

SUPPLEMENT TO ‘SIEVE-BASED INFERENCE FOR INFINITE-VARIANCE LINEAR PROCESSES’

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S.1. Introduction. This supplement contains additional theoretical results and proofs for the theory stated in Cavaliere, Georgiev and Taylor (2015), CGT hereafter. The supplement is organized as follows. Section S.2 provides a lemma with two tail inequalities regarding the series of the coefficients from the AR(∞) representations. Section S.3 reports the proof of Lemma 2 and corollaries from Section 6 in CGT. Section S.4 contains proofs of the results given in Section 7.3.1 of CGT. Finally, Section S.5 discusses the case of multiple restrictions.

S.2. A Tail Inequality. We first establish two inequalities between the tails of the series of autoregressive coefficients and their powers.

LEMMA S.1. *Under Assumption 1, let $k^2/T + 1/k \rightarrow 0$ as $T \rightarrow \infty$. Then for large T , for η in a sufficiently small left neighborhood of $\alpha \wedge 1 := \min\{\alpha, 1\}$ and for $\zeta > 0$ sufficiently small, it holds that*

$$(S.2.1) \quad \tilde{a}_T a_T^{-2} (k \sum_{j=k+1}^{\infty} |\beta_j|^{\eta})^{1/\eta} \leq \tilde{a}_T a_T^{-\zeta-2} + \sum_{j=k+1}^{\infty} |\beta_j|.$$

If $k^3/T + 1/k \rightarrow 0$ as $T \rightarrow \infty$, then also

$$(S.2.2) \quad \tilde{a}_T a_T^{-2} (k \sum_{j=k+1}^{\infty} |\beta_j|^{\eta})^{1/\eta} \leq \tilde{a}_T a_T^{-9/4} + a_T^{-1/5} \sum_{j=k+1}^{\infty} |\beta_j|.$$

PROOF. In the case of a finite-order AR representation the inequality is obvious, so we discuss the opposite case.

From $\sum_{j=1}^{\infty} j^{2/\delta} |\beta_j| < \infty$ it follows that $|\beta_j| \leq j^{-2/\delta}$ for large j . For fixed k , the expression $(\sum_{j=k+1}^{\infty} |\beta_j|)^{-\eta} \sum_{j=k+1}^{\infty} |\beta_j|^{\eta}$ cannot be prolonged by continuity to the zero sequence in ℓ_2 , so we consider separately the sets

$$\begin{aligned} \mathbb{B}_l &:= \{ \{ |\beta_j| \}_{j=k+1}^{\infty} : 0 \leq |\beta_j| \leq a_T^{-\zeta} j^{-2/\delta}, j = k+1, \dots \}, \\ \mathbb{B}_u &:= \{ \{ |\beta_j| \}_{j=k+1}^{\infty} : a_T^{-\zeta} j^{-2/\delta} \leq |\beta_j| \leq j^{-2/\delta}, j = k+1, \dots \}. \end{aligned}$$

Using the upper bound in the standard estimate

$$(S.2.3) \quad \frac{(K+1)^{1-s}}{s-1} \leq \sum_{j=K+1}^{\infty} j^{-s} \leq \frac{K^{1-s}}{s-1} \text{ for } s > 1,$$

we find that on \mathbb{B}_l ,

$$\tilde{a}_T a_T^{-2} (k \sum_{j=k+1}^{\infty} |\beta_j|^{\eta})^{1/\eta} \leq C \tilde{a}_T a_T^{-\zeta-2} (k+1)^{\frac{2(\delta-\eta)}{\delta\eta}} = o(\tilde{a}_T a_T^{-\zeta-2})$$

for $T \geq 2$, $\delta < \eta$ and sufficiently small $\zeta > 0$. If \mathbb{B}_u is equipped with the ℓ_2 metric, it becomes a compact in ℓ_2 , for it is closed, bounded and for every $\epsilon > 0$ there exists an $N \in \mathbb{N}$ such that for all $\{|\beta_j|\}_{j=k+1}^{\infty} \in \mathbb{B}_u$ it holds that $\sum_{j=N}^{\infty} \beta_j^2 \leq \sum_{j=N}^{\infty} j^{-4/\delta} < \epsilon$. The expression $(\sum_{j=k+1}^{\infty} |\beta_j|)^{-\eta} \sum_{j=k+1}^{\infty} |\beta_j|^{\eta}$ defines a continuous real function on \mathbb{B}_u and, hence, attains a maximum there. Let $\{|\beta_j^\sharp|\}_{j=k+1}^{\infty}$ denote a maximizing sequence; by examining directional derivatives, it follows that $|\beta_j^\sharp|$ ($j = k+1, \dots$) satisfy

$$|\beta_j^\sharp| = \begin{cases} a_T^{-\zeta} j^{-2/\delta} & \text{if } B^\sharp < a_T^{-\zeta} j^{-2/\delta} \\ B^\sharp & \text{if } a_T^{-\zeta} j^{-2/\delta} \leq B^\sharp < j^{-2/\delta} \\ j^{-2/\delta} & \text{if } j^{-2/\delta} \leq B^\sharp \end{cases}$$

with $B^\sharp = \left(\sum_{j=k+1}^{\infty} |\beta_j^\sharp| / \sum_{j=k+1}^{\infty} |\beta_j^\sharp|^{\eta} \right)^{\frac{1}{1-\eta}} \neq 0$. We examine this condition without attempting to find all $|\beta_j^\sharp|$ exactly.

As $B^\sharp > 0$ and $j^{-2/\delta}$ is decreasing in j , $|\beta_j^\sharp| = j^{-2/\delta}$ necessarily holds from some index onwards. Let $K_2 \geq k$ be the smallest natural $\geq k$ such that $|\beta_j^\sharp| = j^{-2/\delta}$ for $j \geq K_2 + 1$. Then $(K_2 + 1)^{-2/\delta} \leq B^\sharp$ and, if $K_2 > k$, then $B^\sharp < K_2^{-2/\delta}$ and $|\beta_j^\sharp| \neq j^{-2/\delta}$ for $j = k+1, \dots, K_2$.

Still if $K_2 > k$, then either $|\beta_{k+1}^\sharp| = B^\sharp$ or $|\beta_{k+1}^\sharp| = a_T^{-\zeta} j^{-2/\delta} > B^\sharp$. In the former case it must be that $a_T^{-\zeta} (k+1)^{-2/\delta} \leq B^\sharp$, so $a_T^{-\zeta} j^{-2/\delta} \leq B^\sharp$ for all $j \geq k+1$, the value $a_T^{-\zeta} j^{-2/\delta}$ is never taken by $|\beta_j^\sharp|$ and at K_2 a switch between B^\sharp and $j^{-2/\delta}$ takes place; define $K_1 = k$ in this case. On the other hand, if $|\beta_{k+1}^\sharp| = a_T^{-\zeta} (k+1)^{-2/\delta} > B^\sharp$, let $K_1 < K_2$ be the largest natural j such that $a_T^{-\zeta} j^{-2/\delta} > B^\sharp$ for $j = k+1, \dots, K_1$. Then at K_1 a switch between $a_T^{-\zeta} j^{-2/\delta}$ and B^\sharp or $j^{-2/\delta}$ takes place.

Summarizing,

$$|\beta_j^\sharp| = \begin{cases} a_T^{-\zeta} j^{-2/\delta} & k+1 \leq j \leq K_1 \\ B^\sharp & K_1 + 1 \leq j \leq K_2 \\ j^{-2/\delta} & j \geq K_2 + 1, \dots, \end{cases}$$

with

$$B^\sharp = \left(\frac{a_T^{-\zeta} \sum_{j=k+1}^{K_1} j^{-2/\delta} + (K_2 - K_1) B^\sharp + \sum_{j=K_2+1}^{\infty} j^{-2/\delta}}{a_T^{-\zeta\eta} \sum_{j=k+1}^{K_1} j^{-2\eta/\delta} + (K_2 - K_1) B^{\sharp\eta} + \sum_{j=K_2+1}^{\infty} j^{-2\eta/\delta}} \right)^{\frac{1}{1-\eta}},$$

where the first two conditions may be satisfied by an empty set of j 's. If switches do occur, then $a_T^{-\zeta}(K_1 + 1)^{-2/\delta} \leq B^\sharp \leq a_T^{-\zeta} K_1^{-2/\delta}$ holds at a switch away from the $a_T^{-\zeta} j^{-2/\delta}$ branch, and $(1 + K_2)^{-2/\delta} \leq B^\sharp \leq K_2^{-2/\delta}$ at the start of the $j^{-2/\delta}$ branch.

Solving for B^\sharp in its defining equation gives

$$B^\sharp = \left(\frac{a_T^{-\zeta} \sum_{j=k+1}^{K_1} j^{-2/\delta} + \sum_{j=K_2+1}^{\infty} j^{-2/\delta}}{a_T^{-\zeta\eta} \sum_{j=k+1}^{K_1} j^{-2\eta/\delta} + \sum_{j=K_2+1}^{\infty} j^{-2\eta/\delta}} \right)^{\frac{1}{1-\eta}}$$

and using (S.2.3), it follows that B^\sharp satisfies

$$\begin{aligned} (S.2.4) \quad & \frac{a_T^{-\zeta} \{(k+1)^{1-2/\delta} - K_1^{1-2/\delta}\} + (K_2 + 1)^{1-2/\delta}}{a_T^{-\zeta\eta} (k^{1-2\eta/\delta} - (K_1 + 1)^{1-2\eta/\delta}) + K_2^{1-2\eta/\delta}} \leq \frac{2-\delta}{2\eta-\delta} (B^\sharp)^{1-\eta} \\ & \leq \frac{a_T^{-\zeta} k^{1-2/\delta} + K_2^{1-2/\delta}}{a_T^{-\zeta\eta} \{(k+1)^{1-2\eta/\delta} - K_1^{1-2\eta/\delta}\} + (K_2 + 1)^{1-2\eta/\delta}} \end{aligned}$$

We examine the implications of this inequality and the switching conditions for subsequential limits. Two cases are possible.

1. If two switches occur, then $K_2/K_1 \sim a_T^{\delta\zeta/2}$ from the switching conditions. Let $a_T^{-\zeta}(k/K_2)^{1-2/\delta} \rightarrow c$ as $T \rightarrow \infty$, possibly along a subsequence; we are looking for the values of c that can occur. Passing to the (subsequential) limit in (S.2.4), it follows that

$$c + 1 \leq \frac{2-\delta}{2\eta-\delta} (\lim K_2^{2/\delta} B^\sharp)^{1-\eta} \leq c + 1,$$

and since $K_2^{2/\delta} B^\sharp \rightarrow 1$, the unique subsequential limit is $c = 2(1-\eta)/(2\eta-\delta)$, and thus, it is the limit of $a_T^{-\zeta}(k/K_2)^{1-2/\delta}$ as $T \rightarrow \infty$. Further, since $a_T^2 K_1^{2/\delta} B_\sharp \rightarrow 1$, we find that $K_1 \sim c^{\frac{\delta}{2-\delta}} a_T^{\frac{\zeta\delta^2}{2(2-\delta)}} k$ and $K_2 \sim c^{\frac{\delta}{2-\delta}} a_T^{\frac{\zeta\delta}{2-\delta}} k$. Then

$$\begin{aligned} \frac{(\sum_{j=k+1}^{\infty} |\beta_j|^{\eta})^{1/\eta}}{\sum_{j=k+1}^{\infty} |\beta_j|} &= \frac{(a_T^{-\zeta\eta} \sum_{j=k+1}^{K_1} j^{-2\eta/\delta} + (K_2 - K_1) B^{\sharp\eta} + \sum_{j=K_2+1}^{\infty} j^{-2\eta/\delta})^{1/\eta}}{a_T^{-\zeta} \sum_{j=k+1}^{K_1} j^{-2/\delta} + (K_2 - K_1) B^\sharp + \sum_{j=K_2+1}^{\infty} j^{-2/\delta}} \\ &\sim \frac{(c_1 a_T^{-\zeta\eta} k^{1-2\eta/\delta} + c_2 a_T^{-\frac{\zeta(2\eta-\delta)}{2-\delta}} k^{1-2\eta/\delta})^{1/\eta}}{c_3 a_T^{-\zeta} k^{1-2/\delta}} \sim c_4 (a_T^{\frac{\zeta\delta}{2-\delta}} k)^{\frac{1-\eta}{\eta}} \end{aligned}$$

for some positive $c_{1,2,3,4}$, and

$$\tilde{a}_T a_T^{-2} \frac{(k \sum_{j=k+1}^{\infty} |\beta_j|^{\eta})^{1/\eta}}{\sum_{j=k+1}^{\infty} |\beta_j|} \sim c_4 \tilde{a}_T a_T^{-2} k^{1/\eta} (a_T^{\frac{\zeta\delta}{2-\delta}} k)^{\frac{1-\eta}{\eta}} = o(1)$$

for $k^2/T \rightarrow 0$, η in a sufficiently small left neighbourhood of $\alpha \wedge 1$ and $\zeta > 0$ sufficiently small. The same expression is $o(a_T^{-1/5})$ for $k^3/T \rightarrow 0$ and $\zeta = \frac{1}{4}$.

2. Alternatively, if the value of $a_T^{-\zeta} j^{-2/\delta}$ is not taken by any $|\beta_j^\sharp|$, we are left with

$$\frac{(K_2 + 1)^{1-2/\delta}}{K_2^{1-2\eta/\delta}} \leq \frac{2-\delta}{2\eta-\delta} (B^\sharp)^{1-\eta} \leq \frac{K_2^{1-2/\delta}}{(K_2 + 1)^{1-2\eta/\delta}}$$

or equivalently,

$$\left(\frac{2\eta-\delta}{2-\delta}\right)^{\frac{1}{1-\eta}} \left(1 + \frac{1}{K_2}\right)^{\frac{\delta-2}{\delta(1-\eta)}} K_2^{-\frac{2}{\delta}} \leq B^\sharp \leq \left(\frac{2\eta-\delta}{2-\delta}\right)^{\frac{1}{1-\eta}} \left(1 + \frac{1}{K_2}\right)^{\frac{2-\delta}{\delta(1-\eta)}} (K_2 + 1)^{-\frac{2}{\delta}}.$$

As $(K_2 + 1)^{-2/\delta} \leq B^\sharp \leq K_2^{-2/\delta}$ should also hold, it follows that

$$\frac{2\eta-\delta}{2-\delta} \left(1 + \frac{1}{K_2}\right)^{\frac{\delta-2}{\delta}} \leq 1 \leq \frac{2\eta-\delta}{2-\delta} \left(1 + \frac{1}{K_2}\right)^{\frac{2-\delta}{\delta}},$$

which is equivalent to $K_2 \leq \left[\{(2-\delta)/(2\eta-\delta)\}^{\frac{\delta}{2-\delta}} - 1\right]^{-1}$ for $\eta \in (\frac{\delta}{2}, 1)$. As this is inconsistent with $K_2 > k \rightarrow \infty$ as $T \rightarrow \infty$, for large T the maximizing sequence is $|\beta_j^\sharp| = j^{-2/\delta}$, $j \geq k + 1$. Therefore, $\sum_{j=1}^{\infty} j^{2/\delta} |\beta_j| < \infty$ implies for large T that

$$\frac{\sum_{j=k+1}^T |\beta_j|^\eta}{(\sum_{j=k+1}^T |\beta_j|)^\eta} \leq \frac{\sum_{j=k+1}^T j^{-2\eta/\delta}}{(\sum_{j=k+1}^T j^{-2/\delta})^\eta} \leq \frac{\frac{\delta}{2\eta-\delta} k^{1-2\eta/\delta}}{(\frac{\delta}{2-\delta})^\eta (k+1)^{\eta-2\eta/\delta}} \leq ck^{1-\eta}$$

for obvious choices of c , so

$$\tilde{a}_T a_T^{-2} \frac{(k \sum_{j=k+1}^{\infty} |\beta_j|^{\eta})^{1/\eta}}{\sum_{j=k+1}^{\infty} |\beta_j|} \leq c^{1/\eta} \tilde{a}_T a_T^{-2} k^{2/\eta-1} = \begin{cases} o(1) & \text{if } k^2/T \rightarrow 0 \\ o(a_T^{-1/5}) & \text{if } k^3/T \rightarrow 0 \end{cases}$$

for η smaller than but close to $\alpha \wedge 1$.

This proves the lemma. \square

S.3. Additional proofs. Together with the matrix norms $\|\cdot\|_2$ and $\|\cdot\|$ employed in the the paper, here we also use the linear space matrix norms $\|\cdot\|_1 := \sup_{\|x\|_1=1} \|(\cdot)x\|_1$ and $\|\cdot\|_\infty := \sup_{\|x\|_\infty=1} \|(\cdot)x\|_\infty$ induced respectively by the 1 and max vector norms.

S.3.1. Proof of Lemma 2. Regarding S_{00}^k of part (a), we argue first that $\|S_{00}^k \sigma_T^{-2} - \Sigma_k\|_2 = O_P(l_T \tilde{a}_T a_T^{-2}) \max\{k^\epsilon a_k, k\} = o_P(1)$ when $k^2/T \rightarrow 0$. Then $\lambda_{\min}(S_{00}^k \sigma_T^{-2}) \geq \lambda_{\min}(\Sigma_k) - \|S_{00}^k \sigma_T^{-2} - \Sigma_k\|_2 = \lambda_{\min}(\Sigma_k) + o_P(1)$ by Weyl's inequality (Seber, 2008, p.117), so $\lambda_{\min}(S_{00}^k \sigma_T^{-2})$ is bounded away from zero in probability and $(S_{00}^k)^{-1}$ exists with probability approaching one. Further we use the fact that

$$\|(S_{00}^k)^{-1} \sigma_T^2 - \Sigma_k^{-1}\|_2 \leq \frac{\|\Sigma_k^{-1}\|_2^2 \|S_{00}^k \sigma_T^{-2} - \Sigma_k\|_2}{1 - \|\Sigma_k^{-1}\|_2 \|S_{00}^k \sigma_T^{-2} - \Sigma_k\|_2}$$

if $\|\Sigma_k^{-1}\|_2 \|S_{00}^k \sigma_T^{-2} - \Sigma_k\|_2 < 1$. The latter inequality holds with probability approaching one since $\|\Sigma_k^{-1}\|_2$ is bounded as $k \rightarrow \infty$ and $\|S_{00}^k \sigma_T^{-2} - \Sigma_k\|_2 = o_P(1)$, so we can conclude that also

$$\|(S_{00}^k)^{-1} \sigma_T^2 - \Sigma_k^{-1}\|_2 = O_P(l_T \tilde{a}_T a_T^{-2}) \max\{k^\epsilon a_k, k\}.$$

The proof of part (a) is completed observing that $a_T^{-2} \sigma_T^2$ is bounded away from zero in probability as it converges in distribution to an a.s. positive ($\alpha/2$ -stable) r.v.

We present now the evaluation of $\|S_{00}^k \sigma_T^{-2} - \Sigma_k\|_2$. A generic element of S_{00}^k is $\sum_{t=k}^{T-1} X_{t-i} X_{t-j} = c_{ij}^k + \xi_{ij}^< + \xi_{ij}^>$ (for $0 \leq i, j \leq k-1$), where $c_{ij}^k := \sum_{t=k}^{T-1} \sum_{v=0}^{\infty} \varepsilon_{t-\max(i,j)-v}^2 \gamma_v \gamma_{v+|j-i|}$ and

$$\xi_{ij}^R := \sum_{t=k}^{T-1} \sum_{u,v=0}^{\infty} \mathbb{I}_{\{u \neq v+j-i\}} \gamma_u \gamma_v \varepsilon_{t-i-u} \varepsilon_{t-j-v} \mathbb{I}_{|\varepsilon_{t-i-u} \varepsilon_{t-j-v}| R \tilde{a}_T},$$

$R \in \{>, \leq\}$. With $C^k := (c_{ij}^k)_{i,j=0}^{k-1}$, it holds that

$$\begin{aligned} \|S_{00}^k - C^k\|_2 &\leq \|(\xi_{ij})_{i,j}^<\|_2 + \|(\xi_{0,|i-j|}^>)_{i,j}\|_2 + \|(\xi_{0,|i-j|}^> - \xi_{ij}^>)\|_2 \\ &\leq \|(\xi_{ij})_{i,j}^<\|_2 + \max_{i=0, \dots, k-1} \sum_{j=0}^{k-1} |\xi_{0,|i-j|}^>| + \sum_{i,j=0}^{k-1} |\xi_{0,|i-j|}^> - \xi_{ij}^>| \end{aligned}$$

since $\|(\xi_{0,|i-j|}^>)_{i,j}\|_2 \leq \max_{i=0, \dots, k-1} \sum_{j=0}^{k-1} |\xi_{0,|i-j|}^>|$ as $(\xi_{0,|i-j|}^>)_{i,j}$ is symmetric (in general, $\|\cdot\|_2 \leq \|\cdot\|_1^{1/2} \|\cdot\|_\infty^{1/2}$). Let first $E|\varepsilon_1| = \infty$ (so $\alpha \in (0, 1]$). Since $\|\cdot\|_2 \leq \|\cdot\|$, the inequalities can be continued as

$$\|S_{00}^k - C^k\|_2 \leq (\Xi_1^< + \Xi_2^<)^{1/2} + 2 \sum_{j=0}^{k-1} |\xi_{0,j}^>| + \sum_{i,j=0}^{k-1} |\xi_{0,|i-j|}^> - \xi_{ij}^>|,$$

where

$$\begin{aligned} \Xi_1^< &:= \sum_{i,j} \sum_{s,t=k}^{T-1} \sum_{a,b,u,v=0}^{\infty} \mathbb{I}_A \gamma_a \gamma_b \gamma_u \gamma_v \varepsilon_{s-i-a} \varepsilon_{s-j-b} \varepsilon_{t-i-u} \varepsilon_{t-j-v} \\ &\quad \times \mathbb{I}_{|\varepsilon_{s-i-a} \varepsilon_{s-j-b}| \leq \tilde{a}_T} \mathbb{I}_{|\varepsilon_{t-i-u} \varepsilon_{t-j-v}| \leq \tilde{a}_T}, \\ \Xi_2^< &:= \sum_{i,j} \sum_{s,t=k}^{T-1} \sum_{a,b,u,v=0}^{\infty} \mathbb{I}_{A^c} \gamma_a \gamma_b \gamma_u \gamma_v \varepsilon_{s-i-a} \varepsilon_{s-j-b} \varepsilon_{t-i-u} \varepsilon_{t-j-v} \\ &\quad \times \mathbb{I}_{|\varepsilon_{s-i-a} \varepsilon_{s-j-b}| \leq \tilde{a}_T} \mathbb{I}_{|\varepsilon_{t-i-u} \varepsilon_{t-j-v}| \leq \tilde{a}_T}, \end{aligned}$$

$A := \{\#(\{s - i - a, s - j - b, t - i - u, t - j - v\}) = 4\}$, $A^c := \{\#(\{s - i - a, s - j - b, t - i - u, t - j - v\}) = 2 \text{ or } 3\}$. Further,

$$\mathbb{E} |\Xi_1^\leq| \leq k^2 T^2 \mathbb{E}(|\varepsilon_1 \varepsilon_2| \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T})^2 (\sum_{u=0}^{\infty} |\gamma_u|)^4 = O(k^2 \tilde{a}_T^2)$$

for $\alpha \in (0, 1)$, as $\tilde{a}_T^{-1} T \mathbb{E}(|\varepsilon_1 \varepsilon_2| \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T}) \rightarrow \alpha/(1 - \alpha)$ by Karamata's theorem [KT], and $\mathbb{E} |\Xi_1^\leq| = O(k^2 l_T^2 \tilde{a}_T^2)$ for $\alpha = 1$, as $\tilde{a}_T^{-1} T \mathbb{E}(|\varepsilon_1 \varepsilon_2| \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T})$ is slowly varying in this case. Similarly,

$$\mathbb{E} |\Xi_2^\leq| \leq 4k^2 T \mathbb{E}(|\varepsilon_1^2 \varepsilon_2^2| \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T}) (\sum_{u=0}^{\infty} |\gamma_u|)^4 = O(k^2 \tilde{a}_T^2)$$

by KT, so $\Xi_1^\leq + \Xi_2^\leq = O_P(k^2 l_T^2 \tilde{a}_T^2)$. Also, for every $\eta \in (\delta, \alpha)$,

$$(S.3.1) \quad \mathbb{E} \left| \sum_{j=0}^{k-1} |\xi_{0,j}^>| \right|^{\eta} \leq k T \mathbb{E}(|\varepsilon_1 \varepsilon_2|^{\eta} \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T}) (\sum_{u=0}^{\infty} |\gamma_u|^{\eta})^2 = O(k \tilde{a}_T^{\eta})$$

by KT with $\mathbb{E}(|\varepsilon_1 \varepsilon_2|^{\eta}) < \infty$, so $\sum_{j=0}^{k-1} |\xi_{0,j}^>| = O_P(k^{1/\eta} \tilde{a}_T)$ by Markov's inequality, and by letting $\eta \uparrow \alpha$, $\sum_{j=0}^{k-1} |\xi_{0,j}^>| = o_P(k^{\epsilon} a_k \tilde{a}_T)$ for every $\epsilon > 0$. Similarly, since $|\xi_{0,|i-j|}^> - \xi_{ij}^>|$ does not exceed

$$\left(\sum_{t=k-i \wedge j}^{k-1} + \sum_{t=T-i \wedge j}^{T-1} \right) \sum_{u,v=0}^{\infty} \mathbb{I}_{\{u \neq v + |j-i|\}} |\gamma_u| |\gamma_v| |\varepsilon_{t-u} \varepsilon_{t-|j-i|-v}| \mathbb{I}_{|\varepsilon_{t-u} \varepsilon_{t-|j-i|-v}| > \tilde{a}_T},$$

with $i \wedge j := \min(i, j)$, it follows that

$$(S.3.2) \quad \mathbb{E} \left| \sum_{i,j=0}^{k-1} |\xi_{0,|i-j|}^> - \xi_{ij}^>| \right|^{\eta} \leq (k+1)^3 \mathbb{E}(|\varepsilon_1 \varepsilon_2|^{\eta} \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T}) (\sum_{u=0}^{\infty} |\gamma_u|^{\eta})^2$$

is $O((k^3/T) \tilde{a}_T^{\eta}) = o(k \tilde{a}_T^{\eta})$, so $\sum_{i,j=0}^{k-1} |\xi_{0,|i-j|}^> - \xi_{ij}^>| = o_P(k^{\epsilon} a_k \tilde{a}_T)$ for every $\epsilon > 0$. By combining these results, also $\|S_{00}^k - C^k\|_2 = o_P(k^{\epsilon} a_k l_T \tilde{a}_T)$.

