

# Online Supplement to “Asymptotics for Time–Varying Vector MA( $\infty$ ) Processes”

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In this appendix, we first discuss some impulse response analyses in Appendix B.1, and then present several preliminary lemmas in Appendix B.2, which are helpful to the development of the main results. The omitted proofs of the main results and the proofs of the preliminary lemmas are given in Appendices B.3 and B.4, respectively.

## B.1 Impulse Responses for Time–Varying VARX Models

In this subsection, we focus on the marginal impact of changes in the exogenous variables, which is measured by the impulse response functions, and is of great interest in VAR analysis. We first note that after some trivial but tedious development, successive substitution for lagged  $\mathbf{y}_t$ 's of (4.1) gives

$$\mathbf{y}_t = \mathbf{B}_0(\tau_t)\mathbf{x}_t + \sum_{j=1}^{\infty} \bar{\mathbf{J}} \prod_{k=0}^{j-1} \boldsymbol{\Gamma}(\tau_{t-k}) \boldsymbol{\Upsilon}(\tau_{t-j}) \mathbf{x}_{t-j} + \boldsymbol{\eta}_t + \sum_{j=1}^{\infty} \bar{\mathbf{J}} \prod_{k=0}^{j-1} \boldsymbol{\Gamma}(\tau_{t-k}) \bar{\mathbf{J}}^\top \boldsymbol{\eta}_{t-j},$$

where  $\bar{\mathbf{J}} = [\mathbf{I}_d, \mathbf{0}_{d \times (d(p-1)+mq)}]$ ,  $\boldsymbol{\Upsilon}(\tau) = (\mathbf{B}_0(\tau)^\top, \underbrace{\mathbf{0}_{d \times m}^\top, \dots, \mathbf{0}_{d \times m}^\top}_{p-1 \text{ matrices}}, \mathbf{I}_m, \underbrace{\mathbf{0}_m, \dots, \mathbf{0}_m}_{q-1 \text{ matrices}})^\top$

$$\boldsymbol{\Gamma}(\tau) = \begin{pmatrix} \mathbf{A}_1(\tau) & \cdots & \mathbf{A}_{p-1}(\tau) & \mathbf{A}_p(\tau) & \mathbf{B}_1(\tau) & \cdots & \mathbf{B}_{q-1}(\tau) & \mathbf{B}_q(\tau) \\ \mathbf{I}_d & \cdots & \mathbf{0}_d & \mathbf{0}_d & \mathbf{0}_{d \times m} & \cdots & \mathbf{0}_{d \times m} & \mathbf{0}_{d \times m} \\ \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \mathbf{0}_d & \cdots & \mathbf{I}_d & \mathbf{0}_d & \mathbf{0}_{d \times m} & \cdots & \mathbf{0}_{d \times m} & \mathbf{0}_{d \times m} \\ & & & & \mathbf{0}_m & \cdots & \mathbf{0}_m & \mathbf{0}_m \\ & & & & \mathbf{I}_m & \cdots & \mathbf{0}_m & \mathbf{0}_m \\ & & \mathbf{0} & & \vdots & \ddots & \vdots & \vdots \\ & & & & \mathbf{0}_m & \cdots & \mathbf{I}_m & \mathbf{0}_m \end{pmatrix}.$$

Then it is easy to see that

$$\mathbf{D}_{j,t}^{\mathbf{x}} \equiv \bar{\mathbf{J}} \prod_{k=0}^{j-1} \boldsymbol{\Gamma}(\tau_{t-k}) \boldsymbol{\Upsilon}(\tau_{t-j}) \quad \text{with } j = 1, 2, \dots \quad (\text{B.1})$$

are the impulse response functions associated with the exogenous shocks. We can prove that  $\mathbf{D}_{j,t}^{\mathbf{x}}$  can

be approximated by  $\mathbf{D}_j^{\mathbf{x}}(\tau)$  with negligible errors, where

$$\mathbf{D}_j^{\mathbf{x}}(\tau) = \bar{\mathbf{J}}\mathbf{\Gamma}^j(\tau)\mathbf{\Upsilon}(\tau). \quad (\text{B.2})$$

In addition, we define the nonparametric estimator of (B.2) as

$$\widehat{\mathbf{D}}_j^{\mathbf{x}}(\tau) = \bar{\mathbf{J}}\widehat{\mathbf{\Gamma}}^j(\tau)\widehat{\mathbf{\Upsilon}}(\tau), \quad (\text{B.3})$$

in which we replace  $\mathbf{A}_j(\tau)$  and  $\mathbf{B}_j(\tau)$  of  $\mathbf{\Gamma}(\tau)$  and  $\mathbf{\Upsilon}(\tau)$  of (B.2) with the corresponding nonparametric estimators obtained from (4.5).

The next theorem summarizes the asymptotic properties associated with (B.2) and (B.3).

**Theorem B.1.** *Suppose that Assumptions 4 and 5 hold. As  $T \rightarrow \infty$ ,  $\sup_{t \geq 1} \|\mathbf{D}_{j,t}^{\mathbf{x}} - \mathbf{D}_j^{\mathbf{x}}(\tau_t)\| = O(T^{-1})$ . Moreover, for  $\forall \tau \in (0, 1)$ ,*

$$\sqrt{T}h \left( \text{vec}[\widehat{\mathbf{D}}_j^{\mathbf{x}}(\tau) - \mathbf{D}_j^{\mathbf{x}}(\tau)] - \frac{1}{2}h^2\tilde{c}_2\mathbf{D}_j^{\mathbf{x},(2)}(\tau) \right) \rightarrow_D N(\mathbf{0}, \mathbf{\Delta}_{\mathbf{D}_j^{\mathbf{x}}}(\tau)),$$

where  $\mathbf{D}_j^{\mathbf{x},(2)}(\tau) = \mathbf{G}_j(\tau)\boldsymbol{\beta}^{(2)}(\tau)$ ,  $\mathbf{\Delta}_{\mathbf{D}_j^{\mathbf{x}}}(\tau) = \mathbf{G}_j(\tau)\mathbf{V}(\tau)\mathbf{G}_j^{\top}(\tau)$ ,  $\mathbf{G}_0(\tau) = (\mathbf{0}_{dm \times (d^2p+dmq)}, \mathbf{I}_{dm})$ , and for  $j \geq 1$

$$\mathbf{G}_j(\tau) = \left( \sum_{k=0}^{j-1} \mathbf{\Upsilon}(\tau)^{\top} (\mathbf{\Gamma}(\tau)^{\top})^{j-1-k} \otimes \bar{\mathbf{J}}\mathbf{\Gamma}^k(\tau)\bar{\mathbf{J}}^{\top}, \mathbf{I}_m \otimes \bar{\mathbf{J}}\mathbf{\Gamma}^j(\tau)\bar{\mathbf{J}}^{\top} \right).$$

We now comment on how to construct the confidence interval. By Theorem 4.1, it is straightforward to have

$$\widehat{\mathbf{\Delta}}_{\mathbf{D}_j^{\mathbf{x}}}(\tau) \rightarrow_P \mathbf{\Delta}_{\mathbf{D}_j^{\mathbf{x}}}(\tau), \quad (\text{B.4})$$

where  $\widehat{\mathbf{\Delta}}_{\mathbf{D}_j^{\mathbf{x}}}(\tau)$  has a form identical to  $\mathbf{\Delta}_{\mathbf{D}_j^{\mathbf{x}}}(\tau)$  but replacing  $\mathbf{\Gamma}(\tau)$ ,  $\mathbf{\Upsilon}(\tau)$  and  $\mathbf{V}(\tau)$  with their estimators, respectively.

## B.2 Preliminary Lemmas

**Lemma B.1.** *Suppose  $\{Z_t, \mathcal{F}_t\}$  is a martingale difference sequence,  $S_T = \sum_{t=1}^T Z_t$ ,  $U_T = \sum_{t=1}^T Z_t^2$  and  $s_T^2 = E(U_T^2) = E(S_T^2)$ . If  $s_T^{-2}U_T^2 \rightarrow_P 1$  and  $\sum_{t=1}^T E[Z_{T,t}^2 I(|Z_{T,t}| > \nu)] \rightarrow 0$  for any  $\nu > 0$  with  $Z_{T,t} = s_T^{-1}Z_t$ , then as  $T \rightarrow \infty$ ,  $s_T^{-1}S_T \rightarrow_D N(0, 1)$ .*

Lemma B.1 is Corollary 3.1 of Hall and Heyde (1980).

**Lemma B.2.** *Let  $\{Z_t, \mathcal{F}_t\}$  be a martingale difference sequence. Suppose that  $|Z_t| \leq M$  for a constant*

$M, t = 1, \dots, T$ . Let  $V_T = \sum_{t=1}^T \text{Var}(Z_t | \mathcal{F}_{t-1}) \leq V$  for some  $V > 0$ . Then for any given  $\nu > 0$ ,

$$\Pr \left( \left| \sum_{t=1}^T Z_t \right| > \nu \right) \leq \exp \left\{ -\frac{\nu^2}{2(V + M\nu)} \right\}.$$

Lemma B.2 is Proposition 2.1 of Freedman (1975).

**Lemma B.3.** *The following algebraic decompositions hold true.*

1.  $\mathbb{B}_t(L) = \sum_{j=0}^{\infty} \mathbf{B}_{j,t} L^j$  can be decomposed as  $\mathbb{B}_t(L) = \mathbb{B}_t(1) - (1-L)\tilde{\mathbb{B}}_t(L)$ , where  $\tilde{\mathbb{B}}_t(L) = \sum_{j=0}^{\infty} \tilde{\mathbf{B}}_{j,t} L^j$  and  $\tilde{\mathbf{B}}_{j,t} = \sum_{k=j+1}^{\infty} \mathbf{B}_{k,t}$ .
2.  $\mathbb{B}_t^r(L) = \sum_{j=0}^{\infty} (\mathbf{B}_{j+r,t} \otimes \mathbf{B}_{j,t}) L^j$  can be decomposed as  $\mathbb{B}_t^r(L) = \mathbb{B}_t^r(1) - (1-L)\tilde{\mathbb{B}}_t^r(L)$ , where  $\tilde{\mathbb{B}}_t^r(L) = \sum_{j=0}^{\infty} \tilde{\mathbf{B}}_{j,t}^r L^j$  and  $\tilde{\mathbf{B}}_{j,t}^r = \sum_{k=j+1}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t})$ .

In addition, let Assumption 1 hold, then

3.  $\sup_{t \geq 1} \sum_{j=0}^{\infty} \|\tilde{\mathbf{B}}_{j,t}\| < \infty$ ;
4.  $\lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \|\tilde{\mathbb{B}}_{t+1}(1) - \tilde{\mathbb{B}}_t(1)\| < \infty$ ;
5.  $\sup_{t \geq 1} \sum_{j=0}^{\infty} \|\tilde{\mathbf{B}}_{j,t}^r\| < \infty$ ;
6.  $\sup_{t \geq 1} \sum_{r=1}^{\infty} \|\tilde{\mathbb{B}}_t^r(1)\| < \infty$ ;
7.  $\lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{r=0}^{\infty} \|\tilde{\mathbb{B}}_{t+1}^r(1) - \tilde{\mathbb{B}}_t^r(1)\| < \infty$ .

**Lemma B.4.** *Let Assumptions 1 and 2 hold, and let  $\{\mathbf{W}_{T,t}(\cdot)\}_{t=1}^T$  be a sequence of  $m \times d$  matrices of functions, where  $m \geq 1$  is fixed, and each functional component is Lipschitz continuous and defined on a compact set  $[a, b]$ . Moreover, suppose that (1)  $\sup_{\tau \in [a, b]} \sum_{t=1}^T \|\mathbf{W}_{T,t}(\tau)\| = O(1)$ , and (2)  $T^{\frac{2}{5}} d_T \log T \rightarrow 0$ , where  $d_T = \sup_{\tau \in [a, b], t \geq 1} \|\mathbf{W}_{T,t}(\tau)\|$ . As  $T \rightarrow \infty$ ,  $\sup_{\tau \in [a, b]} \left\| \sum_{t=1}^T \mathbf{W}_{T,t}(\tau) \mathbb{B}_t(1) \boldsymbol{\epsilon}_t \right\| = O_P(\sqrt{d_T \log T})$ .*

**Lemma B.5.** *Let the conditions of Lemma B.4 hold. Suppose  $T^{\frac{4}{5}} d_T \log T \rightarrow 0$ ,  $\sup_{t \geq 1} E[\|\boldsymbol{\epsilon}_t\|^4 | \mathcal{F}_{t-1}] < \infty$  a.s. and  $\sup_{\tau \in [a, b]} \sum_{t=1}^{T-1} \|\mathbf{W}_{T,t+1}(\tau) - \mathbf{W}_{T,t}(\tau)\| = O(d_T)$ . As  $T \rightarrow \infty$*

1.  $\sup_{\tau \in [a, b]} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(\tau)) \mathbb{B}_t^0(1) (\text{vec}[\boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top] - \text{vec}[\mathbf{I}_d]) \right\| = O_P(\sqrt{d_T \log T})$ ;
2.  $\sup_{\tau \in [a, b]} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(\tau)) \boldsymbol{\zeta}_t \boldsymbol{\epsilon}_t \right\| = O_P(\sqrt{d_T \log T})$ ;

where  $\boldsymbol{\zeta}_t = \sum_{r=1}^{\infty} \sum_{s=0}^{\infty} \{\mathbf{B}_{s+r,t} \boldsymbol{\epsilon}_{t-r}\} \otimes \mathbf{B}_{s,t}$ .

**Lemma B.6.** *Under Assumption 5, there exists a time-varying VMA( $\infty$ ) process*

$$\tilde{\mathbf{y}}_t = \boldsymbol{\mu}^*(\tau_t) + \sum_{j=0}^{\infty} \mathbf{D}_j^\epsilon(\tau_t) \boldsymbol{\epsilon}_{t-j} + \sum_{j=0}^{\infty} \mathbf{D}_j^v(\tau_t) \mathbf{v}_{t-j}$$

such that  $\max_{t \geq 1} \{E \|\mathbf{y}_t - \tilde{\mathbf{y}}_t\|^\delta\}^{1/\delta} = O(T^{-1})$ , where

$$\begin{aligned} \boldsymbol{\mu}^*(\tau) &= \sum_{j=0}^{\infty} \boldsymbol{\Psi}_j(\tau) \left( \boldsymbol{\mu}(\tau) + \sum_{l=0}^q \mathbf{B}_l(\tau) \mathbf{g}(\tau) \right), \quad \boldsymbol{\Psi}_j(\tau) = \mathbf{J} \boldsymbol{\Phi}^j(\tau) \mathbf{J}^\top, \\ \boldsymbol{\Phi}(\tau) &= \begin{pmatrix} \mathbf{A}_1(\tau) & \cdots & \mathbf{A}_{p-1}(\tau) & \mathbf{A}_p(\tau) \\ \mathbf{I}_d & \cdots & \mathbf{0}_d & \mathbf{0}_d \\ \vdots & \ddots & \vdots & \vdots \\ \mathbf{0}_d & \cdots & \mathbf{I}_d & \mathbf{0}_d \end{pmatrix}, \quad \mathbf{J} = [\mathbf{I}_d, \mathbf{0}_{d \times d(p-1)}], \\ \mathbf{D}_j^\epsilon(\tau) &= \boldsymbol{\Psi}_j(\tau) \boldsymbol{\omega}(\tau), \quad \mathbf{D}_j^v(\tau) = \sum_{b=\max(0, j-q)}^j \mathbf{D}_{b, j-b}^v(\tau), \\ \mathbf{D}_{j,l}^v(\tau) &= \sum_{k=0}^j \boldsymbol{\Psi}_k(\tau) \mathbf{B}_l(\tau) \mathbf{C}_{j-k}(\tau). \end{aligned}$$

Moreover,  $\tilde{\mathbf{y}}_t$  and  $\mathbf{x}_t$  admit the following expression

$$\begin{pmatrix} \tilde{\mathbf{y}}_t \\ \mathbf{x}_t \end{pmatrix} = \begin{pmatrix} \boldsymbol{\mu}^*(\tau_t) \\ \mathbf{g}(\tau_t) \end{pmatrix} + \sum_{j=0}^{\infty} \mathbf{D}_j(\tau_t) \begin{pmatrix} \boldsymbol{\epsilon}_{t-j} \\ \mathbf{v}_{t+1-j} \end{pmatrix},$$

where  $\mathbf{D}_j(\tau) = \begin{pmatrix} \mathbf{D}_j^\epsilon(\tau) & \mathbf{D}_{j-1}^v(\tau) \\ \mathbf{0} & \mathbf{C}_{j-1}(\tau) \end{pmatrix}$ , and  $\mathbf{D}_j^v(\tau) = 0$  and  $\mathbf{C}_j(\tau) = 0$  for  $j < 0$ . Here, all  $\mathbf{D}_j(\cdot)$  satisfy the same conditions as those in Assumption 3.

