

Generalized Method of Moments

Suppose we have l moment conditions

$$E g(w_i, \theta_0) = 0, \quad \text{where } w_i := (y_i, x_i', z_i')$$

$l \times 1$ $l \times l$

is the data.

Recall the criterion function

$$Q(\theta) = [E g(w_i, \theta)]' A' A [E g(w_i, \theta)]$$

pos def and symmetric

the true parameter θ_0 minimizes $Q(\theta)$ for any choice of norm $A'A$. However, if $E g(w_i, \theta_0) \neq 0$, then different choices of $A'A$ yield a different minimizer. We refer to such value as the pseudo-true value.

GMM focuses on the sample version of the criterion function

$$\hat{\theta}_{GMM}(A_n) := \underset{\theta}{\operatorname{argmin}} \underbrace{\left[\frac{1}{n} \sum_{i=1}^n g(w_i, \theta)' \right] A_n' A_n \left[\frac{1}{n} \sum_{i=1}^n g(w_i, \theta) \right]}_{Q_n(\theta)}$$

In the linear IV model we assume

$$y_i = x_i' \beta + u_i$$

$n \times 1$

$$E z_i u_i = 0, \quad l > n.$$

$l \times 1$

Identification

Assume $E Z_i X_i'$ has full column rank K . To see the implications, let
 $X_i' = \begin{pmatrix} X_{1i}' & X_{2i}' \end{pmatrix}$, and let $Z_i' = \begin{pmatrix} Z_{1i}' & X_{2i}' \end{pmatrix}$.

$1 \times K$ $1 \times k_1$ $1 \times k_2$
included included
endogenous exogenous

1×1 $1 \times k_1$ $1 \times k_2$
excluded included
exogenous exogenous

Let $X_{1i} = \Pi_1 Z_{1i} + \Pi_2 X_{2i}$, which is a linear projection of the element of X_{1i} onto (Z_{1i}, X_{2i}) . Then

$$E Z_i X_i' = E \begin{pmatrix} Z_{1i}' \\ X_{2i}' \end{pmatrix} \begin{pmatrix} X_{1i}' & X_{2i}' \end{pmatrix}$$

$$= \begin{pmatrix} E Z_{1i} Z_{1i}' \Pi_1' + E Z_{1i} X_{2i}' \Pi_2' & E Z_{1i} X_{2i}' \\ E X_{2i} Z_{1i}' \Pi_1' + E X_{2i} X_{2i}' \Pi_2' & E X_{2i} X_{2i}' \end{pmatrix}.$$

Suppose there's a $\theta \neq 0$ such that $E Z_i X_i' \theta = 0$. Then

$$E Z_i X_i' \theta = \begin{pmatrix} E Z_{1i} Z_{1i}' \Pi_1' \theta_1 + E Z_{1i} X_{2i}' \Pi_2' \theta_1 + E Z_{1i} X_{2i}' \theta_2 \\ E X_{2i} Z_{1i}' \Pi_1' \theta_1 + E X_{2i} X_{2i}' \Pi_2' \theta_1 + E X_{2i} X_{2i}' \theta_2 \end{pmatrix}$$

$$= \begin{pmatrix} E Z_{1i} Z_{1i}' & E Z_{1i} X_{2i}' \\ E X_{2i} Z_{1i}' & E X_{2i} X_{2i}' \end{pmatrix} \begin{pmatrix} \Pi_1' \theta_1 \\ \Pi_2' \theta_1 + \theta_2 \end{pmatrix} = 0$$

this second moment matrix is p.d. iff the exogenous variables are all linearly indep

(this requires that Z_{1i} does not form part of X_{2i} , i.e. instruments are not in the outcome model).

\Rightarrow Then this must be 0.

Then

$$\begin{aligned}\Pi_2' \theta_1 + \theta_2 &= 0 \\ \Pi_1' \theta_1 &= 0\end{aligned}$$

Notice that if either θ_j is 0, the other must be zero as well, but this would contradict the fact that $\theta \neq 0$. Hence, suppose $\exists \theta_1 \neq 0$ such that $\Pi_1' \theta_1 = 0$. This means that the matrix Π_1' is rank deficient.

Therefore, we showed that

$$\text{rank}(E Z_i X_i') = k \iff \text{rank}(\Pi_1) = k_1.$$

so that the rank condition is equivalent to another rank condition in terms of the first stage parameters (i.e. sufficiently strong linear relationship between Z_i' and X_i').

