

**TECHNICAL APPENDIX FOR ASYMPTOTIC PROPERTIES OF A ROBUST
VARIANCE MATRIX ESTIMATOR FOR PANEL DATA WHEN T IS LARGE
(TO BE OMITTED FROM PUBLICATION)**

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APPENDIX A. PRELIMINARIES

Throughout the appendix, let $\|A\| = [\text{trace}(A'A)]^{1/2}$ be the Euclidean norm of a matrix A , and let $\sum_i = \sum_{i=1}^n$, $\sum_t = \sum_{t=1}^T$, and $\sum_h = \sum_{h=1}^k$. Repeated use will be made of the following simple results, which are stated here for convenience.

Lemma A.1. *For matrices A and B , $\mathbb{E}\|A \otimes B\|^r \leq (\mathbb{E}\|A\|^{2r}\mathbb{E}\|B\|^{2r})^{1/2}$.*

Proof. $\mathbb{E}\|A \otimes B\|^r = \mathbb{E}\{[\text{trace}(AA' \otimes BB')]^{r/2}\} = \mathbb{E}\{[\text{trace}(AA')]^{r/2}[\text{trace}(BB')]^{r/2}\} = \mathbb{E}(\|A\|^r\|B\|^r) \leq (\mathbb{E}\|A\|^{2r}\mathbb{E}\|B\|^{2r})^{1/2}$ where the equalities follow from the definition of $\|A\|$ and properties of the Kronecker product and the inequality results from the Cauchy-Schwarz inequality. ■

Lemma A.2. *Suppose $\{Z_{i,T}\}$ are independent across i for all T with $\mathbb{E}[Z_{i,T}] = \mu_{i,T}$ and $\mathbb{E}|Z_{i,T}|^{1+\delta} < \Delta < \infty$ for some $\delta > 0$ and all i, T . Then $\frac{1}{n} \sum_{i=1}^n (Z_{i,T} - \mu_{i,T}) \xrightarrow{p} 0$ as $\{n, T\} \rightarrow \infty$ jointly.*

Proof. The proof follows from standard arguments, cf. Chung (2001) Chapter 5. Details are given in Appendix G. ■

Lemma A.3. *For $k \times 1$ vectors $Z_{i,T}$, suppose $\{Z_{i,T}\}$ are independent across i for all T with $\mathbb{E}[Z_{i,T}] = 0$, $\mathbb{E}[Z_{i,T}Z'_{i,T}] = \Omega_{i,T}$, and $\mathbb{E}\|Z_{i,T}\|^{2+\delta} < \Delta < \infty$ for some $\delta > 0$. Assume $\Omega = \lim_{n,T} \frac{1}{n} \sum_{i=1}^n \Omega_{i,T}$ is positive definite with minimum eigenvalue $\lambda_{\min} > 0$. Then $\frac{1}{\sqrt{n}} \sum_{i=1}^n Z_{i,T} \xrightarrow{d} N(0, W)$ as $\{n, T\} \rightarrow \infty$ jointly.*

Proof. The result follows from verifying the Lindeberg condition of Theorem 2 in Phillips and Moon (1999) using an argument similar to that used in the proof of Theorem 3 in Phillips and Moon (1999). Details are given in Appendix G. ■

The final lemma simply restates Doukhan (1994) Theorem 2 with a slight change of notation. Its proof may be found in Doukhan (1994) p. 25-30.

Lemma A.4. *Let $\{z_t\}$ be a strong mixing sequence with $\mathbb{E}[z_t] = 0$, $\mathbb{E}\|z_t\|^{\tau+\epsilon} < \Delta < \infty$, and mixing coefficient $\alpha(m)$ of size $\frac{(1-c)r}{r-c}$ where $c \in 2\mathbb{N}$, $c \geq \tau$, and $r > c$. Then there is a constant C depending only on τ and $\alpha(m)$ such that $\mathbb{E}|\sum_{t=1}^T y_t|^\tau \leq CD(\tau, \epsilon, T)$ with $D(\tau, \epsilon, T)$ defined in Doukhan (1994) and satisfying $D(\tau, \epsilon, T) = O(T)$ if $\tau \leq 2$ and $D(\tau, \epsilon, T) = O(T^{\tau/2})$ if $\tau > 2$.*

Finally note

$$\begin{aligned}\widehat{W} &= \frac{1}{nT} \sum_i x_i' \widehat{\epsilon}_i \widetilde{\epsilon}_i' x_i \\ &= \frac{1}{nT} \sum_i x_i' \epsilon_i \epsilon_i' x_i\end{aligned}\tag{O.1}$$

$$- \frac{1}{nT} \sum_i x_i' x_i (\widehat{\beta} - \beta) \epsilon_i' x_i\tag{O.2}$$

$$- \frac{1}{nT} \sum_i x_i' \epsilon_i (\widehat{\beta} - \beta)' x_i' x_i\tag{O.3}$$

$$+ \frac{1}{nT} \sum_i x_i' x_i (\widehat{\beta} - \beta) (\widehat{\beta} - \beta)' x_i' x_i,\tag{O.4}$$