**Lemma B.7.** Let Assumptions 4–5 hold. Suppose further  $\sup_{t \geq 1} E \left[ \|\mathbf{e}_t\|^4 | \mathcal{F}_{t-1} \right] < \infty$  a.s. and  $\frac{T^{1-4/\delta} h}{\log T} \rightarrow \infty$  as  $T \rightarrow \infty$ ,

1.  $\sup_{\tau \in [h, 1-h]} \left\| \frac{1}{T} \mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \mathbf{Z}_{2,\tau} - \boldsymbol{\Sigma}_{\mathbf{Z}_2}(\tau) \otimes \boldsymbol{\Lambda}_1 \right\| = O_P \left( h^2 + \sqrt{\frac{\log T}{Th}} \right)$ , where  $\boldsymbol{\Lambda}_1 = \text{diag}(\tilde{c}_0, \tilde{c}_2)$  and  $\boldsymbol{\Sigma}_{\mathbf{Z}_2}(\tau) = \boldsymbol{\Sigma}_{\mathbf{z}_2}(\tau) \otimes \mathbf{I}_d$ ;
2.  $\sup_{\tau \in [h, 1-h]} \left\| \frac{1}{T} \mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \mathbf{Z}_1 - \boldsymbol{\Sigma}_{\mathbf{Z}_{1,2}}^\top(\tau) \otimes \boldsymbol{\Lambda}_2 \right\| = O_P \left( h^2 + \sqrt{\frac{\log T}{Th}} \right)$ , where  $\boldsymbol{\Lambda}_2 = [\tilde{c}_0, 0]^\top$  and  $\boldsymbol{\Sigma}_{\mathbf{Z}_{1,2}}(\tau) = \boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}(\tau) \otimes \mathbf{I}_d$ ;
3.  $\sup_{\tau \in [0, 1]} \left\| \frac{1}{T} \mathbf{Z}_\tau \mathbf{K}_\tau \boldsymbol{\eta} \right\| = O_P \left( \sqrt{\frac{\log T}{Th}} \right)$ , where  $\boldsymbol{\eta} = (\boldsymbol{\eta}_1, \dots, \boldsymbol{\eta}_T)^\top$ ;
4.  $\sup_{\tau \in [0, 1]} \|\hat{\boldsymbol{\beta}}(\tau) - \boldsymbol{\beta}(\tau)\| = O_P(h^2 + \sqrt{\log T / (Th)})$ ; and
5.  $\sup_{\tau \in [0, 1]} \left\| \hat{\boldsymbol{\Omega}}(\tau) - \frac{1}{T} \sum_{t=1}^T \boldsymbol{\Omega}(\tau_t) K_h(\tau_t - \tau) \right\| = O_P \left( h^2 + \sqrt{\frac{\log T}{Th}} \right)$ .

**Lemma B.8.** Suppose Assumptions 4–6 hold. As  $T \rightarrow \infty$ ,

1.  $T^{-1} \sum_{t=1}^T \tilde{\mathbf{Z}}_{1t}^\top \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \tilde{\mathbf{Z}}_{1t}^\top \rightarrow_P \int_0^1 (\boldsymbol{\Sigma}_{\mathbf{z}_1}(\tau) - \boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}(\tau) \boldsymbol{\Sigma}_{\mathbf{z}_2}^{-1}(\tau) \boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}^\top(\tau)) \otimes \boldsymbol{\Omega}^{-1}(\tau) d\tau$ ;
2.  $\sum_{t=1}^T \tilde{\mathbf{Z}}_{1t}^\top \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \left( \mathbf{Z}_{2t}^\top \boldsymbol{\theta}(\tau_t) - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) [\boldsymbol{\theta}(\tau_1)^\top \mathbf{Z}_{21}, \dots, \boldsymbol{\theta}(\tau_T)^\top \mathbf{Z}_{2T}]^\top \right) = o_P(\sqrt{T})$ ;

3.  $\sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \boldsymbol{\eta} = o_P(\sqrt{T})$ ; and

4.  $\sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \boldsymbol{\eta}_t = \sum_{t=1}^T \left( \mathbf{Z}_{1t} - \boldsymbol{\Sigma}_{\mathbf{Z}_{1,2}}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{Z}_2}^{-1}(\tau_t) \mathbf{Z}_{2t} \right) \boldsymbol{\Omega}^{-1}(\tau_t) \boldsymbol{\eta}_t + o_P(\sqrt{T})$ .

### B.3 Omitted Proofs of the Main Results

*Proof of Proposition 2.1.*

(1). Let  $\rho$  denote the largest eigenvalue of  $\boldsymbol{\Phi}_t$  uniformly over  $t$ . Then,  $\rho < 1$  by the condition in Proposition 2.1. Similar to the proof of Proposition 2.4 in Dahlhaus and Polonik (2009), we have  $\max_{t \geq 1} \left\| \prod_{i=0}^{j-1} \boldsymbol{\Phi}_{t-i} \right\| \leq M \rho^j$ , which yields that

$$\max_{t \geq 1} \sum_{j=1}^{\infty} j \|\mathbf{B}_{j,t}\| = \max_{t \geq 1} \sum_{j=1}^{\infty} j \left\| \mathbf{J} \prod_{i=0}^{j-1} \boldsymbol{\Phi}_{t-i} \mathbf{J}^\top \right\| \leq M \sum_{j=1}^{\infty} j \rho^j = O(1).$$

In addition, for any conformable matrices  $\{\mathbf{A}_i\}$  and  $\{\mathbf{B}_i\}$ , since

$$\prod_{i=1}^r \mathbf{A}_i - \prod_{i=1}^r \mathbf{B}_i = \sum_{j=1}^r \left( \prod_{k=1}^{j-1} \mathbf{A}_k \right) (\mathbf{A}_j - \mathbf{B}_j) \left( \prod_{k=j+1}^r \mathbf{B}_k \right),$$

we then obtain that

$$\begin{aligned} & \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \left\| \mathbf{J} \left( \prod_{i=0}^{j-1} \boldsymbol{\Phi}_{t+1-i} - \prod_{i=0}^{j-1} \boldsymbol{\Phi}_{t-i} \right) \mathbf{J}^\top \right\| \\ &= \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \left\| \mathbf{J} \sum_{m=1}^j \left( \prod_{k=1}^{m-1} \boldsymbol{\Phi}_{t+2-k} \right) (\boldsymbol{\Phi}_{t+2-m} - \boldsymbol{\Phi}_{t+1-m}) \left( \prod_{k=m}^j \boldsymbol{\Phi}_{t+1-k} \right) \mathbf{J}^\top \right\| \\ &\leq M \sum_{j=1}^{\infty} j^2 \rho^{j-1} \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \|\boldsymbol{\Phi}_{t+1} - \boldsymbol{\Phi}_t\| = O(1) \end{aligned}$$

given the condition in Proposition 2.1. Based on the above two results, we have

$$\max_{t \geq 1} \sum_{b=1}^{\infty} b \|\mathbf{D}_{b,t}\| \leq M \max_{t \geq 1} \sum_{b=1}^{\infty} b \sum_{j=b-q}^b \|\mathbf{B}_{j,t}\| \leq M \sum_{b=1}^{\infty} b \rho^b = O(1).$$

We also have

$$\begin{aligned} & \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{b=1}^{\infty} b \|\mathbf{D}_{b,t+1} - \mathbf{D}_{b,t}\| \leq \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{b=1}^{\infty} b \sum_{j=\max(0,b-q)}^b \|\mathbf{B}_{j,t+1} - \mathbf{B}_{j,t}\| \|\boldsymbol{\Theta}_{b-j,t+1-j}\| \\ &+ \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{b=1}^{\infty} b \sum_{j=\max(0,b-q)}^b \|\mathbf{B}_{j,t}\| \|\boldsymbol{\Theta}_{b-j,t+1-j} - \boldsymbol{\Theta}_{b-j,t-j}\| \\ &\leq \max_{m,t} \|\boldsymbol{\Theta}_{m,t}\| \cdot q \cdot \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{b=1}^{\infty} b \|\mathbf{B}_{b,t+1} - \mathbf{B}_{b,t}\| \\ &+ \left( \max_{t \geq 1} \sum_{b=1}^{\infty} b \sum_{j=\max(0,b-q)}^b \|\mathbf{B}_{j,t}\| \right) \cdot \left( \max_m \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \|\boldsymbol{\Theta}_{m,t+1} - \boldsymbol{\Theta}_{m,t}\| \right) = O(1). \end{aligned}$$

(2). By part (1) and the condition of Proposition 2.1, it suffices to show that  $\max_{t \geq 1} \sum_{j=1}^{\infty} j \|\mathbf{D}_{j,t}\| < \infty$  and  $\lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \|\mathbf{D}_{j,t+1} - \mathbf{D}_{j,t}\| < \infty$ . Write

$$\begin{aligned} \max_{t \geq 1} \sum_{j=1}^{\infty} j \|\mathbf{D}_{j,t}\| &\leq M \max_{t \geq 1} \sum_{j=1}^{\infty} j \sum_{k=0}^j \|\mathbf{B}_{k,t}\| \|\mathbf{C}_{j-k,t-k}\| \\ &= M \max_{t \geq 1} \sum_{k=0}^{\infty} \sum_{j=k}^{\infty} j \|\mathbf{B}_{k,t}\| \|\mathbf{C}_{j-k,t-k}\| = M \max_{t \geq 1} \sum_{k=0}^{\infty} \|\mathbf{B}_{k,t}\| \sum_{j=0}^{\infty} (k+j) \|\mathbf{C}_{j,t-k}\| \\ &= M \max_{t \geq 1} \sum_{j=0}^{\infty} j \|\mathbf{B}_{j,t}\| \sum_{k=1}^{\infty} \|\mathbf{C}_{k,t-j}\| + M \max_{t \geq 1} \sum_{k=0}^{\infty} \|\mathbf{B}_{k,t}\| \sum_{j=0}^{\infty} j \|\mathbf{C}_{j,t-k}\| = O(1). \end{aligned}$$

In addition,

$$\begin{aligned} \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \|\mathbf{D}_{j,t+1} - \mathbf{D}_{j,t}\| &\leq \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \sum_{k=0}^j \|\mathbf{B}_{k,t}\| \cdot \|\Theta_{t-k}\| \cdot \|\mathbf{C}_{j-k,t+1-k} - \mathbf{C}_{j-k,t-k}\| \\ &+ \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \sum_{k=0}^j \|\mathbf{B}_{k,t}\| \cdot \|\Theta_{t+1-k} - \Theta_{t-k}\| \cdot \|\mathbf{C}_{j-k,t+1-k}\| \\ &+ \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \sum_{k=0}^j \|\mathbf{B}_{k,t+1} - \mathbf{B}_{k,t}\| \cdot \|\Theta_{t+1-k}\| \cdot \|\mathbf{C}_{j-k,t+1-k}\| := M_{T,1} + M_{T,2} + M_{T,3}. \end{aligned}$$

We need only to show that  $M_{T,1}$  is bounded below, as the proofs of  $M_{T,2}$  and  $M_{T,3}$  can be established similarly. Observe that

$$\begin{aligned} M_{T,1} &\leq \max_{t \geq 1} \|\Theta_t\| \cdot \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \sum_{k=0}^j \|\mathbf{B}_{k,t}\| \cdot \|\mathbf{C}_{j-k,t+1-k} - \mathbf{C}_{j-k,t-k}\| \\ &= \max_{t \geq 1} \|\Theta_t\| \cdot \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{k=0}^{\infty} \sum_{j=k}^{\infty} j \|\mathbf{B}_{k,t}\| \cdot \|\mathbf{C}_{j-k,t+1-k} - \mathbf{C}_{j-k,t-k}\| \\ &= \max_{t \geq 1} \|\Theta_t\| \cdot \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{k=0}^{\infty} \|\mathbf{B}_{k,t}\| \sum_{j=0}^{\infty} (j+k) \|\mathbf{C}_{j,t+1-k} - \mathbf{C}_{j,t-k}\| \\ &\leq \max_{t \geq 1} \|\Theta_t\| \cdot \max_{t \geq 1} \sum_{k=0}^{\infty} \|\mathbf{B}_{k,t}\| \cdot \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=0}^{\infty} j \|\mathbf{C}_{j,t+1-k} - \mathbf{C}_{j,t-k}\| \\ &\quad + \max_{t \geq 1} \|\Theta_t\| \cdot \max_{t \geq 1} \sum_{k=0}^{\infty} k \|\mathbf{B}_{k,t}\| \cdot \lim_{T \rightarrow \infty} \sum_{t=1}^{T-1} \sum_{j=0}^{\infty} \|\mathbf{C}_{j,t+1-k} - \mathbf{C}_{j,t-k}\| = O(1). \end{aligned}$$

The proof is now completed. □

*Proof of Theorem 3.1.*

Since  $\frac{1}{T} \sum_{t=1}^T K_h(\tau_t - \tau) = 1 + O\left(\frac{1}{Th}\right)$ , we have

$$\hat{\boldsymbol{\mu}}(\tau) - E(\hat{\boldsymbol{\mu}}(\tau)) = \frac{1}{T} \sum_{t=1}^T (\mathbf{x}_t - \boldsymbol{\mu}(\tau_t)) K_h(\tau_t - \tau) + O_P\left(\frac{1}{Th}\right),$$

which follows that  $\sqrt{Th}(\widehat{\boldsymbol{\mu}}(\tau) - E(\widehat{\boldsymbol{\mu}}(\tau))) = \frac{1}{\sqrt{Th}} \sum_{t=1}^T (\mathbf{x}_t - \boldsymbol{\mu}(\tau_t)) K\left(\frac{\tau_t - \tau}{h}\right) + o_P(1)$ .

