

# Online Supplement to “On Time-varying Panel Data Models with Time-varying Interactive Fixed Effects”

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## Abstract

This online supplement is composed of three sections. Sections A and B provide the proofs of the main results in Sections 3 and 4, respectively. Section C provides the proof of the technical lemmas in Appendix A.

## A Proofs of the Main Results in Section 3

This appendix provides the proofs of the main results in Section 3. We need some technical lemmas whose proofs are available in Appendix C.

Recall that  $\hat{V}_{NT}^{(r)}$  denotes the diagonal matrix of the first  $R$  largest eigenvalues of  $(NT)^{-1} \sum_{i=1}^N (Y_i^{(r)} - X_i^{(r)} \hat{\beta}_r)(Y_i^{(r)} - X_i^{(r)} \hat{\beta}_r)'$  arranged in decreasing order along its main diagonal line and  $H^{(r)} = (N^{-1} \Lambda_r' \Lambda_r)(T^{-1} F^{(r)'})' \hat{F}^{(r)} \hat{V}_{NT}^{(r)-1}$ . Let  $V_{NT}$  denote the diagonal matrix of the first  $R$  largest eigenvalues of  $(NT)^{-1} \sum_{i=1}^N (Y_i - X_i \tilde{\beta})(Y_i - X_i \tilde{\beta})'$  arranged in decreasing order along its main diagonal line and  $H = (N^{-1} \Lambda' \Lambda)(T^{-1} F' \tilde{F}) V_{NT}^{-1}$ . Let  $C_{NT} = \min\{\sqrt{T}h, \sqrt{N}, h^{-2}\}$  and  $C_{0NT} = \min\{\sqrt{T}, \sqrt{N}\}$ . Recall that we use  $\mathbf{X}_p^{(r)}$  to denote the  $p$ th sheet of the  $T \times N \times P$  matrix  $\mathbf{X}^{(r)}$  and  $X_i^{(r)} = (X_{i1}^{(r)}, \dots, X_{iT}^{(r)})'$ . With a little bit of abuse of notation, we also write  $X_t^{(r)} = (X_{1t}^{(r)}, \dots, X_{Nt}^{(r)})'$ . Similarly, let  $\mathbf{X}_p^{(r)}$ ,  $X_i^{(r)}$ , and  $X_t^{(r)}$  denote the non-kernel weighted version of  $\mathbf{X}_p$ ,  $X_i$ , and  $X_t$  which are respectively,  $T \times N$ ,  $T \times P$ , and  $N \times P$  matrices. Let  $\Delta_{it}^{(r)} = k_{h,tr}^{*1/2} \Delta_i(t, r)$  and  $\Delta_i^{(r)} = (\Delta_{i1}^{(r)}, \dots, \Delta_{iT}^{(r)})'$ . Define

$$\begin{aligned} B_{1t}^{(r)} &= k_{h,tr}^{*1/2} C_{1t}^{(r)} \frac{t-r}{T} + k_{h,tr}^{*1/2} C_{2t}^{(r)} \left(\frac{t-r}{T}\right)^2, \quad B_{2t,1}^{(r)} = N^{-1} \hat{V}_{NT}^{(r)-1} H^{(r)} E(F_t A_{2,tr}' \Lambda_r F_t k_{h,tr}^{*1/2} h^2 \kappa_2), \\ B_{2t,2}^{(r)} &= \hat{V}_{NT}^{(r)-1} \frac{1}{TN} \sum_s \left(\frac{s-r}{T}\right)^2 k_{h,sr}^* E[\bar{C}_{1s}^{(r)} A_{1,sr}' \Lambda_r F_t^{(r)}], \quad \text{and } B_{2t}^{(r)} = B_{2t,1}^{(r)} + B_{2t,2}^{(r)}, \end{aligned} \quad (\text{A.1})$$

where  $C_{1t}^{(r)}$ ,  $C_{2t}^{(r)}$ ,  $\bar{C}_{1t}^{(r)}$ ,  $A_{1,tr}$  and  $A_{2,tr}$  are defined before Theorem 3.2 in the paper. Let  $\hat{\delta}_r = \hat{\beta}_r - \beta_r$ . Let  $A_{l,it}$  denote the  $i$ th element of  $A_{l,tr} = X_t \beta_r^{(l)} + \Lambda_r^{(l)} F_t$  for  $l = 1, 2$ .

**Lemma A.1** *Let  $\mathcal{F} = \{F \in \mathbb{R}^{T \times R} : F' F / T = \mathbb{I}_R\}$ . Let  $F^{(r)0}$  and  $\lambda_{ir}^0$  denote the true values of  $F^{(r)}$  and  $\lambda_{ir}$ . Suppose that Assumptions A.1–A.4 hold. Then*

$$(i) \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} \varepsilon_i^{(r)} \right\| = O_P(T^{-1/2} + N^{-1/4}),$$

$$\begin{aligned}
(ii) \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir}^{0'} F^{(r)0'} M_{F^{(r)}} \varepsilon_i \right\| &= O_P(T^{-1/2} + N^{-1/4}), \\
(iii) \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N \varepsilon_i^{(r)'} (P_{F^{(r)}} - P_{F^{(r)0}}) \varepsilon_i^{(r)} \right\| &= O_P(T^{-1/2} + N^{-1/4}), \\
(iv) \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N \Delta_i^{(r)'} (P_{F^{(r)}} - P_{F^{(r)0}}) \Delta_i^{(r)} \right\| &= O_P(h^2), \\
(v) \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} \Delta_i^{(r)} \right\| &= O_P(h), \\
(vi) \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N \Delta_i^{(r)'} M_{F^{(r)}} F^{(r)0} \lambda_{ir} \right\| &= O_P(h), \\
(vii) \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N \Delta_i^{(r)'} (P_{F^{(r)}} - P_{F^{(r)0}}) \varepsilon_i^{(r)} \right\| &= O_P((T^{-1/2} + N^{-1/4})h).
\end{aligned}$$

**Lemma A.2** Suppose that Assumptions A.1–A.4 hold. Then

$$\begin{aligned}
(i) T^{-1} \hat{F}^{(r)'} (NT)^{-1} \sum_{i=1}^N (Y_i^{(r)} - X_i^{(r)} \hat{\beta}_r) (Y_i^{(r)} - X_i^{(r)} \hat{\beta}_r)' \hat{F}^{(r)} &= \hat{V}_{NT}^{(r)} = V_r + O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_r\|), \\
(ii) T^{-1} \hat{F}^{(r)'} F^{(r)} &= Q_r + O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_{rr}\|), \\
(iii) H^{(r)} &= Q_r^{-1} + O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_r\|), \\
(iv) H^{(r)} H^{(r)'} &= \Sigma_F^{-1} + O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_r\|), \\
(v) \|P_{\hat{F}^{(r)}} - P_{F^{(r)}}\|^2 &= O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|),
\end{aligned}$$

where  $V_r$  is the diagonal matrix consisting of the eigenvalues of  $\Sigma_{\Lambda_r}^{1/2} \Sigma_F \Sigma_{\Lambda_r}^{1/2}$  in descending order with  $\Upsilon_r$  being the corresponding (normalized) eigenvector matrix, and  $Q_r = V_r^{1/2} \Upsilon_r^{-1} \Sigma_{\Lambda_r}^{-1/2}$ .

**Lemma A.3** Suppose that Assumptions A.1–A.4 hold. Then

$$\begin{aligned}
(i) T^{-1} \left\| \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)} \right\|^2 &= O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|). \\
(ii) T^{-1} \left\| (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)})' F^{(r)} H^{(r)} \right\| &= O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|). \\
(iii) T^{-1} \left\| (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)})' \hat{F}^{(r)} \right\| &= O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|). \\
(iv) T^{-1} \left\| (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)})' X_i \right\| &= O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|) \text{ for } i = 1, 2, \dots, N. \\
(v) T^{-1} \left\| (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)})' \Delta_i^{(r)} \right\| &= O_P(C_{NT}^{-2} h^2) + O_P(\|\hat{\delta}_r\| h^2) \text{ for } i = 1, 2, \dots, N.
\end{aligned}$$

**Lemma A.4** Let  $\psi_{NT}^{(r)} = \frac{1}{N} \sum_{i=1}^N \sum_{k=1}^N \frac{1}{T} X_i^{(r)'} F^{(r)} [\frac{1}{T} F^{(r)'} F^{(r)}]^{-1} (\frac{1}{N} \Lambda_r' \Lambda_r)^{-1} \lambda_{kr} (\frac{1}{T} \sum_{t=1}^T \varepsilon_{it}^{(r)} \varepsilon_{kt}^{(r)})$ . Suppose that Assumptions A.1–A.4 hold. Then

$$\begin{aligned}
(i) N^{-1} T^{-1} \left\| \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} [\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)}] \lambda_{ir}' \right\| &= O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2}), \\
(ii) \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} B^{(r)} \lambda_{ir} &= O_P(h^2) + O_P(\|\hat{\delta}_r\|), \\
(iii) N^{-1} T^{-1} \left\| \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} \right\| &= O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2}), \\
(iv) \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} [\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)}] \varepsilon_i^{(r)} \right\|^2 &= O_P(C_{NT}^{-4} + \|\hat{\delta}_r\|^2), \\
(v) \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \frac{1}{T} X_i^{(r)'} F^{(r)} [\frac{1}{T} F^{(r)'} F^{(r)}]^{-1} [\hat{F}^{(r)} H^{(r)-1} - F^{(r)} - B^{(r)} H^{(r)-1}] \varepsilon_i^{(r)} &= \sqrt{\frac{Th}{N}} \psi_{NT}^{(r)} + O_P(\|\hat{\delta}_r\|) \\
&+ \sqrt{Th} O_P(C_{NT}^{-2}).
\end{aligned}$$

**Lemma A.5** Suppose that Assumptions A.1–A.4 hold. Then

$$\begin{aligned}
&\frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} F^{(r)} \lambda_{ir} \\
&= \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} a_{ji}^{(r)} \hat{\delta}_r - \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N a_{ij}^{(r)} X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} + \frac{1}{Th} \zeta_{NT}^{(r)} \\
&\quad - \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_j^{(r)} a_{ji}^{(r)} + O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-3} + h^4 + C_{NT}^{-1} h^3),
\end{aligned}$$

where  $\zeta_{NT}^{(r)} = -\frac{h}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir}$  and  $G^{(r)} = (T^{-1} F^{(r)'} \hat{F}^{(r)})^{-1} (N^{-1} \Lambda_r' \Lambda_r)^{-1}$ .

**Lemma A.6** Under Assumptions A.1–A.4, we have

$$\begin{aligned} & \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} \right] M_{\hat{F}^{(r)}} \varepsilon_i^{(r)} \\ &= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} \right] M_{F^{(r)}} \varepsilon_i^{(r)} + \sqrt{\frac{Th}{N}} \xi_{NT}^{(r)} \\ & \quad + \sqrt{NT}h O_P((C_{NT}^{-1} + h) \|\hat{\delta}_r\| + \|\hat{\delta}_r\|^{3/2}) + \sqrt{NT}h(Th)^{-1} O_P(\|\hat{\delta}_r\|^{1/2}) + o_P(1), \end{aligned}$$

where  $\xi_{NT}^{(r)} = -\frac{1}{N} \sum_{i=1}^N \sum_{k=1}^N \frac{1}{T} (X_i^{(r)} - V_i^{(r)})' F^{(r)} [\frac{1}{T} F^{(r)'} F^{(r)}]^{-1} [\frac{1}{N} \Lambda_r' \Lambda_r]^{-1} \lambda_{kr} [\frac{1}{T} \sum_{t=1}^T \varepsilon_{it}^{(r)} \varepsilon_{kt}^{(r)}]$  and  $V_i^{(r)} = N^{-1} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)}$ .

**Lemma A.7** Under Assumptions A.1–A.4, we have

$$\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} \right] M_{\hat{F}^{(r)}} \Delta_i^{(r)} = \sqrt{NT}h B_{1\beta}^{(r)} + o_P(1),$$

where

$$\begin{aligned} B_{1\beta}^{(r)} &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr} E[(X_{it} - V_{it,r}) A_{2,itr}] \left(\frac{t-r}{T}\right)^2 \\ & \quad + \frac{1}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,tr} k_{h,sr} (X_{it} - V_{it,r}) F_t' H^{(r)} [H^{(r)'} F_s A_{2,isr} + C_{1s}^{(r)} A_{1,isr}] \left(\frac{s-r}{T}\right)^2. \end{aligned}$$

**Lemma A.8** Suppose that Assumptions A.1–A.5 hold. Then

$$\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} \right] M_{F^{(r)}} \varepsilon_i^{(r)} - \sqrt{\frac{N}{Th}} B_{2\beta}^{(r)} \xrightarrow{d} N(0, \Omega_r),$$

where  $B_{2\beta}^{(r)} = (B_{2\beta,1}^{(r)}, \dots, B_{2\beta,p}^{(r)})'$  with  $B_{2\beta,p}^{(r)} = -\frac{h}{N} \text{tr}\{P_{F^{(r)}} E_C[\mathbf{X}_p^{(r)} \varepsilon^{(r)'}]\}$  for  $p \in [P]$ .

**Lemma A.9** Let  $\zeta_{NT}^{(r)} = -\frac{h}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} (\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} (\frac{1}{N} \Lambda_r' \Lambda_r)^{-1} \lambda_{ir}$ . Suppose that Assumptions A.1–A.4 hold. Then

- (i)  $[D^{(r)}(\hat{F}^{(r)})]^{-1} - [D^{(r)}(F^{(r)})]^{-1} = O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})$ ;
- (ii)  $\sqrt{Th}/N(\xi_{NT}^{(r)} - B_{3\beta}^{(r)}) = o_P(1)$ ;
- (iii)  $\sqrt{N}/(Th)(\zeta_{NT}^{(r)} - B_{4\beta}^{(r)}) = o_P(1)$ .

**Lemma A.10** Suppose that Assumptions A.1–A.4 hold. Then

- (i)  $\frac{\sqrt{NT}h}{NT} \text{tr}\left(P_{F^{(r)}} \left(\tilde{\mathbf{X}}_k^{(r)} \varepsilon^{(r)'} - E_C[\tilde{\mathbf{X}}_k^{(r)} \varepsilon^{(r)'}]\right)\right) = o_P(1)$
- (ii)  $\frac{\sqrt{NT}h}{NT} \text{tr}\left(\tilde{\mathbf{X}}_k^{(r)} P_{\Lambda_r} \varepsilon^{(r)'}\right) = o_P(1)$ ;
- (iii)  $\frac{\sqrt{NT}h}{NT} \text{tr}\left(P_{F^{(r)}} \tilde{\mathbf{X}}_k^{(r)} P_{\Lambda_r} \varepsilon^{(r)'}\right) = o_P(1)$ .

**Lemma A.11** Suppose that Assumptions A.1–A.4 and A.7 hold. Then

- (i)  $\max_t \left\| H^{(t)'} H^{(t)} - (F^{(t)'} F^{(t)})/T \right\|^{-1} = O_P(h^2 + (Th)^{-1} + N^{-1} \ln T)$ ;
- (ii)  $\max_t \left\| \left(\frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t\right)^{-1} - H^{(t)'} \left(\frac{1}{N} \Lambda_t' \Lambda_t\right)^{-1} H^{(t)} \right\| = O_P(h^2 + C_{NT}^{-2} \ln T)$ .

**Proof of Proposition 3.1** (i) Let  $S_{NT}(\beta_r, F^{(r)}) = \frac{1}{NT} \sum_{i=1}^N (Y_i^{(r)} - X_i^{(r)} \beta_r)' M_{F^{(r)}} (Y_i^{(r)} - X_i^{(r)} \beta_r)$ . Only in this proof will we use the superscript 0 to denote the true value. For example,  $F_t^{(r)0} = k_{h,tr}^{*1/2} F_t^0$  and

$F^{(r)0'} = (F_1^{(r)0}, \dots, F_T^{(r)0})'$ , where  $F_t^0$  denotes the true value of  $F_t$ . Similarly,  $\Lambda_r^0 = (\lambda_{1r}^0, \dots, \lambda_{Nr}^0)'$  and  $\beta_r^0$  denote the true values of  $\Lambda_r = (\lambda_{1r}, \dots, \lambda_{Nr})'$  and  $\beta_r$ . Recall that  $\Delta_{it}^{(r)} = k_{h,ir}^{*1/2} \Delta_i(t, r)$  and  $\Delta_i^{(r)} = (\Delta_{i1}^{(r)}, \dots, \Delta_{iT}^{(r)})'$ . Let  $\delta_r = \beta_r - \beta_r^0$  and  $\delta_{F,r} = F^{(r)} - F^{(r)0}$ . Noting that  $Y_{it}^{(r)} = X_{it}^{(r)'} \beta_r^0 + \lambda_{ir}^0 F_t^{(r)0} + \Delta_i^{(r)}(t, r) + \varepsilon_{it}^{(r)}$ , we have

$$\begin{aligned}
& S_{NT}(\beta_r, F^{(r)}) - S_{NT}(\beta_r^0, F^{(r)0}) \\
&= \frac{1}{NT} \sum_{i=1}^N \left[ -X_i^{(r)} \delta_r - \delta_{F,r} \lambda_{ir}^0 + \Delta_i^{(r)} + \varepsilon_i^{(r)} \right]' M_{F^{(r)}} \left[ -X_i^{(r)} \delta_r - \delta_{F,r} \lambda_{ir}^0 + \Delta_i^{(r)} + \varepsilon_i^{(r)} \right] \\
&\quad - \frac{1}{NT} \sum_{i=1}^N \left[ \Delta_i^{(r)} + \varepsilon_i^{(r)} \right]' M_{F^{(r)0}} \left[ \Delta_i^{(r)} + \varepsilon_i^{(r)} \right] \\
&= \frac{1}{NT} \sum_{i=1}^N \left[ X_i^{(r)} \delta_r + \delta_{F,r} \lambda_{ir}^0 \right]' M_{F^{(r)}} \left[ X_i^{(r)} \delta_r + \delta_{F,r} \lambda_{ir}^0 \right] - \frac{2}{NT} \sum_{i=1}^N \left[ X_i^{(r)} \delta_r + \delta_{F,r} \lambda_{ir}^0 \right]' M_{F^{(r)}} \left[ \Delta_i^{(r)} + \varepsilon_i^{(r)} \right] \\
&\quad + \frac{1}{NT} \sum_{i=1}^N \left[ \Delta_i^{(r)} + \varepsilon_i^{(r)} \right]' (P_{F^{(r)0}} - P_{F^{(r)}}) \left[ \Delta_i^{(r)} + \varepsilon_i^{(r)} \right] \\
&= \tilde{S}_{NT}(\beta_r, F^{(r)}) - 2\delta_r' \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} \varepsilon_i^{(r)} + \frac{2}{NT} \sum_{i=1}^N \lambda_{ir}^0 F^{(r)0'} M_{F^{(r)}} \varepsilon_i \\
&\quad + \frac{1}{NT} \sum_{i=1}^N \varepsilon_i^{(r)'} (P_{F^{(r)0}} - P_{F^{(r)}}) \varepsilon_i^{(r)} + \frac{1}{NT} \sum_{i=1}^N \Delta_i^{(r)'} (P_{F^{(r)0}} - P_{F^{(r)}}) \Delta_i^{(r)} - 2\delta_r' \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} \Delta_i^{(r)} \\
&\quad + \frac{2}{NT} \sum_{i=1}^N \Delta_i^{(r)'} M_{F^{(r)}} F^{(r)0} \lambda_{ir} + \frac{2}{NT} \sum_{i=1}^N \Delta_i^{(r)'} (P_{F^{(r)0}} - P_{F^{(r)}}) \varepsilon_i^{(r)},
\end{aligned}$$

where

$$\begin{aligned}
\tilde{S}_{NT}(\beta_r, F^{(r)}) &= \frac{1}{NT} \sum_{i=1}^N \left[ X_i^{(r)} \delta_r + \delta_{F,r} \lambda_{ir}^0 \right]' M_{F^{(r)}} \left[ X_i^{(r)} \delta_r + \delta_{F,r} \lambda_{ir}^0 \right] \\
&= \frac{1}{NT} \sum_{i=1}^N \delta_r' X_i^{(r)'} M_{F^{(r)}} X_i^{(r)} \delta_r + \text{tr} \left[ \frac{F^{(r)0'} M_{F^{(r)}} F^{(r)0}}{T} \frac{\Lambda_r^0 \Lambda_r^0}{N} \right] - \frac{2}{NT} \sum_{i=1}^N \delta_r' X_i^{(r)'} M_{F^{(r)}} F^{(r)0} \lambda_{ir}.
\end{aligned}$$

By Lemma A.1, we have

$$S_{NT}(\beta_r, F^{(r)}) = \tilde{S}_{NT}(\beta_r, F^{(r)}) + O_P(T^{-1/2} + N^{-1/4} + h)$$

uniformly over bounded  $\beta_r$  and over  $F^{(r)} \in \mathcal{F}$ . Clearly,  $\tilde{S}_{NT}(\beta_r^0, F^{(r)0}) H^{(r)} = 0$  because  $M_{F^{(r)0} H^{(r)}} F^{(r)0} = M_{F^{(r)0}} F^{(r)0} = 0$ . Following Bai (2009), we now show that  $S_{NT}(\beta_r, F^{(r)}) > 0$  for any  $(\beta_r, F^{(r)}) \neq (\beta_r^0, F^{(r)0}) H^{(r)}$  where  $H^{(r)}$  can be an arbitrary nonsingular matrix here. That is,  $S_{NT}(\beta_r, F^{(r)})$  attains its unique minimum value 0 at  $(\beta_r^0, F^{(r)0}) H^{(r)}$ . Denote

$$\mathcal{A}^{(r)} = \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} X_i^{(r)}, \quad \mathcal{B}^{(r)} = \frac{\Lambda_r' \Lambda_r}{N} \otimes \mathbb{I}_T, \quad \mathcal{C}^{(r)} = \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \otimes M_{F^{(r)}} X_i^{(r)}, \quad \text{and } \eta^{(r)} = \text{vec}(M_{F^{(r)}} F^{(r)0}).$$

Then, we have

$$\begin{aligned}
\tilde{S}_{NT}(\beta_r, F^{(r)}) &= \delta_r' \mathcal{A}^{(r)} \delta_r + \eta^{(r)'} \mathcal{B}^{(r)} \eta^{(r)} + 2\delta_r' \mathcal{C}^{(r)'} \eta^{(r)} \\
&= \delta_r' (\mathcal{A}^{(r)} - \mathcal{C}^{(r)'} \mathcal{B}^{(r)-1} \mathcal{C}^{(r)}) \delta_r + \theta^{(r)'} \mathcal{B}^{(r)} \theta^{(r)} = \delta_r' D(F^{(r)}) \delta_r + \theta^{(r)'} \mathcal{B}^{(r)} \theta^{(r)},
\end{aligned}$$

where  $\theta^{(r)} \equiv \eta^{(r)} + \mathcal{B}^{(r)-1} \mathcal{C}^{(r)} \delta_r$ . By Assumptions A.1(i) and A.3(i),  $\inf_{F^{(r)} \in \mathcal{F}} D(F^{(r)})$  and  $\mathcal{B}^{(r)}$  are positive definite a.s. Thus  $\tilde{S}_{NT}(\beta_r, F^{(r)}) \geq 0$ . In addition, if either  $\beta_r \neq \beta_r^0$  or  $F^{(r)} \neq F^{(r)0} H^{(r)}$ , then  $\tilde{S}_{NT}(\beta_r, F^{(r)}) > 0$ . Thus,  $\tilde{S}_{NT}(\beta_r, F^{(r)})$  achieves its unique minimum at the true value  $(\beta_r^0, F^{(r)0} H^{(r)})$ . Furthermore, for any  $\|\beta_r - \beta_r^0\| \geq c > 0$ , we have  $\tilde{S}_{NT}(\beta_r, F^{(r)0}) \geq \rho_{\min}^{(r)} c^2 > 0$ , where  $\rho_{\min}^{(r)} = \inf_{F^{(r)} \in \mathcal{F}} \mu_{\min}(D(F^{(r)}))$ . This implies that  $\hat{\beta}_r$  is consistent for  $\beta_r^0$ .

Note that the centered objective function satisfies  $S_{NT}(\beta_r^0, F^{(r)0} H^{(r)}) = 0$  and by definition,  $S_{NT}(\hat{\beta}_r, \hat{F}^{(r)}) \leq 0$ . It follows that

$$0 \geq S_{NT}(\hat{\beta}_r, \hat{F}^{(r)}) = \tilde{S}_{NT}(\hat{\beta}_r, \hat{F}^{(r)}) + O_P(T^{-1/2} + N^{-1/4} + h). \quad (\text{A.2})$$

It must be true that  $\tilde{S}_{NT}(\hat{\beta}_r, \hat{F}^{(r)}) = O_P(T^{-1/2} + N^{-1/4} + h)$  and  $\hat{\beta}_r - \beta_r^0 = O_P(T^{-1/4} + N^{-1/8} + h^{1/2})$ .

(ii) (A.2) also implies that  $\text{tr}[(\frac{1}{T} F^{(r)0'} M_{\hat{F}^{(r)}} F^{(r)0})(\frac{1}{N} \Lambda_r' \Lambda_r)] = O_P(T^{-1/2} + N^{-1/4} + h)$ . This, along with the fact that  $\frac{1}{N} \Lambda_r' \Lambda_r = \Sigma_{\Lambda_r} + O(N^{-1/2})$  by Assumption A.3(i), implies that

$$\frac{1}{T} F^{(r)0'} M_{\hat{F}^{(r)}} F^{(r)0} = \frac{1}{T} F^{(r)0'} F^{(r)0} - \frac{1}{T} F^{(r)0'} \hat{F}^{(r)} \frac{1}{T} \hat{F}^{(r)'} F^{(r)0} = O_P(T^{-1/2} + N^{-1/4} + h). \quad (\text{A.3})$$

By Assumption A.1(iii), we know  $\frac{1}{T} F^{0'} F^0 \xrightarrow{P} \Sigma_F > 0$ . With this, it is standard to show that  $\frac{1}{T} F^{(r)0'} F^{(r)0} = \frac{1}{T} \sum_{t=1}^T k_{tr}^* F_t^0 F_t^{0'} = \Sigma_F + o_P(1)$  under Assumption A.1. It follows that

$$\begin{aligned} T^{-1} \hat{F}^{(r)'} P_{F^{(r)0}} \hat{F}^{(r)} &= T^{-1} \hat{F}^{(r)'} F^{(r)0} (T^{-1} F^{(r)0'} F^{(r)0})^{-1} F^{(r)0'} \hat{F}^{(r)} \\ &= T^{-1} \hat{F}^{(r)'} F^{(r)0} \left[ \frac{1}{T} F^{(r)0'} \hat{F}^{(r)} \frac{1}{T} \hat{F}^{(r)'} F^{(r)0} + O_P(T^{-1/2} + N^{-1/4} + h) \right]^{-1} F^{(r)0'} \hat{F}^{(r)} \\ &= \mathbb{I}_R + O_P(T^{-1/2} + N^{-1/4} + h). \end{aligned}$$

Consequently, we have  $\|P_{\hat{F}^{(r)}} - P_{F^{(r)0}}\|^2 = \text{tr}\{(P_{\hat{F}^{(r)}} - P_{F^{(r)0}})^2\} = 2\text{tr}(\mathbb{I}_R - T^{-1} \hat{F}^{(r)'} P_{F^{(r)0}} \hat{F}^{(r)}) = O_P(T^{-1/2} + N^{-1/4} + h)$ . ■

**Proof of Theorem 3.2.** Noting that  $Y_{it}^{(r)} = X_{it}^{(r)'} \beta_r + \lambda_{ir}' F_t^{(r)} + X_{it}^{(r)'} d_0(t, r) + d_i(t, r)' F_t^{(r)} + \varepsilon_{it}^{(r)} = X_{it}^{(r)'} \beta_r + \lambda_{ir}' F_t^{(r)} + \Delta_i^{(r)}(t, r) + \varepsilon_{it}^{(r)}$ , we have

$$\begin{aligned} \hat{\beta}_r - \beta_r &= \left( \hat{D}^{(r)} \right)^{-1} \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} Y_i^{(r)} - \beta_r \\ &= \left( \hat{D}^{(r)} \right)^{-1} \frac{1}{NT} \left\{ \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} F^{(r)} \lambda_{ir} + \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_i^{(r)} + \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_i^{(r)} \right\}. \quad (\text{A.4}) \end{aligned}$$

where  $\hat{D}^{(r)} \equiv D^{(r)}(\hat{F}^{(r)}) = \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_i^{(r)}$ . By Lemma A.5,

$$\begin{aligned} \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} F^{(r)} \lambda_{ir} &= \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} a_{ji}^{(r)} (\hat{\beta}_r - \beta_r) - \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N a_{ij}^{(r)} X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \\ &\quad - \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} - \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_j^{(r)} a_{ji}^{(r)} \\ &\quad + O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-3} + h^4 + C_{NT}^{-1} h^3). \end{aligned}$$

Substituting this equation to (A.4), combining terms and multiplying by  $\sqrt{NTh}$  yields

$$\hat{D}^{(r)} \sqrt{NTh} (\hat{\beta}_r - \beta_r)$$

$$\begin{aligned}
&= \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} M_{\hat{F}^{(r)}} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} M_{\hat{F}^{(r)}} \right] \varepsilon_i^{(r)} + \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} M_{\hat{F}^{(r)}} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} M_{\hat{F}^{(r)}} \right] \Delta_i^{(r)} \\
&- \frac{\sqrt{NT\bar{h}}}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} + O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|) + o_P(1), \tag{A.5}
\end{aligned}$$

where we use the fact that  $\sqrt{NT\bar{h}}(h^4 + C_{NT}^{-3} + C_{NT}^{-1}h^3) = o(1)$  under Assumption A.4(ii). By Lemma A.7,

$$\frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} \right] M_{\hat{F}^{(r)}} \Delta_i^{(r)} = \sqrt{NT\bar{h}} B_{1\beta}^{(r)} + o_P(1). \tag{A.6}$$

Premultiplying both sides of (A.5) by  $[\hat{D}^{(r)}]^{-1}$ , using (A.6), and rearranging terms, we have

$$\begin{aligned}
\sqrt{NT\bar{h}} \left( \hat{\beta}_r - \beta_r - D(\hat{F}^{(r)})^{-1} B_{1\beta}^{(r)} \right) &= [\hat{D}^{(r)}]^{-1} \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} \right] M_{\hat{F}^{(r)}} \varepsilon_i^{(r)} \\
&+ \sqrt{\frac{N}{T\bar{h}}} [\hat{D}^{(r)}]^{-1} \zeta_{NT}^{(r)} + O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|) + o_P(1).
\end{aligned}$$

By Lemma A.6, we have

$$\begin{aligned}
&\sqrt{NT\bar{h}} \left( \hat{\beta}_r - \beta_r - [\hat{D}^{(r)}]^{-1} B_{1\beta}^{(r)} \right) \\
&= [\hat{D}^{(r)}]^{-1} \left\{ \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} \right] M_{F^{(r)}} \varepsilon_i^{(r)} + \sqrt{\frac{T\bar{h}}{N}} \xi_{NT}^{(r)} + \sqrt{\frac{N}{T\bar{h}}} \zeta_{NT}^{(r)} \right\} + o_P(1).
\end{aligned}$$

By Lemma A.8,

$$\frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} \right] M_{F^{(r)}} \varepsilon_i^{(r)} - \sqrt{\frac{N}{T\bar{h}}} B_{2\beta}^{(r)} \xrightarrow{d} N(0, \Omega_r).$$

By Lemma A.9(ii)-(iii),  $\sqrt{T\bar{h}/N}(\xi_{NT}^{(r)} - B_{3\beta}^{(r)}) = o_P(1)$  and  $\sqrt{N/(T\bar{h})}(\zeta_{NT}^{(r)} - B_{4\beta}^{(r)}) = o_P(1)$ . These results, along with the result in Lemma A.9(i) and Assumption A.2(i), implies that

$$\begin{aligned}
&\sqrt{NT\bar{h}} \left( \hat{\beta}_r - \beta_r - [D^{(r)}(F^{(r)})]^{-1} B_{1\beta}^{(r)} \right) - [D^{(r)}(F^{(r)})]^{-1} \left[ \sqrt{\frac{N}{T\bar{h}}} B_{2\beta}^{(r)} + \sqrt{\frac{T\bar{h}}{N}} B_{3\beta}^{(r)} + \sqrt{\frac{N}{T\bar{h}}} B_{4\beta}^{(r)} \right] \\
&= \sqrt{NT\bar{h}} \left( \hat{\beta}_r - \beta_r - [D^{(r)}(F^{(r)})]^{-1} [B_{1\beta}^{(r)} + \frac{1}{T\bar{h}} B_{2\beta}^{(r)} + \frac{1}{N} B_{3\beta}^{(r)} + \frac{1}{T\bar{h}} B_{4\beta}^{(r)}] \right) \\
&= [D^{(r)}(F^{(r)})]^{-1} \left\{ \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} \right] M_{F^{(r)}} \varepsilon_i^{(r)} - \sqrt{\frac{N}{T\bar{h}}} B_{2\beta}^{(r)} \right\} + o_P(1) \\
&\xrightarrow{d} N\left(0, D_0^{(r)-1} \Omega_r D_0^{(r)-1}\right).
\end{aligned}$$

This complete the proof of Theorem 3.2. ■

**Proof of Theorem 3.3.** (i) Recall that  $\hat{\delta}_r = \hat{\beta}_r - \beta_r$ . Noting that  $Y_s^{(r)} - X_s^{(r)} \hat{\beta}_r = -X_s^{(r)} \hat{\delta}_r + \Lambda_r F_s^{(r)} + \Delta_s^{(r)} + \varepsilon_s^{(r)}$ ,  $\Delta_s^{(r)} = X_s^{(r)} D_0(s, r) + D(s, r) F_s^{(r)}$ , we start with  $(NT)^{-1} (Y^{(r)} - X^{(r)} \hat{\beta}_r) (Y^{(r)} - X^{(r)} \hat{\beta}_r)' \hat{F}^{(r)} \hat{V}_{NT}^{(r)-1} = \hat{F}^{(r)}$

to obtain the following decomposition for  $\hat{F}_t^{(r)} - H^{(r)'} F_t^{(r)} - B_t^{(r)}$  as follows

$$\begin{aligned}
& \hat{F}_t^{(r)} - H^{(r)'} F_t^{(r)} - B_t^{(r)} \\
&= \hat{V}_{NT}^{(r)-1} \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} \left[ -X_s^{(r)} \hat{\delta}_r + \Lambda_r F_s^{(r)} + \Delta_s^{(r)} + \varepsilon_s^{(r)} \right]' \left[ -X_t^{(r)} \hat{\delta}_r + \Lambda_r F_t^{(r)} + \Delta_t^{(r)} + \varepsilon_t^{(r)} \right] - H^{(r)'} F_t^{(r)} - B_t^{(r)} \\
&= \hat{V}_{NT}^{(r)-1} \left\{ \frac{1}{T} \sum_{s=1}^T \hat{F}_s^{(r)} E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}) / N + \frac{1}{T} \sum_s \hat{F}_s^{(r)} \left[ \varepsilon_s^{(r)'} \varepsilon_t^{(r)} / N - E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}) / N \right] + \frac{1}{T} \sum_{s=1}^T \hat{F}_s^{(r)} F_s^{(r)'} \Lambda_r' \varepsilon_t^{(r)} / N \right. \\
&\quad + \frac{1}{T} \sum_{s=1}^T \hat{F}_s^{(r)} F_t^{(r)'} \Lambda_r' \varepsilon_s^{(r)} / N + \left[ \frac{1}{TN} \sum_{s=1}^T \hat{F}_s^{(r)} \Delta_s^{(r)'} \Lambda_r F_t^{(r)} - \hat{V}_{NT}^{(r)} B_{2t}^{(r)} \right] + \frac{1}{TN} \sum_{s=1}^T \hat{F}_s^{(r)} \Delta_s^{(r)'} \Delta_t^{(r)} \\
&\quad + \frac{1}{TN} \sum_{s=1}^T \hat{F}_s^{(r)} \Delta_s^{(r)'} \varepsilon_t^{(r)} + \frac{1}{TN} \sum_{s=1}^T \hat{F}_s^{(r)} \varepsilon_s^{(r)'} \Delta_t^{(r)} + \left[ \frac{1}{TN} \sum_{s=1}^T \hat{F}_s^{(r)} F_s^{(r)'} \Lambda_r' \Delta_t^{(r)} - \hat{V}_{NT}^{(r)} B_{1t}^{(r)} \right] \\
&\quad + \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} \hat{\delta}_r' X_s^{(r)'} X_t^{(r)} \hat{\delta}_r - \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} \hat{\delta}_r' X_s^{(r)'} \Lambda_r F_t^{(r)} - \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} \hat{\delta}_r' X_s^{(r)'} \Delta_t^{(r)} - \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} \hat{\delta}_r' X_s^{(r)'} \varepsilon_t^{(r)} \\
&\quad \left. - \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} F_s^{(r)'} \Lambda_r' X_t^{(r)} \hat{\delta}_r - \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} \Delta_s^{(r)'} X_t^{(r)} \hat{\delta}_r - \frac{1}{NT} \sum_{s=1}^T \hat{F}_s^{(r)} \varepsilon_s^{(r)'} X_t^{(r)} \hat{\delta}_r \right\} \\
&\equiv \sum_{l=1}^{16} A_l(t, r), \tag{A.7}
\end{aligned}$$

where we use the fact that  $B_t^{(r)} = B_{1t}^{(r)} + B_{2t}^{(r)}$ . The first nine terms do not depend on  $\hat{\delta}_r$  explicitly and they have the same expressions as those in Su and Wang (2017, 2020). Following Su and Wang (2017, 2020), we can also show that  $\sqrt{Nh} A_j(t, r) = o_P(1)$  for  $j = 1, 2, 4, 5, 6, 7, 8, 9$ , and that  $\sqrt{Nh} A_3(t, r)$  determines the asymptotic distribution.  $A_5(t, r)$  and  $A_9(t, r)$  are associated with the bias, which we focus on below.