From the moment condition we obtain

$$E Z_i Y_i = E Z_i X_i' \beta$$

$$A \quad E Z_i Y_i = \underbrace{A}_{l \times l} \underbrace{E Z_i X_i'}_{l \times k} \beta$$

$$\text{rank} \leq \min(\text{rank}(A), \text{rank}(E Z_i X_i')) = k$$

$$\text{rank} \geq \text{rank}(A) + \text{rank}(E Z_i X_i') - l = k$$

therefore, the Gram matrix is invertible.

$$E X_i Z_i' A' A E Z_i Y_i = E X_i Z_i' A' A E Z_i X_i' \beta$$

$$\Rightarrow \beta(A) = (E X_i X_i' A' A E Z_i X_i')^{-1} E X_i Z_i' A' A E Z_i Y_i$$

Consistency

Assume

(i) $A_n \xrightarrow{p} A$, where A is a finite matrix.

(ii) $E Z_i X_i'$ has rank k

(iii) $E X_{ij}^2 < \infty$, $j=1, \dots, k$

$E Z_{ij}^2 < \infty$, $j=1, \dots, l$

Then $\hat{\beta}_n(A_n) \xrightarrow{p} \beta$, where $\hat{\beta}_n(A_n)$ is defined as follows.

$$\hat{\beta}_n(A_n) = \left(\frac{1}{n} \sum_{i=1}^n X_i Z_i' A_n' A_n \frac{1}{n} \sum_{i=1}^n Z_i X_i' \right)^{-1} \frac{1}{n} \sum_{i=1}^n X_i Z_i' A_n' A_n \frac{1}{n} \sum_{i=1}^n Z_i y_i$$

Notice that $E Z_i X_i' = O(l)$ because

$$\begin{aligned} E \|Z_i X_i'\| &\leq (E \|Z_i\|^2 E \|X_i\|^2)^{1/2} \\ &= O(l) \text{ by (iii).} \end{aligned}$$

Notice that by (ii) we have that $E X_i Z_i' A' A E Z_i X_i'$ is p.d.

To see this

$$A' E Z_i X_i' v \neq 0 \text{ iff } v \neq 0 \text{ by rank cond.}$$

Hence

$$v' E X_i' Z_i A' A E Z_i X_i' v > 0.$$

The implication of this is that $(E X_i Z_i' A' A E Z_i X_i')^{-1} = O(1)$ because the matrix is bounded away from zero.

Write

$$\begin{aligned}\hat{\beta}_n(A_n) &= \beta + \left(\frac{1}{n} \sum_{i=1}^n z_i x_i' A_n' A_n \frac{1}{n} \sum_{i=1}^n z_i x_i' \right)^{-1} \frac{1}{n} \sum_{i=1}^n z_i x_i' A_n' A_n \frac{1}{n} \sum_{i=1}^n z_i u_i \\ &= \beta + \left\{ [E z_i z_i' + o_p(1)] [A_n' A_n] [E z_i x_i' + o_p(1)] \right\}^{-1} [E z_i x_i' + o_p(1)] [A_n' A_n] o_p(1) \\ &= \beta + \left\{ E z_i z_i' A_n' A_n E z_i x_i' + o_p(1) O(1) \right\}^{-1} [E z_i x_i' A_n' A_n o_p(1) + o_p(1) O(1)] \\ &= \beta + \left\{ (E z_i z_i' A_n' A_n E z_i x_i') + o_p(1) \right\} [E z_i x_i' A_n' A_n o_p(1) + o_p(1)] \\ &= \beta + \left\{ O(1) + o_p(1) \right\} [O(1) o_p(1) + o_p(1)] \\ &= \beta + o_p(1) .\end{aligned}$$

Asymptotic Normality

Assume, in addition, the following

$$(iv) \quad E z_{ij}^4 < \infty \quad \text{for all } j=1, \dots, l. \\ E u_i^4 < \infty$$

$$(v) \quad E u_i^2 z_i z_i' \text{ is positive definite.}$$

Notice that (iv) implies that $E u_i^2 z_i z_i'$ is $O(1)$. To see this

$$\begin{aligned} E \|u_i^2 z_i z_i'\| &\leq (E \|u_i\|^2 E \|z_i z_i'\|^2)^{1/2} \\ &= (E u_i^4 E \|z_i\|^4)^{1/2} \\ &= O(1). \end{aligned}$$

Therefore, by the CLT

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n z_i u_i \xrightarrow{d} N(0, \underbrace{E u_i^2 z_i z_i'}_{\Omega})$$

