

Solution to Chapter 3 Analytical Exercises

1. If \mathbf{A} is symmetric and idempotent, then $\mathbf{A}' = \mathbf{A}$ and $\mathbf{A}\mathbf{A} = \mathbf{A}$. So $\mathbf{x}'\mathbf{A}\mathbf{x} = \mathbf{x}'\mathbf{A}\mathbf{A}\mathbf{x} = \mathbf{x}'\mathbf{A}'\mathbf{A}\mathbf{x} = \mathbf{z}'\mathbf{z} \geq 0$ where $\mathbf{z} \equiv \mathbf{A}\mathbf{x}$.
2. (a) By assumption, $\{x_i, \varepsilon_i\}$ is jointly stationary and ergodic, so by ergodic theorem the first term of (*) converges almost surely to $E(x_i^2 \varepsilon_i^2)$ which exists and is finite by Assumption 3.5.
- (b) $z_i x_i^2 \varepsilon_i$ is the product of $x_i \varepsilon_i$ and $x_i z_i$. By using the Cauchy-Schwartz inequality, we obtain

$$E(|x_i \varepsilon_i \cdot x_i z_i|) \leq \sqrt{E(x_i^2 \varepsilon_i^2) E(x_i^2 z_i^2)}.$$

$E(x_i^2 \varepsilon_i^2)$ exists and is finite by Assumption 3.5 and $E(x_i^2 z_i^2)$ exists and is finite by Assumption 3.6. Therefore, $E(|x_i z_i \cdot x_i \varepsilon_i|)$ is finite. Hence, $E(x_i z_i \cdot x_i \varepsilon_i)$ exists and is finite.

- (c) By ergodic stationarity the sample average of $z_i x_i^2 \varepsilon_i$ converges in probability to some finite number. Because $\hat{\delta}$ is consistent for δ by Proposition 3.1, $\hat{\delta} - \delta$ converges to 0 in probability. Therefore, the second term of (*) converges to zero in probability.
- (d) By ergodic stationarity and Assumption 3.6 the sample average of $z_i^2 x_i^2$ converges in probability to some finite number. As mentioned in (c) $\hat{\delta} - \delta$ converges to 0 in probability. Therefore, the last term of (*) vanishes.

3. (a)

$$\begin{aligned} \mathbf{Q} &\equiv \Sigma'_{xz} \mathbf{S}^{-1} \Sigma_{xz} - \Sigma'_{xz} \mathbf{W} \Sigma_{xz} (\Sigma'_{xz} \mathbf{W} \mathbf{S} \mathbf{W} \Sigma_{xz})^{-1} \Sigma'_{xz} \mathbf{W} \Sigma_{xz} \\ &= \Sigma'_{xz} \mathbf{C}' \mathbf{C} \Sigma_{xz} - \Sigma'_{xz} \mathbf{W} \Sigma_{xz} (\Sigma'_{xz} \mathbf{W} \mathbf{C}^{-1} \mathbf{C}'^{-1} \mathbf{W} \Sigma_{xz})^{-1} \Sigma'_{xz} \mathbf{W} \Sigma_{xz} \\ &= \mathbf{H}' \mathbf{H} - \Sigma'_{xz} \mathbf{W} \Sigma_{xz} (\mathbf{G}' \mathbf{G})^{-1} \Sigma'_{xz} \mathbf{W} \Sigma_{xz} \\ &= \mathbf{H}' \mathbf{H} - \mathbf{H}' \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' \mathbf{H} \\ &= \mathbf{H}' [\mathbf{I}_K - \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}'] \mathbf{H} \\ &= \mathbf{H}' \mathbf{M}_G \mathbf{H}. \end{aligned}$$

- (b) First, we show that \mathbf{M}_G is symmetric and idempotent.

$$\begin{aligned} \mathbf{M}_G' &= \mathbf{I}_K - \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' \\ &= \mathbf{I}_K - \mathbf{G} ((\mathbf{G}' \mathbf{G})^{-1})' \mathbf{G}' \\ &= \mathbf{I}_K - \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' \\ &= \mathbf{M}_G. \end{aligned}$$

$$\begin{aligned} \mathbf{M}_G \mathbf{M}_G &= \mathbf{I}_K \mathbf{I}_K - \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' \mathbf{I}_K - \mathbf{I}_K \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' + \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' \\ &= \mathbf{I}_K - \mathbf{G} (\mathbf{G}' \mathbf{G})^{-1} \mathbf{G}' \\ &= \mathbf{M}_G. \end{aligned}$$

