

1. Reduced form equations:

(a) $y_i = \mathbf{z}'_i \mathbf{\Pi} \boldsymbol{\beta} + v_i$

(b) $v_i = \mathbf{u}'_i \boldsymbol{\beta} + e_i$

(c) Let $\mathbf{w}_i = \mathbf{\Pi}' \mathbf{z}_i$ so that $y_i = \mathbf{w}'_i \boldsymbol{\beta} + v_i$. Since $E(\mathbf{w}_i v_i) = 0$ a simple answer is

$$\begin{aligned} \boldsymbol{\beta} &= (E \mathbf{w}_i \mathbf{w}'_i)^{-1} (E(\mathbf{w}_i y_i)) \\ &= (\mathbf{\Pi}' E(\mathbf{z}_i \mathbf{z}'_i) \mathbf{\Pi})^{-1} (\mathbf{\Pi}' E(\mathbf{z}_i y_i)) \end{aligned}$$

More generally, since $E(\mathbf{z}_i v_i) = 0$ the equation is overidentified. So for any weight matrix \mathbf{W} we can also write the coefficient as

$$\begin{aligned} \boldsymbol{\beta} &= (E(\mathbf{w}_i \mathbf{z}'_i) \mathbf{W} E(\mathbf{z}_i \mathbf{w}'_i))^{-1} (E(\mathbf{w}_i \mathbf{z}'_i) \mathbf{W} E(\mathbf{z}_i y_i)) \\ &= (\mathbf{\Pi}' E(\mathbf{z}_i \mathbf{z}'_i) \mathbf{W} E(\mathbf{z}_i \mathbf{z}'_i) \mathbf{\Pi})^{-1} (\mathbf{\Pi}' E(\mathbf{z}_i \mathbf{z}'_i) \mathbf{W} E(\mathbf{z}_i y_i)) \end{aligned}$$

(d) $\mathbf{\Pi} = E(\mathbf{z}_i \mathbf{z}'_i)^{-1} E(\mathbf{z}_i \mathbf{x}'_i)$

(e) The identification condition is $\text{rank}(\mathbf{\Pi}) = k$

2. Estimation of \mathbf{Q}

(a) $\tilde{\mathbf{Q}} = \frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}'_i$

(b) $\hat{\mathbf{Q}} = \frac{1}{J} \sum_{j=1}^J \mathbf{z}_j \mathbf{z}'_j$

(c) As $n \rightarrow \infty$, $\tilde{\mathbf{Q}} \rightarrow_p E(\mathbf{z}_i \mathbf{z}'_i) = \mathbf{Q}$

As $J \rightarrow \infty$, $\hat{\mathbf{Q}} \rightarrow_p E(\mathbf{z}_j \mathbf{z}'_j) = \mathbf{Q}$

(d) Yes, these two limits are the same, because the distributions in the two samples are identical.

(e) $\tilde{\mathbf{Q}}$ is more efficient if $n > J$.

$\hat{\mathbf{Q}}$ is more efficient if $n < J$.

They are equally efficient if $n = J$

3. Estimation of $\boldsymbol{\beta}$ given $\mathbf{\Pi}$

(a) A simple estimator is

$$\tilde{\boldsymbol{\beta}}_1 = \left(\mathbf{\Pi}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}'_i \right) \mathbf{\Pi} \right)^{-1} \left(\mathbf{\Pi}' \left(\sum_{i=1}^n \mathbf{z}_i y_i \right) \right) \tag{1}$$

A GMM estimator is

$$\tilde{\boldsymbol{\beta}} = \left(\mathbf{\Pi}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}'_i \right) \mathbf{W} \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}'_i \right) \mathbf{\Pi} \right)^{-1} \left(\mathbf{\Pi}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}'_i \right) \mathbf{W} \left(\sum_{i=1}^n \mathbf{z}_i y_i \right) \right)$$

The efficient GMM estimator sets $\mathbf{W} = \hat{\boldsymbol{\Omega}}^{-1}$ where $\hat{\boldsymbol{\Omega}}$ is an estimate of $\boldsymbol{\Omega} = E(\mathbf{z}_i \mathbf{z}'_i v_i^2)$. Notice that the error is v_i from the reduced form, not e_i from the structural form. This is because we are estimating $y_i = \mathbf{z}'_i \mathbf{\Pi} \boldsymbol{\beta} + v_i$ not $y_i = \mathbf{x}'_i \boldsymbol{\beta} + e_i$. Using the preliminary estimate (1) we construct $\hat{v}_i = y_i - \mathbf{z}'_i \mathbf{\Pi} \tilde{\boldsymbol{\beta}}$ and

$$\hat{\boldsymbol{\Omega}} = \frac{1}{n-k} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}'_i \hat{v}_i^2$$