Write

$$\begin{aligned} \sqrt{n} (\hat{\beta}_n(A_n) - \beta_0) &= \left(\frac{1}{n} \sum_{i=1}^n z_i z_i' \quad A_n' A_n \quad \frac{1}{n} \sum_{i=1}^n z_i z_i' \right)^{-1} \frac{1}{n} \sum_{i=1}^n z_i z_i' A_n' A_n \frac{1}{\sqrt{n}} \sum_{i=1}^n z_i u_i \\ &= \left[(\Omega' A' A \Omega)^{-1} \Omega' A' A + o_p(1) \right] \frac{1}{\sqrt{n}} \sum_{i=1}^n z_i u_i \\ &= (\Omega' A' A \Omega)^{-1} \Omega' A' A \frac{1}{\sqrt{n}} \sum_{i=1}^n z_i u_i + o_p(1) O_p(1) \\ &= (\Omega' A' A \Omega)^{-1} \Omega' A' A \frac{1}{\sqrt{n}} \sum_{i=1}^n z_i u_i + o_p(1) \\ &\xrightarrow{d} N(0, (\Omega' A' A \Omega)^{-1} \Omega' A' A \Omega A' A \Omega (\Omega' A' A \Omega)^{-1}). \end{aligned}$$

Efficient GMM

The lower bound on the asymptotic variance is given by

$$(Q' \Omega^{-1} Q)^{-1}.$$

To see why this is true we need to show that

$$\text{Asy Var}(A) - (Q' \Omega^{-1} Q)^{-1} \text{ p.s.d.}$$

\Leftrightarrow

$$Q' \Omega^{-1} Q - (\text{Asy Var}(A))^{-1} \text{ is p.s.d.}$$

Write

$$Q' \Omega^{-1} Q - Q' A' A Q (Q' A' A \Omega A' A Q)^{-1} Q' A' A Q$$

$$= Q' \Omega^{-1/2} \left[I - \underbrace{\Omega^{1/2} A' A Q (Q' A' A \Omega A' A Q)^{-1} Q' A' A \Omega^{1/2}}_H \right] \Omega^{-1/2} Q$$

$$= Q' \Omega^{-1/2} \left[I - \underbrace{H (H' H)^{-1} H'} \right] \Omega^{-1/2} Q .$$

projection matrix = positive semi-definite

The efficient choice of $A = \Omega^{-1/2}$.

Suppose we use a matrix $B_{k \times k}$ that has rank k to linearly transform Z_i into a vector of k instruments $W_i := B Z_i$. We then run IV with W_i .

$$\tilde{\beta}_n(B) = \left(\sum_{i=1}^n W_i X_i' \right)^{-1} \sum_{i=1}^n W_i Y_i = (W'X)^{-1} W'Y$$

$$* W = \begin{pmatrix} W_1' \\ \vdots \\ W_n' \end{pmatrix} = \begin{pmatrix} Z_1' \\ \vdots \\ Z_n' \end{pmatrix} B' = Z B' \quad \begin{matrix} n \times k \\ n \times k \end{matrix}$$

$$\tilde{\beta}_n(B) - \beta = \frac{(W'X)^{-1}}{n} W'u$$

$$\sqrt{n} (\tilde{\beta}_n(B) - \beta) = \frac{(W'X)^{-1}}{n} W'u = \left(B \frac{Z'X}{n} \right)^{-1} B \frac{Z'u}{\sqrt{n}}$$

$$\bullet \frac{Z'u}{\sqrt{n}} = \frac{1}{\sqrt{n}} \sum_{i=1}^n W_i Z_i \xrightarrow{d} N(0, E W_i^2 Z_i Z_i')$$

$$\bullet \frac{Z'X}{n} \xrightarrow{p} E Z_i X_i' := Q$$

$$\text{by Slutsky's theorem } \left(B \frac{Z'X}{n} \right)^{-1} \xrightarrow{p} (B Q)^{-1}$$

Then

$$\begin{aligned} \sqrt{n} (\tilde{\beta}_n(B) - \beta) &\xrightarrow{d} N(0, (B Q)^{-1} B \Omega B' (Q' B')^{-1}) \quad * (ABC)^{-1} = C^{-1} B^{-1} A^{-1} \\ &= N(0, (Q' B' (B \Omega B')^{-1} B Q)^{-1}) \end{aligned}$$

Recall that the asymptotic variance of the efficient GMM is $V^* = (Q' \Omega^{-1} Q)'$.

