^* = K_a + C$  it is  $\mathcal{S}(K_a^*) > \mathcal{S}(K_a Y)$ , where  $C \in R^{k \times n}$ ].

The relations (1b)-(28b) prove.

Theorem 2. Let the assumptions of model  $NM_2$  and (29), (30), (32) be fulfilled. Then

- the estimator  $B_b$  is biased, consistent and it has multivariate normal distribution but  $\|B_b\|^2$  does not have  $\chi^2$  - distribution;
- the ex-post predictor  $\hat{Y}_b$  is biased, consistent, and it has n-variate normal distribution but  $\|\hat{Y}_b\|^2$  does not have  $\chi^2$  - distribution;
- the ex-post residual vector  $E_b$  is biased, consistent, and it has multivariate singular normal distribution but  $\|E_b\|^2$  does not have  $\chi^2$  - distribution;



d) the covariance of  $B_b$  and  $S_b^2$  is different from the zero vector.  $\diamond$

Denoting

$$A_1 = (x' \Omega^{-1} x)^{-1}, \quad A_2 = (x' \Omega^{-1} x + \gamma I)^{-1},$$

it is easily seen that  $A_1 = A_1'$ ,  $A_2 = A_2'$ , and  $A_1 A_2$  are positive definite matrices. Hence  $\det(A_1) > 0$ ,  $\det(A_2) > 0$ ,  $A_1 A_2 = A_2 A_1$ . Therefore, an orthogonal matrix  $T$  diagonalizes simultaneously matrices  $A_1$  and  $A_2$ , i.e.

$$(33) \quad T' A_1 T = \Lambda_1 = \text{diag} \left( \frac{1}{\lambda_1}, \dots, \frac{1}{\lambda_k} \right),$$

where  $\lambda_1, 1 = 1, \dots, k$  are given eigen values of the matrix  $x' \Omega^{-1} x$ , and

$$T' A_2 T = \Lambda_2 = \text{diag} \left( \frac{1}{\lambda_1 + \gamma}, \dots, \frac{1}{\lambda_k + \gamma} \right).$$

Recall that  $K_a = A_1 x' \Omega^{-1}$ ,  $K_b = A_2 x' \Omega^{-1}$ ,  $\mathcal{S}(B_a) = A_1$ ,  $\mathcal{S}(B_b) = A_2$ ,  $K_b \Omega K_b' = A_2 A_1^{-1} A_2' = A_2 A_1^{-1} A_2$ ,  $K_b x = A_2 A_1^{-1} x$ .

Hence

$$\mathcal{E}(B_b) \leq \mathcal{E}(B_a) \iff K_b x \beta \leq \beta \iff A_2 A_1^{-1} \beta \leq \beta \iff \beta > 0$$

$$\beta' (A_2 A_1^{-1}) \beta \leq \beta' \beta \iff \det(A_2 A_1^{-1}) \leq \det(I) \iff$$

$$(34a) \quad \det(A_2 A_1^{-1}) \leq 1.$$

Similarly,

$$(34b) \quad \mathcal{E}(B_b) > \mathcal{E}(B_a) \iff \det(A_2 A_1^{-1}) > 1 \quad \text{under } \beta > 0.$$

By definition of  $\mathcal{S}(B_a)$ ,  $\mathcal{S}(B_b)$  we have

$$(34c) \quad \mathcal{S}(B_b) \geq \mathcal{S}(B_a) \iff A_2 A_1^{-1} A_2 \geq A_1 \iff \det(A_2 A_1^{-1})^2 \geq 1,$$

$$(34d) \quad \mathcal{S}(B_b) < \mathcal{S}(B_a) \iff A_2 A_1^{-1} A_2 < A_1 \iff \det (A_2 A_1^{-1})^2 < 1.$$

By definition of  $MSE(B_a)$ ,  $MSE(B_b)$  we get  $MSE(B_b) = \text{tr}(A_2 A_1^{-1} A_2) + \beta'(A_2 A_1^{-1} - I)'(A_2 A_1^{-1} - I)\beta$ ,  $MSE(B_a) = \text{tr}(A_1)$ .