For  $A_5(t, r)$ , noting that  $\hat{V}_{NT}^{(r)}$  is asymptotically nonsingular by Lemma A.2(i), it suffices to show that  $\bar{A}_5(t, r) \equiv \hat{V}_{NT}^{(r)} A_5(t, r) = o_P((Nh)^{-1/2})$ . Let  $B_{2t,1}^{(r)}$  and  $B_{2t,2}^{(r)}$  be as defined in (A.1). Let  $\varphi_{1,tr} = \hat{F}_t^{(r)} - H^{(r)'} F_t^{(r)} - B_t^{(r)}$ ,  $\varphi_{2,tr} = H^{(r)'} F_t^{(r)}$ ,  $\varphi_{3,tr} = B_t^{(r)}$ ,  $\chi_{21,t}^{(r)} = 0$ ,  $\chi_{22,t}^{(r)} = \hat{V}_{NT}^{(r)} B_{2t,1}^{(r)}$ , and  $\chi_{23,t}^{(r)} = \hat{V}_{NT}^{(r)} B_{2t,2}^{(r)}$ . For  $\bar{A}_5(t, r)$ , we make the following decomposition

$$\bar{A}_5(t, r) = \frac{1}{TN} \sum_{s=1}^T \hat{F}_s^{(r)} \Delta_s^{(r)'} \Lambda_r F_t^{(r)} - \hat{V}_{NT}^{(r)} B_{2t}^{(r)} = \sum_{l=1}^3 \left[ \frac{1}{TN} \sum_s \varphi_{l,sr} \Delta_s^{(r)'} \Lambda_r F_t^{(r)} - \chi_{2l,t}^{(r)} \right] \equiv \sum_{l=1}^3 \bar{A}_{5,l}(t, r).$$

Following the proof of Lemma A.3 in Su and Wang (2020) and Theorem 3.2, we can show that  $\sqrt{Nh} \bar{A}_{5,1}(t, r) = \sqrt{Nh} O_P(C_{NT}^{-2} + \|\hat{\delta}_r\|) = o_P(1)$  as Theorem 3.2 implies that  $\hat{\delta}_r = O_P(N^{-1} + (Th)^{-1}) = o_P((Nh)^{-1/2})$ . For  $\bar{A}_{5,2}(t, r)$ , noting that  $[H^{(r)'}]^{-1} \hat{V}_{NT}^{(r)} B_{2t,1}^{(r)} = N^{-1} E(F_t A_{2,tr}') \Lambda_r F_t k_{h,tr}^{*1/2} h^2 \kappa_2 = N^{-1} E(F_t A_{2,tr}') \Lambda_r F_t h^2 \kappa_2$ , we have

$$\begin{aligned}
\bar{A}_{5,2}(t, r) &= \frac{1}{TN} \sum_{s=1}^T H^{(r)'} F_s^{(r)} \Delta_s^{(r)'} \Lambda_r F_t^{(r)} - \hat{V}_{NT}^{(r)} B_{2t,1}^{(r)} \\
&= H^{(r)'} \frac{1}{TN} \sum_{s=1}^T \sum_{i=1}^N k_{h,sr}^* F_s A_{1,istr} \left( \frac{s-r}{T} \right) \lambda_{ir}' F_t^{(r)} \\
&\quad + H^{(r)'} \left[ \frac{1}{TN} \sum_{s=1}^T \sum_{i=1}^N k_{h,sr}^* F_s A_{2,istr} \left( \frac{s-r}{T} \right)^2 \lambda_{ir} - N^{-1} E(F_t A_{2,tr}') \Lambda_r h^2 \kappa_2 \right] F_t^{(r)} + O_P(h^3) \\
&\equiv H^{(r)'} \bar{A}_{5,21}(t, r) F_t^{(r)} + H^{(r)'} \bar{A}_{5,22}(t, r) F_t^{(r)} + O_P(h^3).
\end{aligned}$$

For  $\bar{A}_{5,21}(t, r)$ , we have

$$\begin{aligned}\bar{A}_{5,21}(t, r) &= \frac{1}{TN} \sum_{s=1}^T \sum_{i=1}^N k_{h, sr}^* F_s X'_{is} \beta_r^{(1)} \lambda'_{ir} \frac{s-r}{T} + \frac{1}{TN} \sum_{s=1}^T \sum_{i=1}^N k_{h, sr}^* F_s F'_s \lambda_{ir}^{(1)} \lambda'_{ir} \frac{s-r}{T} \\ &= \frac{1}{TN} \sum_{s=1}^T \sum_{i=1}^N k_{h, sr}^* F_s X'_{is} \beta_r^{(1)} \lambda'_{ir} \frac{s-r}{T} + \frac{1}{T} \sum_{s=1}^T k_{h, sr}^* F_s F'_s \frac{s-r}{T} (N^{-1} \Lambda_i^{(1)'} \Lambda_i) \equiv \sum_{\ell=1}^2 \bar{A}_{5,21\ell}(t, r).\end{aligned}$$

As in Su and Wang (2020),  $\bar{A}_{5,212}(t, r) = O_P((Th)^{-1/2} h + (Th)^{-1}) = o_P((Nh)^{-1/2})$  by the fact that  $\frac{1}{T} \sum_s k_{h, sr}^* F_s F'_s (\frac{s-r}{T}) = O_P((Th)^{-1/2} h) + O((Th)^{-1})$ . For  $\bar{A}_{5,211}(t, r)$ , we have

$$\begin{aligned}\bar{A}_{5,211}(t, r) &= \frac{1}{NT} \sum_{s=1}^T \sum_{i=1}^N \frac{s-r}{T} k_{h, sr}^* [F_s X'_{is} - E(F_s X'_{is})] \beta_r^{(1)} \lambda'_{ir} + \frac{1}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s=1}^T \frac{s-r}{T} k_{h, sr}^* E(F_s X'_{is}) \right] \beta_r^{(1)} \lambda'_{ir} \\ &= O_P((NT h)^{-1/2} h) + O((Th)^{-1}) = o_P((Nh)^{-1/2}),\end{aligned}$$

where we use the fact that  $\frac{1}{T} \sum_{s=1}^T \frac{s-r}{T} k_{h, sr}^* E(F_s X'_{is}) = \frac{1}{T} \sum_{s=1}^T \frac{s-r}{T} k_{h, sr}^* \Sigma_{FX} = O((Th)^{-1})$  by the property of Riemann integral for interior point  $r$ . Then  $\bar{A}_{5,21}(t, r) = o_P((Nh)^{-1/2})$ . Note that

$$\begin{aligned}& \frac{1}{TN} \sum_{s=1}^T \sum_{i=1}^N k_{h, sr}^* F_s A_{2, isr} (\frac{s-r}{T})^2 \lambda_{ir} - N^{-1} E(F_t A'_{2, tr}) \Lambda_r F_t^{(r)} h^2 \kappa_2 \\ &= \frac{1}{2TN} \sum_{s=1}^T \sum_{i=1}^N k_{h, sr}^* F_s [X'_{is} \beta_r^{(2)} + F'_s \lambda_{ir}^{(2)}] (\frac{s-r}{T})^2 \lambda_{ir} - \frac{1}{2N} E [F_t (\beta_r^{(2)} X'_t + F'_t \Lambda_r^{(2)'})] \Lambda_r h^2 \kappa_2 \\ &= \frac{1}{2N} \sum_{s=1}^T (\frac{s-r}{T})^2 k_{h, sr}^* F_s \beta_r^{(2)} \frac{X'_s \Lambda_r}{N} - \frac{1}{2N} E [F_t \beta_r^{(2)} \frac{X'_t \Lambda_r}{N}] h^2 \kappa_2 \\ &+ \frac{1}{2} \left[ \frac{1}{T} \sum_{s=1}^T (\frac{s-r}{T})^2 k_{h, sr}^* F_s F'_s - E [F_t F'_t] h^2 \kappa_2 \right] \left( \frac{1}{N} \Lambda_r^{(2)'} \Lambda_r \right).\end{aligned}$$

It is easy to show that either term on the right hand side of the above equation is  $O_P((Th)^{-1/2} h^2 + (Th)^{-1} + h^3) = o_P((Nh)^{-1/2})$ . Thus  $\bar{A}_{5,22}(t, r) = o_P((Nh)^{-1/2})$  and  $\bar{A}_{5,2}(t, r) = o_P((Nh)^{-1/2})$ .

Note that  $\Delta_s^{(r)} = X_s^{(r)} d_0(s, r) + D(s, r) F_s^{(r)} = (\Delta_{1s}^{(r)}, \dots, \Delta_{Ns}^{(r)})'$ , and

$$\hat{V}_{NT}^{(r)} B_{2t,2}^{(r)} = \frac{1}{TN} \sum_s (\frac{s-r}{T})^2 k_{h, sr}^* E[\bar{C}_{1s}^{(r)} A'_{1, sr}] \Lambda_r F_t^{(r)} \equiv \alpha_1^{(r)} F_t^{(r)},$$

It follows that

$$\bar{A}_{5,3}(t, r) = \frac{1}{TN} \sum_s B_s^{(r)} \Delta_s^{(r)'} \Lambda_r F_t^{(r)} - \hat{V}_{NT}^{(r)} B_{2t,2}^{(r)} = \left[ \frac{1}{TN} \sum_s \sum_i B_s^{(r)} \Delta_{is}^{(r)} \lambda'_{ir} - \alpha_1^{(r)} \right] F_t^{(r)} \equiv \bar{A}_{5,3}(r) F_t^{(r)}.$$

For  $\bar{A}_{5,3}(r)$ , we have

$$\begin{aligned}\bar{A}_{5,31}(r) &= \frac{1}{TN} \sum_s \sum_i B_s^{(r)} \Delta_{is}^{(r)} \lambda'_{ir} - \alpha_1^{(r)} = \frac{1}{TN} \sum_s \sum_i (\frac{s-r}{T})^2 k_{h, sr}^* C_{1s}^{(r)} A_{1, isr} \lambda'_{ir} - \alpha_{1,1}^{(r)} + O_P(h^3) \\ &= \frac{1}{TN} \sum_s \sum_i (\frac{s-r}{T})^2 k_{h, sr}^* \bar{C}_{1s}^{(r)} A_{1, isr} \lambda'_{ir} - \alpha_{1,1}^{(r)} + O_P(h^3) \\ &= \frac{1}{TN} \sum_s (\frac{s-r}{T})^2 k_{h, sr}^* (\bar{C}_{1s}^{(r)} A'_{1, sr} - E[\bar{C}_{1s}^{(r)} A'_{1, sr}]) \Lambda_r + O_P(h^3)\end{aligned}$$



$$= O_P((N/h)^{-1/2}h^2) + O_P(h^3) = o_P(Nh)^{-1/2}.$$

In sum, we have shown that  $\sqrt{Nh}\bar{A}_5(t, r) = o_P(1)$ .

For  $A_9(t, r)$ , we want to show that  $\bar{A}_9(t, r) \equiv \hat{V}_{NT}^{(r)} A_9(t, r) = o_P(Nh)^{-1/2}$ . Note that

$$\bar{A}_9(t, r) = \frac{1}{TN} \sum_{s=1}^T \hat{F}_s^{(r)} F_s^{(r)'} \Lambda_r' \Delta_t^{(r)} - \hat{V}_{NT}^{(r)} B_{1t}^{(r)} = \sum_{l=1}^3 \left[ \frac{1}{TN} \sum_s \varphi_{l, sr} \Delta_s^{(r)'} \Lambda_r F_t^{(r)} - \chi_{1l, t}^{(r)} \right] \equiv \sum_{l=1}^3 \bar{A}_{9, l}(t, r),$$

where  $\chi_{11, t} = 0$ ,  $\chi_{12, t} = \hat{V}_{NT}^{(r)} B_{1t}^{(r)}$  and  $\chi_{13, t} = 0$ . As in Su and Wang (2020), we can readily show that

$$\begin{aligned} \sqrt{Nh}\bar{A}_{9,1}(t, r) &= \sqrt{Nh}O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})O_P(h^{1/2}) = o_P(1) \text{ and} \\ \sqrt{Nh}\bar{A}_{9,3}(t, r) &= \sqrt{Nh}O_P((Th)^{-1/2}h^{3/2} + (Th)^{-1}h^{1/2} + h^{5/2}) = o_P(1) \end{aligned}$$

under the conditions that  $Nh/T \rightarrow 0$  and  $Nh^6 \rightarrow 0$  in Assumption A.4(ii). Noting that  $\hat{V}_{NT}^{(r)} B_{1t}^{(r)} = \hat{V}_{NT}^{(r)} [k_{h, tr}^{*1/2} C_{1t}^{(r)} \frac{t-r}{T} + k_{h, tr}^{*1/2} C_{3t}^{(r)} (\frac{t-r}{T})^2] = \alpha_{2,1} + \alpha_{2,2}$  with  $\alpha_{2,l} = N^{-1} k_{h, tr}^{*1/2} H^{(r)'} \Sigma_F \Lambda_r' A_{l, tr} (\frac{t-r}{T})^l$  for  $l = 1, 2$ , we have

$$\begin{aligned} \bar{A}_{9,2}(t, r) &= \frac{1}{TN} \sum_{s=1}^T H^{(r)'} F_s^{(r)} F_s^{(r)'} \Lambda_r' \Delta_t^{(r)} - \hat{V}_{NT}^{(r)} B_{1t}^{(r)} = H^{(r)'} \frac{1}{TN} \sum_{s=1}^T k_{h, sr}^* F_s F_s' \Lambda_r' \Delta_t^{(r)} - (\alpha_{2,1} + \alpha_{2,2}) \\ &= H^{(r)'} \frac{1}{N} \Sigma_F \Lambda_r' \Delta_t^{(r)} - (\alpha_{2,1} + \alpha_{2,2}) + O_P((Th)^{-1/2}h) \\ &= H^{(r)'} \Sigma_F \left[ \frac{1}{N} \sum_{i=1}^N \lambda_{ir} \left( X_{it}^{(1)'} \beta_r^{(1)} + \lambda_{ir}^{(1)'} F_t \right) - \frac{1}{N} \Lambda_r' A_{1, tr} \right] \frac{t-r}{T} k_{h, tr}^{*1/2} \\ &\quad + \frac{1}{2} H^{(r)'} \Sigma_F \left[ \frac{1}{N} \sum_{i=1}^N \lambda_{ir} \left( X_{it}^{(2)'} \beta_r^{(2)} + \lambda_{ir}^{(2)'} F_t \right) - \frac{1}{N} \Lambda_r' A_{2, tr} \right] \left( \frac{t-r}{T} \right)^2 k_{h, tr}^{*1/2} + O_P((Th)^{-1/2}h + h^3) \\ &= O_P((Th)^{-1/2}h + h^3) = o_P(Nh)^{-1/2}. \end{aligned}$$

Then  $\sqrt{Nh}A_9(t, r) = o_P(1)$ .

In summary, the bias term  $B_t^{(r)}$  is

$$\begin{aligned} B_t^{(r)} &= B_{1t}^{(r)} + B_{2t,1}^{(r)} + B_{2t,2}^{(r)} \\ &= \left[ k_{h, tr}^{*1/2} C_{1t}^{(r)} \frac{t-r}{T} + k_{h, tr}^{*1/2} C_{3t}^{(r)} \left( \frac{t-r}{T} \right)^2 \right] + \hat{V}_{NT}^{(r)-1} H^{(r)'} N^{-1} E(F_t A_{2, tr}') \Lambda_r F_t^{(r)} h^2 \kappa_2 \\ &\quad + \hat{V}_{NT}^{(r)-1} \frac{1}{TN} \sum_s \left( \frac{s-r}{T} \right)^2 k_{h, sr}^* E[\bar{C}_{1s}^{(r)} A_{1, sr}'] \Lambda_r F_t^{(r)} \\ &= k_{h, tr}^{*1/2} \left[ C_{1t}^{(r)} \frac{t-r}{T} + C_{2t}^{(r)} \left( \frac{t-r}{T} \right)^2 + h^2 \kappa_2 C_{3t}^{(r)} + h^2 \kappa_2 C_{4t}^{(r)} \right], \end{aligned}$$

where  $C_{lt}^{(r)}$ ,  $l \in [4]$ , are as defined in (3.1).

The terms  $A_{10}(t, r)$  to  $A_{17}(t, r)$  are either  $O_P(\|\hat{\delta}_r\|)$  or  $o_P(\|\hat{\delta}_r\|)$ . This, in conjunction with the fact that  $\hat{\delta}_r = O_P(N^{-1} + (Th)^{-1})$  by Theorem 3.2, implies that  $\sqrt{Nh} \sum_{i=10}^{17} A_i(t, r) = o_P(1)$ . In addition, Su and Wang (2017, 2020) have shown that  $K_r^* (\frac{t-r}{Th})^{-1/2} \sqrt{Nh} A_3(t, r) \xrightarrow{d} N(0, V_r^{-1} Q_r \Gamma_{rt} Q_r' V_r^{-1})$ . Combining these results yields the desired result in Theorem 3.3(i).

(ii) Let  $\Upsilon^{(r)} = \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)}$ . Recall that  $\Delta_i^{(r)} = (\Delta_{i,1}^{(r)}, \dots, \Delta_{i,T}^{(r)})'$  with  $\Delta_{i,t}^{(r)} = k_{h, tr}^{*1/2} \Delta_i(t, r)$ . Noting

that  $\hat{\lambda}_{ir} = \frac{1}{T} \hat{F}^{(r)'} (Y_i^{(r)} - X_i^{(r)} \hat{\beta}_r)$  and  $Y_i^{(r)} - X_i^{(r)} \hat{\beta}_r = -X_i^{(r)} \hat{\delta}_r + F^{(r)} \lambda_{ir} + \Delta_i^{(r)} + \varepsilon_i^{(r)}$ , we have

$$\begin{aligned} \hat{\lambda}_{ir} - H^{(r)-1} \lambda_{ir} &= \frac{1}{T} \hat{F}^{(r)'} \left[ -X_i^{(r)} \hat{\delta}_r + F^{(r)} \lambda_{ir} + \Delta_i^{(r)} + \varepsilon_i^{(r)} \right] - H^{(r)-1} \lambda_{ir} \\ &= \frac{1}{T} H^{(r)'} F^{(r)'} \varepsilon_i^{(r)} + \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} - \frac{1}{T} \hat{F}^{(r)'} \Upsilon^{(r)} H^{(r)-1} \lambda_{ir} + \frac{1}{T} \Upsilon^{(r)'} \Delta_i^{(r)} \\ &\quad + \frac{1}{T} H^{(r)'} F^{(r)'} \Delta_i^{(r)} + \frac{1}{T} B^{(r)'} \varepsilon_i^{(r)} - \frac{1}{T} \hat{F}^{(r)'} B^{(r)} H^{(r)-1} \lambda_{ir} + \frac{1}{T} B^{(r)'} \Delta_i^{(r)} \\ &\quad - \frac{1}{T} \Upsilon^{(r)'} X_i^{(r)} \hat{\delta}_r - \frac{1}{T} H^{(r)'} F^{(r)'} X_i^{(r)} \hat{\delta}_r - \frac{1}{T} B^{(r)'} X_i^{(r)} \hat{\delta}_r \equiv \sum_{l=1}^{11} D_l(i, r), \end{aligned} \quad (\text{A.8})$$

Su and Wang (2017, 2020) have studied the terms  $D_1(i, r)$  to  $D_8(i, r)$ . The only difference for these terms is that

$$\begin{aligned} \Delta_{i,t}^{(r)} &= k_{h, tr}^{*1/2} \left\{ X_{it}' [\beta(\frac{t}{T}) - \beta(\frac{r}{T})] + F_t' [\lambda_i(\frac{t}{T}) - \lambda_i(\frac{r}{T})] \right\} \\ &= k_{h, tr}^{*1/2} \left\{ [X_{it}' \beta_r^{(1)} + F_t' \lambda_{ir}^{(1)}] \frac{t-r}{T} + \frac{1}{2} [X_{it}' \beta_r^{(2)} + F_t' \lambda_{ir}^{(2)}] \left( \frac{t-r}{T} \right)^2 + O_P \left( \left( \frac{t-r}{T} \right)^3 \right) \right\} \end{aligned}$$

in this paper, instead of  $\Delta_{i,t}^{(r)} = k_{h, tr}^{*1/2} F_t' [\lambda_i(\frac{t}{T}) - \lambda_i(\frac{r}{T})]$  in Su and Wang (2020). Following the analyses in Su and Wang (2017, 2020), we can show that  $\sqrt{Th} D_l(i, r) = o_P(1)$  for  $l = 2, 3, 4, 6$ , while  $D_5(i, r)$ ,  $D_7(i, r)$  and  $D_8(i, r)$  contribute to the asymptotic bias. Following Su and Wang (2017, 2020), we can show that

$$D_5(i, r) = \frac{1}{T} H^{(r)'} F^{(r)'} \Delta_i^{(r)} = h^2 \kappa_2 H^{(r)'} E(F_t A_{2,itr}) + O_P((Th)^{-1/2} h + h^3).$$

For  $D_7(i, r)$ , we have

$$-D_7(i, r) = \frac{1}{T} \sum_{t=1}^T \Upsilon_t^{(r)'} B_t^{(r)'} H^{(r)-1} \lambda_{ir} + \frac{1}{T} \sum_{t=1}^T H^{(r)'} F_t^{(r)} B_t^{(r)'} H^{(r)-1} \lambda_{ir} + \frac{1}{T} \sum_{t=1}^T B_t^{(r)} B_t^{(r)'} H^{(r)-1} \lambda_{ir} = \sum_{l=1}^3 D_{7,l}(i, r),$$

where  $\Upsilon_t^{(r)} = \hat{F}_t^{(r)} - H^{(r)'} F_t^{(r)} - B_t^{(r)}$ . Note that when  $r \in [[Th], T - [Th]]$ , we have

$$\begin{aligned} \|D_{7,1}(i, r)\| &\lesssim \left\{ \frac{1}{T} \sum_{t=1}^T \|\Upsilon_t^{(r)}\|^2 \right\}^{1/2} \left\{ \frac{1}{T} \sum_{t=1}^T \|B_t^{(r)}\|^2 \right\}^{1/2} = O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|) O_P(h), \\ D_{7,2}(i, r) &= \frac{1}{T} \sum_{t=1}^T k_{h, tr}^* H^{(r)'} F_t \left[ C_{1t}^{(r)} \frac{t-r}{T} + C_{2t}^{(r)} \left( \frac{t-r}{T} \right)^2 + (C_{3t}^{(r)} + C_{4t}^{(r)}) \kappa_2 h^2 \right]' H^{(r)-1} \lambda_{ir} \\ &= \kappa_2 h^2 H^{(r)'} E \left[ F_t (\bar{C}_{2t}^{(r)} + \bar{C}_{3t}^{(r)} + \bar{C}_{4t}^{(r)})' \right] H^{(r)-1} \lambda_{ir} + O_P((Th)^{-1/2} h), \end{aligned}$$

$$\begin{aligned} D_{7,3} &= \frac{1}{T} \sum_{t=1}^T \left[ C_{1t}^{(r)} \frac{t-r}{T} + C_{2t}^{(r)} \kappa_2 h^2 + C_{3t}^{(r)} \left( \frac{t-r}{T} \right)^2 \right] \left[ C_{1t}^{(r)} \frac{t-r}{T} + C_{2t}^{(r)} \kappa_2 h^2 + C_{3t}^{(r)} \left( \frac{t-r}{T} \right)^2 \right]' H^{(r)-1} \lambda_{ir} \\ &= \kappa_2 h^2 E(\bar{C}_{1t}^{(r)} \bar{C}_{1t}^{(r)'}) H^{(r)-1} \lambda_{ir} + O_P((Th)^{-1/2} h + h^3). \end{aligned}$$

and thus  $D_7(i, r) = \kappa_2 h^2 \left\{ H^{(r)'} E[F_t (\bar{C}_{2t}^{(r)} + \bar{C}_{3t}^{(r)} + \bar{C}_{4t}^{(r)})'] + E(\bar{C}_{1t}^{(r)} \bar{C}_{1t}^{(r)'}) \right\} H^{(r)-1} \lambda_{ir} + O_P(T^{-1/2} h^{3/2} + h^3)$ .

When  $r \in [1, [Th]] \cup (T - [Th], T]$ , it is easy to show that the leading bias term is contributed by  $\frac{1}{T} \sum_{t=1}^T k_{h, tr}^* H^{(r)'} E(F_t \bar{C}_{1t}^{(r)'}) \frac{t-r}{T} + O(h)$ .

Similarly,

$$\begin{aligned} D_8(i, r) &= \frac{1}{T} \sum_{t=1}^T \left[ C_{1t}^{(r)} \frac{t-r}{T} + C_{2t}^{(r)} \left( \frac{t-r}{T} \right)^2 + (C_{3t}^{(r)} + C_{4t}^{(r)}) \kappa_2 h^2 \right] \left[ X'_{it} \beta_r^{(1)} + \lambda_{ir}^{(1)'} F_t \right] k_{h, tr}^* \frac{t-r}{T} + O_P(h^3) \\ &= \kappa_2 h^2 E(\bar{C}_{1t}^{(r)} A_{1, itr}) + O_P((Th)^{-1/2} h^2 + h^3). \end{aligned}$$

In addition, noting that  $D_l(i, r) = o_P(\|\hat{\delta}_r\|)$  for  $l = 9, 11$  and  $D_{10}(i, r) = O_P(\|\hat{\delta}_r\|)$ , we have  $\sqrt{Th} \sum_{l=9}^{11} D_l(i, r) = O_P(\|\hat{\delta}_r\|) = o_P(1)$ . In sum, we have shown that

$$\begin{aligned} \sum_{l=2}^{11} D_l(i, r) &= \kappa_2 h^2 \left[ E(\bar{C}_{1t}^{(r)} A_{1, itr}) + H^{(r)'} E(F_t A_{2, itr}) \right] \\ &\quad - \kappa_2 h^2 \left\{ H^{(r)'} E \left[ F_t (\bar{C}_{2t}^{(r)} + \bar{C}_{3t}^{(r)} + \bar{C}_{4t}^{(r)})' \right] + E[\bar{C}_{1t}^{(r)} \bar{C}_{1t}^{(r)'}] \right\} H^{(r)-1} \lambda_{ir} + O_P(T^{-1/2} + h^3). \end{aligned}$$

Finally, by Assumption A.6(ii), we know  $\frac{\sqrt{h}}{\sqrt{T}} \sum_{s=1}^T k_{h, sr} F_s \varepsilon_{is} \xrightarrow{d} N(0, \Omega_{i,r})$ . It follows that  $\sqrt{Th} D_1(i, r) = \frac{1}{\sqrt{Th}} H^{(r)'} \sum_{s=1}^T K_r^* \left( \frac{s-r}{Th} \right) F_s \varepsilon_{is} \xrightarrow{d} N(0, (Q_r^{-1})' \Omega_{i,r} (Q_r^{-1}))$ . This completes the proof of Theorem 3.3(ii).

(iii) Noting that  $Y_{it} - X'_{it} \hat{\beta}_t = -X'_{it} \hat{\delta}_t - [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}]' H^{(t)'} F_t + \hat{\lambda}'_{it} H^{(t)'} F_t + \varepsilon_{it}$ , we have

$$\begin{aligned} \hat{F}_t - H^{(t)'} F_t &= \hat{S}_{\lambda, t}^{-1} \frac{1}{N} \sum_{i=1}^N \hat{\lambda}_{it} (Y_{it} - X'_{it} \hat{\beta}_t) - H^{(t)'} F_t \\ &= \hat{S}_{\lambda, t}^{-1} H^{(t)-1} \frac{1}{N} \sum_{i=1}^N \lambda_{it} \varepsilon_{it} + \hat{S}_{\lambda, t}^{-1} \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}) \varepsilon_{it} \\ &\quad - \hat{S}_{\lambda, t}^{-1} \frac{1}{N} \sum_{i=1}^N \hat{\lambda}_{it} [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}]' H^{(t)'} F_t - \hat{S}_{\lambda, t}^{-1} \frac{1}{N} \sum_{i=1}^N \hat{\lambda}_{it} X'_{it} \hat{\delta}_t \equiv \sum_{\ell=1}^4 A_\ell(t), \end{aligned} \quad (\text{A.9})$$

where  $\hat{S}_{\lambda, t} = \frac{1}{N} \sum_{i=1}^N \hat{\lambda}_{it} \hat{\lambda}'_{it}$ . By Lemma A.5(i)-(ii) in Su and Wang (2020),

$$\hat{S}_{\lambda, t} = \frac{1}{N} \sum_{i=1}^N \hat{\lambda}_{it} \hat{\lambda}'_{it} = Q_t \Sigma_{\Lambda_t} Q_t' + o_P(1) \text{ and } \frac{1}{\sqrt{N}} \sum_{i=1}^N (\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}) \varepsilon_{it} = o_P(1).$$

Then  $\sqrt{N} A_2(t) = o_P(1)$ . For  $A_4(t)$ , we have  $\left\| \sqrt{N} A_4(t) \right\| \leq \left\| \frac{1}{N} \sum_{i=1}^N \hat{\lambda}_{it} X'_{it} \right\| \sqrt{N} \|\hat{\delta}_t\| = O_P(1) \sqrt{N} O_P(N^{-1} + (Th)^{-1} + h^2)$  by Theorem 3.2. Noting that  $NTh^5 = O(1)$  and  $1/(Th) = o(1)$  implies that  $Nh^4 = o(1)$ ,  $\left\| \sqrt{N} A_4(t) \right\| = o_P(1)$  under Assumption A.4(ii) and the additional condition  $NTh^5 = O(1)$ .