As  $\frac{1}{T} \sum_{t=1}^T K_h(\tau_t - \tau) = O(1)$ ,  $\max_t \frac{1}{T} K_h(\tau_t - \tau) = O(1/(Th))$  and  $\frac{1}{T} \sum_{t=1}^{T-1} (K_h(\tau_{t+1} - \tau) - K_h(\tau_t - \tau)) = O(1/(Th))$ , by Lemma 2.4, we have

$$\begin{aligned} & \frac{1}{\sqrt{Th}} \sum_{t=1}^T (\mathbf{x}_t - \boldsymbol{\mu}(\tau_t)) K\left(\frac{\tau_t - \tau}{h}\right) \\ &= \frac{1}{\sqrt{Th}} \sum_{t=1}^T \mathbb{B}_t(1) \boldsymbol{\epsilon}_t K\left(\frac{\tau_t - \tau}{h}\right) + \frac{1}{\sqrt{Th}} \widetilde{\mathbb{B}}_1(L) \boldsymbol{\epsilon}_0 K\left(\frac{\tau_1 - \tau}{h}\right) - \frac{1}{\sqrt{Th}} \widetilde{\mathbb{B}}_T(L) \boldsymbol{\epsilon}_T K\left(\frac{\tau_T - \tau}{h}\right) \\ & \quad + \frac{1}{\sqrt{Th}} \sum_{t=1}^{T-1} \left( \widetilde{\mathbb{B}}_{t+1}(L) K\left(\frac{\tau_{t+1} - \tau}{h}\right) - \widetilde{\mathbb{B}}_t(L) K\left(\frac{\tau_t - \tau}{h}\right) \right) \boldsymbol{\epsilon}_t \\ & \rightarrow_D N \left( \mathbf{0}, \widetilde{v}_0 \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j(\tau) \right\} \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j^\top(\tau) \right\} \right). \end{aligned}$$

For the bias term, we have for any  $\tau \in (0, 1)$

$$\frac{1}{Th} \sum_{t=1}^T \boldsymbol{\mu}(\tau_t) K\left(\frac{\tau_t - \tau}{h}\right) = \boldsymbol{\mu}(\tau) + \frac{1}{2} h^2 \widetilde{c}_2 \boldsymbol{\mu}^{(2)}(\tau) + o(h^2) + O\left(\frac{1}{Th}\right).$$

The proof is now completed. □

*Proof of Theorem 3.2.*

Note that  $\mathbf{x}_t^* = \widetilde{\boldsymbol{\mu}}(\tau_t) + \mathbf{e}_t^*$ , so we can write

$$\widehat{\boldsymbol{\mu}}^*(\tau) - \widetilde{\boldsymbol{\mu}}(\tau) = \left( \sum_{t=1}^T W_{T,t}(\tau) \widetilde{\boldsymbol{\mu}}(\tau_t) - \widetilde{\boldsymbol{\mu}}(\tau) \right) + \sum_{t=1}^T W_{T,t}(\tau) \mathbf{e}_t^* := K_{T,1} + K_{T,2},$$

where  $W_{T,t}(\tau) = K\left(\frac{\tau_t - \tau}{h}\right) / \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right)$ .

We start our investigation from  $K_{T,1}$ , and write

$$\begin{aligned} K_{T,1} &= \left( \frac{1}{Th} \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \frac{1}{Th} \sum_{s=1}^T \boldsymbol{\mu}(\tau_s) K\left(\frac{\tau_s - \tau_t}{h}\right) - \frac{1}{Th} \sum_{s=1}^T \boldsymbol{\mu}(\tau_s) K\left(\frac{\tau_s - \tau}{h}\right) \right) \\ & \quad + \frac{1}{\sqrt{Th}} \left( \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \mathbf{Z}_T(\tau_t) - \mathbf{Z}_T(\tau) \right) + O_P\left(\frac{1}{Th}\right) \\ & := K_{T,11} + K_{T,12} + O_P\left(\frac{1}{Th}\right), \end{aligned}$$

where the definitions of  $K_{T,11}$  and  $K_{T,12}$  should be obvious,  $\mathbf{Z}_T(\tau) = \frac{1}{\sqrt{Th}} \sum_{t=1}^T \mathbf{e}_t K\left(\frac{\tau_t - \tau}{h}\right)$  and  $\mathbf{e}_t = \sum_{j=0}^{\infty} \mathbf{B}_j(\tau_t) \boldsymbol{\epsilon}_{t-j}$ . Similar to the development of Lemma 2.2, we can show that  $\|K_{T,12}\| = O_P((Th)^{-1/2})$ , which, along with the conditions of Theorem 3.2, yields

$$\sqrt{Th} \|K_{T,12}\| = O_P((h/\widetilde{h})^{1/2}) = o_P(1).$$

For  $K_{T,11}$ , by the definition of Riemann integral, we have

$$\begin{aligned} K_{T,11} &= \int_{-1}^1 K(u) \int_{-1}^1 K(v) \left( \boldsymbol{\mu}(\tau + v\tilde{h} + uh) - \boldsymbol{\mu}(\tau + v\tilde{h}) \right) dvdu + O\left(\frac{1}{T\tilde{h}}\right) \\ &= \frac{1}{2} h^2 \tilde{c}_2 \boldsymbol{\mu}^{(2)}(\tau) + O(h^2(h + \tilde{h})) + O\left(\frac{1}{T\tilde{h}}\right). \end{aligned}$$

Thus, we need only to focus on  $K_{T,2}$  and then show that

$$\frac{1}{\sqrt{T\tilde{h}}} \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \mathbf{e}_t^* \rightarrow_{D^*} N\left(\mathbf{0}, \tilde{v}_0 \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j(\tau) \right\} \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j^\top(\tau) \right\}\right).$$

Using the Cramér-Wold device, this is enough to show for any conformable unit vector  $\mathbf{d}$ ,

$$\frac{1}{\sqrt{T\tilde{h}}} \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \mathbf{d}^\top \mathbf{e}_t^* \rightarrow_{D^*} N\left(\mathbf{0}, \tilde{v}_0 \mathbf{d}^\top \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j(\tau) \right\} \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j^\top(\tau) \right\} \mathbf{d}\right).$$

For  $\forall \tau \in [h + \tilde{h}, 1 - h - \tilde{h}]$ , we write

$$\begin{aligned} \frac{1}{\sqrt{T\tilde{h}}} \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \mathbf{d}^\top \tilde{\mathbf{e}}_t \xi_t^* &= Z_T^*(\tau) + \frac{1}{\sqrt{T\tilde{h}}} \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \mathbf{d}^\top (\tilde{\mathbf{e}}_t - \mathbf{e}_t) \xi_t^* \\ &= Z_T^*(\tau) + o_{P^*}(1), \end{aligned} \tag{B.1}$$

where  $Z_T^*(\tau) = \frac{1}{\sqrt{T\tilde{h}}} \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \mathbf{d}^\top \mathbf{e}_t \xi_t^*$ , and the second equality follows from

$$\begin{aligned} & EE^* \left\| \frac{1}{\sqrt{T\tilde{h}}} \sum_{t=1}^T K\left(\frac{\tau_t - \tau}{h}\right) \mathbf{d}^\top (\tilde{\mathbf{e}}_t - \mathbf{e}_t) \xi_t^* \right\|^2 \\ &\leq \max_{[T(\tau-h)] \leq t \leq [T(\tau+h)]} E \|\tilde{\mathbf{e}}_t - \mathbf{e}_t\|^2 \left( \frac{1}{T\tilde{h}} \sum_{t=1}^T \sum_{s=1}^T K\left(\frac{\tau_t - \tau}{h}\right) K\left(\frac{\tau_s - \tau}{h}\right) E^*(\xi_t^* \xi_s^*) \right) \\ &= O(\tilde{h}^4 + 1/(T\tilde{h})) O(l) = o(1), \end{aligned}$$

where  $EE^*[\cdot]$  stands for taking the expectation of the variables with respect to the bootstrap draws, and then taking the expectation with respect to the original sample.

By Lemma 2.5, we already have

$$Z_T^*(\tau) \rightarrow_{D^*} N\left(0, \tilde{v}_0 \mathbf{d}^\top \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j(\tau) \right\} \left\{ \sum_{j=0}^{\infty} \mathbf{B}_j^\top(\tau) \right\} \mathbf{d}\right).$$

Combining the above results, we have

$$\sqrt{T\tilde{h}} (\hat{\boldsymbol{\mu}}^*(\tau) - \tilde{\boldsymbol{\mu}}(\tau)) \rightarrow_{D^*} N(\boldsymbol{\mu}_b(\tau), \tilde{v}_0 \boldsymbol{\Sigma}_\boldsymbol{\mu}(\tau)). \tag{B.2}$$

The proof is now completed.  $\square$



*Proof of Theorem 4.1.*

(1). For notational simplicity, let  $\mathbf{Z}_{\tau,t}$  be the transpose of the  $t^{\text{th}}$  row of  $\mathbf{Z}_\tau$ . Also, we define

$$\begin{aligned}\mathbf{S}_{T,k}(\tau) &= \frac{1}{T} \sum_{t=1}^T \mathbf{Z}_t \mathbf{Z}_t^\top \left( \frac{\tau_t - \tau}{h} \right)^k K_h(\tau_t - \tau) \text{ for } 0 \leq k \leq 3, \\ \mathbf{M}(\tau_t) &= \boldsymbol{\beta}(\tau_t) - \boldsymbol{\beta}(\tau) - \boldsymbol{\beta}^{(1)}(\tau)(\tau_t - \tau) - \frac{1}{2} \boldsymbol{\beta}^{(2)}(\tau)(\tau_t - \tau)^2, \\ \mathbf{S}_T(\tau) &= \begin{pmatrix} \mathbf{S}_{T,0}(\tau) & \mathbf{S}_{T,1}(\tau) \\ \mathbf{S}_{T,1}(\tau) & \mathbf{S}_{T,2}(\tau) \end{pmatrix}.\end{aligned}$$

Since  $\mathbf{y}_t = \mathbf{Z}_t^\top (\boldsymbol{\beta}(\tau) + \boldsymbol{\beta}^{(1)}(\tau)(\tau_t - \tau) + \frac{1}{2} \boldsymbol{\beta}^{(2)}(\tau)(\tau_t - \tau)^2 + \mathbf{M}(\tau_t)) + \boldsymbol{\eta}_t$ , we can write

$$\begin{aligned}\widehat{\boldsymbol{\beta}}(\tau) - \boldsymbol{\beta}(\tau) &= [\mathbf{I}_s, \mathbf{0}_s] \left( \frac{1}{T} \sum_{t=1}^T \mathbf{Z}_{\tau,t} \mathbf{Z}_{\tau,t}^\top K_h(\tau_t - \tau) \right)^{-1} \frac{1}{T} \sum_{t=1}^T \mathbf{Z}_{\tau,t} \mathbf{y}_t K_h(\tau_t - \tau) - \boldsymbol{\beta}(\tau) \\ &= [\mathbf{I}_s, \mathbf{0}_s] \mathbf{S}_T^{-1}(\tau) \begin{bmatrix} \mathbf{S}_{T,2}(\tau) \\ \mathbf{S}_{T,3}(\tau) \end{bmatrix} \frac{1}{2} h^2 \boldsymbol{\beta}^{(2)}(\tau) + [\mathbf{I}_s, \mathbf{0}_s] \mathbf{S}_T^{-1}(\tau) \cdot \frac{1}{T} \sum_{t=1}^T \mathbf{Z}_{\tau,t} \mathbf{Z}_t^\top \mathbf{M}(\tau_t) K_h(\tau_t - \tau) \\ &\quad + [\mathbf{I}_s, \mathbf{0}_s] \mathbf{S}_T^{-1}(\tau) \cdot \frac{1}{T} \sum_{t=1}^T \mathbf{Z}_{\tau,t} \boldsymbol{\eta}_t K_h(\tau_t - \tau) = J_{T,1} + J_{T,2} + J_{T,3},\end{aligned}$$

where the definitions of  $J_{T,1}$  to  $J_{T,3}$  should be obvious.

By Lemma B.6 and Lemma 2.2, we have

$$\begin{aligned}J_{T,1} &= \frac{1}{2} h^2 \widetilde{c}_2 \boldsymbol{\beta}^{(2)}(\tau) + O_P \left( h^2 (h^2 + (Th)^{-1/2}) \right), \\ J_{T,2} &= o_P(h^2).\end{aligned}$$

Thus, we focus on  $J_{T,3}$  below. For any  $\tau \in (0, 1)$ , as  $\{\mathbf{Z}_t \boldsymbol{\eta}_t\}$  is a sequence of martingale differences, by Lemma 2.2 and the martingale central limit theory, we have

$$\sqrt{Th} J_{T,3} = (\boldsymbol{\Sigma}_z^{-1}(\tau) \otimes \mathbf{I}_d) \left( \frac{\sqrt{Th}}{T} \sum_{t=1}^T \mathbf{Z}_t \boldsymbol{\eta}_t K_h(\tau_t - \tau) \right) + o_P(1) \rightarrow_D N(\mathbf{0}, \widetilde{v}_0 \mathbf{V}(\tau)).$$

The proof of the first result of this theorem is now completed.

(2). By Lemma 2.2, we have

$$\left\| \widehat{\boldsymbol{\Sigma}}_z(\tau) - \boldsymbol{\Sigma}_z(\tau) \right\| = o_P(1).$$

Then we need only to focus on the rate associated with  $\widehat{\boldsymbol{\Omega}}(\tau)$ . For notational simplicity, we ignore the  $\frac{1}{T} \sum_{t=1}^T K_h(\tau_t - \tau)$ , because of  $\frac{1}{T} \sum_{t=1}^T K_h(\tau_t - \tau) = 1 + O((Th)^{-1})$ .

Write

$$\begin{aligned}
\widehat{\boldsymbol{\Omega}}(\tau) &= \frac{1}{Th} \sum_{t=1}^T \widehat{\boldsymbol{\eta}}_t \widehat{\boldsymbol{\eta}}_t^\top K\left(\frac{\tau_t - \tau}{h}\right) \\
&= \frac{1}{Th} \sum_{t=1}^T (\boldsymbol{\eta}_t + \widehat{\boldsymbol{\eta}}_t - \boldsymbol{\eta}_t) (\boldsymbol{\eta}_t + \widehat{\boldsymbol{\eta}}_t - \boldsymbol{\eta}_t)^\top K\left(\frac{\tau_t - \tau}{h}\right) \\
&= \frac{1}{Th} \sum_{t=1}^T \boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top K\left(\frac{\tau_t - \tau}{h}\right) + \frac{1}{Th} \sum_{t=1}^T (\widehat{\boldsymbol{\eta}}_t - \boldsymbol{\eta}_t) (\widehat{\boldsymbol{\eta}}_t - \boldsymbol{\eta}_t)^\top K\left(\frac{\tau_t - \tau}{h}\right) \\
&\quad + \frac{1}{Th} \sum_{t=1}^T \boldsymbol{\eta}_t (\widehat{\boldsymbol{\eta}}_t - \boldsymbol{\eta}_t)^\top K\left(\frac{\tau_t - \tau}{h}\right) + \frac{1}{Th} \sum_{t=1}^T (\widehat{\boldsymbol{\eta}}_t - \boldsymbol{\eta}_t) \boldsymbol{\eta}_t^\top K\left(\frac{\tau_t - \tau}{h}\right) \\
&:= J_{T,21} + J_{T,22} + J_{T,23} + J_{T,24}.
\end{aligned}$$

Consider  $J_{T,21}$ . Since  $\{\boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top - E(\boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top)\}$  is a sequence of martingale differences, we have

$$\left\| \frac{1}{T} \sum_{t=1}^T \left[ \boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top - E(\boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top) \right] K_h(\tau_t - \tau) \right\| = o_P(1).$$

Next, consider  $J_{T,22}$ . By Lemma B.7 (4)

$$\|J_{T,22}\| \leq \sup_{\tau \in [0,1]} \|\widehat{\boldsymbol{\beta}}(\tau) - \boldsymbol{\beta}(\tau)\|^2 \cdot \frac{1}{T} \sum_{t=1}^T \|\mathbf{Z}_t\|^2 K_h(\tau_t - \tau) = o_P(1).$$