Hence the condition  $MSE(B_b) \leq MSE(B_a)$  holds iff

$$(34c) \quad \text{tr}(A_2 A_1^{-1} A_2 - A_1) \leq -\beta'(A_2 A_1^{-1} - I)'(A_2 A_1^{-1} - I)\beta.$$

The collected conditions (34a)-(34c) give

**Theorem 3.** Let the assumptions of theorems 1 and 2 be satisfied. Then

$$a) \quad \mathcal{E}(B_b) \leq \mathcal{E}(B_a) \iff \det(A_2 A_1^{-1}) \leq 1 \text{ under } \beta > 0,$$

$$b) \quad \mathcal{E}(B_b) > \mathcal{E}(B_a) \iff \det(A_2 A_1^{-1}) > 1 \text{ under } \beta > 0,$$

c)  $\mathcal{S}(B_b) < \mathcal{S}(B_a)$  in a sense that  $\mathcal{S}(B_b) - \mathcal{S}(B_a)$  is negative definite matrix  $\iff \det(A_2 A_1^{-1})^2 < 1$ ,

d)  $\mathcal{S}(B_b) \geq \mathcal{S}(B_a)$ , i.e.  $\mathcal{S}(B_b) - \mathcal{S}(B_a)$  is non-negative definite matrix  $\iff \det(A_2 A_1^{-1})^2 \geq 1$ ,

$$e) \quad MSE(B_b) \leq MSE(B_a) \iff \text{tr}(A_2 A_1^{-1} A_2 - A_1) \leq -\beta'(A_2 A_1^{-1} - I)'(A_2 A_1^{-1} - I)\beta. \quad \blacklozenge$$

Due to the fact  $A_1, A_2 \in R^{k \times k}$ , by Cauchy's theorem, it follows

$$\det(A_2 A_1^{-1}) = \det(A_2) \det(A_1^{-1}),$$

and by (33)

$$\det(A_2) = \det(T' A_2 T) = \det \Lambda_2 = \prod_{i=1}^k \frac{1}{(\lambda_1 + \gamma)}$$

$$\det(A_1) = \det(T' A_1 T) = \det \Lambda_1 = \prod_{i=1}^k \frac{1}{\lambda_1}.$$



Hence the necessary and sufficient conditions (a)-(d) from the theorem 3 can be replaced as follows

$$\det (A_2 A_1^{-1}) \leq 1 \quad \text{to} \quad \prod_{i=1}^k \lambda_i (\lambda_i + \gamma)^{-1} \leq 1,$$

$$\det (A_2 A_1^{-1}) > 1 \quad \text{to} \quad \prod_{i=1}^k \lambda_i (\lambda_i + \gamma)^{-1} > 1,$$

$$\det (A_2 A_1^{-1})^2 \leq 1 \quad \text{to} \quad \prod_{i=1}^k \left( \frac{\lambda_i}{\lambda_i + \gamma} \right)^2 \leq 1.$$

[Note: for practical purposes one can use instead of an orthogonal matrix T an orthonormal matrix T calculated by the use of Jacobi algorithm].

### 3. Properties of estimator $B_0$

Using the relations (I)-(VI) from § 2, the assumptions of model  $NM_2$  and the definitions of symbols:  $\mathcal{E}$ ,  $\mathcal{S}$ , var, MSE, P, cov, one can find that

$$(1c) \quad B_0 = K_0 Y, \quad K_0 = (x' \Omega^{-1} x + \Gamma)^{-1} x' \Omega^{-1},$$

$$(2c) \quad \mathcal{E}(B_0) = K_0 x \beta, \quad \text{ob}(B_0) = (K_0 x - I) \beta \neq 0,$$

$$(3c) \quad \mathcal{S}(B_0) = K_0 \Omega K_0',$$

$$(4c) \quad P_{B_0} = \mathcal{H}_{B_0}^c (K_0 x \beta, K_0 \Omega K_0'),$$

$$(5c) \quad \text{MSE}(B_0) = \text{tr}(K_0 \Omega K_0') + \beta' (K_0 x - I)' (K_0 x - I) \beta,$$

$$(6c) \quad B_0' B_0 = Y' K_0' K_0 Y,$$

$$(7c) \quad \mathcal{E}(B_0' B_0) = \text{tr}(K_0' K_0 \Omega) + \beta' x' K_0' K_0 x \beta,$$

