

Supplement to
“Accurate and (Almost) Tuning Parameter Free
Inference in Cointegrating Regressions”

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Equation numbers not preceded by “D.” refer to the main article.

C Asymptotic Critical Values in the Presence of Deterministic Regressors

Table C.1: Asymptotic critical values for $\tau_{\text{IM}}(\hat{\eta}_T)$ in the presence of deterministic regressors d_t in (2.1)

%	$m = 1$	$m = 2$		$m = 3$			$m = 4$			
	$s = 1$	$s = 1$	$s = 2$	$s = 1$	$s = 2$	$s = 3$	$s = 1$	$s = 2$	$s = 3$	$s = 4$
intercept ($d_t = 1$)										
90.0	64.13	94.15	168.58	126.65	221.45	305.36	162.08	278.05	382.30	481.15
95.0	95.81	140.55	233.15	187.03	297.11	396.56	236.54	372.79	487.71	596.15
97.5	136.10	190.23	292.64	245.93	375.55	488.35	325.56	458.37	587.85	720.31
99.0	187.13	263.92	381.78	338.59	474.31	602.27	421.68	582.89	719.98	872.07
intercept and linear time trend ($d_t = [1, t]'$)										
90.0	90.44	122.19	209.54	152.66	261.47	363.17	180.25	311.22	434.29	545.37
95.0	134.19	171.46	283.33	219.51	354.08	460.37	258.75	423.39	546.31	688.21
97.5	183.51	231.09	357.66	294.26	433.33	569.84	342.56	524.25	686.44	810.09
99.0	243.72	304.08	460.98	409.03	556.42	713.24	478.05	680.76	821.12	977.09
intercept, linear and square time trend ($d_t = [1, t, t^2]'$)										
90.0	115.13	138.49	245.91	175.40	302.95	418.77	205.72	352.96	479.57	608.35
95.0	166.35	200.65	331.26	255.74	402.90	530.94	303.58	465.28	621.70	764.20
97.5	217.42	268.86	401.63	348.51	509.51	637.29	390.59	589.05	762.57	902.89
99.0	290.63	357.58	513.85	472.49	646.48	800.81	527.20	754.26	923.59	1070.32
intercept, linear, square and cubic time trend ($d_t = [1, t, t^2, t^3]'$)										
90.0	137.70	166.87	292.13	197.84	340.61	465.58	229.38	392.80	533.60	680.84
95.0	198.48	237.82	379.15	288.65	446.27	590.05	334.55	509.11	684.33	858.04
97.5	263.30	308.64	467.71	391.70	565.65	720.19	438.56	645.41	853.07	1004.50
99.0	352.56	406.48	587.03	539.71	726.07	903.53	592.44	846.82	1052.20	1222.78

Notes: s is the number of linearly independent restrictions under the null hypothesis on the coefficients corresponding to the $m \geq s$ integrated regressors.

D Proofs of Auxiliary Results

Proof of Lemma 1. The solution of the sample Yule-Walker equations in the regression of w_t on w_{t-1}, \dots, w_{t-q} , $t = q+1, \dots, T$, can be written in compact form as a $((m+1) \times q(m+1))$ -dimensional matrix

$$\tilde{\Phi}(q) := [\tilde{\Phi}_1(q), \dots, \tilde{\Phi}_q(q)] = \tilde{\Gamma} \tilde{G}^{-1},$$

with the $((m+1) \times q(m+1))$ -dimensional matrix $\tilde{\Gamma} := [\tilde{\Gamma}(1), \dots, \tilde{\Gamma}(q)]$, the $(q(m+1) \times q(m+1))$ -dimensional matrix $\tilde{G} := (\tilde{\Gamma}(s-r))_{r,s=1,\dots,q}$ and the $((m+1) \times (m+1))$ -dimensional empirical autocovariance matrix of w_1, \dots, w_T at lag $-q+1 \leq h \leq q$, given by

$$\tilde{\Gamma}(h) := T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (w_{t+h} - \bar{w}_T)(w_t - \bar{w}_T)',$$

where $\bar{w}_T := T^{-1} \sum_{t=1}^T w_t$. Analogously, the solution of the sample Yule-Walker equations in the regression of \hat{w}_t on $\hat{w}_{t-1}, \dots, \hat{w}_{t-q}$, $t = q+1, \dots, T$, can be written in compact form as

$$\hat{\Phi}(q) := [\hat{\Phi}_1(q), \dots, \hat{\Phi}_q(q)] = \hat{\Gamma} \hat{G}^{-1},$$

with $\hat{\Gamma} := [\hat{\Gamma}(1), \dots, \hat{\Gamma}(q)]$, $\hat{G} := (\hat{\Gamma}(s-r))_{r,s=1,\dots,q}$ and $\hat{\Gamma}(h)$ the empirical autocovariance matrix of $\hat{w}_1, \dots, \hat{w}_T$ at lag $-q+1 \leq h \leq q$. Taking the difference of $\tilde{\Phi}(q)$ and $\hat{\Phi}(q)$, adding and subtracting $\hat{\Gamma} \tilde{G}^{-1}$ and using

$$\tilde{G}^{-1} - \hat{G}^{-1} = \tilde{G}^{-1}(\hat{G} - \tilde{G})\hat{G}^{-1}$$

leads to

$$\hat{\Phi}(q) - \tilde{\Phi}(q) = \hat{\Gamma} \tilde{G}^{-1}(\tilde{G} - \hat{G})\hat{G}^{-1} - (\tilde{\Gamma} - \hat{\Gamma})\tilde{G}^{-1}$$

Hence, we have to consider $\tilde{G} - \hat{G}$ in more detail ($\tilde{\Gamma} - \hat{\Gamma}$ works similarly). A typical

block element of $\tilde{G} - \hat{G}$ is

$$\begin{aligned}
\tilde{\Gamma}(h) - \hat{\Gamma}(h) &= T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (w_{t+h} - \bar{w}_T)(w_t - \bar{w}_T)' \\
&\quad - T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (\hat{w}_{t+h} - \bar{\hat{w}}_T)(\hat{w}_t - \bar{\hat{w}}_T)' \\
&= T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (w_{t+h} - \bar{w}_T)(w_t - \hat{w}_t - (\bar{w}_T - \bar{\hat{w}}_T))' \\
&\quad - T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (\hat{w}_{t+h} - w_{t+h} - (\bar{\hat{w}}_T - \bar{w}_T))(\hat{w}_t - \bar{\hat{w}}_T)' \\
&= A_1(h) - A_2(h),
\end{aligned}$$

with an obvious definition for $A_1(h)$ and $A_2(h)$. Let us consider $A_1(h)$ in more detail ($A_2(h)$ works similarly). Using $\hat{w}_t = [\hat{u}_t, v_t']'$ and $\hat{u}_t = y_t - x_t' \hat{\beta}_{\text{IM}}$ together with the model equations (2.1) and (2.2), we get

$$w_t - \hat{w}_t = \begin{bmatrix} u_t - \hat{u}_t \\ 0_{m \times 1} \end{bmatrix} = \begin{bmatrix} y_t - x_t' \beta - (y_t - x_t' \hat{\beta}_{\text{IM}}) \\ 0_{m \times 1} \end{bmatrix} = \begin{bmatrix} x_t' (\hat{\beta}_{\text{IM}} - \beta) \\ 0_{m \times 1} \end{bmatrix}. \quad (\text{D.1})$$