and

$$\begin{aligned}\widehat{V} &= \frac{1}{nT} \sum_i [(\text{vec}(x_i' \widehat{\epsilon}_i \widetilde{\epsilon}_i' x_i - \widehat{W}))(\text{vec}(x_i' \widehat{\epsilon}_i \widetilde{\epsilon}_i' x_i - \widehat{W}))'] \\ &= \frac{1}{nT} \sum_i (\text{vec}(x_i' \widehat{\epsilon}_i \widetilde{\epsilon}_i' x_i))(\text{vec}(x_i' \widehat{\epsilon}_i \widetilde{\epsilon}_i' x_i))' - \frac{1}{nT} \sum_i (\text{vec}(\widehat{W})) \frac{1}{nT} \sum_i (\text{vec}(\widehat{W}))'\end{aligned}$$

where

$$\begin{aligned}\frac{1}{nT} \sum_i (\text{vec}(x_i' \widehat{\epsilon}_i \widetilde{\epsilon}_i' x_i))(\text{vec}(x_i' \widehat{\epsilon}_i \widetilde{\epsilon}_i' x_i))' \\ = \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i \epsilon_i' x_i))(\text{vec}(x_i' \epsilon_i \epsilon_i' x_i))'\end{aligned}\tag{V.1}$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i \epsilon_i' x_i))(\text{vec}(x_i' \epsilon_i (\widehat{\beta} - \beta)' x_i' x_i))'\tag{V.2}$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i \epsilon_i' x_i))(\text{vec}(x_i' x_i (\widehat{\beta} - \beta) \epsilon_i' x_i))'\tag{V.3}$$

$$+ \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i \epsilon_i' x_i))(\text{vec}(x_i' x_i (\widehat{\beta} - \beta) (\widehat{\beta} - \beta)' x_i' x_i))'\tag{V.4}$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i (\widehat{\beta} - \beta)' x_i' x_i))(\text{vec}(x_i' \epsilon_i \epsilon_i' x_i))'\tag{V.5}$$

$$+ \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i (\widehat{\beta} - \beta)' x_i' x_i))(\text{vec}(x_i' \epsilon_i (\widehat{\beta} - \beta)' x_i' x_i))'\tag{V.6}$$

$$+ \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i (\widehat{\beta} - \beta)' x_i' x_i))(\text{vec}(x_i' x_i (\widehat{\beta} - \beta) \epsilon_i' x_i))'\tag{V.7}$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x_i' \epsilon_i (\widehat{\beta} - \beta)' x_i' x_i))(\text{vec}(x_i' x_i (\widehat{\beta} - \beta) (\widehat{\beta} - \beta)' x_i' x_i))'\tag{V.8}$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x_i' x_i (\widehat{\beta} - \beta) \epsilon_i' x_i))(\text{vec}(x_i' \epsilon_i \epsilon_i' x_i))'\tag{V.9}$$

$$+ \frac{1}{nT} \sum_i (\text{vec}(x'_i x_i (\hat{\beta} - \beta) \epsilon'_i x_i)) (\text{vec}(x'_i \epsilon_i (\hat{\beta} - \beta)' x'_i x_i))' \quad (\text{V.10})$$

$$+ \frac{1}{nT} \sum_i (\text{vec}(x'_i x_i (\hat{\beta} - \beta) \epsilon'_i x_i)) (\text{vec}(x'_i x_i (\hat{\beta} - \beta) \epsilon'_i x_i))' \quad (\text{V.11})$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x'_i x_i (\hat{\beta} - \beta) \epsilon'_i x_i)) (\text{vec}(x'_i x_i (\hat{\beta} - \beta) (\hat{\beta} - \beta)' x'_i x_i))' \quad (\text{V.12})$$

$$+ \frac{1}{nT} \sum_i (\text{vec}(x'_i x_i (\hat{\beta} - \beta) (\hat{\beta} - \beta)' x'_i x_i)) (\text{vec}(x'_i \epsilon_i \epsilon'_i x_i))' \quad (\text{V.13})$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x'_i x_i (\hat{\beta} - \beta) (\hat{\beta} - \beta)' x'_i x_i)) (\text{vec}(x'_i \epsilon_i (\hat{\beta} - \beta)' x'_i x_i))' \quad (\text{V.14})$$

$$- \frac{1}{nT} \sum_i (\text{vec}(x'_i x_i (\hat{\beta} - \beta) (\hat{\beta} - \beta)' x'_i x_i)) (\text{vec}(x'_i x_i (\hat{\beta} - \beta) \epsilon'_i x_i))' \quad (\text{V.15})$$

$$+ \frac{1}{nT} \sum_i (\text{vec}(x'_i x_i (\hat{\beta} - \beta) (\hat{\beta} - \beta)' x'_i x_i)) (\text{vec}(x'_i x_i (\hat{\beta} - \beta) (\hat{\beta} - \beta)' x'_i x_i))'. \quad (\text{V.16})$$

Theorems 1-4 then follow by examining the properties of equations (O.1)-(O.4) and (V.1)-(V.16).

APPENDIX B. PROOF OF THEOREM 1

(i) $\hat{\beta} - \beta \xrightarrow{p} 0$ and $\sqrt{nT}(\hat{\beta} - \beta) \xrightarrow{d} Q^{-1}N(0, W = \lim_n \frac{1}{nT} \sum_{i=1}^n E[x'_i \Omega_i x_i])$ follow immediately under the conditions of Theorem 1 from the Markov LLN and the Liapounov CLT.