Instead, for $\mathbb{E} |\varepsilon_1| < \infty$ (so $\alpha \in [1, 2)$), we write $\|(\xi_{ij})_{i,j}^{\leq}\|_2 \leq \sqrt{2}(\Xi_3^\leq + \Xi_4^\leq)^{1/2}$ with

$$\begin{aligned} \Xi_3^\leq &:= \sum_{i,j} \left\{ \sum_{t=k}^{T-1} \sum_{u,v=0}^{\infty} \mathbb{I}_{\{u \neq v + j-i\}} \gamma_u \gamma_v (\varepsilon_{t-i-u} \varepsilon_{t-j-v} \mathbb{I}_{|\varepsilon_{t-i-u} \varepsilon_{t-j-v}| \leq \tilde{a}_T} - \mu_T) \right\}^2, \\ \Xi_4^\leq &:= \mu_T^2 \sum_{i,j} \left\{ \sum_{t=k}^{T-1} \sum_{u,v=0}^{\infty} \mathbb{I}_{\{u \neq v + j-i\}} \gamma_u \gamma_v \right\}^2 \end{aligned}$$

and $\mu_T := \mathbb{E}(\varepsilon_1 \varepsilon_2 \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T})$, so

$$\|S_{00}^k - C^k\|_2 \leq \sqrt{2}(\Xi_3^\leq + \Xi_4^\leq)^{1/2} + 2 \sum_{j=0}^{k-1} |\xi_{0,j}^>| + \sum_{i,j=0}^{k-1} |\xi_{0,|i-j|}^> - \xi_{ij}^>|.$$

The terms in the upper bound satisfy: (i) $\Xi_3^{\leq} = O_P(k^2 \tilde{a}_T^2)$, as

$$\mathbb{E} |\Xi_3^{\leq}| \leq 4k^2 T \mathbb{E}(\varepsilon_1 \varepsilon_2 \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T} - \mu_T)^2 (\sum |\gamma_u|)^4 \leq O(k^2 T) \mathbb{E}(|\varepsilon_1 \varepsilon_2|^2 \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T})$$

is $O(k^2 \tilde{a}_T^2)$ by independence and KT; (ii) $|\Xi_4^{\leq}| \leq \mu_T^2 k^2 T^2 (\sum |\gamma_u|)^4 = O(k^2 \tilde{a}_T^2)$ for $\alpha \in (1, 2)$ since

$$\begin{aligned} (\text{S.3.3}) \quad |\mu_T| &= \left| \mathbb{E}(\varepsilon_1 \varepsilon_2 \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T}) \right| = \left| -\mathbb{E}(\varepsilon_1 \varepsilon_2 \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T}) \right| \\ &\leq \mathbb{E}(|\varepsilon_1 \varepsilon_2| \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T}) = O(T^{-1} \tilde{a}_T) \end{aligned}$$

using $\mathbb{E}(\varepsilon_1 \varepsilon_2) = 0$ and KT, whereas $|\Xi_4^{\leq}| = O_P(k^2 T^2) = O_P(k^2 l_T^2 a_T^2)$ for $\alpha = 1$ as $\mu_T = O(1)$; (iii) $2 \sum_{j=0}^{k-1} |\xi_{0,j}^>| + \sum_{i,j=0}^{k-1} |\xi_{0,i-j}^> - \xi_{ij}^>| = O_P(k \tilde{a}_T)$ for $\alpha \in (1, 2)$ by (S.3.1)-(S.3.2) with $\eta = 1$, and $2 \sum_{j=0}^{k-1} |\xi_{0,j}^>| + \sum_{i,j=0}^{k-1} |\xi_{0,i-j}^> - \xi_{ij}^>| = O_P(kT) = O_P(k l_T \tilde{a}_T)$ for $\alpha = 1$ using the same displays. Thus, $\|S_{00}^k - C^k\|_2 = O_P(k l_T \tilde{a}_T)$ in the case $\mathbb{E}|\varepsilon_1| < \infty$, and by the earlier argument, $\|S_{00}^k - C^k\|_2 = O_P(l_T \tilde{a}_T) \max\{k^\epsilon a_k, k\}$ for all $\alpha \in (0, 2)$ and $\epsilon > 0$.

In its turn,

$$(\text{S.3.4}) \quad c_{ij}^k = r_{|i-j|} \sum_{t=k}^{T-1} \varepsilon_t^2 + \sum_{v=0}^{\infty} \rho_v^{ij} [\varepsilon_{k-\max(i,j)-v}^2 - \varepsilon_{T-\max(i,j)-1-v}^2],$$

where $\rho_v^{ij} := \sum_{u=v+1}^{\infty} \gamma_u \gamma_{u+|j-i|}$ has $|\rho_v^{ij}| \leq \sum_{u=v+1}^{\infty} \gamma_u^2 := \tilde{\gamma}_v^2$. For δ of Assumption 1(b) it follows that $\sum_{v=0}^{\infty} |\rho_v^{ij}|^{\delta/2} \leq \sum_{v=0}^{\infty} \sum_{u=v+1}^{\infty} |\gamma_u|^\delta = \sum_{u=1}^{\infty} u |\gamma_u|^\delta < \infty$, so the series in (S.3.4) are a.s. convergent because ε_t^2 has tail index $\alpha/2$.

Further, as $r_{|i-j|}^2 \leq r_0^2$,

$$\begin{aligned} (\text{S.3.5}) \quad &\|\Sigma_k \sigma_T^2 - C^k\|_2^2 \leq \|\Sigma_k \sigma_T^2 - C^k\|^2 \\ &\leq 3 \left(\sum_{t=1}^k \varepsilon_t^2 + \sum_{t=T-k}^T \varepsilon_t^2 \right)^2 \sum_{i,j} r_{ij}^2 + 3 \sum_{i,j} \left(\sum_{v=0}^{\infty} \tilde{\gamma}_v^2 \varepsilon_{T-\max(i,j)-1-v}^2 \right)^2 \\ &\quad + 3 \sum_{i,j} \left(\sum_{v=0}^{\infty} \tilde{\gamma}_v^2 \varepsilon_{k-\max(i,j)-v}^2 \right)^2 \leq 3k^2 [O_P(a_k^4) r_{00}^2 \\ &\quad + \max_{i=1,\dots,k} \left(\sum_{v=0}^{\infty} \tilde{\gamma}_v^2 \varepsilon_{T-i-v}^2 \right)^2 + \max_{i=1,\dots,k} \left(\sum_{v=0}^{\infty} \tilde{\gamma}_v^2 \varepsilon_{k-i-v}^2 \right)^2] = O_P(k^2 a_k^4) \end{aligned}$$

using that $\max_{i=1,\dots,k} |a_k^{-2} \sum_{v=0}^{\infty} \tilde{\gamma}_v^2 \varepsilon_{T-i-v}^2|$, $\max_{i=1,\dots,k} |a_k^{-2} \sum_{v=0}^{\infty} \tilde{\gamma}_v^2 \varepsilon_{k-i-v}^2|$, $a_k^{-2} \sum_{t=1}^k \varepsilon_t^2$ and $a_k^{-2} \sum_{t=T-k}^T \varepsilon_t^2$ converge weakly to a.s. finite r.v.'s (see Theorem 3.2 of Davis and Resnick (1985a) for the former two, as ε_t^2 are in the

$\alpha/2$ -stable domain of attraction with normalisation a_T^2). From the triangle inequality and the condition $k^2/T \rightarrow 0$, we conclude that, for every $\epsilon > 0$,

$$(S.3.6) \quad \|S_{00}^k - \Sigma_k \sigma_T^2\|_2 = O_P(l_T \tilde{a}_T) \max\{k^\epsilon a_k, k\}.$$

Regarding $S_{0\varepsilon}^k$ in part (b), first,

$$\left\| \sum_{t=k+1}^T \mathbf{X}_{t-1}^k \varepsilon_t \right\|^2 = \sum_{i=1}^k \left(\sum_{t=k+1}^T \sum_{j=0}^{\infty} \gamma_j \varepsilon_{t-j-i} \varepsilon_t \right)^2.$$

For $\{\varepsilon_t\}$ with $E|\varepsilon_1| = \infty$ (and hence, $\alpha \leq 1$), we write $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \varepsilon_t\|^2 \leq 2\{(\Lambda_1^>)^2 + \Lambda_1 + \Lambda_2\}$ with

$$\begin{aligned} \Lambda_1^> &:= \sum_{i=1}^k \sum_{t=k+1}^T \sum_{j=0}^{\infty} |\gamma_j| |\varepsilon_{t-j-i} \varepsilon_t| \mathbb{I}_{|\varepsilon_{t-j-i} \varepsilon_t| > \tilde{a}_T} \\ \Lambda_1 &:= \sum_{i=1}^k \sum_{s,t=k+1}^T \sum_{h,j=0}^{\infty} \mathbb{I}_B \gamma_h \gamma_j \varepsilon_{s-h-i} \varepsilon_s \varepsilon_{t-j-i} \varepsilon_t \mathbb{I}_{|\varepsilon_{s-h-i} \varepsilon_s| \leq \tilde{a}_T} \mathbb{I}_{|\varepsilon_{t-j-i} \varepsilon_t| \leq \tilde{a}_T}, \\ \Lambda_2 &:= \sum_{i=1}^k \sum_{s,t=k+1}^T \sum_{h,j=0}^{\infty} \mathbb{I}_{B^c} \gamma_h \gamma_j \varepsilon_{s-h-i} \varepsilon_s \varepsilon_{t-j-i} \varepsilon_t \mathbb{I}_{|\varepsilon_{s-h-i} \varepsilon_s| \leq \tilde{a}_T} \mathbb{I}_{|\varepsilon_{t-j-i} \varepsilon_t| \leq \tilde{a}_T}, \end{aligned}$$

$B := \{\#(\{s, t, s-h-i, t-j-i\}) = 4\}$, $B^c := \{\#(\{s, t, s-h-i, t-j-i\}) = 2$ or 3}. Similarly to the evaluations of $(\Xi^>)^2, \Xi_{1,2}$, we find for $\eta \in (\delta, \alpha)$ that $E|\Lambda_1^>|^\eta = O(k\tilde{a}_T^\eta)$, $E|\Lambda_1| = O(kl_T^2 \tilde{a}_T^2)$ and $E|\Lambda_2| = O(k\tilde{a}_T^2)$, so $\Lambda_1^> = o_P(k^\epsilon a_k \tilde{a}_T)$ for every $\epsilon > 0$, $\Lambda_1 = O_P(kl_T^2 \tilde{a}_T^2)$ for l_T as in part (a), and $\Lambda_2 = O_P(k\tilde{a}_T^2)$, giving $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \varepsilon_t\| = o_P(k^\epsilon a_k l_T \tilde{a}_T)$ for every $\epsilon > 0$. On the other hand, in the case $E|\varepsilon_1| < \infty$ (where $\alpha \geq 1$), $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \varepsilon_t\|^2 \leq 3(\Lambda_2^> + \Lambda_3 + \Lambda_4)$ with

$$\Lambda_2^> := \sum_{i=1}^k \left(\sum_{t=k+1}^T \sum_{j=0}^{\infty} |\gamma_j| |\varepsilon_{t-j-i} \varepsilon_t| \mathbb{I}_{|\varepsilon_{t-j-i} \varepsilon_t| > \tilde{a}_T} \right)^2$$

satisfying by KT $E|\Lambda_2^>|^{\eta/2} \leq kT E(|\varepsilon_1 \varepsilon_2|^\eta \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T})(\sum |\gamma_u|^\eta) = O(k\tilde{a}_T^\eta)$ for $\eta \in [1, \alpha]$, $\alpha > 1$, whereas $E|\Lambda_2^>|^{1/2} = O(kT) = O(kl_T \tilde{a}_T)$ for $\alpha = 1$, so $\Lambda_2^> = O_P(k^\epsilon a_k^2 l_T^2 \tilde{a}_T^2)$ by Markov's inequality, and

$$\begin{aligned} \Lambda_3 &:= \sum_{i=1}^k \left\{ \sum_{t=k+1}^T \sum_{j=0}^{\infty} \gamma_j (\varepsilon_{t-j-i} \varepsilon_t \mathbb{I}_{|\varepsilon_{t-j-i} \varepsilon_t| \leq \tilde{a}_T} - \mu_T) \right\}^2 = O_P(k\tilde{a}_T^2), \\ \Lambda_4 &:= \mu_T^2 \sum_{i=1}^k \left(\sum_{t=k+1}^T \sum_{j=0}^{\infty} \gamma_j \right)^2 = O(kl_T^2 \tilde{a}_T^2) \end{aligned}$$

as $\Xi_{3,4}$ earlier. Thus, $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \varepsilon_t\| = O_P(k^\epsilon a_k l_T \tilde{a}_T)$ for every $\epsilon > 0$ in the case $E|\varepsilon_1| < \infty$, and by the previous argument, for all $\alpha \in (0, 2)$.

Second,

$$\begin{aligned} \left\| \sum_{t=k+1}^T \mathbf{X}_{t-1}^k \rho_{t,k} \right\| &= \left\{ \sum_{i=1}^k \left(\sum_{j=k+1}^{\infty} \beta_j \sum_{t=k+1}^T X_{t-i} X_{t-j} \right)^2 \right\}^{1/2} \\ &\leq \sqrt{2} \sum_{i=1}^k \sum_{j=k+1}^{\infty} |\beta_j| |c_{ij}^k| + \sqrt{2} \left\{ \sum_{i=1}^k \left(\sum_{j=k+1}^{\infty} \beta_j \xi_{ij} \right)^2 \right\}^{1/2} \end{aligned}$$

where $c_{ij}^k := \sum_{t=k}^{T-1} \sum_{v=0}^{\infty} \varepsilon_{t-j-v}^2 \gamma_v \gamma_{v+j-i}$ is

$$c_{ij}^k = r_{j-i} \sum_{t=k}^{T-1} \varepsilon_{t-j}^2 - \sum_{v=0}^{\infty} \rho_v^{ij} \varepsilon_{T-j-v-1}^2 + \sum_{v=0}^{\infty} \rho_v^{ij} \varepsilon_{k-j-v}^2$$

and $\xi_{ij} := \sum_{t=k}^{T-1} \sum_{u=0}^{\infty} \sum_{v=0}^{\infty} \mathbb{I}_{\{u \neq v+j-i\}} \gamma_u \gamma_v \varepsilon_{t-i-u} \varepsilon_{t-j-v}$; cf. (S.3.4) with $i < j$. We find that (i)

$$\sum_{i=1}^k \sum_{j=k+1}^{\infty} |\beta_j| |r_{j-i}| \sum_{t=k}^{T-1} \varepsilon_{t-j}^2 \leq \left\{ \sum_{j=k+1}^{\infty} |\beta_j| \right\} \left\{ \sum_{t=k}^{T-1} \sum_{j=k+1}^{\infty} \left(\sum_{i=1}^k |r_{j-i}| \right) \varepsilon_{t-j}^2 \right\},$$

where $\sum_{t=k}^{T-1} \sum_{j=k+1}^{\infty} (\sum_{i=1}^k |r_{j-i}|) \varepsilon_{t-j}^2$ is distributed like

$$\begin{aligned} \sum_{t=0}^{T-k-1} \sum_{j=1}^{\infty} \left(\sum_{i=j}^{j+k-1} |r_i| \right) \varepsilon_{t-j}^2 &\leq T \mathbb{E}[\varepsilon_1^2 \mathbb{I}_{|\varepsilon_1| \leq a_T}] \sum_{j=1}^{\infty} \sum_{i=j}^{\infty} |r_i| \\ &+ \sum_{t=0}^{T-1} \sum_{j=1}^{\infty} \left(\sum_{i=j}^{\infty} |r_i| \right) (\varepsilon_{t-j}^2 - \mathbb{E}[\varepsilon_{t-j}^2 \mathbb{I}_{|\varepsilon_{t-j}| \leq a_T}]) = O_P(a_T^2) \end{aligned}$$

since ε_t^2 have tail index $\alpha/2$ and

$$\sum_{j=1}^{\infty} \left(\sum_{i=j}^{\infty} |r_i| \right)^{\delta/2} \leq \sum_{s=0}^{\infty} |\gamma_s|^{\delta/2} \sum_{j=1}^{\infty} j |\gamma_{s+j}|^{\delta/2} \leq \sum_{s=0}^{\infty} |\gamma_s|^{\delta/2} \sum_{s=0}^{\infty} s |\gamma_s|^{\delta/2} < \infty$$

by Assumption 1(b), so Theorem 4.1 of Davis and Resnick (1985a) applies (with their $c_j = \sum_{i=j}^{\infty} |r_i|$) jointly with KT; (ii)

$$\sum_{i=1}^k \sum_{j=k+1}^{\infty} |\beta_j| \sum_{v=0}^{\infty} |\rho_v^{ij}| \varepsilon_{T-j-v-1}^2 \leq \left\{ \sum_{j=k+1}^{\infty} |\beta_j| \right\} \left\{ \sum_{i=1}^k \sum_{j=k+1}^{\infty} \sum_{v=0}^{\infty} |\rho_v^{ij}| \varepsilon_{T-j-v-1}^2 \right\}$$

with

$$\begin{aligned} \mathbb{E} \left(\sum_{i=1}^k \sum_{j=k+1}^{\infty} \sum_{v=0}^{\infty} |\rho_v^{ij}| \varepsilon_{T-j-v-1}^2 \mathbb{I}_{|\varepsilon_{T-j-v-1}| \leq a_T} \right) &\leq \mathbb{E}(\varepsilon_1^2 \mathbb{I}_{|\varepsilon_1| \leq a_T}) \sum_{i=1}^k \sum_{j=k+1}^{\infty} \sum_{v=0}^{\infty} |\rho_v^{ij}| \\ &\leq O(T^{-1} a_T^2) \sum_{i=1}^k \sum_{j=k+1}^{\infty} \sum_{v=0}^{\infty} \sum_{u=v+1}^{\infty} |\gamma_u| |\gamma_{u+|i-j|}| \leq O(T^{-1} a_T^2) \left(\sum_{u=1}^{\infty} u |\gamma_u| \right)^2 \end{aligned}$$

and similarly, for $\eta \in (\delta, \alpha)$,

$$\mathbb{E} \left(\sum_{i=1}^k \sum_{j=k+1}^{\infty} \sum_{v=0}^{\infty} |\rho_v^{ij}| \varepsilon_{T-j-v-1}^2 \mathbb{I}_{|\varepsilon_{T-j-v-1}| > a_T} \right)^{\eta/2} \leq \mathbb{E}(|\varepsilon_1|^\eta \mathbb{I}_{|\varepsilon_1| \leq a_T}) \left(\sum_{u=1}^{\infty} u |\gamma_u|^{\eta/2} \right)^2$$

is $O(T^{-1}a_T^\eta)$ by KT, so $\sum_{i=1}^k \sum_{j=k+1}^\infty \sum_{v=0}^\infty |\rho_v^{ij}| \varepsilon_{T-j-v-1}^2 = o_P(a_T^2)$; (iii),

$$\sum_{i=1}^k \sum_{j=k+1}^\infty |\beta_j| \sum_{v=0}^\infty |\rho_v^{ij}| \varepsilon_{k-j-v}^2 = o_P(a_T^2) \sum_{j=k+1}^\infty |\beta_j|$$

likewise. Thus, $\sum_{i=1}^k \sum_{j=k+1}^\infty |\beta_j| |c_{ij}^k| = O_P(a_T^2) \sum_{j=k+1}^\infty |\beta_j|$ by combining the previous estimates.

Further, we split $\xi_{ij} = \xi_{ij}^< + \xi_{ij}^>$ as in the proof of part (a):

$$\xi_{ij}^R := \sum_{t=k}^{T-1} \sum_{u,v=0}^\infty \mathbb{I}_{\{u \neq v+j-i\}} \gamma_u \gamma_v \varepsilon_{t-i-u} \varepsilon_{t-j-v} \mathbb{I}_{|\varepsilon_{t-i-u} \varepsilon_{t-j-v}| > \tilde{a}_T}, \quad R \in \{\leq, >\},$$

and for $\{\varepsilon_t\}$ with $E|\varepsilon_1| = \infty$ we find that,

$$E(\sum_{i=1}^k \sum_{j=k+1}^\infty |\beta_j| |\xi_{ij}^<|) \leq Tk E(|\varepsilon_1 \varepsilon_2| \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T}) (\sum_{u=0}^\infty |\gamma_u|)^2 \sum_{j=k+1}^\infty |\beta_j|$$

is $o(a_T^2) \sum_{j=k+1}^\infty |\beta_j|$ by KT, and similarly, for $\eta \in (\delta, \alpha)$,

$$E(\sum_{i=1}^k \sum_{j=k+1}^\infty |\beta_j| |\xi_{ij}^>|)^\eta \leq Tk E(|\varepsilon_1 \varepsilon_2|^\eta \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T}) (\sum_{u=0}^\infty |\gamma_u|^\eta)^2 \sum_{j=k+1}^\infty |\beta_j|^\eta$$

is $O(k \tilde{a}_T^\eta) \sum_{j=k+1}^\infty |\beta_j|^\eta$, so by using (S.2.1), $\{\sum_{i=1}^k (\sum_{j=k+1}^\infty \beta_j \xi_{ij})^2\}^{1/2} \leq \sum_{i=1}^k \sum_{j=k+1}^\infty |\beta_j| |\xi_{ij}| = o_P(a_T^{1-\zeta}) + O_P(a_T^2) \sum_{j=k+1}^\infty |\beta_j|$ for $\zeta > 0$ sufficiently small, and $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \rho_{t,k}\| = o_P(a_T^{1-\zeta}) + O_P(a_T^2) \sum_{j=k+1}^\infty |\beta_j|$ in this case. In the case $E|\varepsilon_1| < \infty$, as in the proof of part (a), we find that, (i),

$$\begin{aligned} (S.3.7) \quad & E(\sum_{i=1}^k (\sum_{j=k+1}^\infty \beta_j \xi_{ij}^<)^2) \leq [\sum_{i=1}^k \sum_{j=k+1}^\infty |\beta_j| E\{(\xi_{ij}^<)^2\}] (\sum_{j=k+1}^\infty |\beta_j|) \\ & \leq k [4T E(\varepsilon_1^2 \varepsilon_2^2 \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T}) + T^2 \mu_T^2] (\sum_{u=0}^\infty |\gamma_u|)^4 (\sum_{j=k+1}^\infty |\beta_j|)^2 \end{aligned}$$

is $O(k l_T^2 \tilde{a}_T^2) (\sum_{j=k+1}^\infty |\beta_j|)^2$, the inequalities respectively from Cauchy-Schwartz and by separating products where some ε is squared from those where all ε 's are distinct, and the magnitude order from KT and (S.3.3), and (ii),

$$E(\sum_{i=1}^k \sum_{j=k+1}^\infty |\beta_j| |\xi_{ij}^>|) \leq kT E(|\varepsilon_1 \varepsilon_2| \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T}) (\sum_{u=0}^\infty |\gamma_u|)^2 \sum_{j=k+1}^\infty |\beta_j|$$

is $O(kl_T \tilde{a}_T) \sum_{j=k+1}^{\infty} |\beta_j|$ by KT. Thus,

$$\left\{ \sum_{i=1}^k \left(\sum_{j=k+1}^{\infty} \beta_j \xi_{ij} \right)^2 \right\}^{1/2} \leq \sqrt{2} \left\{ \sum_{i=1}^k \left(\sum_{j=k+1}^{\infty} \beta_j \xi_{ij}^- \right)^2 \right\}^{1/2} + \sqrt{2} \sum_{i=1}^k \sum_{j=k+1}^{\infty} |\beta_j| |\xi_{ij}^+|$$

is $O_P(kl_T \tilde{a}_T) \sum_{j=k+1}^{\infty} |\beta_j|$ with $kl_T \tilde{a}_T = o(a_T^2)$ when $k^2/T \rightarrow 0$. Finally, $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \rho_{t,k}\| = O_P(a_T^2) \sum_{j=k+1}^{\infty} |\beta_j|$ when $E|\varepsilon_1| < \infty$.

