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

$$\beta_{MM} = (X'X)^{-1}X'y = \beta + (X'X)^{-1}X'u = \beta + \left(\frac{1}{n}X'X\right)^{-1}\left(\frac{1}{n}X'u\right) \longrightarrow \beta.$$

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:

$$\frac{1}{n}\sum_{i=1}^{n} z_i' x_i = \frac{1}{n}Z'X \longrightarrow M_{zx},$$

$$\frac{1}{n}\sum_{i=1}^{n} z_i' z_i = \frac{1}{n}Z'Z \longrightarrow M_{zz},$$

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

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

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

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

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

By the central limit theorem,

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

Note that
$$V(\frac{1}{\sqrt{n}}Z'u) = \frac{1}{n}V(Z'u) = \frac{1}{n}E(Z'uu'Z) = \frac{1}{n}E(E(Z'uu'Z|Z))$$

= $\frac{1}{n}E(Z'E(uu'|Z)Z) = \frac{1}{n}E(\sigma^2Z'Z) = E(\sigma^2\frac{1}{n}Z'Z) \longrightarrow E(\sigma^2M_{zz}) = \sigma^2M_{zz}.$

We obtain the following asymmptotic distribution:

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

Practically, for large n we use the following distribution:

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

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)

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

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

For now, however, suppose that we have z with E(z'u) = 0.

Let z_i be the *i*th realization (i.e., the *i*th data) of z, where z_i is a $1 \times r$ vector and r > k.

Then, using the law of large number, we have the following:

$$\frac{1}{n}Z'u = \frac{1}{n}\sum_{i=1}^{n} z'_{i}u_{i} = \frac{1}{n}\sum_{i=1}^{n} z'_{i}(y_{i} - x_{i}\beta) \longrightarrow E(z'u) = 0.$$

The number of equations (i.e., r) is larger than the number of parameters (i.e., k).

Let us define W as a $r \times r$ weight matrix, which is symmetric.

We solve the following minimization problem:

$$\min_{\beta} \left(\frac{1}{n} \sum_{i=1}^{n} z_i' (y_i - x_i \beta) \right)' W \left(\frac{1}{n} \sum_{i=1}^{n} z_i' (y_i - x_i \beta) \right),$$

which is equivalent to:

$$\min_{\beta} \left(Z'(y - X\beta) \right)' W \left(Z'(y - X\beta) \right),$$

i.e.,

$$\min_{\beta} (y - X\beta)' ZWZ'(y - X\beta).$$

Note that $\sum_{i=1}^{n} z_i'(y_i - x_i\beta) = Z'(y - X\beta)$.

W should be the inverse matrix of the variance-covariance matrix of $Z'(y-X\beta) = Z'u$.

Suppose that $V(u) = \sigma^2 \Omega$.

Then, $V(Z'u) = E(Z'u(Z'u)') = E(Z'uu'Z) = Z'E(uu')Z = \sigma^2 Z'\Omega Z = W^{-1}$.

The following minimization problem should be solved.

$$\min_{\beta} (y - X\beta)' Z (Z'\Omega Z)^{-1} Z' (y - X\beta).$$

The solution of β is given by the GMM estimator, denoted by β_{GMM} .

Remark: For the model: $y = X\beta + u$ and $u \sim (0, \sigma^2 \Omega)$, solving the following minimization problem:

$$\min_{\beta}(y-X\beta)'\Omega^{-1}(y-X\beta),$$

GLS is given by:

$$b = (X'\Omega^{-1}X)^{-1}X'\Omega^{-1}y.$$

Note that b is the best linear unbiased estimator.