We also have that $E\|x'_i x_i / T\|^{2+2\delta} \stackrel{(a)}{\leq} E\|x_i / \sqrt{T}\|^{4+4\delta} \stackrel{(b)}{\leq} \frac{[\sum_t \sum_h (E|x'_{ith}|^{2+2\delta})^{1/(2+2\delta)}]^{2+2\delta}}{T^{2+2\delta}} \stackrel{(c)}{<} \frac{(kT)^{2+2\delta} \Delta}{T^{2+2\delta}}$, where (a) is by the Cauchy-Schwarz inequality, (b) follows from the definition of $\|A\|$ and Minkowski's inequality, and (c) is by $E|x_{ith}|^{4+\delta} < \Delta$. Making use of $E|\epsilon_{ith}|^{4+\delta} < \Delta$, $E\|x'_i \epsilon_i / T\|^{2+2\delta} < k^{2+2\delta} \Delta$ follows similarly. Noting that $E[x'_i x_i \otimes \epsilon'_i x_i] = 0$ by Assumption 3.a, it follows that terms (O.2)-(O.4) of \widehat{W} are $o_p(n^{-1/2})$ by the Markov LLN.¹ The Markov LLN also yields $\frac{1}{nT} \sum_i x'_i \epsilon_i \epsilon'_i x_i \xrightarrow{p} W$, which implies $\widehat{W} \xrightarrow{p} W$.

(ii) In addition, under $E|x_{ith}|^{8+\delta} < \Delta$ and $E|\epsilon_{ith}|^{8+\delta} < \Delta$, $E\|x'_i \epsilon_i / T\|^{4+4\delta} < k^{4+4\delta} \Delta$ and $E\|x'_i x_i / T\|^{4+4\delta} < k^{4+4\delta} \Delta$ follow by an argument similar to that used to show $E\|x'_i x_i / T\|^{2+2\delta} < k^{2+2\delta} \Delta$. Then, using that (O.2)-(O.4) of \widehat{W} are $o_p(n^{-1/2})$, it follows that

$$\sqrt{nT}[\text{vec}(\widehat{W} - W)] = \sqrt{nT}[\text{vec}(\frac{1}{nT} \sum_i x'_i \epsilon_i \epsilon'_i x_i - W)] + o_p(1) \xrightarrow{d} N(0, V)$$

where $V = \lim_n \frac{1}{nT} \sum_{i=1}^n E[(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i - W))(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i - W))']$ by the Liapounov CLT.

It is also straightforward to show that $\hat{\beta} - \beta \xrightarrow{p} 0$, $E\|x'_i \epsilon_i / T\|^{4+4\delta} < k^{4+4\delta} \Delta$, and $E\|x'_i x_i / T\|^{4+4\delta} < k^{4+4\delta} \Delta$ imply that terms (V.2)-(V.16) of \widehat{V} are $o_p(1)$ and that $E\|x'_i \epsilon_i / T\|^{4+4\delta} < k^{4+4\delta} \Delta$ implies

$$\frac{1}{nT} \sum_i [(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i))(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i))' - E[(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i))(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i))']] \xrightarrow{p} 0.$$

¹Under Assumption 3.b, (O.2)-(O.4) will be $o_p(1)$ and consistency of \widehat{W} follows.

We then have $\widehat{V} - V \xrightarrow{p} 0$. ■

APPENDIX C. PROOF OF THEOREM 2

(i) $E\|x'_i x_i / T\|^{2+2\delta} \stackrel{(a)}{\leq} E\|x_i / \sqrt{T}\|^{4+4\delta} \stackrel{(b)}{\leq} \frac{[\sum_t \sum_n (E|x_{itth}|^{2+2\delta})^{1/(2+2\delta)}]^{2+2\delta}}{T^{2+2\delta}} \stackrel{(c)}{<} \frac{(kT)^{2+2\delta} \Delta}{T^{2+2\delta}}$, where (a) is by the Cauchy-Schwarz inequality, (b) follows from the definition of $\|A\|$ and Minkowski's inequality, and (c) is by $E|x_{itth}|^{4+2\delta} < \Delta$. Making use of $E|\epsilon_{itth}|^{4+2\delta} < \Delta$, $E\|x'_i \epsilon_i / T\|^{2+2\delta} < k^{2+2\delta} \Delta$ follows similarly. Then Lemma A.2 gives $\frac{1}{n} \sum_i x'_i x_i / T \xrightarrow{p} Q$ and $\frac{1}{n} \sum_i x'_i \epsilon_i / T \xrightarrow{p} 0$ as $\{n, T\} \rightarrow \infty$ jointly, so $\widehat{\beta} - \beta \xrightarrow{p} 0$. In addition, since $E\|x'_i \epsilon_i / T\|^{2+2\delta} < k^{2+2\delta} \Delta$, we have that $\frac{1}{\sqrt{n}} \sum_i x'_i \epsilon_i / T \xrightarrow{d} N(0, W = \lim_{n,T} \frac{1}{nT^2} \sum_{i=1}^n E[x'_i \Omega_i x_i])$ by Lemma A.3, so $\sqrt{n}(\widehat{\beta} - \beta) \xrightarrow{d} Q^{-1}N(0, W)$.