The magnitude order of $S_{0\varepsilon}^k$ is obtained by combining the magnitude orders of $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \varepsilon_t\|$ and $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \rho_{t,k}\|$. \square

S.3.2. Proof of Corollary 4. Using the fact that $\hat{\Gamma}_k^{-1}$ is the lower triangular Toeplitz matrix with first column $(1 : -\hat{\beta}'_{k-1})'$, it can be checked directly that $\hat{\Gamma}_k^{-1} \gamma_k = (I_k - \Gamma_k) \hat{\beta}_k + \gamma_k$, from where

$$\begin{aligned} \hat{\gamma}_k - \gamma_k &= \hat{\Gamma}_k \hat{\beta}_k - \gamma_k = \hat{\Gamma}_k (\hat{\beta}_k - \hat{\Gamma}_k^{-1} \gamma_k) = \hat{\Gamma}_k (\Gamma_k \hat{\beta}_k - \gamma_k) \\ &= \hat{\Gamma}_k (\Gamma_k \hat{\beta}_k - \Gamma_k \beta_k) = \hat{\Gamma}_k \Gamma_k (\hat{\beta}_k - \beta_k). \end{aligned}$$

Hence, $\|\hat{\gamma}_k - \gamma_k\|_1 \leq \|\hat{\Gamma}_k\|_1 \|\Gamma_k\|_1 \|\hat{\beta}_k - \beta_k\|_1$, where $\|\cdot\|_1$ equals the maximum absolute column sum. As $\|\Gamma_k\|_1 \leq \|\gamma\|_1 := \sum_{i=0}^{\infty} |\gamma_i| < \infty$, and thus,

$$\|\hat{\Gamma}_k\|_1 \leq \|\Gamma_k\|_1 + \|\hat{\Gamma}_k - \Gamma_k\|_1 \leq \|\gamma\|_1 + \|\hat{\gamma}_k - \gamma_k\|_1,$$

it holds further that $\|\hat{\gamma}_k - \gamma_k\|_1 \leq (\|\gamma\|_1 + \|\hat{\gamma}_k - \gamma_k\|_1) \|\gamma\|_1 \|\hat{\beta}_k - \beta_k\|_1$ and, for small $\|\hat{\beta}_k - \beta_k\|_1$, $\|\hat{\gamma}_k - \gamma_k\|_1 \leq \|\hat{\beta}_k - \beta_k\|_1 \|\gamma\|_1^2 / (1 - \|\gamma\|_1 \|\hat{\beta}_k - \beta_k\|_1)$. Hence, $\|\hat{\gamma}_k - \gamma_k\|_1 = O_P(\|\hat{\beta}_k - \beta_k\|_1) = O_P(k^{1/2} \|\hat{\beta}_k - \beta_k\|) = o_P(1)$ by (7.1) with $k^2/T \rightarrow 0$ and $k^{1/2} \sum_{j=k+1}^{\infty} |\beta_j| \leq \sum_{j=k+1}^{\infty} j |\beta_j| = o(1)$.

Similarly, $\|\hat{\gamma}_k - \gamma_k\| \leq \|\hat{\Gamma}_k\|_2 \|\Gamma_k\|_2 \|\hat{\beta}_k - \beta_k\|$, with

$$\begin{aligned} \|\hat{\Gamma}_k\|_2 &\leq \|\Gamma_k\|_2 + \|\hat{\Gamma}_k - \Gamma_k\|_2 \leq \|\Gamma_k\|_2 + \|\hat{\Gamma}_k - \Gamma_k\|_1^{1/2} \|\hat{\Gamma}_k - \Gamma_k\|_{\infty}^{1/2} \\ &\leq \|\Gamma_k\|_2 + \|\hat{\gamma}_k - \gamma_k\|_1, \end{aligned}$$

so $\|\hat{\gamma}_k - \gamma_k\| \leq (\|\Gamma_k\|_2 + \|\hat{\gamma}_k - \gamma_k\|_1) \|\Gamma_k\|_2 \|\hat{\beta}_k - \beta_k\|$. Since $\|\hat{\gamma}_k - \gamma_k\|_1 = o_P(1)$ and $\|\Gamma_k\|_2 \leq \|\Gamma_k\|_1^{1/2} \|\Gamma_k\|_{\infty}^{1/2} \leq \|\gamma\|_1 < \infty$, it follows that $\|\hat{\gamma}_k - \gamma_k\| = O_P(\|\hat{\beta}_k - \beta_k\|) = o_P(1)$. \square

S.3.3. Proof of Corollary 5. It holds that $|\hat{C}_T(\lambda) - C(\lambda)| \leq R_T(\lambda) + I_T(\lambda)$ with

$$\begin{aligned} R_T(\lambda) &:= \left| \left| 1 + \sum_{j=1}^k \hat{\gamma}_j \cos(\lambda j) \right|^2 - \left| 1 + \sum_{j=1}^{\infty} \gamma_j \cos(\lambda j) \right|^2 \right| \\ &\leq \left(2 + \sum_{j=1}^{\infty} |\gamma_j| + \sum_{j=1}^k |\hat{\gamma}_j| \right) \left(\sum_{j=1}^k |\hat{\gamma}_j - \gamma_j| + \sum_{j=k+1}^{\infty} |\gamma_j| \right) \end{aligned}$$

$$\leq \left(2 + 2 \sum_{j=1}^{\infty} |\gamma_j| + \|\hat{\gamma}_k - \gamma_k\|_1 \right) \left(\|\hat{\gamma}_k - \gamma_k\|_1 + \sum_{j=k+1}^{\infty} |\gamma_j| \right) = o_P(1)$$

since $\|\hat{\gamma}_k - \gamma_k\|_1 = o_P(1)$ by the proof of Corollary 4 and $\sum_{j=1}^{\infty} |\gamma_j| < \infty$. Similarly, $I_T(\lambda) := \left| |\sum_{j=1}^{\infty} \gamma_j \sin(\lambda j)|^2 - |\sum_{j=1}^k \hat{\gamma}_j \sin(\lambda j)|^2 \right| = o_P(1)$ using the same upper bounds. As these bounds are independent of λ , convergence is uniform in λ . \square

S.4. Proofs from Section 7.3.1 of CGT. Similarly to Lemma 8.3 of Kreiss (1997), the following bounds can be established for $\hat{\gamma}_j$.

LEMMA S.2. *There exist constants $b_{jk} \geq 0$ and C such that, for large k and uniformly in $j \in \mathbb{N}$, it holds that*

$$(S.4.1) \quad |\hat{\gamma}_j - \gamma_j| \leq \left(1 + \frac{1}{k} \right)^{-j} O_P(\|\hat{\beta}_k - \beta_k\|_1 + \sum_{j=k+1}^{\infty} |\beta_j|) + b_{jk}$$

and $\sum_{j=0}^{\infty} b_{jk} \leq C \sum_{j=k+1}^{\infty} |\beta_j|$.

S.4.1. Proof of Lemma 3.

S.4.1.1. *Preparation.* From $\hat{\varepsilon}_t - \varepsilon_t = (\beta_k - \hat{\beta}_k)' \mathbf{X}_{t-1}^k + \rho_{t,k}$ it follows that

$$(S.4.2) \quad \sum_{t=k+1}^T (\hat{\varepsilon}_t - \varepsilon_t) = \sum_{t=k+1}^T (\mathbf{X}_{t-1}^k)' (\beta_k - \hat{\beta}_k) + \sum_{t=k+1}^T \rho_{t,k},$$

where, (i), $\sum_{t=k+1}^T \mathbf{X}_{t-1}^k = O_P(k^{1/2} a_T)$ is implied by the Beveridge-Nelson decomposition of X_t ($X_t = \gamma(1) \varepsilon_t - \Delta Z_t$, $Z_t := \sum_{j=0}^{\infty} \varepsilon_{t-j} \sum_{i=j+1}^{\infty} \gamma_i$), which yields

$$\begin{aligned} \left\| \sum_{t=k+1}^T \mathbf{X}_{t-1}^k \right\| &\leq k^{1/2} |\gamma(1)| \left\| \sum_{t=k}^{T-1} \varepsilon_t \right\| + k \max_{t \in \{0, \dots, k-1\} \cup \{T-k, \dots, T-1\}} (|\varepsilon_t| + |Z_t|) \\ &= O_P(k^{1/2} a_T l_T + k a_k) = O_P(k^{1/2} a_T l_T) \end{aligned}$$

as $a_T^{-1} \sum_{t=k}^{T-1} \varepsilon_t - a_T^{-1} T \mathbb{E}(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| \leq a_T}) = O_P(1)$, $a_T^{-1} T \mathbb{E}(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| \leq a_T}) = l_T$, and (ii), $\sum_{t=k+1}^T \rho_{t,k} = o_P(l_T)$ by Markov's inequality. Indeed, for $\mathbb{E}|\varepsilon_1| < \infty$,

$$\begin{aligned} \left(\sum_{t=k+1}^T \rho_{t,k} \right)^2 &\leq 2 \left(\sum_{t=k+1}^T \sum_{i=k+1}^{\infty} \beta_i \sum_{j=0}^{\infty} \gamma_j \varepsilon_{t-i-j} \mathbb{I}_{|\varepsilon_{t-i-j}| \leq a_T} \right)^2 \\ &\quad + 2 \left(\sum_{t=k+1}^T \sum_{i=k+1}^{\infty} \beta_i \sum_{j=0}^{\infty} \gamma_j \varepsilon_{t-i-j} \mathbb{I}_{|\varepsilon_{t-i-j}| > a_T} \right)^2 := 2R_k^{\leq} + 2R_k^{>} \end{aligned}$$

with

$$\begin{aligned} \mathbb{E} R_k^{\leq} &\leq \mathbb{E}(\varepsilon_1^2 \mathbb{I}_{|\varepsilon_1| \leq a_T}) \sum_{t=k+1}^T \sum_{i,m=k+1}^{\infty} |\beta_i| |\beta_m| \sum_{j,n=0}^{\infty} |\gamma_j| |\gamma_n| \\ &\quad + \{\mathbb{E}(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| \leq a_T})\}^2 \sum_{t,s=k+1}^T \sum_{i,m=k+1}^{\infty} |\beta_i| |\beta_m| \sum_{j,n=0}^{\infty} |\gamma_j| |\gamma_n| \\ &= O(1) \|\gamma\|_1^2 (a_T \sum_{i=k+1}^{\infty} |\beta_i|)^2 = o(1) \end{aligned}$$

by using $\mathbb{E}(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| \leq a_T}) = -\mathbb{E}(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| > a_T})$ and KT ($\|\gamma\|_1 := \sum_{i=0}^{\infty} |\gamma_i|$), and

$$\mathbb{E}(R_k^>)^{1/2} \leq \mathbb{E}(|\varepsilon_1| \mathbb{I}_{|\varepsilon_1| > a_T}) T \sum_{i=k+1}^{\infty} |\beta_i| \sum_{j=0}^{\infty} |\gamma_j| = O(l_T)(a_T \sum_{i=k+1}^{\infty} |\beta_i|) = o(l_T),$$

so $\sum_{t=k+1}^T \rho_{t,k} = o_P(l_T)$ in this case, whereas for $\mathbb{E}|\varepsilon_1| = \infty$ it holds that

$$\begin{aligned} \sum_{t=k+1}^T |\rho_{t,k}| &\leq \sum_{t=k+1}^T \sum_{i=k+1}^{\infty} |\beta_i| |X_{t-i}| \mathbb{I}_{|X_{t-i}| \leq a_T} + \sum_{t=k+1}^T \sum_{i=k+1}^{\infty} |\beta_i| |X_{t-i}| \mathbb{I}_{|X_{t-i}| > a_T} \\ &=: C_k^{\leq} + C_k^> \end{aligned}$$

with

$$\begin{aligned} \mathbb{E} C_k^{\leq} &\leq T \mathbb{E}(|X_1| \mathbb{I}_{|X_1| \leq a_T}) \sum_{i=k+1}^{\infty} |\beta_i| = O(l_T)(a_T \sum_{i=k+1}^{\infty} |\beta_i|) = o(l_T), \\ \mathbb{E}(C_k^>)^{\eta} &\leq \mathbb{E}\left(\sum_{i=k+1}^{2T-k-1} |X_{T-i}| \mathbb{I}_{|X_{T-i}| > a_T} \sum_{j=k+1}^{\infty} |\beta_i| + \sum_{i=T-k}^{\infty} |X_{-i}| \mathbb{I}_{|X_{-i}| > a_T} \sum_{j=k+1+i}^{T+i} |\beta_j|\right)^{\eta} \\ &\leq T \mathbb{E}(|X_1|^{\eta} \mathbb{I}_{|X_1| > a_T}) \{2\left(\sum_{j=k+1}^{\infty} |\beta_j|\right)^{\eta} + T^{-1} \sum_{i=T-k}^{\infty} \sum_{j=k+1+i}^{T+i} |\beta_j|^{\eta}\} \\ &\leq O(a_T^{\eta}) \{2\left(\sum_{j=k+1}^{\infty} |\beta_j|\right)^{\eta} + \sum_{j=T+1}^{\infty} |\beta_j|^{\eta}\} \end{aligned}$$

by KT for $\eta \in [\delta, \alpha]$, $\alpha \leq 1$, so from $a_T \sum_{j=k+1}^{\infty} |\beta_j| \rightarrow 0$ and $\sum_{j=T+1}^{\infty} |\beta_j|^{\eta} = O(T^{1-2\eta/\delta})$ (under $\sum_{j=1}^{\infty} j^{2/\delta} |\beta_j| < \infty$), it follows that $C_k^> = o_P(1) + O_P(a_T T^{1/\eta - 2/\delta}) = o_P(1)$ as $\eta \in [\delta, \alpha]$ can be chosen such that $1/\alpha + 1/\eta < 2/\delta$; eventually $\sum_{t=k+1}^T |\rho_{t,k}| = o_P(l_T)$ for $\mathbb{E}|\varepsilon_1| = \infty$. Returning to (S.4.2) and using (7.1), it follows that for all $\epsilon > 0$,

$$(S.4.3) \quad \sum_{t=k+1}^T (\hat{\varepsilon}_t - \varepsilon_t) = O_P(k^{1/2} a_k a_T^{\epsilon} + l_T) = O_P(k^{1/2} a_k a_T^{\epsilon}).$$

As $(\hat{\varepsilon}_t - \varepsilon_{t,k})^2 = \{(\hat{\beta}_k - \beta_k)' \mathbf{X}_{t-1}^k\}^2$ and π is a.s. bijective, we find

$$\begin{aligned} \sum_{t=k+1}^T (\hat{\varepsilon}_t - \varepsilon_{t,k})^2 &= (\hat{\beta}_k - \beta_k)' S_{00}^k (\hat{\beta}_k - \beta_k) \leq \|\hat{\beta}_k - \beta_k\|^2 \|S_{00}^k\|_2 \\ &\leq \|\hat{\beta}_k - \beta_k\|^2 (\sigma_T^2 \|\Sigma_k\|_2 + o_P(a_T^2)) = O_P(a_T^2 \|\hat{\beta}_k - \beta_k\|^2) \end{aligned}$$

by Lemma 2(a), and since $\sigma_T^2 = O_P(a_T^2)$ and $\|\Sigma_k\|_2 = O(1)$. Next, from $(\hat{\varepsilon}_t - \varepsilon_t)^2 \leq 2(\hat{\varepsilon}_t - \varepsilon_{t,k})^2 + 2\rho_{t,k}^2$ and the a.s. bijectivity of π , it follows that P^π -a.s. (i.e., conditional on the data and $\{w_t\}_{t=k+1}^T$, with randomness stemming from π alone),

$$\begin{aligned} (S.4.4) \quad \sum_{t=k+1}^T (\hat{\varepsilon}_{\pi(t)} - \varepsilon_{\pi(t)})^2 &= \sum_{t=k+1}^T (\hat{\varepsilon}_t - \varepsilon_t)^2 =: \|\hat{\varepsilon}_T - \varepsilon_T\|^2 \\ &\leq 2 \left\{ \sum_{t=k+1}^T (\hat{\varepsilon}_t - \varepsilon_{t,k})^2 + S_{\rho\rho}^k \right\} = O_P(a_T^2 \|\hat{\beta}_k - \beta_k\|^2) + 2S_{\rho\rho}^k, \end{aligned}$$

where $S_{\rho\rho}^k := \sum_{t=k+1}^T \rho_{t,k}^2 = o_P(l_T)$ by Markov's inequality. In fact, for $E|\varepsilon_1| < \infty$ it holds that

$$\begin{aligned} S_{\rho\rho}^k &\leq 2 \sum_{t=k+1}^T \left(\sum_{i=k+1}^{\infty} \beta_i \sum_{j=0}^{\infty} \gamma_j \varepsilon_{t-i-j} \mathbb{I}_{|\varepsilon_{t-i-j}| \leq a_T} \right)^2 \\ &\quad + 2 \sum_{t=k+1}^T \left(\sum_{i=k+1}^{\infty} \beta_i \sum_{j=0}^{\infty} \gamma_j \varepsilon_{t-i-j} \mathbb{I}_{|\varepsilon_{t-i-j}| > a_T} \right)^2 =: 2S_{\rho\rho}^{\leq} + 2S_{\rho\rho}^{>} \end{aligned}$$

with (i)

$$\begin{aligned} E S_{\rho\rho}^{\leq} &= \sum_{t=k+1}^T E \left(\sum_{i=k+1}^{\infty} \beta_i \sum_{j=0}^{\infty} \gamma_j \varepsilon_{t-i-j} \mathbb{I}_{|\varepsilon_{t-i-j}| \leq a_T} \right)^2 \\ &= E(\varepsilon_1^2 \mathbb{I}_{|\varepsilon_1| \leq a_T}) \sum_{t=k+1}^T \sum_{i=k+1}^{\infty} \left(\sum_{j=k+1}^i \beta_j \gamma_{i-j} \right)^2 \\ &\quad + \{E(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| \leq a_T})\}^2 \sum_{t=k+1}^T \sum_{i,m=k+1}^{\infty} \sum_{j,n=0}^{\infty} \mathbb{I}_{\{i+j \neq m+n\}} \beta_j \beta_m \gamma_j \gamma_n \\ &\leq O(a_T^2) \left(\sum_{i=k+1}^{\infty} \sum_{j=k+1}^i |\beta_j| |\gamma_{i-j}| \right)^2 + O(T^{-1} l_T a_T^2) \left(\sum_{j=k+1}^{\infty} |\beta_j| \right)^2 \|\gamma\|_1^2 \\ &\leq O(1) (a_T^2 \sum_{j=k+1}^{\infty} |\beta_j|)^2 \|\gamma\|_1^2 = o(1) \end{aligned}$$

by using $E(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| \leq a_T}) = -E(\varepsilon_1 \mathbb{I}_{|\varepsilon_1| > a_T})$ and KT, and (ii),

$$E(S_{\rho\rho}^>)^{\frac{1}{2}} \leq E(|\varepsilon_1| \mathbb{I}_{|\varepsilon_1| > a_T}) T \sum_{i=k+1}^{\infty} |\beta_i| \sum_{j=0}^{\infty} |\gamma_j| = O(l_T)(a_T \sum_{i=k+1}^{\infty} |\beta_i|) = o(l_T),$$

whereas for $E|\varepsilon_1| = \infty$ it holds that $(S_{\rho\rho}^k)^{1/2} \leq \sum_{t=k+1}^T |\rho_{t,k}| = o_P(l_T)$ by the earlier argument for (S.4.2), so $S_{\rho\rho}^k = o_P(l_T)$ independently of $E|\varepsilon_1|$ (as the square of slowly varying is slowly varying). Thus, continuing (S.4.4),

$$(S.4.5) \quad \|\hat{\varepsilon}_T - \varepsilon_T\|^2 = O_P(a_T^2 \|\hat{\beta}_k - \beta_k\|^2) + o_P(l_T) = o_P(a_T^2).$$

Further, as $\hat{\varepsilon}_t^2 \leq 2\varepsilon_t^2 + 2(\hat{\varepsilon}_t - \varepsilon_t)^2$, it holds P^π -a.s. that

$$\sum_{t=k+1}^T \hat{\varepsilon}_{\pi(t)}^2 = \sum_{t=k+1}^T \hat{\varepsilon}_t^2 =: \hat{\sigma}_{Tk}^2 \leq 2\sigma_T^2 + 2\|\hat{\varepsilon}_T - \varepsilon_T\|^2 = O_P(a_T^2).$$

S.4.1.2. *Proof of part (a).* After this preparation, we turn to

$$\begin{aligned} S_{00}^{*k} - S_{00}^{\dagger k} &= \sum_{t=k+2}^T \left\{ \sum_{j=0}^{t-k-2} \hat{\gamma}_{j:k} \varepsilon_{t-j-1}^* \sum_{j=0}^{t-k-2} \hat{\gamma}'_{j:k} \varepsilon_{t-j-1}^* \right. \\ &\quad \left. - \sum_{j=0}^{t-k-2} \gamma_{j:k} \varepsilon_{t-j-1}^\dagger \sum_{j=0}^{t-k-2} \gamma'_{j:k} \varepsilon_{t-j-1}^\dagger \right\} \\ &= \sum_{t=k+2}^T \sum_{j,i=0}^{t-k-2} w_{t-j-1} w_{t-i-1} \{ \hat{\gamma}_{j:k} \hat{\gamma}'_{i:k} \hat{\varepsilon}_{\pi(t-j-1)} \hat{\varepsilon}_{\pi(t-i-1)} \\ &\quad - \gamma_{j:k} \gamma'_{i:k} \varepsilon_{\pi(t-j-1)} \varepsilon_{\pi(t-i-1)} \}. \end{aligned}$$

Let $\mathbf{G}_{ji:k} := \text{vec}(\gamma_{j:k} \gamma'_{i:k})$ and $\hat{\mathbf{G}}_{ji:k} := \text{vec}(\hat{\gamma}_{j:k} \hat{\gamma}'_{i:k})$, such that

$$\Delta_{00} := \text{vec}(S_{00}^{*k} - S_{00}^{\dagger k}) = \sum_{t=k+2}^T \sum_{j,i=0}^{t-k-2} w_{t-j-1} w_{t-i-1} (\hat{\mathbf{G}}_{ji:k} \hat{\varepsilon}_{\pi(t-j-1)} \hat{\varepsilon}_{\pi(t-i-1)} \\ - \mathbf{G}_{ji:k} \varepsilon_{\pi(t-j-1)} \varepsilon_{\pi(t-i-1)}).$$