Then we want to show

$$V(B) - V^* \text{ p.s.d.} \iff V^{*-1} - V(B)^{-1} \text{ p.s.d.}$$

$$= Q' \Omega^{-1} Q - Q' B' (B \Omega B')^{-1} B Q$$

$$= \underbrace{Q' \Omega^{-1/2} \left(I_k - \underbrace{\Omega^{-1/2} B'}_H \underbrace{(B \Omega B')^{-1}}_{H'H} \underbrace{B \Omega^{1/2}}_{H'} \right) \Omega^{-1/2} Q}_{\text{projection matrix}}$$

Therefore, it is p.s.d.

Notice that we can find the B such that it attains the lower bound V^* .

$$Q' B' (B' R B')^{-1} B Q = Q' \Omega^{-1} Q$$

we want this to cancel each other
this should be this

⇒
 $B^* = Q' \Omega^{-1}$
 and plug it to confirm :

$$\cancel{Q' \Omega^{-1} Q} (\cancel{Q' \Omega^{-1} R \Omega^{-1} Q})^{-1} \cancel{Q' \Omega^{-1} Q} = V^* \text{ as derived.}$$

We can estimate it as

$$B_n^* = \frac{X' Z}{n} \hat{\Omega}_n^{-1} = \sum_{i=1}^n x_i \frac{z_i'}{n} \hat{\Omega}_n^{-1}$$

where $\hat{\Omega}_n = \frac{1}{n} \sum_{i=1}^n (y_i - x_i' \hat{\beta}_n)^2 z_i z_i'$
↳ some consistent estimator of β

Finally, we can write

$$\begin{aligned} \hat{\beta}_n (B_n^*) &= \left(B_n \sum_{i=1}^n z_i x_i' \right)^{-1} B_n \sum_{i=1}^n x_i y_i \\ &= \left(\sum_{i=1}^n x_i z_i' \hat{\Omega}_n^{-1} \sum_{i=1}^n z_i x_i' \right)^{-1} \sum_{i=1}^n x_i z_i' \hat{\Omega}_n^{-1} \sum_{i=1}^n z_i y_i = \hat{\beta}^{2\text{step}} \end{aligned}$$

Hypothesis Testing

We consider the test

$$H_0: \beta = \beta_0$$

$$H_1: \beta = \beta_0 + \delta, \quad \delta \in \mathbb{R}^k \text{ such that } \delta \neq 0.$$

The proposed statistic is

$$\begin{aligned} W_n &= n (\hat{\beta}_n - \beta_0)' \hat{V}_n^{-1}(A_n) (\hat{\beta}_n - \beta_0) \\ &= \sqrt{n} (\hat{\beta}_n - \beta_0)' \hat{V}_n^{-1}(A_n) \sqrt{n} (\hat{\beta}_n - \beta_0) \\ &= \left[\sqrt{n}^{-1/2} (A_n) \sqrt{n} (\hat{\beta}_n - \beta_0) \right]' \left[\sqrt{n}^{-1/2} (A_n) \sqrt{n} (\hat{\beta}_n - \beta_0) \right] \end{aligned}$$

Under H_0

$$W_n \xrightarrow{d} \chi^2_k \quad \text{and we reject if } W_n(\beta_0) > \chi^2_{1-\alpha, k}.$$

Under H_1

$$\hat{\beta}_n - \beta_0 = \delta + o_p(1), \quad \text{then}$$

$$\begin{aligned} \Pr_{H_1} (W_n > \chi^2_{1-\alpha, k}) &= \Pr_{H_1} \left(\frac{W_n}{n} > \frac{\chi^2_{1-\alpha, k}}{n} \right) \\ &= \Pr_{H_1} \left(\delta' (V(A))^{-1} \delta + o_p(1) > o_p(1) \right) \\ &= \Pr_{H_1} \left(\delta' (V(A))^{-1} \delta + o_p(1) > 0 \right) \end{aligned}$$

Notice that $\delta' (V(A))^{-1} \delta > 0$ because $V(A)$ is pos def and bounded.

Hence by taking limits

$$\lim_{n \rightarrow \infty} \Pr_{H_1} \left(\delta' (V(A))^{-1} \delta + o_p(1) > 0 \right) = 1.$$

In other words, in the limit the test always rejects if H_0 is not true.

