

An eigenvalue problem

Problem 86.2.4

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*Econometric Theory*, 2, 290 (1986)

86.2.3. *Theil's minimum MSE estimator for the standard linear model*, proposed by Heinz Neudecker. Prove that Theil's "estimator" of  $\beta$  in the model  $y = X\beta + \varepsilon$ ,  $E(\varepsilon) = 0$ ,  $D(\varepsilon) = \sigma^2 I$ ;

$$\tilde{\beta} = \beta\beta'X'(X\beta\beta'X' + \sigma^2 I)^{-1}y$$

systematically underestimates  $\beta$ .

#### REFERENCE

H. Theil. *Principles of Econometrics*. (1971), p. 125.

86.2.4. *An Eigenvalue Problem*, proposed by Jan R. Magnus. Let  $a$  and  $b$  be two real  $n \times 1$  vectors, and define the  $n \times n$  matrices  $A = aa' + bb'$  and  $B = aa' - bb'$ .

1. Find the eigenvalues of  $A$  and  $B$ .
2. Show that  $B$  is indefinite (that is, neither positive semidefinite nor negative semidefinite) unless  $a$  and  $b$  are linearly dependent.

Now, let  $x$  and  $y$  be two normally distributed  $n \times 1$  vectors:  $x \cong N(\mu_1, \Omega)$ ,  $y \cong N(\mu_2, \Omega)$  with the same positive definite covariance matrix  $\Omega$ . Let  $V_1$  and  $V_2$  denote the covariance matrices of  $x \otimes x$  and  $y \otimes y$  respectively.

3. Show that  $V_1 - V_2$  is indefinite unless  $\mu_1$  and  $\mu_2$  are linearly dependent.

86.2.5. *The distribution of LIML in the leading case*, proposed by R. W. Farebrother. Let

$$y_1 = Y_2\beta + Z_1\gamma + u$$

be the first equation of a system of simultaneous equations with reduced form

$$Y = Z_1\Pi_1 + Z_2\Pi_2 + V$$

where  $y_1$ ,  $Y_2$  and  $Y = [y_1 \ Y_2]$  are  $T \times 1$ ,  $T \times n$  and  $T \times m$  matrices of observations on the endogenous variables,  $Z_1$ ,  $Z_2$  and  $Z = [Z_1 \ Z_2]$  are  $T \times K_1$ ,  $T \times K_2$  and  $T \times K$  matrices of observations on the predetermined variables,  $u$  and  $V$  are  $T \times 1$  and  $T \times m$  matrices of disturbances, and  $\beta$ ,  $\gamma$ ,  $\Pi_1$  and  $\Pi_2$  are  $n \times 1$ ,  $K_1 \times 1$ ,  $K_1 \times m$ , and  $K_2 \times m$  matrices of parameters. The rows of  $V$  are independent random drawings from a multivariate normal distribution with mean zero and variance  $\Omega$ .

## Solution

*Econometric Theory*, 3, 467–469 (1987)

hence

$$\begin{aligned}\tilde{\beta} &= \frac{1}{\sigma^2} \beta \beta' X' \left( I - \frac{1}{\sigma^2 + \beta' X' X \beta} X \beta \beta' X' \right) y, \\ &= \frac{\beta' X' y}{\sigma^2 + \beta' X' X \beta} \beta.\end{aligned}$$

Therefore,

$$E(\tilde{\beta}) = \frac{\beta' X' X \beta}{\sigma^2 + \beta' X' X \beta} \beta < \beta.$$

*Solution 2* (Proposed by R.W. Farebrother): Let  $z = X\beta$ .

Then

$$(\sigma^2 + z'z)z' = z'(\sigma^2 I + zz'),$$

so that

$$z'(\sigma^2 I + zz')^{-1} = (\sigma^2 + z'z)^{-1}z',$$

and Theil's "estimator"

$$\tilde{\beta} = \beta z'(\sigma^2 I + zz')^{-1}y$$

may be rewritten as

$$\tilde{\beta} = \beta(\sigma^2 + z'z)^{-1}z'y.$$

Now  $E(y) = z$ , so that

$$E(\tilde{\beta}) = \beta z'z(\sigma^2 + z'z)^{-1}$$

and Theil's estimator systematically shrinks  $\beta$  towards zero. (Another good solution has been suggested by Badi. H. Baltagi.)

86.2.4. *An Eigenvalue Problem*—Solution, proposed by Jan R. Magnus. The proof rests on the following lemma which can be proved using high-school algebra.

LEMMA. *The two equations in  $x_1$  and  $x_2$*

$$x_1 + x_2 = p, \quad x_1^2 + x_2^2 = q, \quad (x_1, x_2 \in \mathbb{R}),$$

*with  $p$  and  $q$  in  $\mathbb{R}$  having a solution (in  $\mathbb{R}$ ) if and only if  $p^2 \leq 2q$  in which case the solution is*

$$x_{1,2} = \frac{1}{2}p \pm \frac{1}{2}\sqrt{2q - p^2}.$$

Moreover,

- (1) If  $0 < q < p^2$ , then  $x_{1,2}$  are both positive or both negative.
- (2) If  $0 \leq p^2 < q$ , then one is positive and the other negative.
- (3) If  $0 \leq p^2 = q$ , then  $x_1 = 0$  or  $x_2 = 0$  or both.