For  $A_1(t)$ , we have by Assumption A.6(ii),

$$\begin{aligned} \sqrt{N} A_1(t) &= \hat{S}_{\lambda, t}^{-1} H^{(t)-1} \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \varepsilon_{it} = (Q_t \Sigma_{\Lambda_t} Q_t')^{-1} Q_t \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \varepsilon_{it} + o_P(1) \\ &\xrightarrow{d} N \left( 0, (Q_t \Sigma_{\Lambda_t} Q_t')^{-1} Q_t \Gamma_{tt} Q_t' (Q_t \Sigma_{\Lambda_t} Q_t')^{-1} \right) = N \left( 0, (\Sigma_{\Lambda_t}^{-1} Q_t^{-1})' \Gamma_{tt} \Sigma_{\Lambda_t}^{-1} Q_t^{-1} \right). \end{aligned}$$

Last, we show that  $A_3(t)$  contributes to the asymptotic bias. To this end, we make the following decomposition

$$-\sqrt{N} A_3(t) = \frac{1}{\sqrt{N}} \sum_{i=1}^N [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}] [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}]' H^{(t)'} F_t + \frac{1}{\sqrt{N}} \sum_{i=1}^N H^{(t)-1} \lambda_{it} [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}]' H^{(t)'} F_t$$

$$\equiv A_{3,1}(t) + A_{3,2}(t).$$

Noting that  $\|\hat{\lambda}_{it} - H^{(t)-1}\lambda_{it}\| = O_P((Th)^{-1/2} + h^2)$  by Theorem 3.3(ii), we have:

$$\|A_{3,1}(t)\| \lesssim \frac{1}{\sqrt{N}} \sum_{i=1}^N \|\hat{\lambda}_{it} - H^{(t)-1}\lambda_{it}\|^2 = O_P(N^{1/2}((Th)^{-1} + h^4)) = o_P(1).$$

Next,  $A_{3,2}(t) = \sum_{l=1}^{11} \frac{1}{\sqrt{N}} \sum_{i=1}^N H^{(t)-1}\lambda_{it} D_l(i, r)' H^{(t)'} F_t \equiv \sum_{l=1}^{11} A_{3,2l}(t)$  by (A.8). Su and Wang (2020) have studied the terms  $A_{3,21}(t)$  to  $A_{3,28}(t)$ . Following their analysis, we can show that  $A_{3,2l}(t) = o_P(1)$  for  $l = 1, 2, 3, 4, 6$ . Now, we consider the terms  $A_{3,25}(t)$ ,  $A_{3,27}(t)$ , and  $A_{3,28}(t)$ . First,

$$\begin{aligned} A_{3,25}(t) &= H^{(t)-1} \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \frac{1}{T} \Delta_i^{(t)'} F^{(t)} H^{(t)} H^{(t)'} F_t \\ &= H^{(t)-1} \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \left\{ \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \left( X'_{is} \beta_t^{(1)} + \lambda_{it}^{(1)'} F_s \right) F'_s \frac{s-t}{T} \right\} H^{(t)} H^{(t)'} F_t \\ &\quad + \frac{1}{2\sqrt{N}} \sum_{i=1}^N H^{(t)-1} \lambda_{it} \left\{ \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \left( X'_{is} \beta_t^{(2)} + \lambda_{it}^{(2)'} F_s \right) F'_s \left( \frac{s-t}{T} \right)^2 \right\} H^{(t)} H^{(t)'} F_t + O_P(N^{1/2}h^3) \\ &= \frac{h^2}{2} \sqrt{N} H^{(t)-1} \left\{ \frac{1}{N} \sum_{i=1}^N \lambda_{it} \beta_t^{(2)'} E(X_{is} F'_s) + \frac{\Lambda'_t \Lambda_t^{(2)}}{N} \Sigma_F \right\} \Sigma_F^{-1} \kappa_2 F_t + o_P(1), \end{aligned}$$

where we use the fact that  $\frac{1}{T} \sum_{s=1}^T k_{h,st}^* F'_s [X'_{is} \beta_t^{(1)} + \lambda_{it}^{(1)'} F_s] \frac{s-t}{T} = O_P((Th)^{-1/2} h) = o_P(N^{-1/2})$  and  $H^{(t)} H^{(t)'} = \Sigma_F^{-1} + O_P(C_{NT}^{-1})$  by Lemma A.2(iv) and Theorem 3.2. For  $A_{3,27}(t)$ , we have

$$\begin{aligned} A_{3,27}(t) &= \frac{1}{\sqrt{NT}} \sum_{i=1}^N \sum_{s=1}^T H^{(t)-1} \lambda_{it} \lambda'_{it} (H^{(t)-1})' B_s^{(t)} \hat{F}_s^{(t)'} H^{(t)'} F_t \\ &= \sum_{l=1}^3 H^{(t)-1} \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \lambda'_{it} \right) H^{(t)-1'} \left( \frac{1}{T} \sum_{s=1}^T B_s^{(t)} \phi'_{l,st} \right) H^{(t)'} F_t \equiv \sum_{l=1}^3 A_{3,27l}(t). \end{aligned}$$

For  $A_{3,271}(t)$ , we have

$$\|A_{3,271}(t)\| \lesssim \left\| \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \lambda'_{it} \right\| \left\| \frac{1}{T} \sum_{s=1}^T B_s^{(t)} \left[ \hat{F}_s^{(t)} - H^{(t)'} F_s^{(t)} - B_s^{(t)} \right]' \right\| = N^{1/2} O_P((C_{NT}^{-1} + \|\hat{\delta}_r\|)h) = o_P(1).$$

For  $A_{3,272}(t)$  and  $A_{3,273}(t)$ , we use  $B_s^{(t)} = C_{1s}^{(t)} \frac{s-t}{T} + C_{2s}^{(t)} \left( \frac{s-t}{T} \right)^2 + (C_{3s}^{(t)} + C_{4s}^{(t)}) h^2 \kappa_2$  to obtain

$$\begin{aligned} A_{3,272}(t) &= \frac{1}{\sqrt{NT}} \sum_{i=1}^N \sum_{s=1}^T H^{(t)-1} \lambda_{it} \lambda'_{it} H^{(t)-1'} B_s^{(t)} F_s^{(t)'} H^{(t)} H^{(t)'} F_t \\ &= \sqrt{N} H^{(t)-1} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right) H^{(t)-1'} \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \left[ C_{1s}^{(t)} \frac{s-t}{T} + C_{2s}^{(t)} \left( \frac{s-t}{T} \right)^2 + (C_{3s}^{(t)} + C_{4s}^{(t)}) h^2 \kappa_2 \right] F'_s H^{(t)} H^{(t)'} F_t \\ &= \sqrt{N} h^2 H^{(t)-1} \Sigma_{\Lambda_t} H^{(t)-1'} E[(\bar{C}_{2s}^{(t)} + \bar{C}_{3s}^{(t)} + \bar{C}_{4s}^{(t)}) F'_s] \Sigma_F^{-1} \kappa_2 F_t + o_P(1), \end{aligned}$$

and

$$\begin{aligned}
A_{3,273}(t) &= \frac{1}{\sqrt{NT}} \sum_{i=1}^N \sum_{s=1}^T H^{(t)-1} \lambda_{it} \lambda'_{it} H^{(t)-1'} B_s^{(t)} B_s^{(t)'} H^{(t)'} F_t \\
&= \sqrt{N} H^{(t)-1} \left[ \frac{1}{N} \sum_{i=1}^N \lambda_{it} \lambda'_{it} \right] H^{(t)-1'} \left[ \frac{1}{T} \sum_{s=1}^T k_{h,st}^* C_{1s}^{(t)} C_{1s}^{(t)'} \left( \frac{s-t}{T} \right)^2 \right] H^{(t)'} F_t + o_P(N^{1/2} h^3) \\
&= \sqrt{N} h^2 \kappa_2 H^{(t)-1} \Sigma_{\Lambda_t} H^{(t)-1'} E[\bar{C}_{1s}^{(t)} \bar{C}_{1s}^{(t)'}] H^{(t)'} F_t + o_P(1),
\end{aligned}$$

where  $\bar{C}_{ls}^{(t)}$  is defined analogously to  $C_{ls}^{(t)}$  with  $V_{NT}^{(t)}$  and  $H^{(t)}$  replaced by their probability limits for  $l = 1, 2, 3$ . It follows that  $A_{3,27}(t) = \sqrt{N} h^2 \kappa_2 H^{(t)-1} \Sigma_{\Lambda_t} H^{(t)-1'} \{E[(\bar{C}_{2s}^{(t)} + \bar{C}_{3s}^{(t)} + \bar{C}_{4s}^{(t)}) F_s'] + E(\bar{C}_{1s}^{(t)} \bar{C}_{1s}^{(t)'}) H^{(t)'}\} F_t + o_P(1)$ . For  $A_{3,28}(t)$ , we have

$$\begin{aligned}
A_{3,28}(t) &= H^{(t)-1} \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \frac{1}{T} \Delta_i(t)' B^{(t)} H^{(t)'} F_t \\
&= \sqrt{N} H^{(t)-1} \frac{1}{N} \sum_{i=1}^N \lambda_{it} \lambda_{it}^{(1)'} \frac{1}{T} \sum_{s=1}^T k_{h,st}^* F_s C_{1s}^{(t)'} \left( \frac{s-t}{T} \right)^2 H^{(t)'} F_t + o_P(N^{1/2} h^3) \\
&= \sqrt{N} h^2 \kappa_2 H^{(t)-1} \Sigma_{\Lambda_t} E[F_s \bar{C}_{1s}^{(t)'}] H^{(t)'} F_t + o_P(1).
\end{aligned}$$

In addition, it is easy to see that  $\sqrt{N} \sum_{l=8}^{11} A_{3,2l}(t) = \sqrt{N} O_P(\|\hat{\delta}_t\|) = o_P(1)$ . In sum, we have

$$\begin{aligned}
\frac{1}{\sqrt{N}} \sum_{i=1}^N \hat{\lambda}_{it} [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}]' H^{(t)'} F_t &= \frac{h^2}{2} \sqrt{N} H^{(t)-1} \left\{ \left[ \frac{1}{N} \sum_{i=1}^N \lambda_{it} \beta_r^{(2)'} E(X_{is} F_s') \right] + \frac{\Lambda_t' \Lambda_t^{(2)}}{N} \Sigma_F \right\} \Sigma_F^{-1} \kappa_2 F_t \\
&\quad - \sqrt{N} h^2 \kappa_2 H^{(t)-1} \Sigma_{\Lambda_t} H^{(t)-1'} \left\{ E[(\bar{C}_{2s}^{(t)} + \bar{C}_{3s}^{(t)} + \bar{C}_{4s}^{(t)}) F_s'] + E(\bar{C}_{1s}^{(t)} \bar{C}_{1s}^{(t)'}) H^{(t)'} \right\} F_t \\
&\quad + \sqrt{N} h^2 H^{(t)-1} \Sigma_{\Lambda_t} E(F_s \bar{C}_{1s}^{(t)'}) H^{(t)'} F_t \kappa_2 + o_P(1) \\
&\equiv \sqrt{N} \tilde{B}_F(t) \kappa h^2 F_t + o_P(1),
\end{aligned}$$

where

$$\begin{aligned}
\tilde{B}_F(t) &= H^{(t)-1} \Lambda_t' E(A_{2,st} F_s') \Sigma_F^{-1} - H^{(t)-1} \Sigma_{\Lambda_t} H^{(t)-1'} \{E[(\bar{C}_{2s}^{(t)} + \bar{C}_{3s}^{(t)} + \bar{C}_{4s}^{(t)}) F_s'] + E(\bar{C}_{1s}^{(t)} \bar{C}_{1s}^{(t)'}) H^{(t)'}\} \\
&\quad + H^{(t)-1} \Sigma_{\Lambda_t} E(F_s \bar{C}_{1s}^{(t)'}) H^{(t)'}.
\end{aligned}$$

Therefore, we have

$$\begin{aligned}
\sqrt{N} A_3(t) &= \hat{S}_{\lambda,t}^{-1} \frac{1}{\sqrt{N}} \sum_{i=1}^N \hat{\lambda}_{it} [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}]' H^{(t)'} F_t = \sqrt{N} (Q_t \Sigma_{\Lambda_t} Q_t')^{-1} \tilde{B}_F(t) \kappa h^2 F_t + o_P(\sqrt{N} h^3) \\
&= \sqrt{N} B_F(t) F_t + o_P(1),
\end{aligned}$$

where  $B_F(t) = (Q_t \Sigma_{\Lambda_t} Q_t')^{-1} \tilde{B}_F(t) \kappa h^2$ . Under the additional condition that  $Nh^4 = o(1)$ ,  $\sqrt{N} A_3(t) = o_P(1)$ .

Combining these results, we have  $\sqrt{N} [\hat{F}_t - H^{(t)'} F_t] \xrightarrow{d} N(0, (\Sigma_{\Lambda_t}^{-1} Q_t^{-1})' \Gamma_{tt} \Sigma_{\Lambda_t}^{-1} Q_t^{-1})$ .

(iv) Noting that  $C_{it}^0 = \lambda'_{it} F_t$  and  $\hat{C}_{it} = \hat{\lambda}'_{it} \hat{F}_t$ , we have

$$\begin{aligned}
&\hat{C}_{it} - C_{it}^0 \\
&= \hat{\lambda}'_{it} \hat{F}_t - \lambda'_{it} F_t \\
&= [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it} - B_{\Lambda}(i, t)]' [\hat{F}_t - H^{(t)'} F_t - B_F(t)] + B_{\Lambda}(i, t)' B_F(t) + [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it} - B_{\Lambda}(i, t)]' H^{(t)'} F_t
\end{aligned}$$

$$\begin{aligned}
& + [\hat{\lambda}_{it} - H^{(t)-1}\lambda_{it} - B_\Lambda(i, t)]' B_F(t) + \lambda'_{it}(H^{(t)-1})' [\hat{F}_t - H^{(t)'} F_t - B_F(t)] + \lambda'_{it}(H^{(t)-1})' B_F(t) \\
& + B_\Lambda(i, t)' [\hat{F}_t - H^{(t)'} F_t - B_F(t)] + B_\Lambda(i, t)' H^{(t)'} F_t \\
& \equiv \sum_{l=1}^8 C_{it,l}. \tag{A.10}
\end{aligned}$$

By Theorems 3.3(i)-(ii),  $C_{it,1} = O_P((NTh)^{-1/2})$ ,  $C_{it,2} = O_P(h^4)$ ,  $C_{it,4} = O_P(T^{-1/2}h^{-1/2}h^2)$ ,  $C_{it,7} = O_P(N^{-1/2}h^2)$ . By the proof of Theorem 3.2(ii),  $\sqrt{Th}(\hat{\lambda}_{it} - H^{(t)-1}\lambda_{it} - B_\Lambda(i, t)) = \frac{\sqrt{h}}{\sqrt{T}} H^{(t)'} \sum_{s=1}^T k_{h,st}^* F_s \varepsilon_{is} + o_P(1)$ . By the proof of Theorem 3.3(iii),  $\sqrt{N}(\hat{F}_t - H^{(t)'} F_t - B_F(t)) = (Q_t \Sigma_{\Lambda_t} Q_t')^{-1} Q_t \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \varepsilon_{it} + o_P(1)$ . By Lemma A.2(iii)-(iv) and Theorem 3.2,  $H^{(t)} = Q_t^{-1} + o_P(C_{NT}^{-1})$  and  $H^{(t)'} H^{(t)'} = \Sigma_F^{-1} + o_P(C_{NT}^{-1})$ . Define  $\bar{C}_{NT} = \min\{\sqrt{Th}, \sqrt{N}\}$ . It follows that

$$\begin{aligned}
C_{it,6} + C_{it,8} & = \lambda'_{it}(H^{(t)-1})' B_F(t) + B_\Lambda(i, t)' H^{(t)'} F_t \\
& = \lambda'_{it} Q_t' B_F(t) + B_\Lambda(i, t)' (Q_t^{-1})' F_t + o_P(\bar{C}_{NT}^{-1}) \equiv B_C(i, t) + o_P(\bar{C}_{NT}^{-1}).
\end{aligned}$$

Therefore,

$$\begin{aligned}
& \bar{C}_{NT}[\hat{C}_{it} - C_{it}^0 - B_C(i, t)] \\
& = \frac{\bar{C}_{NT}}{\sqrt{N}} \lambda'_{it} Q_t' (Q_t \Sigma_{\Lambda_t} Q_t')^{-1} Q_t \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \varepsilon_{it} + \frac{\bar{C}_{NT}}{\sqrt{Th}} F_t' H^{(t)} H^{(t)'} \frac{\sqrt{h}}{\sqrt{T}} \sum_{s=1}^T k_{h,st}^* F_s \varepsilon_{is} + o_P(1) \\
& = \frac{\bar{C}_{NT}}{\sqrt{N}} \psi_{1it} + \frac{\bar{C}_{NT}}{\sqrt{Th}} \psi_{2it} + o_P(1),
\end{aligned}$$

where  $\psi_{1it} \equiv \lambda'_{it} \Sigma_{\Lambda_t}^{-1} \frac{1}{\sqrt{N}} \sum_{i=1}^N \lambda_{it} \varepsilon_{it}$  and  $\psi_{2it} \equiv F_t' \Sigma_F^{-1} \frac{\sqrt{h}}{\sqrt{T}} \sum_{s=1}^T k_{h,st}^* F_s \varepsilon_{is}$ . By Assumption A.2,  $\psi_{1it} \xrightarrow{d} N(0, V_{1it})$  and  $\psi_{2it} \xrightarrow{d} N(0, V_{2it})$ . It is easy to show that  $\xi_{1it}$  and  $\xi_{2it}$  are asymptotically independent. Consequently, we have  $V_{it}^{-1/2} \bar{C}_{NT}(\hat{C}_{it} - C_{it}^0 - B_C(i, t)) \xrightarrow{d} N(0, 1)$ , where  $V_{it} = \frac{\bar{C}_{NT}^2}{N} V_{1it} + \frac{\bar{C}_{NT}^2}{Th} V_{2it}$ . ■

**Proof of Theorem 3.4.** By (A.5) in the proof of Theorem 3.2 and the fact that  $\hat{\delta}_t \equiv \hat{\beta}_t - \beta_t = O_P(N^{-1} + (Th)^{-1} + h^2)$  and

$$\begin{aligned}
& D^{(t)}(\hat{F}^{(t)}) \sqrt{NTh}(\hat{\beta}_t - \beta_t) \\
& = \frac{\sqrt{NTh}}{NT} \sum_{i=1}^N \left[ X_i^{(t)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(t)} X_j^{(t)'} \right] M_{\hat{F}^{(t)}} \varepsilon_i^{(t)} + \frac{\sqrt{NTh}}{NT} \sum_{i=1}^N \left[ X_i^{(t)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(t)} X_j^{(t)'} \right] M_{\hat{F}^{(t)}} \Delta_i^{(t)} \\
& - \frac{\sqrt{NTh}}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(t)'} M_{\hat{F}^{(t)}} \varepsilon_j^{(t)} \varepsilon_j^{(t)'} \hat{F}^{(t)'} G^{(t)} \lambda_{it} + o_P(1),
\end{aligned}$$

where we use the condition  $NTh^7 = o(1)$  in Assumption A4(ii) to ensure that  $\sqrt{NTh} O_P(C_{NT}^{-1} + \|\hat{\delta}_t\|^{1/2} + h) O_P(\|\hat{\delta}_t\|) = o_P(1)$ . Let  $\tilde{Z}_i^{(t)} = M_{\hat{F}^{(t)}} [X_i^{(t)} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(t)} X_j^{(t)}]$ . Then

$$\begin{aligned}
& \hat{D}^{(t)}(\check{F}^{(t)}) (\hat{\beta}_t^{bc} - \beta_t) \\
& = \left[ \hat{D}^{(t)}(\check{F}^{(t)}) - D^{(t)}(\hat{F}^{(t)}) \right] [\hat{\beta}_t - \beta_t] + D^{(t)}(\hat{F}^{(t)}) [\hat{\beta}_t - \beta_t] - \left[ \hat{B}_{1\beta}^{(t)} + \frac{1}{Th} \hat{B}_{2\beta}^{(t)} + \frac{1}{N} \hat{B}_{3\beta}^{(t)} + \frac{1}{Th} \hat{B}_{4\beta}^{(t)} \right] \\
& = \left[ \hat{D}^{(t)}(\check{F}^{(t)}) - D^{(t)}(\hat{F}^{(t)}) \right] [\hat{\beta}_t - \beta_t] + \left[ \frac{1}{NT} \sum_{i=1}^N \tilde{Z}_i^{(t)'} \varepsilon_i^{(t)} - \frac{1}{Th} \hat{B}_{2\beta}^{(t)} - \frac{1}{N} \hat{B}_{3\beta}^{(t)} \right] + \frac{1}{NT} \sum_{i=1}^N \tilde{Z}_i^{(t)'} \Delta_i^{(t)}
\end{aligned}$$

$$\begin{aligned}
& - \left[ \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(t)'} M_{\hat{F}^{(t)}} \varepsilon_j^{(t)} \varepsilon_j^{(t)'} \hat{F}^{(t)} G^{(t)} \lambda_{it} - \frac{1}{Th} \hat{B}_{4\beta}^{(t)} \right] + o_P((NTh)^{-1/2}) \\
& \equiv Q_1^{(t)} + Q_2^{(t)} + Q_3^{(t)} + Q_4^{(t)} + O_P(h^4) + o_P((NTh)^{-1/2}).
\end{aligned}$$

As in the proof of Lemma A.9, we can show that

$$\left\| \hat{D}^{(t)}(\check{F}^{(t)}) - D^{(t)}(\hat{F}^{(t)}) \right\| \leq \left\| \hat{D}^{(t)}(\check{F}^{(t)}) - D^{(t)}(\check{F}^{(t)}) \right\| + \left\| D^{(t)}(\check{F}^{(t)}) - D^{(t)}(\hat{F}^{(t)}) \right\| = O_P(C_{NT}^{-2} + \|\hat{\delta}_t\|^{1/2}).$$

As a result,  $Q_1^{(t)} = O_P(C_{NT}^{-2} + \|\hat{\delta}_t\|^{1/2}) \|\hat{\delta}_t\| = o_P((NTh)^{-1/2})$ . This rate can be strengthened to be uniform in  $t$  under Assumption A.7. By Lemmas A.6 and A.8, we have

$$Q_2^{(t)} = \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \mathfrak{X}_{is}^{(t)} \varepsilon_{is}^{(t)} + \frac{1}{N} [\xi_{NT}^{(t)} - \hat{B}_{3\beta}^{(t)}] + \frac{1}{Th} [B_{2\beta}^{(t)} - \hat{B}_{2\beta}^{(t)}] \equiv Q_{2,1}^{(t)} + Q_{2,2}^{(t)} + Q_{2,3}^{(t)}.$$

Note that  $Q_{2,1}^{(t)}$  contributes to the asymptotic variance of  $\hat{\beta}_t - \beta_t : \sqrt{NTh} Q_{2,1}^{(t)} \xrightarrow{d} N(0, \Omega_t)$ . By Assumption A.5,  $\max_t \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \mathfrak{X}_{is}^{(t)} \varepsilon_{is}^{(t)} \right\| = O_P((NTh)^{-1/2} (\ln T)^{1/2})$ . For notational simplicity, let  $\hat{\varepsilon}^{(t)}(i, j) = \frac{1}{T} \sum_{s=1}^T k_{h, st}^* \hat{\varepsilon}_{is} \hat{\varepsilon}_{js}$ , and  $\varepsilon^{(t)}(i, j) = \frac{1}{T} \sum_{s=1}^T k_{h, st}^* \varepsilon_{is} \varepsilon_{js}$ . Then

$$\begin{aligned}
Q_{2,2}^{(t)} &= \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} [X_i^{(t)} - \hat{V}_i^{(t)}]' \check{F}^{(t)} \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) \\
&\quad - \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} [X_i^{(t)} - V_i^{(t)}]' F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \lambda_{jt} \varepsilon^{(t)}(i, j) \\
&= \frac{1}{N^2} \sum_{i,j=1}^N \left[ \frac{1}{T} X_i^{(t)'} \check{F}^{(t)} \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) - \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \lambda_{jt} \varepsilon^{(t)}(i, j) \right] \\
&\quad - \frac{1}{N^2} \sum_{i,j=1}^N \left[ \frac{1}{T} \hat{V}_i^{(t)'} \check{F}^{(t)} \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) - \frac{1}{T} V_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \lambda_{jt} \varepsilon^{(t)}(i, j) \right] \\
&\equiv Q_{2,21}^{(t)} + Q_{2,22}^{(t)}
\end{aligned}$$

We first consider the term  $Q_{2,21}^{(t)}$ . Using the identity  $\hat{a}\hat{b}\hat{c}\hat{d} - abcd = (\hat{a}-a)\hat{b}\hat{c}\hat{d} + a(\hat{b}-b)\hat{c}\hat{d} + ab(\hat{c}-c)\hat{d} + abc(\hat{d}-d)$ , we can decompose  $Q_{2,21}^{(t)}$  into four terms:

$$\begin{aligned}
Q_{2,21}^{(t)} &= \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} [\check{F}^{(t)} - F^{(t)} H^{(t)}] \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) \\
&\quad + \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ H^{(t)} \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} - \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \right] \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) \\
&\quad + \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] \hat{\varepsilon}^{(t)}(i, j) \\
&\quad + \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \lambda_{jt} \left[ \hat{\varepsilon}^{(t)}(i, j) - \varepsilon^{(t)}(i, j) \right] \equiv \sum_{l=1}^4 Q_{2,21l}^{(t)}.
\end{aligned}$$

For  $Q_{2,211}^{(t)}$ , we have

$$\begin{aligned} \left\| Q_{2,211}^{(t)} \right\| &\leq \max_t \left\| \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} [\check{F}^{(t)} - F^{(t)} H^{(t)}] \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) \right\| \\ &\lesssim \max_{i,t} \left\| \frac{1}{T} X_i^{(t)'} [\check{F}^{(t)} - F^{(t)} \tilde{H}^{(t)}] \right\| \max_t \frac{1}{N^2} \sum_{i,j=1}^N \left\| \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) \right\|. \end{aligned}$$

Following the proof of Theorem 3.3(iii) and using Assumption A.7, we can show that  $\frac{1}{T} X_i^{(t)'} [\check{F}^{(t)} - F^{(t)} H^{(t)}] = O_P(N^{-1} + (Th)^{-1} + h^2)$  for each  $(i, t)$  and  $\max_{i,t} \left\| \frac{1}{T} X_i^{(t)'} [\check{F}^{(t)} - F^{(t)} H^{(t)}] \right\| = O_P((N^{-1} + (Th)^{-1}) \ln T + h^2)$ . Next, note that

$$\begin{aligned} &\frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) \right\| \\ &\leq \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N [\hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt}] \hat{\varepsilon}^{(t)}(i, j) \right\| + \left\| H^{(t)-1} \right\| \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N \lambda_{jt} \hat{\varepsilon}^{(t)}(i, j) \right\| \\ &\lesssim \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N [\hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt}] \varepsilon^{(t)}(i, j) \right\| + \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N [\hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt}] [\hat{\varepsilon}^{(t)}(i, j) - \varepsilon^{(t)}(i, j)] \right\| \\ &\quad + \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N \lambda_{jt} \varepsilon^{(t)}(i, j) \right\| + \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N \lambda_{jt} [\hat{\varepsilon}^{(t)}(i, j) - \varepsilon^{(t)}(i, j)] \right\| \equiv \sum_{l=1}^4 a_l^{(t)}. \end{aligned}$$

For  $a_1^{(t)}$ , it suffices to consider the rough probability bound:

$$\begin{aligned} \max_t a_1^{(t)} &= \max_t \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N [\hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt}] \varepsilon^{(t)}(i, j) \right\| \\ &\leq \max_t \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N^2} \sum_{i,j=1}^N \left\| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \varepsilon_{is} \varepsilon_{js} \right\|^2 \right\}^{1/2} \\ &= O_P((Th/\ln T)^{-1/2} + h^2) O_P(N^{-1/2} + (Th/\ln T)^{-1/2}), \end{aligned}$$

where we use Theorem 3.5(i) below and the fact that

$$\begin{aligned} &\max_t \frac{1}{N^2} \sum_{i,j=1}^N \left| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \varepsilon_{is} \varepsilon_{js} \right|^2 \\ &\leq \max_t \frac{2}{N^2} \sum_{i,j=1}^N \left| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* E(\varepsilon_{is} \varepsilon_{js}) \right|^2 + \max_t \frac{2}{N^2} \sum_{i,j=1}^N \left| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* [\varepsilon_{is} \varepsilon_{js} - E(\varepsilon_{is} \varepsilon_{js})] \right|^2 \\ &\leq \max_t \frac{2}{T^2} \sum_{s=1}^T k_{h,st}^* \sum_{r=1}^T k_{h,rt}^* \frac{1}{N^2} \sum_{i,j=1}^N E(\varepsilon_{is} \varepsilon_{js}) E(\varepsilon_{ir} \varepsilon_{jr}) + \max_t \frac{2}{N^2} \sum_{i,j=1}^N \left| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* [\varepsilon_{is} \varepsilon_{js} - E(\varepsilon_{is} \varepsilon_{js})] \right|^2 \\ &\lesssim N^{-1} \max_s \frac{1}{N} \sum_{i,j=1}^N |E(\varepsilon_{is} \varepsilon_{js})| + O_P((Th/\ln T)^{-1}) = O_P(N^{-1} + (Th/\ln T)^{-1}). \end{aligned}$$



It is easy to show that

$$\begin{aligned}
& \max_t a_3^{(t)} \\
& \lesssim \max_t \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N \lambda_{jt} \frac{1}{T} \sum_{s=1}^T k_{h, st}^* \varepsilon_{is} \varepsilon_{js} \right\| \\
& \leq \max_t \frac{1}{N^2} \sum_{i=1}^N \left\| \sum_{j=1}^N \lambda_{jt} \frac{1}{T} \sum_{s=1}^T k_{h, st}^* E(\varepsilon_{is} \varepsilon_{js}) \right\| + \max_t \frac{1}{N^2} \sum_{i=1}^N \left\| \frac{1}{NT} \sum_{j=1}^N \sum_{s=1}^T \lambda_{jt} k_{h, st}^* [\varepsilon_{is} \varepsilon_{js} - E(\varepsilon_{is} \varepsilon_{js})] \right\| \\
& \lesssim \frac{1}{N} \max_{i,s} \frac{1}{N} \sum_{i,j=1}^N |E(\varepsilon_{is} \varepsilon_{js})| + O_P((NTh/\ln T)^{-1/2}) = O_P(N^{-1} + (NTh/\ln T)^{-1/2}).
\end{aligned}$$

Similarly, we can show that  $\max_t a_2^{(t)}$  and  $\max_t a_4^{(t)}$  are of order no larger than  $\max_t a_1^{(t)}$  and  $\max_t a_3^{(t)}$ , respectively. Then

$$\max_t \sum_{l=1}^4 a_l^{(t)} = O_P(Th/\ln T + N^{-1/2}h^2 + (Th/\ln T)^{-1/2}h^2 + N^{-1} + (NTh/\ln T)^{-1/2})$$

and  $\max_t \left\| Q_{2,211}^{(t)} \right\| = O_P((N^{-1} + (Th)^{-1}) \ln T + h^2) O_P(Th/\ln T + N^{-1/2}h^2 + (Th/\ln T)^{-1/2}h^2 + N^{-1} + (NTh/\ln T)^{-1/2}) = o_P((NTh)^{-1/2})$ .