Similarly, for  $J_{T,23}$  and  $J_{T,24}$ , we have

$$\|J_{T,23}\| \leq \sup_{\tau \in [0,1]} \|\widehat{\boldsymbol{\beta}}(\tau) - \boldsymbol{\beta}(\tau)\| \cdot \frac{1}{T} \sum_{t=1}^T \|\mathbf{Z}_t \boldsymbol{\eta}_t\| K_h(\tau_t - \tau) = o_P(1).$$

The proof is now completed. □

*Proof of Theorem 4.2.*

(1). Since

$$\begin{aligned}
\widetilde{\mathbf{y}}_t &= \mathbf{y}_t - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \mathbf{y} \\
&= (\mathbf{Z}_{1t} - \mathbf{Z}_1 \mathbf{s}^\top(\tau_t) \mathbf{Z}_{2t})^\top \mathbf{c} + \boldsymbol{\eta}_t - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \boldsymbol{\eta} + \mathbf{Z}_{2t}^\top \boldsymbol{\theta}(\tau_t) - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \left[ \boldsymbol{\theta}(\tau_1)^\top \mathbf{Z}_{21}, \dots, \boldsymbol{\theta}(\tau_T)^\top \mathbf{Z}_{2T} \right]^\top
\end{aligned}$$

with  $\boldsymbol{\eta} = [\boldsymbol{\eta}_1^\top, \dots, \boldsymbol{\eta}_T^\top]^\top$ , we can decompose  $\widehat{\mathbf{c}}_{WLS} - \mathbf{c}$  into

$$\begin{aligned}
&\sqrt{T} (\widehat{\mathbf{c}}_{WLS} - \mathbf{c}) \\
&= \left( T^{-1} \sum_{t=1}^T \widetilde{\mathbf{Z}}_{1t} \widehat{\boldsymbol{\Omega}}^{-1}(\tau_t) \widetilde{\mathbf{Z}}_{1t}^\top \right)^{-1} T^{-1/2} \sum_{t=1}^T \widetilde{\mathbf{Z}}_{1t} \widehat{\boldsymbol{\Omega}}^{-1}(\tau_t) (\boldsymbol{\eta}_t - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \boldsymbol{\eta})
\end{aligned}$$

$$\begin{aligned}
& + \left( T^{-1} \sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \tilde{\mathbf{Z}}_{1t}^\top \right)^{-1} T^{-1/2} \sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \\
& \times \left( \mathbf{Z}_{2t}^\top \boldsymbol{\theta}(\tau_t) - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \left[ \boldsymbol{\theta}(\tau_1)^\top \mathbf{Z}_{21}, \dots, \boldsymbol{\theta}(\tau_T)^\top \mathbf{Z}_{2T} \right]^\top \right).
\end{aligned}$$

By Lemma B.8, we have

$$\begin{aligned}
(1). \quad & T^{-1} \sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \tilde{\mathbf{Z}}_{1t}^\top \rightarrow_P \boldsymbol{\Delta}_c, \\
(2). \quad & T^{-1/2} \sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \boldsymbol{\eta}_t = T^{-1/2} \sum_{t=1}^T \left( \mathbf{Z}_{1t} - \boldsymbol{\Sigma}_{\mathbf{Z}_{1,2}}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{Z}_2}^{-1}(\tau_t) \mathbf{Z}_{2t} \right) \boldsymbol{\Omega}^{-1}(\tau_t) \boldsymbol{\eta}_t + o_P(1) \\
(3). \quad & T^{-1/2} \sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \boldsymbol{\eta} \rightarrow_P 0, \\
(4). \quad & T^{-1/2} \sum_{t=1}^T \tilde{\mathbf{Z}}_{1t} \hat{\boldsymbol{\Omega}}^{-1}(\tau_t) \left( \mathbf{Z}_{2t}^\top \boldsymbol{\theta}(\tau_t) - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \left[ \boldsymbol{\theta}(\tau_1)^\top \mathbf{Z}_{21}, \dots, \boldsymbol{\theta}(\tau_T)^\top \mathbf{Z}_{2T} \right]^\top \right) \rightarrow_P 0,
\end{aligned}$$

which follows that

$$\sqrt{T} (\hat{\mathbf{c}}_{WLS} - \mathbf{c}) = \boldsymbol{\Delta}_c^{-1} \frac{1}{\sqrt{T}} \sum_{t=1}^T \left( \mathbf{Z}_{1t} - \boldsymbol{\Sigma}_{\mathbf{Z}_{1,2}}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{Z}_2}^{-1}(\tau_t) \mathbf{Z}_{2t} \right) \boldsymbol{\Omega}^{-1}(\tau_t) \boldsymbol{\eta}_t + o_P(1).$$

Since  $\left\{ \left( \mathbf{Z}_{1t} - \boldsymbol{\Sigma}_{\mathbf{Z}_{1,2}}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{Z}_2}^{-1}(\tau_t) \mathbf{Z}_{2t} \right) \boldsymbol{\Omega}^{-1}(\tau_t) \boldsymbol{\eta}_t \right\}$  is a sequence of martingale differences, the result follows by the central limit theorem for martingale differences. Note that the convergence of conditional variance can be proved by Lemma 2.2.

(2). Let  $\boldsymbol{\Theta}(\tau) = [\boldsymbol{\theta}(\tau)^\top, h\boldsymbol{\theta}^{(1)}(\tau)^\top]^\top$ . Note that

$$\begin{aligned}
\hat{\boldsymbol{\theta}}(\tau) - \boldsymbol{\theta}(\tau) & = [\mathbf{I}_{s_2}, \mathbf{0}_{s_2}] (\mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \mathbf{Z}_{2,\tau})^{-1} \mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau (\mathbf{y} - \mathbf{Z}_1 \hat{\mathbf{c}}_{WLS} - \mathbf{Z}_{2,\tau} \boldsymbol{\Theta}(\tau)) \\
& = [\mathbf{I}_{s_2}, \mathbf{0}_{s_2}] (\mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \mathbf{Z}_{2,\tau})^{-1} \mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \mathbf{Z}_1 (\mathbf{c} - \hat{\mathbf{c}}_{WLS}) \\
& \quad + [\mathbf{I}_{s_2}, \mathbf{0}_{s_2}] (\mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \mathbf{Z}_{2,\tau})^{-1} \mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \begin{pmatrix} \left[ \begin{array}{c} \mathbf{Z}_{21}^\top \boldsymbol{\theta}(\tau_1) \\ \vdots \\ \mathbf{Z}_{2T}^\top \boldsymbol{\theta}(\tau_T) \end{array} \right] \\ - \mathbf{Z}_{2,\tau} \boldsymbol{\Theta}(\tau) \end{pmatrix} \\
& \quad + [\mathbf{I}_{s_2}, \mathbf{0}_{s_2}] (\mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \mathbf{Z}_{2,\tau})^{-1} \mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \boldsymbol{\eta} \\
& := J_{T,31} + J_{T,32} + J_{T,33}.
\end{aligned}$$

By part (1), we have  $J_{T,31} = O_P(T^{-1/2})$ . By using standard arguments of the local linear kernel method, we have  $J_{T,32} = \frac{1}{2} h^2 \tilde{\mathbf{c}}_2 \boldsymbol{\theta}^{(2)}(\tau) + o_P(h^2)$ . Then, it suffices to show that

$$[\mathbf{I}_{s_2}, \mathbf{0}_{s_2}] \frac{\sqrt{h}}{\sqrt{T}} \mathbf{Z}_{2,\tau}^\top \mathbf{K}_\tau \boldsymbol{\eta} \rightarrow_D N(\mathbf{0}, \tilde{v}_0 \boldsymbol{\Sigma}_{z_2}(\tau) \otimes \boldsymbol{\Omega}(\tau)),$$

which follows directly from the martingale central limit theorem. Also, the convergence of conditional variance can be proved by Lemma 2.2.

□

*Proof of Theorem B.1.*

(1). To proceed, define

$$\mathbf{\Gamma}(\tau) = \begin{bmatrix} \mathbf{\Phi}(\tau) & * \\ \mathbf{0} & \mathbf{I}^* \end{bmatrix} \quad \text{and} \quad \mathbf{I}^* = \begin{bmatrix} \mathbf{0}_m & \cdots & \mathbf{0}_m & \mathbf{0}_m \\ \mathbf{I}_m & \cdots & \mathbf{0}_m & \mathbf{0}_m \\ \vdots & \ddots & \vdots & \vdots \\ \mathbf{0}_m & \cdots & \mathbf{I}_m & \mathbf{0}_m \end{bmatrix}.$$

Let  $\rho$  denote the largest eigenvalue of  $\mathbf{\Phi}(\tau)$  uniformly over  $\tau \in [0, 1]$ . Then we have  $\rho < 1$  by Assumption 5.1. By construction, it is easy to show that the largest eigenvalue of  $\mathbf{\Gamma}(\tau)$  is  $\rho$  uniformly over  $\tau \in [0, 1]$ , since all eigenvalues of  $\mathbf{I}^*$  are zeros. Similar to the proof of Proposition 2.4 in Dahlhaus and Polonik (2009), we have  $\max_{t \geq 1} \left\| \prod_{m=0}^{j-1} \mathbf{\Gamma}(\tau_{t-m}) \right\| \leq M\rho^j$ .

Note that for any conformable matrices  $\{\mathbf{A}_i\}$  and  $\{\mathbf{B}_i\}$ , since

$$\prod_{i=1}^r \mathbf{A}_i - \prod_{i=1}^r \mathbf{B}_i = \sum_{j=1}^r \left( \prod_{k=1}^{j-1} \mathbf{A}_k \right) (\mathbf{A}_j - \mathbf{B}_j) \left( \prod_{k=j+1}^r \mathbf{B}_k \right),$$

we have

$$\begin{aligned} \left\| \mathbf{D}_{j,t}^{\mathbf{x}} - \mathbf{D}_j^{\mathbf{x}}(\tau_t) \right\| &= \left\| \bar{\mathbf{J}} \prod_{k=0}^{j-1} \mathbf{\Gamma}(\tau_{t-k}) \mathbf{\Upsilon}(\tau_{t-j}) - \bar{\mathbf{J}} \mathbf{\Gamma}^j(\tau_t) \mathbf{\Upsilon}(\tau_t) \right\| \\ &= \left\| \bar{\mathbf{J}} \left( \prod_{k=0}^{j-1} \mathbf{\Gamma}(\tau_{t-k}) - \mathbf{\Gamma}^j(\tau_t) \right) \mathbf{\Upsilon}(\tau_t) + \bar{\mathbf{J}} \prod_{k=0}^{j-1} \mathbf{\Gamma}(\tau_{t-k}) (\mathbf{\Upsilon}(\tau_{t-j}) - \mathbf{\Upsilon}(\tau_t)) \right\| \\ &\leq O(1) \sum_{i=1}^{j-1} \left\| \mathbf{\Gamma}^i(\tau_t) (\mathbf{\Gamma}(\tau_{t-i}) - \mathbf{\Gamma}(\tau_t)) \prod_{k=i+1}^{j-1} \mathbf{\Gamma}(\tau_{t-k}) \right\| + M\rho^j \frac{j}{T} \\ &\leq O(1) \sum_{i=1}^{j-1} \frac{i}{T} \rho^{j-1} + O(1) \rho^j \frac{j}{T} = O(T^{-1}). \end{aligned}$$

The proof of the first result is then complete.

(2). Part (2) follows directly from the Delta method (see Lütkepohl, 2005, p.408) and

$$\mathbf{G}_j(\tau) = \frac{\partial \text{vec} \left[ \mathbf{D}_j^{\mathbf{x}}(\tau) \right]}{\partial \boldsymbol{\beta}(\tau)^\top} = \left[ \sum_{k=0}^{j-1} \mathbf{\Upsilon}(\tau)^\top (\mathbf{\Gamma}(\tau)^\top)^{j-1-k} \otimes \mathbf{J} \mathbf{\Gamma}^k(\tau) \mathbf{J}^\top, \mathbf{I}_m \otimes \mathbf{J} \mathbf{\Gamma}^k(\tau) \mathbf{J}^\top \right].$$

□

## B.4 Proofs of the Preliminary Lemmas

*Proof of Lemma B.3.*

(1). The first result follows from the standard BN decomposition (e.g., Phillips and Solo, 1992), so the details are omitted. (2). For the second decomposition, write

$$\begin{aligned}
(1-L)\tilde{\mathbb{B}}_t^r(L) &= \sum_{j=0}^{\infty} \left( L^j \sum_{k=j+1}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}) - L^{j+1} \sum_{k=j+1}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}) \right) \\
&= \sum_{j=0}^{\infty} \left( L^{j+1} \sum_{k=j+2}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}) - L^{j+1} \sum_{k=j+1}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}) \right) + \sum_{k=1}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}) \\
&= - \sum_{j=0}^{\infty} L^{j+1} (\mathbf{B}_{j+1+r,t} \otimes \mathbf{B}_{j+1,t}) + \sum_{k=1}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}) \\
&= - \sum_{j=0}^{\infty} L^j (\mathbf{B}_{j+r,t} \otimes \mathbf{B}_{j,t}) + \sum_{k=0}^{\infty} (\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}) = \mathbb{B}_t^r(1) - \mathbb{B}_t^r(L).
\end{aligned}$$

(3). By Assumption 1,

$$\max_{t \geq 1} \sum_{j=0}^{\infty} \left\| \tilde{\mathbf{B}}_{j,t} \right\| \leq \max_{t \geq 1} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} \|\mathbf{B}_{k,t}\| = \max_{t \geq 1} \sum_{j=1}^{\infty} j \|\mathbf{B}_{j,t}\| < \infty.$$

(4). By Assumption 1,

$$\sum_{t=1}^{T-1} \left\| \tilde{\mathbb{B}}_{t+1}(1) - \tilde{\mathbb{B}}_t(1) \right\| \leq \sum_{t=1}^{T-1} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} \|\mathbf{B}_{k,t+1} - \mathbf{B}_{k,t}\| = \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \|\mathbf{B}_{j,t+1} - \mathbf{B}_{j,t}\| < \infty.$$

(5). By Assumption 1,

$$\begin{aligned}
\max_{t \geq 1} \sum_{j=0}^{\infty} \left\| \tilde{\mathbf{B}}_{j,t}^r \right\| &\leq \max_{t \geq 1} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} \|\mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t}\| = \max_{t \geq 1} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} \|\mathbf{B}_{k+r,t}\| \cdot \|\mathbf{B}_{k,t}\| \\
&= \max_{t \geq 1} \sum_{j=1}^{\infty} j \|\mathbf{B}_{j+r,t}\| \cdot \|\mathbf{B}_{j,t}\| \leq M \max_{t \geq 1} \sum_{j=1}^{\infty} j \|\mathbf{B}_{j,t}\| < \infty.
\end{aligned}$$

(6). Write

$$\begin{aligned}
\max_{t \geq 1} \sum_{r=1}^{\infty} \left\| \tilde{\mathbb{B}}_t^r(1) \right\| &\leq \max_{t \geq 1} \sum_{r=1}^{\infty} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} \|\mathbf{B}_{k+r,t}\| \cdot \|\mathbf{B}_{k,t}\| \\
&= \max_{t \geq 1} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} \|\mathbf{B}_{k,t}\| \cdot \left( \sum_{r=1}^{\infty} \|\mathbf{B}_{k+r,t}\| \right) \leq \max_{t \geq 1} \left( \sum_{r=1}^{\infty} \|\mathbf{B}_{r,t}\| \right) \cdot \left( \sum_{j=1}^{\infty} j \|\mathbf{B}_{j,t}\| \right) < \infty.
\end{aligned}$$