We split Δ_{00} into $\Delta_{00} = \Delta_{00}^{(1)} + \Delta_{00}^{(2)} + \Delta_{00}^{(3)}$ with $\Delta_{00}^{(1)} := \sum_{t=k+1}^{T-1} \varepsilon_{\pi(t)}^2 c_{t,t}$,

$$\begin{aligned} \Delta_{00}^{(2)} &:= \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} w_s w_t \varepsilon_{\pi(s)} \varepsilon_{\pi(t)} c_{s,t} \\ \Delta_{00}^{(3)} &:= \sum_{t=k+1}^{T-1} \sum_{j,i=0}^{t-k-2} \hat{\mathbf{G}}_{ji:k} w_{t-j} w_{t-i} (\hat{\varepsilon}_{\pi(t-j)} \hat{\varepsilon}_{\pi(t-i)} - \varepsilon_{\pi(t-j)} \varepsilon_{\pi(t-i)}) \end{aligned}$$

$$= \sum_{s,t=k+1}^{T-1} w_s w_t (\hat{\varepsilon}_{\pi(s)} \hat{\varepsilon}_{\pi(t)} - \varepsilon_{\pi(s)} \varepsilon_{\pi(t)}) \hat{d}_{s,t},$$

where

$$c_{s,t} := \sum_{i=\max\{0,s-t\}}^{T-t-1} (\hat{\mathbf{G}}_{t-s+i,i:k} - \mathbf{G}_{t-s+i,i:k}), \quad \hat{d}_{s,t} := \sum_{i=\max\{0,s-t\}}^{T-t-1} \hat{\mathbf{G}}_{t-s+i,i:k}.$$

For an r.p. π , with E^\dagger denoting expectation under P^\dagger , it holds that, first, $E^\dagger \|\Delta_{00}^{(1)}\| \leq \sigma_T^2 \max_{t=k+1,\dots,T} \|c_{t,t}\|$, where $c_{t,t}$ remain to be evaluated. Second, regarding $\Delta_{00}^{(2)}$, for Rademacher w_t it holds that

$$\begin{aligned} (S.4.6) E^\dagger \|\Delta_{00}^{(2)}\|^2 &= \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} E^\dagger (\varepsilon_{\pi(s)}^2 \varepsilon_{\pi(t)}^2) \{c'_{s,t} c_{s,t} + c'_{s,t} c_{t,s}\} \\ &= E^\dagger (\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2) \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} \{c'_{s,t} c_{s,t} + c'_{s,t} c_{t,s}\} \\ &= O_P(T^{-2} a_T^4) \sum_{s,t=k+1}^{T-1} |c'_{s,t} c_{s,t} + c'_{s,t} c_{t,s}| \end{aligned}$$

because

$$\begin{aligned} (S.4.7) E^\dagger (\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2) &= \sum_{u,v=k+1}^T \mathbb{I}_{u \neq v} P\{\pi(k+1) = u, \pi(k+2) = v\} \varepsilon_u^2 \varepsilon_v^2 \\ &= O(T^{-2}) \{ \sigma_T^4 - \sum_{t=k+1}^T \varepsilon_t^4 \} = O_P(T^{-2} a_T^4), \end{aligned}$$

whereas for $w_t = 1$ a.s. (all t),

$$\begin{aligned} E^\dagger \|\Delta_{00}^{(2)}\|^2 &= E^\dagger (\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2) \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} \{c'_{s,t} c_{s,t} + c'_{s,t} c_{t,s}\} \\ &\quad + E^\dagger (\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T)}) \\ &\quad \times \sum_{s,t,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,v\}=3} \{c'_{s,t} c_{s,v} + c'_{s,t} c_{v,s} + c'_{t,s} c_{s,v} + c'_{t,s} c_{v,s}\} \\ &\quad + E^\dagger (\varepsilon_{\pi(k+1)} \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T-1)} \varepsilon_{\pi(T)}) \sum_{s,t,u,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,u,v\}=4} c'_{s,t} c_{u,v}, \end{aligned}$$

where $E^\dagger(\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2) = O_P(T^{-2} a_T^4)$ as before,

$$(S.4.8) \quad E^\dagger(\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T)}) \\ = O(T^{-3}) \{ \sigma_T^2 [(\sum_{t=k+1}^T \varepsilon_t)^2 - \sigma_T^2] - 2[(\sum_{t=k+1}^T \varepsilon_t^3)(\sum_{t=k+1}^T \varepsilon_t) - \sum_{t=k+1}^T \varepsilon_t^4] \}$$

is $O_P(T^{-3} a_T^4 l_T)$ as powers of slowly varying functions vary slowly, and

$$(S.4.9) \leq O(T^{-4}) \{ (\sum_{t=k+1}^T \varepsilon_t)^4 + \sum_{t=k+1}^T \varepsilon_t^4 + 4\sigma_T^4 + 8(\sum_{t=k+1}^T \varepsilon_t^3)(\sum_{t=k+1}^T \varepsilon_t) \}$$

is $O_P(T^{-4} a_T^4 l_T)$ because $\sum_{t=k+1}^T \varepsilon_t = O_P(a_T l_T)$ and $\sum_{t=k+1}^T \varepsilon_t^i = O_P(a_T^i)$ ($i = 2, \dots, 4$), so

$$(S.4.10) \quad E^\dagger \|\Delta_{00}^{(2)}\|^2 = O_P(T^{-2} a_T^4) \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} |c'_{s,t} c_{s,t} + c'_{s,t} c_{t,s}| \\ + O_P(T^{-3} a_T^4 l_T) \sum_{s,t,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,v\}=3} |c'_{s,t} c_{s,v} + c'_{s,t} c_{v,s} + c'_{t,s} c_{s,v} + c'_{t,s} c_{v,s}| \\ + O_P(T^{-4} a_T^4 l_T) \sum_{s,t,u,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,u,v\}=4} |c'_{s,t} c_{u,v}|,$$

and $c_{s,t}$ remain to be evaluated.

Third, regarding $\Delta_{00}^{(3)}$, for Rademacher w_t it holds that

$$\begin{aligned} E^\dagger \|\Delta_{00}^{(3)}\|^2 &= \sum_{s,t=k+1}^{T-1} E^\dagger (\hat{\varepsilon}_{\pi(s)} \hat{\varepsilon}_{\pi(t)} - \varepsilon_{\pi(s)} \varepsilon_{\pi(t)})^2 \{ \hat{d}'_{s,t} \hat{d}_{s,t} + \mathbb{I}_{s \neq t} \hat{d}'_{s,t} \hat{d}_{t,s} \} \\ &\leq E^\dagger (\hat{\varepsilon}_{\pi(k+1)}^2 - \varepsilon_{\pi(k+1)}^2)^2 \sum_{s=k+1}^{T-1} \|\hat{d}_{s,s}\|^2 \\ &\quad + E^\dagger (\hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+1)} \varepsilon_{\pi(k+2)})^2 \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} |\hat{d}'_{s,t} \hat{d}_{s,t} + \hat{d}'_{s,t} \hat{d}_{t,s}|, \end{aligned}$$

where $E^\dagger(\hat{\varepsilon}_{\pi(k+1)}^2 - \varepsilon_{\pi(k+1)}^2)^2$ equals

$$\begin{aligned} O(T^{-1}) \sum_{s=k+1}^T (\hat{\varepsilon}_s^2 - \varepsilon_s^2)^2 &\leq O(T^{-1}) \|\hat{\varepsilon}_T - \varepsilon_T\|^2 (\sigma_T^2 + \|\hat{\varepsilon}_T - \varepsilon_T\|^2) \\ &= O_P(T^{-1} a_T^4 \|\hat{\beta}_k - \beta_k\|^2) + o_P(T^{-1} l_T a_T^2) \end{aligned}$$

and is $O_P(T^{-1}a_k^2a_T^{2+\epsilon})$ for all $\epsilon > 0$, using (S.4.5) and (7.1), and also

$$\begin{aligned} (\text{S.4.11}) \quad & \mathbb{E}^\dagger(\hat{\varepsilon}_{\pi(k+1)}\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+1)}\varepsilon_{\pi(k+2)})^2 \\ & = O(T^{-2}) \sum_{s,t=k+1}^T \mathbb{I}_{s \neq t}(\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t)^2 \leq O(T^{-2})(\hat{\sigma}_{Tk}^2 + \sigma_T^2) \|\hat{\varepsilon}_T - \varepsilon_T\|^2 \\ & = O_P(T^{-2}a_T^4\|\hat{\beta}_k - \beta_k\|^2) + o_P(T^{-2}l_T a_T^2) \end{aligned}$$

is $O_P(T^{-2}a_k^2a_T^{2+\epsilon})$ for all $\epsilon > 0$, so

$$\begin{aligned} (\text{S.4.12}) \quad & \mathbb{E}^\dagger \|\Delta_{00}^{(3)}\|^2 = O_P(T^{-1}a_k^2a_T^{2+\epsilon}) \sum_{s=k+1}^{T-1} \|\hat{d}_{s,s}\|^2 \\ & + O_P(T^{-2}a_k^2a_T^{2+\epsilon}) \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} |\hat{d}'_{s,t}\hat{d}_{s,t} + \hat{d}'_{s,t}\hat{d}_{t,s}| \end{aligned}$$

for all $\epsilon > 0$, where $\hat{d}_{s,t}$ remain to be evaluated. If $w_t = 1$ a.s. (all t),

$$\begin{aligned} \mathbb{E}^\dagger \|\Delta_{00}^{(3)}\|^2 & = \sum_{s,t=k+1}^{T-1} \mathbb{E}^\dagger(\hat{\varepsilon}_{\pi(s)}\hat{\varepsilon}_{\pi(t)} - \varepsilon_{\pi(s)}\varepsilon_{\pi(t)})^2 \{ \hat{d}'_{s,t}\hat{d}_{s,t} + \mathbb{I}_{s \neq t}\hat{d}'_{s,t}\hat{d}_{t,s} \} \\ & + \mathbb{E}^\dagger \{ (\hat{\varepsilon}_{\pi(k+1)}\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+1)}\varepsilon_{\pi(k+2)})(\hat{\varepsilon}_{\pi(k+1)}\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(k+1)}\varepsilon_{\pi(T)}) \} \\ & \quad \times \sum_{s,t,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,v\}=3} (\hat{d}'_{s,t} + \hat{d}'_{s,t})(\hat{d}_{s,v} + \hat{d}_{v,s}) \\ & + \mathbb{E}^\dagger \{ (\hat{\varepsilon}_{\pi(k+1)}\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+1)}\varepsilon_{\pi(k+2)})(\hat{\varepsilon}_{\pi(T-1)}\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T-1)}\varepsilon_{\pi(T)}) \} \\ & \quad \times \sum_{s,t,u,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,u,v\}=4} \hat{d}'_{s,t}\hat{d}_{u,v}, \end{aligned}$$

where $|\mathbb{E}^\dagger \{ (\hat{\varepsilon}_{\pi(k+1)}\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+1)}\varepsilon_{\pi(k+2)})(\hat{\varepsilon}_{\pi(k+1)}\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(k+1)}\varepsilon_{\pi(T)}) \}|$ equals

$$\begin{aligned} & O(T^{-3}) \left| \sum_{s,t,v=k+1}^T \mathbb{I}_{\#\{s,t,v\}=3} (\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t)(\hat{\varepsilon}_s\hat{\varepsilon}_v - \varepsilon_s\varepsilon_v) \right| \\ & = O(T^{-3}) \sum_{s=k+1}^T \left[\left\{ \sum_{t=k+1}^T \mathbb{I}_{s \neq t}(\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t) \right\}^2 - \sum_{t=k+1}^T \mathbb{I}_{s \neq t}(\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t)^2 \right] \\ & \leq O(T^{-3}) \sum_{s=k+1}^T \left\{ \sum_{t=k+1}^T \mathbb{I}_{s \neq t}(\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t) \right\}^2 \\ & = O(T^{-3}) \sum_{s=k+1}^T [\hat{\varepsilon}_s^2 \left\{ \sum_{t=k+1}^T \mathbb{I}_{s \neq t}(\hat{\varepsilon}_t - \varepsilon_t) \right\}^2 + (\hat{\varepsilon}_s - \varepsilon_s)^2 \left(\sum_{t=k+1}^T \mathbb{I}_{s \neq t}\varepsilon_t \right)^2] \end{aligned}$$

$$\leq O(T^{-3})[\hat{\sigma}_{Tk}^2 \{ \sum_{t=k+1}^T (\hat{\varepsilon}_t - \varepsilon_t) \}^2 + \|\hat{\varepsilon}_T - \varepsilon_T\|^2 \{ \hat{\sigma}_{Tk}^2 + \sigma_T^2 + (\sum_{t=k+1}^T \varepsilon_t)^2 \}]$$

is $O_P(T^{-3}ka_k^2a_T^{2+\epsilon})$ for all $\epsilon > 0$, using (S.4.3), (S.4.5) and (7.1), and similarly, $|\mathbb{E}^\dagger \{(\hat{\varepsilon}_{\pi(k+1)}\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+1)}\varepsilon_{\pi(k+2)})(\hat{\varepsilon}_{\pi(T-1)}\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T-1)}\varepsilon_{\pi(T)})\}|$ equals

$$\begin{aligned} O(T^{-4}) & \left| \sum_{s,t,u,v=k+1}^T \mathbb{I}_{\#\{s,t,u,v\}=4} (\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t)(\hat{\varepsilon}_u\hat{\varepsilon}_v - \varepsilon_u\varepsilon_v) \right| \\ & = O(T^{-4}) \left\{ \sum_{s,t=k+1}^T \mathbb{I}_{s \neq t} (\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t) \right\}^2 \\ & - 4 \sum_{s,t,v=k+1}^T \mathbb{I}_{\#\{s,t,v\}=3} (\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t)(\hat{\varepsilon}_s\hat{\varepsilon}_v - \varepsilon_s\varepsilon_v) - 2 \sum_{s,t=k+1}^T \mathbb{I}_{s \neq t} (\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t)^2 \} \\ & \leq O(T^{-4}) \left\{ \sum_{s,t=k+1}^T \mathbb{I}_{s \neq t} (\hat{\varepsilon}_s\hat{\varepsilon}_t - \varepsilon_s\varepsilon_t) \right\}^2 + O_P(T^{-4}ka_k^2a_T^{2+\epsilon}) \end{aligned}$$

using previous evaluations, so further

$$\begin{aligned} & = O(T^{-4}) \{ [\sum_{t=k+1}^T (\hat{\varepsilon}_t - \varepsilon_t)]^2 + 2 \sum_{t=k+1}^T \varepsilon_t \sum_{s=k+1}^T (\hat{\varepsilon}_s - \varepsilon_s) + \sum_{t=k+1}^T (\varepsilon_t^2 - \hat{\varepsilon}_t^2) \}^2 \\ & + O_P(T^{-4}ka_k^2a_T^{2+\epsilon}) = O(T^{-4}) \{ (k^{1/2}a_k a_T^\epsilon + a_T l_T) k^{1/2} a_k a_T^\epsilon \\ & + \|\hat{\varepsilon}_T - \varepsilon_T\|^2 + 2 \|\hat{\varepsilon}_T - \varepsilon_T\| \hat{\sigma}_{Tk} \}^2 + O_P(T^{-4}ka_k^2a_T^{2+\epsilon}) \end{aligned}$$

which is $O_P(T^{-4}ka_k^2a_T^{2+\epsilon})$ for all $\epsilon > 0$. Hence,

$$\begin{aligned} \mathbb{E}^\dagger \|\Delta_{00}^{(3)}\|^2 & = O_P(T^{-1}a_k^2a_T^{2+\epsilon}) \sum_{s=k+1}^{T-1} \|\hat{d}_{s,s}\|^2 \\ & + O_P(T^{-2}a_k^2a_T^{2+\epsilon}) \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} |\hat{d}'_{s,t} \hat{d}_{s,t} + \hat{d}'_{s,t} \hat{d}_{t,s}| \\ & + O_P(T^{-3}ka_k^2a_T^{2+\epsilon}) \sum_{s,t,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,v\}=3} |(\hat{d}'_{s,t} + \hat{d}'_{s,t})(\hat{d}_{s,v} + \hat{d}_{v,s})| \\ (S.4.13) \quad & + O_P(T^{-4}ka_k^2a_T^{2+\epsilon}) \sum_{s,t,u,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,u,v\}=4} |\hat{d}'_{s,t} \hat{d}_{u,v}|. \end{aligned}$$

We now turn to $c_{s,t}$ and $\hat{d}_{s,t}$. As in the proof of Corollary 4,

$$(S.4.14) \quad \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\|_\infty \leq \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\|_1 \\ := \sum_{j=1}^{T-k-2} |\hat{\gamma}_j - \gamma_j| = O_P(\|\hat{\beta}_k - \beta_k\|_1 + \sum_{j=k+1}^{\infty} |\beta_j|),$$

$$(S.4.15) \quad \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\| := (\sum_{j=1}^{T-k-2} |\hat{\gamma}_j - \gamma_j|^2)^{1/2} \\ = O_P(\|\hat{\beta}_k - \beta_k\| + \sum_{j=k+1}^{\infty} |\beta_j|).$$

Using also the identity $\|vec(\mathbf{ab}')\| = \|\mathbf{a}\| \|\mathbf{b}\|$, the triangle inequality and (7.1), we obtain, for all $\epsilon > 0$ and s, t that

$$\begin{aligned} \|c_{s,t}\| &= \|vec\left\{\sum_{i=\max\{0,s-t\}}^{T-t-1} (\hat{\gamma}_{t-s+i:k} - \gamma_{t-s+i:k})\hat{\gamma}'_{i:k} + \gamma_{t-s+i:k}(\hat{\gamma}_{i:k} - \gamma_{i:k})'\right\}\| \\ &\leq (2\|\gamma\| + \|\hat{\gamma}_{T-k-2} - \gamma_{T+k-2}\|) \\ &\quad \times \sum_{i=\max\{0,s-t\}}^{T-t-1} (\|\hat{\gamma}_{t-s+i:k} - \gamma_{t-s+i:k}\| + \|\hat{\gamma}_{i:k} - \gamma_{i:k}\|) \\ &= O_P(k)\|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\| = O_P(ka_k a_T^{\epsilon-1}) \end{aligned}$$

uniformly in s, t . Thus, $E^\dagger \|\Delta_{00}^{(1)}\| = O_P(ka_k a_T^{1+\epsilon})$ and, returning to (S.4.6) and (S.4.10), $E^\dagger \|\Delta_{00}^{(2)}\|^2 = O_P(k^2 a_k^2 a_T^{2+\epsilon})$ for all $\epsilon > 0$.

Further, using (S.4.15), (i),

$$\begin{aligned} \sum_{s=k+1}^{T-1} \|\hat{d}_{s,s}\|^2 &\leq \sum_{s=k+1}^{T-1} (\sum_{i=0}^{T-s-1} \|\hat{\gamma}_{i:k}\|^2)^2 \leq \sum_{s=k+1}^{T-1} (k \sum_{i=0}^{T-s-1} |\hat{\gamma}_i|^2)^2 \\ &\leq 2Tk^2(\|\gamma\|^4 + \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\|^4) = O_P(Tk^2), \end{aligned}$$

and (ii),

$$\begin{aligned} \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} |\hat{d}'_{s,t} \hat{d}_{s,t} + \hat{d}'_{s,t} \hat{d}_{t,s}| &\leq \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} (\|\hat{d}_{s,t}\|^2 + \|\hat{d}_{s,t}\| \|\hat{d}_{t,s}\|) \\ &\leq 2 \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} (\sum_{i=0}^{T-1-\max\{t,s\}} \|\hat{\gamma}_{|t-s|+i:k}\| \|\hat{\gamma}_{i:k}\|)^2 \\ &\leq \|\hat{\gamma}_{T-k-2}\|^2 \sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} (\sum_{i=0}^{T-1-\max\{t,s\}} \|\hat{\gamma}_{|t-s|+i:k}\|)^2 \end{aligned}$$

$$\begin{aligned}
&= O(T^2)(\|\gamma\| + \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\|)^2 (\sum_{i=0}^{T-k-2} \|\hat{\gamma}_{i:k}\|)^2 \\
&= O(k^2 T^2)(\|\gamma\| + \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\|)^4 = O_P(k^2 T^2).
\end{aligned}$$

Before (iii), observe that, by Assumption 1(b), Lemma S.2 and (7.1),

$$\begin{aligned}
(\text{S.4.16}) \quad & \sum_{i=1}^{T-k-2} i \|\hat{\gamma}_{i:k}\| \leq \sum_{i=1}^{T-k-2} i \|\hat{\gamma}_{i:k} - \gamma_{i:k}\| + \sum_{i=1}^{T-k-2} i \|\gamma_{i:k}\| \\
&\leq \sum_{i=1}^{T-k-2} i \{ \sum_{j=0}^{k-1} (\hat{\gamma}_{i-j} - \gamma_{i-j})^2 \}^{1/2} + k \sum_{i=1}^{\infty} i |\gamma_i| + k^2 \|\gamma\|_1 \\
&= O_P(\|\hat{\beta}_k - \beta_k\|_1 + \sum_{j=k+1}^{\infty} |\beta_j|) k^{1/2} \sum_{i=1}^{T-k-2} i (1 + \frac{1}{k})^{i \wedge k-i-1} \\
&\quad + \sqrt{2} \sum_{i=1}^{T-k-2} i \sum_{j=0}^{(k-1) \wedge i} |b_{i-j,k}| + O(k^2) \\
&= O_P(\|\hat{\beta}_k - \beta_k\|_1 + \sum_{j=k+1}^{\infty} |\beta_j|) k^{5/2} + O(kT) \sum_{j=k+1}^{\infty} |\beta_j| \\
&\quad + O(k^2) = O_P(k^3 a_k a_T^{\epsilon-1} + k T a_T^{-1} + k^2)
\end{aligned}$$

for all $\epsilon > 0$, so using also (S.4.15) and (7.1),

$$\begin{aligned}
& \sum_{s,t,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,v\}=3} |(\hat{d}_{s,t} + \hat{d}_{t,s})' (\hat{d}_{s,v} + \hat{d}_{v,s})| \\
&\leq \sum_{s=k+1}^{T-1} \left(\sum_{t=k+1}^{T-1} \mathbb{I}_{s \neq t} \|\hat{d}_{s,t} + \hat{d}_{t,s}\|^2 \right)^2 \\
&\leq 4 \sum_{s=k+1}^{T-1} \left(\sum_{t=k+1}^{T-1} \mathbb{I}_{s \neq t} \sum_{i=0}^{T-1-\max\{t,s\}} \|\hat{\gamma}_{|t-s|+i:k}\| \|\hat{\gamma}_{i:k}\| \right)^2 \\
&\leq 4 \|\hat{\gamma}_{T-k-2}\|^2 \sum_{s=k+1}^{T-1} \left(\sum_{t=k+1}^{T-1} \mathbb{I}_{s \neq t} \sum_{i=0}^{T-1-\max\{t,s\}} \|\hat{\gamma}_{|t-s|+i:k}\| \right)^2 \\
&\leq 4(\|\gamma\| + \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\|)^2 \sum_{s=k+1}^{T-1} (2 \sum_{i=1}^{T-k-2} i \|\hat{\gamma}_{i:k}\|)^2 \\
&= O_P(k^4 T + k^6 a_k^2 T a_T^{\epsilon-2} + k^2 T^3 a_T^{-2})
\end{aligned}$$

for all $\epsilon > 0$, and (iv), similarly,

$$\begin{aligned}
& \sum_{s,t,u,v=k+1}^{T-1} \mathbb{I}_{\#\{s,t,u,v\}=4} |\hat{d}'_{s,t} \hat{d}_{u,v}| \leq \left(\sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} \|\hat{d}_{s,t}\| \right)^2 \\
& \leq \left(\sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} \sum_{i=0}^{T-1-\max\{t,s\}} \|\hat{\gamma}_{|t-s|+i:k}\| \|\hat{\gamma}_{i:k}\| \right)^2 \\
& \leq \|\hat{\gamma}_{T-k-2}\|^2 \left(\sum_{s,t=k+1}^{T-1} \mathbb{I}_{s \neq t} \sum_{i=0}^{T-1-\max\{t,s\}} \|\hat{\gamma}_{|t-s|+i:k}\| \right)^2 \\
& \leq (\|\gamma\| + \|\hat{\gamma}_{T-k-2} - \gamma_{T-k-2}\|)^2 \left(2 \sum_{s=k+1}^{T-1} \sum_{i=1}^{T-k-2} i \|\hat{\gamma}_i\| \right)^2 \\
& = O_P(k^4 T^2 + k^6 a_k^2 T^2 a_T^{\epsilon-2} + k^2 T^4 a_T^{-2}).
\end{aligned}$$

Thus, returning to (S.4.12) and (S.4.13),

$$\begin{aligned}
\mathbf{E}^\dagger \|\Delta_{00}^{(3)}\|^2 &= O_P(k^2 a_k^2 a_T^{2+\epsilon}) \\
&\quad + O_P(T^{-4} k a_k^2 a_T^{2+\epsilon}) O_P(k^4 T^2 + k^6 a_k^2 T^2 a_T^{\epsilon-2} + k^2 T^4 a_T^{-2})
\end{aligned}$$

is $O_P(k^2 a_k^2 a_T^{2+\epsilon})$ for all $\epsilon > 0$, if $k^3/T \rightarrow 0$. As also $\mathbf{E}^\dagger \|\Delta_{00}^{(1)}\| = O_P(k a_k a_T^{1+\epsilon})$ and $\mathbf{E}^\dagger \|\Delta_{00}^{(2)}\|^2 = O_P(k^2 a_k^2 a_T^{2+\epsilon})$ were found to hold, it follows that $\|\Delta_{00}\| = O_{P^\dagger}(k a_k a_T^{1+\epsilon})$ in P -probability for all $\epsilon > 0$, in the case where π is an r.p.