Before we continue, note that the last equality implies

$$\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \leq T^{-1/2} \max_{1 \leq t \leq T} |T^{-1/2} x_t|_F |T (\hat{\beta}_{\text{IM}} - \beta)|_F = O_{\mathbb{P}}(T^{-1/2}),$$

since $\max_{1 \leq t \leq T} |T^{-1/2} x_t|_F = \sup_{0 \leq r \leq 1} |T^{-1/2} x_{\lfloor rT \rfloor}|_F$ converges by Assumption 3 and the continuous mapping theorem to $\sup_{0 \leq r \leq 1} |B_v(r)|_F = O_{\mathbb{P}}(1)$ and $\hat{\beta}_{\text{IM}}$ is rate- T consistent. This proves (B.1).

We now proceed with the proof of (B.2). From (D.1) we obtain

$$\begin{aligned}
A_1(h) &= T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (w_{t+h} - \bar{w}_T) \begin{bmatrix} (x_t - \bar{x}_T)' (\hat{\beta}_{\text{IM}} - \beta) \\ 0_{m \times 1} \end{bmatrix}' \\
&= T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (w_{t+h} - \bar{w}_T) [(x_t - \bar{x}_T)' (\hat{\beta}_{\text{IM}} - \beta), 0_{1 \times m}] \\
&= \tilde{\Gamma}_{w,x}(h) (\hat{\beta}_{\text{IM}} - \beta) e_1',
\end{aligned}$$

where $\tilde{\Gamma}_{w,x}(h) := T^{-1} \sum_{t=\max\{1,1-h\}}^{\min\{T,T-h\}} (w_{t+h} - \bar{w}_T) (x_t - \bar{x}_T)'$ is $((m+1) \times m)$ -dimensional

and $e_1 := (1, 0_{1 \times m})'$ is the first $(m+1)$ -dimensional unit vector. Denoting the part of $\tilde{G} - \hat{G}$ that consists of block-entries $A_1(h)$ by $\tilde{G}_1 - \hat{G}_1$, we get

$$\tilde{G}_1 - \hat{G}_1 = [\tilde{\Gamma}_{w,x}(s-r)(\hat{\beta}_{\text{IM}} - \beta)e'_1]_{r,s=1,\dots,q} = \tilde{\Gamma}_{w,x}(I_q \otimes ((\hat{\beta}_{\text{IM}} - \beta)e'_1)),$$

where $\tilde{\Gamma}_{w,x} := (\tilde{\Gamma}_{w,x}(s-r))_{r,s=1,\dots,q}$ is $(q(m+1) \times qm)$ -dimensional. For the second factor we have

$$\begin{aligned} |I_q \otimes ((\hat{\beta}_{\text{IM}} - \beta)e'_1)|_F &= \sqrt{\text{tr} \left((I_q \otimes ((\hat{\beta}_{\text{IM}} - \beta)e'_1))' I_q \otimes ((\hat{\beta}_{\text{IM}} - \beta)e'_1) \right)} \\ &= \sqrt{\text{tr} \left(I_q \otimes (e_1(\hat{\beta}_{\text{IM}} - \beta)'(\hat{\beta}_{\text{IM}} - \beta)e'_1) \right)} \\ &= \sqrt{\text{tr} \left(I_q \otimes \text{diag} \left(\sum_{i=1}^m (\hat{\beta}_{\text{IM},i} - \beta_i)^2, 0, \dots, 0 \right) \right)} \\ &= \sqrt{q \sum_{i=1}^m (\hat{\beta}_{\text{IM},i} - \beta_i)^2} \\ &= q^{1/2} T^{-1} |T(\hat{\beta}_{\text{IM}} - \beta)|_F \\ &= O_P(q^{1/2} T^{-1}). \end{aligned}$$

Next, let us consider $\tilde{\Gamma}_{w,x}$ in more detail. Recall that $w_t = (u_t, v'_t)'$ and $x_t = \sum_{k=1}^t v_k$. To avoid lengthy index notation, w.l.o.g. we can assume that $m = 1$ and consider the second element of w_t only (the first element works similarly). We thus consider the scalar quantity

$$\tilde{\Gamma}_{w,x}(h) = T^{-1} \sum_{t=\max\{1, 1-h\}}^{\min\{T, T-h\}} (v_{t+h} - \bar{v}_T) \left(\sum_{k=1}^t v_k - T^{-1} \sum_{i=1}^T \sum_{j=1}^i v_j \right).$$

Taking the expectation of the squared Frobenius norm of the corresponding $(q \times q)$ -

dimensional matrix $\tilde{\Gamma}_{w,x}$ and combining the sums over r and s , leads to

$$\begin{aligned}
& \mathbb{E}(|\tilde{\Gamma}_{w,x}|_F^2) \\
&= \sum_{r,s=1}^q T^{-2} \sum_{t_1, t_2=\max\{1, 1-(s-r)\}}^{\min\{T, T-(s-r)\}} \mathbb{E} \left[(v_{t_1+s-r} - \bar{v}_T) \left(\sum_{k_1=1}^{t_1} v_{k_1} - T^{-1} \sum_{i_1=1}^T \sum_{j_1=1}^{i_1} v_{j_1} \right) \right. \\
&\quad \left. \times (v_{t_2+s-r} - \bar{v}_T) \left(\sum_{k_2=1}^{t_2} v_{k_2} - T^{-1} \sum_{i_2=1}^T \sum_{j_2=1}^{i_2} v_{j_2} \right) \right] \\
&= T^{-2} \sum_{h=-q+1}^{q-1} (q - |h|) \sum_{t_1, t_2=\max\{1, 1-h\}}^{\min\{T, T-h\}} \mathbb{E} \left[(v_{t_1+h} - \bar{v}_T) \left(\sum_{k_1=1}^{t_1} v_{k_1} - T^{-1} \sum_{i_1=1}^T \sum_{j_1=1}^{i_1} v_{j_1} \right) \right. \\
&\quad \left. \times (v_{t_2+h} - \bar{v}_T) \left(\sum_{k_2=1}^{t_2} v_{k_2} - T^{-1} \sum_{i_2=1}^T \sum_{j_2=1}^{i_2} v_{j_2} \right) \right].
\end{aligned}$$