(7). Write

$$\begin{aligned}
& \sum_{t=1}^{T-1} \sum_{r=0}^{\infty} \sum_{j=0}^{\infty} \left\| \tilde{\mathbb{B}}_{t+1}^r(1) - \tilde{\mathbb{B}}_t^r(1) \right\| \leq \sum_{t=1}^{T-1} \sum_{r=0}^{\infty} \sum_{j=0}^{\infty} \left\| \sum_{k=j+1}^{\infty} \mathbf{B}_{k+r,t+1} \otimes \mathbf{B}_{k,t+1} - \mathbf{B}_{k+r,t} \otimes \mathbf{B}_{k,t} \right\| \\
& \leq \sum_{t=1}^{T-1} \sum_{r=0}^{\infty} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} (\|\mathbf{B}_{k+r,t+1} - \mathbf{B}_{k+r,t}\| \cdot \|\mathbf{B}_{k,t+1}\| + \|\mathbf{B}_{k+r,t}\| \cdot \|\mathbf{B}_{k,t+1} - \mathbf{B}_{k,t}\|) \\
& = \sum_{t=1}^{T-1} \sum_{j=0}^{\infty} \sum_{k=j+1}^{\infty} \left( \|\mathbf{B}_{k,t+1}\| \cdot \sum_{r=0}^{\infty} \|\mathbf{B}_{k+r,t+1} - \mathbf{B}_{k+r,t}\| + \|\mathbf{B}_{k,t+1} - \mathbf{B}_{k,t}\| \cdot \sum_{r=0}^{\infty} \|\mathbf{B}_{k+r,t}\| \right) \\
& \leq \left( \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \|\mathbf{B}_{r,t+1} - \mathbf{B}_{r,t}\| \right) \cdot \left( \max_{t \geq 1} \sum_{k=1}^{\infty} k \|\mathbf{B}_{k,t}\| \right) \\
& \quad + \left( \sum_{t=1}^{T-1} \sum_{k=1}^{\infty} k \|\mathbf{B}_{k,t+1} - \mathbf{B}_{k,t}\| \right) \cdot \left( \max_{t \geq 1} \sum_{r=1}^{\infty} \|\mathbf{B}_{r,t}\| \right) < \infty.
\end{aligned}$$

The proof is now completed.  $\square$

*Proof of Lemma B.4.*

In the following proof, we cover the interval  $[a, b]$  by a finite number of subintervals  $\{S_l\}$ , which are centred at  $s_l$  with the length denoted by  $\delta_T$ . Denoting the number of these intervals by  $N_T$ , then  $N_T = O(\delta_T^{-1})$ . In addition, let  $\delta_T = O(T^{-1}\gamma_T)$  with  $\gamma_T = \sqrt{d_T \log T}$ .

Write

$$\begin{aligned}
& \sup_{\tau \in [a, b]} \left\| \sum_{t=1}^T \mathbf{W}_{T,t}(\tau) \mathbb{B}_t(1) \boldsymbol{\epsilon}_t \right\| \leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \mathbf{W}_{T,t}(s_l) \mathbb{B}_t(1) \boldsymbol{\epsilon}_t \right\| \\
& \quad + \max_{1 \leq l \leq N_T} \sup_{\tau \in S_l} \left\| \sum_{t=1}^T (\mathbf{W}_{T,t}(\tau) - \mathbf{W}_{T,t}(s_l)) \mathbb{B}_t(1) \boldsymbol{\epsilon}_t \right\| := J_{T,41} + J_{T,42}.
\end{aligned}$$

For  $J_{T,42}$ , since  $\mathbf{W}_{T,t}(\cdot)$  is Lipschitz continuous and  $\max_{t \geq 1} \|\mathbb{B}_t(1)\| < \infty$  by Assumption 1, we have

$$\begin{aligned}
E|J_{T,42}| & \leq \sum_{t=1}^T \max_{1 \leq l \leq N_T} \sup_{\tau \in S_l} \|\mathbf{W}_{T,t}(\tau) - \mathbf{W}_{T,t}(s_l)\| E \|\mathbb{B}_t(1) \boldsymbol{\epsilon}_t\| \\
& \leq MT \delta_T \max_{t \geq 1} E \|\mathbb{B}_t(1) \boldsymbol{\epsilon}_t\| = O(\gamma_T).
\end{aligned}$$

For  $J_{T,41}$ , we apply the truncation method. Define  $\boldsymbol{\epsilon}'_t = \boldsymbol{\epsilon}_t I(\|\boldsymbol{\epsilon}_t\| \leq T^{\frac{1}{\delta}})$  and  $\boldsymbol{\epsilon}''_t = \boldsymbol{\epsilon}_t - \boldsymbol{\epsilon}'_t$ , where  $\delta$  is defined in Assumption 2, and  $I(\cdot)$  is the indicator function. Write

$$\begin{aligned}
J_{T,41} & = \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \mathbf{W}_{T,t}(s_l) \mathbb{B}_t(1) (\boldsymbol{\epsilon}'_t + \boldsymbol{\epsilon}''_t - E(\boldsymbol{\epsilon}'_t + \boldsymbol{\epsilon}''_t | \mathcal{F}_{t-1})) \right\| \\
& \leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \mathbf{W}_{T,t}(s_l) \mathbb{B}_t(1) (\boldsymbol{\epsilon}'_t - E(\boldsymbol{\epsilon}'_t | \mathcal{F}_{t-1})) \right\| + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \mathbf{W}_{T,t}(s_l) \mathbb{B}_t(1) \boldsymbol{\epsilon}''_t \right\|
\end{aligned}$$

$$+ \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \mathbf{W}_{T,t}(s_l) \mathbb{B}_t(1) E(\boldsymbol{\epsilon}_t'' | \mathcal{F}_{t-1}) \right\| := J_{T,51} + J_{T,52} + J_{T,53}.$$

Start from  $J_{T,52}$ . By Hölder's inequality and Markov's inequality,

$$\begin{aligned} E|J_{T,52}| &\leq O(1)d_T \sum_{t=1}^T E \|\boldsymbol{\epsilon}_t''\| = O(1)d_T \sum_{t=1}^T E \|\boldsymbol{\epsilon}_t I(\|\boldsymbol{\epsilon}_t\| \geq T^{\frac{1}{\delta}})\| \\ &\leq O(1)d_T \sum_{t=1}^T \left\{ E \|\boldsymbol{\epsilon}_t\|^\delta \right\}^{\frac{1}{\delta}} \left\{ E I(\|\boldsymbol{\epsilon}_t\| \geq T^{\frac{1}{\delta}}) \right\}^{\frac{\delta-1}{\delta}} \\ &= O(1)d_T \sum_{t=1}^T \left\{ E \|\boldsymbol{\epsilon}_t\|^\delta \right\}^{\frac{1}{\delta}} \left\{ \Pr(\|\boldsymbol{\epsilon}_t\| \geq T^{\frac{1}{\delta}}) \right\}^{\frac{\delta-1}{\delta}} \\ &\leq O(1)d_T \sum_{t=1}^T \left\{ E \|\boldsymbol{\epsilon}_t\|^\delta \right\}^{\frac{1}{\delta}} \left\{ \frac{E \|\boldsymbol{\epsilon}_t\|^\delta}{T} \right\}^{\frac{\delta-1}{\delta}} = O(T^{\frac{1}{\delta}} d_T) = o\left(\sqrt{d_T \log T}\right), \end{aligned}$$

where the second inequality follows from Hölder's inequality, and the third inequality follows from Markov's inequality. Similarly,  $J_{T,53} = O_P(T^{\frac{1}{\delta}} d_T) = o_P\left(\sqrt{d_T \log T}\right)$ .

We now turn to  $J_{T,51}$ . For notational simplicity, let  $\mathbf{Y}_t = \mathbf{W}_{T,t}(s_l) \mathbb{B}_t(1) (\boldsymbol{\epsilon}_t' - E(\boldsymbol{\epsilon}_t' | \mathcal{F}_{t-1}))$  for  $1 \leq t \leq T$  and  $A_T = 2T^{\frac{1}{\delta}} d_T \max_{t \geq 1} \|\mathbb{B}_t(1)\|$ . Simple algebra shows that  $\|\mathbf{Y}_t\| \leq A_T$  uniformly in  $t$  and  $s_l$ . By Assumption 2 and the first condition in the body of this lemma,

$$\max_{1 \leq l \leq N_T} \sum_{t=1}^T E \left( \|\mathbf{Y}_t\|^2 | \mathcal{F}_{t-1} \right) \leq M d_T \max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| E \left( \|\boldsymbol{\epsilon}_t\|^2 | \mathcal{F}_{t-1} \right) = O_{a.s.}(d_T).$$

By Lemma B.2 and  $T^{\frac{2}{\delta}} d_T \log T \rightarrow 0$ , choose some  $\beta > 0$  (such as  $\beta = 4$ ), and write

$$\begin{aligned} &\Pr \left( J_{T,51} > \sqrt{\beta M} \gamma_T \right) \\ &= \Pr \left( J_{T,51} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T \right) \\ &\quad + \Pr \left( J_{T,51} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T \right) \\ &\leq \Pr \left( J_{T,51} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T \right) \\ &\quad + \Pr \left( \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T \right) \\ &\leq N_T \exp \left( -\frac{\beta M \gamma_T^2}{2(M d_T + \gamma_T 2 A_T)} \right) + o(1) \leq N_T \exp \left( -\frac{\beta}{2} \log T \right) + o(1) \rightarrow 0. \end{aligned}$$

Based on the above development, the proof is now completed.  $\square$

*Proof of Lemma B.5.*

(1). Similar to the proof of Lemma B.4, we use a finite number of subintervals  $\{S_l\}$  to cover the

interval  $[a, b]$ , which are centered at  $s_l$  with the length  $\delta_T$ . Denote the number of these intervals by  $N_T$  then  $N_T = O(\delta_T^{-1})$ . In addition, let  $\delta_T = O(T^{-1}\gamma_T)$  with  $\gamma_T = \sqrt{d_T \log T}$ .

$$\begin{aligned}
& \sup_{\tau \in [a, b]} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(\tau)) \mathbb{B}_t^0(1) \left( \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \right) - \text{vec}(\mathbf{I}_d) \right) \right\| \\
& \leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) \mathbb{B}_t^0(1) \left( \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \right) - \text{vec}(\mathbf{I}_d) \right) \right\| \\
& \quad + \max_{1 \leq l \leq N_T} \sup_{\tau \in S_l} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes (\mathbf{W}_{T,t}(\tau) - \mathbf{W}_{T,t}(s_l))) \mathbb{B}_t^0(1) \left( \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \right) - \text{vec}(\mathbf{I}_d) \right) \right\| \\
& := J_{T,61} + J_{T,62}.
\end{aligned}$$

Start from  $J_{T,62}$ . Similar to the proof of Lemma B.4, since

$$\left\| \mathbb{B}_t^0(1) \right\| \leq \sum_{j=0}^{\infty} \left\| \mathbf{B}_{j,t} \right\|^2 \leq \left( \sum_{j=0}^{\infty} \left\| \mathbf{B}_{j,t} \right\| \right)^2 < \infty$$

by Assumption 1, we have

$$E |J_{T,62}| \leq MT \delta_T \max_{t \geq 1} E \left\| \mathbb{B}_t^0(1) \left( \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \right) - \text{vec}(\mathbf{I}_d) \right) \right\| = O(\gamma_T).$$

We then apply the truncation method. Define  $\mathbf{u}_t = \mathbb{B}_t^0(1) \left( \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \right) - \text{vec}(\mathbf{I}_d) \right)$ ,  $\mathbf{u}'_t = \mathbf{u}_t I(\|\mathbf{u}_t\| \leq T^{\frac{2}{\delta}})$  and  $\mathbf{u}''_t = \mathbf{u}_t - \mathbf{u}'_t$ . For  $J_{T,1}$ , write

$$\begin{aligned}
J_{T,61} &= \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) (\mathbf{u}'_t + \mathbf{u}''_t - E(\mathbf{u}'_t + \mathbf{u}''_t | \mathcal{F}_{t-1})) \right\| \\
&\leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) (\mathbf{u}'_t - E(\mathbf{u}'_t | \mathcal{F}_{t-1})) \right\| + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) \mathbf{u}''_t \right\| \\
&\quad + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) E(\mathbf{u}''_t | \mathcal{F}_{t-1}) \right\| := J_{T,71} + J_{T,72} + J_{T,73}.
\end{aligned}$$

As in the proof of Lemma B.4, we can show that  $J_{T,72} = o_P(\sqrt{d_T \log T})$  and  $J_{T,73} = o_P(\sqrt{d_T \log T})$  respectively. We focus on  $J_{T,71}$  below.

For any  $1 \leq l \leq N_T$ , let  $\mathbf{Y}_t = (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) (\mathbf{u}'_t - E(\mathbf{u}'_t | \mathcal{F}_{t-1}))$ . We then have  $E(\mathbf{Y}_t | \mathcal{F}_{t-1}) = 0$  and  $\|\mathbf{Y}_t\| \leq 2T^{2/\delta} d_T \max_t \left\| \mathbb{B}_t^0(1) \right\|$ . Since  $\max_{t \geq 1} E \left( \|\boldsymbol{\epsilon}_t\|^4 | \mathcal{F}_{t-1} \right) < \infty$  a.s., we can write

$$\max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T \max_{t \geq 1} E \left( \|\mathbf{u}_t\|^2 | \mathcal{F}_{t-1} \right) \max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| = O_{a.s.}(d_T).$$

Similar to Lemma B.2, choose  $\beta > 0$  (such as  $\beta = 4$ ). In view of the fact that  $T^{\frac{4}{\delta}} d_T \log T \rightarrow 0$ , we



write

$$\begin{aligned}
\Pr\left(J_{T,71} > \sqrt{\beta M} \gamma_T\right) &= \Pr\left(J_{T,71} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T\right) \\
&\quad + \Pr\left(J_{T,71} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T\right) \\
&\leq \Pr\left(J_{T,71} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T\right) \\
&\quad + \Pr\left(\max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T\right) \\
&\leq N_T \exp\left(-\frac{\beta M \gamma_T^2}{2(M d_T + M \gamma_T T^{\frac{2}{5}} d_T)}\right) + o(1) \\
&\leq N_T \exp\left(-\frac{\beta}{2} \log T\right) + o(1) = N_T T^{-\frac{\beta}{2}} + o(1) = o(1).
\end{aligned}$$

The first result then follows.