Noting that  $H^{(t)} \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t H^{(t)'} - \frac{1}{N} \Lambda'_t \Lambda_t = H^{(t)} (\frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t - \frac{1}{N} \Lambda'_t \Lambda_t) H^{(t)'} = H^{(t)} \frac{1}{N} (\hat{\Lambda}_t - \Lambda_t)' (\hat{\Lambda}_t - \Lambda_t) H^{(t)'} + H^{(t)} \frac{1}{N} (\hat{\Lambda}_t - \Lambda_t)' \Lambda_t H^{(t)'} + H^{(t)} \frac{1}{N} \Lambda'_t (\hat{\Lambda}_t - \Lambda_t) H^{(t)'}$ , it is easy to follow the proof of Theorem 3.3(ii) and using the uniform consistency of  $H^{(t)}$  to show that  $\max_t \left\| H^{(t)} \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t H^{(t)'} - \frac{1}{N} \Lambda'_t \Lambda_t \right\| = O_P((Th)^{-1} + N^{-1} \ln T + h^2)$ . This result, in conjunction with the fact that  $\max_t \left\| (H^{(t)} H^{(t)'})^{-1} - \frac{1}{T} F^{(t)'} F^{(t)} \right\| = O_P((Th)^{-1} + N^{-1} \ln T + h^2)$  established in the proof of Lemma A.11 and the uniform consistency of  $H^{(t)}$  and  $\frac{1}{T} F^{(t)'} F^{(t)}$ , implies that

$$\begin{aligned}
& \left\| H^{(t)} \left( \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t \right)^{-1} - \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right)^{-1} H^{(t)} \right\| \\
& = \left\| \left[ H^{(t)} H^{(t)'} \left( H^{(t)} \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t H^{(t)'} \right)^{-1} - \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right)^{-1} \right] H^{(t)} \right\| \\
& \leq \left\| H^{(t)} H^{(t)'} - \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \right\| \left\| \left( H^{(t)} \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t H^{(t)'} \right)^{-1} \right\| \left\| H^{(t)} \right\| \\
& \quad + \left\| \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \right\| \left\| \left( H^{(t)} \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t H^{(t)'} \right)^{-1} - \left( \frac{1}{N} \Lambda'_t \Lambda_t \right)^{-1} \right\| \left\| H^{(t)} \right\| = O_P((Th)^{-1} + N^{-1} \ln T + h^2).
\end{aligned}$$

Then as in the analysis of  $Q_{2,211}^{(t)}$ , we can readily show that

$$\begin{aligned}
& \max_t \left\| Q_{2,212}^{(t)} \right\| \\
& \lesssim \max_t \left\| H^{(t)} \left( \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t \right)^{-1} - \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right)^{-1} H^{(t)} \right\| \frac{1}{N^2} \sum_{i=1}^N \left| \sum_{j=1}^N \hat{\lambda}_{jt} \hat{\varepsilon}^{(t)}(i, j) \right| \\
& = O_P((Th)^{-1} + N^{-1} \ln T + h^2) O_P(Th/\ln T + N^{-1/2}h^2 + (Th/\ln T)^{-1/2}h^2 + N^{-1} + (NTh/\ln T)^{-1/2}) \\
& = o_P((NTh)^{-1/2}).
\end{aligned}$$

Now, we study  $Q_{2,213}^{(t)}$ . Note that

$$\begin{aligned}
Q_{2,213}^{(t)} &= \frac{1}{N} \sum_{i=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \frac{1}{N} \sum_{j=1}^N \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] \varepsilon^{(t)}(i, j) \\
&\quad + \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] \left[ \hat{\varepsilon}^{(t)}(i, j) - \varepsilon^{(t)}(i, j) \right] \\
&\equiv Q_{2,213a}^{(t)} + Q_{2,213b}^{(t)}.
\end{aligned}$$

Following the proof of Theorem 3.3(ii) and using the fact that  $\max_{i \neq j} \max_t \left| \frac{1}{T} \sum_{s=1}^T k_{h, st}^* \varepsilon_{is} \varepsilon_{js} \right| = O_P((Th/\ln T)^{-1/2})$ , we can readily show that uniformly in  $t$  and  $i$ ,

$$\begin{aligned}
&\frac{1}{N} \sum_{j=1}^N \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] \varepsilon^{(t)}(i, j) \\
&= \frac{1}{N} \sum_{j=1}^N D_1(j, t) \frac{1}{T} \sum_{s=1}^T k_{h, st}^* \varepsilon_{is} \varepsilon_{js} + o_P((NTh)^{-1/2}) \\
&= \frac{1}{NT^2} H^{(t)'} \sum_{j=1}^N \sum_{r=1}^T k_{h, rt}^* F_r \varepsilon_{jr} \sum_{s=1}^T k_{h, st}^* \varepsilon_{is} \varepsilon_{js} + o_P((NTh)^{-1/2}) = H^{(t)'} b_i^{(t)} + o_P((NTh)^{-1/2}),
\end{aligned}$$

where  $b_i^{(t)} = \frac{1}{NT^2} \sum_{j=1}^N \sum_{r=1}^T \sum_{s=1}^T k_{h, st}^* k_{h, rt}^* F_r \varepsilon_{jr} \varepsilon_{is} \varepsilon_{js}$ . Note that

$$\begin{aligned}
\max_{i,t} \|b_i^{(t)}\| &\leq \max_{i,t} \frac{1}{NT^2} \sum_{j=1}^N \sum_{r=1}^T \sum_{s=1}^T k_{h, rt}^* k_{h, st}^* |E(F_r \varepsilon_{jr} \varepsilon_{is} \varepsilon_{js})| \\
&\quad + \max_{i,t} \left\| \frac{1}{NT^2} \sum_{j=1}^N \sum_{r=1}^T \sum_{s=1}^T k_{h, rt}^* k_{h, st}^* [F_r \varepsilon_{jr} \varepsilon_{is} \varepsilon_{js} - E(F_r \varepsilon_{jr} \varepsilon_{is} \varepsilon_{js})] \right\| \\
&\lesssim N^{-1} T^{-1} h^{-2} \max_i \frac{1}{T} \sum_{j=1}^N \sum_{r=1}^T \sum_{s=1}^T |E(F_r \varepsilon_{jr} \varepsilon_{is} \varepsilon_{js})| \\
&\quad + \max_{i,t} \left\| \frac{1}{NT^2} \sum_{j=1}^N \sum_{r=1}^T \sum_{s=1}^T k_{h, rt}^* k_{h, st}^* [F_r \varepsilon_{jr} \varepsilon_{is} \varepsilon_{js} - E(F_r \varepsilon_{jr} \varepsilon_{is} \varepsilon_{js})] \right\| \\
&= O(N^{-1} T^{-1} h^{-2}) + O_P(N^{-1/2} T^{-1} h^{-1} (\ln T)^{1/2}) = o_P((NTh)^{-1/2}).
\end{aligned}$$

Then  $\max_{i,t} \frac{1}{N} \sum_{j=1}^N \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] \varepsilon^{(t)}(i, j) = \frac{1}{N} \sum_{j=1}^N \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] \varepsilon^{(t)}(i, j)$ . With this, we can readily show that  $\max_t \|Q_{2,213a}^{(t)}\| = o_P((NTh)^{-1/2})$ . Noting that

$$\begin{aligned}
\hat{\varepsilon}^{(t)}(i, j) - \varepsilon^{(t)}(i, j) &= \frac{1}{T} \sum_{s=1}^T [\hat{\varepsilon}_{is}^{(t)} \hat{\varepsilon}_{js}^{(t)} - \varepsilon_{is}^{(t)} \varepsilon_{js}^{(t)}] \\
&= \frac{1}{T} \sum_{s=1}^T \{ [\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)}] [\hat{\varepsilon}_{js}^{(t)} - \varepsilon_{js}^{(t)}] + [\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)}] \varepsilon_{js}^{(t)} + \varepsilon_{is}^{(t)} [\hat{\varepsilon}_{js}^{(t)} - \varepsilon_{js}^{(t)}] \} \equiv \sum_{l=1}^3 c_{l,ij}^{(t)},
\end{aligned}$$

we have

$$\begin{aligned}
Q_{2,213b}^{(t)} &= \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] \left[ \hat{\varepsilon}^{(t)}(i,j) - \varepsilon^{(t)}(i,j) \right] \\
&= \sum_{l=1}^3 \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \left[ \hat{\lambda}_{jt} - H^{(t)-1} \lambda_{jt} \right] c_{l,ij}^{(t)} \equiv \sum_{l=1}^3 Q_{2,213}^{(t)}(l).
\end{aligned}$$

Using

$$\begin{aligned}
&\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)} \\
&= -X_{is}^{(t)'} \hat{\delta}_s - (\check{F}_s^{(t)} - H^{(t)'} F_s^{(t)})' H^{(t)-1} \lambda_{is}^{(t)} - (\check{F}_s^{(t)} - H^{(t)'} F_s^{(t)})' (\hat{\lambda}_{is} - H^{(t)-1} \lambda_{is}) - F_s^{(t)'} H^{(t)} (\hat{\lambda}_{is} - H^{(t)-1} \lambda_{is}) \\
&\equiv \sum_{l=1}^4 e_{is}^{(t)}(l),
\end{aligned}$$

we can readily show that

$$\begin{aligned}
&\max_{i,j,t} \left| \frac{1}{T} \sum_{s=1}^T [\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)}] \varepsilon_{js}^{(t)} \right| \\
&\leq \max_{i,j,t} \left| \frac{1}{T} \sum_{s=1}^T \varepsilon_{js}^{(t)} X_{is}^{(t)'} \hat{\delta}_s \right| + \max_{i,j,t} \left| \frac{1}{T} \sum_{s=1}^T (\check{F}_s^{(t)} - H^{(t)'} F_s^{(t)})' H^{(t)-1} \lambda_{js}^{(t)} \right| \\
&\quad + \max_{i,j,t} \left| \frac{1}{T} \sum_{s=1}^T (\check{F}_s^{(t)} - H^{(t)'} F_s^{(t)})' (\hat{\lambda}_{is} - H^{(t)-1} \lambda_{is}) \varepsilon_{js}^{(t)} \right| + \max_{i,j,t} \left| \frac{1}{T} \sum_{s=1}^T F_s^{(t)'} H^{(t)} (\hat{\lambda}_{is} - H^{(t)-1} \lambda_{is}) \varepsilon_{js}^{(t)} \right| \\
&= O_P((N^{-1} + (Th)^{-1}) \ln T + h^4)
\end{aligned}$$

and similarly,

$$\begin{aligned}
&\max_t \frac{1}{T} \sum_{s=1}^T \left\| \frac{1}{NT} \sum_{i=1}^N [\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)}] \omega' X_i^{(t)'} F^{(t)} \right\| \\
&= \max_t \frac{1}{T} \sum_{s=1}^T \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{r=1}^T k_{h,rt}^* [\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)}] \omega' X_{ir} F_r \right\| = O_P((N^{-1} + (Th)^{-1}) \ln T + h^2).
\end{aligned}$$

With the first result, it is easy to show that  $\max_t \left\| Q_{2,213b}^{(t)}(l) \right\| = o_P((NTh)^{-1/2})$  for  $l = 1, 2, 3$ . Then  $\max_t \left\| Q_{2,213b}^{(t)} \right\| = o_P((NTh)^{-1/2})$  and  $\max_t \left\| Q_{2,213}^{(t)} \right\| = o_P((NTh)^{-1/2})$ .

For  $Q_{2,214}^{(t)}$ , we have

$$\begin{aligned}
Q_{2,214}^{(t)} &= \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \lambda_{jt} \left[ \hat{\varepsilon}^{(t)}(i,j) - \varepsilon^{(t)}(i,j) \right] \\
&= \sum_{l=1}^3 \frac{1}{N^2} \sum_{i,j=1}^N \frac{1}{T} X_i^{(t)'} F^{(t)} \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \lambda_{jt} c_{l,ij}^{(t)} \equiv \sum_{l=1}^3 Q_{2,214}^{(t)}(l).
\end{aligned}$$

We can show that  $\max_t \left\| Q_{2,214}^{(t)} \right\| = o_P((NTh)^{-1/2})$  by showing that  $\max_t \left\| Q_{2,214}^{(t)}(l) \right\| = o_P((NTh)^{-1/2})$  for

$l = 1, 2, 3$ . For example, note that

$$\begin{aligned}
& \max_t \left| \omega' Q_{2,214}^{(t)}(2) \right| \\
&= \max_t \left| \text{tr} \left( \left[ \frac{1}{T} F^{(t)'} F^{(t)} \right]^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \frac{1}{T} \sum_{s=1}^T \frac{1}{N} \sum_{j=1}^N \lambda_{jt} \varepsilon_{js}^{(t)} \frac{1}{N} \sum_{i=1}^N [\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)}] \frac{1}{T} \omega' X_i^{(t)'} F^{(t)} \right) \right| \\
&\lesssim \max_t \left\| \frac{1}{N^2 T^2} \sum_{s=1}^T \sum_{r=1}^T \sum_{j=1}^N k_{h,rt}^* \lambda_{jt} \varepsilon_{js}^{(t)} \sum_{i=1}^N [\hat{\varepsilon}_{is}^{(t)} - \varepsilon_{is}^{(t)}] \omega' X_{ir} F_r \right\| \\
&\leq \sum_{l=1}^4 \max_t \left\| \frac{1}{N^2 T^2} \sum_{s=1}^T \sum_{r=1}^T \sum_{j=1}^N k_{h,rt}^* \lambda_{jt} \varepsilon_{js}^{(t)} \sum_{i=1}^N e_{is}^{(t)}(l) \omega' X_{ir} F_r \right\| \equiv \sum_{l=1}^4 E_l,
\end{aligned}$$

and we can show that  $E_l = o_P((NTh)^{-1/2})$  for  $l = 1, 2, 3, 4$ . In sum, we have  $\max_t \left\| Q_{2,21}^{(t)} \right\| = o_P((NTh)^{-1/2})$ .

Now, we consider the  $Q_{2,22}^{(t)}$  term. The only difference between  $Q_{2,21}^{(t)}$  and  $Q_{2,22}^{(t)}$  is that  $X_i^{(t)}$  is replaced by  $\hat{V}_i^{(t)}$ . The proof will be virtually the same if  $\hat{V}_i^{(t)}$  is substituted by  $V_i^{(t)}$ . It is easy to argue that this substitution does not cause any difficulty in the analysis of  $Q_{2,22}^{(t)}$ . To see this, note that  $\hat{V}_i^{(t)} - V_i^{(t)} = N^{-1} \sum_{k=1}^N (\hat{a}_{ik}^{(t)} - a_{ik}^{(t)}) X_k^{(t)}$ , where

$$\begin{aligned}
\hat{a}_{ik}^{(t)} - a_{ik}^{(t)} &= [\hat{\lambda}_{it} - H^{(t)-1} \lambda_{it}]' (N^{-1} \hat{\Lambda}_t' \hat{\Lambda}_t)^{-1} \hat{\lambda}_{kt} + \lambda_{it}' H^{(t)'}{}^{-1} \left[ (N^{-1} \hat{\Lambda}_t' \hat{\Lambda}_t)^{-1} - H^{(t)'} (N^{-1} \Lambda_t' \Lambda_t)^{-1} H^{(t)} \right] \hat{\lambda}_{kt} \\
&\quad + \lambda_{it}' (N^{-1} \Lambda_t' \Lambda_t)^{-1} H^{(t)} (\hat{\lambda}_{kt} - H^{(t)-1} \lambda_{kt}).
\end{aligned} \tag{A.11}$$

Using Theorem 3.5(ii) below, we can readily show that  $\max_{i,k,t} \left| \hat{a}_{ik}^{(t)} - a_{ik}^{(t)} \right| = O_P((Th/\ln T)^{-1/2} + h^2)$ , which implies that  $\max_{i,s,t} \left| \hat{V}_{is}^{(t)} - V_{is}^{(t)} \right| = O_P((Th/\ln T)^{-1/2} + h^2)$ . Using the above decomposition, we can also argue that

$$\max_t \left\| \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N \frac{1}{T} (\hat{V}_i^{(t)} - V_i^{(t)})' F^{(t)} \left( \frac{1}{T} F^{(t)'} F^{(t)} \right)^{-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} \lambda_{jt} \frac{1}{T} \sum_{s=1}^T \varepsilon_{is}^{(t)} \varepsilon_{js}^{(t)} \right\| = o_P((NTh)^{-1/2})$$

and similarly for other differences arising by replacing  $V_i^{(t)}$  by  $\hat{V}_i^{(t)}$ . Then  $\max_t \left\| Q_{2,22}^{(t)} \right\| = o_P((NTh)^{-1/2})$  and  $\max_t \left\| Q_{2,2}^{(t)} \right\| = o_P((NTh)^{-1/2})$ .

Now, we consider the  $Q_{2,3}^{(t)}$  term. Note that  $\hat{B}_{2\beta}^{(t)}$  is analogous to  $\hat{\mathcal{B}}_{3NT}$  defined in Su and Ju (2018, SJ hereafter) and the only difference is that both the residual  $\hat{\varepsilon}_{ir}^{(t)}$  and the regressor  $X_{i,s}^{(t)}$  are now kernel-weighted in this paper and they are non-weighted in SJ. Following the proof of consistency of  $\hat{\mathcal{B}}_{3NT}$  in that of Corollary 3.9 of SJ, we can also show that  $\max_t \left\| Q_{2,3}^{(t)} \right\| = \max_t \frac{1}{Th} \left\| B_{2\beta}^{(t)} - \hat{B}_{2\beta}^{(t)} \right\| = o_P((NTh)^{-1/2})$ .

In sum, we have shown that  $\max_t \left\| Q_2^{(t)} \right\| = o_P((NTh)^{-1/2})$ .

Note that  $\hat{B}_{4\beta}^{(t)}$  is analogous to  $\hat{\mathcal{B}}_{1NT}$  defined in SJ. There are two main differences: 1) both the residual  $\hat{\varepsilon}_{ir}^{(t)}$  and the regressor  $X_{i,s}^{(t)}$  are now kernel weighted while they are non-weighted in SJ, and 2) the definitions in SJ use the fact that the error terms are serially uncorrelated conditional on the common sigma-field  $\mathcal{C}$  while we do not assume so. Despite these differences, we can follow SJ's proofs of Corollary 3.9 closely and show that

$$\max_t \left\| Q_4^{(t)} \right\| = \max_t \frac{1}{Th} \left\| B_{4\beta}^{(t)} - \hat{B}_{4\beta}^{(t)} \right\| = o_P((NTh)^{-1/2}).$$

By Lemma A.7,  $Q_3^{(t)} = \frac{1}{NT} \sum_{i=1}^N \tilde{Z}_i^{(t)'} \Delta_i^{(t)} = B_{1\beta}^{(t)} + o_P((NT)h)^{-1/2}$ . It is easy to show that  $\max_t \|B_{1\beta}^{(t)}\| = O_P(h^2) = O_P((NT)h/\ln T)^{-1/2}$  when  $h = O((NT)^{-1/5})$ .

It follows that  $\max_t \|\hat{\beta}_t^{bc} - \hat{\beta}_t\| = O_P((NT)h/\ln T)^{-1/2}$  and  $\sqrt{NT}h(\hat{\beta}_t^{bc} - \hat{\beta}_t - [D^{(t)}(F^{(t)})]^{-1}B_{1\beta}^{(t)})$  is asymptotically normal with mean zero and variance-covariance matrix given in Theorem 3.2. ■

**Proof of Theorem 3.5.** Given the fast convergence rate of  $\hat{\beta}_t^{bc}$  in Theorem 3.4, the proof is almost the same as that of Theorem 2.5 in Su and Wang (2020) and is thus omitted. ■

**Proof of Theorem 3.6.** The proof is almost the same as that of Theorem 3.6 in Su and Wang (2017), and is thus omitted. ■

## B Proofs of Theorems in Section 4

To prove Theorems 4.1 and 4.3, we need the following lemma from Su and Chen (2013, Lemma A.5). It is also a conditional version of Lemma 2.1 in Sun and Chiang (1997).

**Lemma B.1** *Let  $\{\xi_t, t \geq 1\}$  be an  $l$ -dimensional strong mixing process conditional on  $\mathcal{C}$  with mixing coefficient  $\alpha_{\mathcal{C}}(\cdot)$  and distribution function  $F_t(\cdot|\mathcal{C})$ . Let the integers  $(t_1, \dots, t_m)$  be such that  $1 \leq t_1 < t_2 < \dots < t_m \leq T$ . Suppose that  $\max\{\int |\vartheta(v_1, \dots, v_m)|^{1+\tilde{\eta}} dF_{t_1, \dots, t_m}(v_1, \dots, v_m|\mathcal{C}), \int |\theta(v_1, \dots, v_m)|^{1+\tilde{\eta}} dF_{t_1, \dots, t_j}(v_1, \dots, v_j|\mathcal{C}) dF_{t_{j+1}, \dots, t_m}(v_{j+1}, \dots, v_m|\mathcal{C})\} \leq C_C(t_1, \dots, t_m)$  for some  $\tilde{\eta} > 0$ , where, e.g.,  $F_{t_1, \dots, t_m}(v_1, \dots, v_m|\mathcal{C})$  denotes the distribution function of  $(\xi_{t_1}, \dots, \xi_{t_m})$  given  $\mathcal{C}$ . Then*

$$\left| \int \theta(v_1, \dots, v_m) dF_{t_1, \dots, t_m}(v_1, \dots, v_m|\mathcal{C}) - \int \theta(v_1, \dots, v_m) dF_{t_1, \dots, t_j}^{(1)}(v_1, \dots, v_j|\mathcal{C}) dF_{t_{j+1}, \dots, t_m}(v_{j+1}, \dots, v_m) \right| \leq 4C_C(t_1, \dots, t_m)^{1/(1+\tilde{\eta})} \alpha_{\mathcal{C}}(t_{j+1} - t_j)^{\tilde{\eta}/(1+\tilde{\eta})}.$$

**Proof of Theorem 4.1** (i) For the convenience of proving Theorem 4.3 below, we prove that under  $\mathbb{H}_A^{(1)}(a_{1NT})$  with  $a_{1NT} = N^{-1/4}T^{-1/2}h^{-1/4}$ , we have

$$J_{NT}^{(1)} = TN^{1/2}h^{1/2}\hat{M}^{(1)} - \mathbb{B}_{NT}^{(1)} - \Pi^{(1)} \xrightarrow{d} N(0, \mathbb{V}_0^{(1)}).$$

Noting that  $\hat{\lambda}'_t \hat{F}_t = (\hat{\lambda}_t - H^{(t)-1}\lambda_t)' H^{(t)'} F_t + \lambda'_{it} (H^{(t)-1})' (\hat{F}_t - H^{(t)'} F_t) + (\hat{\lambda}_t - H^{(t)-1}\lambda_t)' (\hat{F}_t - H^{(t)'} F_t) + \lambda'_{it} F_t$  and  $\tilde{\lambda}'_{i0} \tilde{F}_t = (\tilde{\lambda}_{i0} - H^{-1}\lambda_{i0})' H' F_t + \lambda'_{i0} (H^{-1})' (\tilde{F}_t - H' F_t) + (\tilde{\lambda}_{i0} - H^{-1}\lambda_{i0})' (\tilde{F}_t - H' F_t) + \lambda'_{i0} F_t$ , we have  $\hat{\lambda}'_t \hat{F}_t - \tilde{\lambda}'_{i0} \tilde{F}_t = d_{1it} + d_{2it} + d_{3it}$ , where

$$\begin{aligned} d_{1it} &= F_t' H^{(t)} (\hat{\lambda}_t - H^{(t)-1}\lambda_t) - F_t' H (\tilde{\lambda}_{i0} - H^{-1}\lambda_{i0}) + \lambda'_{it} (H^{(t)-1})' (\hat{F}_t - H^{(t)'} F_t) - \lambda'_{i0} (H^{-1})' (\tilde{F}_t - H' F_t), \\ d_{2it} &= (\lambda_{it} - \lambda_{i0})' F_t, \text{ and} \\ d_{3it} &= (\hat{\lambda}_t - H^{(t)-1}\lambda_t)' (\hat{F}_t - H^{(t)'} F_t) - (\tilde{\lambda}_{i0} - H^{-1}\lambda_{i0})' (\tilde{F}_t - H' F_t). \end{aligned}$$

Let  $d_{4it} = X'_{it}(\hat{\beta}_t^{bc} - \tilde{\beta})$ . Then

$$\begin{aligned} TN^{1/2}h^{1/2}\hat{M}^{(1)} &= N^{-1/2}h^{1/2} \sum_{i=1}^N \sum_{t=1}^T [d_{1it} + d_{2it} + d_{3it} + d_{4it}]^2 \\ &= N^{-1/2}h^{1/2} \sum_{i=1}^N \sum_{t=1}^T [d_{1it}^2 + d_{2it}^2 + d_{3it}^2 + 2d_{1it}d_{2it} + 2d_{1it}d_{3it} + 2d_{2it}d_{3it} + d_{4it}^2 + 2d_{1it}d_{4it} + 2d_{2it}d_{4it} + 2d_{3it}d_{4it}] \\ &\equiv M_1 + M_2 + M_3 + 2M_4 + 2M_5 + 2M_6 + M_7 + 2M_8 + 2M_9 + 2M_{10}. \end{aligned}$$

The terms  $M_1$  to  $M_6$  have been analyzed by Su and Wang (2017, 2020). Following their analyses, we can also show that  $M_2 = \Pi_2^{(1)} + o_P(1)$ , and  $M_j = o_P(1)$  for  $j = 3, 4, 5, 6$ . It remains to show that (i1)  $M_1 - \mathbb{B}_{NT}^{(1)} - \Pi_1^{(1)} \xrightarrow{d} N(0, \mathbb{V}_0^{(1)})$ , (i2)  $M_7 = \Pi_3^{(1)} + o_P(1)$ , and (i3)  $M_j = o_P(1)$  for  $j = 8, 9, 10$ .

First, we show (i1). The asymptotic bias and variance of  $M_1$  are different from that of the corresponding object in Su and Wang (2017) because we do not have the m.d.s. assumption on the error term. As in Su and Wang (2017), let  $L_{st} = k_{h,st}^* H^{(t)} H^{(t)'} - HH'$  and  $\bar{L}_{st} = (k_{h,st}^* - 1)H_0 H_0'$ , where  $H_0$  is the probability limit of both  $H^{(t)}$  and  $H$  under the local alternative. By Lemma B.1(ii) and B.2(vi) in Su and Wang (2017),

$$\max_t \left\| H^{(t)} - H_0 \right\| = O_P(C_{NT}^{-1} (\ln T)^{1/2}) \quad \text{and} \quad \|H - H_0\| = O_P(N^{-1/2} + T^{-1/2}). \quad (\text{B.1})$$

Following their analysis, we can show that the leading term in  $d_{1it}$  is given by  $d_{1it,1} + d_{1it,2}$  such that  $M_1 = M_{1,1} + M_{1,2} + o_P(1)$ , where  $d_{1it,1} \equiv F_t' \frac{1}{T} \sum_{s=1}^T L_{st} F_s \varepsilon_{is}$ ,  $d_{1it,2} \equiv a_{1NT} [F_t' H H' \frac{1}{T} \sum_{s=1}^T F_s g_{is}^\dagger + \lambda_{i0}' (H^{-1})' V_{NT}^{-1} \times (\frac{1}{T} \tilde{F}' F) \Lambda_0' g_t^\dagger / N]$ , and  $M_{1,l} = N^{-1/2} h^{1/2} \sum_{i=1}^N \sum_{t=1}^T d_{1it,l}^2$  for  $l = 1, 2$ . Following Su and Wang (2017), we can show that  $M_{1,2} = \Pi_1^{(1)} + o_P(1)$ . Let  $\bar{L}_{st} = (k_{h,st}^* - 1)H_0 H_0'$ . For  $M_{1,1}$ , we can decompose it as follows:

$$\begin{aligned} M_{1,1} &= \frac{h^{1/2}}{N^{1/2} T^2} \sum_{i=1}^N \sum_{t,s,r=1}^T F_t' L_{st} F_s F_r' L_{rt}' F_t \varepsilon_{is} \varepsilon_{ir} \\ &= \frac{h^{1/2}}{N^{1/2} T^2} \sum_{i=1}^N \sum_{t,s,r=1}^T \{ F_t' L_{st} F_s F_r' L_{rt}' F_t E_C(\varepsilon_{i,sr}) + F_t' \bar{L}_{st} F_s F_r' \bar{L}_{rt}' F_t \tilde{\varepsilon}_{i,sr} \\ &\quad + 2F_t' (L_{st} - \bar{L}_{st}) F_s F_r' \bar{L}_{rt}' F_t \tilde{\varepsilon}_{i,sr} + F_t' (L_{st} - \bar{L}_{st}) F_s F_r' (L_{rt} - \bar{L}_{rt})' F_t \tilde{\varepsilon}_{i,sr} \} \equiv \sum_{\ell=1}^4 M_{1,1}^{(\ell)}, \end{aligned}$$

where  $\varepsilon_{i,sr} \equiv \varepsilon_{is} \varepsilon_{ir}$  and  $\tilde{\varepsilon}_{i,sr} = \varepsilon_{is} \varepsilon_{ir} - E_C(\varepsilon_{is} \varepsilon_{ir})$ . Note that  $M_{1,1}^{(1)} = \mathbb{B}_{NT}^{(1)}$ . Following the analysis of  $M_{1,1}^{(2)}$  below, we can readily show that  $M_{1,1}^{(l)} = o_P(1)$  for  $l = 3, 4$ . For  $M_{1,1}^{(2)}$ , we can further decompose it as follows

$$\begin{aligned} M_{1,1}^{(2)} &= \frac{h^{1/2}}{N^{1/2} T} \sum_{i=1}^N \sum_{s,r=1}^T \text{tr} \{ F_s F_r' \frac{1}{T} \sum_{t=1}^T \bar{L}_{rt}' F_t F_t' \bar{L}_{st} \} \tilde{\varepsilon}_{i,sr} \\ &= \frac{h^{1/2}}{N^{1/2} T} \sum_{i=1}^N \sum_{s,r=1}^T \frac{1}{T} \sum_{t=1}^T \text{tr} \{ F_s F_r' H_0 H_0' (k_{h,st}^* k_{h,rt}^* - 2k_{h,st} + 1) F_t F_t' H_0 H_0' \} \tilde{\varepsilon}_{i,sr} \\ &\equiv M_{1,1}^{(2,1)} - 2M_{1,1}^{(2,2)} + M_{1,1}^{(2,3)}, \quad \text{say.} \end{aligned}$$

We will show that  $M_{1,1}^{(2,1)}$  is  $O_P(1)$  and contributes to the asymptotic variance and similar analyses show that  $M_{1,1}^{(2,2)}$  and  $M_{1,1}^{(2,3)}$  are both  $O_P(h^{1/2})$ . Let  $\xi_{h,sr} \equiv F_r' H_0 H_0' \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* F_t F_t' H_0 H_0' F_s$ . Note that  $\xi_{h,sr}$  is  $\mathcal{C}$ -measurable and

$$M_{1,1}^{(2,1)} = \sum_{i=1}^N \frac{h^{1/2}}{N^{1/2} T} \sum_{s,r=1}^T \xi_{h,sr} \tilde{\varepsilon}_{i,sr} \equiv \sum_{i=1}^N Z_{NT,i},$$

where  $Z_{NT,i} = \frac{h^{1/2}}{N^{1/2} T} \sum_{s,r=1}^T \xi_{h,sr} \tilde{\varepsilon}_{i,sr}$ . By the CLT for independent but non-identically distributed variables (conditional on  $\mathcal{C}$ ), it suffices to prove  $\mathbb{V}_{NT}^{-1/2} M_{1,1}^{(2,1)} \xrightarrow{d} N(0, 1)$  by showing that

$$\mathcal{Z} \equiv \sum_{i=1}^N E_C(Z_{NT,i}^4) = o_P(1) \quad \text{and} \quad \sum_{i=1}^N Z_{NT,i}^2 - \mathbb{V}_{NT}^{(1)} = o_P(1). \quad (\text{B.2})$$

First, we verify the first part of (B.2). Note that

$$\mathcal{Z} \leq \frac{16h^2}{N^2T^4} \sum_{i=1}^N E_C \left[ \sum_{s=1}^T \xi_{h,ss} \tilde{\varepsilon}_{i,ss} \right]^4 + \frac{256h^2}{N^2T^4} \sum_{i=1}^N E_C \left[ \sum_{1 \leq s < r \leq T} \xi_{h,sr} \tilde{\varepsilon}_{i,sr} \right]^4 \equiv 16\mathcal{Z}_1 + 256\mathcal{Z}_2.$$