Note that the last expectation is of the form $\mathbb{E}(ABCD)$ with $\mathbb{E}(A) = \mathbb{E}(B) = \mathbb{E}(C) = \mathbb{E}(D) = 0$. Hence, by using common rules for joint cumulants of centered random variables (see, e. g., Brillinger, 1981), we get

$$\begin{aligned}
& \mathbb{E}(ABCD) \\
&= \text{cum}(A, B, C, D) + \mathbb{E}(AB)\mathbb{E}(CD) + \mathbb{E}(AC)\mathbb{E}(BD) + \mathbb{E}(AD)\mathbb{E}(BC) \\
&= \text{cum}(A, B, C, D) + \text{Cov}(A, B)\text{Cov}(C, D) + \text{Cov}(A, C)\text{Cov}(B, D) \\
&\quad + \text{Cov}(A, D)\text{Cov}(B, C),
\end{aligned}$$

where $\text{Cov}(\cdot, \cdot)$ denotes the covariance of two random variables. Hence, the first term corresponding to the fourth-order cumulant becomes

$$\begin{aligned}
& T^{-2} \sum_{h=-q+1}^{q-1} (q - |h|) \sum_{t_1, t_2=\max\{1, 1-h\}}^{\min\{T, T-h\}} \text{cum} \left(v_{t_1+h} - \bar{v}_T, \sum_{k_1=1}^{t_1} v_{k_1} - T^{-1} \sum_{i_1=1}^T \sum_{j_1=1}^{i_1} v_{j_1}, \right. \\
&\quad \left. v_{t_2+h} - \bar{v}_T, \sum_{k_2=1}^{t_2} v_{k_2} - T^{-1} \sum_{i_2=1}^T \sum_{j_2=1}^{i_2} v_{j_2} \right)
\end{aligned}$$

leading to $2^4 = 16$ terms when expanding the cumulant. Exemplarily, for the absolute value of the first one (the others work similarly), we get from the common

calculation rules for cumulants

$$\begin{aligned}
& |T^{-2} \sum_{h=-q+1}^{q-1} (q - |h|) \sum_{t_1, t_2 = \max\{1, 1-h\}}^{\min\{T, T-h\}} \text{cum} \left(v_{t_1+h}, \sum_{k_1=1}^{t_1} v_{k_1}, v_{t_2+h}, \sum_{k_2=1}^{t_2} v_{k_2} \right) | \\
& \leq T^{-2} \sum_{h=-q+1}^{q-1} |q - |h|| \sum_{t_1, t_2 = \max\{1, 1-h\}}^{\min\{T, T-h\}} \sum_{k_1=1}^{t_1} \sum_{k_2=1}^{t_2} |\text{cum}(v_{t_1+h}, v_{k_1}, v_{t_2+h}, v_{k_2})| \\
& \leq \frac{q}{T^2} \sum_{h=-q+1}^{q-1} \sum_{t_1, t_2=1}^T \sum_{k_1, k_2=1}^T |\text{cum}(v_{t_1+h}, v_{k_1}, v_{t_2+h}, v_{k_2})|.
\end{aligned}$$

By combining the sums over t_1 and t_2 and those over k_1 and k_2 , respectively, the above term becomes

$$\begin{aligned}
& \frac{q}{T^2} \sum_{h=-q+1}^{q-1} \sum_{l=-(T-1)}^{T-1} \sum_{s=\max\{1, 1-l\}}^{\min\{T, T-l\}} \sum_{i=-(T-1)}^{T-1} \sum_{j=\max\{1, 1-i\}}^{\min\{T, T-i\}} |\text{cum}(v_{s+l+h}, v_{j+i}, v_{s+h}, v_j)| \\
& \leq \frac{q}{T^2} \sum_{h=-q+1}^{q-1} \sum_{i, l=-(T-1)}^{T-1} \sum_{s, j=1}^T |\text{cum}(v_{s+l+h}, v_{j+i}, v_{s+h}, v_j)| \\
& \leq \frac{q}{T^2} \sum_{h=-q+1}^{q-1} \sum_{i, l=-(T-1)}^{T-1} \sum_{k=-(T-1)}^{T-1} \sum_{r=\max\{1, 1-k\}}^{\min\{T, T-k\}} |\text{cum}(v_{r+k+l+h}, v_{r+i}, v_{r+k+h}, v_r)| \\
& \leq \frac{q}{T} \sum_{h=-q+1}^{q-1} \sum_{i, l=-(T-1)}^{T-1} \sum_{k=-(T-1)}^{T-1} |\text{cum}(v_{k+l+h}, v_i, v_{k+h}, v_0)|,
\end{aligned}$$

where we also combined the sums over s and j and made use of the (strict) stationarity of $\{v_t\}_{t \in \mathbb{Z}}$. Finally, combining the sums over h and k , we get the bound

$$\begin{aligned}
& \frac{q}{T} \sum_{r=-(T+q-2)}^{T+q-2} (2q-1) \sum_{i, l=-(T-1)}^{T-1} |\text{cum}(v_{r+l}, v_i, v_r, v_0)| \\
& \leq 2 \frac{q^2}{T} \sum_{j=-(2T+q-3)}^{2T+q-3} \sum_{r=-(T+q-2)}^{T+q-2} \sum_{i=-(T-1)}^{T-1} |\text{cum}(v_j, v_i, v_r, v_0)| \\
& \leq 2 \frac{q^2}{T} \sum_{j, i, r=-\infty}^{\infty} |\text{cum}(v_j, v_i, v_r, v_0)| = O(q^2 T^{-1})
\end{aligned}$$

due to the summability condition imposed on the fourth order cumulants in Assumption 2 and, hence, vanishes for $T \rightarrow \infty$. However, the leading term is the

term corresponding to $\text{Cov}(A, B)\text{Cov}(C, D)$. That is, we have to consider

$$\begin{aligned}
& T^{-2} \sum_{h=-q+1}^{q-1} (q - |h|) \sum_{t_1, t_2=\max\{1, 1-h\}}^{\min\{T, T-h\}} \text{Cov}(v_{t_1+h} - \bar{v}_T, \sum_{k_1=1}^{t_1} v_{k_1} - T^{-1} \sum_{i_1=1}^T \sum_{j_1=1}^{i_1} v_{j_1}) \\
& \quad \times \text{Cov}(v_{t_2+h} - \bar{v}_T, \sum_{k_2=1}^{t_2} v_{k_2} - T^{-1} \sum_{i_2=1}^T \sum_{j_2=1}^{i_2} v_{j_2}) \\
& = \sum_{h=-q+1}^{q-1} (q - |h|) \left(T^{-1} \sum_{t=\max\{1, 1-h\}}^{\min\{T, T-h\}} \text{Cov}(v_{t+h} - \bar{v}_T, \sum_{k=1}^t v_k - T^{-1} \sum_{i=1}^T \sum_{j=1}^i v_j) \right)^2.
\end{aligned}$$

Hence, we have to compute

$$T^{-1} \sum_{t=\max\{1, 1-h\}}^{\min\{T, T-h\}} \text{Cov}(v_{t+h} - \bar{v}_T, \sum_{k=1}^t v_k - T^{-1} \sum_{i=1}^T \sum_{j=1}^i v_j).$$

This leads to four terms to consider, which can be treated with similar arguments.