Thus, \mathbf{M}_G is symmetric and idempotent. For any L -dimensional vector \mathbf{x} ,

$$\begin{aligned} \mathbf{x}' \mathbf{Q} \mathbf{x} &= \mathbf{x}' \mathbf{H}' \mathbf{M}_G \mathbf{H} \mathbf{x} \\ &= \mathbf{z}' \mathbf{M}_G \mathbf{z} \quad (\text{where } \mathbf{z} \equiv \mathbf{H} \mathbf{x}) \\ &\geq 0 \quad (\text{since } \mathbf{M}_G \text{ is positive semidefinite}). \end{aligned}$$

Therefore, \mathbf{Q} is positive semidefinite.

4. (the answer on p. 254 of the book simplified) If \mathbf{W} is as defined in the hint, then

$$\mathbf{WSW} = \mathbf{W} \quad \text{and} \quad \boldsymbol{\Sigma}'_{xz} \mathbf{W} \boldsymbol{\Sigma}_{xz} = \boldsymbol{\Sigma}_{zz} \mathbf{A}^{-1} \boldsymbol{\Sigma}_{zz}.$$

So (3.5.1) reduces to the asymptotic variance of the OLS estimator. By (3.5.11), it is no smaller than $(\boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz})^{-1}$, which is the asymptotic variance of the efficient GMM estimator.

5. (a) From the expression for $\widehat{\boldsymbol{\delta}}(\widehat{\mathbf{S}}^{-1})$ (given in (3.5.12)) and the expression for $\mathbf{g}_n(\widetilde{\boldsymbol{\delta}})$ (given in (3.4.2)), it is easy to show that $\mathbf{g}_n(\widehat{\boldsymbol{\delta}}(\widehat{\mathbf{S}}^{-1})) = \widehat{\mathbf{B}}\mathbf{s}_{xy}$. But $\widehat{\mathbf{B}}\mathbf{s}_{xy} = \widehat{\mathbf{B}}\bar{\mathbf{g}}$ because

$$\begin{aligned} \widehat{\mathbf{B}}\mathbf{s}_{xy} &= (\mathbf{I}_K - \mathbf{S}_{xz}(\mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1} \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1}) \mathbf{s}_{xy} \\ &= (\mathbf{I}_K - \mathbf{S}_{xz}(\mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1} \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1}) (\mathbf{S}_{xz} \boldsymbol{\delta} + \bar{\mathbf{g}}) \quad (\text{since } y_i = \mathbf{z}'_i \boldsymbol{\delta} + \varepsilon_i) \\ &= (\mathbf{S}_{xz} - \mathbf{S}_{xz}(\mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1} \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1} \mathbf{S}_{xz}) \boldsymbol{\delta} + (\mathbf{I}_K - \mathbf{S}_{xz}(\mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1} \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1}) \bar{\mathbf{g}} \\ &= (\mathbf{S}_{xz} - \mathbf{S}_{xz}) \boldsymbol{\delta} + \widehat{\mathbf{B}}\bar{\mathbf{g}} \\ &= \widehat{\mathbf{B}}\bar{\mathbf{g}}. \end{aligned}$$

(b) Since $\widehat{\mathbf{S}}^{-1} = \mathbf{C}'\mathbf{C}$, we obtain $\widehat{\mathbf{B}}'\widehat{\mathbf{S}}^{-1}\widehat{\mathbf{B}} = \widehat{\mathbf{B}}'\mathbf{C}'\mathbf{C}\widehat{\mathbf{B}} = (\mathbf{C}\widehat{\mathbf{B}})'(\mathbf{C}\widehat{\mathbf{B}})$. But