For  $\mathcal{Z}_1$ , we have

$$\begin{aligned} \mathcal{Z}_1 &= \frac{h^2}{N^2T^4} \sum_{i=1}^N \sum_{s_l \in [T], l=1:4} \xi_{h,s_1s_1} \xi_{h,s_2s_2} \xi_{h,s_3s_3} \xi_{h,s_4s_4} E_C(\tilde{\varepsilon}_{i,s_1s_1} \tilde{\varepsilon}_{i,s_2s_2} \tilde{\varepsilon}_{i,s_3s_3} \tilde{\varepsilon}_{i,s_4s_4}) \\ &\lesssim \frac{h^2}{N^2T^4} \sum_{i=1}^N \left[ \sum_{s=1}^T \xi_{h,ss}^4 E_C(\tilde{\varepsilon}_{i,ss}^4) + \sum_{1 \leq s_1 \neq s_2 \leq T} \xi_{h,s_1s_1}^2 \xi_{h,s_2s_2}^2 E_C(\tilde{\varepsilon}_{i,s_1s_1}^2 \tilde{\varepsilon}_{i,s_2s_2}^2) \right. \\ &+ \sum_{1 \leq s_1 \neq s_2 \leq T} \xi_{h,s_1s_1}^3 \xi_{h,s_2s_2} E_C(\tilde{\varepsilon}_{i,s_1s_1}^3 \tilde{\varepsilon}_{i,s_2s_2}) + \sum_{1 \leq s_1 \neq s_2 \neq s_3 \leq T} \xi_{h,s_1s_1}^2 \xi_{h,s_2s_2} \xi_{h,s_3s_3} E_C(\tilde{\varepsilon}_{i,s_1s_1}^2 \tilde{\varepsilon}_{i,s_2s_2} \tilde{\varepsilon}_{i,s_3s_3}) \\ &\left. + \sum_{1 \leq s_1 \neq s_2 \neq s_3 \neq s_4 \leq T} \xi_{h,s_1s_1} \xi_{h,s_2s_2} \xi_{h,s_3s_3} \xi_{h,s_4s_4} E_C(\tilde{\varepsilon}_{i,s_1s_1} \tilde{\varepsilon}_{i,s_2s_2} \tilde{\varepsilon}_{i,s_3s_3} \tilde{\varepsilon}_{i,s_4s_4}) \right] \\ &\equiv \mathcal{Z}_{1,1} + \mathcal{Z}_{1,2} + \mathcal{Z}_{1,3} + \mathcal{Z}_{1,4} + \mathcal{Z}_{1,5}. \end{aligned}$$

By straightforward moment calculations, we can show that  $\mathcal{Z}_{1,1} = O_P(N^{-1}T^{-3}h^{-1})$ ,  $\mathcal{Z}_{1,2} = O_P(N^{-1}T^{-2})$ ,  $\mathcal{Z}_{1,3} = O_P(N^{-1}T^{-2})$ ,  $\mathcal{Z}_{1,4} = O_P(N^{-1}T^{-1}h)$  and  $\mathcal{Z}_{1,5} = O_P(N^{-1}h^2)$ . It follows that  $\mathcal{Z}_1 = o_P(1)$ . Next,

$$\begin{aligned} \mathcal{Z}_2 &= \frac{h^2}{N^2T^4} \sum_{i=1}^N E_C \left[ \sum_{1 \leq s < r \leq T} \xi_{h,sr} \tilde{\varepsilon}_{i,sr} \right]^4 \\ &\leq \frac{16h^2}{N^2T^4} \sum_{i=1}^N E_C \left[ \sum_{1 \leq s < r \leq T} \xi_{h,sr} \varepsilon_{i,sr} \right]^4 + \frac{16h^2}{N^2T^4} \sum_{i=1}^N \left[ \sum_{1 \leq s < r \leq T} \xi_{h,sr} E_C(\varepsilon_{i,sr}) \right]^4 \equiv 16(\mathcal{Z}_{2,1} + \mathcal{Z}_{2,2}). \end{aligned}$$

We first consider  $\mathcal{Z}_{2,1} = \frac{h^2}{N^2T^4} \sum_{i=1}^N \sum_{s_l, r_l \in [T], s_l < r_l, l=1:4} \xi_{h,s_1r_1} \xi_{h,s_2r_2} \xi_{h,s_3r_3} \xi_{h,s_4r_4} E_C(\varepsilon_{i,s_1r_1} \varepsilon_{i,s_2r_2} \varepsilon_{i,s_3r_3} \varepsilon_{i,s_4r_4})$ . For the summation over  $s_l$  and  $r_l$ ,  $l \in [4]$ , in the last expression, we consider seven cases depending on the cardinality ( $\#$ ) of the set  $\mathcal{S} \equiv \{s_1, r_1, \dots, s_4, r_4\}$ :  $\#\mathcal{S} = 8, 7, 6, 5, 4, 3, 2$ , and write  $\mathcal{Z}_2 = \sum_{l=1}^7 \mathcal{Z}_{2,l}$ , where

$$\mathcal{Z}_{2,1l} = \frac{h^2}{N^2T^4} \sum_{i=1}^N \sum_{s_r, r_r \in [T], s_r < r_r, r=1:4, \#\mathcal{S}=9-l} \xi_{h,s_1r_1} \xi_{h,s_2r_2} \xi_{h,s_3r_3} \xi_{h,s_4r_4} E_C(\varepsilon_{i,s_1r_1} \varepsilon_{i,s_2r_2} \varepsilon_{i,s_3r_3} \varepsilon_{i,s_4r_4}) \text{ for } l \in [7].$$

We will show  $\mathcal{Z}_{2,11} = o_P(1)$  as the other cases are similar or simpler. Now, we can rewrite  $\mathcal{Z}_{2,11}$  as follows

$$\mathcal{Z}_{2,11} = \frac{h^2}{N^2T^4} \sum_{i=1}^N \sum_{s_1, \dots, s_8 \in [T], s_1, \dots, s_8 \text{ are distinct}} \xi_{h,s_1s_2} \xi_{h,s_3s_4} \xi_{h,s_5s_6} \xi_{h,s_7s_8} E_C(\varepsilon_{i,s_1} \varepsilon_{i,s_2} \dots \varepsilon_{i,s_8}).$$

Let  $1 \leq r_1 < r_2 < \dots < r_8 \leq T$  be the permutation of  $s_1, s_2, \dots, s_8$  in descending order and let  $d_c$  be the  $c$ -th largest difference among  $r_{j+1} - r_j$ ,  $j \in [7]$ . For notational simplicity, let  $L_i(r_1, \dots, r_8) = \xi_{h,s_1s_2} \xi_{h,s_3s_4} \xi_{h,s_5s_6} \xi_{h,s_7s_8} \times E_C(\varepsilon_{i,s_1} \varepsilon_{i,s_2} \dots \varepsilon_{i,s_8})$ . Note that  $\xi_{h,sr} \lesssim \|F_r\| \|F_s\| h^{-1} k^{**}(\frac{s-r}{h})$ , where  $k^{**}(\cdot)$  is the 2-fold convolution kernel of

$k^*(\cdot)$ . Then by Lemma B.1 (with  $\tilde{\eta} = \eta/2$ )

$$\begin{aligned}
& \frac{h^2}{N^2 T^4} \sum_{i=1}^N \sum_{1 \leq r_1 < \dots < r_8 \leq T, r_2 - r_1 = d_1} |L_i(r_1, \dots, r_8)| \\
& \leq \frac{4M_{NT} h^2}{N^2 T^4} \sum_{i=1}^N \sum_{r_1=1}^{T-7} \sum_{r_2=r_1+\max_{j \geq 3} \{r_j - r_{j-1}\}}^{T-6} \sum_{r_3=k_2+1}^{T-5} \dots \sum_{r_8=r_7+1}^T |\xi_{h,s_1 s_2} \xi_{h,s_3 s_4} \xi_{h,s_5 s_6} \xi_{h,s_7 s_8} [\alpha_C(r_2 - r_1)]^{\frac{\eta}{2+\eta}}| \\
& \lesssim \max_r \|F_r\|^7 h^{-4} \frac{4M_{NT} h^2}{N T^4} \sum_{r_1=1}^{T-7} \sum_{r_2=r_1+\max_{j \geq 3} \{r_j - r_{j-1}\}}^{T-6} \|F_{r_1}\| (r_2 - r_1)^6 [\alpha_C(r_2 - r_1)]^{\frac{\eta}{2+\eta}} \\
& \lesssim \frac{T^{7/(8+4\eta)} h^{-2}}{N T^3} \sum_{r=1}^{\infty} r^6 [\alpha_C(r)]^{\frac{\eta}{2+\eta}} = O_P(N^{-1} T^{-3+7/(8+4\eta)} h^{-2}) = o_P(1),
\end{aligned}$$

where  $M_{NT} = [\max_{i,t} E_C(|\varepsilon_{it}|^{8+4\eta})]^{\eta/(2+\eta)} = O_P(1)$ . Similarly,  $\frac{h^2}{N^2 T^4} \sum_{i=1}^N \sum_{1 \leq r_1 < \dots < r_8 \leq T, r_8 - r_7 = d_1} |L_i(r_1, \dots, r_8)| = o_P(1)$ . If for some  $l_\alpha$  with  $2 \leq l_\alpha \leq 6$  and  $1 \leq \alpha \leq 4$ ,  $r_{l_\alpha+1} - r_{l_\alpha} = d_\alpha$ , then we can show that

$$\begin{aligned}
& \frac{h^2}{N^2 T^4} \sum_{i=1}^N \sum_{1 \leq r_1 < \dots < r_8 \leq T, r_{l_\alpha+1} - r_{l_\alpha} = d_\alpha, 2 \leq l_\alpha \leq 6, 1 \leq \alpha \leq 4} |L_i(r_1, \dots, r_8)| \\
& \lesssim \frac{\max_r \|F_r\|^6 h^{-1}}{N T^2} \sum_{r=1}^{\infty} r^5 \alpha^{\frac{\eta}{2+\eta}}(r) + \frac{\max_r \|F_r\|^5}{N T} \sum_{r=1}^{\infty} r^4 \alpha^{\frac{\eta}{2+\eta}}(r) + \frac{\max_r \|F_r\|^4 h}{N} \sum_{r=1}^{\infty} r^3 \alpha^{\frac{\eta}{2+\eta}}(r) \\
& = O_P(N^{-1} T^{-2+6/(8+4\eta)} h^{-1}) + O_P(N^{-1} T^{-1+5/(8+4\eta)}) + O_P(N^{-1} T^{1/(2+\eta)} h) = o_P(1).
\end{aligned}$$

In sum,  $\mathcal{Z}_{2,11} = o_P(1)$ . Analogously,  $\mathcal{Z}_{2,l} = o_P(1)$  for  $l = 2, \dots, 7$ . It follows that  $\mathcal{Z}_{2,1} = o_P(1)$ . For  $\mathcal{Z}_{2,2}$ , in view of the fact that  $\xi_{h,sr} \equiv F_r' H_0 H_0' \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* F_t F_t' H_0 H_0' F_s$  and  $\bar{k}_{h,sr}^* = \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^*$

$$\begin{aligned}
& \frac{1}{T} \left| \sum_{1 \leq s < r \leq T} \xi_{h,sr} E_C(\varepsilon_{i,sr}) \right| \\
& \asymp \frac{1}{T} \left| \sum_{1 \leq s < r \leq T} F_r' H_0 H_0' \Sigma_F H_0 H_0' F_s \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* E_C(\varepsilon_{i,sr}) \right| \\
& = \frac{1}{T} \left| \sum_{1 \leq s < r \leq T} \bar{k}_{h,sr}^* F_r' H_0 H_0' \Sigma_F H_0 H_0' F_s E_C(\varepsilon_{i,sr}) \right| \lesssim \frac{1}{T} \sum_{s=1}^T \sum_{r=1}^T \bar{k}_{h,sr}^* (\|F_r\|^2 + \|F_s\|^2) |E_C(\varepsilon_{i,sr})| \\
& \lesssim \frac{1}{T} \sum_{r=1}^T \max_s \bar{k}_{h,sr}^* \|F_r\|^2 \sum_{s=1}^T |E_C(\varepsilon_{i,sr})| \lesssim \max_{i,r} \sum_{s=1}^T |E_C(\varepsilon_{is} \varepsilon_{ir})| \frac{1}{T} \sum_{r=1}^T \max_s \bar{k}_{h,sr}^* \|F_r\|^2 = O_P(1),
\end{aligned}$$

we have  $\mathcal{Z}_{2,2} = \frac{h^2}{N^2} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{1 \leq s < r \leq T} \xi_{h,sr} E_C(\varepsilon_{i,sr}) \right]^4 = O_P(N^{-1} h^2) = o_P(1)$ . Consequently,  $\mathcal{Z}_2 = o_P(1)$  and  $\mathcal{Z} = o_P(1)$ .

Now, we verify the second part of (B.2). It suffices to show that (I)  $\sum_{i=1}^N E_C(Z_{NT,i}^2) = \mathbb{V}_{NT} + o_P(1)$ , and (II)  $\text{Var}_C \left( \sum_{i=1}^N Z_{NT,i}^2 \right) = o_P(1)$  by conditional Chebyshev inequality. (I) follows because

$$\sum_{i=1}^N E_C(Z_{NT,i}^2) = \frac{h}{N T^2} \sum_{i=1}^N E_C \left( \sum_{s,r=1}^T \xi_{h,sr} \tilde{\varepsilon}_{i,sr} \right)^2$$



$$= \frac{h}{N} \sum_{i=1}^N E_{\mathcal{C}} \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^2 - \frac{h}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} E_{\mathcal{C}}(\varepsilon_{is} \varepsilon_{ir}) \right)^2 = \mathbb{V}_{NT}^{(1)} - o_P(1).$$

(II) follows because

$$\begin{aligned} \text{Var}_{\mathcal{C}} \left( \sum_{i=1}^N Z_{NT,i}^2 \right) &= \sum_{i=1}^N \text{Var}_{\mathcal{C}} (Z_{NT,i}^2) = \frac{h^2}{N^2 T^4} \sum_{i=1}^N E \left[ \text{Var}_{\mathcal{C}} \left( \sum_{s,r=1}^T \xi_{h,sr} \tilde{\varepsilon}_{i,sr} \right)^2 \right] \leq \frac{h^2}{N^2 T^4} \sum_{i=1}^N E_{\mathcal{C}} \left| \sum_{s,r=1}^T \xi_{h,sr} \tilde{\varepsilon}_{i,sr} \right|^4 \\ &= o_P(1). \end{aligned}$$

In sum, we have shown that  $M_{1,1}^{(2)} - \mathbb{B}_{NT}^{(1)} \xrightarrow{d} N(0, \mathbb{V}_0)$  and  $M_1 - \mathbb{B}_{NT}^{(1)} - \Pi_1^{(1)} \xrightarrow{d} N(0, \mathbb{V}_0^{(1)})$ .

Now, we show (i2) and (i3). Note that

$$d_{4it} = X'_{it}(\hat{\beta}_t^{bc} - \tilde{\beta}) = X'_{it}(\hat{\beta}_t^{bc} - \beta_t) - X'_{it}(\tilde{\beta} - \beta_0) + X'_{it}(\beta_t - \beta_0).$$

Apparently,  $X'_{it}(\beta_t - \beta_0) = 0$  under  $\mathbb{H}_0^{(1)}$ , and  $X'_{it}(\beta_t - \beta_0) = a_{1NT} X'_{it} g_{0t}$  under  $\mathbb{H}_A^{(1)}(a_{1NT})$ . Then, we can further decompose  $M_7$  as follows

$$\begin{aligned} M_7 &= N^{-1/2} h^{1/2} \sum_{i=1}^N \sum_{t=1}^T d_{4it}^2 \\ &= N^{-1/2} h^{1/2} \sum_{i=1}^N \sum_{t=1}^T \left\{ [X'_{it}(\hat{\beta}_t^{bc} - \beta_t)]^2 + [X'_{it}(\tilde{\beta} - \beta_0)]^2 + [X'_{it}(\beta_t - \beta_0)]^2 \right. \\ &\quad \left. - 2X'_{it}(\hat{\beta}_t^{bc} - \beta_t)(\tilde{\beta} - \beta_0)' X_{it} + 2X'_{it}(\hat{\beta}_t^{bc} - \beta_t)(\beta_t - \beta_0)' X_{it} - 2X'_{it}(\tilde{\beta} - \beta_0)(\beta_t - \beta_0)' X_{it} \right\} \\ &\equiv M_{7,1} + M_{7,2} + M_{7,3} - 2M_{7,4} + 2M_{7,5} - 2M_{7,6}. \end{aligned}$$

By Theorem 3.4(ii), we have  $\max_t \|\hat{\beta}_t^{bc} - \beta_t\| = O_P((NT h / \ln T)^{-1/2})$ . Following the proof of Lemma A.1 in Su and Chen (2013), we can show that under  $\mathbb{H}_A^{(1)}(a_{1NT})$ ,

$$\tilde{\beta} - \beta_0 = a_{1NT} D(F)^{-1} \pi_{NT} + o_P(a_{1NT}), \quad (\text{B.3})$$

where  $D(F)$  is defined in Section 4.4. Let  $\pi_{NT} = (\pi_{NT,1}, \dots, \pi_{NT,P})'$ ,  $\pi_{NT,p} = \frac{1}{NT} \text{tr}(M_F \mathbf{X}_p M_{\Lambda_0} \Delta'_0)$  for  $p = 1, \dots, P$ , and  $\Delta_0$  is a  $T \times N$  matrix with  $(t, i)$ th element given by  $X'_{it} g_{0t}$ . With these results, we can readily show that  $M_{7,1} = O_P((Nh)^{-1/2} \ln T) = o_P(1)$ ,

$$\begin{aligned} M_{7,2} &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left[ X'_{it} D(F)^{-1} \pi_{NT} \right]^2 + o_P(1), \quad M_{7,3} = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T [X'_{it} g_{0t}]^2, \quad \text{and} \\ M_{7,6} &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T X'_{it} D(F)^{-1} \pi_{NT} g'_{0t} X_{it} + o_P(1). \end{aligned}$$

Then  $M_{7,4} \leq \{M_{7,1} M_{7,2}\}^{1/2} = o_P(1)$  and  $M_{7,5} \leq \{M_{7,1} M_{7,3}\}^{1/2} = o_P(1)$  by CS inequality. Consequently, we have

$$\begin{aligned} M_7 &= M_{7,2} + M_{7,3} - 2M_{7,6} + o_P(1) \\ &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left[ X'_{it} (g_{0t} - D(F)^{-1} \pi_{NT}) \right]^2 + o_P(1) = \Pi_3^{(1)} + o_P(1). \end{aligned}$$

Similarly, we can show that  $M_l = o_P(1)$  for  $l = 8$  and  $9$ . For  $M_{10}$ , we have  $M_{10} \leq \{M_3 M_4\}^{1/2} = o_P(1)$  by CS inequality.

In sum, we have shown that  $J_{NT}^{(1)} \equiv TN^{1/2}h^{1/2}\hat{M}^{(1)} - \mathbb{B}_{NT}^{(1)} - \Pi^{(1)} \xrightarrow{d} N(0, \mathbb{V}_0^{(1)})$ , where  $\Pi^{(1)} = \sum_{\ell=1}^3 \Pi_\ell^{(1)}$ .

(ii) As in the proof of (i), we prove that under  $\mathbb{H}_A^{(2)}(a_{2NT}) : \beta_t = \beta_0 + a_{2NT}g_{0t}$  with  $a_{2NT} = (NT)^{-1/2}h^{-1/4}$ , we have  $J_{NT}^{(2)} = TNh^{1/2}\hat{M}^{(2)} - \mathbb{B}_{NT}^{(2)} \xrightarrow{d} N(0, \mathbb{V}_0^{(2)})$ . First, we decompose  $TNh^{1/2}\hat{M}^{(2)}$  as follows:

$$\begin{aligned} TNh^{1/2}\hat{M}^{(2)} &= Nh^{1/2} \sum_{t=1}^T \left\| (\hat{\beta}_t^{bc} - \beta_t) - (\bar{\beta}^{bc} - \beta_t) \right\|^2 \\ &= Nh^{1/2} \sum_{t=1}^T \left\{ \left\| \hat{\beta}_t^{bc} - \beta_t \right\|^2 + \left\| \bar{\beta}^{bc} - \beta_t \right\|^2 - 2(\hat{\beta}_t^{bc} - \beta_t)'(\bar{\beta}^{bc} - \beta_t) \right\} \equiv A_1 + A_2 - 2A_3. \end{aligned}$$

Following the proofs of Theorems 3.2 and 3.4, we can argue that  $\hat{\beta}_t^{bc} - \beta_t = [D^{(t)}(F^{(t)})]^{-1} \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \mathfrak{X}_{is}^{(t)} \varepsilon_{is}^{(t)} + R_\beta(t)$  and  $R_\beta(t) = o_P(a_{2NT})$  uniformly in  $t$  under  $\mathbb{H}_A^{(2)}(a_{2NT})$  under the additional bandwidth condition  $NT h^{9/2} = o(1)$ . For  $A_1$ , we have

$$\begin{aligned} A_1 &= Nh^{1/2} \sum_{t=1}^T \left\{ \left\| [D^{(t)}(F^{(t)})]^{-1} \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \mathfrak{X}_{is}^{(t)} \varepsilon_{is}^{(t)} \right\|^2 + \|R_\beta(t)\|^2 + 2 \left[ [D^{(t)}(F^{(t)})]^{-1} \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \mathfrak{X}_{is}^{(t)} \varepsilon_{is}^{(t)} \right]' R_\beta(t) \right\} \\ &\equiv A_{1,1} + A_{1,2} + 2A_{1,3}. \end{aligned}$$

We will show that the first term is the leading term, while the second and third terms are higher order terms. Before that, we first analyze the  $\mathfrak{X}_{is}^{(t)}$  term in detail. Note that the  $(s, r)$ th element of  $M_{F^{(t)}}$  is

$$[M_{F^{(t)}}]_{s,r} = \mathbf{1}\{s=r\} - T^{-1}F_s^{(t)'}[T^{-1}F^{(t)'}F^{(t)}]^{-1}F_r^{(t)} = \mathbf{1}\{s=r\} - T^{-1}b_{sr}^{(t)},$$

and the  $(i, j)$ th element of  $M_{\Lambda_t}$  is given by  $[M_{\Lambda_t}]_{i,j} = \mathbf{1}\{i=j\} - N^{-1}a_{ij}^{(t)}$ , where  $a_{ij}^{(t)} = \lambda_{it}'(\Lambda_t' \Lambda_t / N)^{-1} \lambda_{jt}$  and  $b_{sr}^{(t)} = F_s^{(t)'}(T^{-1} \sum_{l=1}^T F_l^{(t)'} F_l^{(t)})^{-1} F_r^{(t)}$ . Let  $\bar{b}_{sr}^{(t)} = F_s'[T^{-1} \sum_{l=1}^T k_{h,lt}^* F_l F_l']^{-1} F_r$ . The  $(s, i)$ th element of the  $T \times N$  matrix  $\mathfrak{X}^{(t)}$  is given by

$$\begin{aligned} \mathfrak{X}_{is}^{(t)} &= \sum_{j=1}^N \sum_{r=1}^T k_{h,rt}^{*1/2} [M_{F^{(t)}}]_{s,r} E_C(X_{jr}) [M_{\Lambda_t}]_{j,i} + [X_{is}^{(t)} - E_C(X_{is}^{(t)})] \\ &= \sum_{j=1}^N \sum_{r=1}^T k_{h,rt}^{*1/2} \left[ \mathbf{1}\{s=r\} - T^{-1}b_{sr}^{(t)} \right] E_C(X_{jr}) \left[ \mathbf{1}\{i=j\} - N^{-1}a_{ij}^{(t)} \right] + [X_{is}^{(t)} - E_C(X_{is}^{(t)})] \\ &= X_{is}^{(t)} - \frac{1}{N} \sum_{j=1}^N k_{h,st}^{*1/2} E_C(X_{jr}) a_{ij}^{(t)} - \frac{1}{T} \sum_{r=1}^T k_{h,rt}^{*1/2} b_{sr}^{(t)} E_C(X_{jr}) + \frac{1}{NT} \sum_{j=1}^N \sum_{r=1}^T k_{h,rt}^{*1/2} b_{sr}^{(t)} E_C(X_{jr}) a_{ij}^{(t)} \\ &= k_{h,st}^{*1/2} \left[ X_{is} - \frac{1}{N} \sum_{j=1}^N E_C(X_{js}) a_{ji}^{(t)} - \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* \bar{b}_{sr}^{(t)} E_C(X_{jr}) + \frac{1}{NT} \sum_{r=1}^T \sum_{j=1}^N k_{h,rt}^* \bar{b}_{sr}^{(t)} a_{ji}^{(t)} E_C(X_{jr}) \right] \\ &\equiv k_{h,st}^{*1/2} \mathcal{X}_{is}^{(t)}. \end{aligned}$$

For  $A_{1,1}$ , we can readily show that

$$A_{1,1} = Nh^{1/2} \sum_{t=1}^T \frac{1}{N^2 T^2} \sum_{i,j=1}^N \sum_{s,r=1}^T \varepsilon_{is}^{(t)} \varepsilon_{jr}^{(t)} \mathfrak{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathfrak{X}_{jr}^{(t)}$$

$$\begin{aligned}
&= \frac{h^{1/2}}{NT^2} \sum_{t=1}^T \sum_{i=1}^N \sum_{s,r=1}^T k_{h,st}^* k_{h,rt}^* \varepsilon_{is} \varepsilon_{ir} \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathfrak{X}_{ir}^{(t)} + \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \varepsilon_{is} \varepsilon_{js} \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)} \\
&+ \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s \neq r} k_{h,st}^* k_{h,rt}^* \varepsilon_{is} \varepsilon_{jr} \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{jr}^{(t)} \equiv \sum_{\ell=1}^3 A_{1,1\ell},
\end{aligned}$$

where  $\mathbb{D}^{(t)} = [D^{(t)}(F^{(t)})]^{-1} [D^{(t)}(F^{(t)})]^{-1}$  and  $\sum_{s \neq r} = \sum_{1 \leq s \neq r \leq T}$ . We will show that  $A_{1,11}$  and  $A_{1,12}$  contribute to the asymptotic bias and  $A_{1,13}$  contributes the asymptotic variance. Apparently,

$$\begin{aligned}
A_{1,11} &= \frac{h^{1/2}}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s,r=1}^T k_{h,st}^* k_{h,rt}^* \varepsilon_{is} \varepsilon_{ir} \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{ir}^{(t)} = \mathbb{B}_{1,NT}^{(2)} \text{ and} \\
A_{1,12} &= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \varepsilon_{is} \varepsilon_{js} \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)} = \mathbb{B}_{2,NT}^{(2)},
\end{aligned}$$

both of which are of  $O_P(h^{-1/2})$ . Let  $\xi_{it} = (\varepsilon_{it}, X'_{it})'$  and  $\xi_i = (\xi_{i1}, \dots, \xi_{iT})'$ . Let  $\varsigma_{ij,sr} = \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{jr}^{(t)}$  and  $W_{ij} \equiv W(\xi_i, \xi_j) = \frac{2h^{1/2}}{NT} \sum_{s,r=1}^T \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}$ . Then  $A_{1,1}^{(b)} = \sum_{1 \leq i < j \leq N} W_{ij}$ . Define

$$\begin{aligned}
G_1 &= \sum_{1 \leq i < j \leq N} E_C(W_{ij}^4), \quad G_2 = \sum_{1 \leq i < j < k \leq N} E_C(W_{ij}^2 W_{ik}^2 + W_{ji}^2 W_{jk}^2 + W_{ki}^2 W_{kj}^2), \text{ and} \\
G_3 &= \sum_{1 \leq i < j < k < l \leq N} E_C(W_{ij} W_{ik} W_{lj} W_{lk} + W_{ij} W_{il} W_{kj} W_{kl} + W_{ik} W_{il} W_{jk} W_{jl}).
\end{aligned}$$

Let  $\mathbb{V}_{NT}^{(2)} = \text{Var}_C(A_{1,1}^{(b)})$ . Note that  $\mathbb{V}_{NT}^{(2)} = \sum_{1 \leq i < j \leq N} E_C(W_{ij}^2) = \frac{4h}{N^2 T^2} \sum_{1 \leq i < j \leq N} E_C\left(\sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}\right)^2 \xrightarrow{P} \mathbb{V}_0^{(2)} > 0$ . By straightforward moment calculations and the repeated use of Lemma B.1, we can show that

$$\begin{aligned}
G_1 &= \frac{16h^2}{N^4 T^4} \sum_{1 \leq i < j \leq N} E_C\left(\sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}\right)^4 = O_P(N^{-2} h^{-2}), \\
G_2 &\lesssim \frac{16h^2}{N^4 T^4} \sum_{i,j,k \text{ are all distinct}} E_C\left[\left(\sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}\right)^2 \left(\sum_{s \neq r} \varepsilon_{is} \varepsilon_{kr} \varsigma_{ik,sr}\right)^2\right] = O_P(N^{-1}),
\end{aligned}$$

and similarly,  $G_3 = O_P(h^2)$ . Consequently,  $G_l = o_P((\mathbb{V}_{NT}^{(2)})^2)$  for  $l = 1, 2, 3$ . Then by a conditional version of the CLT in Proposition 3.2 in de Jong (1987), we have  $(\mathbb{V}_{NT}^{(2)})^{-1/2} A_{1,1}^{(b)} \xrightarrow{d} N(0, 1)$  conditional on  $\mathcal{C}$ . The unconditional limiting law is also given by  $N(0, 1)$ . Then  $A_{1,1} - \mathbb{B}_{NT}^{(2)} \xrightarrow{d} N(0, \mathbb{V}_0^{(2)})$ .

Under the additional bandwidth condition that  $NT h^{9/2} = o(1)$ , we have that  $A_{1,2} = NT h^{1/2} o_P(a_{2NT}^2) = o_P(1)$ . With some tedious calculations, we can show that  $A_{1,3} = 2Nh^{1/2} \sum_{t=1}^T \left[ D(F^{(t)})^{-1} \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \mathfrak{X}_{is}^{(t)} \varepsilon_{is}^{(t)} \right]' R_\beta(t) = o_P(1)$ . Then  $A_1 - \mathbb{B}_{NT}^{(2)} \xrightarrow{d} N(0, \mathbb{V}_0^{(2)})$ .

Now, we consider the  $A_2$  term under the local alternative  $\mathbb{H}_A^{(2)}(a_{2NT})$  with  $a_{2NT} = (NT)^{-1/2} h^{-1/4}$ :

$$A_2 = Nh^{1/2} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{s=1}^T \hat{\beta}_s^{bc} - \beta_t \right\|^2 = Nh^{1/2} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{s=1}^T (\hat{\beta}_s^{bc} - \beta_s) - \frac{1}{T} \sum_{s=1}^T a_{2NT} (g_{0t} - g_{0s}) \right\|^2$$

$$\begin{aligned}
&= Nh^{1/2} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{s=1}^T (\hat{\beta}_s^{bc} - \beta_s) \right\|^2 + Nh^{1/2} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{s=1}^T a_{2NT} (g_{0t} - g_{0s}) \right\|^2 \\
&\quad - 2T^{-2} Nh^{1/2} \sum_{t=1}^T \sum_{s=1}^T \sum_{r=1}^T a_{2NT} (\hat{\beta}_s^{bc} - \beta_s)' (g_{0t} - g_{0r}) \equiv A_{2,1} + A_{2,2} - 2A_{2,3}.
\end{aligned}$$

Under  $\mathbb{H}_A^{(2)}(a_{2NT})$ , we can follow the proofs of Theorems 3.2 and 3.4 and show that  $\frac{1}{T} \sum_{s=1}^T (\hat{\beta}_s^{bc} - \beta_s) = O_P((NT)^{-1/2} + h^2)$ . Then  $A_{2,1} = NT h^{1/2} O_P((NT)^{-1} + h^4) = O_P(h^{1/2} + NT h^9/2) = o_P(1)$ . For  $A_{2,2}$ , we have

$$\begin{aligned}
A_{2,2} &= \frac{1}{T^3} \sum_{t=1}^T \sum_{s=1}^T \sum_{r=1}^T (g_{0t} - g_{0s})' (g_{0t} - g_{0r}) = \frac{1}{T} \sum_{t=1}^T g'_{0t} g_{0t} + O(T^{-1}) \\
&= \int_0^1 \|g_0(\tau)\|^2 d\tau + O_P(T^{-1}) \equiv \Pi_2 + O(T^{-1}),
\end{aligned}$$

where the second equality holds because  $\frac{1}{T} \sum_{s=1}^T g_{0s} = \frac{1}{T} \sum_{s=1}^T g_0(\frac{s}{T}) = \int_0^1 g_0(\tau) d\tau + O(\frac{1}{T}) = O(\frac{1}{T})$  under our normalization restriction on  $g_0(\cdot)$ . By CS inequality, we have  $|A_{2,3}| \leq \{A_{2,1}\}^{1/2} \{A_{2,2}\}^{1/2} = o_P(1)$ . It follows that  $A_2 = \Pi_2 + o_P(1)$ .

Now, we show that  $A_3 = o_P(1)$ . Noting that  $\bar{\beta}^{bc} - \beta_t = \frac{1}{T} \sum_{s=1}^T (\hat{\beta}_s^{bc} - \beta_s) - \frac{1}{T} \sum_{s=1}^T a_{2NT} (g_{0t} - g_{0s})$ , we can decompose  $A_3$  as follows:

$$A_3 = Nh^{1/2} \sum_{t=1}^T (\hat{\beta}_t^{bc} - \beta_t)' \frac{1}{T} \sum_{s=1}^T (\hat{\beta}_s^{bc} - \beta_s) - Nh^{1/2} \sum_{t=1}^T (\hat{\beta}_t^{bc} - \beta_t)' \frac{1}{T} \sum_{s=1}^T a_{2NT} (g_{0t} - g_s) \equiv A_{3,1} - A_{3,2}.$$

$A_{3,1}$  is the same as  $A_{2,1}$ , which is  $o_P(1)$ . For  $A_{3,2}$ , we have

$$\begin{aligned}
A_{3,2} &= (NT)^{1/2} h^{1/4} \frac{1}{T} \sum_{t=1}^T (\hat{\beta}_t^{bc} - \beta_t) \frac{1}{T} \sum_{s=1}^T (g_{0t} - g_{0s}) \\
&= (NT)^{1/2} h^{1/4} \frac{1}{T} \sum_{t=1}^T (\hat{\beta}_t^{bc} - \beta_t)' g_{0t} - (NT)^{1/2} h^{1/4} \frac{1}{T} \sum_{t=1}^T (\hat{\beta}_t^{bc} - \beta_t) O(T^{-1}).
\end{aligned}$$

Noting that  $g_{0t}$  is nonrandom, we can follow the proofs of Theorems 3.2 and 3.4 and show that  $\frac{1}{T} \sum_{t=1}^T (\hat{\beta}_t^{bc} - \beta_t)' g_{0t} = O_P((NT)^{-1/2} + h^2)$ . Then  $A_{3,2} = O_P(h^{1/4} + (NT)^{1/2} h^9/4) = o_P(1)$  and  $A_3 = o_P(1)$ . In sum, we have shown that  $TNh^{1/2} \hat{M}^{(2)} - \mathbb{B}_{NT}^{(2)} - \Pi_2 \xrightarrow{d} N(0, \mathbb{V}_0^{(2)})$ , where  $\Pi_2 = 0$  under  $\mathbb{H}_0^{(2)}$  and  $\Pi_2 = \int_0^1 \|g_0(\tau)\|^2 d\tau$  under  $\mathbb{H}_A^{(2)}(a_{2NT})$ .

