

Homework 8: Suggested Solutions

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11.14

JHU ...Take the model

$$\begin{aligned}
y_i &= \pi_i' \beta + e_i \\
\pi_i &= \mathbb{E}[\mathbf{x}_i | \mathbf{z}_i] = \Gamma' \mathbf{z}_i \\
\mathbb{E}[e_i | \mathbf{z}_i] &= 0
\end{aligned}$$

where y_i is scalar, \mathbf{x}_i is a k vector and \mathbf{z}_i is an ℓ vector. β and π_i are $k \times 1$ and Γ is $\ell \times k$. The sample is $(y_i, \mathbf{x}_i, \mathbf{z}_i : i = 1, \dots, n)$ with π_i unobserved.

Consider the estimator $\hat{\beta}$ for β by OLS of y_i on $\hat{\pi}_i = \hat{\Gamma}' \mathbf{z}_i$, where $\hat{\Gamma}$ is the OLS coefficient from the multivariate regression of \mathbf{x}_i on \mathbf{z}_i .

- Show that $\hat{\beta}$ is consistent for β .
- Find the asymptotic distribution $\sqrt{n}(\hat{\beta} - \beta)$ as $n \rightarrow \infty$, assuming that $\beta = \mathbf{0}$.
- Why is the assumption $\beta = \mathbf{0}$ an important simplifying condition in part (b)?
- Using the result in (c), construct an appropriate asymptotic test for the hypothesis $H_0: \beta = \mathbf{0}$.

Proof. (a)

Since $\mathbf{y} = \mathbf{Z}\Gamma\beta + \mathbf{e}$,

$$\begin{aligned}
\hat{\beta} &= (\hat{\Gamma}' \mathbf{Z}' \mathbf{Z} \hat{\Gamma})^{-1} \hat{\Gamma}' \mathbf{Z}' \mathbf{y} \\
&= (\hat{\Gamma}' \mathbf{Z}' \mathbf{Z} \hat{\Gamma})^{-1} \hat{\Gamma}' \mathbf{Z}' \mathbf{Z} \Gamma \beta + (\hat{\Gamma}' \mathbf{Z}' \mathbf{Z} \hat{\Gamma})^{-1} \hat{\Gamma}' \mathbf{Z}' \mathbf{e} \\
&= \left(\hat{\Gamma}' \left(\frac{1}{n} \mathbf{Z}' \mathbf{Z} \right) \hat{\Gamma} \right)^{-1} \hat{\Gamma}' \left(\frac{1}{n} \mathbf{Z}' \mathbf{Z} \right) \Gamma \beta + \left(\hat{\Gamma}' \left(\frac{1}{n} \mathbf{Z}' \mathbf{Z} \right) \hat{\Gamma} \right)^{-1} \hat{\Gamma}' \cdot \frac{1}{n} \mathbf{Z}' \mathbf{e}.
\end{aligned}$$

By $\mathbb{E}[\mathbf{x}_i | \mathbf{z}_i] = \Gamma' \mathbf{z}_i$, $\hat{\Gamma} \xrightarrow{P} \Gamma$. Also, $\frac{1}{n} \mathbf{Z}' \mathbf{Z} \xrightarrow{P} \mathbf{Q}_{zz} \equiv \mathbb{E}[\mathbf{z}_i \mathbf{z}_i']$ and $\frac{1}{n} \mathbf{Z}' \mathbf{e} \xrightarrow{P} \mathbf{0}$ by $\mathbb{E}[e_i | \mathbf{z}_i] = 0$.

Therefore, we find

$$\hat{\beta} \xrightarrow{P} (\Gamma' \mathbf{Q}_{zz} \Gamma)^{-1} (\Gamma' \mathbf{Q}_{zz} \Gamma) \beta \equiv \beta$$

(b)

Following (a), we have

$$(1) \quad \sqrt{n}(\hat{\beta}^* - \beta) = \left(\hat{\Gamma}^* \left(\frac{1}{n} \mathbf{Z}^* \mathbf{Z}^* \right)^* \hat{\Gamma}^* \right)^* \hat{\Gamma}^* \left(\frac{1}{n} \mathbf{Z}^* \mathbf{Z}^* \right) \Gamma \beta^* - \beta$$

$$(2) \quad + \left(\hat{\Gamma}^* \left(\frac{1}{n} \mathbf{Z}^* \mathbf{Z}^* \right)^* \right)^* \hat{\Gamma}^* \left(\frac{1}{\sqrt{n}} \mathbf{Z}^* \mathbf{e} \right)$$

