# A simple derivation of the identity between double $h$ vector class and three stage least square estimator 

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#### Abstract

A simple derivation of the identity connecting DhVC and 3 SLS estimators is presented in this article which will be used for studying the properties of other estimators.


Key Words: Three Stage Least Squares (3 SLS), double h - vector class (DhVC), h vector class.
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## INTRODUCTION

Schraf (1976, 1980) presented a full-information relationship between three stage least squares (3SLS) and double h-vector class (DhVC) estimators. The DhVC is full-information generalization of well known h-Class estimators. Following Srivastava and Tiwari (1977), a simple derivation of the identity connecting DhVC and 3 SLS estimators is presented in this article which will useful for the study of the properties the other estimators.

## DOUBLE h-VECTOR CLASS AND THREESTAGE LEAST SQUARES ESTIMATORS

Consider a complete system of T-structural equations in N jointly dependent and $\Lambda$ predetermined variables.
Let the $\mathrm{i}^{\text {th }}$ structural equation be
$y_{i}=\gamma_{i} Y_{i}+\beta_{i} X_{i}+u_{i}$
(1)
or
$\widehat{V}_{l}=\left(\widehat{\mathrm{V}}_{i} O\right)=\left(M Y_{1}, O\right)(5)$

$$
y_{i}=\delta_{i} z_{i}+u_{i} ; Z_{i}=\left(Y_{i} X_{i}\right), \delta_{i}=\left(\gamma_{i}^{\prime} \beta_{i}^{\prime}\right)^{\prime}
$$

(2)

Where, $y_{i}$ is a column vector of T-observations on the jointly dependent variables to be explained, $Y_{i}$ and $X_{i}$ are the matrices of T-observations on $n_{i}$ explanatory jointly dependent and $l_{i}$ explanatory predetermined variables respectively, $\gamma_{i}$ and $\beta_{i}$ are the associated coefficient vector and $u_{i}$ is the column vector of T-structural disturbances. These disturbances are assumed to be temporally independent and generated by a stationary multivariate stochastic process with
$E\left(u_{i}\right)=0$
$E\left(u_{i} u_{j}\right)=\sigma_{i j} I ;(\mathrm{i}, \mathrm{j}=1,2, \ldots, \mathrm{~N})$
Where, $I$ is the identity matrix of order T. Similarly, the predetermined variables are also assumed to be generated by a multivariate stochastic process, independently of the process generating disturbances, with a non-singular moment's matrix. We assume that all the structural equations of the system are identifiable so that the system has been solved initially to eliminate all identities. Finally, we assume that the system can be solved for the jointly dependent variable. Let
$y_{i}=X \Pi_{i}+V_{i}$
bethe reduced form corresponding to the explanatory jointly dependent variable of (1), where X is a $T \times$ $\Lambda$ matrix, assumed to be of full column rank, of the values taken by all $\Lambda$ - predetermined variables and $\Pi_{i}$ is a $\Lambda \times n_{i}$ matrix of parent reduced form coefficients. Let us define the following matrices.
whereO is a $T \times l_{i}$ null matrices, $l_{i}$ being the number of predetermined variables in (1) and
$M=I_{T}-X\left(X^{\prime} X\right)^{-1} X^{\prime}$
writing $\delta^{\prime}=\left(\delta_{1}^{\prime}, \delta_{2}^{\prime}, \ldots, \delta_{N}\right)^{\prime}$, the double h-vector class estimators of the parameter vector $\delta$ is given by
$\hat{\delta}\left(h_{1}, h_{2}\right)=\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right]^{-1}\left[\sum_{k=1}^{N} \sigma^{i k}\left\{Z_{i}^{\prime}-\left(1-h_{i 1}\left(1+h_{k 2}\right)\right) \hat{\bar{V}}_{i}^{\prime}\right\} y_{k}\right]$
where $h_{1}=\left(h_{11}, h_{21}, \ldots, h_{N 1}\right)^{\prime}$ and $h_{2}=\left(h_{12}, h_{22}, \ldots, h_{N 2}\right)^{\prime}$ and are $N \times 1$ vectors which fix DhVC estimator. If $h_{1}=0$ i.e. $h_{i 1}=0, \forall i=1,2, \ldots, N$.
We get 3SLS estimator given by

$$
\begin{equation*}
\hat{\delta}_{3 S L S}=\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right]_{i, k}^{-1}\left[\sum_{k=1}^{N} \sigma^{i k}\left\{Z_{i}^{\prime}-\hat{\bar{V}}_{i}^{\prime}\right\} y_{k}\right]_{i, 1} \tag{8}
\end{equation*}
$$

