

Long-Run Variance Estimation

Model:

Let $\{\varepsilon_t\}$ be an AR(∞)-process of the form

$$\varepsilon_t = \sum_{j=1}^{\infty} \alpha_j \varepsilon_{t-j} + \eta_t$$

with $\{\eta_t\}$ i.i.d. and $A(z) = 1 - \sum_{j=1}^{\infty} \alpha_j z^j \neq 0$
 for all $|z| \leq 1+d$ with some small $d > 0$.

Remarks:

(i.) The coefficients α_j decay to zero exp. fast, i.e.,
 $|\alpha_j| \leq C j^{-\delta}$ for some $\delta \in (0, 1)$ and $C < \infty$.

(ii.) $\{\varepsilon_t\}$ has an MA(∞)-representation of the form

$$\varepsilon_t = \sum_{j=0}^{\infty} c_j \eta_{t-j}.$$

The coefficients c_j decay exp. fast to zero, i.e.,
 $|c_j| \leq C j^{-\delta}$ for some $\delta \in (0, 1)$ and $C < \infty$.

(iii) From (ii), it follows that $\{\varepsilon_t\}$ is stationary
 and causal.

(iv.) If $\alpha_j = 0$ for all $j > p^*$ and $\alpha_{p^*} \neq 0$, then
 we have the special case of an AR(p^*)-process
 of finite order p^* .

(v.) The coefficients c_j can be computed iteratively from the following equations:

$$\underbrace{(1 - \alpha_1 L - \alpha_2 L^2 - \dots)}_{A(L)} \underbrace{(c_0 + c_1 L + c_2 L^2 + \dots)}_{C(L)} m_A = m_A,$$

that is,

$$c_0 = 1$$

$$c_1 - \alpha_1 c_0 = 0$$

$$c_2 - \alpha_1 c_1 - \alpha_2 c_0 = 0$$

$$c_3 - \alpha_1 c_2 - \alpha_2 c_1 - \alpha_3 c_0 = 0,$$

⋮

that is,

$$c_j - \sum_{l=1}^j \alpha_l c_{j-l} = 0 \quad \text{for } j \geq 1$$

with $c_l = 0$ for $l < 0$ and $\alpha_l = 0$ for $l > p$.

Yule-Walker equations:

Consider the process $\{\Delta_q \varepsilon_t\}$ with $\Delta_q \varepsilon_t = \varepsilon_t - \varepsilon_{t-q}$.
 $\{\Delta_q \varepsilon_t\}$ is an ARMA (α, q) -process of the form

$$\begin{aligned} \Delta_q \varepsilon_t - \sum_{j=1}^{\infty} \alpha_j \Delta_q \varepsilon_{t-j} \\ = A(L) \Delta_q \varepsilon_t = A(L) \varepsilon_t - A(L) \varepsilon_{t-q} = \eta_t - \eta_{t-q}. \end{aligned}$$

The corresponding Yule-Walker equations are as follows:

$$\begin{aligned} & E[\Delta_q \varepsilon_t \Delta_q \varepsilon_{t+l}] - \underbrace{\sum_{j=1}^{\infty} \alpha_j E[\Delta_q \varepsilon_{t-j} \Delta_q \varepsilon_{t+l}]}_{-\gamma_q(l-j)} \\ & = \gamma_q(l) \\ & = E[\eta_t \Delta_q \varepsilon_{t+l}] - E[\eta_{t-q} \Delta_q \varepsilon_{t+l}] \\ & = E[\eta_t \varepsilon_{t+l}] - E[\eta_t \varepsilon_{t+l-q}] \\ & \quad - E[\eta_{t-q} \varepsilon_{t+l}] + E[\eta_{t-q} \varepsilon_{t+l-q}] \\ & = 2 \cdot E[\eta_t \varepsilon_{t+l}] - E[\eta_t \varepsilon_{t+l-q}] - E[\eta_{t-q} \varepsilon_{t+l+q}] \end{aligned}$$

With

$$E[\eta_t \varepsilon_{t+l}] = \sum_{j=0}^{\infty} c_j E[\eta_t \eta_{t+l-j}] = c_l \frac{\sigma^2}{E[\eta_0^2]}.$$

Hence,

$$\gamma_q(l) - \sum_{j=1}^{\infty} a_j \gamma_q(l-j) =: d_q(l) = \begin{cases} [2c_{1l} - c_{1l+q} - c_{1l+q}]v^2, & l \leq 0 \\ -c_{q-l}v^2 & , 1 \leq l \leq q \\ 0 & , l > q. \end{cases}$$

Combining these equations for $l=1, \dots, p$, we get

$$\gamma_q - \Gamma_q a - r_q = d_q \text{, i.e.}$$

$$\Gamma_q a = \gamma_q - d_q - r_q$$

with

$$\Gamma_q = (\gamma_q(i-j) : 1 \leq i, j \leq p)$$

$$\gamma_q = (\gamma_q(1), \dots, \gamma_q(p))^T$$

$$d_q = (d_q(1), \dots, d_q(p))^T$$

$$r_q = \left(\sum_{j=p+1}^{\infty} a_j \gamma_q(l-j) : l=1, \dots, p \right)^T$$

$$a = (a_1, \dots, a_p)^T.$$

Note that $\|r_q\|_2 \leq C \sum_{j=p+1}^{\infty} |a_j| \leq C \xi^p$

and thus $\|r_q\|_2 \leq C \sqrt{p} \xi^p$.