For the first term we get

$$T^{-1} \sum_{t=\max\{1, 1-h\}}^{\min\{T, T-h\}} \text{Cov}(v_{t+h}, \sum_{k=1}^t v_k) = T^{-1} \sum_{t=\max\{1, 1-h\}}^{\min\{T, T-h\}} \sum_{k=1}^t \gamma_v(t+h-k),$$

where $\gamma_v(h)$ is the covariance of the one-dimensional process (still assumed for notational brevity) $\{v_t\}_{t \in \mathbb{Z}}$ at lag h . E. g., for $h \geq 0$, this can be exactly computed to equal

$$T^{-1} \sum_{t=1}^{T-h} \sum_{k=1}^t \gamma_v(t+h-k) = T^{-1} \sum_{j=h}^{T-1} (T-j) \gamma_v(j) \leq \sum_{j=h}^{T-1} \gamma_v(j)$$

and its absolute value can be bounded by $\sum_{j=-\infty}^{\infty} |\gamma_v(j)| < \infty$ due to the second-order cumulant condition imposed in Assumption 2. Similar arguments yield the same bound for $h < 0$ and for the other three terms. Hence, the term of $\mathbb{E}(|\tilde{\Gamma}_{w,x}|_F^2)$ that corresponds to $\text{Cov}(A, B)\text{Cov}(C, D)$ is of order $O(q^2)$.

In total we thus have $\mathbb{E}(|\tilde{\Gamma}_{w,x}|_F^2) = O(q^2)$. Note that we have proven this result for $m = 1$. However, as m is fixed, the result also holds for $m > 1$. Therefore, we obtain for the $(q(m+1) \times qm)$ -dimensional matrix $\tilde{\Gamma}_{w,x}$ that $|\tilde{\Gamma}_{w,x}|_F = O_{\mathbb{P}}(q)$. It follows that $|\tilde{G}_1 - \hat{G}_1|_F = O_{\mathbb{P}}(q) O_P(q^{1/2} T^{-1}) = O_{\mathbb{P}}(q^{3/2} T^{-1})$ and similarly also $|\tilde{G} - \hat{G}|_F = O_{\mathbb{P}}(q) O_{\mathbb{P}}(q^{1/2} T^{-1}) = O_{\mathbb{P}}(q^{3/2} T^{-1})$.

Further, we have to consider \tilde{G}^{-1} . In the following, let $\mu_{\min}(A)$ and $\mu_{\max}(A)$

denote the smallest and largest eigenvalue of a matrix A , respectively and define $G := (\Gamma(s - r))_{r,s=1,\dots,q} \in \mathbb{R}^{q(m+1) \times q(m+1)}$. Similar to the above, using the fourth-order cumulant condition from Assumption 2, we can show that $|\tilde{G} - G|_F = O_{\mathbb{P}}(qT^{-1/2}) = o_{\mathbb{P}}(1)$. Then, to show boundedness in probability of \tilde{G}^{-1} (similar for \hat{G}^{-1}), for all $\epsilon > 0$, we have to find a $K < \infty$ and a $T_0 < \infty$ both large enough such that for all $T > T_0$, it holds that

$$\mathbb{P}(|\tilde{G}^{-1}|_2 > K) < \epsilon,$$

where $|A|_2$ denotes the spectral norm of a matrix A . Let $\epsilon > 0$. Then, due to positive semi-definiteness of \tilde{G} by construction and invertibility, see Meyer and Kreiss (2015, Lemma 3.4 and Remark 3.2), we have positive definiteness of \tilde{G} and, consequently, of \tilde{G}^{-1} . Hence, we get

$$\begin{aligned} \mathbb{P}(|\tilde{G}^{-1}|_2 > K) &= \mathbb{P}(\mu_{\max}(\tilde{G}^{-1}) > K) = \mathbb{P}(\mu_{\min}^{-1}(\tilde{G}) > K) \\ &= \mathbb{P}(\mu_{\min}(\tilde{G}) < \frac{1}{K}, |\tilde{G} - G|_2 \geq \delta) \\ &\quad + \mathbb{P}(\mu_{\min}(\tilde{G}) < \frac{1}{K}, |\tilde{G} - G|_2 < \delta). \end{aligned}$$

Further, as $|\tilde{G} - G|_2 \leq |\tilde{G} - G|_F = o_{\mathbb{P}}(1)$, for any $\delta > 0$, we can choose T_0 large enough to have $\mathbb{P}(|\tilde{G} - G|_2 \geq \delta) \leq \epsilon$. Then the first term on the last right-hand side can be bounded by $\mathbb{P}(|\tilde{G} - G|_2 \geq \delta) \leq \epsilon$. For the second term, as \tilde{G} , G and hence also $\tilde{G} - G$ are symmetric with real-valued entries, these matrices are Hermitian such that Weyl's theorem (see, e.g., Theorem 4.3.1 in Horn and Johnson, 2012) applies, leading to the inequality $\mu_{\min}(G) + \mu_{\min}(\tilde{G} - G) \leq \mu_{\min}(\tilde{G})$. It follows that

$$\begin{aligned} &\mathbb{P}(\mu_{\min}(\tilde{G}) < \frac{1}{K}, |\tilde{G} - G|_2 < \delta) \\ &\leq \mathbb{P}(\mu_{\min}(G) + \mu_{\min}(\tilde{G} - G) < \frac{1}{K}, |\tilde{G} - G|_2 < \delta) \\ &= \mathbb{P}(\mu_{\min}(G) < \frac{1}{K} - \mu_{\min}(\tilde{G} - G), |\tilde{G} - G|_2 < \delta). \end{aligned}$$

From symmetry of $\tilde{G} - G$ we get that the eigenvalues of $(\tilde{G} - G)'(\tilde{G} - G)$ are

exactly the squared eigenvalues of $\tilde{G} - G$. Hence, the bound

$$|\tilde{G} - G|_2 = \sqrt{\mu_{\max}((\tilde{G} - G)'(\tilde{G} - G))} < \delta$$

implies also $\mu_{\min}(\tilde{G} - G) \geq -\delta$ such that the last right-hand side can be bounded by

$$\mathbb{P}(\mu_{\min}(G) < \frac{1}{K} + \delta, |\tilde{G} - G|_2 < \delta) \leq \mathbb{P}(\mu_{\min}(G) < \frac{1}{K} + \delta). \quad (\text{D.2})$$