For $\pi = id$ (and Rademacher w_t), $\mathbf{E}^\dagger \|\Delta_{00}^{(1)}\| \leq \sigma_T^2 \max_{t=k+1,\dots,T} \|c_{t,t}\|$ is $O_P(k a_k a_T^{1+\epsilon})$ as previously,

$$\mathbf{E}^\dagger \|\Delta_{00}^{(2)}\|^2 \leq \sum_{s,t=k+1}^{T-1} \varepsilon_s^2 \varepsilon_t^2 \mathbb{I}_{s \neq t} |c'_{s,t} c_{s,t} + c'_{s,t} c_{t,s}| \leq \sigma_T^4 O_P(k^2 a_k^2 a_T^{\epsilon-2})$$

is $O_P(k^2 a_k^2 a_T^{2+\epsilon})$ using the previous uniform estimate of $\|c_{s,t}\|$, and finally,

$$\begin{aligned}
\mathbf{E}^\dagger \|\Delta_{00}^{(3)}\|^2 &= \mathbf{E}^\dagger \left\| \sum_{s,t=k+1}^{T-1} w_s w_t (\hat{\varepsilon}_s \hat{\varepsilon}_t - \varepsilon_s \varepsilon_t) \hat{d}_{s,t} \right\|^2 \\
&\leq \sum_{s,t=k+1}^{T-1} (\hat{\varepsilon}_s \hat{\varepsilon}_t - \varepsilon_s \varepsilon_t)^2 \{ \hat{d}'_{s,t} \hat{d}_{s,t} + \hat{d}'_{s,t} \hat{d}_{t,s} \} \\
&\leq \left\{ \sum_{s,t=k+1}^{T-1} (\hat{\varepsilon}_s \hat{\varepsilon}_t - \varepsilon_s \varepsilon_t)^2 \right\} \left(\max_{s,t=k+1,\dots,T-1} \|\hat{d}_{s,t}\| \right)^2,
\end{aligned}$$

where $\sum_{s,t=k+1}^{T-1} (\hat{\varepsilon}_s \hat{\varepsilon}_t - \varepsilon_s \varepsilon_t)^2 = O_P(a_k^2 a_T^{2+\epsilon})$ for $\epsilon > 0$ as in (S.4.11) and

$$\begin{aligned} \|\hat{d}_{s,t}\| &\leq \sum_{i=\max\{0,s-t\}}^{T-t-1} \|\mathbf{G}_{t-s+i,i:k}\| + \|c_{s,t}\| \\ &\leq \sum_{i=\max\{0,s-t\}}^{T-t-1} \|\boldsymbol{\gamma}_{t-s+i:k}\| \|\boldsymbol{\gamma}_{i:k}\| + \|c_{s,t}\| \leq k \|\boldsymbol{\gamma}\|^2 + O_P(k a_k a_T^{\epsilon-1}) \end{aligned}$$

is $O_P(k)$ uniformly in s, t , so also $E^\dagger \|\Delta_{00}^{(3)}\|^2 = O_P(k^2 a_k^2 a_T^{2+\epsilon})$ for every $\epsilon > 0$. By combining the evaluations of $\|\Delta_{00}^{(i)}\|$ ($i = 1, 2, 3$) and applying Markov's inequality, the first statement in part (a) is proved also for the wild bootstrap scheme.

Regarding the lower bound for $S_{00}^{\dagger k}$, let

$$\begin{aligned} \tau &:= \min \left\{ t : k+1 \leq t \leq T, |\varepsilon_t| = \max_{k+1 \leq s \leq T} |\varepsilon_s| \right\}, \\ \mathcal{T} &:= \left\{ \tau = \min \{t : k+1 \leq t \leq T, |\varepsilon_t| = \max_{k+1 \leq s \leq T-1} |\varepsilon_{\pi(s)}| \} \right\}; \end{aligned}$$

then $P(\mathcal{T}) \rightarrow 1$. By considerations of positive semi-definiteness, for outcomes in \mathcal{T} it holds that

$$(S.4.17) \quad \lambda_{\min}(S_{00}^{\dagger k}) \geq \lambda_{\min} \left(\varepsilon_\tau^2 \sum_{j=0}^{T-\pi^{-1}(\tau)-1} \boldsymbol{\gamma}_{j:k} \boldsymbol{\gamma}'_{j:k} + \varepsilon_\tau w_{\pi^{-1}(\tau)} \Delta_\lambda \right),$$

where the right-hand side matrix collects the terms of $S_{00}^{\dagger k}$ involving ε_τ , with

$$\Delta_\lambda := \sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq \pi(t)} w_t \varepsilon_{\pi(t)} (d_{\pi^{-1}(\tau),t} + d_{t,\pi^{-1}(\tau)}),$$

$d_{s,t} := \sum_{i=\max\{0,s-t\}}^{T-t-1} \boldsymbol{\gamma}_{t-s+i:k} \boldsymbol{\gamma}'_{i:k}$. We evaluate Δ_λ for the three bootstrap schemes.

If π is an r.p., then

$$\begin{aligned} E^\dagger \|\Delta_\lambda\|^2 &= \sum_{t=k+1}^{T-1} E^\dagger (\mathbb{I}_{\tau \neq \pi(t)} \varepsilon_{\pi(t)}^2 \|d_{\pi^{-1}(\tau),t} + d_{t,\pi^{-1}(\tau)}\|^2) \\ &\quad + \sum_{t,s=k+1}^{T-1} E^\dagger [\mathbb{I}_{\pi(t) \neq \tau \neq \pi(s), t \neq s} w_t w_s \varepsilon_{\pi(t)} \varepsilon_{\pi(s)} \\ &\quad \times \text{tr}\{(d_{\pi^{-1}(\tau),t} + d_{t,\pi^{-1}(\tau)})' (d_{\pi^{-1}(\tau),s} + d_{s,\pi^{-1}(\tau)})\}]. \end{aligned}$$

Next, if further w_t are Rademacher, this reduces to

$$\begin{aligned}
\mathbb{E}^\dagger \|\Delta_\lambda\|^2 &= \sum_{t=k+1}^{T-1} \mathbb{E}^\dagger (\mathbb{I}_{\tau \neq \pi(t)} \varepsilon_{\pi(t)}^2 \|d_{\pi^{-1}(\tau),t} + d_{t,\pi^{-1}(\tau)}\|^2) \\
&= \sum_{t=k+1}^{T-1} \mathbb{E}^\dagger \left(\sum_{u,v=k+1}^T \mathbb{I}_{t \neq v} \mathbb{I}_{\pi(t)=u, \pi(v)=\tau} \varepsilon_u^2 \|d_{v,t} + d_{t,v}\|^2 \right) \\
&= \sum_{t=k+1}^{T-1} \sum_{u,v=k+1}^T \mathbb{I}_{t \neq v} \mathbb{P}^\dagger(\pi(t)=u, \pi(v)=\tau) \varepsilon_u^2 \|d_{v,t} + d_{t,v}\|^2 \\
&= O(T^{-2}) \sum_{t=k+1}^{T-1} \sum_{u,v=k+1}^T \mathbb{I}_{t \neq v} \mathbb{I}_{u \neq \tau} \varepsilon_u^2 \|d_{v,t} + d_{t,v}\|^2 \\
&= O(T^{-2}) \sigma_T^2 \sum_{v=k+1}^T \sum_{t=k+1}^{T-1} \mathbb{I}_{t \neq v} \|d_{v,t} + d_{t,v}\|^2
\end{aligned}$$

because $\mathbb{P}^\dagger(\pi(t)=u, \pi(v)=\tau) = (T-k)^{-1} (T-k-1)^{-1}$ for $t \neq v, u \neq \tau$ (τ is \dagger -measurable). Further,

$$\begin{aligned}
\sum_{t=k+1}^{T-1} \mathbb{I}_{t \neq v} \|d_{v,t}\|^2 &\leq \sum_{t=k+1}^{T-1} \mathbb{I}_{t \neq v} \left\| \sum_{i=\max\{0, v-t\}}^{T-t-1} \gamma_{t-v+i:k} \gamma'_{i:k} \right\|^2 \\
(S.4.18) \quad &\leq 2k \|\gamma\|^2 (k^2 \|\gamma\|^2 + k \sum_{i=1}^{\infty} i \gamma_i^2) = O(k^3)
\end{aligned}$$

and similarly for $\sum_{t=k+1}^{T-1} \mathbb{I}_{t \neq v} \|d_{t,v}\|^2$, so for the permuted wild bootstrap, $\mathbb{E}^\dagger \|\Delta_\lambda\|^2 = O_P(T^{-1}k^3)\sigma_T^2 = O_P(T^{-1}k^3a_T^2) = o_P(a_T^2)$ as $k^3/T \rightarrow 0$, and $\|\Delta_\lambda\| = o_{P^\dagger}(a_T)$ in P -probability. If $w_t = 1$ a.s. (all t), the estimate of $\sum_{t=k+1}^{T-1} \mathbb{E}^\dagger (\mathbb{I}_{\tau \neq \pi(t)} \varepsilon_{\pi(t)}^2 \|d_{\pi^{-1}(\tau),t} + d_{t,\pi^{-1}(\tau)}\|^2)$ remains valid. Additionally,

$$\begin{aligned}
&\mathbb{E}^\dagger [\mathbb{I}_{\pi(t) \neq \tau \neq \pi(s), t \neq s} \varepsilon_{\pi(t)} \varepsilon_{\pi(s)} \text{tr}\{(d_{\pi^{-1}(\tau),t} + d_{t,\pi^{-1}(\tau)})'(d_{\pi^{-1}(\tau),s} + d_{s,\pi^{-1}(\tau)})\}] \\
&= \mathbb{E}^\dagger \sum_{u,v,w=k+1}^T \mathbb{I}_{\#\{t,s,w\}=3} \mathbb{I}_{\pi(t)=u, \pi(s)=v, \pi(w)=\tau} \varepsilon_u \varepsilon_v \\
&\quad \times \text{tr}\{(d_{w,t} + d_{t,w})'(d_{w,s} + d_{s,w})\} \\
&= \sum_{u,v,w=k+1}^T \mathbb{I}_{\#\{t,s,w\}=3} \mathbb{P}^\dagger(\pi(t)=u, \pi(s)=v, \pi(w)=\tau) \\
&\quad \times \varepsilon_u \varepsilon_v \text{tr}\{(d_{w,t} + d_{t,w})'(d_{w,s} + d_{s,w})\}
\end{aligned}$$

$$= O(T^{-3}) \sum_{u,v=k+1}^T \varepsilon_u \varepsilon_v \mathbb{I}_{\#\{u,v,\tau\}=3} \sum_{w=k+1}^T \mathbb{I}_{\#\{t,s,w\}=3} \\ \times \text{tr}\{(d_{w,t} + d_{t,w})'(d_{w,s} + d_{s,w})\}$$

uniformly in t, s because $\mathbb{P}^\dagger(\pi(t) = u, \pi(s) = v, \pi(w) = \tau) = O(T^{-3}) \mathbb{I}_{\#\{u,v,\tau\}=3}$ for $\#\{t, s, w\} = 3$. Hence,

$$\left| \sum_{t,s=k+1}^{T-1} \mathbb{E}^\dagger[\mathbb{I}_{\pi(t) \neq \tau \neq \pi(s), t \neq s} \varepsilon_{\pi(t)} \varepsilon_{\pi(s)} \text{tr}\{(d_{\pi^{-1}(\tau),t} + d_{t,\pi^{-1}(\tau)})'(d_{\pi^{-1}(\tau),s} + d_{s,\pi^{-1}(\tau)})\}] \right|$$

$$\leq O(T^{-3}) \left[\left(\sum_{u=k+1}^T \varepsilon_u \right)^2 + \sigma_T^2 \right] \sum_{w=k+1}^T \left| \sum_{t,s=k+1}^{T-1} \mathbb{I}_{\#\{t,s,w\}=3} \right. \\ \left. \times \text{tr}\{(d_{w,t} + d_{t,w})'(d_{w,s} + d_{s,w})\} \right|$$

is $O_P(T^{-2}k^4 a_T^2 l_T) = o_P(a_T^2)$ for $k^3/T \rightarrow 0$, since

$$\left| \sum_{t,s=k+1}^{T-1} \mathbb{I}_{\#\{t,s,w\}=3} \text{tr}\{(d_{w,t} + d_{t,w})'(d_{w,s} + d_{s,w})\} \right| \\ \leq 2^4 \|\gamma\|^2 (k^2 \|\gamma\|_1 + k \sum_{i=1}^{\infty} i |\gamma_i|)^2 = O(k^4).$$

Therefore, $\mathbb{E}^\dagger \|\Delta_\lambda\|^2 = o_P(a_T^2)$, and $\|\Delta_\lambda\| = o_{P^\dagger}(a_T)$ in P -probability, also for the permutation bootstrap.

For π equal to the identity (wild bootstrap), it holds that

$$(S.4.19) \quad \mathbb{E}^\dagger \|\Delta_\lambda\|^2 = \sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \varepsilon_t^2 \|d_{\tau,t} + d_{t,\tau}\|^2,$$

where

$$(S.4.20) \quad \begin{aligned} & \mathbb{E} \sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \mathbb{I}_{|\varepsilon_t| \leq a_T} \varepsilon_t^2 \|d_{\tau,t} + d_{t,\tau}\|^2 \\ &= \mathbb{E} \mathbb{E} \left(\sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \mathbb{I}_{|\varepsilon_t| \leq a_T} \varepsilon_t^2 \|d_{\tau,t} + d_{t,\tau}\|^2 | \tau \right) \\ &= \mathbb{E} \sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \mathbb{E}(\mathbb{I}_{|\varepsilon_t| \leq a_T} \varepsilon_t^2 | \tau) \|d_{\tau,t} + d_{t,\tau}\|^2 \\ &\leq \mathbb{E}(\mathbb{I}_{|\varepsilon_1| \leq a_T} \varepsilon_1^2) \mathbb{E} \sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \|d_{\tau,t} + d_{t,\tau}\|^2 \end{aligned}$$

is $O(T^{-1}k^3a_T^2) = o(a_T^2)$ because $\mathbb{I}_{\tau \neq t} \mathbb{E}(\mathbb{I}_{|\varepsilon_t| \leq a_T} \varepsilon_t^2 | \tau) \leq \mathbb{I}_{\tau \neq t} \mathbb{E}(\mathbb{I}_{|\varepsilon_t| \leq a_T} \varepsilon_t^2) = \mathbb{I}_{\tau \neq t} \mathbb{E}(\mathbb{I}_{|\varepsilon_1| \leq a_T} \varepsilon_1^2)$ a.s. by the maximizing property of τ , $\mathbb{E}(\mathbb{I}_{|\varepsilon_t| \leq a_T} \varepsilon_t^2) = O(T^{-1}a_T^2)$ by KT,

$$\sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \|d_{\tau,t} + d_{t,\tau}\|^2 \leq 4k\|\gamma\|^2(k^2\|\gamma\|^2 + k \sum_{i=1}^{\infty} i\gamma_i^2) = O(k^3)$$

with a deterministic bound (see equation (S.4.18)), and $k^3/T \rightarrow 0$; similarly,

$$\begin{aligned} & \mathbb{E} \left(\sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \mathbb{I}_{|\varepsilon_t| > a_T} \varepsilon_t^2 \|d_{\tau,t} + d_{t,\tau}\|^2 \right)^{\eta/2} \\ & \leq \mathbb{E}(\mathbb{I}_{|\varepsilon_1| > a_T} |\varepsilon_1|^\eta) \mathbb{E} \sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \|d_{\tau,t} + d_{t,\tau}\|^\eta \end{aligned}$$

is $O(T^{-1}k^2a_T^\eta) = o(a_T^\eta)$ for $\eta \in [\delta, \alpha]$ and δ from Assumption 1(b), by KT and since, with a deterministic upper bound and $k^2/T \rightarrow 0$,

$$\sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \|d_{\tau,t} + d_{t,\tau}\|^\eta \leq O(1) (k^2 \sum_{i=0}^{\infty} |\gamma_i|^\eta + k \sum_{i=1}^{\infty} i|\gamma_i|^\eta) = O(k^2),$$

so $\sum_{t=k+1}^{T-1} \mathbb{I}_{\tau \neq t} \mathbb{I}_{|\varepsilon_t| > a_T} \varepsilon_t^2 \|d_{\tau,t} + d_{t,\tau}\|^2 = o_P(a_T^2)$. Recalling also (S.4.19) and (S.4.20), it follows that $\|\Delta_\lambda\| = o_{P^\dagger}(a_T)$ in P -probability also for the wild bootstrap.

By Weyl's inequality (Seber, 2008, p.117), the estimate of $\|\Delta_\lambda\|$ yields

$$\begin{aligned} (S.4.21) \quad & |\lambda_{\min} \left(\varepsilon_\tau^2 \sum_{j=0}^{T-\pi^{-1}(\tau)-1} \gamma_{j:k} \gamma'_{j:k} + \varepsilon_\tau w_{\pi^{-1}(\tau)} \Delta_\lambda \right) \\ & - \varepsilon_\tau^2 \lambda_{\min} \left(\sum_{j=0}^{T-\pi^{-1}(\tau)-1} \gamma_{j:k} \gamma'_{j:k} \right)| \leq |\varepsilon_\tau| \|\Delta_\lambda\| = o_{P^\dagger}(a_T^2) \end{aligned}$$

in P -probability, because $a_T^{-1}\varepsilon_\tau$ converges in distribution under P . Again by Weyl's inequality and matrix symmetry,

$$\begin{aligned} & \left| \lambda_{\min} \left(\sum_{j=0}^{T-\pi^{-1}(\tau)-1} \gamma_{j:k} \gamma'_{j:k} \right) - \lambda_{\min}(\Sigma_k) \right| \leq \left\| \sum_{j=T-\pi^{-1}(\tau)}^{\infty} \gamma_{j:k} \gamma'_{j:k} \right\|_2 \\ & \leq \sum_{j=T-\pi^{-1}(\tau)}^{\infty} \|\gamma_{j:k} \gamma'_{j:k}\|_1 \leq k \left(\max_i |\gamma_i| \right) \sum_{j=\max\{0, T-\pi^{-1}(\tau)-k+1\}}^{\infty} |\gamma_j| \end{aligned}$$

is $o_{P^\dagger}(1)$ in P -probability, the magnitude order because $P^\dagger(T - \pi^{-1}(\tau) - k + 1 \geq k) \xrightarrow{P} 1$ and $k \sum_{j=k}^{\infty} |\gamma_j| \leq \sum_{j=k}^{\infty} j |\gamma_j| \rightarrow 0$ as $k \rightarrow \infty$ as the tail of a convergent series. Since $\lambda_{\min}(\Sigma_k)$ is bounded away from zero, it follows that $\lambda_{\min}(\sum_{j=0}^{T-\pi^{-1}(\tau)-1} \gamma_{j:k} \gamma'_{j:k})$ is bounded away from zero in P^\dagger -probability, and as further $a_T^{-1} \varepsilon_\tau$ converges weakly under P to a distribution with no atom at zero, it follows that $\lambda_{\min}(a_T^{-2} \varepsilon_\tau^2 \sum_{j=0}^{T-\pi^{-1}(\tau)-1} \gamma_{j:k} \gamma'_{j:k})$ is bounded away from zero in P^\dagger , then P , probability. Recalling (S.4.17) and (S.4.21), we can conclude that also $\lambda_{\min}(a_T^{-2} S_{00}^{\dagger k})$ is bounded away from zero in P^\dagger , then P , probability.

S.4.1.3. Proof of part (b). Write $S_{0\varepsilon}^{*k} - S_{0\varepsilon}^{\dagger k} = \sigma_1 + \sigma_2^* - \sigma_2^\dagger + \sigma_3$ with $\sigma_2^* := \sum_{t=k+1}^T \mathbf{X}_{t-1}^{*k} \rho_{\pi(t),k} w_t$, $\sigma_2^\dagger := \sum_{t=k+1}^T \mathbf{X}_{t-1}^{\dagger k} \rho_{\pi(t),k}^\dagger$ and

$$\begin{aligned} \sigma_3 &:= \sum_{t=k+1}^T (\mathbf{X}_{t-1}^{*k} - \mathbf{X}_{t-1}^{\dagger k}) \varepsilon_{\pi(t)} w_t \\ &= \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} (\hat{\gamma}_{j:k} - \gamma_{j:k}) \hat{\varepsilon}_{\pi(t-j-1)} \varepsilon_{\pi(t)} w_{t-j-1} w_t \\ &\quad + \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \gamma_{j:k} (\hat{\varepsilon}_{\pi(t-j-1)} - \varepsilon_{\pi(t-j-1)}) \varepsilon_{\pi(t)} w_{t-j-1} w_t. \end{aligned}$$

If π is an r.p., we discuss $\sigma_1 + \sigma_2^*$, σ_2^\dagger and σ_3 , no matter how w_t are specified, whereas if π is the identity, we evaluate σ_2^* , σ_2^\dagger and σ_3 .