Estimadores:

Definie

$$\hat{Y}_q(l) = \frac{1}{F_q} \sum_{k=q+l+1}^T \Delta_q Y_{k,T} \Delta_q Y_{t-l,T}$$

$$\hat{Y}_q^*(l) = \frac{1}{F_q} \sum_{k=q+l+1}^T \Delta_q \epsilon_k \Delta_q \epsilon_{t-l}.$$

We have

$$\hat{Y}_q(l) = \hat{Y}_q^*(l) + R_{q,1}(l) + R_{q,2}(l) + R_{q,3}(l)$$

with

$$R_{q,1}(l) = \frac{1}{F_q} \sum_{k=q+l+1}^T \underbrace{\Delta_q m_k}_{= m(A_T) - m(A_{q,T})} \Delta_q \epsilon_{t-l}$$

$$R_{q,2}(l) = \frac{1}{F_q} \sum_{k=q+l+1}^T \Delta_q \epsilon_k \underbrace{\Delta_q m_{t-l}}_{m(A_T) - m(A_{q,T})}$$

$$R_{q,3}(l) = \frac{1}{F_q} \sum_{k=q+l+1}^T \Delta_q m_k \Delta_q m_{t-l}.$$

Write

$$\hat{Y}_q = (\hat{Y}_q(1), \dots, \hat{Y}_q(P))^T$$

$$\Gamma_q = (\hat{Y}_q(i-j) : 1 \leq i, j \leq P).$$

Define

$$\tilde{\alpha}_q = \hat{\Gamma}_q^{-1} \hat{y}_q = (\tilde{\alpha}_1, \dots, \tilde{\alpha}_p)^T.$$

Moreover, let

$$\tilde{\nu}^2 = \frac{1}{2T} \sum_{k=p+2}^T \tilde{\pi}_{k,T}^2$$

with $\tilde{\pi}_{k,T} = \Delta_k y_{k,T} - \sum_{j=1}^p \tilde{\alpha}_j \Delta_j y_{kj,T}$ and set

$$\tilde{\sigma}^2 = \frac{\tilde{\nu}^2}{(1 - \sum_{j=1}^p \tilde{\alpha}_j)^2}.$$

Finally, let the estimates \tilde{c}_j ($j \geq 0$) be defined by the recursions $\tilde{c}_0 = 1$ and

$$\tilde{c}_j = \sum_{\ell=1}^j \tilde{\alpha}_\ell \tilde{c}_{j-\ell} \text{ for } j \geq 1.$$

Lemma 1: Let $1 \leq p \ll \sqrt{T}$ and $1 \leq q \ll \sqrt{T}$. Then
for any $1 \leq l \leq p$,

$$E(\hat{\gamma}_q(l) - \gamma_q(l))^2 \leq C \left(\frac{1}{Tq} + \left(\frac{p}{Tq} \right)^2 + \left(\frac{q}{T} \right)^2 \right)$$

with a fixed constant C indep. of l, p, q, T .

Proof:

$$\begin{aligned} & E(\hat{\gamma}_q(l) - \gamma_q(l))^2 \\ & \leq 16 \left\{ E(\hat{\gamma}_q^*(l) - \gamma_q(l))^2 + \sum_{k=1}^3 E R_{q,k}^2(l) \right\}. \end{aligned}$$

We first have a close look at $E(\hat{\gamma}_q^*(l) - \gamma_q(l))^2$:

$$\begin{aligned} & \hat{\gamma}_q^*(l) - \gamma_q(l) = \frac{1}{Tq} \sum_{t=q+l+1}^T \{ \Delta_q \varepsilon_t \Delta_q \varepsilon_{t-l} - E \Delta_q \varepsilon_t \Delta_q \varepsilon_{t-l} \} \\ & \quad - \left(1 - \frac{T-q-l}{T-q} \right) \gamma_q(l) \\ & = \frac{1}{Tq} \sum_{t=q+l+1}^T \{ \varepsilon_t \varepsilon_{t-l} - E \varepsilon_t \varepsilon_{t-l} \} \\ & \quad - \frac{1}{Tq} \sum_{t=q+l+1}^T \{ \varepsilon_t \varepsilon_{t+l-q} - E \varepsilon_t \varepsilon_{t+l-q} \} \\ & \quad - \frac{1}{Tq} \sum_{t=q+l+1}^T \{ \varepsilon_{t-q} \varepsilon_{t-l} - E \varepsilon_{t-q} \varepsilon_{t-l} \} \\ & \quad + \frac{1}{Tq} \sum_{t=q+l+1}^T \{ \varepsilon_{t-q} \varepsilon_{t+l-q} - E \varepsilon_{t-q} \varepsilon_{t+l-q} \} \\ & \quad - \left(1 - \frac{T-q-l}{T-q} \right) \gamma_q(l) \\ & =: \Delta_{q,1}(l) + \dots + \Delta_{q,4}(l) - \left(1 - \frac{T-q-l}{T-q} \right) \gamma_q(l). \end{aligned}$$

and thus

$$E(\hat{Y}_q^*(l) - Y_q(l))^2 \leq 25 \left\{ \sum_{k=1}^4 E \Delta_{q,k}^2(l) + \underbrace{\left(1 - \frac{F_q - l}{F_q}\right)^2 Y_q^2(l)}_{\leq C \left(\frac{P}{F_q}\right)^2} \right\}.$$

It holds that

$$\begin{aligned} E \Delta_{q,1}^2(l) &= \frac{1}{(F_q)^2} \sum_{k,t=k+l+1}^T E[\epsilon_k \epsilon_{t-l} \epsilon_{t'} \epsilon_{t'-l}] \\ &\quad - \underbrace{\frac{1}{(F_q)^2} \sum_{k,t'=k+l+1}^T E[\epsilon_k \epsilon_{t-l}] E[\epsilon_{t'} \epsilon_{t'-l}]}_{= \frac{(F_q - l)^2}{(F_q)^2} Y_E^2(l)} \\ &= \frac{(F_q - l)^2}{(F_q)^2} Y_E^2(l) \end{aligned}$$