(2). Let  $\{S_l\}$  be a finite number of subintervals covering the interval  $[a, b]$ , which are centered at  $s_l$  with the length  $\delta_T$ . Denote the number of these intervals by  $N_T$  then  $N_T = O(\delta_T^{-1})$ . In addition, let  $\delta_T = O(T^{-1} \gamma_T)$  with  $\gamma_T = \sqrt{d_T \log T}$ . Then

$$\begin{aligned}
\sup_{\tau \in [a, b]} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(\tau)) \zeta_t \epsilon_t \right\| &\leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) \zeta_t \epsilon_t \right\| \\
&\quad + \max_{1 \leq l \leq N_T} \sup_{\tau \in S_l} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes (\mathbf{W}_{T,t}(\tau) - \mathbf{W}_{T,t}(s_l))) \zeta_t \epsilon_t \right\| := J_{T,81} + J_{T,82}.
\end{aligned}$$

Consider  $J_{T,82}$ . By the fact that  $|\text{tr}(\mathbf{A})| \leq d \|\mathbf{A}\|$  for any  $d \times d$  matrix  $\mathbf{A}$  and Assumption 1,

$$\begin{aligned}
E \|\zeta_t \epsilon_t\| &= E \left\| \sum_{r=1}^{\infty} \left( \sum_{s=0}^{\infty} \mathbf{B}_{s+r,t} \otimes \mathbf{B}_{s,t} \right) \text{vec} \left( \epsilon_t \epsilon_{t-r}^\top \right) \right\| \\
&\leq \left( E \left\| \sum_{r=1}^{\infty} \left( \sum_{s=0}^{\infty} \mathbf{B}_{s+r,t} \otimes \mathbf{B}_{s,t} \right) \text{vec} \left( \epsilon_t \epsilon_{t-r}^\top \right) \right\|^2 \right)^{1/2} \\
&\leq \left\{ \text{tr} \left[ \sum_{r=1}^{\infty} \left( \sum_{s=0}^{\infty} \mathbf{B}_{s+r,t} \otimes \mathbf{B}_{s,t} \right) \cdot (\mathbf{I}_d \otimes \mathbf{I}_d) \cdot \left( \sum_{s=0}^{\infty} \mathbf{B}_{s+r,t} \otimes \mathbf{B}_{s,t} \right)^\top \right] \right\}^{1/2} \\
&\leq M \left( \sum_{r=1}^{\infty} \left\| \sum_{s=0}^{\infty} \mathbf{B}_{s+r,t} \otimes \mathbf{B}_{s,t} \right\|^2 \right)^{1/2} \leq M \left( \sum_{r=1}^{\infty} \left( \sum_{s=0}^{\infty} \|\mathbf{B}_{s+r,t}\|^2 \right) \cdot \left( \sum_{s=0}^{\infty} \|\mathbf{B}_{s,t}\|^2 \right) \right)^{1/2} \\
&= M \left( \left( \sum_{r=1}^{\infty} r \|\mathbf{B}_{r,t}\|^2 \right) \cdot \left( \sum_{s=0}^{\infty} \|\mathbf{B}_{s,t}\|^2 \right) \right)^{1/2} \\
&\leq M \left( \left( \sum_{r=1}^{\infty} r \|\mathbf{B}_{r,t}\|^2 \right)^2 \cdot \left( \sum_{s=0}^{\infty} \|\mathbf{B}_{s,t}\|^2 \right) \right)^{1/2} < \infty.
\end{aligned}$$

Similarly, we have  $E|J_{T,82}| \leq MT\delta_T \max_{t \geq 1} E \|\zeta_t \epsilon_t\| = O(\gamma_T)$ .

Before investigating  $J_{T,81}$ , we first show that

$$\max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| E \left( \|\zeta_t \epsilon_t\|^2 | \mathcal{F}_{t-1} \right) = O_P(1). \quad (\text{B.1})$$

Note that

$$\begin{aligned} & \max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| E \left( \|\zeta_t \epsilon_t\|^2 | \mathcal{F}_{t-1} \right) \\ & \leq \max_{1 \leq l \leq N_T} \left| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \left( E \left( \|\zeta_t \epsilon_t\|^2 | \mathcal{F}_{t-1} \right) - E \|\zeta_t \epsilon_t\|^2 \right) \right| \\ & \quad + \max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| E \|\zeta_t \epsilon_t\|^2 \end{aligned}$$

and  $\max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| E \|\zeta_t \epsilon_t\|^2 = O(1)$ . Thus, to prove (B.1), it is sufficient to show

$$\max_{1 \leq l \leq N_T} \left| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \left( E \left( \|\zeta_t \epsilon_t\|^2 | \mathcal{F}_{t-1} \right) - E \|\zeta_t \epsilon_t\|^2 \right) \right| = o_P(1).$$

In order to do so, we write

$$\begin{aligned} & \max_{1 \leq l \leq N_T} \left| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \left( E \left( \|\zeta_t \epsilon_t\|^2 | \mathcal{F}_{t-1} \right) - E \|\zeta_t \epsilon_t\|^2 \right) \right| \\ & = \max_{1 \leq l \leq N_T} \left| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \operatorname{tr} \left( \sum_{r,r^*=1}^{\infty} \mathbb{B}_t^r(1) (\epsilon_{t-r} \epsilon_{t-r^*}^\top \otimes \mathbf{I}_d) \mathbb{B}_t^{r^*,\top}(1) - \sum_{r=1}^{\infty} \mathbb{B}_t^r(1) \mathbb{B}_t^{r,\top}(1) \right) \right| \\ & \leq d^2 \cdot \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \sum_{r=1}^{\infty} (\mathbb{B}_t^r(1) \otimes \mathbb{B}_t^r(1)) \left( \operatorname{vec} \left( \epsilon_{t-r} \epsilon_{t-r}^\top \otimes \mathbf{I}_d \right) - \operatorname{vec}(\mathbf{I}_{d^2}) \right) \right\| \\ & \quad + 2d^2 \cdot \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \sum_{r=1}^{\infty} \sum_{j=1}^{\infty} \left( \mathbb{B}_t^{r+j}(1) \otimes \mathbb{B}_t^r(1) \right) \operatorname{vec} \left( \epsilon_{t-r} \epsilon_{t-r-j}^\top \otimes \mathbf{I}_d \right) \right\| \\ & := J_{T,91} + J_{T,92}. \end{aligned}$$

Let  $\mathbb{F}_{r,t}(L) = \sum_{j=1}^{\infty} \mathbb{B}_t^{r+j}(1) \otimes \mathbb{B}_t^j(1) L^j$ . Similar to the second result of Lemma B.3, we have

$$\mathbb{F}_{r,t}(L) = \mathbb{F}_{r,t}(1) - (1-L) \tilde{\mathbb{F}}_{r,t}(L) \quad (\text{B.2})$$

where  $\tilde{\mathbb{F}}_{r,t}(L) = \sum_{j=1}^{\infty} \tilde{\mathbb{F}}_{r,j,t} L^j$  and  $\tilde{\mathbb{F}}_{r,j,t} = \sum_{k=j+1}^{\infty} \mathbb{B}_t^{r+k}(1) \otimes \mathbb{B}_t^k(1)$ . For notational simplicity, denote

$$\begin{aligned}\tilde{X}_{at} &= \sum_{j=1}^{\infty} \left( \mathbb{B}_t^j(1) \otimes \mathbb{B}_t^j(1) \right) \text{vec} \left( \boldsymbol{\epsilon}_{t-j} \boldsymbol{\epsilon}_{t-j}^\top \otimes \mathbf{I}_d \right), \\ \tilde{X}_{bt} &= \sum_{r=1}^{\infty} \sum_{j=1}^{\infty} \left( \mathbb{B}_t^{r+j}(1) \otimes \mathbb{B}_t^j(1) \right) \text{vec} \left( \boldsymbol{\epsilon}_{t-j} \boldsymbol{\epsilon}_{t-r-j}^\top \otimes \mathbf{I}_d \right).\end{aligned}$$

Applying (B.2) to  $\tilde{X}_{at}$  and  $\tilde{X}_{bt}$  yields that

$$\begin{aligned}\tilde{X}_{at} &= \mathbb{F}_{0,t}(1) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \otimes \mathbf{I}_d \right) - (1-L) \tilde{\mathbb{F}}_{0,t}(L) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \otimes \mathbf{I}_d \right), \\ \tilde{X}_{bt} &= \sum_{r=1}^{\infty} \mathbb{F}_{r,t}(1) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_{t-r}^\top \otimes \mathbf{I}_d \right) - (1-L) \sum_{r=1}^{\infty} \tilde{\mathbb{F}}_{r,t}(L) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_{t-r}^\top \otimes \mathbf{I}_d \right).\end{aligned}$$

For  $J_{T,91}$ , summing up  $\tilde{X}_{at}$  over  $t$  yields

$$\begin{aligned}& \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \sum_{r=1}^{\infty} \left( \mathbb{B}_t^r(1) \otimes \mathbb{B}_t^r(1) \right) \left( \text{vec} \left( \boldsymbol{\epsilon}_{t-r} \boldsymbol{\epsilon}_{t-r}^\top \otimes \mathbf{I}_d \right) - \text{vec}(\mathbf{I}_{d^2}) \right) \right\| \\ & \leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \mathbb{F}_{0,t}(1) \left( \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \otimes \mathbf{I}_d \right) - \text{vec}(\mathbf{I}_{d^2}) \right) \right\| \\ & \quad + \max_{1 \leq l \leq N_T} \left\| \|\mathbf{W}_{T,1}(s_l)\| \tilde{\mathbb{F}}_{0,1}(L) \text{vec} \left( \boldsymbol{\epsilon}_0 \boldsymbol{\epsilon}_0^\top \otimes \mathbf{I}_d \right) \right\| + \sup_{0 \leq r \leq 1} \left\| \|\mathbf{W}_{T,T}(s_l)\| \tilde{\mathbb{F}}_{0,T}(L) \text{vec} \left( \boldsymbol{\epsilon}_T \boldsymbol{\epsilon}_T^\top \otimes \mathbf{I}_d \right) \right\| \\ & \quad + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^{T-1} \left( \|\mathbf{W}_{T,t+1}(s_l)\| \tilde{\mathbb{F}}_{0,t+1}(L) - \|\mathbf{W}_{T,t}(s_l)\| \tilde{\mathbb{F}}_{0,t}(L) \right) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_t^\top \otimes \mathbf{I}_d \right) \right\| \\ & := J_{T,101} + J_{T,102} + J_{T,103} + J_{T,104}.\end{aligned}$$

Similar to the proof of Lemma B.5.1, we can show that  $J_{T,101} = O_P(\sqrt{d_T \log T})$ , since

$$\begin{aligned}\max_{t \geq 1} \|\mathbb{F}_{0,t}(1)\| & \leq \max_{t \geq 1} \sum_{j=1}^{\infty} \left\| \sum_{k=0}^{\infty} \mathbf{B}_{k+j,t} \otimes \mathbf{B}_{k,t} \right\|^2 \leq \max_{t \geq 1} \sum_{j=1}^{\infty} \left( \sum_{k=0}^{\infty} \|\mathbf{B}_{k+j,t}\|^2 \right) \left( \sum_{k=0}^{\infty} \|\mathbf{B}_{k,t}\|^2 \right) \\ & \leq \max_{t \geq 1} \left( \sum_{k=0}^{\infty} \|\mathbf{B}_{k,t}\|^2 \right) \left( \sum_{j=1}^{\infty} j \|\mathbf{B}_{j,t}\|^2 \right) < \infty.\end{aligned}$$

Also, we can show that  $J_{T,102} = O_P(d_T)$  and  $J_{T,103} = O_P(d_T)$ , since

$$\begin{aligned}\max_{t \geq 1} \|\tilde{\mathbb{F}}_{0,t}(1)\| & \leq \max_{t \geq 1} \sum_{j=1}^{\infty} \sum_{k=j+1}^{\infty} \left\| \mathbb{B}_t^k(1) \right\|^2 \leq \max_{t \geq 1} \sum_{r=1}^{\infty} \sum_{k=r+1}^{\infty} \left( \sum_{j=0}^{\infty} \|\mathbf{B}_{j+k,t}\|^2 \right) \left( \sum_{j=0}^{\infty} \|\mathbf{B}_{j,t}\|^2 \right) \\ & \leq \max_{t \geq 1} \left( \sum_{j=0}^{\infty} \|\mathbf{B}_{j,t}\|^2 \right) \left( \sum_{r=1}^{\infty} \sum_{k=r+1}^{\infty} (k-r) \|\mathbf{B}_{k,t}\|^2 \right) \\ & \leq \max_{t \geq 1} \left( \sum_{j=0}^{\infty} \|\mathbf{B}_{j,t}\|^2 \right) \left( \sum_{r=1}^{\infty} \frac{r(r+1)}{2} \|\mathbf{B}_{r+1,t}\|^2 \right)\end{aligned}$$

$$\leq \max_{t \geq 1} \left( \sum_{j=0}^{\infty} \|\mathbf{B}_{j,t}\|^2 \right) \left( \sum_{j=1}^{\infty} j^2 \|\mathbf{B}_{j,t}\|^2 \right) < \infty.$$

We can easily show  $J_{T,104} = o_P(1)$ , since

$$\sup_{\tau \in [a,b]} \left( \sum_{t=1}^{T-1} \|\mathbf{W}_{T,t+1}(\tau)\| - \|\mathbf{W}_{T,t}(\tau)\| \right) \leq \sup_{\tau \in [a,b]} \sum_{t=1}^{T-1} \|\mathbf{W}_{T,t+1}(\tau) - \mathbf{W}_{T,t}(\tau)\| = o(1)$$

and

$$\begin{aligned} & \sum_{t=1}^{T-1} \left\| \tilde{\mathbb{F}}_{0,t+1}(1) - \tilde{\mathbb{F}}_{0,t}(1) \right\| \leq \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \sum_{k=r+1}^{\infty} \left\| \mathbb{B}_{t+1}^k(1) \otimes \mathbb{B}_{t+1}^k(1) - \mathbb{B}_t^k(1) \otimes \mathbb{B}_t^k(1) \right\| \\ & \leq \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \sum_{k=r+1}^{\infty} \left\| \mathbb{B}_{t+1}^k(1) - \mathbb{B}_t^k(1) \right\| \cdot \left( \left\| \mathbb{B}_{t+1}^k(1) \right\| + \left\| \mathbb{B}_t^k(1) \right\| \right) \\ & \leq M \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \sum_{k=r+1}^{\infty} \sum_{j=0}^{\infty} \left\| \mathbf{B}_{j+k,t+1} \otimes \mathbf{B}_{j,t+1} - \mathbf{B}_{j+k,t} \otimes \mathbf{B}_{j,t} \right\| \\ & \leq M \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \sum_{k=r+1}^{\infty} \sum_{j=0}^{\infty} \left( \left\| \mathbf{B}_{j+k,t+1} - \mathbf{B}_{j+k,t} \right\| \cdot \left\| \mathbf{B}_{j,t+1} \right\| + \left\| \mathbf{B}_{j,t+1} - \mathbf{B}_{j,t} \right\| \cdot \left\| \mathbf{B}_{j+k,t} \right\| \right) \\ & \leq M \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \sum_{k=r+1}^{\infty} \left( \left\| \mathbf{B}_{k,t+1} - \mathbf{B}_{k,t} \right\| \cdot \sum_{j=0}^{\infty} \left\| \mathbf{B}_{j,t+1} \right\| + \left\| \mathbf{B}_{k,t} \right\| \cdot \sum_{j=0}^{\infty} \left\| \mathbf{B}_{j,t+1} - \mathbf{B}_{j,t} \right\| \right) \\ & \leq M \left( \sum_{t=1}^{T-1} \sum_{k=1}^{\infty} k \left\| \mathbf{B}_{k,t+1} - \mathbf{B}_{k,t} \right\| \right) \cdot \left( \max_{t \geq 1} \sum_{j=0}^{\infty} \left\| \mathbf{B}_{j,t+1} \right\| \right) \\ & \quad + M \left( \max_{t \geq 1} \sum_{k=1}^{\infty} k \left\| \mathbf{B}_{k,t} \right\| \right) \cdot \left( \sum_{t=1}^{T-1} \sum_{j=0}^{\infty} \left\| \mathbf{B}_{j,t+1} - \mathbf{B}_{j,t} \right\| \right) = O(1). \end{aligned}$$