$$(8c) \quad \text{var} (B'_c B_c) = 2 \text{tr} (K'_c K_c \Omega)^2 + 4 \beta' x' K'_c K_c \Omega K'_c K_c x \beta,$$

$$(9c) \quad P_{B'_c B_c / \sigma^2} \neq \chi^2(\dots) \text{ since } \Omega K'_c K_c \Omega K'_c K_c \Omega \neq \Omega K'_c K_c \Omega,$$

$$(10c) \quad \hat{Y}_c = x B_c = x K_c Y,$$

$$(11c) \quad E(Y_c) = x K_c x \beta, \quad \text{ob}(\hat{Y}_c) = (x K_c - I) x \beta \neq 0,$$

$$(12c) \quad S(Y_c) = x K_c \Omega K'_c x',$$

$$(13c) \quad P_{\hat{Y}_c} = \mathcal{N}_{\hat{Y}_c}^c(x K_c x \beta, x K_c \Omega K'_c x'),$$

$$(14c) \quad \text{MSE}(\hat{Y}_c) = \text{tr}(x K_c \Omega K'_c x') + \beta' x' (x K_c - I)' (x K_c - I) x \beta,$$

$$(15c) \quad \hat{Y}'_c \hat{Y}_c = Y' K'_c x' x K_c Y,$$

$$(16c) \quad E(\hat{Y}'_c \hat{Y}_c) = 2 \text{tr}(K'_c x' x K_c \Omega) + \beta' x' K'_c x' x K_c x \beta,$$

$$(17c) \quad \text{var}(\hat{Y}'_c \hat{Y}_c) = 2 \text{tr}(K'_c x' x K_c \Omega)^2 + 4 \beta' x' K'_c x' x K_c \Omega K'_c x' x K_c x \beta,$$

$$(18c) \quad P_{\hat{Y}'_c \hat{Y}_c / \sigma^2} \neq \chi^2(\dots) \text{ since } \Omega K'_c x' x K_c \Omega K'_c x' x K_c \Omega \neq \Omega K'_c x' x K_c \Omega,$$

$$(19c) \quad E_c = (I - x K_c) Y = M'_c Y, \quad M_c = I - x K_c,$$

$$(20c) \quad E(E_c) = M_c x \beta \neq 0, \quad \text{ob}(E_c) = M_c x \beta \neq 0,$$

$$(21c) \quad S(E_c) = M_c \Omega M'_c,$$

$$(22c) \quad P_{E_c} = \mathcal{N}_{E_c}^c(M_c x \beta, M_c \Omega M'_c),$$

$$(23c) \quad \text{MSE}(E_c) = \text{tr}(M_c \Omega M'_c) + \beta' x' M'_c M_c x \beta,$$

$$(24c) \quad E'_c E_c = Y' M'_c M_c Y,$$

$$(25c) \quad E(E'_c E_c) = \text{tr}(M'_c M_c \Omega) + \beta' x' M'_c M_c x \beta,$$

$$(26c) \quad \text{var}(E'_c E_c) = 2\sigma^2 (M'_c M_c \Omega)^2 + 4\beta' x' M'_c M_c M'_c M_c x \beta,$$

$$(27c) \quad P_{E'_c E_c / \sigma^2} \neq \chi^2(\dots), \text{ since } \Omega M'_c M_c \Omega M'_c M_c \Omega \neq \Omega M'_c M_c \Omega,$$

$$(28c) \quad \text{cov}(B_c, S_{E_c}^2) = \frac{2}{n_c} K_c \Omega M'_c M_c x \beta \neq 0, \quad n_c = \text{tr}(M'_c M_c \Phi), \\ \Omega = \sigma^2 \Phi.$$

The relations (1c)-(28c) prove the following.