Since $\epsilon_t = \sum_{k=0}^{\infty} c_k \eta_{t+k}$ and $Y_E(l) = (\sum_{k=0}^{\infty} c_k c_{k+l}) \nu^2$,

$$E[\epsilon_k \epsilon_{t-l} \epsilon_{t'} \epsilon_{t'-l}]$$

$$= \sum_{k,k',j,j'} c_k c_{k'} c_j c_{j'} \underbrace{E[m_{k,k+j} m_{t-k,t+l-j} m_{t'-k',t'+l-j'}]}_{\neq 0 \text{ if}}$$

$$(i.) k = l+j, k' = l+j'$$

$$(ii.) k = k + (t-k), j = j + (t-k)$$

$$(iii.) k' = (l+j) + (t-k), k = (l+j') + (t-k')$$

$$(iv.) k = l+j, k' = k + (t-k), j' = k - l$$

$$\begin{aligned}
&= \left(\sum_j c_j c_{j+l} \right)^2 v^4 + \left(\sum_k c_k c_{k+(k-l)} \right)^2 v^4 \\
&\quad + \left(\sum_j c_j c_{(l+j)+(k-l)} \sum_{j'} c_{j'} c_{(l+j')+(k-l')} \right) v^4 \\
&\quad + \left(\sum_j c_j c_{l+j} c_{(l+j)+(k-l)} c_{(l+j)+(k-l)-l} \right) \underbrace{\left(E \eta_0^4 - 3v^4 \right)}_{= K} \\
&= \gamma_\varepsilon^2(l) + \gamma_\varepsilon^2(k-l) + \gamma_\varepsilon(k-l+l) \gamma_\varepsilon(k-l-l) \\
&\quad + \left(\sum_j c_j c_{j+(k-l)} c_{j+l} c_{j+l+(k-l)} \right) K.
\end{aligned}$$

Hence,

$$\begin{aligned}
E \Delta_{q,1}^2(l) &= \underbrace{\frac{1}{(F_q)^2} \sum_{A, k' = q+l+1}^T \gamma_\varepsilon^2(k'-l)}_{= \frac{1}{(F_q)^2} \sum_r \# \{ k'-l = r \} \gamma_\varepsilon^2(r)} \\
&= \frac{1}{(F_q)^2} \sum_r \# \{ k'-l = r \} \gamma_\varepsilon^2(r) \\
&\leq \frac{1}{F_q} \sum_{r=\infty}^{\infty} \gamma_\varepsilon^2(r) \leq C/F_q \\
&\quad + \underbrace{\frac{1}{(F_q)^2} \sum_{A, k' = q+l+1}^T \gamma_\varepsilon(k'-l+l) \gamma_\varepsilon(k'-l-l)}_{= 1 \cdot 1 \leq \frac{1}{(F_q)^2} \sum_r \# \{ k'-l = r \} |\gamma_\varepsilon(r+l) \gamma_\varepsilon(r-l)|} \\
&\leq \frac{C}{F_q} \sum_{r=\infty}^{\infty} |\gamma_\varepsilon(r)| = C/F_q
\end{aligned}$$

$$+ \frac{1}{(Fq)^2} \sum_{\substack{j, k \\ j+k=q+l+r}}^T \left(\sum_j c_j c_{j+(k-l)} c_{j+r} c_{j+l+(k'-l)} \right) \cdot K$$

$$\begin{aligned} | \cdot | &\leq \frac{1}{(Fq)^2} \sum_r \#\{k'-l=r\} \sum_j |c_j c_{j+r} c_{j+r} c_{j+l+r}| \\ &\leq \frac{1}{Fq} \sum_j |c_j c_{j+r}| \sum_r |c_{j+r} c_{j+l+r}| \\ &\leq \frac{(\max_j |c_j|)^2}{Fq} \sum_j |c_j| \sum_r |c_r| \\ &\leq C_{Fq}, \end{aligned}$$

which implies that

$$\mathbb{E} \Delta_{q,\varepsilon}^2(l) \leq C_{Fq} \quad (*)$$

for $\varepsilon < 1$ with some fixed constant (C_{Fq}) does not depend on l, p, q, T . Analogous calculations show that $(*)$ holds for $\varepsilon = 2, 3, 4$ as well. As a result, we get that

$$E(\hat{\gamma}_q^*(l) - \gamma_q(l))^2 \leq C \left(\frac{1}{Fq} + \left(\frac{p}{Fq} \right)^2 \right)$$

with some fixed constant independent of l, p, q, T .

Moreover,

$$\begin{aligned} \text{ER}_{q,1}^2(l) &\leq \underbrace{\left(\frac{1}{T-q} \sum_{t=q+l+1}^T \{\Delta_{q,m_t}\}^2 \right)}_{\leq C(\eta_T)^2} \underbrace{\left(\frac{1}{T-q} \sum_{t=q+l+1}^T \mathbb{E} \{\Delta_q \varepsilon_{t,q}\}^2 \right)}_{=C} \\ &\leq C(\eta_T)^2 \end{aligned}$$

and analogously $\text{ER}_{q,2}^2(l) = C(\eta_T)^2$ as well as
 $\text{ER}_{q,3}^2(l) = C(\eta_T)^4$ with C indep. of l, p, q, T .

Putting everything together, we arrive at

$$\mathbb{E} (\hat{f}_q(l) - f_q(l))^2 \leq C \left(\frac{1}{T-q} + \left(\frac{P}{T-q} \right)^2 + (\eta_T)^2 \right)$$

with C indep. of l, p, q, T . \square

Lemma 2: For $1 \leq p \ll \sqrt{T}$ and $1 \leq q \ll \sqrt{T}$,

$$E\|\hat{f}_q - f_q\|_2 \leq C_p^{1/2} \left(\frac{1}{T_q} + \left(\frac{P}{T_q}\right)^2 + \left(\frac{q}{T}\right)^2 \right)^{1/2}$$

$$E\|\hat{\Gamma}_q - \Gamma_q\|_2 \leq C_p \left(\frac{1}{T_q} + \left(\frac{P}{T_q}\right)^2 + \left(\frac{q}{T}\right)^2 \right)^{1/2}$$

with some fixed constant C independent of P, q, T .

