

- However, β_{MM} is inconsistent when $E(x'u) \neq 0$, i.e.,

$$\beta_{MM} = (X'X)^{-1}X'y = \beta + (X'X)^{-1}X'u = \beta + \left(\frac{1}{n}X'X\right)^{-1}\left(\frac{1}{n}X'u\right) \not\rightarrow \beta.$$

Note as follows:

$$\frac{1}{n}X'u = \frac{1}{n} \sum_{i=1}^n x_i'u_i \longrightarrow E(x'u) \neq 0.$$

In order to obtain a consistent estimator of β , we find the instrumental variable z which satisfies $E(z'u) = 0$.

Let z_i be the i th realization of z , where z_i is a $1 \times k$ vector.

Then, we have the following:

$$\frac{1}{n}Z'u = \frac{1}{n} \sum_{i=1}^n z_i'u_i \longrightarrow E(z'u) = 0.$$

The β which satisfies $\frac{1}{n} \sum_{i=1}^n z_i'u_i = 0$ is denoted by β_{IV} , i.e., $\frac{1}{n} \sum_{i=1}^n z_i'(y_i - x_i\beta_{IV}) = 0$.

Thus, β_{IV} is obtained as:

$$\beta_{IV} = \left(\frac{1}{n} \sum_{i=1}^n z_i' x_i \right)^{-1} \left(\frac{1}{n} \sum_{i=1}^n z_i' y_i \right) = (Z'X)^{-1} Z'y.$$

Note that $Z'X$ is a $k \times k$ square matrix, where we assume that the inverse matrix of $Z'X$ exists.

Assume that as n goes to infinity there exist the following moment matrices:

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n z_i' x_i &= \frac{1}{n} Z'X \longrightarrow M_{zx}, \\ \frac{1}{n} \sum_{i=1}^n z_i' z_i &= \frac{1}{n} Z'Z \longrightarrow M_{zz}, \\ \frac{1}{n} \sum_{i=1}^n z_i' u_i &= \frac{1}{n} Z'u \longrightarrow 0. \end{aligned}$$

As n goes to infinity, β_{IV} is rewritten as:

$$\begin{aligned}\beta_{IV} &= (Z'X)^{-1}Z'y = (Z'X)^{-1}Z'(X\beta + u) = \beta + (Z'X)^{-1}Z'u \\ &= \beta + \left(\frac{1}{n}Z'X\right)^{-1}\left(\frac{1}{n}Z'u\right) \longrightarrow \beta + M_{zx} \times 0 = \beta,\end{aligned}$$

Thus, β_{IV} is a consistent estimator of β .

● We consider the asymptotic distribution of β_{IV} .

By the central limit theorem,

$$\frac{1}{\sqrt{n}}Z'u \longrightarrow N(0, \sigma^2 M_{zz})$$

Note that $V\left(\frac{1}{\sqrt{n}}Z'u\right) = \frac{1}{n}V(Z'u) = \frac{1}{n}E(Z'uu'Z) = \frac{1}{n}E\left(E(Z'uu'Z|Z)\right) = \frac{1}{n}E\left(Z'E(uu'|Z)Z\right) = \frac{1}{n}E(\sigma^2 Z'Z) = E\left(\sigma^2 \frac{1}{n}Z'Z\right) \longrightarrow E(\sigma^2 M_{zz}) = \sigma^2 M_{zz}.$

We obtain the following asymptotic distribution:

$$\sqrt{n}(\beta_{IV} - \beta) = \left(\frac{1}{n}Z'X\right)^{-1}\left(\frac{1}{\sqrt{n}}Z'u\right) \longrightarrow N(0, \sigma^2 M_{zx}^{-1} M_{zz} M_{zx}^{-1'})$$

Practically, for large n we use the following distribution:

$$\beta_{IV} \sim N\left(\beta, s^2(Z'X)^{-1}Z'Z(Z'X)^{-1'}\right),$$

where $s^2 = \frac{1}{n-k}(y - X\beta_{IV})'(y - X\beta_{IV})$.

● In the case where z_i is a $1 \times r$ vector for $r > k$, $Z'X$ is a $r \times k$ matrix, which is not a square matrix. \implies **Generalized Method of Moments (GMM, 一般化積率法)**

4.2 Generalized Method of Moments (GMM, 一般化積率法)

Consider the following regression model:

$$Z'y = Z'X\beta + Z'u,$$

where Z , y , X , β and u are $n \times r$, $n \times 1$, $n \times k$, $k \times 1$ and $n \times 1$ matrices or vectors.

Note that $r \geq k$.

$y^* = Z'y$, $X^* = Z'X$ and $u^* = Z'u$ denote $r \times 1$, $r \times k$ and $r \times 1$ matrices or vectors, where $r \geq k$.