Next, note that $\mu_{\min}(G) \geq \tilde{c}$ for some constant $\tilde{c} > 0$ by Assumption 1. Therefore, the right-hand side in (D.2) becomes zero if we choose $\delta < \tilde{c}/2$ small enough and $K > 2/\tilde{c}$ large enough, such that $\frac{1}{K} + \delta < \tilde{c}$. This completes the proof of $|\tilde{G}^{-1}|_2 = O_{\mathbb{P}}(1)$. Furthermore, from Assumption 1 we also get $|\tilde{G}|_2 = O_{\mathbb{P}}(1)$ and similarly $|\hat{G}|_2 = O_{\mathbb{P}}(1)$ and $|\hat{G}|_2 = O_{\mathbb{P}}(1)$. Altogether, we get

$$\begin{aligned} |\tilde{\Phi}(q) - \hat{\Phi}(q)|_2 &\leq |\tilde{\Gamma} - \hat{\Gamma}|_2 |\tilde{G}^{-1}|_2 + |\hat{\Gamma}|_2 |\tilde{G}^{-1}|_2 |\tilde{G} - \hat{G}|_2 |\hat{G}^{-1}|_2 \\ &\leq |\tilde{G} - \hat{G}|_2 (|\tilde{G}^{-1}|_2 + |\hat{G}|_2 |\tilde{G}^{-1}|_2 |\hat{G}^{-1}|_2) \\ &\leq |\tilde{G} - \hat{G}|_F (|\tilde{G}^{-1}|_2 + |\hat{G}|_2 |\tilde{G}^{-1}|_2 |\hat{G}^{-1}|_2) \\ &= O_{\mathbb{P}}(q^{3/2}T^{-1}) (O_{\mathbb{P}}(1) + O_{\mathbb{P}}(1)O_{\mathbb{P}}(1)O_{\mathbb{P}}(1)) \\ &= O_{\mathbb{P}}(q^{3/2}T^{-1}). \end{aligned}$$

Since $\tilde{\Phi}(q) - \hat{\Phi}(q)$ is $((m+1) \times q(m+1))$ -dimensional and m is fixed, it holds that (see, e. g. Gentle, 2007)

$$|\tilde{\Phi}(q) - \hat{\Phi}(q)|_F \leq \sqrt{m+1} |\tilde{\Phi}(q) - \hat{\Phi}(q)|_2 = O_{\mathbb{P}}(q^{3/2}T^{-1}).$$

This implies that

$$q^{1/2} \sum_{j=1}^q |\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F \leq q^{3/2} |\tilde{\Phi}(q) - \hat{\Phi}(q)|_F = O_{\mathbb{P}}(q^3T^{-1}),$$

which is $o_{\mathbb{P}}(1)$ since $q^3T^{-1} = o(1)$ by Assumption 4. \square

In the proofs of Lemma 2 and 3 we repeatedly use the fact that by convexity, $|\sum_{i=1}^k z_i|^a \leq k^{a-1} \sum_{i=1}^k |z_i|^a$, for all $a, k \geq 1$.

Proof of Lemma 2. Let $\tilde{\varepsilon}_t(q) := w_t - \sum_{j=1}^q \tilde{\Phi}_j(q)w_{t-j}$, $t = q+1, \dots, T$, denote

the Yule-Walker residuals in the regression of w_t on w_{t-1}, \dots, w_{t-q} , $t = q+1, \dots, T$ and define $\bar{\tilde{\varepsilon}}_T(q) := (T-q)^{-1} \sum_{t=q+1}^T \tilde{\varepsilon}_t(q)$.²³ For $q+1 \leq t \leq T$ we have

$$\begin{aligned}
& |\hat{\varepsilon}_t(q) - \bar{\tilde{\varepsilon}}_T(q)|_F^a \\
& \leq \left(|\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F + |\tilde{\varepsilon}_t(q) - \bar{\tilde{\varepsilon}}_T(q)|_F + |\bar{\tilde{\varepsilon}}_T(q) - \tilde{\varepsilon}_T(q)|_F \right)^a \\
& \leq \left(|\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F + |\tilde{\varepsilon}_t(q) - \bar{\tilde{\varepsilon}}_T(q)|_F + (T-q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F \right)^a \\
& \leq 3^{a-1} \left(|\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F^a + |\tilde{\varepsilon}_t(q) - \bar{\tilde{\varepsilon}}_T(q)|_F^a + \left((T-q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F \right)^a \right).
\end{aligned}$$

Hence,

$$\begin{aligned}
& (T-q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\tilde{\varepsilon}}_T(q)|_F^a \\
& \leq 3^{a-1} \left((T-q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F^a + (T-q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_t(q) - \bar{\tilde{\varepsilon}}_T(q)|_F^a \right. \\
& \quad \left. + \left((T-q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F \right)^a \right) \\
& = 3^{a-1} \left(F_{T,a} + (T-q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_t(q) - \bar{\tilde{\varepsilon}}_T(q)|_F^a + (F_{T,1})^a \right),
\end{aligned}$$

where $F_{T,a} := (T-q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F^a$.

²³In the following, a denotes the fixed $a > 2$ from Assumption 1. However, the results also hold for $1 \leq \tilde{a} < a$.

We now consider $F_{T,a}$ in more detail. For $q+1 \leq t \leq T$ we have

$$\begin{aligned}
& |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F \\
&= |\hat{w}_t - \sum_{j=1}^q \hat{\Phi}_j(q) \hat{w}_{t-j} - (w_t - \sum_{j=1}^q \tilde{\Phi}_j(q) w_{t-j})|_F \\
&= |\hat{w}_t - w_t - \sum_{j=1}^q (\hat{\Phi}_j(q) - \tilde{\Phi}_j(q) + \tilde{\Phi}_j(q)) (\hat{w}_{t-j} - w_{t-j} + w_{t-j}) + \sum_{j=1}^q \tilde{\Phi}_j(q) w_{t-j}|_F \\
&= |\hat{w}_t - w_t - \sum_{j=1}^q (\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)) (\hat{w}_{t-j} - w_{t-j} + w_{t-j}) - \sum_{j=1}^q \tilde{\Phi}_j(q) (\hat{w}_{t-j} - w_{t-j})|_F \\
&\leq \max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F + \sum_{j=1}^q |\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F |\hat{w}_{t-j} - w_{t-j}|_F \\
&\quad + \sum_{j=1}^q |\tilde{\Phi}_j(q)|_F |\hat{w}_{t-j} - w_{t-j}|_F + \sum_{j=1}^q |\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F |w_{t-j}|_F \\
&\leq \max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F + \max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \sum_{j=1}^q |\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F \\
&\quad + \max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \sum_{j=1}^q |\tilde{\Phi}_j(q)|_F + \sqrt{m+1} q |\hat{\Phi}(q) - \tilde{\Phi}(q)|_F q^{-1} \sum_{j=1}^q |w_{t-j}|_F,
\end{aligned}$$

where we have used that

$$\begin{aligned}
|\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F &\leq |\hat{\Phi}(q) - \tilde{\Phi}(q)|_F |[0, I_{m+1}, 0]']_F \\
&= \sqrt{m+1} |\hat{\Phi}(q) - \tilde{\Phi}(q)|_F.
\end{aligned}$$