Proof:

$$\begin{aligned} E\|\hat{f}_q - f_q\|_2^2 &= E \left(\sum_{l=-P}^P (\hat{f}_q(l) - f_q(l))^2 \right) \\ &= \sum_{l=-P}^P E (\hat{f}_q(l) - f_q(l))^2 \\ &\leq C_p \left(\frac{1}{T_q} + \left(\frac{P}{T_q}\right)^2 + \left(\frac{q}{T}\right)^2 \right) \end{aligned}$$

by Lemma 1. Moreover, by Gershgorin's theorem,

$$\begin{aligned} \|\hat{\Gamma}_q - \Gamma_q\|_2 &= \max_{1 \leq i \leq P} \left(\sum_{j=1}^P |\hat{f}_q(i-j) - f_q(i-j)| \right) \\ &\leq \sum_{l=-P}^P |\hat{f}_q(l) - f_q(l)|, \end{aligned}$$

which immediately implies the second statement when combined with Lemma 1. \square

Lemma 3: Let $p \rightarrow \infty$ (in an arbitrary way) and suppose that $(1+\delta)p \leq q \ll \sqrt{T}$ for some small $\delta > 0$. Then

- (i.) $\|\Gamma_q^{-1}\|_2 \leq C$ for suff. large T with C indep. of p, q, T
- (ii.) $\|\hat{\Gamma}_q^{-1} - \Gamma_q^{-1}\|_2 = O_p(p/\sqrt{T})$.
- (iii) $\|\hat{\Gamma}_q^{-1}\|_2 = O_p(1)$.

Proof:

(i.) Consider $\Gamma_q = (\gamma_{q(i-j)} : 1 \leq i, j \leq p)$. Since $\gamma_q(l) = 2\gamma_\varepsilon(l) - \gamma_\varepsilon(q-l) - \gamma_\varepsilon(q+l)$ with $\gamma_\varepsilon(l) = \text{Cov}(\varepsilon_t, \varepsilon_{t+l})$, we have $\Gamma_q = 2\Gamma - R$, where $\Gamma = (\gamma_\varepsilon(i-j) : 1 \leq i, j \leq p)$ and $R = (\gamma_\varepsilon(q+(i-j)) + \gamma_\varepsilon(q-(i-j)) : 1 \leq i, j \leq p)$.

The spectral density f_ε of $\{\varepsilon_t\}$ is bounded away from zero and infinity, i.e., $0 < c_p \leq f(w) \leq C_p < \infty$ for all w and some constants c_p, C_p . This implies that the eigenvalues of Γ lie in some interval $[c_p, C_p]$ with constants $0 < c_p \leq C_p < \infty$ indep. of p [Grenander & Szegő (1958), Toeplitz Forms and Their Applications, Sec. 5.2; Brockwell & Davis (1981), Prop. 4.5.3.]

From this, it immediately follows that
 $\|\Gamma^{-1}\|_2 \leq C$ with same C indep. of P .

For general matrices A and B , it holds that
 $\tilde{A}^T - \tilde{B}^T = (A^T - B^T + B^T)(B - A)B^T$. Since
 $\|AB\|_2 \leq \|A\|_2 \|B\|_2$, this implies that

$$\|\tilde{A}^T - \tilde{B}^T\|_2 \leq (\|\tilde{A}^T - B^T\|_2 + \|B^T\|_2) \|B - A\|_2 \|B^T\|_2$$

and thus

$$\|\tilde{A}^T - \tilde{B}^T\|_2 \leq \frac{\|B^T\|_2^2 \|B - A\|_2}{1 - \|B - A\|_2 \|B^{-1}\|_2}. \quad (*)$$

Setting $A = \Gamma_q/2$ and $B = \Gamma$, we get that

$$\|\left(\Gamma_{q/2}\right)^{-1} - \Gamma^{-1}\|_2 \leq \frac{\|\Gamma^{-1}\|_2^2 \|\Gamma_{q/2} - \Gamma\|_2}{1 - \|\Gamma_{q/2} - \Gamma\|_2 \|\Gamma^{-1}\|_2} \leq C$$

for suff. large T with C indep. of P, q, T , since
 $\|\Gamma^{-1}\|_2 \leq C$ and

$$\begin{aligned} \|\Gamma_{q/2} - \Gamma\|_2 - \|R/2\|_2 &\leq \sum_{l=-P}^P |\gamma_\varepsilon(q+l) + \gamma_\varepsilon(q-l)|/2 \\ &\leq C_P \cdot \delta^{q+P} \stackrel{(+)}{=} o(1). \end{aligned}$$

From this, it is easy to conclude that $\|\Gamma_q^{-1}\|_2 \leq C$ with C indep. of P, q, T for suff. large T .

(ii.) With (*), we get that

$$\begin{aligned}\|\hat{\Gamma}_q^{-1} - \Gamma_q^{-1}\|_2 &\leq \frac{\|\Gamma_q^{-1}\|_2^2 \|\hat{\Gamma}_q - \Gamma_q\|_2}{1 - \|\hat{\Gamma}_q - \Gamma_q\|_2 \|\Gamma_q^{-1}\|_2} \\ &= O_p(P/\sqrt{F}),\end{aligned}$$

since $\|\Gamma_q^{-1}\|_2 \leq C$ by (i) and $\|\hat{\Gamma}_q - \Gamma_q\|_2 = O_p(P/\sqrt{F})$ by Lemma 2.

(iii.) follows directly from (i.) and (ii.). \square

(+) : Use rule of l'Hopital: (c) $f, g: \mathbb{R} \rightarrow \mathbb{R}$ diff.

with $g' \neq 0$ and $\lim_{x \rightarrow \infty} f(x) = \lim_{x \rightarrow \infty} g(x) = \pm \infty$.