Rewrite as follows:

$$y^* = X^*\beta + u^*,$$

$\Rightarrow r$ is taken as sample size.

Mean and variance of u^* are given by:

$$E(u^*) = 0 \quad \text{and} \quad V(u^*) = E(u^* u^{*\prime}) = \sigma^2 Z'Z = \sigma^2 \Omega.$$

Using GLS, GMM is obtained as:

$$\beta_{GMM} = (X^{*\prime} \Omega^{-1} X^*)^{-1} X^{*\prime} \Omega^{-1} y^* = (X'Z(Z'Z)^{-1}Z'X)^{-1} X'Z(Z'Z)^{-1}Z'y.$$

β_{GMM} is rewritten as:

$$\begin{aligned} \beta_{GMM} &= (X'Z(Z'Z)^{-1}Z'X)^{-1} X'Z(Z'Z)^{-1}Z'y \\ &= (X'Z(Z'Z)^{-1}Z'X)^{-1} X'Z(Z'Z)^{-1}Z'(X\beta + u) \\ &= \beta + (X'Z(Z'Z)^{-1}Z'X)^{-1} X'Z(Z'Z)^{-1}Z'u. \end{aligned}$$

Assume:

$$\frac{1}{n}X'Z \longrightarrow M_{xz}, \quad \frac{1}{n}Z'Z \longrightarrow M_{zz}, \quad \frac{1}{n}Z'u \longrightarrow 0.$$

Then, β_{GMM} is a consistent estimator of β , which is shown as follows:

$$\begin{aligned}\beta_{GMM} &= \beta + \left(\left(\frac{1}{n} X'Z \right) \left(\frac{1}{n} Z'Z \right)^{-1} \left(\frac{1}{n} Z'X \right) \right)^{-1} \left(\frac{1}{n} X'Z \right) \left(\frac{1}{n} Z'Z \right)^{-1} \left(\frac{1}{n} Z'u \right) \\ &\longrightarrow \beta + (M_{xz} M_{zz}^{-1} M'_{xz})^{-1} M_{xz} M_{zz}^{-1} \times 0 = \beta.\end{aligned}$$

● We derive the asymptotic distribution of β_{GMM} .

From the central limit theorem,

$$\frac{1}{\sqrt{n}} Z'u \longrightarrow N(0, \sigma^2 M_{zz}).$$

Accordingly, β_{GMM} is distributed as:

$$\begin{aligned}\sqrt{n}(\beta_{GMM} - \beta) &= \left(\left(\frac{1}{n} X'Z \right) \left(\frac{1}{n} Z'Z \right)^{-1} \left(\frac{1}{n} Z'X \right) \right)^{-1} \left(\frac{1}{n} X'Z \right) \left(\frac{1}{n} Z'Z \right)^{-1} \left(\frac{1}{\sqrt{n}} Z'u \right) \\ &\longrightarrow N\left(0, \sigma^2 (M_{xz} M_{zz}^{-1} M'_{xz})^{-1}\right).\end{aligned}$$

Practically, for large n we use the following distribution:

$$\beta_{GMM} \sim N\left(\beta, s^2(X'Z(Z'Z)^{-1}Z'X)^{-1}\right),$$

where $s^2 = \frac{1}{n-k}(y - X\beta_{GMM})'(y - X\beta_{GMM})$.

● The above GMM is equivalent to 2SLS.

$X: n \times k, \quad Z: n \times r, \quad r > k.$

Assume:

$$\frac{1}{n}X'u = \frac{1}{n} \sum_{i=1}^n x'_i u_i \longrightarrow E(x'u) \neq 0,$$

$$\frac{1}{n}Z'u = \frac{1}{n} \sum_{i=1}^n z'_i u_i \longrightarrow E(z'u) = 0.$$

Regress X on Z , i.e., $X = Z\Gamma + V$ by OLS, where Γ is a $r \times k$ unknown parameter matrix and V is an error term,

Denote the predicted value of X by $\hat{X} = Z\hat{\Gamma} = Z(Z'Z)^{-1}Z'X$, where $\hat{\Gamma} = (Z'Z)^{-1}Z'X$.

Note that 2SLS is equivalent to IV in the case of $Z = \hat{X}$, where this Z is different from the above Z .

This Z is a $n \times k$ matrix, while the above Z is a $n \times r$ matrix.

When Z is a $n \times k$ instrumental variable, the IV estimator is given by:

$$\beta_{IV} = (Z'X)^{-1}Z'y,$$

Z is replaced by \hat{X} . Then,

$$\beta_{2SLS} = (\hat{X}'X)^{-1}\hat{X}'y = (X'Z(Z'Z)^{-1}Z'X)^{-1}X'Z(Z'Z)^{-1}Z'y = \beta_{GMM}.$$

GMM is interpreted as the GLS applied to MM.