Proof of (1): The matrix  $A$  has eigenvalues  $\lambda_1, \lambda_2$ , and zero (multiplicity  $n - 2$ ). Further,

$$a'a + b'b = \text{tr}A = \lambda_1 + \lambda_2,$$

$$(a'a)^2 + (b'b)^2 + 2(a'b)^2 = \text{tr}A^2 = \lambda_1^2 + \lambda_2^2.$$

Hence, using the Lemma,

$$\lambda_{1,2} = \frac{a'a + b'b}{2} \pm \frac{1}{2} \sqrt{(a'a - b'b)^2 + 4(a'b)^2}.$$

Similarly,  $B$  has eigenvalues  $\mu_1, \mu_2$ , and zero ( $n - 2$  times) and

$$\mu_{1,2} = \frac{a'a - b'b}{2} \pm \frac{1}{2} \sqrt{(a+b)'(a+b) \cdot (a-b)'(a-b)}.$$

Proof of (2): We have  $p \equiv \text{tr} B = a'a - b'b$  and

$$q \equiv \text{tr} B^2 = (a'a)^2 + (b'b)^2 - 2(a'b)^2.$$

Hence,

$$p^2 - q = -2((a'a)(b'b) - (a'b)^2) \leq 0$$

with equality if and only if  $a$  and  $b$  are linearly dependent.

Proof of (3): From Magnus and Neudecker [1, Lemma 9], we have

$$V_1 = 2N(\Omega \otimes \Omega + \Omega \otimes \mu_1 \mu'_1 + \mu_1 \mu'_1 \otimes \Omega),$$

where  $N = \frac{1}{2}(I + K)$  and  $K$  is the commutation matrix. Hence,

$$\begin{aligned} V_1 - V_2 &= 2N(\Omega \otimes (\mu_1 \mu'_1 - \mu_2 \mu'_2) + (\mu_1 \mu'_1 - \mu_2 \mu'_2) \otimes \Omega), \\ &= 4N(\Omega \otimes B)N, \end{aligned}$$

where  $B = \mu_1 \mu'_1 - \mu_2 \mu'_2$ . If  $\mu_1$  and  $\mu_2$  are linearly dependent, then  $B$  is positive (or negative) semidefinite and thus  $V_1 - V_2$  also. If  $\mu_1$  and  $\mu_2$  are linearly independent, then (2) implies that  $B$  can be written as

$$B = \lambda a a' - \mu b b'$$

with  $\lambda > 0, \mu > 0, a'a = b'b = 1$ , and  $a'b = 0$ . Let  $S = I - a a'$  and  $T = I - b b'$ . Then

$$SBS = -\mu b b', \quad TBT = \lambda a a',$$

so that

$$(\text{vec } S)' (V_1 - V_2) \text{vec } S = 4 \text{tr } \Omega SBS = -4\mu b' \Omega b < 0$$

and

$$(\text{vec } T)' (V_1 - V_2) \text{vec } T = 4 \text{tr } \Omega TBT = 4\lambda a' \Omega a > 0.$$

REFERENCE

1. Magnus, J.R. & H. Neudecker. Symmetry, 0-1 matrices and Jacobians: A review. *Econometric Theory* 2 (1986): 157-190.
2. Teräsvirta T. Superiority comparisons of homogeneous linear estimators. *Communications in Statistics: Theory and Methods*. A11 (1982), 1595-1601.

Note: Heinz Neudecker has noted to the Editor, after he sent the text to the publisher, that part of the questions of this problem have been answered by Timo Teräsvirta [2].

86.2.5. *The Distribution of LIML in the Leading Case* – Solution, proposed by Peter C.B. Phillips. Let  $X = [\overset{1}{X}_1, \overset{n}{X}_2]' \equiv N(0, I_m)$ , where  $m = n + 1$  and where we use the symbol “ $\equiv$ ” to signify equality in distribution. The vector  $h = X(X'X)^{-1/2}$  is distributed uniformly on the unit sphere in  $\mathbb{R}^m$  and we may therefore write the LIML estimator in the form:

$$\tilde{\beta} \equiv -h_2/h_1 = -X_2/X_1 \equiv N(0, I_n)/N(0,1) \equiv \text{multivariate Cauchy.}$$

The final  $\equiv$  follows by elementary integration. Thus, setting  $r = -X_2/X_1$  and  $s = X_1^2$ , we have

$$\begin{aligned} \text{pdf}(r) &= (2\pi)^{-m/2} \int_0^\infty e^{-(1+r'r)s/2} s^{m/2-1} ds, \\ &= \Gamma(m/2) \pi^{-m/2} (1 + r'r)^{-m/2} \end{aligned}$$

as required.

86.2.6. *Moments of OLS and 2SLS Via Fractional Calculus* – Solution, proposed by John L. Knight.

$$\begin{aligned} \text{(i)} \quad \psi(t) &= \int_{-\infty}^\infty \int_{-\infty}^\infty \exp(itz'z_1/z_1'z_1) \cdot \text{pdf}(z_1) \text{pdf}(z) dz_1 dz, \\ &= \int_{-\infty}^\infty \int_{-\infty}^\infty \exp[itx' \partial x / \Delta_x] e^{x'z_1} \text{pdf}(z_1) \cdot \text{pdf}(z) dz_1 dz \Big|_{x=0}. \end{aligned}$$

Integrating with respect to  $z_1$ , we have

$$\int_{-\infty}^\infty e^{x'z_1} \text{pdf}(z_1) dz_1 = e^{x' \bar{z}_1 + x'x/2}$$

due to the normality of  $z_1$ .