$$\begin{aligned} \mathbf{C}\widehat{\mathbf{B}} &= \mathbf{C}(\mathbf{I}_K - \mathbf{S}_{xz}(\mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1} \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{S}}^{-1}) \\ &= \mathbf{C} - \mathbf{C}\mathbf{S}_{xz}(\mathbf{S}'_{xz} \mathbf{C}'\mathbf{C}\mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \mathbf{C}'\mathbf{C} \\ &= \mathbf{C} - \mathbf{A}(\mathbf{A}'\mathbf{A})^{-1} \mathbf{A}'\mathbf{C} \quad (\text{where } \mathbf{A} \equiv \mathbf{C}\mathbf{S}_{xz}) \\ &= [\mathbf{I}_K - \mathbf{A}(\mathbf{A}'\mathbf{A})^{-1} \mathbf{A}']\mathbf{C} \\ &\equiv \mathbf{M}\mathbf{C}. \end{aligned}$$

So $\widehat{\mathbf{B}}'\widehat{\mathbf{S}}^{-1}\widehat{\mathbf{B}} = (\mathbf{M}\mathbf{C})'(\mathbf{M}\mathbf{C}) = \mathbf{C}'\mathbf{M}'\mathbf{M}\mathbf{C}$. It should be routine to show that \mathbf{M} is symmetric and idempotent. Thus $\widehat{\mathbf{B}}'\widehat{\mathbf{S}}^{-1}\widehat{\mathbf{B}} = \mathbf{C}'\mathbf{M}\mathbf{C}$.

The rank of \mathbf{M} equals its trace, which is

$$\begin{aligned} \text{trace}(\mathbf{M}) &= \text{trace}(\mathbf{I}_K - \mathbf{A}(\mathbf{A}'\mathbf{A})^{-1} \mathbf{A}') \\ &= \text{trace}(\mathbf{I}_K) - \text{trace}(\mathbf{A}(\mathbf{A}'\mathbf{A})^{-1} \mathbf{A}') \\ &= \text{trace}(\mathbf{I}_K) - \text{trace}(\mathbf{A}'\mathbf{A}(\mathbf{A}'\mathbf{A})^{-1}) \\ &= K - \text{trace}(\mathbf{I}_L) \\ &= K - L. \end{aligned}$$

(c) As defined in (b), $\mathbf{C}'\mathbf{C} = \widehat{\mathbf{S}}^{-1}$. Let \mathbf{D} be such that $\mathbf{D}'\mathbf{D} = \mathbf{S}^{-1}$. The choice of \mathbf{C} and \mathbf{D} is not unique, but it would be possible to choose \mathbf{C} so that $\text{plim } \mathbf{C} = \mathbf{D}$. Now,

$$\mathbf{v} \equiv \sqrt{n}(\mathbf{C}\bar{\mathbf{g}}) = \mathbf{C}(\sqrt{n}\bar{\mathbf{g}}).$$

By using the Ergodic Stationary Martingale Differences CLT, we obtain $\sqrt{n}\bar{\mathbf{g}} \rightarrow_d N(\mathbf{0}, \mathbf{S})$. So

$$\mathbf{v} = \mathbf{C}(\sqrt{n}\bar{\mathbf{g}}) \rightarrow_d N(\mathbf{0}, \text{Avar}(\mathbf{v}))$$

where

$$\begin{aligned} \text{Avar}(\mathbf{v}) &= \mathbf{D}\mathbf{S}\mathbf{D}' \\ &= \mathbf{D}(\mathbf{D}'\mathbf{D})^{-1}\mathbf{D}' \\ &= \mathbf{D}\mathbf{D}^{-1}\mathbf{D}^{-1'}\mathbf{D}' \\ &= \mathbf{I}_K. \end{aligned}$$

