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## ECONOMETRIC INSTITUTE

## OLS ESTIMATION IN A MODEL WHERE A MICROVARIABLE IS EXPLAINED BY AGGREGATES AND CONTEMPORANEOUS DISTURBANCES ARE EQUICORRELATED

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OLS ESTIMATION IN A MODEL WHERE A MICROVARIABLE IS EXPIAINED BY AGGREGATES AND CONTEMPORANEOUS DISTURBANCES ARE EQUICORRELATED *

T. Kloek

## Abstract

In the model $y=X \beta+u$ with $E u=0$ and $E u u^{\prime}=\sigma^{2} G$ it is possible that the OLS and GLS estimators are identical, even if $G \neq I$. However, the conditions for this identity do not necessarily imply the second equality sign in $V\left(\hat{\beta}_{O L S}\right)=\sigma^{2}\left(X^{\prime} X\right)^{-1} X^{\prime} G X\left(X^{\prime} X\right)^{-1}=\sigma^{2}\left(X^{\prime} X\right)^{-1}$, the latter being the usual formula for the OLS covariance matrix. This problem is illustrated for a particular model which may be applicable when a microvariable is explained by aggregates and contemporaneous disturbances are equicorrelated.

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[^0]
## 1. INTRODUCTION

In the general case of a linear model $\mathrm{y}=\mathrm{X} \beta+\mathrm{u}$ with $\mathrm{E} u=0$, $E$ uu' $=\sigma^{2} G$ (where $G$ is positive-definite symmetric and $X$ is fixed of order $n \times k$ with full colimn rank) it can be shown (Rao [1]), that the OLS and GLS estimators are identical if and only if matrices $C$ and $D$ exist such that

$$
\begin{equation*}
G=X C X^{\prime}+Z D Z^{\prime}+\sigma^{2} I \tag{1.1}
\end{equation*}
$$

where $Z$ is an $n \times(n-k)$ matrix with full column rank, satisfying $Z^{\prime} X=0$. If both estimators are identical they have the same covariance matrix which may be written as

$$
\begin{equation*}
V\left(\hat{\beta}_{O L S}\right)=\sigma^{2}\left(X^{\prime} X\right)^{-1} X^{\prime} G X\left(X^{\prime} X\right)^{-1}=\sigma^{2} C+\sigma^{2}\left(X^{\prime} X\right)^{-1} \tag{1.2}
\end{equation*}
$$

Hence, the usual formula for the OLS covariance matrix applies if and only if $C=0$, which is true only in a subset of the cases where (1.1) holds. In the present paper we shall consider two particular models. In the first (1.1) does not hold exactly but perhaps approximately, while in the second (which is a special case of the first) (1.1) holds with $c \neq 0$.

In the models we consider we can partition $y, X$ and $u$ as

$$
\mathrm{y}=\left[\begin{array}{c}
\mathrm{y}_{1}  \tag{1.3}\\
\vdots \\
\dot{y}_{\mathrm{T}}
\end{array}\right] \quad \mathrm{x}=\left[\begin{array}{c}
e_{1} x_{1}^{\prime} \\
\vdots \\
e_{\mathrm{T}} \mathrm{x}_{\mathrm{T}}^{\prime}
\end{array}\right] \quad u=\left[\begin{array}{c}
u_{1} \\
\vdots \\
\dot{u}_{T}
\end{array}\right]
$$

$y_{t}, e_{t}$ and $u_{t}$ are column vectors each consisting of $m_{t}$ components $(t=1, \ldots, T) ; e_{t}^{\prime}=[1 \ldots 1]$ and $x_{t}$ is a column vector consisting of $k$ components. So the model may be applicable in case we have $m_{t}$ observations on micro-variables in time period $t$ which are all explained by the same vector of aggregates $x_{t}$. An empirical example can be found in Riddell [2, equations (1) through (12)], where $y_{t}$ is a vector of money wage changes in $m_{t}$ wage contracts for individual decision making units in period t. The explanatory variables are (functions of) aggregates, such as national unemployment and the (expected) consumer price index. So, all elements of $y_{t}$ are explained by identical rows in the $X$ matrix.