Based on the above development, we conclude that  $J_{T,91} = o_P(1)$ . Next, we focus on  $J_{T,92}$ , and write

$$\begin{aligned} & \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \sum_{r=1}^{\infty} \sum_{j=1}^{\infty} \mathbb{B}_t^{r+j}(1) \otimes \mathbb{B}_t^r(1) \text{vec} \left( \boldsymbol{\epsilon}_{t-r} \boldsymbol{\epsilon}_{t-r-j}^\top \otimes \mathbf{I}_d \right) \right\| \\ & \leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| \sum_{r=1}^{\infty} \mathbb{F}_{r,t}(1) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_{t-r}^\top \otimes \mathbf{I}_d \right) \right\| \\ & \quad + \max_{1 \leq l \leq N_T} \left\| \|\mathbf{W}_{T,1}(s_l)\| \sum_{r=1}^{\infty} \tilde{\mathbb{F}}_{r,1}(L) \text{vec} \left( \boldsymbol{\epsilon}_0 \boldsymbol{\epsilon}_{-r}^\top \otimes \mathbf{I}_d \right) \right\| \\ & \quad + \max_{1 \leq l \leq N_T} \left\| \|\mathbf{W}_{T,T}(s_l)\| \sum_{r=1}^{\infty} \tilde{\mathbb{F}}_{r,T}(L) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_{T-r}^\top \otimes \mathbf{I}_d \right) \right\| \\ & \quad + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \left( \|\mathbf{W}_{T,t+1}(s_l)\| \tilde{\mathbb{F}}_{r,t+1}(L) - \|\mathbf{W}_{T,t}(s_l)\| \tilde{\mathbb{F}}_{r,t}(L) \right) \text{vec} \left( \boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_{t-r}^\top \otimes \mathbf{I}_d \right) \right\| \\ & := J_{T,111} + J_{T,112} + J_{T,113} + J_{T,114}. \end{aligned}$$

We can show that  $J_{T,112}$  and  $J_{T,113}$  are  $O_P(d_T)$ , since

$$\begin{aligned}
& \max_{t \geq 1} \left\| \sum_{r=1}^{\infty} \tilde{\mathbb{F}}_{r,t}(1) \right\| \leq \max_{t \geq 1} \sum_{r=1}^{\infty} \sum_{j=1}^{\infty} \left\| \tilde{\mathbb{F}}_{rj,t} \right\| \leq \max_{t \geq 1} \sum_{r=1}^{\infty} \sum_{j=1}^{\infty} \sum_{k=j+1}^{\infty} \left\| \mathbb{B}_t^{r+k}(1) \right\| \left\| \mathbb{B}_t^k(1) \right\| \\
& \leq \max_{t \geq 1} \left( \sum_{r=1}^{\infty} \left\| \mathbb{B}_t^r(1) \right\| \right) \left( \sum_{j=1}^{\infty} j \left\| \mathbb{B}_t^j(1) \right\| \right) \leq M \max_{t \geq 1} \sum_{j=1}^{\infty} j \sum_{k=0}^{\infty} \left\| \mathbf{B}_{k+j,t} \right\| \left\| \mathbf{B}_{k,t} \right\| \\
& \leq M \max_{t \geq 1} \left( \sum_{j=1}^{\infty} j \left\| \mathbf{B}_{j,t} \right\| \right) \left( \sum_{k=0}^{\infty} \left\| \mathbf{B}_{k,t} \right\| \right) < \infty.
\end{aligned}$$

Similar to  $J_{T,104}$ , we have  $J_{T,114} = o_P(1)$ , since

$$\begin{aligned}
& \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \left\| \tilde{\mathbb{F}}_{r,t+1}(1) - \tilde{\mathbb{F}}_{r,t}(1) \right\| \leq \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \sum_{j=1}^{\infty} \sum_{k=j+1}^{\infty} \left\| \mathbb{B}_{t+1}^{r+k}(1) \otimes \mathbb{B}_{t+1}^k(1) - \mathbb{B}_t^{r+k}(1) \otimes \mathbb{B}_t^k(1) \right\| \\
& \leq \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \sum_{j=1}^{\infty} \sum_{k=j+1}^{\infty} \left( \left\| \mathbb{B}_{t+1}^{r+k}(1) - \mathbb{B}_t^{r+k}(1) \right\| \left\| \mathbb{B}_{t+1}^k(1) \right\| + \left\| \mathbb{B}_{t+1}^k(1) - \mathbb{B}_t^k(1) \right\| \left\| \mathbb{B}_t^{r+k}(1) \right\| \right) \\
& \leq \left( \sum_{t=1}^{T-1} \sum_{r=1}^{\infty} \left\| \mathbb{B}_{t+1}^r(1) - \mathbb{B}_t^r(1) \right\| \right) \left( \max_{t \geq 1} \sum_{j=1}^{\infty} j \left\| \mathbb{B}_t^j \right\| \right) \\
& \quad + \left( \sum_{t=1}^{T-1} \sum_{j=1}^{\infty} j \left\| \mathbb{B}_{t+1}^j(1) - \mathbb{B}_t^j(1) \right\| \right) \left( \max_{t \geq 1} \sum_{r=1}^{\infty} \left\| \mathbb{B}_t^r \right\| \right) = O(1).
\end{aligned}$$

Now consider term  $J_{T,111}$ . Define  $\mathbf{u}_t = \sum_{r=1}^{\infty} \mathbb{F}_{r,t}(1) \text{vec}(\boldsymbol{\epsilon}_t \boldsymbol{\epsilon}_{t-r}^\top \otimes \mathbf{I}_d)$ ,  $\mathbf{u}'_t = \mathbf{u}_t I(\|\mathbf{u}_t\| \leq T^{\frac{2}{\delta}})$  and  $\mathbf{u}''_t = \mathbf{u}_t - \mathbf{u}'_t$ . Then we have

$$\begin{aligned}
J_{T,111} &= \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \left\| \mathbf{W}_{T,t}(s_l) \right\| (\mathbf{u}'_t + \mathbf{u}''_t - E(\mathbf{u}'_t + \mathbf{u}''_t | \mathcal{F}_{t-1})) \right\| \\
&\leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \left\| \mathbf{W}_{T,t}(s_l) \right\| (\mathbf{u}'_t - E(\mathbf{u}'_t | \mathcal{F}_{t-1})) \right\| + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \left\| \mathbf{W}_{T,t}(s_l) \right\| \mathbf{u}''_t \right\| \\
&\quad + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T \left\| \mathbf{W}_{T,t}(s_l) \right\| E(\mathbf{u}''_t | \mathcal{F}_{t-1}) \right\| := J_{T,121} + J_{T,122} + J_{T,123}.
\end{aligned}$$

Using the same argument as that used in the proof of  $J_{T,72}$  in Lemma B.4, we can show that  $J_{T,122}$  and  $J_{T,123}$  are  $O_P(T^{\frac{2}{\delta}} d_T)$ . Next, consider  $J_{T,121}$ . For any  $1 \leq l \leq N_T$ , let  $\mathbf{Y}_t = \left\| \mathbf{W}_{T,t}(s_l) \right\| (\mathbf{u}'_t - E(\mathbf{u}'_t | \mathcal{F}_{t-1}))$ . We then have  $E(\mathbf{Y}_t | \mathcal{F}_{t-1}) = 0$  and  $\|\mathbf{Y}_t\| \leq 2T^{2/\delta} d_T$ . In addition, we have

$$\begin{aligned}
& \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq 4 \max_{1 \leq l \leq N_T} \sum_{t=1}^T \left\| \mathbf{W}_{T,t}(s_l) \right\|^2 E(\|\mathbf{u}_t\|^2 | \mathcal{F}_{t-1}) \\
& \leq M \cdot d_T \max_{1 \leq l \leq N_T} \sum_{t=1}^T \left\| \mathbf{W}_{T,t}(s_l) \right\| \sum_{r=1}^{\infty} \left\| \mathbb{F}_{r,t}(1) \right\|^2 \|\boldsymbol{\epsilon}_{t-r}\|^2 \leq M \cdot d_T \max_{t \geq 1} \sum_{r=1}^{\infty} \left\| \mathbb{F}_{r,t}(1) \right\|^2 \|\boldsymbol{\epsilon}_{t-r}\|^2
\end{aligned}$$

$$\leq M \cdot d_T \sum_{r=1}^{\infty} \max_{t \geq 1} \|\mathbb{F}_{r,t}(1)\|^2 \left( \sum_{t=1}^T \|\epsilon_{t-r}\|^\delta \right)^{\frac{2}{\delta}} = O_P \left( d_T T^{\frac{2}{\delta}} \right).$$

Therefore, we have  $\max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| = O_P(d_T T^{\frac{2}{\delta}})$ . By Lemma B.2, and choosing  $\beta = 4$ , we have

$$\begin{aligned} & \Pr \left( J_{T,121} > \sqrt{\beta M} \sqrt{d_T T^{\frac{2}{\delta}} \log T} \right) \\ &= \Pr \left( J_{T,121} > \sqrt{\beta M} \sqrt{d_T T^{\frac{2}{\delta}} \log T}, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T T^{\frac{2}{\delta}} \right) \\ & \quad + \Pr \left( J_{T,121} > \sqrt{\beta M} \sqrt{d_T T^{\frac{2}{\delta}} \log T}, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T T^{\frac{2}{\delta}} \right) \\ &\leq \Pr \left( J_{T,121} > \sqrt{\beta M} \sqrt{d_T T^{\frac{2}{\delta}} \log T}, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T T^{\frac{2}{\delta}} \right) \\ & \quad + \Pr \left( \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T T^{\frac{2}{\delta}} \right) \\ &\leq N_T \exp \left( - \frac{\beta M d_T T^{\frac{2}{\delta}} \log T}{2(M d_T T^{\frac{2}{\delta}} + M \sqrt{d_T T^{\frac{2}{\delta}} \log T T^{\frac{2}{\delta}} d_T})} \right) + o(1) \\ &\leq N_T \exp \left( - \frac{\beta}{2} \log T \right) = N_T T^{-\frac{\beta}{2}} = o(1) \end{aligned}$$

given  $d_T T^{\frac{4}{\delta}} \log T \rightarrow 0$ . Hence, we have  $J_{T,611} = O_P(\{d_T T^{\frac{2}{\delta}} \log T\}^{1/2})$ . Combining the above results, we have proved that  $\sup_{\tau \in [a,b]} \left| \sum_{t=1}^T \|\mathbf{W}_{T,t}(\tau)\| E \left( \|\zeta_t \epsilon_t\|^2 | \mathcal{F}_{t-1} \right) \right| = O_P(1)$ .

Finally, we turn to  $J_{T,81}$ , and apply the truncation method. Let  $\mathbf{u}_t = \zeta_t \epsilon_t$ ,  $\mathbf{u}'_t = \mathbf{u}_t I \left( \|\mathbf{u}_t\| \leq T^{\frac{2}{\delta}} \right)$  and  $\mathbf{u}''_t = \mathbf{u}_t - \mathbf{u}'_t$ . Then we have

$$\begin{aligned} J_{T,81} &= \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) (\mathbf{u}'_t + \mathbf{u}''_t - E(\mathbf{u}'_t + \mathbf{u}''_t | \mathcal{F}_{t-1})) \right\| \\ &\leq \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) (\mathbf{u}'_t - E(\mathbf{u}'_t | \mathcal{F}_{t-1})) \right\| + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) \mathbf{u}''_t \right\| \\ &\quad + \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) E(\mathbf{u}''_t | \mathcal{F}_{t-1}) \right\| = J_{T,131} + J_{T,132} + J_{T,133}. \end{aligned}$$

It's easy to show that  $J_{T,132} = O_P(T^{\frac{2}{\delta}} d_T)$  and  $J_{T,133} = O_P(T^{\frac{2}{\delta}} d_T)$ . Thus, we focus on  $J_{T,131}$ .

For any  $1 \leq l \leq N_T$ , let  $\mathbf{Y}_t = (\mathbf{I}_d \otimes \mathbf{W}_{T,t}(s_l)) (\mathbf{u}'_t - E(\mathbf{u}'_t | \mathcal{F}_{t-1}))$ , then we have  $E(\mathbf{Y}_t | \mathcal{F}_{t-1}) = 0$  and  $\|\mathbf{Y}_t\| \leq 2T^{2/\delta} d_T$ . Also,

$$\max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| E \left( \|\zeta_t \epsilon_t\|^2 | \mathcal{F}_{t-1} \right) = O_P(1),$$

which yields

$$\max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T \max_{1 \leq l \leq N_T} \sum_{t=1}^T \|\mathbf{W}_{T,t}(s_l)\| E(\|\mathbf{u}_t\|^2 | \mathcal{F}_{t-1}) = O_P(d_T).$$

Therefore, we have  $\max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| = O_P(d_T)$ . By Lemma B.2 and choosing  $\beta = 4$ , we have

$$\begin{aligned} \Pr\left(J_{T,131} > \sqrt{\beta M} \gamma_T\right) &= \Pr\left(J_{T,31} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T\right) \\ &+ \Pr\left(J_{T,131} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T\right) \\ &\leq \Pr\left(J_{T,131} > \sqrt{\beta M} \gamma_T, \max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| \leq M d_T\right) \\ &+ \Pr\left(\max_{1 \leq l \leq N_T} \left\| \sum_{t=1}^T E(\mathbf{Y}_t \mathbf{Y}_t^\top | \mathcal{F}_{t-1}) \right\| > M d_T\right) \leq N_T \exp\left(-\frac{\beta M \gamma_T^2}{2(M d_T + M \gamma_T T^{\frac{2}{\delta}} d_T)}\right) + o(1) \\ &\leq N_T \exp\left(-\frac{\beta}{2} \log T\right) = N_T T^{-\frac{\beta}{2}} = o(1) \end{aligned}$$

given  $d_T T^{\frac{4}{\delta}} \log T \rightarrow 0$ .

We now have completed the proof of the second result.  $\square$

*Proof of Lemma B.6.*

Let  $\Psi_j(\tau) = \mathbf{J} \Phi^j(\tau) \mathbf{J}^\top$ , where

$$\Phi(\tau) = \begin{pmatrix} \mathbf{A}_1(\tau) & \cdots & \mathbf{A}_{p-1}(\tau) & \mathbf{A}_p(\tau) \\ \mathbf{I}_d & \cdots & \mathbf{0}_d & \mathbf{0}_d \\ \vdots & \ddots & \vdots & \vdots \\ \mathbf{0}_d & \cdots & \mathbf{I}_d & \mathbf{0}_d \end{pmatrix}$$

and  $\mathbf{J} = [\mathbf{I}_d, \mathbf{0}_{d \times d(p-1)}]$ .

To proceed, we write  $\mathbf{y}_t$  as a time-varying VMA( $\infty$ ):

$$\mathbf{y}_t = \sum_{j=0}^{\infty} \Psi_{j,t} \left( \sum_{l=0}^q \mathbf{B}_l(\tau_{t-j}) \mathbf{x}_{t-l-j} + \boldsymbol{\eta}_{t-j} \right) = \boldsymbol{\mu}_t^* + \sum_{j=0}^{\infty} \mathbf{D}_{j,t}^\epsilon \boldsymbol{\epsilon}_{t-j} + \sum_{l=0}^q \sum_{j=0}^{\infty} \mathbf{D}_{j,l,t}^v \mathbf{v}_{t-l-j},$$

where  $\boldsymbol{\mu}_t^* = \sum_{j=0}^{\infty} \Psi_{j,t}(\boldsymbol{\mu}(\tau_{t-j}) + \sum_{l=0}^q \mathbf{B}_l(\tau_{t-j}) \mathbf{g}(\tau_{t-l-j}))$ ,  $\Psi_{j,t} = \mathbf{J} \prod_{m=0}^{j-1} \Phi(\tau_{t-m}) \mathbf{J}^\top$ ,  $\mathbf{D}_{j,t}^\epsilon = \Psi_{j,t} \boldsymbol{\omega}(\tau_{t-j})$ , and  $\mathbf{D}_{j,l,t}^v = \sum_{k=0}^j \Psi_{k,t} \mathbf{B}_l(\tau_{t-k}) \mathbf{C}_{j-k}(\tau_{t-l-k})$ .