(d)

$$\begin{aligned}
J(\widehat{\boldsymbol{\delta}}(\widehat{\mathbf{S}}^{-1}), \widehat{\mathbf{S}}^{-1}) &= n \cdot \mathbf{g}_n(\widehat{\boldsymbol{\delta}}(\widehat{\mathbf{S}}^{-1})) \widehat{\mathbf{S}}^{-1} \mathbf{g}_n(\widehat{\boldsymbol{\delta}}(\widehat{\mathbf{S}}^{-1})) \\
&= n \cdot (\widehat{\mathbf{B}}\bar{\mathbf{g}})' \widehat{\mathbf{S}}^{-1} (\widehat{\mathbf{B}}\bar{\mathbf{g}}) \quad (\text{by (a)}) \\
&= n \cdot \bar{\mathbf{g}}' \widehat{\mathbf{B}}' \widehat{\mathbf{S}}^{-1} \widehat{\mathbf{B}}\bar{\mathbf{g}} \\
&= n \cdot \bar{\mathbf{g}}' \mathbf{C}' \mathbf{M} \mathbf{C} \bar{\mathbf{g}} \quad (\text{by (b)}) \\
&= \mathbf{v}' \mathbf{M} \mathbf{v} \quad (\text{since } \mathbf{v} \equiv \sqrt{n} \mathbf{C} \bar{\mathbf{g}}).
\end{aligned}$$

Since $\mathbf{v} \rightarrow_d N(\mathbf{0}, \mathbf{I}_K)$ and \mathbf{M} is idempotent, $\mathbf{v}' \mathbf{M} \mathbf{v}$ is asymptotically chi-squared with degrees of freedom equaling the rank of $\mathbf{M} = K - L$.

6. From Exercise 5, $J = n \cdot \bar{\mathbf{g}}' \widehat{\mathbf{B}}' \widehat{\mathbf{S}}^{-1} \widehat{\mathbf{B}}\bar{\mathbf{g}}$. Also from Exercise 5, $\widehat{\mathbf{B}}\bar{\mathbf{g}} = \widehat{\mathbf{B}}\mathbf{s}_{xy}$.

7. For the most parts, the hints are nearly the answer. Here, we provide answers to (d), (f), (g), (i), and (j).

(d) As shown in (c), $J_1 = \mathbf{v}'_1 \mathbf{M}_1 \mathbf{v}_1$. It suffices to prove that $\mathbf{v}_1 = \mathbf{C}_1 \mathbf{F}' \mathbf{C}^{-1} \mathbf{v}$.

$$\begin{aligned}
\mathbf{v}_1 &\equiv \sqrt{n} \mathbf{C}_1 \bar{\mathbf{g}}_1 \\
&= \sqrt{n} \mathbf{C}_1 \mathbf{F}' \bar{\mathbf{g}} \\
&= \sqrt{n} \mathbf{C}_1 \mathbf{F}' \mathbf{C}^{-1} \mathbf{C} \bar{\mathbf{g}} \\
&= \mathbf{C}_1 \mathbf{F}' \mathbf{C}^{-1} \sqrt{n} \mathbf{C} \bar{\mathbf{g}} \\
&= \mathbf{C}_1 \mathbf{F}' \mathbf{C}^{-1} \mathbf{v} \quad (\text{since } \mathbf{v} \equiv \sqrt{n} \mathbf{C} \bar{\mathbf{g}}).
\end{aligned}$$

(f) Use the hint to show that $\mathbf{A}' \mathbf{D} = \mathbf{0}$ if $\mathbf{A}'_1 \mathbf{M}_1 = \mathbf{0}$. It should be easy to show that $\mathbf{A}'_1 \mathbf{M}_1 = \mathbf{0}$ from the definition of \mathbf{M}_1 .

(g) By the definition of \mathbf{M} in Exercise 5, $\mathbf{M} \mathbf{D} = \mathbf{D} - \mathbf{A}(\mathbf{A}' \mathbf{A})^{-1} \mathbf{A}' \mathbf{D}$. So $\mathbf{M} \mathbf{D} = \mathbf{D}$ since $\mathbf{A}' \mathbf{D} = \mathbf{0}$ as shown in the previous part. Since both \mathbf{M} and \mathbf{D} are symmetric, $\mathbf{D} \mathbf{M} = \mathbf{D}' \mathbf{M}' = (\mathbf{M} \mathbf{D})' = \mathbf{D}' = \mathbf{D}$. As shown in part (e), \mathbf{D} is idempotent. Also, \mathbf{M} is idempotent as shown in Exercise 5. So $(\mathbf{M} - \mathbf{D})^2 = \mathbf{M}^2 - \mathbf{D} \mathbf{M} - \mathbf{M} \mathbf{D} + \mathbf{D}^2 = \mathbf{M} - \mathbf{D}$. As shown in Exercise 5, the trace of \mathbf{M} is $K - L$. As shown in (e), the trace of \mathbf{D} is $K_1 - L$. So the trace of $\mathbf{M} - \mathbf{D}$ is $K - K_1$. The rank of a symmetric and idempotent matrix is its trace.