Now consider term (O.2) of \widehat{W}/T :

$$\text{vec}\left(\frac{1}{nT^2} \sum_i x'_i x_i (\widehat{\beta} - \beta) \epsilon'_i x_i\right) = \frac{1}{\sqrt{n}} \left(\frac{1}{nT^2} \sum_i x'_i \epsilon_i \otimes x'_i x_i\right) \sqrt{n}(\widehat{\beta} - \beta) = \frac{1}{\sqrt{n}} O_p(1) O_p(1)$$

by Lemma A.2 since $E\|x'_i \epsilon_i / T \otimes x'_i x_i / T\|^{1+\delta} \leq (E\|x'_i \epsilon_i / T\|^{2+2\delta} E\|x'_i x_i / T\|^{2+2\delta})^{1/2} < C$ by Lemma A.1 and the argument in the preceding paragraph. That (O.3) is $O_p(1/\sqrt{n})$ and (O.4) is $O_p(1/n)$ follow similarly. Finally, $E\|x'_i \epsilon_i \epsilon'_i x_i / T^2\|^{1+\delta} \leq E\|x'_i \epsilon_i / T\|^{2+2\delta} < C$ where the first inequality follows from the Cauchy-Schwarz inequality and the second inequality is proven above. Hence, by Lemma A.2 $\frac{1}{nT^2} \sum_i \text{vec}(x'_i \epsilon_i \epsilon'_i x_i - E[x'_i \epsilon_i \epsilon'_i x_i]) = o_p(1)$. It follows immediately that $\widehat{W}/T - W \xrightarrow{p} 0$.

(ii) Under Assumption 3.a, we also have that $\frac{1}{\sqrt{n}} \left(\frac{1}{nT^2} \sum_i x'_i \epsilon_i \otimes x'_i x_i\right) \sqrt{n}(\widehat{\beta} - \beta) = \frac{1}{\sqrt{n}} o_p(1) O_p(1) = o_p\left(\frac{1}{\sqrt{n}}\right)$ since $E[x'_i \epsilon_i \otimes x'_i x_i] = 0$. Similarly, $\frac{1}{\sqrt{n}} \left(\frac{1}{nT^2} \sum_i x'_i x_i \otimes \epsilon'_i x_i\right) = o_p\left(\frac{1}{\sqrt{n}}\right)$. It follows from Lemma A.3 that $\sqrt{n}(\widehat{W}/T - W) = \sqrt{n} \left(\frac{1}{nT^2} \sum_i \text{vec}(x'_i \epsilon_i \epsilon'_i x_i - E[x'_i \epsilon_i \epsilon'_i x_i])\right) + o_p(1) \xrightarrow{d} N(0, V)$ where

$$V = \lim_{n,T} \frac{1}{nT^4} \sum_{i=1}^n E[(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i - W))(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i - W))']$$

since $E\|x'_i \epsilon_i \epsilon'_i x_i / T^2\|^{2+\delta} \leq E\|x'_i \epsilon_i / T\|^{4+2\delta} < C$ by an argument similar to that used to show $E\|x'_i \epsilon_i / T\|^{2+2\delta} < k^{2+2\delta} \Delta$ as long as $E|x_{itth}|^{8+\delta} < \Delta$ and $E|\epsilon_{itth}|^{8+\delta} < \Delta$.

To show $\widehat{V}/T^3 - V \xrightarrow{p} 0$, consider

$$E\|\text{vec}(x'_i \epsilon_i \epsilon'_i x_i / T^2) \text{vec}(x'_i \epsilon_i \epsilon'_i x_i / T^2)'\|^{1+\delta} \leq E\|x'_i \epsilon_i / T\|^{4+4\delta} < C$$

where the first inequality is by repeated application of the Cauchy-Schwarz inequality and the second is by an argument similar to that used to show $E\|x'_i \epsilon_i / T\|^{2+2\delta} < k^{2+2\delta} \Delta$ which holds if $E|x_{itth}|^{8+\delta} < \Delta$ and $E|\epsilon_{itth}|^{8+\delta} < \Delta$. It then follows by Lemma A.2 that

$$\frac{1}{nT^4} \sum_i [\text{vec}(x'_i \epsilon_i \epsilon'_i x_i) \text{vec}(x'_i \epsilon_i \epsilon'_i x_i)' - E[\text{vec}(x'_i \epsilon_i \epsilon'_i x_i) \text{vec}(x'_i \epsilon_i \epsilon'_i x_i)']] \xrightarrow{p} 0.$$

Turning to (V.2), we have

$$\begin{aligned} & \text{vec}[(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i / T^2))(\text{vec}(x'_i \epsilon_i (\widehat{\beta} - \beta)' x'_i x_i / T^2))'] \\ &= [(x'_i \epsilon_i / T \otimes x'_i \epsilon_i / T) \otimes (x'_i x_i / T \otimes \epsilon'_i x_i / T)] \text{vec}(\widehat{\beta} - \beta). \end{aligned}$$

$\mathbb{E}\|(x'_i \epsilon_i / T \otimes x'_i \epsilon_i / T) \otimes (x'_i x_i / T \otimes \epsilon'_i x_i / T)\|^{1+\delta} \leq [(\mathbb{E}\|x'_i \epsilon_i / T\|^{4+4\delta})^3 \mathbb{E}\|x'_i x_i\|^{4+4\delta}]^{1/4} < C$ where the first inequality is by repeated application of Lemma A.1 and the second inequality is by the moment conditions. It then follows from Lemma A.2 that $\frac{1}{n} \sum_i (x'_i \epsilon_i / T \otimes x'_i \epsilon_i / T) \otimes (x'_i x_i / T \otimes \epsilon'_i x_i / T) = O_p(1)$, so (V.2) is $o_p(1)$. Using similar arguments, it is also straightforward to show that terms (V.3)-(V.16) are $o_p(1)$, and the conclusion follows. ■