Let π be the identity (and w_t be Rademacher). For the discussion of σ_2^* , define modified $\rho_{t,kj} := \rho_{t,k} - \varepsilon_{t-j-1} \sum_{i=k+1}^{j+1} \beta_i \gamma_{j+1-i}$, ($j = k, \dots, T - k - 2$), as $\rho_{t,k}$ 'cleaned' from the contribution of ε_{t-j-1} , and $\rho_{t,kj} := \rho_{t,k}$ for $j = 0, \dots, k - 1$. It holds that

$$\begin{aligned} \mathbb{E}^\dagger \|\sigma_2^*\|^2 &= \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k}\|^2 \hat{\varepsilon}_{t-j-1}^2 \rho_{t,k}^2 \\ &\leq 4 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} (\|\gamma_{j:k}\|^2 + \|\hat{\gamma}_{j:k} - \gamma_{j:k}\|^2) (\varepsilon_{t-j-1}^2 + (\hat{\varepsilon}_{t-j-1} - \varepsilon_{t-j-1})^2) \rho_{t,k}^2 \\ &\leq 8 \left(\sum_{t=k+2}^T \sum_{j=k}^{t-k-2} \|\gamma_{j:k}\| \varepsilon_{t-j-1}^2 \left| \sum_{i=k+1}^{j+1} \beta_i \gamma_{j+1-i} \right|^2 \right. \\ &\quad \left. + 8 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \varepsilon_{t-j-1}^2 \rho_{t,kj}^2 + 4 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k} - \gamma_{j:k}\|^2 \varepsilon_{t-j-1}^2 \rho_{t,k}^2 \right) \end{aligned}$$

$$\begin{aligned}
& + 4 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} (\|\boldsymbol{\gamma}_{j:k}\|^2 + \|\hat{\boldsymbol{\gamma}}_{j:k} - \boldsymbol{\gamma}_{j:k}\|^2) (\hat{\varepsilon}_{t-j-1} - \varepsilon_{t-j-1})^2 \rho_{t,k}^2 \\
& =: 8(\varsigma^2 + \sigma_{21}^*) + 4(\sigma_{22}^* + \sigma_{23}^*).
\end{aligned}$$

Here

$$\begin{aligned}
\varsigma &= \sum_{t=k+2}^T \sum_{j=k}^{t-k-2} \|\boldsymbol{\gamma}_{j:k}\| \varepsilon_{t-j-1}^2 (\mathbb{I}_{\{|\varepsilon_{t-j-1}| \leq a_T\}} + \mathbb{I}_{\{|\varepsilon_{t-j-1}| > a_T\}}) \sum_{i=k+1}^{j+1} |\beta_i \gamma_{j+1-i}| \\
&:= \varsigma^{\leq} + \varsigma^> = o_P(a_T)
\end{aligned}$$

by Markov's inequality, since

$$\begin{aligned}
\mathbb{E}(\varsigma^{\leq}) &\leq T \mathbb{E}(\varepsilon_1^2 \mathbb{I}_{\{|\varepsilon_1| \leq a_T\}}) \|\boldsymbol{\gamma}\| \sum_{j=k}^{T-k-2} \|\boldsymbol{\gamma}_{j:k}\| \sum_{i=k+1}^{\infty} |\beta_i| \\
&\leq T \mathbb{E}(\varepsilon_1^2 \mathbb{I}_{\{|\varepsilon_1| \leq a_T\}}) \|\boldsymbol{\gamma}\| (\sum_{i=1}^{\infty} i |\gamma_i| + k \sum_{i=k+1}^{\infty} |\gamma_i|) \sum_{i=k+1}^{\infty} |\beta_i| = o_P(a_T)
\end{aligned}$$

by KT, Assumption 1(b) and the condition $\sum_{i=k+1}^{\infty} |\beta_i| = o(a_T^{-1})$ and, for $\eta \in [\delta, \alpha)$, by using the same facts,

$$\begin{aligned}
\mathbb{E} |\varsigma^>|^{\frac{\eta}{2}} &\leq T \mathbb{E}(|\varepsilon_1|^{\eta} \mathbb{I}_{\{|\varepsilon_1| > a_T\}}) \|\boldsymbol{\gamma}\|^{\frac{\eta}{2}} \sum_{j=k}^{T-k-2} \|\boldsymbol{\gamma}_{j:k}\|^{\frac{\eta}{2}} (\sum_{i=k+1}^{\infty} |\beta_i|)^{\frac{\eta}{2}} \\
&\leq T \mathbb{E}(|\varepsilon_1|^{\eta} \mathbb{I}_{\{|\varepsilon_1| > a_T\}}) \|\boldsymbol{\gamma}\|^{\frac{\eta}{2}} (\sum_{i=1}^{\infty} i |\gamma_i|^{\frac{\eta}{2}} + k \sum_{i=k+1}^{\infty} |\gamma_i|^{\frac{\eta}{2}}) (\sum_{i=k+1}^{\infty} |\beta_i|)^{\frac{\eta}{2}}
\end{aligned}$$

is $o_P(a_T^{\eta/2})$. Further, for $\eta \in [\delta, \alpha)$ and all t , by independence of ε_{t-j-1} and $\rho_{t,kj}$,

$$\mathbb{E} (\sigma_{21}^*)^{\frac{\eta}{2}} \leq \mathbb{E} |\varepsilon_1|^{\eta} (\sup_{t,j} \mathbb{E} |\rho_{t,kj}|^{\eta}) \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\boldsymbol{\gamma}_{j:k}\|^{\eta},$$

where the sup is over $t = k+2, \dots, T$ and $j = 0, \dots, t-k-2$. By Hölder's inequality, for $\alpha > 1$ and $\eta \in [1, \alpha)$ it holds that

$$\begin{aligned}
\mathbb{E} |\rho_{t,kj}|^{\eta} &= \mathbb{E} |\rho_{k+1,k}|^{\eta} \leq (\sum_{i=k+1}^{\infty} |\beta_i|)^{\eta-1} \sum_{i=k+1}^{\infty} |\beta_i| \mathbb{E} |X_{k+1-i}|^{\eta} \\
&= (\sum_{i=k+1}^{\infty} |\beta_i|)^{\eta} \mathbb{E} |X_1|^{\eta} = o(a_T^{-\eta})
\end{aligned}$$

if $j = 0, \dots, k - 1$, and similarly,

$$\begin{aligned} \mathbb{E} |\rho_{t,kj}|^\eta &\leq (\sum_{i=k+1}^{\infty} |\beta_i|)^{\eta-1} \sum_{i=k+1}^{\infty} |\beta_i| \mathbb{E} |\sum_{l=0}^{\infty} \mathbb{I}_{l \neq j+1} \gamma_l \varepsilon_{t-l}|^\eta \\ &\leq (\sum_{i=k+1}^{\infty} |\beta_i|)^\eta \|\boldsymbol{\gamma}\|^\eta \mathbb{E} |\varepsilon_1|^\eta = o(a_T^{-\eta}) \end{aligned}$$

for $j = k, \dots, t - k - 2$. Hence, a common, in t and j , $o(a_T^{-\eta})$ upper bound exists for $\mathbb{E} |\rho_{t,kj}|^\eta$, yielding

$$\mathbb{E} (\sigma_{21}^*)^{\frac{\eta}{2}} \leq o(kTa_T^{-\eta}) (\sum_{j=0}^{\infty} |\gamma_j|^\eta) = o(kTa_T^{-\eta}), \quad \eta \in [1, \alpha].$$

As $k^2/T \rightarrow 0$, this yields $\sigma_{21}^* = o_P(a_T^2)$. On the other hand, for $\alpha \leq 1$ and $\eta \in [\delta, \alpha]$ it holds that $\mathbb{E} |\rho_{t,kj}|^\eta = \mathbb{E} |\rho_{k+1,k}|^\eta \leq \mathbb{E} |\varepsilon_1|^\eta \sum_{i=k+1}^{\infty} |\beta_i|^\eta$ if $j = 0, \dots, k - 1$, and

$$\mathbb{E} |\rho_{t,kj}|^\eta \leq \sum_{i=k+1}^{\infty} |\beta_i|^\eta \mathbb{E} |\sum_{l=0}^{\infty} \mathbb{I}_{l \neq j+1} \gamma_l \varepsilon_{t-l}|^\eta \leq \mathbb{E} |\varepsilon_1|^\eta \sum_{i=0}^{\infty} |\gamma_i|^\eta \sum_{i=k+1}^{\infty} |\beta_i|^\eta$$

if $j = k, \dots, t - k - 2$, so $\mathbb{E} (\sigma_{21}^*)^{\eta/2} = O(kT \sum_{i=k+1}^{\infty} |\beta_i|^\eta) \sum_{j=k+1}^{\infty} |\gamma_j|^\eta$ and, by (S.2.2) and Markov's inequality, $\sigma_{21}^* = o_P(\tilde{a}_T^2)$ under $k^3/T \rightarrow 0$.

Next,

$$\sigma_{22}^* \leq \sigma_T^2 S_{\rho\rho}^k \sum_{j=0}^{T-k-2} \|\hat{\boldsymbol{\gamma}}_{j:k} - \boldsymbol{\gamma}_{j:k}\|^2 = o_P(l_T a_T^2) \sum_{j=0}^{T-k-2} \sum_{i=0}^{k-1} |\hat{\gamma}_{j-i} - \gamma_{j-i}|^2$$

since $S_{\rho\rho}^k = o_P(l_T)$ was proved in the preparation, so from (S.4.1) and (7.1), for all $\epsilon > 0$,

$$\begin{aligned} \sigma_{22}^* &= o_P(kl_T a_T^2) \{ [\|\hat{\boldsymbol{\beta}}_k - \boldsymbol{\beta}_k\|_1^2 + (\sum_{j=k+1}^{\infty} |\beta_j|)^2] \sum_{j=0}^{\infty} \left(1 + \frac{1}{k}\right)^{-2j} + \sum_{j=0}^{\infty} b_{jk}^2 \} \\ &= o_P(kl_T a_T^2) \{ k^2 a_k^2 a_T^{\epsilon-2} + (\sum_{j=k+1}^{\infty} |\beta_j|)^2 \} = o_P(k^3 a_k^2 T^\epsilon) = o_P(\tilde{a}_T^2) \end{aligned}$$

if $k^4/T \rightarrow 0$ by choosing small $\epsilon > 0$, and similarly, from (S.4.5) and (7.1),

$$\begin{aligned} \sigma_{23}^* &\leq \|\hat{\boldsymbol{\varepsilon}}_T - \boldsymbol{\varepsilon}_T\|^2 S_{\rho\rho}^k \sum_{j=0}^{T-k-2} (\|\boldsymbol{\gamma}_{j:k}\|^2 + \|\hat{\boldsymbol{\gamma}}_{j:k} - \boldsymbol{\gamma}_{j:k}\|^2) \\ &= o_P(kl_T a_T^2 \|\hat{\boldsymbol{\beta}}_k - \boldsymbol{\beta}_k\|^2) (\|\boldsymbol{\gamma}\|^2 + o_P(1)) = o_P(\tilde{a}_T^2) \end{aligned}$$

if $k^2/T \rightarrow 0$. We conclude that $\mathbb{E}^\dagger \|\sigma_2^*\|^2 = o_P(\tilde{a}_T^2)$ if $k^4/T \rightarrow 0$ and, hence, $\|\sigma_2^*\| = o_{P^\dagger}(\tilde{a}_T)$ in P -probability if π is the identity.

Regarding $\sigma_2^\dagger = \sum_{t=k+1}^T \mathbf{X}_{t-1}^{\dagger k} \rho_{t,k}^\dagger$ ($\rho_{t,k}^\dagger = \sum_{i=k+1}^{t-k-1} \beta_i X_{t-i}^\dagger$), we reuse several steps of the evaluation of $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \rho_{t,k}\|$ in the proof of Lemma 2. Namely, all the evaluations of expressions in $|\varepsilon_t|$ and ε_t^2 (equal to $|\varepsilon_t^\dagger|$ and $(\varepsilon_t^\dagger)^2$, resp.) can be used as there, upon replacement of ε_t ($t \leq k$) by zeroes. A minor modification is needed only for (what is now)

$$\begin{aligned} \xi_{ij}^{\leq, \dagger} := & \sum_{t=k+1+\max\{i,j\}}^{T-1} \sum_{u=0}^{t-i-k-1} \sum_{v=0}^{t-j-k-1} \mathbb{I}_{\{u \neq v+j-i\}} \\ & \times \mathbb{I}_{|\varepsilon_{t-i-u} \varepsilon_{t-j-v}| \leq \tilde{a}_T} \varepsilon_{t-i-u}^\dagger \varepsilon_{t-j-v}^\dagger \gamma_u \gamma_v \end{aligned}$$

with

$$\begin{aligned} \mathbb{E}^\dagger (\xi_{ij}^{\leq, \dagger})^2 \leq & 2 \|\gamma\|^2 \sum_{t=k+1+\max\{i,j\}}^{T-1} \sum_{u=0}^{t-i-k-1} \sum_{v=0}^{t-j-k-1} \mathbb{I}_{\{u \neq v+j-i\}} \\ & \times \mathbb{I}_{|\varepsilon_{t-i-u} \varepsilon_{t-j-v}| \leq \tilde{a}_T} \varepsilon_{t-i-u}^2 \varepsilon_{t-j-v}^2 |\gamma_u| |\gamma_v| \end{aligned}$$

possessing, by KT, $\mathbb{E} \mathbb{E}^\dagger (\xi_{ij}^{\leq, \dagger})^2 \leq O(\tilde{a}_T^2) \|\gamma\|^4$ uniformly in i, j , so in place of (S.3.7),

$$\mathbb{E} \mathbb{E}^\dagger \left(\sum_{i=1}^k \left(\sum_{j=k+1}^{\infty} \beta_j \xi_{ij}^{\leq, \dagger} \right)^2 \right) \leq \left[\sum_{i=1}^k \sum_{j=k+1}^{\infty} |\beta_j| \mathbb{E} \mathbb{E}^\dagger \{ (\xi_{ij}^{\leq, \dagger})^2 \} \right] \left(\sum_{j=k+1}^{\infty} |\beta_j| \right)$$

is $O(k \tilde{a}_T^2) (\sum_{j=k+1}^{\infty} |\beta_j|)^2$. Thus, as in the discussion of $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^k \rho_{t,k}\|$ in Lemma 2, we conclude that $\|\sum_{t=k+1}^T \mathbf{X}_{t-1}^{\dagger k} \rho_{t,k}^\dagger\| = O_{P^\dagger}(a_T^2) \sum_{j=k+1}^{\infty} |\beta_j| = o_{P^\dagger}(a_T)$ in P -probability.

Further,

$$\begin{aligned} \mathbb{E}^\dagger \|\sigma_3\|^2 \leq & 4 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k} - \gamma_{j:k}\|^2 \underbrace{\{\varepsilon_{t-j-1,k}^2 + (\hat{\varepsilon}_{t-j-1} - \varepsilon_{t-j-1,k})^2\} \varepsilon_t^2}_{=: e_{tj}} \\ & + 4 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \{(\hat{\varepsilon}_{t-j-1} - \varepsilon_{t-j-1,k})^2 + \rho_{t-j-1,k}^2\} \varepsilon_t^2 := 4(\sigma_{31} + \sigma_{32} + \sigma_{33}). \end{aligned}$$

First,

$$\sigma_{31} = \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k} - \gamma_{j:k}\|^2 e_{tj} = \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{k-1} (\hat{\gamma}_{j-i} - \gamma_{j-i})^2 e_{tj}$$

$$\begin{aligned} &\leq kO_P(\|\hat{\beta}_k - \beta_k\|_1^2 + (\sum_{j=k}^{\infty} |\beta_j|)^2) \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} e_{tj} \\ &\quad + 2 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{(k-1)\wedge j} b_{j-i,k}^2 e_{tj} \end{aligned}$$

contains, (i),

$$\begin{aligned} \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} e_{tj} &\leq 2 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} \{\varepsilon_{t-j-1}^2 + \rho_{t-j-1,k}^2\} \varepsilon_t^2 \\ &\quad + 2\|\hat{\beta}_k - \beta_k\|^2 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} \|\mathbf{X}_{t-j-2}^k\|^2 \varepsilon_t^2 \end{aligned}$$

is $O_P(a_k^{2+\epsilon} \tilde{a}_T^2)$ for all $\epsilon > 0$ since, (i.i), $\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} (1 + \frac{1}{k})^{-2j} \varepsilon_{t-j-1}^2 \varepsilon_t^2 = O_P(a_k^{2+\epsilon} \tilde{a}_T^2)$ for all $\epsilon > 0$, as

$$\begin{aligned} &\mathbb{E} \left(\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} \varepsilon_{t-j-1}^2 \varepsilon_t^2 \mathbb{I}_{\{|\varepsilon_{t-j-1} \varepsilon_t| \leq \tilde{a}_T\}} \right) \\ &\leq T \mathbb{E}(\varepsilon_1^2 \varepsilon_2^2 \mathbb{I}_{\{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T\}}) \sum_{j=0}^{\infty} \left(1 + \frac{1}{k}\right)^{-2j} = O(k \tilde{a}_T^2), \\ &\mathbb{E} \left(\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} \varepsilon_{t-j-1}^2 \varepsilon_t^2 \mathbb{I}_{\{|\varepsilon_{t-j-1} \varepsilon_t| > \tilde{a}_T\}} \right)^{\frac{\eta}{2}} \\ &\leq T \mathbb{E}(|\varepsilon_1 \varepsilon_2|^{\eta} \mathbb{I}_{\{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T\}}) \sum_{j=0}^{\infty} \left(1 + \frac{1}{k}\right)^{-j\eta} = O(k \tilde{a}_T^{\eta}) \end{aligned}$$

by KT for all $\eta \in (0, \alpha)$, (i.ii), similarly,

$$\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} \|\mathbf{X}_{t-j-2}^k\|^2 \varepsilon_t^2 = O_P(a_k^{4+\epsilon} \tilde{a}_T^2)$$

for all $\epsilon > 0$, and is multiplied by $\|\hat{\beta}_k - \beta_k\|^2 = O_P(a_k^2 a_T^{\epsilon-2}) = O_P(a_k^{-2})$ for sufficiently small $\epsilon > 0$ by (7.1) and $k^3/T \rightarrow 0$, (i.iii),

$$\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} \rho_{t-j-1,k}^2 \varepsilon_t^2 = o_P(\tilde{a}_T^2)$$

since, for $\eta \in [\delta, \alpha]$,

$$\mathbb{E} \left(\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \left(1 + \frac{1}{k}\right)^{-2j} \rho_{t-j-1,k}^2 \varepsilon_t^2 \right)^{\frac{\eta}{2}} \leq$$

$$\leq T \mathbb{E} |\varepsilon_1|^\eta \mathbb{E} |\rho_{k+1,k}|^\eta \sum_{j=0}^{\infty} \left(1 + \frac{1}{k}\right)^{-j\eta} = O(kT) \mathbb{E} |\rho_{k+1,k}|^\eta$$

is $O_P(kTa_T^{-\eta})$ as $\mathbb{E} |\rho_{k+1,k}|^\eta$ was evaluated in the discussion of σ_{21}^* and $k^2/T \rightarrow 0$, and (ii),

$$\begin{aligned} & \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{(k-1)\wedge j} b_{j-i,k}^2 e_{tj} \\ & \leq 2 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{(k-1)\wedge j} b_{j-i,k}^2 \{\varepsilon_{t-j-1}^2 + \rho_{t-j-1,k}^2\} \varepsilon_t^2 \\ & \quad + 2\|\hat{\beta}_k - \beta_k\|^2 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{(k-1)\wedge j} b_{j-i,k}^2 \|\mathbf{X}_{t-j-2}^k\|^2 \varepsilon_t^2 \\ & = O_P(kT^{-1}\tilde{a}_T^{4+\epsilon} + \|\hat{\beta}_k - \beta_k\|^2 a_k^2 kT^{-1}\tilde{a}_T^{4+\epsilon})(\sum_{i=k+1}^{\infty} |\beta_i|)^2 = o_P(a_T^2) \end{aligned}$$

for all $\epsilon > 0$, by taking expectations as in (i) and using $\sum_{j=0}^{t-k-2} \sum_{i=0}^{k-1} b_{j-i,k}^2 \leq k \sum_{j=0}^{t-k-2} b_{jk}^2 \leq k(\sum_{j=0}^{\infty} b_{jk})^2 \leq Ck(\sum_{i=k+1}^{\infty} |\beta_i|)^2$, (7.1) and the conditions $k^3/T \rightarrow 0$ and $a_T \sum_{i=k+1}^{\infty} |\beta_i| \rightarrow 0$; for example,

$$\begin{aligned} & \mathbb{E} \left(\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{k-1} b_{j-i,k}^2 \varepsilon_{t-j-1}^2 \varepsilon_t^2 \mathbb{I}_{\{|\varepsilon_{t-j-1}\varepsilon_t| \leq \tilde{a}_T\}} \right) \\ & \leq T \mathbb{E} (\varepsilon_1^2 \varepsilon_2^2 \mathbb{I}_{\{|\varepsilon_1\varepsilon_2| \leq \tilde{a}_T\}}) \sum_{j=0}^{T-k-2} \sum_{i=0}^{k-1} b_{j-i,k}^2 = O(k\tilde{a}_T^2)(\sum_{i=k+1}^{\infty} |\beta_i|)^2 \end{aligned}$$

and, by Hölder's inequality for $\eta \in (0, \alpha)$,

$$\begin{aligned} & \mathbb{E} \left(\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{k-1} b_{j-i,k}^2 \varepsilon_{t-j-1}^2 \varepsilon_t^2 \mathbb{I}_{\{|\varepsilon_{t-j-1}\varepsilon_t| > \tilde{a}_T\}} \right)^{\frac{\eta}{2}} \\ & \leq T \mathbb{E} (|\varepsilon_1\varepsilon_2|^\eta \mathbb{I}_{\{|\varepsilon_1\varepsilon_2| > \tilde{a}_T\}}) \times \sum_{j=0}^{T-k-2} \left(\sum_{i=0}^{k-1} b_{j-i,k}^2 \right)^{\frac{\eta}{2}} \\ & \leq O(\tilde{a}_T^\eta) T^{1-\frac{\eta}{2}} \left(\sum_{j=0}^{T-k-2} \sum_{i=0}^{k-1} b_{j-i,k}^2 \right)^{\frac{\eta}{2}} = \{O(kT^{\frac{2}{\eta}-1}\tilde{a}_T^2)(\sum_{i=k+1}^{\infty} |\beta_i|)^2\}^{\frac{\eta}{2}}. \end{aligned}$$

By combining (i), (ii) and (7.1), for all $\epsilon > 0$,

$$\sigma_{31} = O_P(ka_k^{2+\epsilon}\tilde{a}_T^2)(\|\hat{\beta}_k - \beta_k\|_1^2 + (\sum_{j=k}^{\infty} |\beta_j|)^2) + 2 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \sum_{i=0}^{k-1} b_{j-i,k}^2 e_{tj}$$

is $O_P(k^2 a_k^4 a_T^\epsilon) + o_P(\tilde{a}_T^2) = o_P(\tilde{a}_T^2)$ since $k^4/T \rightarrow 0$ and $a_T \sum_{i=k+1}^\infty |\beta_i| \rightarrow 0$. Second,

$$\sigma_{32} \leq \|\hat{\beta}_k - \beta_k\|^2 \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \|\mathbf{X}_{t-j-2}^k\|^2 \varepsilon_t^2 = \|\hat{\beta}_k - \beta_k\|^2 O_P(a_k^{4+\epsilon} \tilde{a}_T^2)$$

is $O_P(a_k^6 a_T^\epsilon) = o_P(\tilde{a}_T^2)$ using (7.1) and $k^4/T \rightarrow 0$, again by taking expectations and replacing the geometric series in powers of $1 + \frac{1}{k}$ by

$$\sum_{j=0}^{T-k-2} \|\gamma_{j:k}\|^2 \leq k \sum_{j=0}^\infty \gamma_j^2 = O(k), \quad \sum_{j=0}^{T-k-2} \|\gamma_{j:k}\|^\eta \leq k \sum_{j=0}^\infty |\gamma_j|^\eta = O(k)$$

for $\eta \geq \delta$. Third, $\sigma_{33} := \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \rho_{t-j-1,k}^2 \varepsilon_t^2 = o_P(\tilde{a}_T^2)$, similarly to σ_{21}^* .

Returning to the initial decomposition of $E^\dagger \|\sigma_3\|^2$, we can conclude that $E^\dagger \|\sigma_3\|^2 = o_P(\tilde{a}_T^2)$ for $k^4/T \rightarrow 0$. By combining it with the evaluations of σ_2^* and σ_2^\dagger , we complete the proof of part (b) for π equal to the identity.