Theorem 4. Let the assumptions of model  $NM_2$  and the assumptions (29), (31), (32) be satisfied. Then

a) the estimator  $B_c$  is biased, consistent, it has multivariate normal distribution but  $\|B_c\|^2$  does not have  $\chi^2$ -distribution;

b) the predictor  $\hat{Y}_c$  is biased, consistent, it has n-variate normal distribution, but the square of length of it does not have  $\chi^2$  distribution;

c) the residual vector  $E_c$  is biased, consistent, it has n-variate normal distribution with  $\|E_c\|^2$  not  $\chi^2$ -distributed;

d) the covariance of  $B_c$  and  $S_{E_c}^2$  is different from the zero vector. ♦

Since the matrix  $A_3$ ,  $A_3 = (x' \Omega^{-1} x + \Gamma)^{-1}$ , is symmetric and positive definite, therefore  $\det(A_3) > 0$ . Simultaneous diagonalization is possible only in the case of diagonality of matrix  $\Gamma$ . For non-diagonal matrices  $\Gamma$  it holds  $A_1 A_3 \neq A_3 A_1$ . There, however, exists (see th. 6 in ch. 4 of Bellman book [2]) a non-singular matrix  $\tilde{T}$  such that, due to symmetry and positive definiteness of  $A_1$ ,  $A_3$ ,

$$(35) \quad \tilde{T}' A_1 \tilde{T} = I, \quad \tilde{T}' A_3 \tilde{T} = \tilde{\Lambda}, \quad \tilde{\Lambda} = \text{diag}(\tilde{\lambda}_1, \dots, \tilde{\lambda}_k).$$

Noticing that  $K_c = A_3 x' \Omega^{-1}$ ,  $K_c \Omega K'_c = A_3 A_1^{-1} A_3 = \mathcal{L}(B_c)$ ,  $K_c x = A_3 A_1^{-1}$ ,  $\det(A_3 A_1^{-1}) > 0$ ,  $\det(A_3^2 A_1^{-1}) > 0$  it is seen that

$$(36a) \quad \mathcal{E}(B_c) \leq \mathcal{E}(B_a) \stackrel{\beta > 0}{\iff} A_3 A_1^{-1} \beta \leq \beta \iff \det(A_3 A_1^{-1}) \leq 1,$$

$$(36b) \quad \mathcal{E}(B_c) > \mathcal{E}(B_a) \stackrel{\beta > 0}{\iff} \det(A_3 A_1^{-1}) > 1,$$



$$(36c) \quad \mathcal{S}(B_c) \leq \mathcal{S}(B_a) \iff A_3 A_1^{-1} A_3 \leq A_1 \iff \det (A_3 A_1^{-1})^2 \leq 1$$

in the sense that the matrix  $\mathcal{S}(B_c) - \mathcal{S}(B_a)$  is non-positive (positive) definite,

$$(36d) \quad \text{MSE}(B_c) \leq \text{MSE}(B_a) \iff \text{tr}(A_3 A_1^{-1} A_3 - A_1) \leq -\beta'(A_3 A_1^{-1} - I)' \\ (A_3 A_1^{-1} - I)\beta.$$

In the case of diagonal matrix  $\Gamma = \text{diag}(\gamma_1, \dots, \gamma_k)$  the conditions of th. 4 can be rewritten as.

$$\det(A_3 A_1^{-1}) \geq 1 \quad \text{to} \quad \prod_{i=1}^k \frac{\lambda_i}{\lambda_i + \gamma_i} \geq 1, \\ \det(A_3 A_1^{-1})^2 \leq 1 \quad \text{to} \quad \prod_{i=1}^k \left( \frac{\lambda_i}{\lambda_i + \gamma_i} \right)^2 \leq 1,$$

where

$\lambda_i, i = \overline{1, k}$  denote eigen values of  $A_1$ .

It is convenient to put

$$\tilde{A}_1 = \tilde{T}' A_1 \tilde{T} = I, \quad \tilde{A}_3 = \tilde{T}' A_3 \tilde{T} = \Lambda.$$