APPENDIX D. PROOF OF THEOREM 3

(i) $\mathbb{E}\|x'_i x_i / T\|^{2+2\delta} \stackrel{(a)}{\leq} \mathbb{E}\|x_i / \sqrt{T}\|^{4+4\delta} \stackrel{(b)}{\leq} \frac{[\sum_t \sum_h (\mathbb{E}|x_{it}^2|^{2+2\delta})^{1/(2+2\delta)}]^{2+2\delta}}{T^{2+2\delta}} \stackrel{(c)}{<} \frac{(kT)^{2+2\delta} \Delta}{T^{2+2\delta}}$, where (a) is by the Cauchy-Schwarz inequality, (b) by definition of $\|A\|$ and Minkowski's inequality, and (c) by $\mathbb{E}|x_{it}|^{r+\delta} < \Delta$. Also, $\mathbb{E}\|x'_i \epsilon_i / \sqrt{T}\|^{2+2\delta} < C$ by Lemma A.4, $\mathbb{E}|\epsilon_{it}|^{r+\delta} < \Delta$, and the mixing condition that α is of size $-3r/(r-4)$ for $r > 4$. It follows by Lemmas A.2 and A.3 that $\sqrt{nT}(\widehat{\beta} - \beta) \xrightarrow{d} Q^{-1}N(0, W = \lim_{n,T} \frac{1}{nT} \sum_{i=1}^n \mathbb{E}[x'_i \Xi_i x_i])$ as $\{n, T\} \rightarrow \infty$.

Now consider term (O.2) of \widehat{W} : $\text{vec}(\frac{1}{nT} \sum_i x'_i x_i (\widehat{\beta} - \beta) \epsilon'_i x_i) = (\frac{1}{(nT)^{3/2}} \sum_i x'_i \epsilon_i \otimes x'_i x_i) \sqrt{nT}(\widehat{\beta} - \beta) = \frac{1}{\sqrt{n}} o_p(1) O_p(1)$ by Lemma A.2 since $\mathbb{E}[x'_i \epsilon_i \otimes x'_i x_i] = 0$ by Assumption 3.a and $\mathbb{E}\|x'_i \epsilon_i / \sqrt{T} \otimes x'_i x_i / T\|^{1+\delta} \leq (\mathbb{E}\|x'_i \epsilon_i / \sqrt{T}\|^{2+2\delta} \mathbb{E}\|x'_i x_i / T\|^{2+2\delta})^{1/2} < C$ by Lemma A.1 and the argument in the preceding paragraph. That (O.3) is $o_p(1/\sqrt{n})$ and (O.4) is $O_p(1/n)$ follow similarly.² Finally, $\mathbb{E}\|x'_i \epsilon_i \epsilon'_i x_i / T\|^{1+\delta} \leq \mathbb{E}\|x'_i \epsilon_i / \sqrt{T}\|^{2+2\delta} < C$ by the Cauchy-Schwarz inequality and the preceding argument. It then follows that $\widehat{W} - W \xrightarrow{p} 0$ by Lemma A.2.

(ii) In addition, under the mixing condition that α is of size $-7r/(r-8)$ for $r > 8$, $\mathbb{E}|x_{it}|^{r+\delta} < \Delta$, and $\mathbb{E}|\epsilon_{it}|^{r+\delta} < \Delta$, $\mathbb{E}\|x'_i \epsilon_i / \sqrt{T}\|^{4+4\delta} < C$ follows by an argument similar to that used to show $\mathbb{E}\|x'_i \epsilon_i / \sqrt{T}\|^{2+2\delta} < C$. Then, since $\mathbb{E}\|x'_i \epsilon_i \epsilon'_i x_i / T\|^{2+\delta} \leq \mathbb{E}\|x'_i \epsilon_i / \sqrt{T}\|^{4+4\delta} < C$ and using that (O.2)-(O.4) of \widehat{W} are $o_p(n^{-1/2})$, the second conclusion follows from Lemma A.3 since

$$\sqrt{n}[\text{vec}(\widehat{W} - W)] = \sqrt{n}[\text{vec}(\frac{1}{nT} \sum_i x'_i \epsilon_i \epsilon'_i x_i - W)] + o_p(1) \xrightarrow{d} N(0, V)$$

where

$$V = \lim_{n,T} \frac{1}{nT^2} \sum_{i=1}^n \mathbb{E}[(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i - W))(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i - W))']$$

To show $\widehat{V}/T - V \xrightarrow{p} 0$, consider

$$\mathbb{E}\|\text{vec}(x'_i \epsilon_i \epsilon'_i x_i / T) \text{vec}(x'_i \epsilon_i \epsilon'_i x_i / T)'\|^{1+\delta} \leq \mathbb{E}\|x'_i \epsilon_i / \sqrt{T}\|^{4+4\delta} < C$$

where the first inequality is by repeated application of the Cauchy-Schwarz inequality and the second is by an argument similar to that used above which holds if $\mathbb{E}|\epsilon_{it}|^{8+\delta} < \Delta < \infty$, $\mathbb{E}|x_{it}|^{8+\delta} < \Delta < \infty$, and the strong mixing coefficient α is of size $-7r/(r-8)$ for $r > 8$. It then follows from Lemma A.2 that

$$\frac{1}{nT^2} \sum_i [\text{vec}(x'_i \epsilon_i \epsilon'_i x_i) \text{vec}(x'_i \epsilon_i \epsilon'_i x_i)' - \mathbb{E}[\text{vec}(x'_i \epsilon_i \epsilon'_i x_i) \text{vec}(x'_i \epsilon_i \epsilon'_i x_i)']] \xrightarrow{p} 0.$$

²Under Assumption 3.b, (O.2)-(O.4) will be $o_p(1)$ and consistency of \widehat{W} follows.

