

We assume that

$$\frac{1}{n}X'Z \longrightarrow M_{xz} \quad \text{and} \quad \frac{1}{n}Z'\Omega Z \longrightarrow M_{z\Omega z},$$

which are  $k \times r$  and  $r \times r$  matrices.

From the assumption of  $\frac{1}{n}Z'u \longrightarrow 0$ , we have the following result:

$$\beta_{GMM} \longrightarrow \beta + (M_{xz}M_{z\Omega z}^{-1}M'_{xz})^{-1}M_{xz}M_{z\Omega z}^{-1} \times 0 = \beta.$$

Thus,  $\beta_{GMM}$  is a consistent estimator of  $\beta$  (i.e., asymptotically unbiased estimator).

The variance of  $\beta_{GMM}$  is asymptotically obtained as follows:

$$\begin{aligned}
 V(\beta_{GMM}) &= E\left((\beta_{GMM} - E(\beta_{GMM}))(\beta_{GMM} - E(\beta_{GMM}))'\right) \approx E\left((\beta_{GMM} - \beta)(\beta_{GMM} - \beta)'\right) \\
 &= E\left((X'Z(Z'\Omega Z)^{-1}Z'X)^{-1}X'Z(Z'\Omega Z)^{-1}Z'u((X'Z(Z'\Omega Z)^{-1}Z'X)^{-1}X'Z(Z'\Omega Z)^{-1}Z'u)'\right) \\
 &= E\left((X'Z(Z'\Omega Z)^{-1}Z'X)^{-1}X'Z(Z'\Omega Z)^{-1}Z'uu'Z(Z'\Omega Z)^{-1}Z'X(X'Z(Z'\Omega Z)^{-1}Z'X)^{-1}\right) \\
 &\approx (X'Z(Z'\Omega Z)^{-1}Z'X)^{-1}X'Z(Z'\Omega Z)^{-1}Z'E(uu')Z(Z'\Omega Z)^{-1}Z'X(X'Z(Z'\Omega Z)^{-1}Z'X)^{-1} \\
 &= \sigma^2(X'Z(Z'\Omega Z)^{-1}Z'X)^{-1}.
 \end{aligned}$$

Note that  $\beta_{GMM} \rightarrow \beta$  implies  $E(\beta_{GMM}) \rightarrow \beta$  in the 1st line.

$\approx$  in the 4th line indicates that  $Z$  and  $X$  are treated as exogenous variables although they are stochastic.

We assume that  $E(uu') = \sigma^2\Omega$  from the 4th line to the 5th line.

- We derive the asymptotic distribution of  $\beta_{GMM}$ .

From the central limit theorem,

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

Accordingly,  $\beta_{GMM}$  is asymptotically distributed as:

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

Practically, we use:  $\beta_{GMM} \sim N(\beta, s^2(X'Z(Z'\Omega Z)^{-1}Z'X)^{-1})$ ,

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

We may use  $n$  instead of  $n - k$ .

### Identically and Independently Distributed Errors:

- If  $u_1, u_2, \dots, u_n$  are mutually independent and  $u_i$  is distributed with mean zero and variance  $\sigma^2$ , the mean and variance of  $u^*$  are given by:

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

Using GLS, GMM is obtained as:

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

- We derive the asymptotic distribution of  $\beta_{GMM}$ .

From the central limit theorem,

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

Accordingly,  $\beta_{GMM}$  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  $\Gamma$  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  $\hat{\Gamma} = (Z'Z)^{-1}Z'X$ .

**Review — IV estimator:** Consider the regression model is:

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

Assumption:  $E(X'u) \neq 0$  and  $E(Z'u) = 0$ .

The  $n \times k$  matrix  $Z$  is called the instrumental variable (IV).

The IV estimator is given by:

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