(iii) As above, we show the asymptotic distribution of  $TN^{1/2} h^{1/2} \hat{M}^{(3)}$  under  $\mathbb{H}_A^{(3)}(a_{1NT})$  with  $a_{1NT} = N^{-1/4} T^{-1/2} h^{-1/4}$ . As in the proof of part (i), we can make the following decomposition:

$$\hat{\lambda}_{it}^{W'} \hat{F}_t^W - \tilde{\lambda}_{i0}^{W'} \tilde{F}_t^W = d_{lit}^W + d_{2it}^W + d_{3it}^W,$$

where for  $l = 1, 2, 3$ ,  $d_{lit}^W$  is defined analogously to  $d_{lit}$  with  $\hat{\lambda}_{it}$ ,  $\hat{F}_t$ ,  $\tilde{\lambda}_{i0}$  and  $\tilde{F}_t$  replaced by  $\hat{\lambda}_{it}^W$ ,  $\hat{F}_t^W$ ,  $\tilde{\lambda}_{i0}^W$  and  $\tilde{F}_t^W$ , respectively. Given the fast convergence rate of  $\hat{\beta}_t^{bc} - \beta_t$  under our bandwidth condition, the error in the estimation of  $\beta_t$  is asymptotically negligible so that we can readily show that under  $\mathbb{H}_A^{(3)}(a_{1NT})$ ,

$$TN^{1/2} h^{1/2} \hat{M}^{(3)} = M_1 + M_2 + o_P(1),$$

where  $M_1 - \mathbb{B}_{NT}^{(1)} - \Pi_1^{(1)} \xrightarrow{d} N(0, \mathbb{V}_0^{(3)})$ ,  $M_2 = \Pi_1^{(2)} + o_P(1)$ , and  $\mathbb{V}_0^{(3)} = \mathbb{V}_0^{(1)}$ . Consequently, we have

$$TN^{1/2}h^{1/2}\hat{M}^{(3)} - \mathbb{B}_{NT}^{(3)} - \Pi_3 \xrightarrow{d} N(0, \mathbb{V}_0^{(3)}),$$

where  $\mathbb{B}_{NT}^{(3)} = \mathbb{B}_{NT}^{(1)}$  and  $\Pi_3 = \Pi_1^{(1)} + \Pi_1^{(2)}$ . This completes the proof of the theorem.  $\blacksquare$

**Proof of Theorem 4.2.** (i) By Theorem 4.1(i), it suffices to show (i1)  $\hat{\mathbb{B}}_{NT}^{(1)} - \mathbb{B}_{NT}^{(1)} = o_P(1)$  and (i2)  $\hat{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)} = o_P(1)$ . For the use in the proof of Theorem 4.3(i), we prove the claims under  $\mathbb{H}_1^{(1)}(a_{1NT})$ . Recall that

$$\begin{aligned} \hat{L}_{st} &= k_{h, st}^* \hat{F}'_s \hat{F}_t - \tilde{F}'_s \tilde{F}_t, \quad \hat{\Gamma}_{it, j} = T^{-1} \sum_{s=j+1}^T \hat{L}_{st} \hat{L}_{s-j, t} \varepsilon_{is} \varepsilon_{i, s-j}, \quad \text{and} \quad \hat{\Xi}_{it} = \hat{\Gamma}_{it, 0} + 2 \sum_{j=1}^{l_T} w_{Tj} \hat{\Gamma}_{it, j}. \\ \hat{\mathbb{B}}_{NT}^{(1)} &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T \hat{\eta}_{it, s}^2 + \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=1}^{l_T} w_{Tj} \sum_{s=j+1}^T \hat{\eta}_{it, s} \hat{\eta}_{it, s-j}. \end{aligned}$$

Let  $\eta_{it, s} = F'_t L_{st} F_s \varepsilon_{is}$  and  $\hat{\eta}_{it, s} = (k_{h, st}^* \hat{F}'_s \hat{F}_t - \tilde{F}'_s \tilde{F}_t) \varepsilon_{is}$  and  $\bar{\eta}_{it, s} = (k_{h, st}^* \tilde{F}'_s \tilde{F}_t - \tilde{F}'_s \tilde{F}_t) \varepsilon_{is}$ . Then

$$\begin{aligned} \mathbb{B}_{NT}^{(1)} &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T E_C \left( \sum_{s=1}^T \eta_{it, s} \right)^2 \\ &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T E_C(\eta_{it, s}^2) + \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=1}^{T-1} \sum_{s=j+1}^T E_C(\eta_{it, s} \eta_{it, s-j}). \end{aligned}$$

Let  $\bar{\mathbb{B}}_{NT}^{(1)} = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T \eta_{it, s}^2 + \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=1}^{l_T} w_{Tj} \sum_{s=j+1}^T \eta_{it, s} \eta_{it, s-j}$ . Then

$$\begin{aligned} \hat{\mathbb{B}}_{NT}^{(1)} - \mathbb{B}_{NT}^{(1)} &= \left[ \hat{\mathbb{B}}_{NT}^{(1)} - \bar{\mathbb{B}}_{NT}^{(1)} \right] + \left[ \bar{\mathbb{B}}_{NT}^{(1)} - \mathbb{B}_{NT}^{(1)} \right] \\ &= \left[ \hat{\mathbb{B}}_{NT}^{(1)} - \bar{\mathbb{B}}_{NT}^{(1)} \right] + \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T [\eta_{it, s}^2 - E_C(\eta_{it, s}^2)] \\ &\quad + \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=1}^{l_T} w_{Tj} \sum_{s=j+1}^T [\eta_{it, s} \eta_{it, s-j} - E_C(\eta_{it, s} \eta_{it, s-j})] \\ &\quad + \left[ \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \left\{ \sum_{s=1}^T [E_C(\eta_{it, s}^2) + 2 \sum_{j=1}^{l_T} w_{Tj} \sum_{s=j+1}^T E_C(\eta_{it, s} \eta_{it, s-j})] \right\} - \mathbb{B}_{NT}^{(1)} \right] \equiv \sum_{\ell=1}^4 B_\ell. \end{aligned}$$

It suffices to show that  $B_\ell = o_P(1)$  for  $\ell \in [4]$ .

We first show that  $B_1 = o_P(1)$ . Note that

$$\begin{aligned} B_1 &= \hat{\mathbb{B}}_{NT}^{(1)} - \bar{\mathbb{B}}_{NT}^{(1)} \\ &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\hat{\eta}_{it, s}^2 - \eta_{it, s}^2) + \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=1}^{l_T} w_{Tj} \sum_{s=j+1}^T (\hat{\eta}_{it, s} \hat{\eta}_{it, s-j} - \eta_{it, s} \eta_{it, s-j}) \\ &\equiv B_{1,1} + 2B_{1,2}. \end{aligned}$$

We will show that  $B_1 = o_P(1)$  by showing that  $B_{1,1} = o_P(l_T^{-1})$  and  $B_{1,2} = o_P(1)$ . Using  $a^2 - b^2 = (a-b)^2 + 2(a-b)b$ , we obtain

$$B_{1,1} = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\hat{\eta}_{it, s}^2 - \bar{\eta}_{it, s}^2) + \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\bar{\eta}_{it, s}^2 - \eta_{it, s}^2)$$

$$\begin{aligned}
&= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\hat{\eta}_{it,s} - \bar{\eta}_{it,s})^2 + \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\hat{\eta}_{it,s} - \bar{\eta}_{it,s}) \bar{\eta}_{it,s} \\
&\quad + \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\bar{\eta}_{it,s} - \eta_{it,s})^2 + \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\bar{\eta}_{it,s} - \eta_{it,s}) \eta_{it,s} \\
&\equiv B_{1,11} + 2B_{1,12} + B_{1,13} + 2B_{1,14}.
\end{aligned}$$

Using  $\hat{\eta}_{it,s} - \bar{\eta}_{it,s} = (k_{h,st}^* \hat{F}'_s \hat{F}_t - \tilde{F}'_s \tilde{F}_t) (\hat{\varepsilon}_{is} - \varepsilon_{is})$ ,  $\bar{\eta}_{it,s} - \eta_{it,s} = [(k_{h,st}^* \hat{F}'_s \hat{F}_t - \tilde{F}'_s \tilde{F}_t) - F'_t L_{st} F_s] \varepsilon_{is}$ , the expansion for  $\hat{\varepsilon}_{is} - \varepsilon_{is}$ , and the mean square convergence of  $\hat{F}_t$  and  $\tilde{F}_t$  under the local alternatives, one can readily show that  $B_{1,l} = o_P(l_T^{-1})$  for  $l \in [4]$ . For example, for  $B_{1,13}$ , noting that

$$\begin{aligned}
\bar{\eta}_{it,s} - \eta_{it,s} &= \left[ (k_{h,st}^* \hat{F}'_s \hat{F}_t - \tilde{F}'_s \tilde{F}_t) - F'_t \bar{L}_{st} F_s \right] \varepsilon_{is} \\
&= \left[ (k_{h,st}^* \hat{F}'_s \hat{F}_t - \tilde{F}'_s \tilde{F}_t) - F'_t L_{st} F_s \right] \varepsilon_{is} + F'_t (L_{st} - \bar{L}_{st}) F_s \varepsilon_{is} \\
&= k_{h,st}^* \left( \hat{F}'_s \hat{F}_t - F'_s H^{(t)} H^{(t)'} F_t \right) \varepsilon_{is} - (\tilde{F}'_s \tilde{F}_t - F'_s H H' F_t) \varepsilon_{is} \\
&\quad + k_{h,st}^* F'_t \left( H^{(t)} H^{(t)'} - H_0 H_0' \right) F_s \varepsilon_{is} - F'_t (H H' - H_0 H_0') F_s \varepsilon_{is} \equiv \sum_{l=1}^4 \eta_{it,s}(l),
\end{aligned}$$

we have

$$B_{1,13} = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (\bar{\eta}_{it,s} - \eta_{it,s})^2 \leq 4 \sum_{l=1}^4 \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T [\eta_{it,s}(l)]^2 \equiv 4 \sum_{l=1}^4 B_{1,13}(l).$$

Noting that  $\hat{F}'_t \hat{F}_s - F'_t H^{(t)} H^{(t)'} F_s = (\hat{F}_t - H^{(t)'} F_t)' (\hat{F}_s - H^{(t)'} F_s) + (\hat{F}_t - H^{(t)'} F_t)' H^{(t)'} F_s + F'_t H^{(t)} (\hat{F}_s - H^{(t)'} F_s) \equiv \sum_{l=1}^3 \varsigma_{l,ts}$

$$\begin{aligned}
B_{1,13}(1) &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} [\hat{F}'_s \hat{F}_t - F'_s H^{(t)} H^{(t)'} F_t]^2 \varepsilon_{is}^2 \leq 3 \sum_{l=1}^3 \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \varsigma_{st,l}^2 \varepsilon_{is}^2 \\
&\equiv \sum_{l=1}^3 \varsigma_l.
\end{aligned}$$

Using the fact that  $H^{(t)} = H + O_P(C_{0NT}^{-1})$  under  $\mathbb{H}_1^{(1)}(a_{1NT})$  and  $\max_t \frac{1}{T} \sum_{s=1}^T k_{h,st}^{*2} \left\| \hat{F}_s - H^{(t)'} F_s \right\|^2 = O(h^{-1} C_{NT}^{-2} \ln N)$ , we can readily show that  $\varsigma_1 = N^{1/2} h^{1/2} O_P(h^{-1} C_{NT}^{-4} \ln N) = o_P(l_T^{-1})$ , and  $\varsigma_l = N^{1/2} h^{1/2} \times O_P(h^{-1} C_{NT}^{-2} \ln N) = o_P(l_T^{-1})$  for  $l = 2, 3$ . Then  $B_{1,13}(1) = o_P(l_T^{-1})$ . By the same token,  $B_{1,13}(2) = o_P(l_T^{-1})$ . For  $B_{1,13}(3)$  and  $B_{1,13}(4)$ , we have

$$B_{1,13}(3) = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} [F'_t (H^{(t)} H^{(t)'} - H_0 H_0') F_s]^2 \varepsilon_{is}^2 = N^{1/2} h^{1/2} O_P(h^{-1} C_{NT}^{-2} \ln N) = o_P(l_T^{-1})$$

and

$$B_{1,13}(4) = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T [F'_t (H H' - H_0 H_0') F_s]^2 \varepsilon_{is}^2 = N^{1/2} h^{1/2} O_P(C_{0NT}^{-2}) = o_P(l_T^{-1}).$$

Consequently,  $B_{1,13} = o_P(l_T^{-1})$ . It follows that  $B_{1,1} = o_P(l_T^{-1})$ .

By the uniform boundedness of  $w_{Tj}$  and arguments similar to those used in the analysis of  $B_{1,1}$ , we can show that  $B_{1,2} = l_T o_P(l_T^{-1}) = o_P(1)$ . Then  $B_1 = o_P(1)$ .

Next, we consider  $B_2$  :

$$B_2 = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T [\eta_{it,s}^2 - E_C(\eta_{it,s}^2)] = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T (k_{h,st}^* - 1)^2 (F'_t F_s)^2 [\varepsilon_{is}^2 - E_C(\varepsilon_{is}^2)].$$

Note that  $E_C(B_2) = 0$  and

$$\begin{aligned} \text{Var}_C(B_2) &= \frac{h}{NT^4} \sum_{i=1}^N \text{Var}_C \left( \sum_{t=1}^T \sum_{s=1}^T (k_{h,st}^* - 1)^2 (F'_t F_s)^2 [\varepsilon_{is}^2 - E_C(\varepsilon_{is}^2)] \right) \\ &\leq \frac{h}{NT^4} \sum_{i=1}^N \text{Var}_C \left( \sum_{s=1}^T (k_{h,ss}^* - 1)^2 (F'_s F_s)^2 [\varepsilon_{is}^2 - E_C(\varepsilon_{is}^2)] \right) \\ &\quad + \frac{4h}{NT^4} \sum_{i=1}^N \text{Var}_C \left( \sum_{1 \leq t < s \leq T} (k_{h,st}^* - 1)^2 (F'_t F_s)^2 [\varepsilon_{is}^2 - E_C(\varepsilon_{is}^2)] \right). \end{aligned}$$

For the first term, we can readily obtain a rough bound  $O_P(T^{-2}h^{-3}) = o_P(1)$ ; for the second term, an application of Lemma B.1 yields the probability order  $O_P(T^{-2}h^{-1} + T^{-1}h) = o_P(1)$ . Thus  $B_2 = o_P(1)$ .

Next, we consider  $B_3$  :

$$B_3 = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=1}^{l_T} w_{Tj} \sum_{s=j+1}^T [\eta_{it,s} \eta_{it,s-j} - E_C(\eta_{it,s} \eta_{it,s-j})] = \sum_{j=1}^{l_T} w_{Tj} B_{3,j},$$

where  $B_{3,j} = \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=j+1}^T (k_{h,st}^* - 1)(k_{h,s-j,t}^* - 1) F'_t F_s F'_t F_{s-j} [\varepsilon_{is} \varepsilon_{i,s-j} - E_C(\varepsilon_{is} \varepsilon_{i,s-j})]$ . By moment calculations, we can show that  $E[B_{3,j}^2] = E\{E_C[B_{3,j}^2]\} \leq CT^{-1}$  for some constant  $C$ . Then by the union bound and Chebyshev inequality, we have that for any  $\epsilon > 0$

$$\begin{aligned} P(|B_3| \geq \epsilon) &= P\left(\left|\sum_{j=1}^{l_T} w_{Tj} B_{3,j}\right| \geq \epsilon\right) \leq P\left(\sum_{j=1}^{l_T} |w_{Tj}| |B_{3,j}| \geq \epsilon\right) \leq P\left(\sum_{j=1}^{l_T} |B_{3,j}| \geq \epsilon/c_w\right) \\ &\leq \sum_{j=1}^{l_T} P\left(\sum_{j=1}^{l_T} |B_{3,j}| \geq \epsilon/(c_w l_T)\right) \leq \frac{(c_w l_T)^2}{\epsilon^2} \sum_{j=1}^{l_T} E[B_{3,j}^2] = O(l_T^3 T^{-1}) = o(1). \end{aligned}$$

Then  $B_3 = o_P(1)$ .

For  $B_4$ , we have

$$\begin{aligned} B_4 &= \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \left[ \sum_{j=1}^{l_T} w_{Tj} \sum_{s=j+1}^T E_C(\eta_{it,s} \eta_{it,s-j}) - \sum_{j=1}^{T-1} \sum_{s=j+1}^T E_C(\eta_{it,s} \eta_{it,s-j}) \right] \\ &= \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=1}^{l_T} (w_{Tj} - 1) \sum_{s=j+1}^T E_C(\eta_{it,s} \eta_{it,s-j}) + \frac{2h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=l_T+1}^{T-1} \sum_{s=j+1}^T E_C(\eta_{it,s} \eta_{it,s-j}) \\ &\equiv 2B_{4,1} + 2B_{4,2}. \end{aligned}$$

For  $B_{4,2}$ , we have

$$E|B_{4,2}| = \frac{2h^{1/2}}{N^{1/2}T^2} E \left| \sum_{i=1}^N \sum_{t=1}^T \sum_{j=l_T+1}^{T-1} \sum_{s=j+1}^T (k_{h,st}^* - 1)(k_{h,s-j,t}^* - 1) F'_t F_s F'_t F_{s-j} E_C(\varepsilon_{is} \varepsilon_{i,s-j}) \right|$$

$$\begin{aligned}
&\lesssim \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=l_T+1}^{T-1} E \left| \sum_{s=j+1}^T (k_{h,st}^* - 1)(k_{h,s-j,t}^* - 1) F_t' F_s F_t' F_{s-j} \alpha^{\frac{3+2\eta}{4+2\eta}}(j) \right| \\
&\lesssim \frac{h^{-1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{j=l_T+1}^{T-1} \sum_{t=j+1}^T \alpha^{\frac{3+2\eta}{4+2\eta}}(j) + \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{j=l_T+1}^{T-1} \sum_{s=j+1, s \neq t}^T \alpha^{\frac{3+2\eta}{4+2\eta}}(j) \\
&\lesssim \frac{N^{1/2}h^{-1/2}}{Tl_T^a} \sum_{j=l_T+1}^T j^{a_0} \alpha^{\frac{3+2\eta}{4+2\eta}}(j) + \frac{N^{1/2}h^{1/2}}{l_T^a} \sum_{j=l_T+1}^T j^{a_0} \alpha^{\frac{3+2\eta}{4+2\eta}}(j) = o(1).
\end{aligned}$$

For  $B_{4,1}$ , we have

$$|B_{4,1}| \leq \frac{h^{1/2}}{N^{1/2}T^2} \sum_{j=1}^{l_T} |w_{Tj} - 1| \left| \sum_{i=1}^N \sum_{t=1}^T \sum_{s=j+1}^T E_C(\eta_{it,s} \eta_{it,s-j}) \right|,$$

where the right hand side is dominated by  $\bar{B}_{4,1} \equiv \frac{h^{1/2}}{N^{1/2}T^2} \sum_{j=1}^{l_T} |w_{Tj} - 1| \left| \sum_{i=1}^N \sum_{t=1}^T \sum_{s=j+1}^T E_C(\eta_{it,s} \eta_{it,s-j}) \right|$ .

$$\begin{aligned}
\bar{B}_{4,1} &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{j=1}^{l_T} \left| \sum_{i=1}^N \sum_{t=1}^T \sum_{s=j+1}^T E_C(\eta_{it,s} \eta_{it,s-j}) \right| \\
&= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{j=1}^{l_T} \left| \sum_{i=1}^N \sum_{t=1}^T \sum_{s=j+1}^T (k_{h,st}^* - 1)(k_{h,s-j,t}^* - 1) F_t' F_s F_t' F_{s-j} E_C(\varepsilon_{is} \varepsilon_{i,s-j}) \right| \\
&= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{t=1}^{l_T} \left| \sum_{i=1}^N \sum_{s=t+1}^T |(k_{h,st}^* - 1)(k_{h,s-t,t}^* - 1)| F_t' F_s F_t' F_{s-t} E_C(\varepsilon_{is} \varepsilon_{i,s-t}) \right| \\
&\quad + \frac{h^{1/2}}{N^{1/2}T^2} \sum_{j=1}^{l_T} \left| \sum_{i=1}^N \sum_{s=j+1}^T |(k_{h,ss}^* - 1)(k_{h,s-j,s}^* - 1)| F_t' F_s F_t' F_{s-j} E_C(\varepsilon_{is} \varepsilon_{i,s-j}) \right| \\
&\quad + \frac{h^{1/2}}{N^{1/2}T^2} \sum_{j=1}^{l_T} \left| \sum_{i=1}^N \sum_{s=j+1}^T \sum_{t=1, t \neq s, j}^T |(k_{h,st}^* - 1)(k_{h,s-j,t}^* - 1)| F_t' F_s F_t' F_{s-j} E_C(\varepsilon_{is} \varepsilon_{i,s-j}) \right| \equiv \sum_{\ell=1}^3 \bar{B}_{4,1\ell}.
\end{aligned}$$

By Lemma B.1 with  $\tilde{\eta} = 3 + 2\eta$ , we can readily show that

$$\begin{aligned}
E[\bar{B}_{4,11}] &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{t=1}^{l_T} \sum_{i=1}^N \sum_{s=t+1}^T |(k_{h,st}^* - 1)(k_{h,s-t,t}^* - 1)| E[F_t' F_s F_t' F_{s-t} E_C(\varepsilon_{is} \varepsilon_{i,s-t})] \\
&\lesssim \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{t=1}^{l_T} \sum_{s=t+1}^T |(k_{h,st}^* - 1)(k_{h,s-t,t}^* - 1)| \alpha^{\frac{3+2\eta}{4+2\eta}}(t) = O(N^{1/2}T^{-1}h^{-1/2}) = o(1).
\end{aligned}$$

Then  $\bar{B}_{4,11} = o_P(1)$ . Similarly,  $\bar{B}_{4,12} = o_P(1)$ . For  $\bar{B}_{4,13}$ , we have

$$\begin{aligned}
\bar{B}_{4,13} &= \frac{h^{1/2}}{N^{1/2}T^2} \sum_{j=1}^{l_T} \sum_{i=1}^N \sum_{s=j+1}^T \sum_{t=1, t \neq s, j}^T |(k_{h,st}^* - 1)(k_{h,s-j,t}^* - 1)| |F_t' F_s F_t' F_{s-j} E_C(\varepsilon_{is} \varepsilon_{i,s-j})| \\
&\lesssim \frac{h^{1/2}}{N^{1/2}T^2} \sum_{i=1}^N \sum_{r=1}^{T-1} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=r+1}^{(r+l_T) \wedge T} |(k_{h,st}^* - 1)(k_{h,rt}^* - 1)| \|F_t\|^2 (\|F_s\|^2 + \|F_r\|^2) |E_C(\varepsilon_{is} \varepsilon_{ir})| \\
&= O_P(N^{1/2}T^{-1}h^{-1/2}l_T) = o_P(1).
\end{aligned}$$



Then  $\bar{B}_{4,1} = o_P(1)$ . Since  $\lim_{T \rightarrow \infty} w_{Tj} = 1$  for each  $j$ , it follows from the dominated convergence theorem that  $B_{4,1} = o_P(1)$ .

In sum, we have shown that  $\hat{\mathbb{B}}_{NT}^{(1)} - \mathbb{B}_{NT}^{(1)} = o_p(1)$ .

Now, we prove  $\hat{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)} = o_p(1)$ . Let  $\bar{\mathbb{V}}_{NT}^{(1)} = \frac{h}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^2$ .

$$\begin{aligned}
& \hat{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)} \\
= & \hat{\mathbb{V}}_{NT}^{(1)} - \bar{\mathbb{V}}_{NT}^{(1)} + (\bar{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)}) \\
= & \frac{h}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{s,r=1}^T \hat{\xi}_{h,sr} \hat{\varepsilon}_{ir} \hat{\varepsilon}_{is} \right)^2 - \frac{h}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^2 + (\bar{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)}) \\
= & \frac{h}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} \hat{\varepsilon}_{ir} \hat{\varepsilon}_{is} - \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir}) \right]^2 + \frac{2h}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} \hat{\varepsilon}_{ir} \hat{\varepsilon}_{is} - \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir}) \right) \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right) \\
& + (\bar{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)}) \equiv V_1 + 2V_2 + V_3,
\end{aligned}$$

where  $\hat{\xi}_{h,sr} \equiv \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \tilde{F}_r' \hat{F}_t \hat{F}_t' \tilde{F}_s$  and  $\xi_{h,sr} \equiv \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* F_r' H_0 H_0' F_t F_t' H_0 H_0' F_s$ . Note that

$$\begin{aligned}
\hat{\xi}_{h,sr} \hat{\varepsilon}_{ir} \hat{\varepsilon}_{is} - \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} &= (\hat{\xi}_{h,sr} - \xi_{h,sr})(\hat{\varepsilon}_{ir} \hat{\varepsilon}_{is} - \varepsilon_{is} \varepsilon_{ir}) + (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} \varepsilon_{ir} + \xi_{h,sr} (\hat{\varepsilon}_{ir} \hat{\varepsilon}_{is} - \varepsilon_{is} \varepsilon_{ir}) \\
&= (\hat{\xi}_{h,sr} - \xi_{h,sr})(\hat{\varepsilon}_{ir} - \varepsilon_{ir})(\hat{\varepsilon}_{is} - \varepsilon_{is}) + (\hat{\xi}_{h,sr} - \xi_{h,sr})(\hat{\varepsilon}_{ir} - \varepsilon_{ir}) \varepsilon_{is} \\
&\quad + (\hat{\xi}_{h,sr} - \xi_{h,sr})(\hat{\varepsilon}_{is} - \varepsilon_{is}) \varepsilon_{ir} + (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} \varepsilon_{ir} \\
&\quad + \xi_{h,sr} (\hat{\varepsilon}_{ir} - \varepsilon_{ir})(\hat{\varepsilon}_{is} - \varepsilon_{is}) + \xi_{h,sr} (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) \varepsilon_{is} + \xi_{h,sr} (\hat{\varepsilon}_{is} - \varepsilon_{is}) \varepsilon_{ir} \\
&\equiv \sum_{l=1}^7 \varphi_{l, isr},
\end{aligned} \tag{B.4}$$

and

$$\begin{aligned}
\hat{\xi}_{h,sr} - \xi_{h,sr} &= \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \left( \tilde{F}_r' \hat{F}_t \hat{F}_t' \tilde{F}_s - F_r' H_0 H_0' F_t F_t' H_0 H_0' F_s \right) \\
&= \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \{ (\tilde{F}_r - F_r H_0)' (\hat{F}_t \hat{F}_t' - H_0' F_t F_t' H_0) \tilde{F}_s + (\tilde{F}_r - F_r H_0)' H_0' F_t F_t' H_0 (\tilde{F}_s - H_0' F_s) \\
&\quad + F_r' H_0 (\hat{F}_t \hat{F}_t' - H_0' F_t F_t' H_0) (\tilde{F}_s - H_0' F_s) + (\tilde{F}_r - F_r H_0)' H_0' F_t F_t' H_0 H_0' F_s \\
&\quad + F_r' H_0 (\hat{F}_t \hat{F}_t' - H_0' F_t F_t' H_0) H_0' F_s + F_r' H_0 H_0' F_t F_t' H_0 (\tilde{F}_s - H_0' F_s) \} \\
&\equiv \sum_{l=1}^6 \xi_{lh, sr}.
\end{aligned} \tag{B.5}$$

In addition,

$$\begin{aligned}
& \varepsilon_{is} - \hat{\varepsilon}_{is} \\
= & X'_{is} (\hat{\beta}_s^{bc} - \beta_s) + (\hat{\lambda}'_{is} \hat{F}_s - \lambda'_{is} F_s) \\
= & X'_{is} (\hat{\beta}_s^{bc} - \beta_s) + (\hat{\lambda}_{is} - H^{(s)-1} \lambda_{is})' H^{(s)'} F_s + \lambda'_{is} (H^{(s)-1})' (\hat{F}_s - H^{(s)'} F_s) + (\hat{\lambda}_{is} - H^{(s)-1} \lambda_{is})' (\hat{F}_s - H^{(s)'} F_s) \\
& \equiv \sum_{l=1}^4 \varkappa_{l, is}.
\end{aligned}$$

We can readily show that

$$\frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T (\hat{\varepsilon}_{is} - \varepsilon_{is})^2 \leq \sum_{l=1}^4 \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \varkappa_{lis}^2 = O_P(C_{NT}^{-2} \ln T)$$

where, e.g.,  $\frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T \varkappa_{2is}^2 \lesssim \max_s \frac{1}{N} \sum_{i=1}^N \left\| \hat{\lambda}_{is} - H^{(s)-1} \lambda_{is} \right\|^2 \frac{1}{T} \sum_{s=1}^T \|F_s\|^2 = O_P(C_{NT}^{-2} \ln T)$ . Similarly, we can show that

$$\begin{aligned} \max_t \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T k_{h,st}^* (\hat{\varepsilon}_{is} - \varepsilon_{is})^2 &\leq \max_t \sum_{l=1}^4 \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T k_{h,st}^* \varkappa_{lis}^2 = O_P(C_{NT}^{-2} \ln T), \\ \max_t \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T k_{h,st}^* F_s (\hat{\varepsilon}_{is} - \varepsilon_{is}) &= \max_t \sum_{l=1}^4 \frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T k_{h,st}^* F_s \varkappa_{lis} = O_P(C_{NT}^{-2} \ln T), \\ \frac{1}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{r=1}^T (\hat{\varepsilon}_{ir} - \varepsilon_{ir})^2 \right]^2 &= O_P(C_{NT}^{-4} (\ln T)^2) \text{ and } \frac{1}{T^2} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr})^2 = O_P(C_{NT}^{-2}). \end{aligned}$$

where, e.g.,  $\frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T k_{h,st}^* \varkappa_{2is}^2 \lesssim \max_s \frac{1}{N} \sum_{i=1}^N \left\| \hat{\lambda}_{is} - H^{(s)-1} \lambda_{is} \right\|^2 \max_t \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \|F_s\|^2 = O_P(C_{NT}^{-2} \ln T)$  and  $\frac{1}{NT} \sum_{i=1}^N \sum_{s=1}^T k_{h,st}^* F_s \varkappa_{2is} \lesssim \max_s \left| \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_{is} - H^{(s)-1} \lambda_{is}) \right| \max_t \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \|F_s\|^2 = O_P(C_{NT}^{-2} \ln T) O_P(1)$ .