The disturbances are assumed to be homoskedastic, equicorrelated within time periods and uncorrelated across time periods. So the covariance matrix $\sigma^{2} G$ is block diagonal and the t-in diagonal block can be written as

$$
\begin{equation*}
E u_{t} u_{t}^{\prime}=G_{t}=(1-\rho) I_{t}+\rho e_{t} e_{t}^{\prime} \tag{1.4}
\end{equation*}
$$

where $I_{t}$ is the unit matrix of order $m_{t}$. This concludes the assumptions for the first model to be considered. The second model is a particular case of the first, the additional assumption being that $m_{t}=m$ ( $\mathrm{t}=1, \ldots, \mathrm{~T}$ ).

For these models we start to derive simple expressions for the OLS and GLS estimators and their covariance matrices. Then we show that these estimators are identical in the second model. Finally we show for the second model that the usual covariance formula for the OLS estimator underestimates the true covariance matrix. The error made in this way may be serious if $\rho m$ is greater than two or three.

## 2. DERIVATION OF RESULTS

The following properties of $G_{t}$ and $G_{t}^{-1}$ are readily verified:

$$
\begin{equation*}
G_{t}^{-1}=\left[I_{t}-\left(\rho / \tau_{t}\right) e_{t} e_{t}^{\prime}\right] /(1-\rho) \tag{2.1}
\end{equation*}
$$

where

$$
\begin{equation*}
\tau_{t}=\rho\left(m_{t}-1\right)+1 \tag{2.2}
\end{equation*}
$$

Furthermore we have

$$
\begin{array}{ll}
G_{t} e_{t}=\tau_{t} e_{t} & e_{t}^{\prime} G_{t} e_{t}=m_{t} \tau_{t} \\
G_{t}^{-1} e_{t}=\left(1 / \tau_{t}\right) e_{t} & e_{t}^{\prime} G_{t}^{-1} e_{t}=m_{t} / \tau_{t} \tag{2.4}
\end{array}
$$

These results may be used to obtain simple expressions for the OLS and GLS estimators, as follows:

$$
\begin{equation*}
\hat{\beta}_{O L S}=\left[\Sigma m_{t} x_{t} x_{t}^{\prime}\right]^{-1} \Sigma m_{t} x_{t} \bar{y}_{t} \tag{2.5}
\end{equation*}
$$

$$
\begin{equation*}
\hat{\beta}_{G L S}=\left[\Sigma\left(m_{t} / \tau_{t}\right) x_{t} x_{t}^{\prime}\right]^{-1} \Sigma\left(m_{t} / \tau_{t}\right) x_{t} \bar{y}_{t} \tag{2.6}
\end{equation*}
$$

where all summations are over $t=1, \ldots, T$ and where $\bar{y}_{t}=e_{t}^{\prime} y_{t} / m_{t}$. It is seen that the individual observations on the dependent variable only enter into (2.5) and (2.6) via the averages $\bar{y}_{t}$. In order to obtain efficient estimates one need not use the individual observations, provided the means are weighted with the appropriate weights $\mathrm{m}_{t} / \tau_{t}$. The corresponding covariance matrices are given by

$$
\begin{gather*}
V_{O L S}=\sigma^{2}\left(\Sigma m_{t} x_{t} x_{t}^{\prime}\right)^{-1} \Sigma m_{t} \tau_{t} x_{t} x_{t}^{\prime}\left(\Sigma m_{t} x_{t} x_{t}^{\prime}\right)^{-1}  \tag{2.7}\\
V_{G L S}=\sigma^{2}\left[\Sigma\left(m_{t} / \tau_{t}\right) x_{t} x_{t}^{\prime}\right]^{-1}
\end{gather*}
$$

The estimators (2.5) and (2.6) are identical if and only if

$$
\begin{equation*}
\left[\Sigma\left(m_{t} / \tau_{t}\right) x_{t} x_{t}^{\prime}\right]^{-1} x_{s}=\left(\Sigma m_{t} x_{t} x_{t}^{\prime}\right)^{-1} \tau_{s} x_{s} \tag{2.9}
\end{equation*}
$$

for $s=1, \ldots, T . A$ special case where this occurs is $m_{t}=m(a l l t)$. One might conjecture that, if (2.9) is mildiy violated, (2.5) will yield a good approximation to (2.6).

Suppose next that the OLS estimator (2.5) has been used and that the standard errors have not been estimated according to (2.7) but using the traditional formula $\sigma^{2}\left(X^{\prime} X\right)^{-1}$ which in the present case amounts to ${ }^{1}$

$$
\begin{equation*}
V_{O L S}^{*}=\sigma^{2}\left(\sum m_{t} x_{t} x_{t}^{\prime}\right)^{-1} \tag{2.10}
\end{equation*}
$$

We analyse the consequences of this for the case of the second model where $m_{t}=m(a l l t)$.