Let  $\rho$  denote the largest eigenvalue of  $\Phi(\tau)$  uniformly over  $\tau \in [0, 1]$ . Then we have  $\rho < 1$  by Assumption 1.1. In addition, similar to the proof of Proposition 2.4 in Dahlhaus and Polonik (2009),

we have  $\max_{t \geq 1} \|\prod_{m=0}^{j-1} \Phi(\tau_{t-m})\| \leq M\rho^j$ .

Next, we will show that  $\mathbf{y}_t$  can be approximated by a time-varying MA( $\infty$ ) process  $\tilde{\mathbf{y}}_t$  satisfying  $\{E\|\mathbf{y}_t - \tilde{\mathbf{y}}_t\|^\delta\}^{1/\delta} = O(T^{-1})$ , where  $\tilde{\mathbf{y}}_t$  has been defined in the body of this lemma. It follows that

$$\begin{aligned} \{E\|\mathbf{y}_t - \tilde{\mathbf{y}}_t\|^\delta\}^{1/\delta} &\leq M \left( \|\boldsymbol{\mu}_t^* - \boldsymbol{\mu}^*(\tau_t)\| + \sum_{j=0}^{\infty} \|\mathbf{D}_{j,t}^\epsilon - \mathbf{D}_j^\epsilon(\tau_t)\| + \sum_{l=0}^q \sum_{j=0}^{\infty} \|\mathbf{D}_{j,l,t}^v - \mathbf{D}_{j,l}^v(\tau_t)\| \right) \\ &:= O(1) \cdot (J_{T,141} + J_{T,142} + J_{T,143}), \end{aligned}$$

where the definitions of  $J_{T,141}$ ,  $J_{T,142}$ , and  $J_{T,143}$  are obvious.

Consider  $J_{T,141}$ . Note that for any conformable matrices  $\{\mathbf{A}_i\}$  and  $\{\mathbf{B}_i\}$ , since

$$\prod_{i=1}^r \mathbf{A}_i - \prod_{i=1}^r \mathbf{B}_i = \sum_{j=1}^r \left( \prod_{k=1}^{j-1} \mathbf{A}_k \right) (\mathbf{A}_j - \mathbf{B}_j) \left( \prod_{k=j+1}^r \mathbf{B}_k \right),$$

we obtain

$$\begin{aligned} \|\Psi_{j,t} - \Psi_j(\tau_t)\| &= \left\| \mathbf{J} \prod_{m=0}^{j-1} \Phi(\tau_{t-m}) \mathbf{J}^\top - \mathbf{J} \Phi^j(\tau_t) \mathbf{J}^\top \right\| \\ &\leq M \sum_{i=1}^{j-1} \left\| \Phi^i(\tau_t) (\Phi(\tau_{t-i}) - \Phi(\tau_t)) \prod_{m=i+1}^{j-1} \Phi(\tau_{t-m}) \right\| \leq M \sum_{i=1}^{j-1} \frac{i}{T} \rho^{j-1}. \end{aligned}$$

Hence, we have

$$\begin{aligned} J_{T,141} &\leq \sum_{j=0}^{\infty} \|\Psi_{j,t} - \Psi_j(\tau_t)\| \cdot \|\boldsymbol{\mu}(\tau_{t-j})\| + \sum_{j=0}^{\infty} \|\Psi_j(\tau_t)\| \cdot \|\boldsymbol{\mu}(\tau_{t-j}) - \boldsymbol{\mu}(\tau_t)\| \\ &\quad + \sum_{j=0}^{\infty} \|\Psi_{j,t} - \Psi_j(\tau_t)\| \cdot \sum_{l=0}^q \|\mathbf{B}_l(\tau_{t-j}) \mathbf{g}(\tau_{t-l-j})\| \\ &\quad + \sum_{j=0}^{\infty} \|\Psi_j(\tau_t)\| \cdot \sum_{l=0}^q \|\mathbf{B}_l(\tau_{t-j}) \mathbf{g}(\tau_{t-l-j}) - \mathbf{B}_l(\tau_t) \mathbf{g}(\tau_t)\| \\ &\leq M \sum_{j=0}^{\infty} \sum_{i=1}^{j-1} \frac{i}{T} \rho^{j-1} + M \sum_{j=0}^{\infty} \rho^j \frac{j}{T} = O(T^{-1}), \end{aligned}$$

where we have used the facts that  $\|\Psi_j(\tau)\| \leq M\rho^j$  and

$$\begin{aligned} &\|\mathbf{B}_l(\tau_{t-j}) \mathbf{g}(\tau_{t-l-j}) - \mathbf{B}_l(\tau_t) \mathbf{g}(\tau_t)\| \\ &= \|\mathbf{B}_l(\tau_{t-j}) \mathbf{g}(\tau_{t-l-j}) - \mathbf{B}_l(\tau_t) \mathbf{g}(\tau_{t-l-j}) + \mathbf{B}_l(\tau_t) \mathbf{g}(\tau_{t-l-j}) - \mathbf{B}_l(\tau_t) \mathbf{g}(\tau_t)\| \\ &\leq \|\mathbf{B}_l(\tau_{t-j}) - \mathbf{B}_l(\tau_t)\| \cdot \|\mathbf{g}(\tau_{t-l-j})\| + \|\mathbf{B}_l(\tau_t)\| \cdot \|\mathbf{g}(\tau_{t-l-j}) - \mathbf{g}(\tau_t)\| \leq M \frac{j}{T}. \end{aligned}$$

Similarly, we have  $J_{T,142} = O(T^{-1})$ .



For  $J_{T,143}$ , we have

$$\begin{aligned}
J_{T,143} &\leq \sum_{l=0}^q \sum_{j=0}^{\infty} \sum_{k=0}^j \|\Psi_{k,t} \mathbf{B}_l(\tau_{t-k}) \mathbf{C}_{j-k}(\tau_{t-l-k}) - \Psi_k(\tau_t) \mathbf{B}_l(\tau_t) \mathbf{C}_{j-k}(\tau_t)\| \\
&= \sum_{l=0}^q \sum_{j=0}^{\infty} \sum_{k=0}^{\infty} \|\Psi_{j,t} \mathbf{B}_l(\tau_{t-j}) \mathbf{C}_k(\tau_{t-l-j}) - \Psi_j(\tau_t) \mathbf{B}_l(\tau_t) \mathbf{C}_k(\tau_t)\| \\
&\leq \sum_{l=0}^q \sum_{j=0}^{\infty} \sum_{k=0}^{\infty} \|\Psi_{j,t} - \Psi_j(\tau_t)\| \cdot \|\mathbf{B}_l(\tau_{t-j})\| \cdot \|\mathbf{C}_k(\tau_{t-l-j})\| \\
&\quad + \sum_{l=0}^q \sum_{j=0}^{\infty} \sum_{k=0}^{\infty} \|\Psi_j(\tau_t)\| \cdot \|\mathbf{B}_l(\tau_{t-j}) - \mathbf{B}_l(\tau_t)\| \cdot \|\mathbf{C}_k(\tau_{t-l-j})\| \\
&\quad + \sum_{l=0}^q \sum_{j=0}^{\infty} \sum_{k=0}^{\infty} \|\Psi_j(\tau_t)\| \cdot \|\mathbf{B}_l(\tau_t)\| \cdot \|\mathbf{C}_k(\tau_{t-l-j}) - \mathbf{C}_k(\tau_t)\| \\
&\leq M \sum_{j=0}^{\infty} \sum_{i=1}^{j-1} \frac{i}{T} \rho^{j-1} \cdot \sup_{\tau \in [0,1]} \sum_{k=0}^{\infty} \|\mathbf{C}_k(\tau)\| \\
&\quad + \frac{M}{T} \left( \sum_{j=0}^{\infty} j \rho^j \right) \left( \sup_{\tau \in [0,1]} \sum_{k=0}^{\infty} \|\mathbf{C}_k(\tau)\| + \sup_{\tau \in [0,1]} \sum_{k=0}^{\infty} \|\mathbf{C}_k^{(1)}(\tau)\| \right) = O(T^{-1}).
\end{aligned}$$

In addition, it is straightforward to verify that

$$\sup_{\tau \in [0,1]} \sum_{j=0}^{\infty} j \|\mathbf{D}_j^{\epsilon,(k)}(\tau)\| < \infty \quad \text{and} \quad \sup_{\tau \in [0,1]} \sum_{j=0}^{\infty} j \|\mathbf{D}_{j,l}^{\mathbf{v},(k)}(\tau)\| < \infty$$

for  $k = 0, 1$  (see, the proof of Propositions 2.1, for example.)

The proof is now completed. □

*Proof of Lemma B.7.*

(1)–(2). To prove parts (1) and (2), it suffices to show that

$$\sup_{\tau \in [0,1]} \left\| \frac{1}{T} \sum_{t=1}^T \left( \mathbf{Z}_t \mathbf{Z}_t^\top - E(\mathbf{Z}_t \mathbf{Z}_t^\top) \right) \left( \frac{\tau_t - \tau}{h} \right)^k K_h(\tau_t - \tau) \right\| = O_P \left( \sqrt{\frac{\log T}{Th}} \right)$$

for  $k = 0, 1, 2$ . Since  $\mathbf{Z}_t$  can be approximated by a time-varying VMA( $\infty$ ) process by Lemma B.6, then the uniform convergence results follow directly from Lemma 2.3.

(3). Part (3) follows directly from Lemmas B.6 and 2.3.

(4). The uniform convergence rate for  $\hat{\beta}(\tau)$  follows directly from the proof of Theorem 4.1 (1), Lemma 2.3 and part (3).

(5). By the proof of Theorem 4.1 (2), we have

$$\sup_{\tau \in [0,1]} \left\| \hat{\Omega}(\tau) - T^{-1} \sum_{t=1}^T \boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top K_h(\tau_t - \tau) \right\| = O_P(h^2 + \sqrt{\log T / (Th)}).$$

Similar to the proof of Lemma A.5 (1), we have

$$\sup_{\tau \in [0,1]} \left\| T^{-1} \sum_{t=1}^T \left[ \boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top - E(\boldsymbol{\eta}_t \boldsymbol{\eta}_t^\top) \right] K_h(\tau_t - \tau) \right\| = O_P(\sqrt{\log T / (Th)}).$$

The proof of part (5) is now complete.  $\square$

*Proof of Lemma B.8.*

(1). By Lemma B.7, we have

$$\sup_{\tau \in [0,1]} \left\| \mathbf{s}(\tau) \mathbf{Z}_1 - \boldsymbol{\Sigma}_{\mathbf{Z}_2}^{-1}(\tau) \boldsymbol{\Sigma}_{\mathbf{Z}_{1,2}}^\top(\tau) \right\| = o_P(1)$$

and  $\sup_{\tau \in [0,1]} \left\| \widehat{\boldsymbol{\Omega}}(\tau) - \boldsymbol{\Omega}(\tau) \right\| = o_P(1)$ . Hence, we have

$$\begin{aligned} & T^{-1} \sum_{t=1}^T \widetilde{\mathbf{Z}}_{1t} \widehat{\boldsymbol{\Omega}}^{-1}(\tau_t) \widetilde{\mathbf{Z}}_{1t}^\top \\ &= T^{-1} \sum_{t=1}^T \left( \mathbf{z}_{1t} - \boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{z}_2}^{-1}(\tau_t) \mathbf{z}_{2t}^\top \right) \left( \mathbf{z}_{1t}^\top - \mathbf{z}_{2t}^\top \boldsymbol{\Sigma}_{\mathbf{z}_2}^{-1}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}^\top(\tau_t) \right) \otimes \boldsymbol{\Omega}^{-1}(\tau_t) + o_P(1) \end{aligned}$$

Note that each element of  $\boldsymbol{\Sigma}_{\mathbf{z}_2}(\tau)$  and  $\boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}(\tau)$  is Lipschitz continuous. Thus, by Lemmas 2.2 and B.6, the result holds.

(2). Let  $\rho_T = h^2 + \sqrt{\log T / (Th)}$ . By Lemma B.7 (1), we have

$$\mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \left[ \boldsymbol{\theta}(\tau_1)^\top \mathbf{Z}_{21}, \dots, \boldsymbol{\theta}(\tau_T)^\top \mathbf{Z}_{2T} \right]^\top = \mathbf{Z}_{2t}^\top \boldsymbol{\theta}(\tau_t) (1 + O_P(\rho_T))$$

uniformly over  $1 \leq t \leq T$ . In addition, by Lemma B.7 (4), we have

$$\begin{aligned} & \sum_{t=1}^T \widetilde{\mathbf{Z}}_{1t} \widehat{\boldsymbol{\Omega}}^{-1}(\tau_t) \left( \mathbf{Z}_{2t}^\top \boldsymbol{\theta}(\tau_t) - \mathbf{Z}_{2t}^\top \mathbf{s}(\tau_t) \left[ \boldsymbol{\theta}(\tau_1)^\top \mathbf{Z}_{21}, \dots, \boldsymbol{\theta}(\tau_T)^\top \mathbf{Z}_{2T} \right]^\top \right) \\ &= \sum_{t=1}^T \left[ \left( \mathbf{z}_{1t} \mathbf{z}_{2t}^\top - \boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{z}_2}^{-1}(\tau_t) \mathbf{z}_{2t} \mathbf{z}_{2t}^\top (1 + O_P(\rho_T)) \right) \otimes \left( \frac{1}{T} \sum_{s=1}^T \boldsymbol{\Omega}(\tau_s) K_h(\tau_s - \tau_t) + O_P(\rho_T) \right) \right] \\ &\times \boldsymbol{\theta}(\tau_t) \cdot O_P(\rho_T) = \sum_{t=1}^T \left[ \left( \mathbf{z}_{1t} \mathbf{z}_{2t}^\top - \boldsymbol{\Sigma}_{\mathbf{z}_{1,2}}(\tau_t) \boldsymbol{\Sigma}_{\mathbf{z}_2}^{-1}(\tau_t) \mathbf{z}_{2t} \mathbf{z}_{2t}^\top \right) \otimes \left( \frac{1}{T} \sum_{s=1}^T \boldsymbol{\Omega}(\tau_s) K_h(\tau_s - \tau_t) \right) \right] \\ &\times \boldsymbol{\theta}(\tau_t) \cdot O_P(\rho_T) + O_P(T \rho_T^2) = O_P(\sqrt{T} \rho_T) + O_P(T \rho_T^2) \end{aligned}$$

where the last equality follows from Lemma 2.2. Finally, the result holds since  $O_P(T \rho_T^2) = o_P(\sqrt{T})$  by Assumption 6.

(3)–(4). By Lemma B.7, we have  $\sup_{\tau \in [0,1]} \|\mathbf{s}(\tau)\boldsymbol{\eta}\| = O_P\left(\sqrt{\log T/(Th)}\right)$  and

$$\sup_{\tau \in [0,1]} \left\| \widehat{\boldsymbol{\Omega}}(\tau) - \frac{1}{T} \sum_{t=1}^T \boldsymbol{\Omega}(\tau_t) K_h(\tau_t - \tau) \right\| = O_P\left(h^2 + \sqrt{\frac{\log T}{Th}}\right).$$

Then, parts (3) and (4) can be proved by using similar arguments of part (2).  $\square$

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