For  $V_1$ , we have

$$V_1 = \frac{h}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} \hat{\varepsilon}_{ir} \hat{\varepsilon}_{is} - \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir}) \right]^2 \leq 7 \sum_{l=1}^7 \frac{h}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s,r=1}^T \varphi_{l,isr} \right]^2 \equiv \sum_{l=1}^7 V_{1l}.$$

Noting that  $V_{13} = V_{12}$  and  $V_{17} = V_{16}$ , it suffices to study  $V_{1l}$  for  $l = 1, 2, 4, 5, 6$ . For  $V_{11}$ , it suffices to consider a rough bound:

$$\begin{aligned} V_{11} &= \frac{h}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr}) (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) (\hat{\varepsilon}_{is} - \varepsilon_{is}) \right]^2 \\ &\leq T^2 h \frac{1}{T^2} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr})^2 \frac{1}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{r=1}^T (\hat{\varepsilon}_{ir} - \varepsilon_{ir})^2 \right]^2 = T^2 h O_P(C_{NT}^{-2}) O_P(C_{NT}^{-4} (\ln T)^2) = o_P(1). \end{aligned}$$

To study  $V_{1l}$  for  $l = 2, 4, 5, 6$ , we first notice that

$$\max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} \right| \leq \sum_{l=1}^6 \max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T \xi_{hl,sr} \varepsilon_{is} \right| \equiv \sum_{l=1}^6 I_l, \text{ say.}$$

We focus on the study of  $I_6$  as the other terms are of the same or smaller order:

$$\begin{aligned} I_6 &= \max_{i,r} \left| \frac{1}{T} \sum_{t=1}^T k_{h,rt}^* F_r' H_0 H_0' F_t F_t' H_0 \frac{1}{T} \sum_{s=1}^T k_{h,st}^* (\tilde{F}_s - H_0' F_s) \varepsilon_{is} \right| \\ &\lesssim \max_r \|F_r\| \max_r \frac{1}{T} \sum_{t=1}^T k_{h,rt}^* \|F_t\|^2 \left\| \max_{i,t} \frac{1}{T} \sum_{s=1}^T k_{h,st}^* (\tilde{F}_s - H_0' F_s) \varepsilon_{is} \right\| = T^{1/(8+4\eta)} O_P(C_{NT}^{-2} \ln T). \end{aligned}$$

Then  $\max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} \right| = O_P(T^{1/(8+4\eta)} C_{NT}^{-2} \ln T)$ . Similarly, we have

$$\begin{aligned} \max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T \xi_{h,sr} (\hat{\varepsilon}_{is} - \varepsilon_{is}) \right| &= \max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* F_r' H_0 H_0' F_t F_t' H_0 H_0' F_s (\hat{\varepsilon}_{is} - \varepsilon_{is}) \right| \\ &\lesssim \max_r \|F_r\| \max_r \frac{1}{T} \sum_{t=1}^T k_{h,rt}^* \|F_t F_t'\| \max_{i,t} \left| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* F_s (\hat{\varepsilon}_{is} - \varepsilon_{is}) \right| \\ &= O_P(T^{1/(8+4\eta)}) O_P(1) O_P(C_{NT}^{-2} \ln T), \\ \max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T \xi_{h,sr} \varepsilon_{is} \right| &= \max_{i,r} \left| \frac{1}{T} \sum_{t=1}^T k_{h,rt}^* F_r' H_0 H_0' F_t F_t' H_0 H_0' \frac{1}{T} \sum_{s=1}^T k_{h,st}^* F_s \varepsilon_{is} \right| \\ &\lesssim \max_r \|F_r\| \max_r \frac{1}{T} \sum_{t=1}^T k_{h,rt}^* \|F_t F_t'\| \max_{i,t} \left| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* F_s \varepsilon_{is} \right| \\ &= O_P(T^{1/(8+4\eta)}) O_P(1) O_P((Th)^{-1/2} (\ln T)^{1/2}). \end{aligned}$$

It follows that

$$\begin{aligned} V_{12} &= \frac{h}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) \right]^2 \\ &\leq \frac{T^2 h}{N} \sum_{i=1}^N \left[ \max_r \left| \frac{1}{T} \sum_{s=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} \right| \frac{1}{T} \sum_{r=1}^T |\hat{\varepsilon}_{ir} - \varepsilon_{ir}| \right]^2 \\ &\leq T^2 h \max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} \right|^2 \frac{1}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{r=1}^T |\hat{\varepsilon}_{ir} - \varepsilon_{ir}| \right]^2 \\ &= T^2 h O_P(T^{1/(4+2\eta)} C_{NT}^{-4} (\ln T)^2) O_P(C_{NT}^{-2} \ln T) = o_P(1), \end{aligned}$$

$$\begin{aligned} V_{15} &= \frac{h}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) (\hat{\varepsilon}_{is} - \varepsilon_{is}) \right]^2 \\ &\leq T^2 h \max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T \xi_{h,sr} (\hat{\varepsilon}_{is} - \varepsilon_{is}) \right|^2 \frac{1}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{r=1}^T |\hat{\varepsilon}_{ir} - \varepsilon_{ir}| \right]^2 \\ &= T^2 h O_P(T^{1/(4+2\eta)} C_{NT}^{-4} (\ln T)^2) O_P(C_{NT}^{-2} \ln T) = o_P(1), \end{aligned}$$

and

$$\begin{aligned} V_{16} &= \frac{h}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) \varepsilon_{is} \right]^2 = T^2 h \max_{i,r} \left| \frac{1}{T} \sum_{s=1}^T \xi_{h,sr} \varepsilon_{is} \right|^2 \frac{1}{N} \sum_{i=1}^N \left[ \frac{1}{T} \sum_{r=1}^T |\hat{\varepsilon}_{ir} - \varepsilon_{ir}| \right]^2 \\ &= T^2 h O_P(T^{1/(4+2\eta)} C_{NT}^{-4} (\ln T)^2) O_P(C_{NT}^{-2} \ln T) = o_P(1). \end{aligned}$$

For  $V_{14}$ , we use the decomposition in (B.5) to obtain

$$V_{14} = \frac{T^2 h}{N} \sum_{i=1}^N \left[ \frac{1}{T^2} \sum_{s,r=1}^T (\hat{\xi}_{h,sr} - \xi_{h,sr}) \varepsilon_{is} \varepsilon_{ir} \right]^2 \leq 6 \sum_{l=1}^6 \frac{T^2 h}{N} \sum_{i=1}^N \left[ \frac{1}{T^2} \sum_{s,r=1}^T \xi_{lh,sr} \varepsilon_{is} \varepsilon_{ir} \right]^2 \equiv 6 \sum_{l=1}^6 V_{14,l}.$$

We can prove  $V_{14} = o_P(1)$  by showing that  $V_{14,l} = o_P(1)$  for  $l = 1, 2, \dots, 6$ . For example,

$$\begin{aligned}
V_{14,6} &= \frac{T^2 h}{N} \sum_{i=1}^N \left[ \frac{1}{T^2} \sum_{s,r=1}^T \xi_{6h,sr} \varepsilon_{is} \varepsilon_{ir} \right]^2 \\
&= \frac{T^2 h}{N} \sum_{i=1}^N \left[ \text{tr} \left( H_0 H_0' \frac{1}{T} \sum_{t=1}^T F_t F_t' H_0 \frac{1}{T} \sum_{s=1}^T k_{h,st}^* (\tilde{F}_s - H_0' F_s) \varepsilon_{is} \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* F_r' \varepsilon_{ir} \right) \right]^2 \\
&\lesssim T^2 h \max_{i,t} \left\| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* (\tilde{F}_s - H_0' F_s) \varepsilon_{is} \right\|^2 \max_t \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* \varepsilon_{ir} \right\|^2 \\
&= T^2 h O_P(C_{NT}^{-4} (\ln T)^2) O_P((Th)^{-1} \ln T) = o_P(1).
\end{aligned}$$

In sum, we have shown that  $V_1 = o_P(1)$ .

Similarly, we can show that  $V_2 = o_P(1)$ . Next, noting that

$$V_3 = \bar{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)} = \frac{h}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^2 - \frac{h}{N} \sum_{i=1}^N E_C \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^2,$$

we have  $E_C(V_3) = 0$  and

$$\begin{aligned}
\text{Var}_C(V_3) &= \frac{h^2}{N^2} \sum_{i=1}^N \text{Var}_C \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^2 \leq \frac{h^2}{N^2} \sum_{i=1}^N E_C \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^4 \\
&= \frac{h^2}{N^2} \sum_{i=1}^N E_C \left( \frac{1}{T} \sum_{s,r=1}^T \xi_{h,sr} \varepsilon_{is} \varepsilon_{ir} \right)^4 = o_P(1).
\end{aligned}$$

Then  $V_3 = o_P(1)$  by conditional Chebyshev inequality. So we have shown that  $\hat{\mathbb{V}}_{NT}^{(1)} - \mathbb{V}_{NT}^{(1)} = o_P(1)$ .

(ii) By Theorem 4.1(ii), it suffices to show (ii1)  $\hat{\mathbb{B}}_{NT}^{(2)} - \mathbb{B}_{NT}^{(2)} = o_p(1)$  and (ii2)  $\hat{\mathbb{V}}_{NT}^{(2)} - \mathbb{V}_{NT}^{(2)} = o_p(1)$ . For the use in the proof of Theorem 4.3(ii), we prove the claims under  $\mathbb{H}_1^{(2)}(a_{2NT})$ .

We first prove (ii1) by showing that  $\hat{\mathbb{B}}_{1,NT}^{(2)} - \mathbb{B}_{1,NT}^{(2)} = o_p(1)$  and  $\hat{\mathbb{B}}_{2,NT}^{(2)} - \mathbb{B}_{2,NT}^{(2)} = o_p(1)$ . We make the following two decompositions:

$$\begin{aligned}
&\hat{\varsigma}_{ij,sr} \hat{\varepsilon}_{jr} \hat{\varepsilon}_{is} - \varsigma_{h,sr} \varepsilon_{is} \varepsilon_{ir} \\
&= (\hat{\varsigma}_{ij,sr} - \varsigma_{ij,sr})(\hat{\varepsilon}_{jr} \hat{\varepsilon}_{is} - \varepsilon_{is} \varepsilon_{jr}) + (\hat{\varsigma}_{ij,sr} - \varsigma_{ij,sr}) \varepsilon_{is} \varepsilon_{jr} + \varsigma_{ij,sr} (\hat{\varepsilon}_{jr} \hat{\varepsilon}_{is} - \varepsilon_{is} \varepsilon_{jr}) \\
&= (\hat{\varsigma}_{ij,sr} - \varsigma_{ij,sr})(\hat{\varepsilon}_{jr} - \varepsilon_{jr})(\hat{\varepsilon}_{is} - \varepsilon_{is}) + (\hat{\varsigma}_{ij,sr} - \varsigma_{ij,sr})(\hat{\varepsilon}_{jr} - \varepsilon_{jr}) \varepsilon_{is} \\
&\quad + (\hat{\varsigma}_{ij,sr} - \varsigma_{ij,sr})(\hat{\varepsilon}_{is} - \varepsilon_{is}) \varepsilon_{jr} + (\hat{\varsigma}_{ij,sr} - \varsigma_{ij,sr}) \varepsilon_{is} \varepsilon_{jr} \\
&\quad + \varsigma_{ij,sr} (\hat{\varepsilon}_{jr} - \varepsilon_{jr})(\hat{\varepsilon}_{is} - \varepsilon_{is}) + \varsigma_{ij,sr} (\hat{\varepsilon}_{jr} - \varepsilon_{jr}) \varepsilon_{is} + \varsigma_{ij,sr} (\hat{\varepsilon}_{is} - \varepsilon_{is}) \varepsilon_{jr} \equiv \sum_{l=1}^7 \phi_{l,ij,sr} \quad (\text{B.6})
\end{aligned}$$

and

$$\begin{aligned}
\hat{\varsigma}_{ii,sr} - \varsigma_{ii,sr} &= \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \left[ \hat{\mathcal{X}}_{is}^{(t)'} \hat{D}^{(t)} (\hat{F}^{(t)})^{-1} \hat{D}^{(t)} (\hat{F}^{(t)})^{-1} \hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{is}^{(t)'} D(F^{(t)})^{-1} D(F^{(t)})^{-1} \mathcal{X}_{ir}^{(t)} \right] \\
&= \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \{ [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] [\hat{D}^{(t)} (\hat{F}^{(t)})^{-1} \hat{D}^{(t)} (\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \}
\end{aligned}$$

$$\begin{aligned}
& + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] [\hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] \mathcal{X}_{ir}^{(t)} \\
& + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] D(F^{(t)})^{-1} D(F^{(t)})^{-1} [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] D(F^{(t)})^{-1} D(F^{(t)})^{-1} \mathcal{X}_{ir}^{(t)} \\
& + \mathcal{X}_{is}^{(t)} [\hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \\
& + \mathcal{X}_{is}^{(t)} [\hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] \mathcal{X}_{ir}^{(t)} \\
& + \mathcal{X}_{is}^{(t)'} D(F^{(t)})^{-1} D(F^{(t)})^{-1} [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \} \equiv \sum_{l=1}^7 \varsigma_{l,ii,sr}
\end{aligned}$$

and

$$\begin{aligned}
\hat{\varsigma}_{ii,sr} - \varsigma_{ii,sr} &= \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \left[ \hat{\mathcal{X}}_{is}^{(t)'} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{is}^{(t)'} D(F^{(t)})^{-1} D(F^{(t)})^{-1} \mathcal{X}_{ir}^{(t)} \right] \\
&= \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \{ [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] [\hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \\
&\quad + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] [\hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] \mathcal{X}_{ir}^{(t)} \\
&\quad + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] D(F^{(t)})^{-1} D(F^{(t)})^{-1} [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)}] D(F^{(t)})^{-1} D(F^{(t)})^{-1} \mathcal{X}_{ir}^{(t)} \\
&\quad + \mathcal{X}_{is}^{(t)} [\hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \\
&\quad + \mathcal{X}_{is}^{(t)} [\hat{D}^{(t)}(\hat{F}^{(t)})^{-1} \hat{D}^{(t)}(\hat{F}^{(t)})^{-1} - D(F^{(t)})^{-1} D(F^{(t)})^{-1}] \mathcal{X}_{ir}^{(t)} \\
&\quad + \mathcal{X}_{is}^{(t)'} D(F^{(t)})^{-1} D(F^{(t)})^{-1} [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \} \equiv \sum_{l=1}^7 \varsigma_{l,ii,sr} \tag{B.7}
\end{aligned}$$

Note that

$$\hat{\mathbb{B}}_{1,NT}^{(2)} - \mathbb{B}_{NT}^{(2)} = \frac{h^{1/2}}{NT} \sum_{i=1}^N \sum_{s,r=1}^T (\hat{\varepsilon}_{is} \hat{\varepsilon}_{ir} \hat{\varsigma}_{ii,sr} - \varepsilon_{is} \varepsilon_{ir} \varsigma_{ii,sr}) = \sum_{l=1}^7 \frac{h^{1/2}}{NT} \sum_{i=1}^N \sum_{s,r=1}^T \phi_{l,ii,sr} \equiv \sum_{l=1}^7 B_l^{(2)}$$

Noting that  $B_2^{(2)} = B_3^{(2)}$  and  $B_6^{(2)} = B_7^{(2)}$ , it suffices to show  $B_l^{(2)} = o_P(1)$  for  $l = 1, 2, 4, 5, 6$ . The proofs of these claims are straightforward so that we only demonstrate that  $B_4^{(2)} = o_P(1)$  and  $B_6^{(2)} = o_P(1)$ . For  $B_4^{(2)}$ , we make the following decomposition:

$$B_4^{(2)} = \frac{h^{1/2}}{NT} \sum_{i=1}^N \sum_{s,r=1}^T (\hat{\varsigma}_{ii,sr} - \varsigma_{ii,sr}) \varepsilon_{is} \varepsilon_{ir} = \sum_{l=1}^7 \frac{h^{1/2}}{NT} \sum_{i=1}^N \sum_{s,r=1}^T \varsigma_{l,ii,sr} \varepsilon_{is} \varepsilon_{ir} \equiv \sum_{l=1}^7 B_{4,l}^{(2)}.$$

We can show that  $B_{4,l}^{(2)} = o_P(1)$  for  $l = 1, \dots, 7$ . For example,

$$\begin{aligned}
|B_{4,7}^{(2)}| &= \frac{Th^{1/2}}{N} \left| \sum_{i=1}^N \frac{1}{T} \sum_{t=1}^T \text{tr} \left\{ D(F^{(t)})^{-1} D(F^{(t)})^{-1} \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \varepsilon_{ir} \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \varepsilon_{is} \mathcal{X}_{is}^{(t)'} \right\} \right| \\
&\lesssim \frac{Th^{1/2}}{N} \sum_{i=1}^N \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \varepsilon_{ir} \right\| \left\| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \varepsilon_{is} \mathcal{X}_{is}^{(t)'} \right\| \\
&= Th^{1/2} O_P(C_{NT}^{-2} \ln T) O_P((Th)^{-1/2}) = o_P(1),
\end{aligned}$$

where we use the fact that  $\max_{i,t} \left\| \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* [\hat{\mathcal{X}}_{ir}^{(t)} - \mathcal{X}_{ir}^{(t)}] \varepsilon_{ir} \right\| = O_P(C_{NT}^{-2} \ln T)$  by straightforward verifications. For  $B_6^{(2)}$ , we have

$$\begin{aligned}
|B_6^{(2)}| &= \frac{h^{1/2}}{NT} \left| \sum_{i=1}^N \sum_{s,r=1}^T \varsigma_{ii,sr} (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) \varepsilon_{is} \right| \\
&= \frac{Th^{1/2}}{N} \left| \sum_{i=1}^N \frac{1}{T} \sum_{t=1}^T \text{tr} \left( D(F^{(t)})^{-1} D(F^{(t)})^{-1} \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* \mathcal{X}_{ir}^{(t)} (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \varepsilon_{is} \mathcal{X}_{is}^{(t)'} \right) \right| \\
&\lesssim \frac{Th^{1/2}}{N} \sum_{i=1}^N \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* \mathcal{X}_{ir}^{(t)} (\hat{\varepsilon}_{ir} - \varepsilon_{ir}) \right\| \left\| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* \varepsilon_{is} \mathcal{X}_{is}^{(t)'} \right\| \\
&= Th^{1/2} O_P(C_{NT}^{-2} \ln T) O_P((Th)^{-1/2}) = o_P(1).
\end{aligned}$$

Consequently, we have established that  $\hat{\mathbb{B}}_{1,NT}^{(2)} - \mathbb{B}_{1,NT}^{(2)} = o_p(1)$ .

Next,

$$\begin{aligned}
\hat{\mathbb{B}}_{2,NT}^{(2)} - \mathbb{B}_{2,NT}^{(2)} &= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} [\hat{\varepsilon}_{is} \hat{\varepsilon}_{js} \hat{\mathcal{X}}_{is}^{(t)'} \hat{\mathbb{D}}^{(t)} \hat{\mathcal{X}}_{js}^{(t)} - \varepsilon_{is} \varepsilon_{js} \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)}] \\
&= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \{ [\hat{\varepsilon}_{is} \hat{\varepsilon}_{js} - \varepsilon_{is} \varepsilon_{js}] \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)} + \varepsilon_{is} \varepsilon_{js} [\hat{\mathcal{X}}_{is}^{(t)'} \hat{\mathbb{D}}^{(t)} \hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)}] \\
&\quad + [\hat{\varepsilon}_{is} \hat{\varepsilon}_{js} - \varepsilon_{is} \varepsilon_{js}] [\hat{\mathcal{X}}_{is}^{(t)'} \hat{\mathbb{D}}^{(t)} \hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)}] \} \equiv 2B_{2,1} + 2B_{2,2} + B_{2,3}.
\end{aligned}$$

For  $B_{2,1}$ , we make the following decomposition.

$$\begin{aligned}
B_{2,1} &= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \{ [\hat{\varepsilon}_{is} \hat{\varepsilon}_{js} - \varepsilon_{is} \varepsilon_{js}] \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)} + \varepsilon_{is} \varepsilon_{js} [\hat{\mathcal{X}}_{is}^{(t)'} \hat{\mathbb{D}}^{(t)} \hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)}] \} \\
&\equiv B_{2,11} + B_{2,12}.
\end{aligned}$$

Noting that  $\hat{\varepsilon}_{is} \hat{\varepsilon}_{js} - \varepsilon_{is} \varepsilon_{js} = (\hat{\varepsilon}_{is} - \varepsilon_{is})(\hat{\varepsilon}_{js} - \varepsilon_{js}) + (\hat{\varepsilon}_{is} - \varepsilon_{is})\varepsilon_{js} + \varepsilon_{is}(\hat{\varepsilon}_{js} - \varepsilon_{js})$ , we have

$$\begin{aligned}
B_{2,11} &= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} [\hat{\varepsilon}_{is} \hat{\varepsilon}_{js} - \varepsilon_{is} \varepsilon_{js}] \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)} \\
&= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} [(\hat{\varepsilon}_{is} - \varepsilon_{is})(\hat{\varepsilon}_{js} - \varepsilon_{js}) + (\hat{\varepsilon}_{is} - \varepsilon_{is})\varepsilon_{js} + \varepsilon_{is}(\hat{\varepsilon}_{js} - \varepsilon_{js})] \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)} \equiv \sum_{l=1}^3 B_{2,11l}.
\end{aligned}$$

We can readily show that  $B_{2,11} = o_P(1)$  by showing that  $B_{2,11l} = o_P(1)$  for  $l = 1, 2, 3$ . For example,

$$\begin{aligned}
|B_{2,112}| &= \left| \frac{h^{1/2}}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} (\hat{\varepsilon}_{is} - \varepsilon_{is}) \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \sum_{j=1}^N \mathcal{X}_{js}^{(t)} \varepsilon_{js} \right| + o_P(1) \\
&\leq \left| \frac{h^{1/2}}{NT^2} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \sum_{i=1}^N (\hat{\varepsilon}_{is} - \varepsilon_{is}) \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \sum_{j=1}^N \mathcal{X}_{js}^{(t)} \varepsilon_{js} \right| + o_P(1) \\
&\lesssim N \left\{ \frac{h}{T^2} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \left\| \frac{1}{N} \sum_{i=1}^N (\hat{\varepsilon}_{is} - \varepsilon_{is}) \mathcal{X}_{is}^{(t)'} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{T^2} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \left\| \frac{1}{N} \sum_{j=1}^N \mathcal{X}_{js}^{(t)} \varepsilon_{js} \right\|^2 \right\}^{1/2} + o_P(1)
\end{aligned}$$

$$= NO_P((NTh)^{-1/2})O_P(h^{-1/2}N^{-1/2}) + o_P(1) = O_P(T^{-1/2}h^{-1}) + o_P(1) = o_P(1),$$

as we can show that  $\frac{h}{T^2} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \left\| \frac{1}{N} \sum_{i=1}^N (\hat{\varepsilon}_{is} - \varepsilon_{is}) \mathcal{X}_{is}^{(t)'} \right\|^2 = O_P((NTh)^{-1} + h^4) = O_P((NTh)^{-1})$ . Similarly, we can show that  $B_{2,12} = o_P(1)$ .

For  $B_{2,2}$ , we make the following decomposition:

$$\begin{aligned} B_{2,2} &= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \varepsilon_{is} \varepsilon_{js} [\hat{\mathcal{X}}_{is}^{(t)'} \hat{\mathbb{D}}^{(t)} \hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)}] \\ &= \frac{2h^{1/2}}{NT^2} \sum_{1 \leq j < i \leq N} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \varepsilon_{is} \varepsilon_{js} \{ [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)'}]' [\hat{\mathbb{D}}^{(t)} - \mathbb{D}^{(t)}] [\hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{js}^{(t)}] \\ &\quad + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)'}]' [\hat{\mathbb{D}}^{(t)} - \mathbb{D}^{(t)}] \mathcal{X}_{js}^{(t)} + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)'}]' \mathbb{D}^{(t)} [\hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{js}^{(t)}] \\ &\quad + [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)'}]' \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)} + \mathcal{X}_{is}^{(t)'} (\hat{\mathbb{D}}^{(t)} - \mathbb{D}^{(t)}) (\hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{js}^{(t)}) + \mathcal{X}_{is}^{(t)'} (\hat{\mathbb{D}}^{(t)} - \mathbb{D}^{(t)}) \mathcal{X}_{js}^{(t)} \\ &\quad + \mathcal{X}_{is}^{(t)'} \mathbb{D}^{(t)} [\hat{\mathcal{X}}_{js}^{(t)} - \mathcal{X}_{js}^{(t)}] \} \equiv \sum_{l=1}^7 B_{2,2l}. \end{aligned}$$

It is easy to show that  $B_{2,2l} = o_P(1)$  for  $l = 1, 2, 3, 5$ , and  $6$ . Next,  $B_{2,24} = \bar{B}_{2,24} + o_P(1)$ , where  $\bar{B}_{2,24} = \frac{h^{1/2}}{NT^2} \sum_{i,j} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \varepsilon_{is} \varepsilon_{js} [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)'}]' \mathbb{D}^{(t)} \mathcal{X}_{js}^{(t)}$ . Note that

$$\begin{aligned} |\bar{B}_{2,24}| &= \left| \frac{Nh^{1/2}}{T^2} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \left\{ \frac{1}{N} \sum_{i=1}^N \varepsilon_{is} [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)'}]' \right\} \mathbb{D}^{(t)} \left\{ \frac{1}{N} \sum_{j=1}^N \mathcal{X}_{js}^{(t)} \varepsilon_{js} \right\} \right| \\ &\lesssim N \left\{ \frac{h}{T^2} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \left\| \frac{1}{N} \sum_{i=1}^N \varepsilon_{is} [\hat{\mathcal{X}}_{is}^{(t)} - \mathcal{X}_{is}^{(t)'}]' \right\|^2 \right\}^{1/2} \left\{ \frac{1}{T^2} \sum_{t=1}^T \sum_{s=1}^T k_{h,st}^{*2} \left\| \frac{1}{N} \sum_{j=1}^N \mathcal{X}_{js}^{(t)} \varepsilon_{js} \right\|^2 \right\}^{1/2} \\ &= NO_P((NTh)^{-1/2})O_P((Nh)^{-1/2}) = o_P(1) \end{aligned}$$

Similarly,  $B_{2,27} = o_P(1)$ . Then  $B_{2,2} = o_P(1)$ . Lastly, one can also show that  $B_{2,3} = o_P(1)$ . Then  $\hat{\mathbb{B}}_{2,NT}^{(2)} - \mathbb{B}_{2,NT}^{(2)} = o_P(1)$ .

Let  $\bar{\mathbb{V}}_{NT}^{(2)} = \frac{4h}{N^2 T^2} \sum_{1 \leq i < j \leq N} \left( \sum_{s,r=1}^T \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr} \right)^2$ . Note that

$$\begin{aligned} \hat{\mathbb{V}}_{NT}^{(2)} - \mathbb{V}_{NT}^{(2)} &= \frac{2h}{N^2} \sum_{1 \leq i \neq j \leq N} \left[ \left( \frac{1}{T} \sum_{s \neq r} \hat{\varepsilon}_{is} \hat{\varepsilon}_{jr} \hat{\varsigma}_{ij,sr} \right)^2 - \left( \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr} \right)^2 \right] + [\bar{\mathbb{V}}_{NT}^{(2)} - \mathbb{V}_{NT}^{(2)}] \\ &= \frac{2h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} [\hat{\varepsilon}_{is} \hat{\varepsilon}_{jr} \hat{\varsigma}_{ij,sr} - \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}] \right)^2 \\ &\quad + \frac{4h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} [\hat{\varepsilon}_{is} \hat{\varepsilon}_{jr} \hat{\varsigma}_{ij,sr} - \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}] \right) \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr} + [\bar{\mathbb{V}}_{NT}^{(2)} - \mathbb{V}_{NT}^{(2)}] \\ &\equiv 2V_4 + 4V_5 + V_6. \end{aligned}$$

Note that  $V_4 \leq 7 \sum_{l=1}^7 \frac{h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} \phi_{l,ij,sr} \right)^2 \equiv 7 \sum_{l=1}^7 V_{4,l}$ . We can readily show that  $V_4 = o_P(1)$  by proving that  $V_{4,l} = o_P(1)$  for  $l = 1, \dots, 7$ . For example,

$$\begin{aligned}
V_{4,7} &= \frac{h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} \varsigma_{ij,sr} (\hat{\varepsilon}_{is} - \varepsilon_{is}) \varepsilon_{jr} \right)^2 \\
&= \frac{h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \mathcal{X}_{is}^{(t)'} D(F^{(t)})^{-1} D(F^{(t)})^{-1} \mathcal{X}_{jr}^{(t)} (\hat{\varepsilon}_{is} - \varepsilon_{is}) \varepsilon_{jr} \right)^2 \\
&= \frac{T^2 h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{t=1}^T \frac{1}{T} \sum_{s=1}^T k_{h,st}^* (\hat{\varepsilon}_{is} - \varepsilon_{is}) \mathcal{X}_{is}^{(t)'} D(F^{(t)})^{-1} D(F^{(t)})^{-1} \frac{1}{T} \sum_{s \neq r} k_{h,rt}^* \mathcal{X}_{jr}^{(t)} \varepsilon_{jr} \right)^2 \\
&\lesssim \frac{T^2 h}{N^2} \sum_{1 \leq i \neq j \leq N} \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{s=1}^T k_{h,st}^* (\hat{\varepsilon}_{is} - \varepsilon_{is}) \mathcal{X}_{is}^{(t)'} \right\|^2 \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{T} \sum_{s \neq r} k_{h,rt}^* \mathcal{X}_{jr}^{(t)} \varepsilon_{jr} \right\|^2 \\
&= T^2 h O_P(C_{NT}^{-4} (\ln T)^2) O_P((Th)^{-1}) = o_P(1).
\end{aligned}$$

For  $V_5$ , we apply CS inequality to obtain

$$\begin{aligned}
V_5 &= \frac{h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} [\hat{\varepsilon}_{is} \hat{\varepsilon}_{jr} \hat{\varsigma}_{ij,sr} - \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}] \right) \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr} \\
&\leq \{V_4\}^{1/2} \left\{ \frac{h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr} \right)^2 \right\}^{1/2} = o_P(1) O_P(1) = o_P(1),
\end{aligned}$$

where we also use the fact that

$$\begin{aligned}
&\frac{h}{N^2} \sum_{1 \leq i \neq j \leq N} E_C \left( \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr} \right)^2 \\
&\leq \frac{2h}{N^2} \sum_{1 \leq i \neq j \leq N} \left( \frac{1}{T} \sum_{s \neq r} E_C (\varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}) \right)^2 + \frac{2h}{N^2} \sum_{1 \leq i \neq j \leq N} \text{Var}_C \left( \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr} \right)^2 = O_P(1) + O_P(1)
\end{aligned}$$

and

$$\begin{aligned}
\left| \frac{1}{T} \sum_{s \neq r} E_C (\varepsilon_{is} \varepsilon_{jr} \varsigma_{ij,sr}) \right| &= \left| \frac{1}{T} \sum_{s \neq r} \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* E_C [\varepsilon_{is} \mathcal{X}_{is}^{(t)'} D(F^{(t)})^{-1} D(F^{(t)})^{-1} \mathcal{X}_{jr}^{(t)} \varepsilon_{jr}] \right| \\
&\lesssim \left| \frac{1}{T} \sum_{s \neq r} \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* \text{tr} \left( E_C [\mathcal{X}_{jr}^{(t)} \varepsilon_{jr} \varepsilon_{is} \mathcal{X}_{is}^{(t)'}] \right) \right| \\
&\asymp \left| \frac{1}{T} \sum_{s \neq r} \frac{1}{T} \sum_{t=1}^T k_{h,st}^* k_{h,rt}^* E_C [\varepsilon_{is} \mathcal{X}_{is}' \mathcal{X}_{jr} \varepsilon_{jr}] \right| \\
&\leq \frac{1}{T} \sum_{r=1}^T \frac{1}{T} \sum_{t=1}^T \max_s k_{h,st}^* k_{h,rt}^* \sum_{s=1}^T |E_C [\varepsilon_{is} \mathcal{X}_{is}' \mathcal{X}_{jr} \varepsilon_{jr}]|
\end{aligned}$$



$$\leq \max_{i,j,r} \sum_{s=1}^T |E_C [\varepsilon_{is} \mathcal{X}'_{is} \mathcal{X}_{jr} \varepsilon_{jr}]| \frac{1}{T} \sum_{t=1}^T \max_s k_{h,st}^* \frac{1}{T} \sum_{r=1}^T k_{h,rt}^* = O_P(1).$$

Note that  $V_6 = \frac{4h}{N^2 T^2} \sum_{1 \leq i < j \leq N} \left[ \left( \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij, sr} \right)^2 - E_C \left( \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij, sr} \right)^2 \right]$ . It is easy to see that  $E_C(V_6) = 0$  and

$$\text{Var}_C(V_6) = \frac{16h^2}{N^4} \sum_{1 \leq i < j \leq N} \text{Var}_C \left( \left( \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij, sr} \right)^2 \right) \leq \frac{16h^2}{N^4} \sum_{1 \leq i < j \leq N} E_C \left[ \left( \frac{1}{T} \sum_{s \neq r} \varepsilon_{is} \varepsilon_{jr} \varsigma_{ij, sr} \right)^4 \right] = o_P(1).$$

(iii) The estimators are the same as those in (i) and the proof is similar to that of (i) and thus omitted. ■

**Proof of Theorem 4.3.** By the proof of Theorem 4.1, we have  $J_{NT}^{(l)} \xrightarrow{d} N(\pi^{(l)}, 1)$  under the corresponding local alternatives for  $l = 1, 2, 3$ . By the proof of Theorem 4.2, we have  $\hat{\mathbb{B}}_{NT}^{(l)} = \mathbb{B}_{NT}^{(l)} + o_P(1)$ , and  $\hat{\mathbb{V}}_{NT}^{(l)} = \mathbb{V}_{NT}^{(l)} + o_P(1)$  under the corresponding local alternatives for  $l = 1, 2, 3$ . It follows that  $\hat{J}_{NT}^{(l)} \xrightarrow{d} N(\pi^{(l)}, 1)$  under the corresponding local alternatives. ■

**Proof of Theorem 4.4.** The proof of this theorem is almost the same as that of Theorem 4.5 in Su and Wang (2017). We do not repeat it here. ■

## C Proofs of the Technical Lemmas in Appendix A

**Proof of Lemma A.1.** (i) Note that  $\frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} \varepsilon_i^{(r)} = \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} \varepsilon_i^{(r)} - \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} P_{F^{(r)}} \varepsilon_i^{(r)}$ . For the first term, we can apply Chebyshev inequality to show that  $\frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} \varepsilon_i^{(r)} = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{tr}^* X_{it} \varepsilon_{it} = O_P((NTh)^{-1/2})$ . For the second term, noting that  $P_{F^{(r)}} = T^{-1} F^{(r)} F^{(r)'}$ , we have

$$\begin{aligned} \frac{1}{NT} \left\| \sum_{i=1}^N X_i^{(r)'} P_{F^{(r)}} \varepsilon_i^{(r)} \right\| &= \left\| \frac{1}{N} \sum_{i=1}^N \frac{X_i^{(r)'} F^{(r)}}{T} \frac{1}{T} \sum_{t=1}^T F_t^{(r)} \varepsilon_{it}^{(r)} \right\| \leq \frac{1}{N} \sum_{i=1}^N \left\| \frac{X_i^{(r)'} F^{(r)}}{T} \right\| \left\| \frac{1}{T} \sum_{t=1}^T F_t^{(r)} \varepsilon_{it}^{(r)} \right\| \\ &\leq \left( \frac{1}{N} \sum_{i=1}^N \left\| \frac{X_i^{(r)'} F^{(r)}}{T} \right\|^2 \right)^{1/2} \left( \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T F_t^{(r)} \varepsilon_{it}^{(r)} \right\|^2 \right)^{1/2}, \end{aligned}$$

where the first inequality holds by the submultiplicative property of the Frobenius norm and the second inequality holds by Cauchy-Schwarz (CS) inequality. It is easy to apply Markov inequality to show that

$$\sup_{F^{(r)} \in \mathcal{F}} \frac{1}{N} \sum_{i=1}^N \left\| \frac{X_i^{(r)'} F^{(r)}}{T} \right\|^2 \leq \sup_{F^{(r)} \in \mathcal{F}} \frac{1}{T} \|F^{(r)}\|^2 \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \|X_{it}^{(r)}\|^2 = \frac{R}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{tr}^* \|X_{it}\|^2 = O_P(1).$$

In addition

$$\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T F_t^{(r)} \varepsilon_{it}^{(r)} \right\|^2 = \frac{1}{NT^2} \sum_{i=1}^N \sum_{t=1}^T F_t^{(r)'} F_t^{(r)} (\varepsilon_{it}^{(r)})^2 + \frac{1}{NT^2} \sum_{i=1}^N \sum_{t \neq s} F_t^{(r)'} F_s^{(r)} \varepsilon_{it}^{(r)} \varepsilon_{is}^{(r)} \equiv I_1 + I_2.$$

For  $I_1$ , we have  $I_1 \leq \sup_{F^{(r)} \in \mathcal{F}} \frac{1}{T^2} \sum_{t=1}^T \|F_t^{(r)}\|^2 \max_t \frac{1}{N} \sum_{i=1}^N k_{tr}^* \varepsilon_{it}^2 \leq \frac{R}{T} \max_t \frac{1}{N} \sum_{i=1}^N k_{tr}^* \varepsilon_{it}^2 = O_P(T^{-1})$  as it is standard to show that  $\max_t \frac{1}{N} \sum_{i=1}^N k_{tr}^* \varepsilon_{it}^2 \leq \max_t \frac{1}{N} \sum_{i=1}^N k_{tr}^* E(\varepsilon_{it}^2) + \max_t \frac{1}{N} \sum_{i=1}^N k_{tr}^* [\varepsilon_{it}^2 - E(\varepsilon_{it}^2)] = O_P(1)$ .