Turning to (V.2), we have

$$\begin{aligned} & \frac{1}{n^{3/2}} \sum_i \text{vec}[(\text{vec}(x'_i \epsilon_i \epsilon'_i x_i / T)) (\text{vec}(x'_i \epsilon_i (\sqrt{nT}(\hat{\beta} - \beta))' x'_i x_i / T^{3/2}))'] \\ &= \frac{1}{n^{3/2}} \sum_i [(x'_i \epsilon_i / \sqrt{T} \otimes x'_i \epsilon_i / \sqrt{T}) \otimes (x'_i x_i / T \otimes \epsilon'_i x_i / \sqrt{T})] \text{vec}(\sqrt{nT}(\hat{\beta} - \beta)). \end{aligned}$$

By Lemma A.2, we have that $\frac{1}{n} \sum_i (x'_i \epsilon_i / \sqrt{T} \otimes x'_i \epsilon_i / \sqrt{T}) \otimes (x'_i x_i / T \otimes \epsilon'_i x_i / \sqrt{T}) = O_p(1)$ since

$$\mathbb{E} \|(x'_i \epsilon_i / \sqrt{T} \otimes x'_i \epsilon_i / \sqrt{T}) \otimes (x'_i x_i / T \otimes \epsilon'_i x_i / \sqrt{T})\|^{1+\delta} \leq [(\mathbb{E} \|x'_i \epsilon_i / \sqrt{T}\|^{4+4\delta})^3 \mathbb{E} \|x'_i x_i\|^{4+4\delta}]^{1/4} < C$$

where the first inequality is by repeated application of Lemma A.1 and the second inequality is by the moment and mixing conditions. It then follows that (V.2) is $O_p(1/\sqrt{n})$. It can similarly be shown that terms (V.3)-(V.16) are $O_p(1/\sqrt{n})$, and the conclusion follows. ■

APPENDIX E. PROOF OF THEOREM 4

Under the hypotheses of the theorem, $\sqrt{n}(\hat{\beta} - \beta) \xrightarrow{d} Q^{-1}N(0, W)$, $x'_i x_i / T - Q_i \xrightarrow{p} 0$, and $x'_i \epsilon_i / \sqrt{T} \xrightarrow{d} N(0, W_i)$ are immediate from a LLN and CLT for mixing sequences, cf. White (2001) Theorems 3.47 and 5.20. The conclusion then follows from the definition of \widehat{W} and $\widehat{\epsilon}_i$.

APPENDIX F. PROOF OF COROLLARY 4.1

Consider $t^* = \frac{\sqrt{nT}(R\hat{\beta} - r)}{\sqrt{R\widehat{Q}^{-1}\widehat{W}\widehat{Q}^{-1}R'}}$. Under the null hypothesis, $R\beta = r$, so the numerator of t^* is $\sqrt{nT}R(\hat{\beta} - \beta) = R(\frac{1}{nT} \sum_i x'_i x_i)^{-1} (\frac{1}{\sqrt{nT}} \sum_i x'_i \epsilon_i) \xrightarrow{d} RQ^{-1}\Lambda \sum_i B_i / \sqrt{n}$. From Theorem 4 and the hypotheses of the Corollary, the denominator of t^* converges in distribution to $\sqrt{RQ^{-1}\frac{1}{n}\Lambda(\sum_{i=1}^n B_i B'_i - \frac{1}{n} \sum_{i=1}^n B_i \sum_{i=1}^n B'_i)\Lambda Q^{-1}R'}$. It follows from the Continuous Mapping Theorem that

$$t^* \xrightarrow{d} \frac{RQ^{-1}\Lambda \sum_i B_i / \sqrt{n}}{\sqrt{\frac{1}{n}RQ^{-1}\Lambda(\sum_{i=1}^n B_i B'_i - \frac{1}{n} \sum_{i=1}^n B_i \sum_{i=1}^n B'_i)\Lambda Q^{-1}R'}}$$

Define $\delta = (RQ^{-1}\Lambda\Lambda Q^{-1}R')^{1/2}$, so

$$\begin{aligned} t^* \xrightarrow{d} U &= \frac{\delta \sum_i B_{1,i} / \sqrt{n}}{\sqrt{\frac{\delta^2}{n} (\sum_{i=1}^n B_{1,i} B'_{1,i} - \frac{1}{n} \sum_{i=1}^n B_{1,i} \sum_{i=1}^n B'_{1,i})}} \\ &= \frac{\widetilde{B}_{1,n}}{\sqrt{\frac{1}{n} (\sum_i B_{1,i}^2 - \widetilde{B}_{1,n}^2)}}. \end{aligned}$$

It is straightforward to show that $\widetilde{B}_{1,n} \sim N(0, 1)$, that $\sum_i B_{1,i}^2 - \widetilde{B}_{1,n}^2 \sim \chi_{n-1}^2$, and that $\sum_i B_{1,i}^2 - \widetilde{B}_{1,n}^2$ and $\widetilde{B}_{1,n}$ are independent, from which it follows that

$$U = \left(\frac{n}{n-1}\right)^{1/2} \frac{\widetilde{B}_{1,n}}{\sqrt{(\sum_i B_{1,i}^2 - \widetilde{B}_{1,n}^2)/(n-1)}} \sim \left(\frac{n}{n-1}\right)^{1/2} t_{n-1}.$$