SIMPLE DERIVATION OF IDENTITY BETWEEN DHVC AND 3SLS ESTIMATORS
Scharf's identity between DhVC and 3SLS estimators is given by

$$
\begin{align*}
& \hat{\delta}\left(h_{1}, h_{2}\right)=\hat{\delta}_{3 S L S} \\
&  \tag{9}\\
& \quad+\left[\sigma^{i k} Z_{i}^{\prime}\left\{I-\left(1-h_{i 1} h_{k 1}\right) M\right\} Z_{k}\right]_{i, k}^{-1}\left[\left\{\sum_{k=1}^{N} \sigma^{i k} h_{i 1}\left(1+h_{k 2}\right) Z_{i}^{\prime} M y_{k}\right\}_{i, 1}-\left(\sigma^{i k} h_{i 1} h_{k 1} Z_{i}^{\prime} M Z_{k}\right)_{i, k} \hat{\delta}_{3 S L S}\right]
\end{align*}
$$

we present a simple derivation of this identity.
Noticing that for any two matrices $Q_{1}$ and $Q_{2}$

$$
\begin{equation*}
\left(Q_{1}+Q_{2}\right)^{-1}=\left[I-\left(Q_{1}+Q_{2}\right)^{-1} Q_{2}\right] Q_{1}^{-1} \tag{10}
\end{equation*}
$$

provided $Q_{1}$ is invertible. Consider

$$
\begin{align*}
& {\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right]_{j, k}^{-1}} \\
& =\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}+\sigma^{i k} h_{i 1} h_{k 1} \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right]_{j, k}^{-1}  \tag{11}\\
& =\left[I-\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right\}_{j, k}^{-1}\left(\sigma^{i k} h_{i 1} h_{k 1} \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right)_{i, k}\right]\left\{\sigma^{i k}\left(Z_{i}^{\prime} Z_{k}-\hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right)\right\}_{i, k}^{-1}
\end{align*}
$$

## So that

$$
\begin{align*}
& {\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime}, \hat{V}_{k}\right)\right]_{j, k}^{-1}\left\{\sum_{k=1}^{N} \sigma^{i k}\left(Z_{i}^{\prime}-\hat{\bar{V}}_{i}^{\prime}\right) y_{k}\right\}_{i, 1}} \\
& =\left[I-\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right\}_{j, k}^{-1}\left(\sigma^{i k} h_{i 1} h_{k 1} \hat{\bar{V}}_{i}^{\prime} \cdot \hat{V}_{k}\right)_{i, k}\right] \hat{\delta}_{3 S L S}  \tag{12}\\
& =\hat{\delta}_{3 S L S}-\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right]_{, k}^{-1}\left(\sigma^{i k} h_{i 1} h_{k 1} \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right)_{i, k} \hat{\delta}_{3 S L S}
\end{align*}
$$

expressing
$\left[\sum_{k=1}^{N} \sigma^{i k}\left\{Z_{i}^{\prime}-\left(1-h_{i 1}\left(1+h_{k 2}\right) \hat{\bar{V}}_{i}^{\prime}\right\} y_{k}\right]_{i, 1}=\left[\sum_{k=1}^{N} \sigma^{i k}\left(Z_{i}^{\prime}-\hat{\bar{V}}_{i}^{\prime}\right) y_{k}\right]_{i, 1}+\left\{\sum_{k=1}^{N} \sigma^{i k} h_{i 1}\left(1+h_{k 2}\right) \hat{\bar{V}}_{1}^{\prime} y_{k}\right\}_{i, 1}\right.$
and using (13), we find from (7)

$$
\begin{aligned}
& \hat{\delta}\left(h_{1}, h_{2}\right)= \hat{\delta}_{3 S L S}-\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right]_{i, k}^{-1}\left(\sigma^{i k} h_{i 1} h_{k 1} \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right)_{i, k} \hat{\delta}_{3 S L S} \\
&+\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right]_{i, k}^{-1}\left[\sum_{k=1}^{N} \sigma^{i k} h_{i 1}\left(1+h_{K 2}\right) \hat{\bar{V}}_{i}^{\prime} y_{k}\right]_{i, 1} \\
&=\hat{\delta}_{3 S L S}+\left[\sigma^{i k}\left\{Z_{i}^{\prime} Z_{k}-\left(1-h_{i 1} h_{k 1}\right) \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right\}\right]_{i, k}^{-1}\left[\left\{\sum_{k=1}^{N} \sigma^{i k} h_{i 1}\left(1+h_{K 2}\right) \hat{\bar{V}}_{i}^{\prime} y_{k}\right\}_{i, 1} \hat{\delta}_{3 S L S}\right](14) \\
& \quad-\left(\sigma^{i k} h_{i 1} h_{k 1} \hat{\bar{V}}_{i}^{\prime} \hat{\bar{V}}_{k}\right)_{i, k} \hat{\delta}_{3 S L S} \\
&=\hat{\delta}_{3 S L S}+\left[\sigma^{i k} Z_{i}^{\prime}\left\{I-\left(1-h_{i 1} h_{k 1}\right) M\right\} Z_{k}\right]_{i, k}\left[\sum_{k=1}^{N} \sigma^{i k} h_{i 1}\left(1+h_{K 2}\right) Z_{i}^{\prime} M y_{k}\right]_{i, 1} \\
& \quad\left(\sigma^{i k} h_{i 1} h_{k 1} Z_{i}^{\prime} M Z_{k}\right)_{i, k} \hat{\delta}_{3 S L S}
\end{aligned}
$$

Which is same as (9).

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