Under the assumption $\beta = \mathbf{0}$, it reduces to

$$\sqrt{n}(\hat{\beta}^* - \beta) = \left(\hat{\Gamma}^* \left(\frac{1}{n} \mathbf{Z}^* \mathbf{Z}^* \right)^* \right)^* \hat{\Gamma}^* \left(\frac{1}{\sqrt{n}} \mathbf{Z}^* \mathbf{e} \right)$$

By the CLT, $\frac{1}{\sqrt{n}} \mathbf{Z}^* \mathbf{e} \xrightarrow{d} \mathcal{N}(\mathbf{0}, \Sigma)$, where $\Sigma = \mathbb{E}[\mathbf{z}_i \mathbf{z}_i' e_i^2]$. Hence,

$$\sqrt{n}(\hat{\beta}^* - \beta) \xrightarrow{d} \mathcal{N}\left(\mathbf{0}, \left(\Gamma^* \mathbf{Q}_{zz} \Gamma \right)^* \left(\Gamma^* \Omega \Gamma \right) \left(\Gamma^* \mathbf{Q}_{zz} \Gamma \right)^* \right)$$

(c)

Without $\beta = \mathbf{0}$, the term Eq. (1) is $O_p(1)$ and therefore contaminates the asymptotic distribution in (b)

(d)

Under $\beta = \mathbf{0}$, apply the result in (b), a Wald statistic can be used

$$\mathcal{W}_n = n \hat{\beta}^* \hat{\mathbf{V}}_{\beta}^* \hat{\beta}^*$$

where $\hat{\mathbf{V}}_{\beta} = \left(\hat{\Gamma}^* \hat{\mathbf{Q}}_{zz} \hat{\Gamma}^* \right)^* \left(\hat{\Gamma}^* \hat{\Omega} \hat{\Gamma}^* \right) \left(\hat{\Gamma}^* \hat{\mathbf{Q}}_{zz} \hat{\Gamma}^* \right)^*$. □

12.2

In the linear model

$$y_i = \mathbf{x}_i' \beta + e_i$$

$$\mathbb{E}[e_i | \mathbf{x}_i] = 0$$

suppose $\sigma_i^2 = \mathbb{E}[e_i^2 | \mathbf{x}_i]$ is known. Show that the GLS estimator of β can be written as an IV estimator using some instrument \mathbf{z}_i . (Find an expression for \mathbf{z}_i .)

Proof. The GLS estimator can be written as :

$$\begin{aligned}\hat{\beta}_{GLS} &= (\mathbf{X}' \mathbf{\Omega}^{-1} \mathbf{X})^{-1} (\mathbf{X}' \mathbf{\Omega}^{-1} \mathbf{y}) \\ &= (\mathbf{W}' \mathbf{X})^{-1} \mathbf{W}' \mathbf{y}\end{aligned}$$

where $\mathbf{\Omega} = \text{diag}(\sigma_1^2, \sigma_2^2, \dots, \sigma_n^2)$ and $\mathbf{W} = \mathbf{\Omega}^{-1} \mathbf{X}$. □

12.3

Take the linear model

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{e}$$

Let the OLS estimator for $\boldsymbol{\beta}$ be $\hat{\boldsymbol{\beta}}$ and the OLS residual be $\hat{\mathbf{e}} = \mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}}$.

Let the IV estimator for $\boldsymbol{\beta}$ using some instrument \mathbf{Z} be $\tilde{\boldsymbol{\beta}}$ and the IV residual be $\tilde{\mathbf{e}} = \mathbf{y} - \mathbf{X}\tilde{\boldsymbol{\beta}}$. If \mathbf{X} is indeed endogenous, will IV “fit” better than OLS, in the sense that $\tilde{\mathbf{e}}' \tilde{\mathbf{e}} < \hat{\mathbf{e}}' \hat{\mathbf{e}}$?

Proof. Define $\tilde{\mathbf{M}} = \mathbf{I} - \mathbf{X}(\mathbf{X}' \mathbf{P}_Z \mathbf{X})^{-1} \mathbf{X}' \mathbf{P}_Z$, where $\mathbf{P}_Z = \mathbf{Z}(\mathbf{Z}' \mathbf{Z})^{-1} \mathbf{Z}'$.

Then $\tilde{\mathbf{e}} = \tilde{\mathbf{M}}\mathbf{e}$, and the proof is similar to that in Problem 6 of Midterm. Observing $\tilde{\mathbf{M}}' \tilde{\mathbf{M}} = \tilde{\mathbf{M}}$, it is easy to show

$$\tilde{\mathbf{e}}' \tilde{\mathbf{e}} \geq \hat{\mathbf{e}}' \hat{\mathbf{e}}$$

□