The result for F^* is obtained through a similar argument. ■

APPENDIX G. PROOF OF LEMMA A.2 AND LEMMA A.3

Proof of Lemma A.2 Instead of proving Lemma A.2 directly, we prove the following: If $\{Z_{i,T}\}$ are independent across i for all T with $E[Z_{i,T}] = 0$ and $E|Z_{i,T}|^{1+\delta} < \Delta < \infty$ for some $\delta > 0$ and all i, T , $\frac{1}{n} \sum_i Z_{i,T} \xrightarrow{P} 0$ as $\{n, T\} \rightarrow \infty$ jointly. Lemma A.2 then follows immediately from an argument similar to that used to prove White (2001) Corollary 3.9.

Define $Y_{i,T} = Z_{i,T}1(|Z_{i,T}| \leq n)$. Then $Var(\sum_i Y_{i,T}/n) = \sum_i Var(Y_{i,T}/n)$ by independence of $Z_{i,T}$ across i . Now $\sum_i Var(Y_{i,T}/n) \leq E(Y_{i,T}/n)^2 = \sum_i \int_{|Z| \leq n} (Z^2/n^2) dF_{i,T}(Z)$ where $F_{i,T}$ is the distribution function of $Z_{i,T}$. $Z^2/n^2 \leq Z^{1+\delta}/n^{1+\delta}$ for $|Z| \leq n$ implies

$$\sum_i \int_{|Z| \leq n} (Z^2/n^2) dF_{i,T}(Z) \leq \sum_i \int_{|Z| \leq n} (Z^{1+\delta}/n^{1+\delta}) dF_{i,T}(Z) \leq \sum_i \int (Z^{1+\delta}/n^{1+\delta}) dF_{i,T}(Z) < \Delta/n^\delta$$

where the last inequality results from $E|Z_{i,T}|^{1+\delta} < \Delta$. It follows that

$$Var(\sum_i Y_{i,T}/n) < \Delta/n^\delta \rightarrow 0 \tag{1}$$

as $\{n, T\} \rightarrow \infty$ jointly.

Now consider

$$\begin{aligned} |E \frac{1}{n} \sum_i Y_{i,T}| &= |\sum_i \int_{|Z| \leq n} (Z/n) dF_{i,T}(Z)| \\ &= |\sum_i \int (Z/n) dF_{i,T}(Z) - \sum_i \int_{|Z| > n} (Z/n) dF_{i,T}(Z)| \\ &= |\sum_i \int_{|Z| > n} (Z/n) dF_{i,T}(Z)| \\ &\leq \sum_i \int_{|Z| > n} (|Z|/n) dF_{i,T}(Z) \end{aligned}$$

by the Triangle inequality and Jensen's inequality. For $|Z| > n$, $|Z|/n \leq |Z|^{1+\delta}/n^{1+\delta}$, so

$$\sum_i \int_{|Z| > n} \frac{|Z|}{n} dF_{i,T}(Z) \leq \sum_i \int_{|Z| > n} \frac{|Z|^{1+\delta}}{n^{1+\delta}} dF_{i,T}(Z) \leq \sum_i \int \frac{|Z|^{1+\delta}}{n^{1+\delta}} dF_{i,T}(Z) \leq \Delta/n^\delta \rightarrow 0, \tag{2}$$

which yields

$$|E \frac{1}{n} \sum_i Y_{i,T}| \rightarrow 0. \tag{3}$$

By Chebyshev's inequality and (1),

$$\lim_{n,T} P(|\sum_i (Y_{i,T} - E[Y_{i,T}])/n| \geq \epsilon) \leq \lim_{n,T} Var(\sum_i Y_{i,T}/n)/\epsilon^2 = 0,$$

so $\frac{1}{n} \sum_i Y_{i,T} - \frac{1}{n} \sum_i E[Y_{i,T}] \xrightarrow{P} 0$, which implies, using (3), that $\frac{1}{n} \sum_i Y_{i,T} \xrightarrow{P} 0$.

Finally, consider

$$\begin{aligned} P\left(\left|\frac{1}{n} \sum_i Z_{i,T} - \frac{1}{n} \sum_i Y_{i,T}\right| \geq \epsilon\right) &= P\left(\left|\frac{1}{n} \sum_i (1 - \mathbf{1}(|Z_{i,T}| \leq n)) Z_{i,T}\right| \geq \epsilon\right) \\ &\leq \mathbb{E}\left|\frac{1}{n} \sum_i (1 - \mathbf{1}(|Z_{i,T}| \leq n)) Z_{i,T}\right|/\epsilon \end{aligned}$$

by the Markov inequality. $\mathbb{E}\left|\frac{1}{n} \sum_i (1 - \mathbf{1}(|Z_{i,T}| \leq n)) Z_{i,T}\right| \leq \frac{1}{n} \sum_i \mathbb{E}|(1 - \mathbf{1}(|Z_{i,T}| \leq n)) Z_{i,T}|$ by the Triangle inequality, and

$$\begin{aligned} \frac{1}{n} \sum_i \mathbb{E}|(1 - \mathbf{1}(|Z_{i,T}| \leq n)) Z_{i,T}| &= \frac{1}{n} \sum_i \left[\int |Z| dF_{i,T}(Z) - \int_{|Z| \leq n} |Z| dF_{i,T}(Z) \right] \\ &= \sum_i \int_{|Z| > n} \frac{|Z|}{n} dF_{i,T}(Z) \rightarrow 0 \end{aligned}$$

by (2). It then follows that $\frac{1}{n} \sum_i Z_{i,T} - \frac{1}{n} \sum_i Y_{i,T} \xrightarrow{p} 0$ which, with $\frac{1}{n} \sum_i Y_{i,T} \xrightarrow{p} 0$, implies $\frac{1}{n} \sum_i Z_{i,T} \xrightarrow{p} 0$. \blacksquare