For an r.p. π , we start from

$$\sigma_1 + \sigma_2^* = \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \hat{\gamma}_{j:k} \hat{\varepsilon}_{\pi(t-j-1)} (\hat{\varepsilon}_{\pi(t)} - \varepsilon_{\pi(t)}) w_{t-j-1} w_t.$$

If w_t are Rademacher, then

$$\begin{aligned} E^\dagger \|\sigma_1 + \sigma_2^*\|^2 &= \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k}\|^2 E^\dagger \{\hat{\varepsilon}_{\pi(t-j-1)}^2 (\hat{\varepsilon}_{\pi(t)} - \varepsilon_{\pi(t)})^2\} \\ &= E^\dagger \{\hat{\varepsilon}_{\pi(k+1)}^2 (\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+2)})^2\} \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k}\|^2, \end{aligned}$$

with $E^\dagger \{\hat{\varepsilon}_{\pi(k+1)}^2 (\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+2)})^2\} \leq O(T^{-2}) \hat{\sigma}_{Tk}^2 \|\hat{\varepsilon}_T - \varepsilon_T\|^2 = O_P(T^{\epsilon-2} a_k^2 a_T^2)$, so

$$\begin{aligned} E^\dagger \|\sigma_1 + \sigma_2^*\|^2 &= O_P(T^{\epsilon-1} a_k^2 a_T^2) \sum_{j=0}^{T-k-2} \|\hat{\gamma}_{j:k}\|^2 \\ &= O_P(T^{\epsilon-1} k a_k^2 a_T^2) (\|\gamma\|^2 + \|\hat{\gamma}_{T+k-2} - \gamma_{T+k-2}\|^2) \end{aligned}$$

is $O_P(T^{\epsilon-1} k a_k^2 a_T^2)$ for all $\epsilon > 0$, using equation (S.4.5) and (7.1). On the other hand, if $w_t = 1$ P^\dagger -a.s. (all t), then

$$E^\dagger \|\sigma_1 + \sigma_2^*\|^2 = \sum_{s,t=k+2}^T \sum_{i=0}^{s-k-2} \sum_{j=0}^{t-k-2} \hat{\gamma}'_{i:k} \hat{\gamma}_{j:k} \times$$

$$\begin{aligned}
& \times E^\dagger \{ \hat{\varepsilon}_{\pi(s-i-1)} \hat{\varepsilon}_{\pi(t-j-1)} (\hat{\varepsilon}_{\pi(s)} - \varepsilon_{\pi(s)}) (\hat{\varepsilon}_{\pi(t)} - \varepsilon_{\pi(t)}) \} \\
& = E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)}^2 (\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+2)})^2 \} \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k}\|^2 \\
& + E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} (\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T)})^2 \} \sum_{t=k+2}^T \sum_{i,j=0}^{t-k-2} \mathbb{I}_{\{i \neq j\}} \hat{\gamma}'_{i:k} \hat{\gamma}_{j:k} \\
& + E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)}^2 (\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+2)}) (\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T)}) \} \\
& \times \sum_{s,t=k+2}^T \sum_{i=\max\{0,s-t\}}^{s-k-2} \hat{\gamma}'_{i:k} \hat{\gamma}_{t-s+i:k} \mathbb{I}_{\#\{s,t,s-i-1\}=3} \\
& + E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} (\hat{\varepsilon}_{\pi(k+3)} - \varepsilon_{\pi(k+3)}) (\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T)}) \} \\
& \times \sum_{s,t=k+2}^T \sum_{i=0}^{s-k-2} \sum_{j=0}^{t-k-2} \mathbb{I}_{\#\{s-i-1,t-j-1,s,t\}=4} \hat{\gamma}'_{i:k} \hat{\gamma}_{j:k}
\end{aligned}$$

by separating according to the possible subscript repetitions. Here, first, $E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)}^2 (\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+2)})^2 \} = O_P(T^{\epsilon-2} a_k^2 a_T^2)$ for all $\epsilon > 0$, as found earlier. Second,

$$\begin{aligned}
& E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} (\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T)})^2 \} = O(T^{-3}) \sum_{u,v,s=k+1}^T \mathbb{I}_{\#\{u,v,s\}=3} \hat{\varepsilon}_u \hat{\varepsilon}_v (\hat{\varepsilon}_s - \varepsilon_s)^2 \\
& \leq O(T^{-3}) \{ \left(\sum_{u=k+1}^T \hat{\varepsilon}_u \right)^2 \|\hat{\varepsilon}_T - \varepsilon_T\|^2 - 2 \left(\sum_{u=k+1}^T \hat{\varepsilon}_u \right) \sum_{s=k+1}^T \hat{\varepsilon}_s (\hat{\varepsilon}_s - \varepsilon_s)^2 \\
& \quad + \sum_{u=k+1}^T \hat{\varepsilon}_u^2 (\hat{\varepsilon}_u - \varepsilon_u)^2 \} \\
& \leq O(T^{-3}) \{ \left(\sum_{u=k+1}^T \hat{\varepsilon}_u \right)^2 + 2 \hat{\sigma}_{Tk} \left| \sum_{u=k+1}^T \hat{\varepsilon}_u \right| + \hat{\sigma}_{Tk}^2 \} \|\hat{\varepsilon}_T - \varepsilon_T\|^2
\end{aligned}$$

Hence, using $\sum_{t=k+1}^T \varepsilon_t = O_P(a_T l_T)$, equation (S.4.3) and $k^3/T \rightarrow 0$,

$$\begin{aligned}
& E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} (\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T)})^2 \} = O_P(T^{-3}) \{ a_T^2 l_T + k a_k^2 T^\epsilon + k^{1/2} a_k a_T T^\epsilon \} \\
& \quad \times \|\hat{\varepsilon}_T - \varepsilon_T\|^2 = O_P(T^{-3} a_T^2 l_T) \|\hat{\varepsilon}_T - \varepsilon_T\|^2 = O_P(T^{\epsilon-3} a_k^2 a_T^2)
\end{aligned}$$

for all $\epsilon > 0$. Third, with $|\sum_{u=k+1}^T \hat{\varepsilon}_u^2 (\hat{\varepsilon}_u - \varepsilon_u)| \leq \hat{\sigma}_{Tk}^2 \|\hat{\varepsilon}_T - \varepsilon_T\|$, we find

$$\begin{aligned}
& E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)}^2 (\hat{\varepsilon}_{\pi(k+2)} - \varepsilon_{\pi(k+2)}) (\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T)}) \} \\
& = O(T^{-3}) \sum_{u,v,s=k+1}^T \mathbb{I}_{\#\{u,v,s\}=3} \hat{\varepsilon}_u^2 (\hat{\varepsilon}_v - \varepsilon_v) (\hat{\varepsilon}_s - \varepsilon_s)
\end{aligned}$$

$$\begin{aligned}
&\leq O(T^{-3})[\hat{\sigma}_{Tk}^2 \{ \sum_{v=k+1}^T (\hat{\varepsilon}_v - \varepsilon_v) \}^2 \\
&\quad - 2 \sum_{u=k+1}^T \hat{\varepsilon}_u^2 (\hat{\varepsilon}_u - \varepsilon_u) \sum_{u=k+1}^T (\hat{\varepsilon}_u - \varepsilon_u) + \sum_{u=k+1}^T \hat{\varepsilon}_u^2 (\hat{\varepsilon}_u - \varepsilon_u)^2] \\
&= O_P(T^{\epsilon-3} k a_k^2 a_T^2 + T^{\epsilon-3} k^{1/2} a_k a_T^2 \| \hat{\varepsilon}_T - \varepsilon_T \| + T^{-3} \hat{\sigma}_{Tk}^2 \| \hat{\varepsilon}_T - \varepsilon_T \|^2) \\
&= O_P(T^{\epsilon-3} k a_k^2 a_T^2 + T^{\epsilon-3} k^{1/2} a_k^2 a_T + T^{\epsilon-3} a_k^2) = O_P(T^{\epsilon-3} k a_k^2 a_T^2)
\end{aligned}$$

for all $\epsilon > 0$. Fourth,

$$\begin{aligned}
&\mathbb{E}^\dagger \{ \hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} (\hat{\varepsilon}_{\pi(k+3)} - \varepsilon_{\pi(k+3)}) (\hat{\varepsilon}_{\pi(T)} - \varepsilon_{\pi(T)}) \} \\
&= O(T^{-4}) \sum_{u,v,s,t=k+1}^T \mathbb{I}_{\#\{u,v,s,t\}=4} \hat{\varepsilon}_u \hat{\varepsilon}_t (\hat{\varepsilon}_v - \varepsilon_v) (\hat{\varepsilon}_s - \varepsilon_s) \\
&= O(T^{-4}) \{ \sum_{u,v=k+1}^T \mathbb{I}_{u \neq v} \hat{\varepsilon}_u (\hat{\varepsilon}_v - \varepsilon_v) \}^2 - \sum_{u,v,s=k+1}^T \mathbb{I}_{\#\{u,v,s\}=3} \hat{\varepsilon}_u \hat{\varepsilon}_v (\hat{\varepsilon}_s - \varepsilon_s)^2 \\
&\quad - \sum_{u,v,s=k+1}^T \mathbb{I}_{\#\{u,v,s\}=3} \hat{\varepsilon}_u^2 (\hat{\varepsilon}_v - \varepsilon_v) (\hat{\varepsilon}_s - \varepsilon_s) - \sum_{u,v=k+1}^T \mathbb{I}_{u \neq v} \hat{\varepsilon}_u^2 (\hat{\varepsilon}_v - \varepsilon_v)^2 \},
\end{aligned}$$

where the magnitude order of the last three sums was determined above, so

$$\begin{aligned}
&= O(T^{-4}) \{ \sum_{u=k+1}^T \hat{\varepsilon}_u \sum_{v=k+1}^T (\hat{\varepsilon}_v - \varepsilon_v) - \sum_{u=k+1}^T \hat{\varepsilon}_u (\hat{\varepsilon}_u - \varepsilon_u) \}^2 + O_P(T^{\epsilon-4} k a_k^2 a_T^2) \\
&\leq O(T^{-4}) [(\sum_{u=k+1}^T \hat{\varepsilon}_u)^2 \{ \sum_{v=k+1}^T (\hat{\varepsilon}_v - \varepsilon_v) \}^2 + \hat{\sigma}_{Tk}^2 \| \hat{\varepsilon}_T - \varepsilon_T \|^2] + O_P(T^{\epsilon-4} k a_k^2 a_T^2) \\
&= O_P(T^{-4} (a_T^2 l_T + k a_k^2 T^\epsilon) k a_k^2 T^\epsilon) + O_P(T^{\epsilon-4} k a_k^2 a_T^2) = O_P(T^{\epsilon-4} k a_k^2 a_T^2)
\end{aligned}$$

for all $\epsilon > 0$ if $k^3/T \rightarrow 0$. Returning to $\|\sigma_1 + \sigma_2^*\|$,

$$\begin{aligned}
\mathbb{E}^\dagger \|\sigma_1 + \sigma_2^*\|^2 &= O_P(T^{\epsilon-1} a_k^2 a_T^2) \sum_{j=0}^{T-k-2} \|\hat{\gamma}_{j:k}\|^2 \\
&\quad + O_P(T^{\epsilon-3} k a_k^2 a_T^2) \sum_{t=k+2}^T \sum_{i,j=0}^{t-k-2} \mathbb{I}_{\{i \neq j\}} \hat{\gamma}'_{i:k} \hat{\gamma}_{j:k} \\
&\quad + O_P(T^{\epsilon-3} k a_k^2 a_T^2) \sum_{s,t=k+2}^T \sum_{i=\max\{0,s-t\}}^{s-k-2} \hat{\gamma}'_{i:k} \hat{\gamma}_{t-s+i:k} \mathbb{I}_{\{s,t,s-i-1\}=3} \\
&\quad + O_P(T^{\epsilon-4} k a_k^2 a_T^2) \sum_{s,t=k+2}^T \sum_{i=0}^{s-k-2} \sum_{j=0}^{t-k-2} \mathbb{I}_{\{s-i-1,t-j-1,s,t\}=4} \hat{\gamma}'_{i:k} \hat{\gamma}_{j:k}
\end{aligned}$$

$$= O_P(T^{\epsilon-1} k a_k^2 a_T^2) \|\hat{\gamma}\|^2 + O_P(T^{\epsilon-2} k^2 a_k^2 a_T^2) \|\hat{\gamma}\|_1^2 \\ + O_P(T^{\epsilon-2} k^3 a_k^2 a_T^2 + T^{\epsilon-1} k^2 a_k^2 a_T) + O_P(T^{\epsilon-2} k^3 a_k^2 a_T^2) \|\hat{\gamma}\|_1^2$$

is $O_P(T^{\epsilon-1} k a_k^2 a_T^2)$ for all $\epsilon > 0$ if $k^3/T \rightarrow 0$, the magnitude orders using (S.4.14), (S.4.16) and reasoning applied previously. Hence, $\|\sigma_1 + \sigma_2^*\| = o_{P^\dagger}(T^{\epsilon-1/2} k^{1/2} a_k a_T)$ in P -probability for all $\epsilon > 0$ and an r.p. π .

Next, with $\varpi_{kj} := \sum_{m=k+1}^j \beta_m \gamma_{j-m}$ ($j = k+1, \dots, T-k-2$), σ_2^\dagger can be written as

$$\begin{aligned} \sigma_2^\dagger &= \sum_{t=2k+2}^T \left[\sum_{i=1}^{t-k-1} \gamma_{i-1:k} \varepsilon_{t-i}^\dagger \right] \left[\sum_{j=k+1}^{t-k-1} \varpi_{kj} \varepsilon_{t-j}^\dagger \right] \\ &= \sum_{t=2k+2}^T \sum_{i=k+1}^{t-k-1} \gamma_{i-1:k} \varpi_{ki} \varepsilon_{\pi(t-i)}^2 + \sum_{t=2k+2}^T \sum_{i=1}^{t-k-1} \sum_{j=k+1}^{t-k-1} \mathbb{I}_{i \neq j} \gamma_{i-1:k} \varpi_{kj} \varepsilon_{t-i}^\dagger \varepsilon_{t-j}^\dagger, \end{aligned}$$

where, independently of how w_t are specified,

$$\begin{aligned} \mathbf{E}^\dagger \left\| \sum_{t=2k+2}^T \sum_{i=k+1}^{t-k-1} \gamma_{i-1:k} \varpi_{ki} \varepsilon_{\pi(t-i)}^2 \right\| &\leq \mathbf{E}^\dagger \{ \varepsilon_{\pi(T)}^2 \} \sum_{t=2k+2}^T \sum_{i=k+1}^{t-k-1} \|\gamma_{i-1:k}\| |\varpi_{ki}| \\ &= \sigma_T^2 \|\gamma\|_1^2 \sum_{m=k+1}^\infty |\beta_m| = o_P(a_T) \end{aligned}$$

under Assumption 1(b) and the condition $\sum_{m=k+1}^\infty |\beta_m| = o(a_T^{-1})$. Further, with $f_{s,t} := \sum_{j=1+\max\{k,s-t\}}^{T-t} \gamma_{j+t-s-1:k} \varpi_{kj}$, the term

$$\sigma_{2\times}^\dagger := \sum_{t=2k+2}^T \sum_{i=1}^{t-k-1} \sum_{j=k+1}^{t-k-1} \mathbb{I}_{i \neq j} \gamma_{i-1:k} \varpi_{kj} \varepsilon_{t-i}^\dagger \varepsilon_{t-j}^\dagger = \sum_{s=2k+1}^{T-1} \sum_{t=k+1}^{T-k-1} \mathbb{I}_{s \neq t} \varepsilon_s^\dagger \varepsilon_t^\dagger f_{s,t}$$

has, for Rademacher w_t ,

$$\mathbf{E}^\dagger \|\sigma_{2\times}^\dagger\|^2 = \mathbf{E}^\dagger (\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2) \left(\sum_{s=2k+1}^{T-1} \sum_{t=k+1}^{T-k-1} \mathbb{I}_{s \neq t} \|f_{s,t}\|^2 + \sum_{s,t=2k+1}^{T-k-1} f'_{s,t} f_{t,s} \right) := \mathbf{e}_{2\times}^\dagger,$$

where

$$\begin{aligned} \sum_{s=2k+1}^{T-1} \sum_{t=k+1}^{T-k-1} \mathbb{I}_{s \neq t} \|f_{s,t}\|^2 &\leq \|\gamma\|^2 \sum_{s=2k+1}^{T-1} \sum_{t=k+1}^{T-k-1} \mathbb{I}_{s \neq t} \left(\sum_{j=1+\max\{k,s-t\}}^{T-t} |\varpi_{kj}| \right)^2 \\ &\leq T^2 \|\gamma\|^2 \left(\sum_{j=k+1}^{T-k-1} \sum_{m=k+1}^j |\beta_m| |\gamma_{j-m}| \right)^2 \\ &\leq T^2 \|\gamma\|_1^4 \left(\sum_{m=k+1}^\infty |\beta_m| \right)^2 = o(T^2 a_T^{-2}), \end{aligned}$$

and similarly, also $\sum_{s,t=2k+1}^{T-k-1} f'_{s,t} f_{t,s} = o(T^2 a_T^{-2})$, so

$$\mathbb{E}^\dagger \|\sigma_{2\times}^\dagger\|^2 = \mathbf{e}_{2\times}^\dagger = o(T^2 a_T^{-2}) \mathbb{E}^\dagger(\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2) = o_P(a_T^2)$$

using equation (S.4.7). On the other hand, for $w_t = 1$ a.s. (all t),

$$\begin{aligned} \mathbb{E}^\dagger \|\sigma_{2\times}^\dagger\|^2 &= \mathbf{e}_{2\times}^\dagger + \mathbb{E}^\dagger(\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T)}) \\ &\times \left(\sum_{s=2k+1}^{T-1} \sum_{t,u=k+1}^{T-k-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{su} + \sum_{s=2k+1}^{T-k-1} \sum_{t=k+1}^{T-k-1} \sum_{u=2k+1}^{T-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{us} \right. \\ &+ \left. \sum_{s=2k+1}^{T-1} \sum_{t=2k+1}^{T-k-1} \sum_{u=k+1}^{T-k-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{tu} + \sum_{s,u=2k+1}^{T-1} \sum_{t=k+1}^{T-k-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{ut} \right) \\ &+ \mathbb{E}^\dagger(\varepsilon_{\pi(k+1)} \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T-1)} \varepsilon_{\pi(T)}) \sum_{s,u=2k+1}^{T-1} \sum_{t,v=k+1}^{T-k-1} \mathbb{I}_{\#\{u,v,s,t\}=4} f'_{st} f_{uv} \end{aligned}$$

is $o_P(a_T^2)$ as

$$\begin{aligned} &\left| \sum_{s=2k+1}^{T-1} \sum_{t,u=k+1}^{T-k-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{su} \right| \leq \sum_{s=2k+1}^{T-1} \left(\sum_{t=k+1}^{T-k-1} \mathbb{I}_{s \neq t} \|f_{st}\| \right)^2 \\ &\leq \sum_{s=2k+1}^{T-1} \left(\sum_{t=k+1}^{T-k-1} \mathbb{I}_{s \neq t} \sum_{j=1+\max\{k,s-t\}}^{T-t} \|\gamma_{j+t-s-1:k}\| |\varpi_{kj}| \right)^2 \\ &\leq T^3 \|\gamma\|^2 \left(\sum_{j=k+1}^{T-k-1} \sum_{m=k+1}^j |\beta_m| |\gamma_{j-m}| \right)^2 \leq T^3 \|\gamma\|_1^4 \left(\sum_{j=k+1}^{\infty} |\beta_j| \right)^2 = o(T^3 a_T^{-2}), \end{aligned}$$

and similarly $\sum_{s=2k+1}^{T-k-1} \sum_{t=k+1}^{T-k-1} \sum_{u=2k+1}^{T-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{us} = o(T^3 a_T^{-2})$, also $\sum_{s=2k+1}^{T-1} \sum_{t=2k+1}^{T-k-1} \sum_{u=k+1}^{T-k-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{tu} = o(T^3 a_T^{-2})$, and eventually, likewise, $\sum_{s,u=2k+1}^{T-1} \sum_{t=k+1}^{T-k-1} \mathbb{I}_{\#\{u,s,t\}=3} f'_{st} f_{ut} = o(T^3 a_T^{-2})$, whereas

$$\begin{aligned} &\left| \sum_{s,u=2k+1}^{T-1} \sum_{t,v=k+1}^{T-k-1} \mathbb{I}_{\#\{u,v,s,t\}=4} f'_{st} f_{uv} \right| \leq \left(\sum_{s=2k+1}^{T-1} \sum_{t=k+1}^{T-k-1} \mathbb{I}_{s \neq t} \|f_{st}\| \right)^2 \\ &\leq T^4 \|\gamma\|^2 \left(\sum_{j=k+1}^{T-k-1} \sum_{m=k+1}^j |\beta_m| |\gamma_{j-m}| \right)^2 = o(T^4 a_T^{-2}) \end{aligned}$$

and the expectations were evaluated in (S.4.8) and (S.4.9). By combining the above results with Markov's inequality, it follows that $\sigma_2^\dagger = o_{P^\dagger}(a_T)$ in P -probability.

Finally, we consider σ_3 , still in the case where π is an r.p.:

$$\begin{aligned} \|\sigma_3\| &\leq \left\| \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} (\hat{\gamma}_{j:k} - \gamma_{j:k}) \hat{\varepsilon}_{\pi(t-j-1)} \varepsilon_{\pi(t)} w_{t-j-1} w_t \right\| \\ &\quad + \left\| \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \gamma_{j:k} (\hat{\varepsilon}_{\pi(t-j-1)} - \varepsilon_{\pi(t-j-1)}) \varepsilon_{\pi(t)} w_{t-j-1} w_t \right\|, \end{aligned}$$

where the second norm on the right-hand side is of the same form as $\|\sigma_1 + \sigma_2^*\|$, with $\gamma_{j:k}$ in place of $\hat{\gamma}_{j:k}$, and is $o_{P^\dagger}(T^{\epsilon-1/2} k^{1/2} a_k a_T)$ in P -probability, for all $\epsilon > 0$, by a similar argument as for $\sigma_1 + \sigma_2^*$. Regarding the other norm, say $\|\sigma_{31}\|$, for Rademacher $\{w_t\}$ it holds that

$$\begin{aligned} E^\dagger \|\sigma_{31}\|^2 &= E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2 \} \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k} - \gamma_{j:k}\|^2 \\ &\leq O(T^{-1} k) \sigma_T^2 \delta_{Tk}^2 \|\hat{\gamma}_{T+k-2} - \gamma_{T+k-2}\|^2 \\ &= O(T^{-1} k a_T^4) \|\hat{\beta}_k - \beta_k\|^2 = O_P(T^{\epsilon-1} k a_k^2 a_T^2) \end{aligned}$$

using equations (S.4.14), (S.4.5) and (7.1), for all $\epsilon > 0$. Similarly, for $w_t = 1$ P^\dagger -a.s. (all t),

$$\begin{aligned} E^\dagger \sigma_{31} &= \sum_{s,t=k+2}^T \sum_{i=0}^{s-k-2} \sum_{j=0}^{t-k-2} (\hat{\gamma}_{i:k} - \gamma_{i:k})' (\hat{\gamma}_{j:k} - \gamma_{j:k}) \\ &\quad \times E^\dagger \{ \hat{\varepsilon}_{\pi(s-i-1)} \hat{\varepsilon}_{\pi(t-j-1)} \varepsilon_{\pi(s)} \varepsilon_{\pi(t)} \} \\ &= E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2 \} \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\hat{\gamma}_{j:k} - \gamma_{j:k}\|^2 \\ &\quad + E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} \varepsilon_{\pi(T)}^2 \} \sum_{t=k+2}^T \sum_{i,j=0}^{t-k-2} \mathbb{I}_{\{i \neq j\}} (\hat{\gamma}_{i:k} - \gamma_{i:k})' (\hat{\gamma}_{j:k} - \gamma_{j:k}) \\ &\quad + E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T)} \} \\ &\quad \times \sum_{s,t=k+2}^T \sum_{i=\max\{0,s-t\}}^{s-k-2} (\hat{\gamma}_{i:k} - \gamma_{i:k})' (\hat{\gamma}_{t-s+i:k} - \gamma_{t-s+i:k}) \mathbb{I}_{\#\{s,t,s-i-1\}=3} \\ &\quad + E^\dagger \{ \hat{\varepsilon}_{\pi(k+1)} \hat{\varepsilon}_{\pi(k+2)} \varepsilon_{\pi(k+3)} \varepsilon_{\pi(T)} \} \\ &\quad \times \sum_{s,t=k+2}^T \sum_{i=0}^{s-k-2} \sum_{j=0}^{t-k-2} \mathbb{I}_{\#\{s-i-1, t-j-1, s, t\}=4} (\hat{\gamma}_{i:k} - \gamma_{i:k})' (\hat{\gamma}_{j:k} - \gamma_{j:k}) \end{aligned}$$

implying that

$$\begin{aligned} \mathbb{E}^\dagger \sigma_{31} &= O_P(T^{\epsilon-1} k a_k^2 a_T^2) + O_P(T^{-2} k^2 a_T^4 l_T) \|\hat{\gamma}_{T+k-2} - \gamma_{T+k-2}\|^2 \\ &\quad + O_P(T^{-2} k^3 a_k a_T^{3+\epsilon} l_T + T^{-1} k a_T^3 l_T) \|\hat{\gamma}_{T+k-2} - \gamma_{T+k-2}\| \\ &= O_P(T^{\epsilon-1} k a_k^2 a_T^2) + O_P(T^{-2} k^2 a_T^4 l_T) \|\hat{\beta}_k - \beta_k\|^2 \\ &\quad + O_P(T^{-2} k^3 a_k a_T^{3+\epsilon} l_T + T^{-1} k a_T^3 l_T) \|\hat{\beta}_k - \beta_k\| = O_P(T^{\epsilon-1} k a_k^2 a_T^2), \end{aligned}$$

for all $\epsilon > 0$, as

$$\begin{aligned} &\sum_{s,t=k+2}^T \sum_{i=\max\{0,s-t\}}^{s-k-2} \|\hat{\gamma}_{i:k} - \gamma_{i:k}\| \|\hat{\gamma}_{t-s+i:k} - \gamma_{t-s+i:k}\| \mathbb{I}_{\#\{s,t,s-i-1\}=3} \\ &\leq 2T \|\hat{\gamma}_{T+k-2} - \gamma_{T+k-2}\| \sum_{i=0}^{T-k-2} i \|\hat{\gamma}_{i:k} - \gamma_{i:k}\| \\ &= \|\hat{\gamma}_{T+k-2} - \gamma_{T+k-2}\| O_P(k^3 a_k T a_T^{\epsilon-1} + k T^2 a_T^{-1}) \end{aligned}$$

by (S.4.16) and $k^3/T \rightarrow 0$, so $\|\sigma_3\| = o_{P^\dagger}(T^{\epsilon-1/2} k^{1/2} a_k a_T)$ in P -probability.