Therefore,

$$\begin{aligned}
& |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F^a \\
&\leq 4^{a-1} \left(\left(\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \right)^a + \left(\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \right)^a \left(\sum_{j=1}^q |\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F \right)^a \right. \\
&\quad + \left(\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \right)^a \left(\sum_{j=1}^q |\tilde{\Phi}_j(q)|_F \right)^a \\
&\quad \left. + \left(\sqrt{m+1} q |\hat{\Phi}(q) - \tilde{\Phi}(q)|_F \right)^a \left(q^{-1} \sum_{j=1}^q |w_{t-j}|_F \right)^a \right).
\end{aligned}$$

It follows that

$$\begin{aligned}
F_{T,a} \leq & 4^{a-1} \left(\left(\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \right)^a + \left(\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \right)^a \left(\sum_{j=1}^q |\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F \right)^a \right. \\
& + \left(\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F \right)^a \left(\sum_{j=1}^q |\tilde{\Phi}_j(q)|_F \right)^a \\
& \left. + \left(\sqrt{m+1}q |\hat{\Phi}(q) - \tilde{\Phi}(q)|_F \right)^a (T-q)^{-1} \sum_{t=q+1}^T \left(q^{-1} \sum_{j=1}^q |w_{t-j}|_F \right)^a \right).
\end{aligned}$$

From Lemma 1 we have $\max_{1 \leq t \leq T} |\hat{w}_t - w_t|_F = O_{\mathbb{P}}(T^{-1/2})$ and $\sum_{j=1}^q |\hat{\Phi}_j(q) - \tilde{\Phi}_j(q)|_F = O_{\mathbb{P}}(q^{5/2}T^{-1})$. Moreover, the proof of Lemma 1 shows that $q|\hat{\Phi}(q) - \tilde{\Phi}(q)|_F = O_{\mathbb{P}}(q^{5/2}T^{-1})$. Further, by (B.3), (B.4) and Assumption 1 we have

$$\begin{aligned}
\sum_{j=1}^q |\tilde{\Phi}_j(q)|_F & \leq \sum_{j=1}^q |\tilde{\Phi}_j(q) - \Phi_j(q)|_F + \sum_{j=1}^q |\Phi_j(q) - \Phi_j|_F + \sum_{j=1}^q |\Phi_j|_F \\
& \leq q \sup_{1 \leq j \leq q} |\tilde{\Phi}_j(q) - \Phi_j(q)|_F + c \sum_{j=q+1}^{\infty} |\Phi_j|_F + \sum_{j=1}^{\infty} |\Phi_j|_F \\
& = O_{\mathbb{P}}(1).
\end{aligned}$$

Finally, note that

$$\begin{aligned}
(T-q)^{-1} \sum_{t=q+1}^T \left(q^{-1} \sum_{j=1}^q |w_{t-j}|_F \right)^a & \leq (T-q)^{-1} \sum_{t=q+1}^T q^{-a} q^{a-1} \sum_{j=1}^q |w_{t-j}|_F^a \\
& = (T-q)^{-1} q^{-1} \sum_{t=q+1}^T \sum_{j=1}^q |w_{t-j}|_F^a \\
& \leq (T-q)^{-1} \sum_{t=1}^{T-1} |w_t|_F^a,
\end{aligned}$$

where the last inequality follows from the fact that each element in the double sum occurs at most q times, i. e., $\sum_{t=q+1}^T \sum_{j=1}^q |w_{t-j}|_F^a \leq q \sum_{t=1}^{T-1} |w_t|_F^a$. From $\sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) < \infty$ and Markov's inequality, it follows that $(T-q)^{-1} \sum_{t=1}^{T-1} |w_t|_F^a = O_{\mathbb{P}}(1)$. In total, we thus have $F_{T,a} = O_{\mathbb{P}}((q^{5/2}T^{-1})^a) = o_{\mathbb{P}}(1)$. Therefore,

$$(T-q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_T(q)|_F^a \leq 3^{a-1} \left((T-q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_t(q) - \tilde{\varepsilon}_T(q)|_F^a + o_{\mathbb{P}}(1) \right).$$

It thus remains to show that $(T - q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_t(q) - \tilde{\varepsilon}_T(q)|_F^a = O_{\mathbb{P}}(1)$. We now follow Park (2002, Proof of Lemma 3.2) and Palm *et al.* (2010, Proof of Lemma 2). Define $\varepsilon_t(q) := \varepsilon_t + \sum_{j=q+1}^{\infty} \Phi_j w_{t-j}$ and note that

$$\begin{aligned} & (T - q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_t(q) - \tilde{\varepsilon}_T(q)|_F^a \\ &= (T - q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_t(q) - \varepsilon_t(q) + \varepsilon_t(q) - \varepsilon_t + \varepsilon_t - \tilde{\varepsilon}_T(q)|_F^a \\ &\leq 4^{a-1} (A_{T,a} + B_{T,a} + C_{T,a} + D_{T,a}), \end{aligned}$$

where

$$\begin{aligned} A_{T,a} &:= (T - q)^{-1} \sum_{t=q+1}^T |\varepsilon_t|_F^a, \\ B_{T,a} &:= (T - q)^{-1} \sum_{t=q+1}^T |\varepsilon_t(q) - \varepsilon_t|_F^a = (T - q)^{-1} \sum_{t=q+1}^T \left| \sum_{j=q+1}^{\infty} \Phi_j w_{t-j} \right|_F^a, \\ C_{T,a} &:= (T - q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_t(q) - \varepsilon_t(q)|_F^a, \\ D_{T,a} &:= (T - q)^{-1} \sum_{t=q+1}^T |\tilde{\varepsilon}_T(q)|_F^a = |\tilde{\varepsilon}_T(q)|_F^a = |(T - q)^{-1} \sum_{t=q+1}^T \tilde{\varepsilon}_t(q)|_F^a. \end{aligned}$$