Proof of Lemma A.3 Define $\xi_{i,n,T} = \Omega_{n,T}^{-1/2} Z_{i,T}$ where $\Omega_{n,T} = \sum_i \Omega_{i,T}$. By the Cramer-Wold device, $\sum_i \xi_{i,n,T} \xrightarrow{d} N(0, I_k)$ as $\{n, T\} \rightarrow \infty$ jointly if $\forall c \in \mathbb{R}^k$ with $\|c\| = 1$, $c' \sum_i \xi_{i,n,T} \xrightarrow{d} N(0, 1)$ as $\{n, T\} \rightarrow \infty$ jointly. Then, by $\Omega = \lim_{n,T} \frac{1}{n} \sum_{i=1}^n \Omega_{i,T} > 0$, $\frac{1}{\sqrt{n}} \sum_{i=1}^n Z_{i,T} \xrightarrow{d} N(0, \Omega)$ as $\{n, T\} \rightarrow \infty$ jointly.

To establish $c' \sum_i \xi_{i,n,T} \xrightarrow{d} N(0, 1)$, it is sufficient to verify

$$\lim_{n,T} \sum_i \mathbb{E}[\xi_{i,n,T}^2 \mathbf{1}(|\xi_{i,n,T}^2| > \epsilon)] = 0. \quad (4)$$

For a given $\epsilon > 0$ and $c \in \mathbb{R}^k$ with $\|c\| = 1$,

$$\begin{aligned} c' \sum_i \mathbb{E}[\xi_{i,n,T} \xi'_{i,n,T} \mathbf{1}(|c' \xi_{i,n,T} \xi'_{i,n,T} c| > \epsilon)] c \\ = c' \Omega_{n,T}^{-1/2} \sum_i \mathbb{E}[Z_{i,T} Z'_{i,T} \mathbf{1}(|c' \Omega_{n,T}^{-1/2} Z_{i,T} Z'_{i,T} \Omega_{n,T}^{-1/2} c| > \epsilon)] \Omega_{n,T}^{-1/2} c. \end{aligned}$$

Considering first the indicator function, we have

$$\begin{aligned} \mathbf{1}(|c' \Omega_{n,T}^{-1/2} Z_{i,T} Z'_{i,T} \Omega_{n,T}^{-1/2} c| > \epsilon) &\leq \mathbf{1}(\|c\| \|\Omega_{n,T}^{-1/2} Z_{i,T} Z'_{i,T} \Omega_{n,T}^{-1/2}\| > \epsilon) \\ &\leq \mathbf{1}(\lambda_{\max}(\Omega_{n,T}^{-1}) \|Z_{i,T}\|^2 > \epsilon) \\ &= \mathbf{1}(\|Z_{i,T}\|^2 > \epsilon \lambda_{\min}(\Omega_{n,T})) \end{aligned}$$

Then

$$\begin{aligned}
& c' \Omega_{n,T}^{-1/2} \sum_i \mathbb{E}[Z_{i,T} Z'_{i,T} 1(|c' \Omega_{n,T}^{-1/2} Z_{i,T} Z'_{i,T} \Omega_{n,T}^{-1/2} c| > \epsilon)] \Omega_{n,T}^{-1/2} c \\
& \leq \|c\|^2 \|\Omega_{n,T}^{-1/2}\| \sum_i \mathbb{E}[Z_{i,T} Z'_{i,T} 1(|c' \Omega_{n,T}^{-1/2} Z_{i,T} Z'_{i,T} \Omega_{n,T}^{-1/2} c| > \epsilon)] \Omega_{n,T}^{-1/2} \| \\
& \leq \lambda_{\max}(\Omega_{n,T}^{-1}) \|\sum_i \mathbb{E}[Z_{i,T} Z'_{i,T} 1(\|Z_{i,T}\|^2 > \epsilon \lambda_{\min}(\Omega_{n,T}))]\| \\
& \leq \frac{1}{\lambda_{\min}(\Omega_{n,T})} \sum_i \mathbb{E}[\|Z_{i,T}\|^2 1(\|Z_{i,T}\|^2 > \epsilon \lambda_{\min}(\Omega_{n,T}))] \\
& \leq \frac{1}{\lambda_{\min}(\Omega_{n,T})} \sum_i \frac{\mathbb{E}\|Z_{i,T}\|^{2+\delta}}{(\epsilon \lambda_{\min}(\Omega_{n,T}))^\delta} \\
& \leq \frac{n\Delta}{\epsilon^\delta [n\lambda_{\min}(\frac{1}{n} \sum_i \Omega_{i,T})]^{1+\delta}} = \frac{\Delta}{\epsilon^\delta n^\delta \lambda_{\min}^{1+\delta}} \rightarrow 0
\end{aligned}$$

as $\{n, T\} \rightarrow \infty$ jointly, and it follows that (4) $\rightarrow 0$ as $\{n, T\} \rightarrow \infty$ jointly.

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