Then $\lim_{x \rightarrow \infty} f(x)/g(x) = \lim_{x \rightarrow \infty} f'(x)/g'(x)$.

$$\text{Hoe: } P \cdot \zeta^{q-p} \leq P \cdot \zeta^{dp} = P/\zeta^{-dp}$$

$$(f(p) = p, g(p) = \zeta^{-dp})$$

$$\lim_{p \rightarrow \infty} \frac{P}{\zeta^{-dp}} = \lim_{p \rightarrow \infty} \frac{1}{-d\zeta^{-dp}} = -\frac{1}{d} \lim_{p \rightarrow \infty} \zeta^{dp} = 0.$$

Lemma 4: Let $C \log T \leq p \ll q$ and $q \ll \sqrt{T}$.
It holds that

$$\|\tilde{\alpha}_q - \alpha\|_2 = O_p(\sqrt{pT}).$$

Proof:

$$\begin{aligned}\tilde{\alpha}_q - \alpha &= \hat{\Gamma}_q^{-1} \hat{\gamma}_q - \alpha \\ &= \hat{\Gamma}_q^{-1} ((\hat{\gamma}_q - \gamma_q) + (\gamma_q - \Gamma_q \alpha) + (\Gamma_q - \hat{\Gamma}_q) \alpha)\end{aligned}$$

and thus

$$\begin{aligned}\|\tilde{\alpha}_q - \alpha\|_2 &\leq \underbrace{\|\hat{\Gamma}_q^{-1}\|_2}_{= O_p(1) \text{ by Lemma 3}} \left\{ \underbrace{\|\hat{\gamma}_q - \gamma_q\|_2}_{= O_p(\sqrt{pT}) \text{ by Lemma 2}} + \underbrace{\|\gamma_q - \Gamma_q \alpha\|_2}_{= \|\lambda_q + \varepsilon_q\|_2} \right. \\ &\quad \left. + \underbrace{\|\Gamma_q - \hat{\Gamma}_q\|_2 \alpha\|_2}_{\stackrel{(*)}{=} O_p(\sqrt{pT})} \right\} \\ &= O_p(\sqrt{pT} + \sqrt{p} \|\varepsilon^{q-p}\| + \sqrt{p} \|\varepsilon^p\|).\end{aligned}$$

It remains to prove (*):

$$\begin{aligned}
 & \|(\hat{f}_q - \hat{f}_{q'})_{\alpha}\|_2^2 \\
 &= \sum_{i=1}^P \left(\sum_{j=1}^P \{f_q(i-j) - \hat{f}_{q'}(i-j)\} a_j \right)^2 \\
 &= \sum_{j,j'=1}^P \left(\sum_{i=1}^P \{f_q(i-j) - \hat{f}_{q'}(i-j)\} \{f_q(i-j') - \hat{f}_{q'}(i-j')\} \right) a_j a_{j'} \\
 &\leq \sum_{j,j'=1}^P |a_j a_{j'}| \left(\sum_{i=1}^P \{f_q(i-j) - \hat{f}_{q'}(i-j)\}^2 \right)^{1/2} \\
 &\quad \times \left(\sum_{i=1}^P \{f_q(i-j') - \hat{f}_{q'}(i-j')\}^2 \right)^{1/2} \\
 &\leq \max_{1 \leq j \leq P} \sum_{i=1}^P \{f_q(i-j) - \hat{f}_{q'}(i-j)\}^2 \cdot \underbrace{\sum_{j,j'=1}^P |a_j a_{j'}|}_{=C} \\
 &\leq C \cdot \sum_{l=-P}^P \{f_q(l) - \hat{f}_{q'}(l)\}^2 \\
 &= O_p(P^{-1})
 \end{aligned}$$

by Lemma 1. □

Lemma 5: Let $C \log T \leq p \ll \min\{T^{1/3}, q\}$ and $q \ll \sqrt{T}$.

Then

$$\tilde{\nu}^2 = \nu^2 + O_p(\sqrt{pT}).$$

Proof:

$$\begin{aligned}\tilde{\pi}_{t,T} &= (\Delta_1 \epsilon_t - \sum_{j=1}^p \tilde{\alpha}_j \Delta_1 \epsilon_{t-j}) \\ &\quad + (\Delta_1 m_t - \sum_{j=1}^p \tilde{\alpha}_j \Delta_1 m_{t-j})\end{aligned}$$

$$\begin{aligned}\tilde{\nu}^2 &= \frac{1}{2T} \sum_{k=p+2}^T \left\{ \Delta_1 \epsilon_k - \sum_{j=1}^p \tilde{\alpha}_j \Delta_1 \epsilon_{k-j} \right\}^2 \\ &\quad + \frac{1}{2T} \sum_{k=p+2}^T \left\{ \Delta_1 m_k - \sum_{j=1}^p \tilde{\alpha}_j \Delta_1 m_{k-j} \right\}^2 \\ &\quad + \frac{1}{T} \sum_{k=p+2}^T \left\{ \Delta_1 \epsilon_k - \sum_{j=1}^p \tilde{\alpha}_j \Delta_1 \epsilon_{k-j} \right\} \left\{ \Delta_1 m_k - \sum_{j=1}^p \tilde{\alpha}_j \Delta_1 m_{k-j} \right\} \\ &=: \tilde{\nu}_1^2 + \tilde{\nu}_2^2 + \tilde{\nu}_3^2\end{aligned}$$

$$\begin{aligned}\tilde{\nu}_1^2 &= \frac{1}{2T} \sum_{k=p+2}^T \left\{ \Delta_1 \epsilon_k - \sum_{j=1}^p \alpha_j \Delta_1 \epsilon_{k-j} \right\}^2 \\ &\quad + \frac{1}{2T} \sum_{k=p+2}^T \left\{ \sum_{j=1}^p (\alpha_j - \tilde{\alpha}_j) \Delta_1 \epsilon_{k-j} \right\}^2 \\ &\quad + \frac{1}{T} \sum_{k=p+2}^T \left\{ \Delta_1 \epsilon_k - \sum_{j=1}^p \alpha_j \Delta_1 \epsilon_{k-j} \right\} \left\{ \sum_{j=1}^p (\alpha_j - \tilde{\alpha}_j) \Delta_1 \epsilon_{k-j} \right\} \\ &=: \tilde{\nu}_{1,A}^2 + \tilde{\nu}_{1,B}^2 + \tilde{\nu}_{1,C}^2.\end{aligned}$$