We first consider $B_{T,a}$. Note that $\mathbb{E}(|B_{T,a}|_F) \leq \sup_{t \in \mathbb{Z}} \mathbb{E}(|\varepsilon_t(q) - \varepsilon_t|_F^a)$. Using Minkowski's inequality, we have

$$\begin{aligned} \mathbb{E}(|\varepsilon_t(q) - \varepsilon_t|_F^a) &= \mathbb{E} \left(\left| \sum_{j=q+1}^{\infty} \Phi_j w_{t-j} \right|_F^a \right) = \left(\left[\mathbb{E} \left(\left| \sum_{j=q+1}^{\infty} \Phi_j w_{t-j} \right|_F^a \right) \right]^{1/a} \right)^a \\ &\leq \left(\sum_{j=q+1}^{\infty} [\mathbb{E}(|\Phi_j w_{t-j}|_F^a)]^{1/a} \right)^a \leq \left(\sum_{j=q+1}^{\infty} [\mathbb{E}(|\Phi_j|_F^a |w_{t-j}|_F^a)]^{1/a} \right)^a \\ &\leq \left(\sum_{j=q+1}^{\infty} |\Phi_j|_F [\mathbb{E}(|w_{t-j}|_F^a)]^{1/a} \right)^a \leq \left(\sum_{j=q+1}^{\infty} |\Phi_j|_F \left[\sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) \right]^{1/a} \right)^a \\ &\leq \sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) \left(\sum_{j=q+1}^{\infty} |\Phi_j|_F \right)^a. \end{aligned}$$

From Assumption 1 we have $\sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) < \infty$ and $\sum_{j=q+1}^{\infty} |\Phi_j|_F = o(1)$.

Markov's inequality thus yields $B_{T,a} = o_{\mathbb{P}}(1)$. Analogously, $\mathbb{E}(|A_{T,a}|_F) \leq \sup_{t \in \mathbb{Z}} \mathbb{E}(|\varepsilon_t|_F^a)$.

Using Minkowski's inequality, we have as above

$$\begin{aligned}
\mathbb{E}(|\varepsilon_t|_F^a) &= \mathbb{E}\left(|w_t - \sum_{j=1}^{\infty} \Phi_j w_{t-j}|_F^a\right) \leq 2^{a-1} \left(\mathbb{E}(|w_t|_F^a) + \mathbb{E}\left(|\sum_{j=1}^{\infty} \Phi_j w_{t-j}|_F^a\right) \right) \\
&\leq 2^{a-1} \left(\sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) + \mathbb{E}\left(|\sum_{j=1}^{\infty} \Phi_j w_{t-j}|_F^a\right) \right) \\
&\leq 2^{a-1} \sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) \left(1 + \left(\sum_{j=1}^{\infty} |\Phi_j|_F \right)^a \right) < \infty.
\end{aligned}$$

Using Markov's inequality we conclude that $A_{T,a} = O_{\mathbb{P}}(1)$. We now turn to $C_{T,a}$. By definition,

$$\begin{aligned}
\tilde{\varepsilon}_t(q) &= w_t - \sum_{j=1}^q \tilde{\Phi}_j(q) w_{t-j} = \varepsilon_t(q) - \sum_{j=1}^q (\tilde{\Phi}_j(q) - \Phi_j) w_{t-j} \\
&= \varepsilon_t(q) - \sum_{j=1}^q (\tilde{\Phi}_j(q) - \Phi_j(q)) w_{t-j} - \sum_{j=1}^q (\Phi_j(q) - \Phi_j) w_{t-j}.
\end{aligned}$$

Hence,

$$|\tilde{\varepsilon}_t(q) - \varepsilon_t(q)|_F^a \leq 2^{a-1} \left(\left| \sum_{j=1}^q (\tilde{\Phi}_j(q) - \Phi_j(q)) w_{t-j} \right|_F^a + \left| \sum_{j=1}^q (\Phi_j(q) - \Phi_j) w_{t-j} \right|_F^a \right).$$

It follows that $C_{T,a} = 2^{a-1} (C_{1T,a} + C_{2T,a})$, where

$$\begin{aligned}
C_{1T,a} &:= (T-q)^{-1} \sum_{t=q+1}^T \left| \sum_{j=1}^q (\tilde{\Phi}_j(q) - \Phi_j(q)) w_{t-j} \right|_F^a, \\
C_{2T,a} &:= (T-q)^{-1} \sum_{t=q+1}^T \left| \sum_{j=1}^q (\Phi_j(q) - \Phi_j) w_{t-j} \right|_F^a.
\end{aligned}$$

We consider both terms separately. First note that

$$\begin{aligned}
C_{1T,a} &\leq q^{a-1} (T-q)^{-1} \sum_{t=q+1}^T \sum_{j=1}^q |\tilde{\Phi}_j(q) - \Phi_j(q)|_F^a |w_{t-j}|_F^a \\
&\leq q^{a-1} \left(\sup_{1 \leq j \leq q} |\tilde{\Phi}_j(q) - \Phi_j(q)|_F \right)^a (T-q)^{-1} \sum_{t=q+1}^T \sum_{j=1}^q |w_{t-j}|_F^a \\
&\leq \left(q \sup_{1 \leq j \leq q} |\tilde{\Phi}_j(q) - \Phi_j(q)|_F \right)^a (T-q)^{-1} \sum_{t=1}^{T-1} |w_t|_F^a,
\end{aligned}$$

where the third inequality follows again from the fact that $\sum_{t=q+1}^T \sum_{j=1}^q |w_{t-j}|_F^a \leq q \sum_{t=1}^{T-1} |w_t|_F^a$. As $(T-q)^{-1} \sum_{t=1}^{T-1} |w_t|_F^a = O_{\mathbb{P}}(1)$ it follows from (B.3) that $C_{1T,a} = o_{\mathbb{P}}(1)$. Moreover, using Minkowski's inequality, we obtain

$$\begin{aligned}
\mathbb{E}(|C_{2T,a}|_F) &= (T-q)^{-1} \sum_{t=q+1}^T \mathbb{E} \left(\left| \sum_{j=1}^q (\Phi_j(q) - \Phi_j) w_{t-j} \right|_F^a \right) \\
&= (T-q)^{-1} \sum_{t=q+1}^T \left(\left[\mathbb{E} \left(\left| \sum_{j=1}^q (\Phi_j(q) - \Phi_j) w_{t-j} \right|_F^a \right) \right]^{1/a} \right)^a \\
&\leq (T-q)^{-1} \sum_{t=q+1}^T \left(\sum_{j=1}^q [\mathbb{E}(|(\Phi_j(q) - \Phi_j) w_{t-j}|_F^a)]^{1/a} \right)^a \\
&\leq (T-q)^{-1} \sum_{t=q+1}^T \sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) \left(\sum_{j=1}^q |\Phi_j(q) - \Phi_j|_F \right)^a \\
&= \sup_{t \in \mathbb{Z}} \mathbb{E}(|w_t|_F^a) \left(\sum_{j=1}^q |\Phi_j(q) - \Phi_j|_F \right)^a.
\end{aligned}$$