(i) It has been shown in Exercise 6 that $\bar{\mathbf{g}}' \mathbf{C}' \mathbf{M} \mathbf{C} \bar{\mathbf{g}} = \mathbf{s}'_{xy} \mathbf{C}' \mathbf{M} \mathbf{C} \mathbf{s}_{xy}$ since $\mathbf{C}' \mathbf{M} \mathbf{C} = \widehat{\mathbf{B}}' \widehat{\mathbf{S}}^{-1} \widehat{\mathbf{B}}$. Here, we show that $\bar{\mathbf{g}}' \mathbf{C}' \mathbf{D} \mathbf{C} \bar{\mathbf{g}} = \mathbf{s}'_{xy} \mathbf{C}' \mathbf{D} \mathbf{C} \mathbf{s}_{xy}$.

$$\begin{aligned}
\bar{\mathbf{g}}' \mathbf{C}' \mathbf{D} \mathbf{C} \bar{\mathbf{g}} &= \bar{\mathbf{g}}' \mathbf{F} \mathbf{C}'_1 \mathbf{M}_1 \mathbf{C}_1 \mathbf{F}' \bar{\mathbf{g}} \quad (\mathbf{C}' \mathbf{D} \mathbf{C} = \mathbf{F} \mathbf{C}'_1 \mathbf{M}_1 \mathbf{C}_1 \mathbf{F}' \text{ by the definition of } \mathbf{D} \text{ in (d)}) \\
&= \bar{\mathbf{g}}' \mathbf{F} \widehat{\mathbf{B}}'_1 (\widehat{\mathbf{S}}_{11})^{-1} \widehat{\mathbf{B}}_1 \mathbf{F}' \bar{\mathbf{g}} \quad (\text{since } \mathbf{C}'_1 \mathbf{M}_1 \mathbf{C}_1 = \widehat{\mathbf{B}}'_1 (\widehat{\mathbf{S}}_{11})^{-1} \widehat{\mathbf{B}}_1 \text{ from (a)}) \\
&= \bar{\mathbf{g}}'_1 \widehat{\mathbf{B}}'_1 (\widehat{\mathbf{S}}_{11})^{-1} \widehat{\mathbf{B}}_1 \bar{\mathbf{g}}_1 \quad (\text{since } \bar{\mathbf{g}}_1 = \mathbf{F}' \bar{\mathbf{g}}).
\end{aligned}$$

From the definition of $\widehat{\mathbf{B}}_1$ and the fact that $\mathbf{s}_{x_1 y} = \mathbf{S}_{x_1 z} \boldsymbol{\delta} + \bar{\mathbf{g}}_1$, it follows that $\widehat{\mathbf{B}}_1 \bar{\mathbf{g}}_1 = \widehat{\mathbf{B}}_1 \mathbf{s}_{x_1 y}$. So

$$\begin{aligned}
\bar{\mathbf{g}}'_1 \widehat{\mathbf{B}}'_1 (\widehat{\mathbf{S}}_{11})^{-1} \widehat{\mathbf{B}}_1 \bar{\mathbf{g}}_1 &= \mathbf{s}'_{x_1 y} \widehat{\mathbf{B}}'_1 (\widehat{\mathbf{S}}_{11})^{-1} \widehat{\mathbf{B}}_1 \mathbf{s}_{x_1 y} \\
&= \mathbf{s}'_{xy} \mathbf{F} \widehat{\mathbf{B}}'_1 (\widehat{\mathbf{S}}_{11})^{-1} \widehat{\mathbf{B}}_1 \mathbf{F}' \mathbf{s}_{xy} \quad (\text{since } \mathbf{s}_{x_1 y} = \mathbf{F}' \mathbf{s}_{xy}) \\
&= \mathbf{s}'_{xy} \mathbf{F} \mathbf{C}'_1 \mathbf{M}_1 \mathbf{C}_1 \mathbf{F}' \mathbf{s}_{xy} \quad (\text{since } \widehat{\mathbf{B}}'_1 (\widehat{\mathbf{S}}_{11})^{-1} \widehat{\mathbf{B}}_1 = \mathbf{C}'_1 \mathbf{M}_1 \mathbf{C}_1 \text{ from (a)}) \\
&= \mathbf{s}'_{xy} \mathbf{C}' \mathbf{D} \mathbf{C} \mathbf{s}_{xy}.
\end{aligned}$$