$$\begin{aligned}
\tilde{\nu}_{1,A}^2 &= \frac{1}{2T} \sum_{t=p+2}^T (\gamma_t - \gamma_{t-1})^2 \\
&\quad + \frac{1}{2T} \sum_{t=p+2}^T \left(\sum_{j=p+1}^{\infty} a_j \Delta_1 \epsilon_{t-j} \right)^2 \\
&\quad \underbrace{\qquad\qquad\qquad}_{E[\cdot] \leq \sum_{j=p+1}^{\infty} |a_j| |a_j| E[|\Delta_1 \epsilon_{t-j} \Delta_1 \epsilon_{t-j}|]} \\
&\quad \leq C \cdot \beta^{2P} \\
&\quad + \frac{1}{T} \sum_{t=p+2}^T (\gamma_t - \gamma_{t-1}) \left(\sum_{j=p+1}^{\infty} a_j \Delta_1 \epsilon_{t-j} \right) \\
&\quad \underbrace{\qquad\qquad\qquad}_{E|\cdot| \leq C \cdot \sum_{j=p+1}^{\infty} |a_j|} \leq C \cdot \beta^P \\
&= v^2 + O_p(1/\sqrt{T}) + O_p(\beta^P)
\end{aligned}$$

$$\begin{aligned}
\tilde{\nu}_{1,B}^2 &\leq \frac{1}{2T} \sum_{t=p+2}^T \left\{ \sum_{j=1}^P (a_j - \hat{a}_j)^2 \right\} \left\{ \sum_{j=1}^P (\Delta_1 \epsilon_{t-j})^2 \right\} \\
&= \|\hat{a}_q - a\|_2^2 \cdot O_p(P) \\
&= O_p(P^2/T).
\end{aligned}$$

$$\tilde{\nu}_{1,C}^2 = \left(\sum_{t=p+2}^T \sum_{j=1}^P (a_j - \hat{a}_j)^2 \right)^2 = C \cdot \left(\frac{P^2}{T} \right).$$

$$\begin{aligned}
\tilde{\sigma}_{ik}^2 &= \frac{1}{T} \sum_{t=p+2}^T \left\{ (\gamma_t - \gamma_{t-1}) + \sum_{j=p+1}^{\infty} \alpha_j \Delta_1 \epsilon_{t-j} \right\} \\
&\quad \times \left\{ \sum_{j=1}^p (\alpha_j - \tilde{\alpha}_j) \Delta_1 \epsilon_{t-j} \right\} \\
&= \sum_{j=1}^p (\alpha_j - \tilde{\alpha}_j) \left\{ \frac{1}{T} \sum_{t=p+2}^T (\gamma_t - \gamma_{t-1}) \Delta_1 \epsilon_{t-j} \right\} \\
&\quad + \frac{1}{T} \sum_{t=p+2}^T \left\{ \sum_{j=p+1}^{\infty} \alpha_j \Delta_1 \epsilon_{t-j} \right\} \left\{ \sum_{j=1}^p (\alpha_j - \tilde{\alpha}_j) \Delta_1 \epsilon_{t-j} \right\} \\
&\quad \underbrace{\qquad\qquad\qquad}_{1 \cdot 1} \leq \left(\frac{1}{T} \sum_{t=p+2}^T \left\{ \sum_{j=p+1}^{\infty} \alpha_j \Delta_1 \epsilon_{t-j} \right\}^2 \right)^{1/2} \\
&\quad = O_p(\sqrt{T}) \\
&\quad \times \left(\frac{1}{T} \sum_{t=p+2}^T \left\{ \sum_{j=1}^p (\alpha_j - \tilde{\alpha}_j) \Delta_1 \epsilon_{t-j} \right\}^2 \right)^{1/2} \\
&\leq \|\alpha - \tilde{\alpha}\|_2 \cdot \left(\frac{1}{T} \sum_{t=p+2}^T \left\{ \sum_{j=1}^p (\Delta_1 \epsilon_{t-j})^2 \right\} \right)^{1/2} \\
&= O_p(\sqrt{p/T}) \cdot O_p(\sqrt{p}) \\
&= O_p(\sqrt{p} \cdot p/\sqrt{T})
\end{aligned}$$

$$= \underbrace{(\alpha_1 - \tilde{\alpha}_1)}_{= O_p(1/\sqrt{T}) \text{ by}} \left\{ \frac{1}{T} \sum_{A=p+2}^T (\gamma_A - \gamma_{A-1}) \Delta_1 \varepsilon_{A-1} \right\}$$

$= O_p(1)$

is (most) similar
to those for Lemma 4

$$+ \underbrace{\sum_{j=2}^P (\alpha_j - \tilde{\alpha}_j) \left\{ \frac{1}{T} \sum_{A=p+2}^T (\gamma_A - \gamma_{A-1}) \Delta_1 \varepsilon_{t-j} \right\}}_{1.1 \leq \sum_{j=2}^P |\alpha_j - \tilde{\alpha}_j| \cdot \max_{2 \leq j \leq P} \left| \frac{1}{T} \sum_{A=p+2}^T (\gamma_A - \gamma_{A-1}) \Delta_1 \varepsilon_{t-j} \right|} + O_p(P^2 \cdot P/\sqrt{T})$$

$$\leq \sqrt{P} \|\alpha - \tilde{\alpha}\|_2 = O_p(P/\sqrt{T})$$

$$= O_p(1/\sqrt{T} + P^2/T + P^2 \cdot P/\sqrt{T}).$$

Hence,

$$\tilde{\nu}_1^2 = \nu^2 + O_p(\sqrt{P}/T).$$

(noting that $P^2/T \leq \sqrt{P}/T$ for $P \leq T^{1/3}$).