From (B.4) and Markov's inequality it follows that $C_{2T,a} = o_{\mathbb{P}}(1)$. In total we thus have $C_{T,a} = o_{\mathbb{P}}(1)$. Finally, we consider $D_{T,a}$. It holds that $(T-q)^{-1} \sum_{t=q+1}^T \tilde{\varepsilon}_t(q) = D_{1T} + D_{2T} + D_{3T}$, where

$$\begin{aligned}
D_{1T} &:= (T-q)^{-1} \sum_{t=q+1}^T \varepsilon_t, \\
D_{2T} &:= (T-q)^{-1} \sum_{t=q+1}^T (\varepsilon_t(q) - \varepsilon_t), \\
D_{3T} &:= (T-q)^{-1} \sum_{t=q+1}^T (\tilde{\varepsilon}_t(q) - \varepsilon_t(q)).
\end{aligned}$$

By Chebyshev's weak law of large numbers (White, 2001, p. 25), $D_{1T} \xrightarrow{p} \mathbb{E}(\varepsilon_t) = 0$, i. e., $D_{1T} = o_{\mathbb{P}}(1)$. Moreover, $|D_{2T}|_F \leq B_{T,1} = o_{\mathbb{P}}(1)$ and $|D_{3T}|_F \leq C_{T,1} = o_{\mathbb{P}}(1)$. By the continuous mapping theorem we thus have $D_T = o_{\mathbb{P}}(1)$. This completes the proof. \square

Proof of Lemma 3. It follows from Assumption 2 that

$$|(T-q)^{-1} \sum_{t=q+1}^T \varepsilon_t \varepsilon_t' - \Sigma|_F = o_{\mathbb{P}}(1).$$

Therefore,

$$|\mathbb{E}^* (\varepsilon_t^* \varepsilon_t^{*'}) - \Sigma|_F \leq |\mathbb{E}^* (\varepsilon_t^* \varepsilon_t^{*'}) - (T - q)^{-1} \sum_{t=q+1}^T \varepsilon_t \varepsilon_t'|_F + o_{\mathbb{P}}(1).$$

Moreover,

$$\begin{aligned} & |\mathbb{E}^* (\varepsilon_t^* \varepsilon_t^{*'}) - (T - q)^{-1} \sum_{t=q+1}^T \varepsilon_t \varepsilon_t'|_F \\ &= |(T - q)^{-1} \sum_{t=q+1}^T \left(\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) \right) \left(\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) \right)' - \varepsilon_t \varepsilon_t'|_F \\ &= |(T - q)^{-1} \sum_{t=q+1}^T \left(\left[\left(\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) \right) - \varepsilon_t \right] \left(\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) \right)' \right. \\ &\quad \left. + \varepsilon_t \left[\left(\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) \right) - \varepsilon_t \right]' \right)|_F \\ &\leq E_{1T} + E_{2T}, \end{aligned}$$

where

$$\begin{aligned} E_{1T} &= (T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q)|_F, \\ E_{2T} &= (T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F |\varepsilon_t|_F. \end{aligned}$$

The Cauchy-Schwarz inequality yields

$$\begin{aligned} E_{1T} &\leq \left((T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F^2 (T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q)|_F^2 \right)^{1/2}, \\ E_{2T} &\leq \left((T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F^2 (T - q)^{-1} \sum_{t=q+1}^T |\varepsilon_t|_F^2 \right)^{1/2}. \end{aligned}$$

From the proof of Lemma 2 we have $(T - q)^{-1} \sum_{t=q+1}^T |\varepsilon_t|_F^2 = A_{T,2} = O_{\mathbb{P}}(1)$ and $(T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q)|_F^2 = O_{\mathbb{P}}(1)$. It thus remains to show that

$(T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F^2 = o_{\mathbb{P}}(1)$. To this end note that

$$\begin{aligned}
& |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F \\
& \leq |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F + |\tilde{\varepsilon}_t(q) - (w_t - \sum_{j=1}^q \Phi_j(q) w_{t-j})|_F \\
& \quad + |w_t - \sum_{j=1}^q \Phi_j(q) w_{t-j} - \varepsilon_t|_F + |\bar{\varepsilon}_T(q)|_F \\
& = |\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F + |\sum_{j=1}^q (\tilde{\Phi}_j(q) - \Phi_j(q)) w_{t-j}|_F \\
& \quad + |\sum_{j=1}^q (\Phi_j(q) - \Phi_j) w_{t-j}|_F + |\sum_{j=q+1}^{\infty} \Phi_j w_{t-j}|_F + |\bar{\varepsilon}_T(q)|_F.
\end{aligned}$$

Hence,

$$\begin{aligned}
& |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F^2 \\
& \leq 5 \left(|\hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F^2 + |\sum_{j=1}^q (\tilde{\Phi}_j(q) - \Phi_j(q)) w_{t-j}|_F^2 \right. \\
& \quad \left. + |\sum_{j=1}^q (\Phi_j(q) - \Phi_j) w_{t-j}|_F^2 + |\sum_{j=q+1}^{\infty} \Phi_j w_{t-j}|_F^2 + |\bar{\varepsilon}_T(q)|_F^2 \right).
\end{aligned}$$

In the notation of the proof of Lemma 2, we obtain

$$\begin{aligned}
(T - q)^{-1} \sum_{t=q+1}^T |\hat{\varepsilon}_t(q) - \bar{\varepsilon}_T(q) - \varepsilon_t|_F^2 & \leq 5 \left(F_{T,2} + C_{1T,2} + C_{2T,2} + B_{T,2} + |\bar{\varepsilon}_T(q)|_F^2 \right) \\
& = 5|\bar{\varepsilon}_T(q)|_F^2 + o_{\mathbb{P}}(1).
\end{aligned}$$

From $\hat{\varepsilon}_t(q) = \hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q) + \tilde{\varepsilon}_t(q) - \varepsilon_t(q) + \varepsilon_t(q) - \varepsilon_t + \varepsilon_t$, with $\varepsilon_t(q)$ as defined in the proof of Lemma 2, it follows that

$$\begin{aligned}
|\bar{\varepsilon}_T(q)|_F & \leq |(T - q)^{-1} \sum_{t=q+1}^T \hat{\varepsilon}_t(q) - \tilde{\varepsilon}_t(q)|_F + |(T - q)^{-1} \sum_{t=q+1}^T \tilde{\varepsilon}_t(q) - \varepsilon_t(q)|_F \\
& \quad + |(T - q)^{-1} \sum_{t=q+1}^T \varepsilon_t(q) - \varepsilon_t|_F + |(T - q)^{-1} \sum_{t=q+1}^T \varepsilon_t|_F \\
& \leq F_{T,1} + C_{T,1} + B_{T,1} + |D_{1T}|_F = o_{\mathbb{P}}(1).
\end{aligned}$$

This completes the proof. \square

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