Combined with the evaluations of $\|\sigma_1 + \sigma_2^*\|$ and σ_2^\dagger , this proves part (b) in the r.p. case.

S.4.1.4. Proof of part (c). We consider the bootstrap schemes separately for $S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger = \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \gamma_{j:k} \varepsilon_{\pi(t-j-1)} \varepsilon_{\pi(t)} w_{t-j-1} w_t$, and we use previous evaluations for $\|\sigma_2^\dagger\| = o_{P^\dagger}(a_T)$ in P -probability.

For π equal to the identity it holds that

$$\begin{aligned} \mathbb{E}^\dagger \|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\|^2 &= \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \varepsilon_{t-j-1}^2 \varepsilon_t^2 \\ &= \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \varepsilon_{t-j-1}^2 \varepsilon_t^2 (\mathbb{I}_{|\varepsilon_{t-j-1} \varepsilon_t| \leq \tilde{a}_T} + \mathbb{I}_{|\varepsilon_{t-j-1} \varepsilon_t| > \tilde{a}_T}) \end{aligned}$$

with

$$\mathbb{E} \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \varepsilon_{t-j-1}^2 \varepsilon_t^2 \mathbb{I}_{|\varepsilon_{t-j-1} \varepsilon_t| \leq \tilde{a}_T} \leq T k \mathbb{E}(\varepsilon_1^2 \varepsilon_2^2 \mathbb{I}_{|\varepsilon_1 \varepsilon_2| \leq \tilde{a}_T}) \sum_{j=0}^\infty \gamma_j^2,$$

which is $O(\tilde{a}_T^2 k)$, and for $\eta \in [\delta, \alpha]$,

$$\mathbb{E} \left(\sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \varepsilon_{t-j-1}^2 \varepsilon_t^2 \mathbb{I}_{|\varepsilon_{t-j-1} \varepsilon_t| > \tilde{a}_T} \right)^{\frac{\eta}{2}} \leq T k \mathbb{E}(|\varepsilon_1 \varepsilon_2|^\eta \mathbb{I}_{|\varepsilon_1 \varepsilon_2| > \tilde{a}_T}) \sum_{j=0}^\infty |\gamma_j|^\eta$$

which is $O(\tilde{a}_T^\eta k)$ by KT, so

$$\mathbb{E}^\dagger \|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\|^2 = O_P(\tilde{a}_T^2 k^{2/\eta}) \text{ and } \|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\| = o_{P^\dagger}(\tilde{a}_T a_k^{1+\epsilon})$$

in P -probability, for $\epsilon > 0$, by Markov's inequality. Adding $\|\sigma_2^\dagger\| = o_{P^\dagger}(a_T)$ completes the case π equal to the identity.

If π is an r.p., for $S_{0\varepsilon}^{k\dagger}$ it holds that

$$\begin{aligned} \mathbb{E}^\dagger \|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\| &\leq \mathbb{E}^\dagger \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\| |\varepsilon_{\pi(t-j-1)}| |\varepsilon_{\pi(t)}| \\ &\leq O_P(T^{-1}k) \left(\sum_{t=k+1}^T |\varepsilon_t| \right)^2 = O_P(T^{-1}k \max\{T^2, a_T^2 l_T^2\}). \end{aligned}$$

Hence, by Markov's inequality, $\|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\| = O_P(T^{-1}k \max\{T^2, a_T^2 l_T^2\})$ in P -probability. For large α this can be sharpened slightly by evaluating the conditional variance. For Rademacher $\{w_t\}$,

$$\mathbb{E}^\dagger \|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\|^2 = \mathbb{E}^\dagger \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \varepsilon_{\pi(t-j-1)}^2 \varepsilon_{\pi(t)}^2 \leq O_P(T^{-1}k) \sigma_T^4$$

is $O_P(T^{-1}ka_T^4)$, whereas for constant w_t ,

$$\begin{aligned} \mathbb{E}^\dagger \|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\|^2 &= \sum_{s,t=k+2}^T \sum_{i=0}^{s-k-2} \sum_{j=0}^{t-k-2} \gamma'_{i:k} \gamma_{j:k} \mathbb{E}^\dagger \{\varepsilon_{\pi(s-i-1)} \varepsilon_{\pi(t-j-1)} \varepsilon_{\pi(s)} \varepsilon_{\pi(t)}\} \\ &= \mathbb{E}^\dagger \{\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)}^2\} \sum_{t=k+2}^T \sum_{j=0}^{t-k-2} \|\gamma_{j:k}\|^2 \\ &\quad + \mathbb{E}^\dagger \{\varepsilon_{\pi(k+1)} \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T)}^2\} \sum_{t=k+2}^T \sum_{i,j=0}^{t-k-2} \mathbb{I}_{\{i \neq j\}} \gamma'_{i:k} \gamma_{j:k} \\ &\quad + \mathbb{E}^\dagger \{\varepsilon_{\pi(k+1)}^2 \varepsilon_{\pi(k+2)} \varepsilon_{\pi(T)}\} \\ &\quad \times \sum_{s,t=k+2}^T \sum_{i=\max\{0,s-t\}}^{s-k-2} \gamma'_{i:k} \gamma_{t-s+i:k} \mathbb{I}_{\{s,t,s-i-1\}=3} \\ &\quad + \mathbb{E}^\dagger \{\varepsilon_{\pi(k+1)} \varepsilon_{\pi(k+2)} \varepsilon_{\pi(k+3)} \varepsilon_{\pi(T)}\} \\ &\quad \times \sum_{s,t=k+2}^T \sum_{i=0}^{s-k-2} \sum_{j=0}^{t-k-2} \mathbb{I}_{\{s-i-1, t-j-1, s, t\}=4} \gamma'_{i:k} \gamma_{j:k} \\ &= O_P(T^{-1}ka_T^4) \|\gamma\|^2 + O_P(T^{-2}k^2 a_T^4 l_T) = O_P(T^{-1}ka_T^4) \end{aligned}$$

using $\sum_{i=1}^{\infty} i \|\gamma_{i:k}\| \leq k^2 \|\gamma\|_1 + k \sum_{i=1}^{\infty} i |\gamma_i| = O(k^2)$, so $\|S_{0\varepsilon}^{k\dagger} - \sigma_2^\dagger\| = O_{P^\dagger}(T^{-1/2} k^{1/2} a_T^2)$ in P -probability. Adding $\|\sigma_2^\dagger\| = o_{P^\dagger}(a_T)$ completes the proof in the case where π is equal to the identity. \square

S.5. Multiple Restrictions. Consider the Wald statistic W and its bootstrap counterparts W^* and W_Δ^* defined in Remark 4.2(ix). As in the proof of Theorem 3, and using repeatedly the notation introduced there in what follows, define the r.v.'s

$$\begin{aligned} W^\dagger &:= T\sigma_T^{-2} L_k(S_{00}^{\dagger k})^{-1} S_{0\varepsilon}^{\dagger k} [L_k(S_{00}^{\dagger k})^{-1} L'_k]^{-1} L_k(S_{00}^{\dagger k})^{-1} S_{0\varepsilon}^{\dagger k}, \\ \check{W} &:= T\sigma_T^{-2} L_k(\check{S}_{00}^k)^{-1} \check{S}_{0\varepsilon}^k [L_k(\check{S}_{00}^k)^{-1} L'_k]^{-1} L_k(\check{S}_{00}^k)^{-1} \check{S}_{0\varepsilon}^k, \\ W^{\pi k} &:= T\sigma_T^{-2} L_k(S_{00}^{\pi k})^{-1} S_{0\varepsilon}^{\pi k} [L_k(S_{00}^{\pi k})^{-1} L'_k]^{-1} L_k(S_{00}^{\pi k})^{-1} S_{0\varepsilon}^{\pi k}. \end{aligned}$$

The following sequential distances can then be evaluated in place of those in the proof of Theorem 3.

1. As in step 1 of that proof, it holds that $\rho_L(\mathcal{L}^*(W^*), \mathcal{L}^\dagger(W^*)) = 0$ and $\rho_L(\mathcal{L}^*(W_\Delta^*), \mathcal{L}^\dagger(W_\Delta^*)) = 0$.
2. We argue below that, for the wild bootstrap, $W_\Delta^* = W^\dagger + o_{P^\dagger}(T \tilde{a}_T^2 a_T^{-4})$ in P -probability, so

$$(S.5.1) \quad \rho_L(\mathcal{L}^\dagger(a_T^4 \tilde{a}_T^{-2} T^{-1} W_\Delta^*), \mathcal{L}^\dagger(a_T^4 \tilde{a}_T^{-2} T^{-1} W^\dagger)) = o_P(1),$$

whereas for an r.p. π , $W^* = W^\dagger + o_{P^\dagger}(T \tilde{a}_T^2 a_T^{-4})$ in P -probability, so

$$(S.5.2) \quad \rho_L(\mathcal{L}^\dagger(a_T^4 \tilde{a}_T^{-2} T^{-1} W^*), \mathcal{L}^\dagger(a_T^4 \tilde{a}_T^{-2} T^{-1} W^\dagger)) = o_P(1).$$

3(a). As in step 3 of Theorem 3's proof, under bootstrap schemes \mathbf{w}_R it holds that $\rho_L(\mathcal{L}^\varepsilon(S_{00}^{\dagger k}), \mathcal{L}^{|\varepsilon|}(\check{S}_{00}^k)) = 0$ and $\rho_L(\mathcal{L}^\varepsilon(S_{0\varepsilon}^{\dagger k}), \mathcal{L}^{|\varepsilon|}(\check{S}_{0\varepsilon}^k)) = 0$ for symmetric ε_t 's. As a result, $\rho_L(\mathcal{L}^\dagger(a_T^4 \tilde{a}_T^{-2} T^{-1} W^\dagger), \mathcal{L}^{|\varepsilon|}(a_T^4 \tilde{a}_T^{-2} T^{-1} \check{W})) = 0$.

3(b). Under scheme $(\boldsymbol{\pi}_R, \mathbf{w}_1)$, it holds that $W^\dagger = \check{W}$ algebraically.

4(a). Under symmetry of the distribution of ε_t , similarly to step 2,

$$L_k(\check{S}_{00}^k)^{-1} \check{S}_{0\varepsilon}^k = L_k(S_{00}^{\pi k})^{-1} S_{0\varepsilon}^{\pi k} + o_{P^{|\varepsilon|}}(a_T^{-2} \tilde{a}_T + \mathbb{I}_{\pi=r.p.} T^{\varepsilon-1/2} a_T^{-1} k^{1/2} a_k),$$

$$\begin{aligned} [L_k(\check{S}_{00}^k)^{-1} L'_k]^{-1} L_k(\check{S}_{00}^k)^{-1} \check{S}_{0\varepsilon}^k &= [L_k(S_{00}^{\pi k})^{-1} L'_k]^{-1} L_k(S_{00}^{\pi k})^{-1} S_{0\varepsilon}^{\pi k} \\ &\quad + o_{P^{|\varepsilon|}}(\tilde{a}_T + \mathbb{I}_{\pi=r.p.} T^{\varepsilon-1/2} a_T k^{1/2} a_k), \end{aligned}$$

in P -probability, so $\rho_L(\mathcal{L}^{|\varepsilon|}(a_T^4 \tilde{a}_T^{-2} T^{-1} \check{W}), \mathcal{L}^{|\varepsilon|}(a_T^4 \tilde{a}_T^{-2} T^{-1} W^{\pi k})) = o_P(1)$. In the case where π is the identity, using

$$\begin{aligned} \hat{\sigma}_{Tk}^{-2} \sigma_T^2 &= 1 + \hat{\sigma}_{Tk}^{-2} (\|\hat{\varepsilon} - \varepsilon\|^2 + 2 \sum \hat{\varepsilon}_t (\varepsilon_t - \hat{\varepsilon}_t)) \\ &\leq 1 + \hat{\sigma}_{Tk}^{-2} (\|\hat{\varepsilon} - \varepsilon\|^2 + 2 \hat{\sigma}_{Tk} \|\hat{\varepsilon} - \varepsilon\|) \\ &= 1 + O_P(a_T^{-2} (a_k^2 a_T^\varepsilon + a_k a_T^{1+\varepsilon})) = 1 + O_P(a_k a_T^{\varepsilon-1}) \end{aligned}$$

for $\epsilon > 0$, we can conclude that $W^{\pi k}$ equals $\hat{\sigma}_{Tk}^{-2}\sigma_T^2 W = W + o_{P|\epsilon|}(W) = W + o_{P|\epsilon|}(a_T^{-4}\tilde{a}_T^2 T)$, resulting in

$$\rho_L(\mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}\check{W}), \mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1).$$

In the case of an r.p. π , $\hat{\sigma}_{Tk}^{-2}\sigma_T^2 W$ conditional on $\{|\varepsilon_t|\}_{t=-\infty}^k$ and the order statistics of $\{|\varepsilon_t|\}_{t=k+1}^T$ is distributed like $W^{\pi k}$ conditional on $\{|\varepsilon_i|\}_{i=-\infty}^T$, so

$$\rho_L(\mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}\check{W}), \mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1).$$

As $\hat{\sigma}_{Tk}^{-2}\sigma_T^2 TW = TW + o_{P|\epsilon|}(W)$, also

$$\rho_L(\mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}\check{W}), \mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1).$$

4(b). Under scheme $(\boldsymbol{\pi}_R, \mathbf{w}_1)$, similarly to step 2,

$$L_k(\check{S}_{00}^k)^{-1}\check{S}_{0\varepsilon}^k = L_k(S_{00}^{\pi k})^{-1}S_{0\varepsilon}^{\pi k} + o_{P^\dagger}(a_T^{-2}\tilde{a}_T + \mathbb{I}_{\pi=r,p.}T^{\epsilon-1/2}a_T^{-1}k^{1/2}a_k),$$

$$\begin{aligned} [L_k(\check{S}_{00}^k)^{-1}L'_k]^{-1}L_k(\check{S}_{00}^k)^{-1}\check{S}_{0\varepsilon}^k &= [L_k(S_{00}^{\pi k})^{-1}L'_k]^{-1}L_k(S_{00}^{\pi k})^{-1}S_{0\varepsilon}^{\pi k} \\ &\quad + o_{P^\dagger}(\tilde{a}_T + \mathbb{I}_{\pi=r,p.}T^{\epsilon-1/2}a_Tk^{1/2}a_k) \end{aligned}$$

in P -probability. As $\hat{\sigma}_{Tk}^{-2}\sigma_T^2 W$ conditional on $\{|\varepsilon_t|\}_{t=-\infty}^k$ and the order statistics of $\{|\varepsilon_i|\}_{i=k+1}^T$ is distributed like $W^{\pi k}$ under P^\dagger , it follows that

$$\rho_L(\mathcal{L}^\dagger(a_T^4\tilde{a}_T^{-2}T^{-1}\check{W}), \mathcal{L}^\dagger(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1).$$

Finally, $\hat{\sigma}_{Tk}^{-2}\sigma_T^2 W = W + o_{P^\dagger}(W)$, so also

$$\rho_L(\mathcal{L}^\dagger(a_T^4\tilde{a}_T^{-2}T^{-1}\check{W}), \mathcal{L}^\dagger(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1).$$

Then we combine the previous distances using the triangle inequality to conclude that

$$\begin{aligned} \text{for } \boldsymbol{\pi}_{id} &: \rho_L(\mathcal{L}^*(a_T^4\tilde{a}_T^{-2}T^{-1}W_\Delta^*), \mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1), \\ \text{for } \boldsymbol{\pi}_R, \mathbf{w}_R &: \rho_L(\mathcal{L}^*(a_T^4\tilde{a}_T^{-2}T^{-1}W^*), \mathcal{L}^{|\epsilon|}(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1), \\ \text{for } \boldsymbol{\pi}_R, \mathbf{w}_{id} &: \rho_L(\mathcal{L}^*(a_T^4\tilde{a}_T^{-2}T^{-1}W^*), \mathcal{L}^\dagger(a_T^4\tilde{a}_T^{-2}T^{-1}W)) = o_P(1). \end{aligned}$$

These are equivalent to the convergence asserted in Remark 4.2.(ix).

The argument for Step 2 above can be structured as follows. First, the following refinement of Lemma 3(c) can be proved similarly to that lemma (we skip the details).

LEMMA S.3. *Let $k^4/T + 1/k \rightarrow 0$ and Assumption 1 hold. Then it holds that:*

(a) *For an r.p. π , $\|S_{0\varepsilon}^{k\dagger}\| = O_{P^\dagger}((k + a_k^{1+\epsilon})\tilde{a}_T)$ in P probability, for every $\epsilon > 0$.*

Moreover, as in Lemma 1, let the selection matrix L has δ' -summable rows under linear weighting (i.e. such that $\sum_{j=1}^{\infty} j|l_{ij}|^{\delta'} < \infty$, $i = 1, \dots, m$) for some $\delta' \in (\delta, \frac{2\alpha}{2+\alpha})$, with δ as defined in Assumption 1. Then:

(b) *For an r.p. π , $\|(L\Sigma^{-1})_k S_{0\varepsilon}^{k\dagger}\| = O_{P^\dagger}(\tilde{a}_T l_T)$ in P probability, where $l_T = 1$ for $\alpha \neq 1$ and l_T is slowly varying for $\alpha = 1$.*

(c) *For π equal to the identity, $\|(L\Sigma^{-1})_k S_{0\varepsilon}^{k\dagger}\| = O_{P^\dagger}(a_T^{1+\epsilon})$ in P probability, for every $\epsilon > 0$.*

Next, it can be used to derive the following expansions.

LEMMA S.4. *Under Assumption 1 and the δ' -summability assumption of Lemma 1, it holds in P probability that, for small $\epsilon > 0$,*

$$L_k(S_{00}^{k\dagger})^{-1} S_{0\varepsilon}^{k\dagger} = \sigma_T^{-2} (L\Sigma^{-1})_k S_{0\varepsilon}^{k\dagger} + o_{P^\dagger}(a_T^{-\epsilon-1}) = o_{P^\dagger}(a_T^{\epsilon-1})$$

if π is the identity and $k^4/T + 1/k \rightarrow 0$, whereas

$$L_k(S_{00}^{k\dagger})^{-1} S_{0\varepsilon}^{k\dagger} = \sigma_T^{-2} (L\Sigma^{-1})_k S_{0\varepsilon}^{k\dagger} + o_{P^\dagger}(a_T^{-2}\tilde{a}_T) = o_{P^\dagger}(a_T^{-2}\tilde{a}_T l_T)$$

if π is an r.p., $k^5/T + 1/k \rightarrow 0$ and for $\alpha \leq 1$, also $k^{3+2/\alpha+\zeta}/T \rightarrow 0$ for some $\zeta > 0$.

Finally, (S.5.1) and (S.5.2) can be obtained as follows. Similarly to the argument for Step 2 in the proof of Theorem 3, it can be shown that

$$(S.5.3) L_k(\hat{\beta}_k^* - \hat{\beta}_k + \mathbb{I}_{\pi=id}\Delta\hat{\beta}_k^*) = L_k(S_{00}^{*k})^{-1}(S_{0\varepsilon}^{*k} - \sigma_1^{id})$$

$$= L_k(S_{00}^{k\dagger})^{-1} S_{0\varepsilon}^{k\dagger} + o_{P^\dagger}(a_T^{-1-\epsilon} + \mathbb{I}_{\pi=r.p.} T^{\epsilon-1/2} a_T^{-1} k^{1/2} a_k),$$

$$(S.5.4) [L_k(S_{00}^{*k})^{-1} L'_k]^{-1} L_k(\hat{\beta}_k^* - \hat{\beta}_k + \mathbb{I}_{\pi=id}\Delta\hat{\beta}_k^*)$$

$$= [L_k(S_{00}^{k\dagger})^{-1} L'_k]^{-1} L_k(S_{00}^{k\dagger})^{-1} S_{0\varepsilon}^{k\dagger} + o_{P^\dagger}(a_T^{-\epsilon} + \mathbb{I}_{\pi=r.p.} T^{\epsilon-1/2} a_T k^{1/2} a_k),$$

It also holds that

$$\begin{aligned} |\sigma_{Tk}^{*2} - \hat{\sigma}_{Tk}^2| &\leq (\hat{\beta}_k^* - \hat{\beta}_k)' S_{00}^{*k} (\hat{\beta}_k^* - \hat{\beta}_k) + 2\hat{\sigma}_{Tk}[(\hat{\beta}_k^* - \hat{\beta}_k)' S_{00}^{*k} (\hat{\beta}_k^* - \hat{\beta}_k)]^{1/2} \\ &= S_{\varepsilon 0}^{*k} (S_{00}^{*k})^{-1} S_{0\varepsilon}^{*k} + 2\hat{\sigma}_{Tk}[S_{\varepsilon 0}^{*k} (S_{00}^{*k})^{-1} S_{0\varepsilon}^{*k}]^{1/2} \end{aligned}$$

so $|\sigma_{Tk}^{*2} - \hat{\sigma}_{Tk}^2| = o_{P^\dagger}(a_k^{1+\epsilon}\tilde{a}_T)$ for $\pi = id$ and $|\sigma_k^{*2} - \hat{\sigma}_k^2| = O_{P^\dagger}((k + a_k)\tilde{a}_T)$ for an r.p. π , in P -probability, for all $\epsilon > 0$. As $|\hat{\sigma}_{Tk}^2 - \sigma_T^2| = o_P(a_k a_T^{1+\epsilon})$, all $\epsilon > 0$ (see equation (7.1) and (S.4.5)), it follows that

$$(S.5.5) \quad \sigma_k^{*2} = \sigma_T^2 + o_{P^\dagger}(a_k a_T^{1+\epsilon} + \mathbb{I}_{\pi=r.p.} k a_T^{1+\epsilon}).$$

From (S.5.3) and (S.5.4), using that the eigenvalues of $S_{00}^{\dagger k}$ have exact magnitude order a_T^2 in P^\dagger, P probability, and $L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k} = O_{P^\dagger}(a_T^{\epsilon-1})$ for $\pi = id$ and $\epsilon > 0$ (by Lemma S.4), it follows that for small $\epsilon > 0$,

$$T^{-1}\sigma_{Tk}^{*2}W_\Delta^* = L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k}[L_k(S_{00}^{\dagger k})^{-1}L'_k]^{-1}L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k} + o_{P^\dagger}(a_T^{-\epsilon}) = o_{P^\dagger}(a_T^\epsilon)$$

for the wild bootstrap, if $k^4/T \rightarrow 0$. Upon division by σ_k^{*2} and its approximation by σ_T^2 according to (S.5.5), it is obtained that $W_\Delta^* = W^\dagger + o_{P^\dagger}(T\tilde{a}_T^2a_T^{-4})$ and, as a consequence, equation (S.5.1) holds.

If π is an r.p., independently of the specification of $\{w_t\}_{t=k+1}^T$, equations (S.5.3) and (S.5.4) specialize to

$$L_k(\hat{\beta}_k^* - \hat{\beta}_k) = L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k} + o_{P^\dagger}(a_T^{-1-\epsilon} + T^{\epsilon-1/2}a_T^{-1}k^{1/2}a_k),$$

$$\begin{aligned} [L_k(S_{00}^{*k})^{-1}L'_k]^{-1}L_k(\hat{\beta}_k^* - \hat{\beta}_k) &= [L_k(S_{00}^{\dagger k})^{-1}L'_k]^{-1}L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k} \\ &\quad + o_{P^\dagger}(a_T^{1-\epsilon} + T^{\epsilon-1/2}k^{1/2}a_k a_T). \end{aligned}$$

As now $L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k} = O_{P^\dagger}(a_T^{-2}\tilde{a}_T l_T)$ under the hypotheses of Lemma S.4, it follows that, if $k^5/T \rightarrow 0$ (and for $\alpha \leq 1$, also $k^{3+2/\alpha+\zeta}/T \rightarrow 0$ for some $\zeta > 0$), then for sufficiently small $\epsilon > 0$,

$$T^{-1}\sigma_{Tk}^{*2}W^* = L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k}[L_k(S_{00}^{\dagger k})^{-1}L'_k]^{-1}L_k(S_{00}^{\dagger k})^{-1}S_{0\varepsilon}^{\dagger k} + o_{P^\dagger}(a_T^{-\epsilon})$$

is $o_{P^\dagger}(a_T^\epsilon)$ in P probability. As previously, jointly with (S.5.5) this leads to $W^* = W^\dagger + o_{P^\dagger}(T\tilde{a}_T^2a_T^{-4})$ and, hence, (S.5.2) follows.

Additional References.

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