(j) $\mathbf{M} - \mathbf{D}$ is positive semi-definite because it is symmetric and idempotent.

8. (a) Solve the first-order conditions in the hint for $\bar{\delta}$ to obtain

$$\bar{\delta} = \widehat{\delta}(\widehat{\mathbf{W}}) - \frac{1}{2n} (\mathbf{S}'_{xz} \widehat{\mathbf{W}} \mathbf{S}_{xz})^{-1} \mathbf{R}' \boldsymbol{\lambda}.$$

Substitute this into the constraint $\mathbf{R}\bar{\delta} = \mathbf{r}$ to obtain the expression for $\boldsymbol{\lambda}$ in the question. Then substitute this expression for $\boldsymbol{\lambda}$ into the above equation to obtain the expression for $\bar{\delta}$ in the question.

(b) The hint is almost the answer.

(c) What needs to be shown is that $n \cdot (\widehat{\delta}(\widehat{\mathbf{W}}) - \bar{\delta})' (\mathbf{S}'_{xz} \widehat{\mathbf{W}} \mathbf{S}_{xz}) (\widehat{\delta}(\widehat{\mathbf{W}}) - \bar{\delta})$ equals the Wald statistic. But this is immediate from substitution of the expression for $\bar{\delta}$ in (a).

9. (a) By applying (3.4.11), we obtain

$$\begin{bmatrix} \sqrt{n}(\widehat{\delta}_1 - \boldsymbol{\delta}) \\ \sqrt{n}(\widehat{\delta}_2 - \boldsymbol{\delta}) \end{bmatrix} = \begin{bmatrix} (\mathbf{S}'_{xz} \widehat{\mathbf{W}}_1 \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{W}}_1 \\ (\mathbf{S}'_{xz} \widehat{\mathbf{W}}_2 \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{W}}_2 \end{bmatrix} \sqrt{n} \bar{\mathbf{g}}.$$

By using Billingsley CLT, we have

$$\sqrt{n} \bar{\mathbf{g}} \xrightarrow{d} N(\mathbf{0}, \mathbf{S}).$$

Also, we have

$$\begin{bmatrix} (\mathbf{S}'_{xz} \widehat{\mathbf{W}}_1 \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{W}}_1 \\ (\mathbf{S}'_{xz} \widehat{\mathbf{W}}_2 \mathbf{S}_{xz})^{-1} \mathbf{S}'_{xz} \widehat{\mathbf{W}}_2 \end{bmatrix} \xrightarrow{p} \begin{bmatrix} \mathbf{Q}_1^{-1} \boldsymbol{\Sigma}'_{xz} \mathbf{W}_1 \\ \mathbf{Q}_2^{-1} \boldsymbol{\Sigma}'_{xz} \mathbf{W}_2 \end{bmatrix}.$$

Therefore, by Lemma 2.4(c),

$$\begin{aligned} \begin{bmatrix} \sqrt{n}(\widehat{\delta}_1 - \boldsymbol{\delta}) \\ \sqrt{n}(\widehat{\delta}_2 - \boldsymbol{\delta}) \end{bmatrix} &\xrightarrow{d} N\left(\mathbf{0}, \begin{bmatrix} \mathbf{Q}_1^{-1} \boldsymbol{\Sigma}'_{xz} \mathbf{W}_1 \\ \mathbf{Q}_2^{-1} \boldsymbol{\Sigma}'_{xz} \mathbf{W}_2 \end{bmatrix} \mathbf{S} (\mathbf{W}_1 \boldsymbol{\Sigma}_{xz} \mathbf{Q}_1^{-1} \quad \mathbf{W}_2 \boldsymbol{\Sigma}_{xz} \mathbf{Q}_2^{-1})\right) \\ &= N\left(\mathbf{0}, \begin{bmatrix} \mathbf{A}_{11} & \mathbf{A}_{12} \\ \mathbf{A}_{21} & \mathbf{A}_{22} \end{bmatrix}\right). \end{aligned}$$