There are two effects: the traditional estimator

$$
\begin{equation*}
\hat{\sigma}^{2}=\frac{1}{m T-k} \sum \hat{u}_{t}^{\prime} \hat{u}_{t} \tag{2.11}
\end{equation*}
$$

is biased downward, while the matrix $\left(\sum m x_{t} x_{t}^{\prime}\right)^{-1}$ underestimates the
1 Upon comparing (1.2), (2.7) and (2.8) it follows that, if $m_{t}=m$ (all $t$ ), $C=(\tau-1)\left(\sum m x_{t} x_{t}^{\prime}\right)^{-1}$. Given this result one may also construct $Z$ and $D$ matrices satisfying (1.1), but this is tedious and not very illuminating.
the correct matrix $\left[\Sigma(m / \tau) x_{t} x_{t}^{1}\right]^{-1}$. The former effect is usually is a minor one, the latter may be quite important, as we shall proceed to show now. We first consider the bias of the traditional estimator (2.11). If Euu' $=\sigma^{2} G$ and $\hat{u}$ is a vector of least-squares residuals, it is easily seen that

$$
\begin{equation*}
\operatorname{E} \hat{u}^{\prime} \hat{u}=\sigma^{2}\left[\operatorname{tr} G-\operatorname{tr}\left(X^{\prime} X\right)^{-1} X^{\prime} G X\right] \tag{2.12}
\end{equation*}
$$

In the present particular case this amounts to

$$
\begin{equation*}
E \hat{u} \cdot \hat{u}=\sigma^{2}(m T-k \tau) \tag{2.13}
\end{equation*}
$$

compare (2.7). So (2.11) has to be multiplied by ( $\mathrm{mT}-\mathrm{k}$ ) $/(\mathrm{mT}-\mathrm{k} \tau)$ to get an unbiased estimator. If $m T$ is not too small this effect will usually not be important. Therefore we shall ignore it in the next paragraph.

The underestimation of the standard errors may be far more serious because of the second effect, i.e. the omission of the factor $\tau$, which was missing in the matrix $\left(\Sigma m x_{t} x_{t}^{\prime}\right)^{-1}$. In Table 1 we have tabulated $\sqrt{ } \tau$, the factor by which the standard errors obtained from (2.10) have to be multiplied.

TABLE 1. $\sqrt{\tau}$ AS A FUNCTION OF $\rho$ AND $m$

| $\rho=$ | 0 | .05 | .10 | .20 | .30 | .50 |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
| $m=10$ | 1 | 1.20 | 1.38 | 1.67 | 1.92 | 2.35 |
| $m=30$ | 1 | 1.57 | 1.97 | 2.61 | 3.11 | 3.94 |
| $m=50$ | 1 | 1.86 | 2.43 | 3.29 | 3.96 | 5.05 |

It is seen that $\rho$ cannot be ignored without serious consequences for the conclusions. ${ }^{2}$ In particular, if $m$ is large, the effect is sizable even for small values of $\rho$ such as .05 .

[^1][1] RAO, C.R., "Least Squares Theory Using an Estimated Dispersion Matrix and Its Application to Measurement of Signals," Proceedings of the Fifth Berkeley Symposium, Vol. I, 355-372, University of California Press (1967).
[2] RIDDELL, W.C., "The Empirical Foundations of the Phillips Curve: Evidence from Canadian Wage Contract Data," Econometrica, 47 (1979), 1-24.

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[^0]:    * An earlier version of this paper was written in the form of a comment on a paper by Riddell [2]. In its present form it is more self-contained so that reading [2] is no longer a prerequisite. The author gratefully acknowledges an extensive comment by Riddell. He is also indebted to R. Harkema, J. Kmenta, P. Kooiman, A. Kunstman, S. Schim van der Loeff, H.K. van Dijk and a referee for valuable comments and sucgestions. None of these persons should be held responsible for remaining errors.

[^1]:    ${ }^{2}$ In Riddell's case the assumptions of the first model are applicable. Since the variation in the $\mathrm{m}_{\mathrm{t}}$ is not too large, we conjecture that the properties of the second model hold approximately. As Riddell's $m_{t}$ are of the order of 10 , the first row of Table 1 has to be used.