Moreover,

$$\begin{aligned}\tilde{\gamma}_2^2 &= \frac{1}{T} \sum_{t=p+2}^T \left\{ \gamma_t \left(1 + \sum_{j=1}^p |\tilde{\alpha}_j| \right) \right\}^2 \\ &\leq C/T^2 \left(1 + \sum_{j=1}^p |\tilde{\alpha}_j| \right)^2 = O_p\left(\frac{1}{T^2}\right)\end{aligned}$$

and

$$\begin{aligned}|\tilde{\gamma}_3^2| &\leq \left(\frac{1}{T} \sum_{t=p+2}^T \left\{ D_1 \varepsilon_t - \sum_{j=1}^p \tilde{\alpha}_j D_1 \varepsilon_{t-j} \right\}^2 \right)^{1/2} \\ &\quad \times \left(\frac{1}{T} \sum_{t=p+2}^T \left\{ D_1 m_t - \sum_{j=1}^p \tilde{\alpha}_j D_1 m_{t-j} \right\}^2 \right)^{1/2} \\ &= O_p\left(\frac{1}{T}\right).\end{aligned}$$

Putting everything together, we arrive at

$$\tilde{\gamma}^2 = \gamma^2 + O_p\left(\sqrt{p/T}\right).$$

□

Lemma 6: Let $(\log T \leq p \ll \min\{T^{1/3}, q\})$ and $q \ll \sqrt{T}$. Then

$$\tilde{C}^2 = C^2 + O_p(P/\sqrt{T}).$$

Proof:

$$\begin{aligned} (1 - \sum_{j=1}^P \tilde{\alpha}_j)^2 &= (1 - \underbrace{\sum_{j=1}^P \alpha_j}_{\rightarrow C^* \neq 0} + \sum_{j=1}^P (\alpha_j - \tilde{\alpha}_j))^2 \\ &= \underbrace{(1 - \sum_{j=1}^P \alpha_j)^2}_{1 \cdot 1 \leq \sqrt{P} \|\alpha - \tilde{\alpha}\|_2} \\ &\quad + 2 \cdot (1 - \sum_{j=1}^P \alpha_j) \underbrace{\left(\sum_{j=1}^P (\alpha_j - \tilde{\alpha}_j) \right)}_{= O_p(P/\sqrt{T})} \end{aligned}$$

$$\begin{aligned} &\quad + \underbrace{\left(\sum_{j=1}^P (\alpha_j - \tilde{\alpha}_j) \right)^2}_{= O_p(P^2/\sqrt{T})} \\ &= \underbrace{(1 - \sum_{j=1}^P \alpha_j)^2}_{\rightarrow C^* \neq 0} + O_p(P/\sqrt{T}). \end{aligned}$$

Thus,

$$\begin{aligned}\tilde{\sigma}^2 - \sigma^2 &= \frac{\tilde{\nu}^2}{(1 - \sum_{j=1}^P \tilde{\alpha}_j)^2} - \frac{\nu^2}{(1 - \sum_{j=1}^P \alpha_j)^2} \\&= \underbrace{\frac{\tilde{\nu}^2 - \nu^2}{(1 - \sum_{j=1}^P \tilde{\alpha}_j)^2}}_{= O_p(\sqrt{P})} + \nu^2 \left(\underbrace{\frac{1}{(1 - \sum_{j=1}^P \tilde{\alpha}_j)^2} - \frac{1}{(1 - \sum_{j=1}^P \alpha_j)^2}}_{= \frac{(1 - \sum_{j=1}^P \alpha_j)^2 - (1 - \sum_{j=1}^P \tilde{\alpha}_j)^2}{(1 - \sum_{j=1}^P \alpha_j)(1 - \sum_{j=1}^P \tilde{\alpha}_j)}} \right) \\&= O_p(P/\sqrt{P})\end{aligned}$$

□

Lemma 7: (e) the estimators \tilde{c}_j be defined by the recursions $\tilde{c}_0 = 1$ and

$$\tilde{c}_j = \sum_{k=1}^j \hat{\alpha}_k \tilde{c}_{j-k} \text{ for } j \geq 1.$$

If $(\log T \leq p \ll \min\{T^{1/3}, q\})$ and $q \ll \sqrt{T}$, then

$$\max_{1 \leq j \leq p} |\tilde{c}_j - c_j| = O_p(p/\sqrt{T}).$$

Proof:

By assumption, $A(z) \neq 0$ for all $|z| < 1+d$. Hence, $\tilde{A}'(z)$ is analytic (complex diff.) on $D = \{z \in \mathbb{C} : |z| < 1+d\}$ and has a unique Laurent series representation in the ring segment $R = \{z \in \mathbb{C} : 0 < d' < |z| < 1+d\}$ of the form $\tilde{A}'(z) = \sum_{j=-\infty}^{\infty} \phi_j z^j$

with

$$\phi_j = \frac{1}{2\pi i} \int_{|S|=1} \frac{\tilde{A}'(\xi)}{\xi^{j+1}} d\xi.$$

As

$$\tilde{A}'(z) = \sum_{j=1}^{\infty} c_j z^j \text{ for } |z| < 1+d,$$

The uniqueness of the Laurent series implies that

$$c_j = \frac{1}{2\pi i} \int_{|S|=1} \frac{\tilde{A}'(\xi)}{\xi^{j+1}} d\xi.$$

By Prop. 5.1.1 and Problems 7.11 and 8.7 in Breckwell & Davis, we have the following:

If $\hat{f}_g(0) > 0$, then

$$\tilde{A}(z) = 1 - \sum_{j=1}^P \tilde{a}_j z^j + O \quad \forall |z| \leq 1,$$

which implies (since $\tilde{A}(z)$ cannot have more than P roots) that

$$\tilde{A}(z) = 1 - \sum_{j=1}^P \tilde{a}_j z^j + O \quad \forall |z| < 1 + \tilde{\delta},$$

where $\tilde{\delta} = \tilde{\delta}(\tilde{a}_1, \dots, \tilde{a}_P) > 0$ is random.