Under local alternatives $\beta = \beta_0 + d/\sqrt{n}$ we have that

$$\sqrt{n}(\hat{\beta}_n - \beta_0) = d + (Q' A' A Q)^{-1} Q' A' A \frac{1}{\sqrt{n}} \sum_{i=1}^n z_i u_i + o_p(1)$$
$$\xrightarrow{d} N(d, V(A))$$

Therefore

$$W_n(\beta_0) \xrightarrow{d} Z_d' Z_d \sim \chi_k^2(d' V^{-1}(A) d)$$

The power function indexed by d is given by

$$\alpha \stackrel{\|d\| \rightarrow 0}{\leftarrow} \Pr(\chi_k^2(d' V(A) d) > \chi_{1-\alpha, k}^2) \stackrel{\|d\| \rightarrow \infty}{\rightarrow} 1$$

$\pi(d' V(A) d)$

Overid Test

Suppose we want to test $E \tilde{z}_i \tilde{u}_i = 0$.

We may use the criterion function

$$J_n = n \left(\frac{1}{n} \sum_{i=1}^n \tilde{z}_i \hat{u}_i \right)' \hat{\Omega}_n^{-1} \left(\frac{1}{n} \sum_{i=1}^n \tilde{z}_i \hat{u}_i \right)$$

⊕ we need \hat{u}_i
so we'll lose
degrees of
freedom

$$= \underbrace{\left(\Omega^{-1/2} \frac{1}{\sqrt{n}} \sum_{i=1}^n \tilde{z}_i \hat{u}_i \right)'}_{\hat{g}_n(\beta_n^*)} \hat{\Omega}_n^{-1} \Omega^{1/2} \left(\Omega^{-1/2} \frac{1}{\sqrt{n}} \sum_{i=1}^n \tilde{z}_i \hat{u}_i \right)$$

where

$$\begin{aligned} \hat{g}_n(\beta) &= \Omega^{-1/2} \frac{z'u}{\sqrt{n}} - \Omega^{-1/2} z'X \sqrt{n}(\beta_n^* - \beta) \\ &= \Omega^{-1/2} \frac{z'u}{\sqrt{n}} - \Omega^{-1/2} Q(Q'\Omega^{-1}Q)^{-1}Q'\Omega^{-1} \frac{z'u}{\sqrt{n}} + o_p(1), \\ &= \left\{ I_k - \Omega^{-1/2} Q(Q'\Omega^{-1}Q)^{-1}Q'\Omega^{-1} \right\} \Omega^{-1/2} \frac{z'u}{\sqrt{n}} + o_p(1). \\ &= \left\{ I_k - \underbrace{\Omega^{-1/2} Q}_{R} \underbrace{(Q'\Omega^{-1}Q)^{-1}}_{R'} \underbrace{Q'\Omega^{-1}}_{R} \right\} \Omega^{-1/2} \frac{z'u}{\sqrt{n}} + o_p(1). \\ &= \underbrace{\left\{ I_k - R(R'R)^{-1}R' \right\}}_{\substack{\text{idempotent and} \\ \text{symmetric}}} \Omega^{-1/2} \frac{z'u}{\sqrt{n}} + o_p(1). \end{aligned}$$

= projection

Write

$$\begin{aligned} \bar{J}_n &= Z' [I_L - R(R'R)^{-1}R'] [I_{L+op(1)}] [I_L - R(R'R)^{-1}R'] Z \\ &= Z' [I_L - R(R'R)^{-1}R'] Z + op(1) \\ &= Z' (I_L - R(R'R)^{-1}R') Z + op(1) \\ &= Z' (H \Delta H') Z + op(1) \\ &= \tilde{Z}' \Delta \tilde{Z} + op(1) \end{aligned}$$

Δ is orthogonal, i.e. pure rotation (normality is preserved)

Δ diagonal of 0 and 1s. $\text{Tr}(I_L - R(R'R)^{-1}R') = L - K$

$$= \sum_{j=1}^{L-K} \tilde{z}_j^2 + op(1)$$
$$\xrightarrow{d} \chi^2_{L-K}.$$

Under local alternatives we get normals around $\Omega^{-1/2} \delta$.
Hence the non-centrality parameter is

$$\begin{aligned} &\delta' \Omega^{-1/2} (I_L - R(R'R)^{-1}R') \Omega^{-1/2} \delta \\ &= \delta' (\Omega^{-1} - \Omega^{-1} \Omega^{-1/2} (\Omega^{-1/2} \Omega^{-1} \Omega^{-1/2})^{-1} \Omega^{-1/2} \Omega^{-1}) \delta \end{aligned}$$

if $\delta = \Omega^{-1/2} \pi$ then the non-centrality parameter collapses to zero (trivial power),