Then the efficient estimator is

$$\tilde{\beta}_2 = \left(\mathbf{\Pi}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \mathbf{\Pi} \right)^{-1} \left(\mathbf{\Pi}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\sum_{i=1}^n \mathbf{z}_i y_i \right) \right)$$

(b) Sample 1

(c) As $n \rightarrow \infty$,

$$\tilde{\beta}_1 = \left(\mathbf{\Pi}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \mathbf{\Pi} \right)^{-1} \left(\mathbf{\Pi}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i y_i \right) \right) \xrightarrow{p} (\mathbf{\Pi}' \mathbf{Q} \mathbf{\Pi})^{-1} \mathbf{\Pi}' E(\mathbf{z}_i y_i) = \beta$$

as defined in 1(c). The asymptotic approximation is as n goes to infinity. Also,

$$\begin{aligned} \tilde{\beta}_2 &= \left(\mathbf{\Pi}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \mathbf{\Pi} \right)^{-1} \left(\mathbf{\Pi}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i y_i \right) \right) \\ &\xrightarrow{p} (\mathbf{\Pi}' \mathbf{Q} \mathbf{\Omega}^{-1} \mathbf{Q} \mathbf{\Pi})^{-1} \mathbf{\Pi}' \mathbf{Q} \mathbf{\Omega}^{-1} E(\mathbf{z}_i y_i) = \beta \end{aligned}$$

4. Estimation of $\mathbf{\Pi}$

(a) $\hat{\mathbf{\Pi}} = \left(\sum_{j=1}^J \mathbf{z}_j \mathbf{z}_j' \right)^{-1} \left(\sum_{j=1}^J \mathbf{z}_j \mathbf{x}_j' \right)$

(b) Sample 2

(c) As $J \rightarrow \infty$,

$$\hat{\mathbf{\Pi}} = \left(\frac{1}{J} \sum_{j=1}^J \mathbf{z}_j \mathbf{z}_j' \right)^{-1} \left(\frac{1}{J} \sum_{j=1}^J \mathbf{z}_j \mathbf{x}_j' \right) \xrightarrow{p} E(\mathbf{z}_j \mathbf{z}_j')^{-1} E(\mathbf{z}_j \mathbf{x}_j') = \mathbf{\Pi}$$

as defined in 1(d). The asymptotics is as J goes to infinity.

5. Estimation of β . when $\mathbf{\Pi}$ unknown

(a) $\hat{\beta}_1 = \left(\hat{\mathbf{\Pi}}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Pi}} \right)^{-1} \left(\hat{\mathbf{\Pi}}' \left(\sum_{i=1}^n \mathbf{z}_i y_i \right) \right)$ or

$$\tilde{\beta}_2 = \left(\hat{\mathbf{\Pi}}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Pi}} \right)^{-1} \left(\hat{\mathbf{\Pi}}' \left(\sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\sum_{i=1}^n \mathbf{z}_i y_i \right) \right)$$

This is just the answer in 3(a), replacing the known $\mathbf{\Pi}$ with the estimate $\hat{\mathbf{\Pi}}$

(b) As $\min(n, J) \rightarrow \infty$

$$\hat{\beta}_1 = \left(\hat{\mathbf{\Pi}}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Pi}} \right)^{-1} \left(\hat{\mathbf{\Pi}}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i y_i \right) \right) \xrightarrow{p} (\mathbf{\Pi}' \mathbf{Q} \mathbf{\Pi})^{-1} \mathbf{\Pi}' E(\mathbf{z}_i y_i) = \beta$$

$$\begin{aligned} \tilde{\beta}_2 &= \left(\hat{\mathbf{\Pi}}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Pi}} \right)^{-1} \left(\hat{\mathbf{\Pi}}' \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i \mathbf{z}_i' \right) \hat{\mathbf{\Omega}}^{-1} \left(\frac{1}{n} \sum_{i=1}^n \mathbf{z}_i y_i \right) \right) \\ &\xrightarrow{p} (\mathbf{\Pi}' \mathbf{Q} \mathbf{\Omega}^{-1} \mathbf{Q} \mathbf{\Pi})^{-1} \mathbf{\Pi}' \mathbf{Q} \mathbf{\Omega}^{-1} E(\mathbf{z}_i y_i) = \beta \end{aligned}$$