Since $\hat{f}_g(0) > 0$ holds with prob. approaching 1, we obtain that

$$\tilde{A}(z) = 1 - \sum_{j=1}^P \tilde{a}_j z^j + O \quad \forall |z| < 1 + \tilde{\delta}$$

with $\tilde{\delta} > 0$ with prob. approaching 1.

By construction of \tilde{c}_j , it holds that

$$\tilde{A}'(z) = \sum_{j=0}^{\infty} \tilde{c}_j z^j \quad \text{for } |z| < 1 + \tilde{\delta}.$$

As before, we can argue that

$$\tilde{c}_j = \frac{1}{2\pi i} \int_{|S|=1} \frac{\tilde{A}'(s)}{s^{j+1}} ds \quad \text{for } j \geq 0$$

with prob. approaching 1.

It holds that

$$\begin{aligned}
 & \max_{1 \leq j \leq P} |\tilde{c}_j - c_j| \\
 &= \max_{1 \leq j \leq P} \left| \frac{1}{2\pi n} \int_{|\beta|=1} \frac{\tilde{A}'(\beta) - A'(\beta)}{\beta^{j+1}} d\beta \right| \\
 &\leq \max_{1 \leq j \leq P} \frac{1}{2\pi} \int_{|\beta|=1} \underbrace{\frac{|\tilde{A}'(\beta) - A'(\beta)|}{|\beta|^{j+1}}}_{< 1} d\beta \\
 &\leq \sup_{|\beta|=1} |\tilde{A}'(\beta) - A'(\beta)| \cdot 2\pi \\
 &\leq \sup_{|\beta|=1} |\tilde{A}'(\beta) - A'(\beta)| \\
 &= \sup_{|\beta|=1} \left| \frac{1}{1 - \sum_{j=1}^P \tilde{a}_j \beta^j} - \frac{1}{1 - \sum_{j=1}^{\infty} a_j \beta^j} \right| \\
 &= \sup_{|\beta|=1} \frac{\left| \sum_{j=1}^P (\tilde{a}_j - a_j) \beta^j - \sum_{j=P+1}^{\infty} a_j \beta^j \right|}{\underbrace{|1 - \sum_{j=1}^P \tilde{a}_j \beta^j| |1 - \sum_{j=1}^{\infty} a_j \beta^j|}_{\geq C^* > 0}} \\
 &\leq \left(\underbrace{\sum_{j=1}^P |\tilde{a}_j - a_j| + \sum_{j=P+1}^{\infty} |a_j|} \right) \cdot \frac{1}{C^*} \cdot \sup_{|\beta|=1} \frac{1}{|1 - \sum_{j=1}^P \tilde{a}_j \beta^j|} \\
 &= O_p(P/\sqrt{f} + \xi^P) \quad =: (*)
 \end{aligned}$$

with

$$\begin{aligned} (*) &= \sup_{|S|=l} \frac{1}{|1 - \sum_{j=1}^P \tilde{\alpha}_j \xi_j^S|} \\ &= \sup_{|S|=l} \frac{1}{|1 - \sum_{j=1}^P \alpha_j \xi_j^S + \sum_{j=p+1}^P \alpha_j \xi_j^S + \sum_{j=1}^P (\alpha_j - \tilde{\alpha}_j) \xi_j^S|} \\ &\leq \sup_{|S|=l} \frac{1}{\underbrace{|1 - \sum_{j=1}^P \alpha_j \xi_j^S|}_{\geq c^* > 0} - \underbrace{|\sum_{j=p+1}^P \alpha_j|}_{\substack{\leq c^{*/2} \text{ for } p \\ \text{large enough}}} - \underbrace{|\sum_{j=1}^P (\alpha_j - \tilde{\alpha}_j)|}_{O_p(P/\sqrt{P})}} \\ &= O_p(1). \end{aligned}$$

Hence,

$$\max_{1 \leq j \leq P} |\tilde{c}_j - c_j| = O_p(P/\sqrt{P} + \xi^P). \quad \square$$

Lemma 8: Let $(\log T \leq p \ll \min\{T^{1/3}, q\})$, $q \ll \sqrt{T}$ and $r = (1+d)p$ for some small $d > 0$. Then

$$\|\hat{a}_r - a\|_2 = O_p(p\sqrt{p_f}),$$

where $\hat{a}_r = \hat{\Gamma}_r^{-1}(\hat{f}_r - \hat{d}_r)$.

Proof:

As in the proof of Lemma 4,

$$\begin{aligned} \|\hat{a}_r - a\|_2 &\leq \underbrace{\|\hat{\Gamma}_r^{-1}\|_2}_{=O_p(1)} \left\{ \underbrace{\|\hat{f}_r - f_r\|_2}_{=O_p(\sqrt{p_f})} + \underbrace{\|f_r - \Gamma_r a - \hat{d}_r\|_2}_{=: (*)} \right. \\ &\quad \left. + \underbrace{\|(\Gamma_r - \hat{\Gamma}_r)a\|_2}_{=O_p(\sqrt{p_f})} \right\}. \end{aligned}$$

Noting that $f_r - \Gamma_r a = d_r + r_r$,

$$(*) \leq \underbrace{\|\hat{d}_r - d_r\|_2}_{=O_p(p\sqrt{p_f})} + \underbrace{\|r_r\|_2}_{\leq C\sqrt{p}\beta^p} = O_p(p\sqrt{p_f} + \sqrt{p}\beta^p).$$

by Lemma 7

Hence,

$$\|\hat{a}_r - a\|_2 = O_p(p\sqrt{p_f}).$$

Repeating the arguments from Lemmas 5 and 6,
 one can finally prove the following: Let
 $C \log T \leq p \ll \min\{T^{1/5}, q\}$, $q \ll \sqrt{T}$ and
 $r = (1+\delta) p$ for some small $\delta > 0$. Then

$$\hat{v}^2 = v^2 + O_p(p \sqrt{pq})$$

$$\hat{c}^2 = c^2 + O_p(p^2/\sqrt{T}).$$