The matrix  $T$  can be obtained as, f.e.,  $\tilde{T} = S_1 S_2 S_3$ , where  $S_1$  is an orthogonal matrix that diagonalizes  $A_1$  (but does not diagonalize  $A_3$  except in the case of diagonal matrix  $\Gamma$ ),  $S_2$  is a diagonal matrix, i.e.  $S_2 = \text{diag}(\frac{1}{\sqrt{\lambda_1}}, \dots, \frac{1}{\sqrt{\lambda_k}})$ ,  $\lambda_i, i = \overline{1, k}$ , are eigen values, of  $A_1$ ,  $S_3$  is an orthogonal matrix that diagonalizes the matrix  $S_2' S_1' A_3 S_1 S_2$ . By (35) and Cauchy's theorem it follows

$$\det(\tilde{A}_3) = \det(\tilde{T}' A_3 \tilde{T}) = \det(\tilde{\Lambda}) = \prod_{i=1}^k \tilde{\lambda}_i, \quad \det(\tilde{A}_1) = 1.$$

Thus

$$(36e) \quad (\det(A_3 A_1^{-1}) \geq 1) \iff \left( \prod_{i=1}^k \tilde{\lambda}_i \geq 1 \right)$$

$$(36f) \quad (\det (A_3 A_1^{-1})^2 \leq 1) \Leftrightarrow (\prod_{i=1}^k \tilde{\lambda}_i^2 \leq 1),$$

where  $\tilde{\lambda}_i, i = 1, k$ , are eigen values of the matrix  $S_2' \cdot S_1' A_3 S_1 S_2$ . It was proved.

Theorem 5. Let the assumptions of th. 1 and 4 be satisfied. Then

$$a) \quad \varepsilon(B_o) \leq \varepsilon(B_a) \Leftrightarrow \prod_{i=1}^k \tilde{\lambda}_i \leq 1, \text{ under } \beta > 0,$$

$$b) \quad \mathcal{L}(B_o) \leq \mathcal{L}(B_a) \Leftrightarrow \prod_{i=1}^k \tilde{\lambda}_i^2 \leq 1, \text{ under } \beta > 0.$$

$$c) \quad \text{MSE}(B_o) \leq \text{MSE}(B_a) \Leftrightarrow \text{tr}(A_3 A_1^{-1} A_3 - A_1) \leq -\beta'(A_3 A_1^{-1} - I)' (A_3 A_1^{-1} - I) \beta. \quad \blacklozenge$$

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## O OBCIĄŻONYCH ESTYMATORACH REGULARYZUJĄCYCH. CZĘŚĆ II

Celem pracy jest:

- analiza niektórych własności statystycznych regularyzujących estymatorów postaci  $B_b = (x' \Omega^{-1} x + \gamma I)^{-1} x \Omega^{-1} Y$  oraz  $B_c = (x' \Omega^{-1} x + \Gamma)^{-1} x \Omega^{-1} Y$ ,
  - opis zmodyfikowanych planów eksperymentów, których celem będzie rozszerzona analiza własności numeryczno-statystycznych pewnej rodziny regularyzujących estymatorów.
- Samodzielnie udowodniono 5 nowych twierdzeń. Orzekają one:
- a) o obciążoności (nieobciążoności), zgodności wielowymiarowej normalności rozkładów estymatorów  $B_b$ ,  $B_c$  i  $B_a$ , predyktorów  $\hat{y}_a$ ,  $\hat{y}_b$ ,  $\hat{y}_c$  ex post wg  $B_a$ ,  $B_b$ ,  $B_c$ , wektorów reszt ex post wg  $B_a$ ,  $B_b$ ,  $B_c$ ;
  - b) o rozkładach kwadratów długości wektorów  $B_a$ ,  $B_b$ ,  $B_c$ ,  $\hat{y}_a$ ,  $\hat{y}_b$ ,  $\hat{y}_c$ ,  $E_a$ ,  $E_b$ ,  $E_c$ , które nie są rozkładami  $\chi^2$ ;
  - c) o kowariancjach par  $(B_a, S_{E_a}^2)$ ,  $(B_b, S_{E_b}^2)$ ,  $(B_c, S_{E_c}^2)$ , które są odpowiednio wektorem zerowym i wektorami różnymi od zera;
  - d) o warunkach koniecznych i dostatecznych nierówności lub równości między  $\xi(B_{(j)})$  a  $\xi(B_a)$ ,  $\mathcal{S}(B_{(j)})$  a  $\mathcal{S}(B_a)$ ,  $MSE(B_{(j)})$  a  $MSE(B_a)$ , gdzie  $j = b, c$ .