(b) $\sqrt{n}\mathbf{q}$ can be rewritten as

$$\sqrt{n}\mathbf{q} = \sqrt{n}(\widehat{\delta}_1 - \widehat{\delta}_2) = \sqrt{n}(\widehat{\delta}_1 - \boldsymbol{\delta}) - \sqrt{n}(\widehat{\delta}_2 - \boldsymbol{\delta}) = \begin{bmatrix} 1 & -1 \end{bmatrix} \begin{bmatrix} \sqrt{n}(\widehat{\delta}_1 - \boldsymbol{\delta}) \\ \sqrt{n}(\widehat{\delta}_2 - \boldsymbol{\delta}) \end{bmatrix}.$$

Therefore, we obtain

$$\sqrt{n}\mathbf{q} \xrightarrow{d} N(\mathbf{0}, \text{Avar}(\mathbf{q})).$$

where

$$\text{Avar}(\mathbf{q}) = \begin{bmatrix} 1 & -1 \end{bmatrix} \begin{bmatrix} \mathbf{A}_{11} & \mathbf{A}_{12} \\ \mathbf{A}_{21} & \mathbf{A}_{22} \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \mathbf{A}_{11} + \mathbf{A}_{22} - \mathbf{A}_{12} - \mathbf{A}_{21}.$$

(c) Since $\mathbf{W}_2 = \mathbf{S}^{-1}$, \mathbf{Q}_2 , \mathbf{A}_{12} , \mathbf{A}_{21} , and \mathbf{A}_{22} can be rewritten as follows:

$$\begin{aligned}
\mathbf{Q}_2 &= \boldsymbol{\Sigma}'_{xz} \mathbf{W}_2 \boldsymbol{\Sigma}_{xz} \\
&= \boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz}, \\
\mathbf{A}_{12} &= \mathbf{Q}_1^{-1} \boldsymbol{\Sigma}'_{xz} \mathbf{W}_1 \mathbf{S} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz} \mathbf{Q}_2^{-1} \\
&= \mathbf{Q}_1^{-1} (\boldsymbol{\Sigma}'_{xz} \mathbf{W}_1 \boldsymbol{\Sigma}_{xz}) \mathbf{Q}_2^{-1} \\
&= \mathbf{Q}_1^{-1} \mathbf{Q}_1 \mathbf{Q}_2^{-1} \\
&= \mathbf{Q}_2^{-1}, \\
\mathbf{A}_{21} &= \mathbf{Q}_2^{-1} \boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \mathbf{S} \mathbf{W}_1 \boldsymbol{\Sigma}_{xz} \mathbf{Q}_1^{-1} \\
&= \mathbf{Q}_2^{-1}, \\
\mathbf{A}_{22} &= (\boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz})^{-1} \boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \mathbf{S} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz} (\boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz})^{-1} \\
&= (\boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz})^{-1} \\
&= \mathbf{Q}_2^{-1}.
\end{aligned}$$

Substitution of these into the expression for $\text{Avar}(\mathbf{q})$ in (b), we obtain

$$\begin{aligned}
\text{Avar}(\mathbf{q}) &= \mathbf{A}_{11} - \mathbf{Q}_2^{-1} \\
&= \mathbf{A}_{11} - (\boldsymbol{\Sigma}'_{xz} \mathbf{S}^{-1} \boldsymbol{\Sigma}_{xz})^{-1} \\
&= \text{Avar}(\widehat{\boldsymbol{\delta}}(\widehat{\mathbf{W}}_1)) - \text{Avar}(\widehat{\boldsymbol{\delta}}(\widehat{\mathbf{S}}^{-1})).
\end{aligned}$$

10. (a)

$$\begin{aligned}
\sigma_{xz} \equiv \text{E}(x_i z_i) &= \text{E}(x_i(x_i \beta + v_i)) \\
&= \beta \text{E}(x_i^2) + \text{E}(x_i v_i) \\
&= \beta \sigma_x^2 \neq 0 \quad (\text{by assumptions (2), (3), and (4)}).
\end{aligned}$$

(b) From the definition of $\widehat{\boldsymbol{\delta}}$,

$$\widehat{\boldsymbol{\delta}} - \boldsymbol{\delta} = \left(\frac{1}{n} \sum_{i=1}^n x_i z_i \right)^{-1} \frac{1}{n} \sum_{i=1}^n x_i \varepsilon_i = s_{xz}^{-1} \frac{1}{n} \sum_{i=1}^n x_i \varepsilon_i.$$

We have $x_i z_i = x_i(x_i \beta + v_i) = x_i^2 \beta + x_i v_i$, which, being a function of $(x_i, \boldsymbol{\eta}_i)$, is ergodic stationary by assumption (1). So by the Ergodic theorem, $s_{xz} \rightarrow_{\text{p}} \sigma_{xz}$. Since $\sigma_{xz} \neq 0$ by (a), we have $s_{xz}^{-1} \rightarrow_{\text{p}} \sigma_{xz}^{-1}$. By assumption (2), $\text{E}(x_i \varepsilon_i) = 0$. So by assumption (1), we have $\frac{1}{n} \sum_{i=1}^n x_i \varepsilon_i \rightarrow_{\text{p}} 0$. Thus $\widehat{\boldsymbol{\delta}} - \boldsymbol{\delta} \rightarrow_{\text{p}} 0$.

(c)

$$\begin{aligned}
s_{xz} &\equiv \frac{1}{n} \sum_{i=1}^n x_i z_i \\
&= \frac{1}{n} \sum_{i=1}^n (x_i^2 \beta + x_i v_i) \\
&= \frac{1}{\sqrt{n}} \frac{1}{n} \sum_{i=1}^n x_i^2 + \frac{1}{n} \sum_{i=1}^n x_i v_i \quad (\text{since } \beta = \frac{1}{\sqrt{n}}) \\
&\rightarrow_{\text{p}} 0 \cdot \text{E}(x_i^2) + \text{E}(x_i v_i) \\
&= 0
\end{aligned}$$

(d)

$$\sqrt{n}s_{xz} = \frac{1}{n} \sum_{i=1}^n x_i^2 + \frac{1}{\sqrt{n}} \sum_{i=1}^n x_i v_i.$$

By assumption (1) and the Ergodic Theorem, the first term of RHS converges in probability to $E(x_i^2) = \sigma_x^2 > 0$. Assumption (2) and the Martingale Differences CLT imply that

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n x_i v_i \xrightarrow{d} a \sim N(0, s_{22}).$$

Therefore, by Lemma 2.4(a), we obtain

$$\sqrt{n}s_{xz} \xrightarrow{d} \sigma_x^2 + a.$$

(e) $\hat{\delta} - \delta$ can be rewritten as

$$\hat{\delta} - \delta = (\sqrt{n}s_{xz})^{-1} \sqrt{n}\bar{g}_1.$$

From assumption (2) and the Martingale Differences CLT, we obtain

$$\sqrt{n}\bar{g}_1 \xrightarrow{d} b \sim N(0, s_{11}).$$

where s_{11} is the (1, 1) element of \mathbf{S} . By using the result of (d) and Lemma 2.3(b),

$$\hat{\delta} - \delta \xrightarrow{d} (\sigma_x^2 + a)^{-1} b.$$

(a, b) are jointly normal because the joint distribution is the limiting distribution of

$$\sqrt{n}\bar{\mathbf{g}} = \begin{bmatrix} \sqrt{n}\bar{g}_1 \\ \sqrt{n}(\frac{1}{n} \sum_{i=1}^n x_i v_i) \end{bmatrix}.$$

(f) Because $\hat{\delta} - \delta$ converges in distribution to $(\sigma_x^2 + a)^{-1} b$ which is not zero, the answer is No.