For  $I_2$ , we have by CS inequality:

$$\begin{aligned}
& \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT^2} \sum_{i=1}^N \sum_{t \neq s} F_t^{(r)'} F_s^{(r)} \varepsilon_{it}^{(r)} \varepsilon_{is}^{(r)} \right\| \\
&= \sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{T^2} \sum_{t \neq s} F_t^{(r)'} F_s^{(r)} (k_{tr}^* k_{sr}^*)^{1/2} \frac{1}{N} \sum_{i=1}^N \varepsilon_{it} \varepsilon_{is} \right\| \\
&\leq N^{-1/2} \left( \sup_{F^{(r)} \in \mathcal{F}} \frac{1}{T^2} \sum_{t \neq s} \|F_s^{(r)}\|^2 \|F_t^{(r)}\|^2 \right)^{1/2} \left( \frac{1}{T^2} \sum_{t \neq s} (k_{tr}^* k_{sr}^*)^{1/2} \left[ \frac{1}{\sqrt{N}} \sum_{i=1}^N \varepsilon_{it} \varepsilon_{is} \right]^2 \right)^{1/2} = O_P(N^{-1/2}),
\end{aligned}$$

where we use the fact that  $\frac{1}{T^2} \sum_{t \neq s} (k_{tr}^* k_{sr}^*)^{1/2} \left[ \frac{1}{\sqrt{N}} \sum_{i=1}^N \varepsilon_{it} \varepsilon_{is} \right]^2 = O_P(1)$  by Markov equality and Assumption A.2(i)-(iii). It follows that

$$\sup_{F^{(r)} \in \mathcal{F}} \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T F_t^{(r)} \varepsilon_{it}^{(r)} \right\|^2 = O_P(T^{-1/2} + N^{-1/4}) \quad (\text{C.1})$$

and  $\sup_{F^{(r)} \in \mathcal{F}} \frac{1}{NT} \left\| \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} \varepsilon_i^{(r)} \right\| = O_P((NT h)^{-1/2} + T^{-1/2} + N^{-1/4}) = O_P(T^{-1/2} + N^{-1/4})$ .

(ii) and (iii) The proof is similar to that of (i) and thus omitted here.

(iv) Note that  $\sup_{F^{(r)} \in \mathcal{F}} \frac{1}{NT} \left\| \sum_{i=1}^N \Delta_i^{(r)'} (P_{F^{(r)}} - P_{F^{(r)0}}) \Delta_i^{(r)} \right\| \leq \frac{1}{NT} \sum_{i=1}^N \left\| \Delta_i^{(r)'} \Delta_i^{(r)} \right\|$ . The result follows because

$$\frac{1}{NT} \sum_{i=1}^N \left\| \Delta_i^{(r)'} \Delta_i^{(r)} \right\| = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h, tr}^* [X_{it}' d_0(t, r) + F_t^0 d_i(t, r)]^2 = O_P(h^2). \quad (\text{C.2})$$

(v) Note that

$$\sup_{F^{(r)} \in \mathcal{F}} \left\| \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{F^{(r)}} \Delta_i^{(r)} \right\| \leq \left\{ \frac{1}{NT} \sum_{i=1}^N \|X_i^{(r)}\|^2 \right\}^{1/2} \left\{ \frac{1}{NT} \sum_{i=1}^N \|\Delta_i^{(r)}\|^2 \right\}^{1/2} = O_P(1) O_P(h) = O_P(h).$$

(vi) The proof is quite similar to that of (v) and thus omitted here.

(vii) We only show that  $\sup_{F^{(r)} \in \mathcal{F}} \frac{1}{NT} \left\| \sum_{i=1}^N \Delta_i^{(r)'} P_{F^{(r)}} \varepsilon_i^{(r)} \right\| = O_P((T^{-1/2} + N^{-1/4})h)$  as it is easier to show that  $\frac{1}{NT} \left\| \sum_{i=1}^N \Delta_i^{(r)'} P_{F^{(r)0}} \varepsilon_i^{(r)} \right\| = O_P((T^{-1/2} + N^{-1/4})h)$ . By (C.1) and (C.2),

$$\begin{aligned}
\sup_{F^{(r)} \in \mathcal{F}} \frac{1}{NT} \left\| \sum_{i=1}^N \Delta_i^{(r)'} P_{F^{(r)}} \varepsilon_i^{(r)} \right\| &= \sup_{F^{(r)} \in \mathcal{F}} \frac{1}{NT^2} \left\| \sum_{i=1}^N \Delta_i^{(r)'} F^{(r)} F^{(r)'} \varepsilon_i^{(r)} \right\| \\
&\leq \sup_{F^{(r)} \in \mathcal{F}} T^{-1/2} \|F^{(r)}\| \left\{ \frac{1}{NT} \sum_{i=1}^N \|\Delta_i^{(r)}\|^2 \right\}^{1/2} \sup_{F^{(r)} \in \mathcal{F}} \left\{ \frac{1}{NT^2} \sum_{i=1}^N \|F^{(r)'} \varepsilon_i^{(r)}\|^2 \right\}^{1/2} \\
&\leq R^{1/2} O_P(h) O_P(T^{-1/2} + N^{-1/4}) = O_P((T^{-1/2} + N^{-1/4})h).
\end{aligned}$$

It follows that  $\sup_{F^{(r)} \in \mathcal{F}} \frac{1}{NT} \left\| \sum_{i=1}^N \Delta_i^{(r)'} (P_{F^{(r)}} - P_{F^{(r)0}}) \varepsilon_i^{(r)} \right\| = O_P((T^{-1/2} + N^{-1/4})h)$ . ■

In the following proofs, we suppress the superscript 0 for the true parameters  $\beta_r^0$ ,  $F_t^{(r)0}$ ,  $F^{(r)0}$ ,  $\lambda_{ir}^0$  and  $\Lambda_r^0$  unless confusion may arise.

**Proof of Lemma A.2.** This Lemma is parallel to Lemma A.1 in Su and Wang (2017) and we try to be brief in the proof.

(i) Noting that  $Y_i^{(r)} - X_i^{(r)} \hat{\beta}_r = -X_i^{(r)} \hat{\delta}_r + F^{(r)} \lambda_{ir} + \Delta_i^{(r)} + \varepsilon_i^{(r)}$  with  $\hat{\delta}_r = \hat{\beta}_r - \beta_r$ , we start with the eigenvalue problem  $\left[ \frac{1}{NT} (Y^{(r)} - X^{(r)} \hat{\beta}_r)(Y^{(r)} - X^{(r)} \hat{\beta}_r)' \right] \hat{F}^{(r)} = \hat{F}^{(r)} \hat{V}_{NT}^{(r)}$  to obtain the following decomposition:

$$\begin{aligned}
& \hat{F}^{(r)} \hat{V}_{NT}^{(r)} - \frac{1}{NT} \sum_{i=1}^n F^{(r)} \lambda_{ir} \lambda'_{ir} F^{(r)} \hat{F}^{(r)} \\
&= \frac{1}{NT} \sum_{i=1}^n X_i^{(r)} \hat{\delta}_r \hat{\delta}_r' X_i^{(r)'} \hat{F}^{(r)} - \frac{1}{NT} \sum_{i=1}^n X_i^{(r)} \hat{\delta}_r \lambda'_{ir} F^{(r)'} \hat{F}^{(r)} - \frac{1}{NT} \sum_{i=1}^n X_i^{(r)} \hat{\delta}_r \varepsilon_i^{(r)'} \hat{F}^{(r)} \\
&\quad - \frac{1}{NT} \sum_{i=1}^n F^{(r)} \lambda_{ir} \hat{\delta}_r' X_i^{(r)'} \hat{F}^{(r)} - \frac{1}{NT} \sum_{i=1}^n \varepsilon_i^{(r)} \hat{\delta}_r' X_i^{(r)'} \hat{F}^{(r)} - \frac{1}{NT} \sum_{i=1}^n X_i^{(r)} \hat{\delta}_r' \Delta_i^{(r)'} \hat{F}^{(r)} - \frac{1}{NT} \sum_{i=1}^n \Delta_i^{(r)} \hat{\delta}_r' X_i^{(r)'} \hat{F}^{(r)} \\
&\quad + \frac{1}{NT} \sum_{i=1}^n F^{(r)} \lambda_{ir} \varepsilon_i^{(r)'} \hat{F}^{(r)} + \frac{1}{NT} \sum_{i=1}^n \varepsilon_i^{(r)} \lambda'_{ir} F^{(r)'} \hat{F}^{(r)} + \frac{1}{NT} \sum_{i=1}^n \varepsilon_i^{(r)} \varepsilon_i^{(r)'} \hat{F}^{(r)} + \frac{1}{NT} \sum_{i=1}^n F^{(r)} \lambda_{ir} \Delta_i^{(r)'} \hat{F}^{(r)} \\
&\quad + \frac{1}{NT} \sum_{i=1}^n \Delta_i^{(r)} \Delta_i^{(r)'} \hat{F}^{(r)} + \frac{1}{NT} \sum_{i=1}^n \varepsilon_i^{(r)} \Delta_i^{(r)'} \hat{F}^{(r)} + \frac{1}{NT} \sum_{i=1}^n \Delta_i^{(r)} \lambda'_{ir} F^{(r)'} \hat{F}^{(r)} + \frac{1}{NT} \sum_{i=1}^n \Delta_i^{(r)} \varepsilon_i^{(r)'} \hat{F}^{(r)} \\
&\equiv \sum_{s=1}^{15} I_s^{(r)}. \tag{C.3}
\end{aligned}$$

Pre-multiplying both sides of (C.3) by  $(N^{-1} \Lambda_r' \Lambda_r)^{1/2} T^{-1} F^{(r)'}$ , we have

$$\left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{1/2} \left( \frac{F^{(r)'} \hat{F}^{(r)}}{T} \right) \hat{V}_{NT}^{(r)} = \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{1/2} \left( \frac{F^{(r)'} F^{(r)}}{T} \right) \frac{\Lambda_r' \Lambda_r}{N} \left( \frac{F^{(r)'} \hat{F}^{(r)}}{T} \right) + d_{NT}^{(r)}, \tag{C.4}$$

where  $d_{NT}^{(r)} \equiv \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{1/2} \frac{1}{T} F^{(r)'} \sum_{s=1}^{15} I_s^{(r)}$ . We note that Su and Wang (2017, 2020) have shown that  $\left\| \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{1/2} \times \frac{1}{T} F^{(r)'} \sum_{s=8}^{15} I_s^{(r)} \right\| = O_P(T^{-1} h^{-1} + N^{-1/2} + h^2)$ . This, along with the asymptotic nonsingularity of  $\frac{\Lambda_r' \Lambda_r}{N}$ , implies that it suffices to study the order of  $\frac{1}{T} F^{(r)'} I_s^{(r)}$  for  $s \in [7]$ . Noting that  $\frac{1}{T} \left\| \hat{F}^{(r)} \right\|^2 = \text{tr} \left( \frac{1}{T} \hat{F}^{(r)'} \hat{F}^{(r)} \right) = R$ , we have by the triangle and CS inequalities,

$$\left\| \frac{1}{T} F^{(r)'} I_1^{(r)} \right\| \leq R^{1/2} \frac{1}{T^{1/2}} \left\| F^{(r)} \right\| \frac{1}{NT} \sum_{i=1}^n \|X_i^{(r)}\|^2 \left\| \hat{\delta}_r \right\|^2 = O_P(\|\hat{\delta}_r\|^2).$$

Similarly, we can readily show that  $\left\| \frac{1}{T} F^{(r)'} I_l^{(r)} \right\| = O_P(\|\hat{\delta}_r\|)$  for  $l = 2, 3, 4, 5$ . Next,

$$\left\| \frac{1}{T} F^{(r)'} I_6^{(r)} \right\| \leq R^{1/2} \left[ \frac{1}{N} \sum_{i=1}^n \left\| \frac{F^{(r)'} X_i^{(r)}}{T} \right\|^2 \right]^{1/2} \left[ \frac{1}{N} \sum_{i=1}^n \left\| \frac{\Delta_i^{(r)}}{T} \right\|^2 \right]^{1/2} \left\| \hat{\delta}_r \right\| = O_P(\|\hat{\delta}_r\| h)$$

where we use the fact that  $\frac{1}{NT} \sum_{i=1}^n \left\| \Delta_i^{(r)} \right\|^2 = \frac{1}{NT} \sum_{i=1}^n \sum_{t=1}^T \left\| \Delta_i^{(r)}(t, r) \right\|^2 = O_P(h^2)$ . Similarly, we can show that  $\left\| \frac{1}{T} F^{(r)'} I_7^{(r)} \right\| = O_P(\|\hat{\delta}_r\| h)$ . In sum, we have shown that  $d_{NT}^{(r)} = O_P(T^{-1} h^{-1} + N^{-1/2} + h^2) + O_P(\|\hat{\delta}_r\|)$ .

Now, letting

$$B_{NT}^{(r)} = \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{1/2} \left( \frac{F^{(r)'} F^{(r)}}{T} \right) \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{1/2} \quad \text{and} \quad R_{NT}^{(r)} = \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{1/2} \left( \frac{F^{(r)'} \hat{F}^{(r)}}{T} \right),$$

we can rewrite (C.4) as follows:  $[B_{NT}^{(r)} + d_{NT}^{(r)} R_{NT}^{(r)-1}] R_{NT}^{(r)} = R_{NT}^{(r)} \hat{V}_{NT}^{(r)}$ . Hence, each column of  $R_{NT}^{(r)}$  is a non-standardized eigenvector of the matrix  $B_{NT}^{(r)} + d_{NT}^{(r)} R_{NT}^{(r)-1}$ . Let  $\hat{V}_{NT}^{(r)}$  be a diagonal matrix consisting of the

diagonal elements of  $R_{NT}^{(r)'} R_{NT}^{(r)}$ . Denote the standardized eigenvector  $\Upsilon_{NT}^{(r)} = R_{NT}^{(r)} \tilde{V}_{NT}^{(r)-1/2}$ . Hence, we have  $[B_{NT}^{(r)} + d_{NT}^{(r)} R_{NT}^{(r)-1}] \Upsilon_{NT}^{(r)} = \Upsilon_{NT}^{(r)} \hat{V}_{NT}^{(r)}$ . That is,  $\hat{V}_{NT}^{(r)}$  contains the eigenvalues of  $B_{NT}^{(r)} + d_{NT}^{(r)} R_{NT}^{(r)-1}$  with the corresponding normalized eigenvectors contained in  $\Upsilon_{NT}^{(r)}$ . It is trivial to show that

$$\left\| B_{NT}^{(r)} + d_{NT}^{(r)} R_{NT}^{(r)-1} - B_r \right\| = O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_r\|), \quad (\text{C.5})$$

where  $B_r$  denotes the probability of  $B_{NT}^{(r)}$ , i.e.,  $B_r = \Sigma_{\Lambda_r}^{1/2} \Sigma_F \Sigma_{\Lambda_r}^{1/2}$ . By the perturbation theory for eigenvalues of Hermitian matrices (e.g., Stewart and Sun (1990, p. 203)),

$$\left| \mu_j(B_{NT}^{(r)} + d_{NT}^{(r)} R_{NT}^{(r)-1}) - \mu_j(B_r) \right| \leq \left\| B_{NT}^{(r)} + d_{NT}^{(r)} R_{NT}^{(r)-1} - B_r \right\| = O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_r\|),$$

where  $\mu_j(A)$  denotes the  $j$ th largest eigenvalue of a symmetric matrix  $A$  and  $j = 1, \dots, R$ . That is,  $\left\| V_{NT}^{(r)} - V_r \right\| = O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_r\|)$ .

(ii)-(iv) The proofs are essentially the same as those of Lemma A.1(ii)-(iv) in Su and Wang (2017) and thus omitted here.

(v) As in the proof of Proposition 3.1, we have  $\left\| P_{\hat{F}^{(r)}} - P_{F^{(r)}} \right\|^2 = 2\text{tr}(\mathbb{I}_R - T^{-1} \hat{F}^{(r)'} P_{F^{(r)}} \hat{F}^{(r)})$ . Using the decomposition that  $\hat{F}^{(r)} = (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)}) + F^{(r)} H^{(r)} + B^{(r)}$  and Lemma A.2 (iv) and Lemma A.3 (ii)-(iii) below, we can show that  $\mathbb{I}_R - T^{-1} \hat{F}^{(r)'} P_{F^{(r)}} \hat{F}^{(r)} = \mathbb{I}_R - T^{-1} \hat{F}^{(r)'} F^{(r)} (F^{(r)'} F^{(r)})^{-1} F^{(r)} \hat{F}^{(r)} = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|)$ . ■

**Proof of Lemma A.3.** (i) We consider the decomposition in (A.7). By CS inequality, we have

$$\frac{1}{T} \sum_{t=1}^T \left\| \hat{F}_t^{(r)} - H^{(r)'} F_t^{(r)} - B_t^{(r)} \right\|^2 \leq \left\| \hat{V}_{NT}^{(r)-1} \right\|^2 \frac{16}{T} \sum_{t=1}^T \sum_{l=1}^{16} \left\| \hat{V}_{NT}^{(r)} A_l(t, r) \right\|^2.$$

By Lemma A.2(i), we can bound  $T^{-1} \sum_{t=1}^T \|A_l(t, r)\|^2$  by determining the probability order of  $T^{-1} \sum_{t=1}^T \left\| \hat{V}_{NT}^{(r)} A_l(t, r) \right\|^2$ . The terms  $A_1(t, r)$  to  $A_9(t, r)$  are almost the same as those defined in the proof of Theorem 2.1 of Su and Wang (2020), except that their bias term  $D(s, r) F_s^{(r)}$  has been replaced by  $\Delta_s^{(r)} = X_s^{(r)} L(s, r) + D(s, r) F_s^{(r)}$  here. Following the proof of Lemma A.2 in Su and Wang (2017, 2020), we can show that  $\sum_{l=1}^9 T^{-1} \sum_{t=1}^T \left\| \hat{V}_{NT}^{(r)} A_l(t, r) \right\|^2 = O_P(C_{NT}^{-2})$ . Similar to the proof of Lemma A.2(i), it is easy to show that  $\sum_{l=10}^{15} T^{-1} \sum_{t=1}^T \left\| \hat{V}_{NT}^{(r)} A_l(t, r) \right\|^2 = O_P(\|\hat{\delta}_r\|)$ . Then  $T^{-1} \sum_{t=1}^T \left\| \hat{F}_t^{(r)} - H^{(r)'} F_t^{(r)} - B_t^{(r)} \right\|^2 = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|)$ .

(ii) By (A.7), we obtain  $\frac{1}{T} (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)})' F^{(r)} H^{(r)} = \hat{V}_{NT}^{(r)-1} \sum_{l=1}^{16} \bar{A}_l(r) H^{(r)}$ , where  $\bar{A}_l(r) = \frac{1}{T} \sum_{t=1}^T \hat{V}_{NT}^{(r)} A_l(t, r) F_t^{(r)'}$ . Su and Wang (2017, 2020) have studied the terms  $\bar{A}_l(r)$ ,  $l \in [9]$ . Following the results there, we have  $\sum_{l=1}^9 \bar{A}_l(r) = O_P(C_{NT}^{-2})$ . For  $\bar{A}_{10}(r)$  and  $\bar{A}_{11}(r)$ , we have

$$\begin{aligned} \left\| \bar{A}_{10}(r) \right\| &\leq R^{1/2} T^{-1/2} \left\| F^{(r)} \right\| \frac{1}{NT} \sum_{i=1}^N \|X_i^{(r)}\|^2 \|\hat{\delta}_r\|^2 = O_P(\|\hat{\delta}_r\|^2), \text{ and} \\ \left\| \bar{A}_{11}(r) \right\| &\leq R^{1/2} \left\{ \frac{1}{N^2 T} \sum_{s=1}^T \left\| X_s^{(r)'} \Lambda_r \right\|^2 \right\}^{1/2} \frac{1}{T} \sum_{t=1}^T \left\| F_t^{(r)} \right\|^2 \|\hat{\delta}_r\| = O_P(\|\hat{\delta}_r\|). \end{aligned}$$

Similarly, we can readily show that  $\left\| \bar{A}_l(r) \right\| = O_P(\|\hat{\delta}_r\|)$  for  $l = 12, 13$ , and  $14$ , and  $\left\| \bar{A}_{15}(r) \right\| = O_P(\|\hat{\delta}_r\| h)$  for  $l = 15$  and  $16$ . Combining the above results and using Lemma A.2(i) yield the claim in part (ii) of the lemma.

(iii) This follows from (i) and (ii) and the triangle inequality.

(iv)-(v) The proof is almost the same as part (ii) of the lemma and omitted for brevity. ■

**Proof of Lemma A.4.** The proof is similar to that of Lemma A.3 and we only sketch it here.

(i) Let  $\Upsilon^{(r)} = \hat{F}^{(r)} - F^{(r)}H^{(r)} - B^{(r)}$ . Then

$$\begin{aligned} & (NT)^{-1} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Upsilon^{(r)} \lambda'_{ir} \\ &= (NT)^{-1} \sum_{i=1}^N X_i^{(r)'} \Upsilon^{(r)} \lambda'_{ir} - (NT)^{-1} \sum_{i=1}^N X_i^{(r)'} P_{F^{(r)}} \Upsilon^{(r)} \lambda'_{ir} - (NT)^{-1} \sum_{i=1}^N X_i^{(r)'} (P_{\hat{F}^{(r)}} - P_{F^{(r)}}) \Upsilon^{(r)} \lambda'_{ir} \\ &\equiv II_1 - II_2 - II_3. \end{aligned}$$

Following the proof of Lemma A.3(ii), we can show that  $\|II_l\| = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|)$  for  $l = 1, 2$ . For  $II_3$ , we can readily apply Lemmas A.2(v) and A.3(i) to obtain

$$\|II_3\| \leq \left\{ N^{-1} T^{-1/2} \sum_{i=1}^N \left\| X_i^{(r)} \right\| \left\| \lambda_{ir} \right\| \right\} \|P_{\hat{F}^{(r)}} - P_{F^{(r)}}\| \frac{1}{T^{1/2}} \left\| \Upsilon^{(r)} \right\| = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|).$$

It follows that  $(NT)^{-1} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Upsilon^{(r)} \lambda'_{ir} = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|)$ .

(ii) Let  $\omega$  be an arbitrary nonrandom  $P$ -vector with  $\|\omega\| = 1$ . Note that  $\frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} B^{(r)} = \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} B^{(r)} - \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} P_{F^{(r)}} B^{(r)} - \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} (P_{\hat{F}^{(r)}} - P_{F^{(r)}}) B^{(r)} \equiv II_4 - II_5 - II_6$ . Recall that  $B_t^{(r)} = k_{h,tr}^{1/2} C_{1t}^{(r)} \frac{t-r}{T} + k_{h,tr}^{1/2} C_{2t}^{(r)} \left(\frac{t-r}{T}\right)^2 + k_{h,tr}^{1/2} C_{3:4,t}^{(r)} h^2 \kappa_2$  where  $C_{3:4,t}^{(r)} = C_{3t}^{(r)} + C_{4t}^{(r)}$ , we can readily show that

$$\begin{aligned} II_4 &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \lambda_{ir} \omega' X_{it}^{(r)} B_t^{(r)'} = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* \lambda_{ir} \omega' X_{it} \left[ C_{1t}^{(r)} \frac{t-r}{T} + C_{2t}^{(r)} \left(\frac{t-r}{T}\right)^2 + C_{3:4,t}^{(r)} h^2 \kappa_2 \right]' \\ &= \bar{II}_4 + O_P(h^2), \end{aligned}$$

where  $\bar{II}_4 = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* \lambda_{ir} X_{it} \omega' C_{1t}^{(r)'} \frac{t-r}{T}$ . Noting that  $C_{1t}^{(r)} = \hat{V}_{NT}^{(r)-1} H^{(r)'} \Sigma_F (\Lambda_r' A_{1,tr} / N)$  and  $A_{1,tr} = X_t \beta_r^{(1)} + \Lambda_r^{(1)} F_t$ , it is easy to see that the probability order of  $\bar{II}_4$  is determined by that of  $\bar{II}_{4,1}$  and  $\bar{II}_{4,2}$ , where

$$\begin{aligned} \bar{II}_{4,1} &= \frac{1}{NT} \sum_{t=1}^T k_{h,tr}^* \left( \frac{1}{N} \sum_{i=1}^N \lambda_{ir} X_{it}' \right) \omega \beta_r^{(1)'} X_t' \Lambda_r \frac{t-r}{T} \text{ and} \\ \bar{II}_{4,2} &= \frac{1}{NT} \sum_{t=1}^T k_{h,tr}^* \left( \frac{1}{N} \sum_{i=1}^N \lambda_{ir} X_{it}' \right) \omega F_t' \Lambda_r^{(1)} \frac{t-r}{T}. \end{aligned}$$

Using the fact that  $\frac{1}{N} \sum_{i=1}^N \lambda_{ir} X_{it}' = \frac{1}{N} \sum_{i=1}^N \lambda_{ir} E(X_{i1}') + O_P((N/\ln T)^{-1/2})$  uniformly in  $t$ , we can show that

$$\bar{II}_{4,1} = \frac{1}{N} \sum_{i=1}^N \lambda_{ir} E(X_{i1}') \omega \beta_r^{(1)'} \frac{1}{N} \sum_{j=1}^N \frac{1}{T} \sum_{t=1}^T k_{h,tr}^* X_{jt} \lambda'_{jr} \frac{t-r}{T} + O_P((N/\ln T)^{-1/2} h).$$

This, in conjunction with the fact that

$$\begin{aligned} \frac{1}{T} \sum_{t=1}^T k_{h,tr}^* X_{jt} \frac{t-r}{T} &= \frac{1}{T} \sum_{t=1}^T k_{h,tr}^* E(X_{j1}) \frac{t-r}{T} + \frac{1}{T} \sum_{t=1}^T k_{h,tr}^* [X_{jt} - E(X_{jt})] \frac{t-r}{T} \\ &= O(T^{-1}) + O_P((Th/\ln T)^{-1/2} h) \text{ uniformly in } j, \end{aligned}$$

implies that  $\bar{II}_{4,1} = O_P((N/\ln T)^{-1/2} h + (Th/\ln T)^{-1/2} h + T^{-1})$ . Similarly,  $\bar{II}_{4,2} = O_P((N/\ln T)^{-1/2} h +$

$(Th/\ln T)^{-1/2}h + T^{-1}$ ). Thus  $\bar{II}_4 = O_P((N/\ln T)^{-1/2}h + (Th/\ln T)^{-1/2}h + T^{-1})$  and

$$II_4 = O_P((N/\ln T)^{-1/2}h + (Th/\ln T)^{-1/2}h + T^{-1}) + O_P(h^2) = O_P(h^2).$$

Similarly, we can show that  $II_5 = O_P(h^2)$ . Following the analysis of  $II_3$  in the proof of (ii), we can show that

$$\begin{aligned} \|II_6\| &= \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} (P_{\hat{F}^{(r)}} - P_{F^{(r)}}) B^{(r)} \right\| \lesssim \|P_{\hat{F}^{(r)}} - P_{F^{(r)}}\| \frac{1}{NT^{1/2}} \sum_{i=1}^N \|X_i^{(r)}\| \frac{1}{T^{1/2}} \|B^{(r)}\| \\ &= O_P((C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})h). \end{aligned}$$

It follows that  $\frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} B^{(r)} = O_P(h^2) + O_P(\|\hat{\delta}_r\|)$  and the conclusion follows.

(iii)  $N^{-1}T^{-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} = N^{-1}T^{-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \Upsilon^{(r)} + N^{-1}T^{-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} F^{(r)} H^{(r)} + N^{-1}T^{-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} B^{(r)} \equiv II_7 + II_8 + II_9$ . Following the proof of Lemma A.3(ii), we can show that  $II_7 = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|)$ . For  $II_8$ , we have  $N^{-1}T^{-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} F^{(r)} = N^{-1}T^{-1} \sum_{j=1}^N \sum_{t=1}^T k_{tr}^* \lambda_{jr} \varepsilon_{jt} F_t = O_P((NT h)^{-1/2})$ , implying  $II_8 = O_P((NT h)^{-1/2})$ . Similarly,  $II_9 = O_P((NT h)^{-1/2}h)$ . Consequently,

$$N^{-1}T^{-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|) + O_P((NT h)^{-1/2}) = O_P(C_{NT}^{-2}) + O_P(\|\hat{\delta}_r\|).$$

(iv) The result is related to Lemma A.4 of Bai (2009). Following his arguments, one expect that  $\frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} = (Th)^{-1/2} O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2})$  for each  $i$ . Here we prove the corresponding mean square error result. By (A.7),

$$\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\|^2 = \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T \Upsilon_t^{(r)} \varepsilon_{it}^{(r)} \right\|^2 \leq 16 \left\| \hat{V}_{NT}^{(r)-1} \right\| \sum_{l=1}^{16} A_l^{(r)},$$

where  $A_l^{(r)} = \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T \hat{V}_{NT}^{(r)} A_l(t, r) \varepsilon_{it}^{(r)} \right\|^2$  for  $l \in [16]$ . Following the proof of Lemma A.4 of Bai (2009) and the proof of Lemma A.3, and using the results in Lemmas A.2-A.3, we can readily show that  $\frac{1}{T} \sum_{t=1}^T \hat{V}_{NT}^{(r)} A_l(t, r) \varepsilon_{it}^{(r)} = O_P(C_{NT}^{-4} + \|\hat{\delta}_r\|^2)$  for  $l \in [9]$ . In addition, it is trivial to show that  $\frac{1}{T} \sum_{t=1}^T \hat{V}_{NT}^{(r)} A_l(t, r) \varepsilon_{it}^{(r)} = O_P(\|\hat{\delta}_r\|^2)$ . Then  $\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\|^2 = O_P(C_{NT}^{-4} + \|\hat{\delta}_r\|^2)$ .

(v) By (A.7) we have

$$\begin{aligned} & \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \frac{1}{T} X_i^{(r)'} F^{(r)} \left[ \frac{1}{T} F^{(r)'} F^{(r)} \right]^{-1} \left[ \hat{F}^{(r)} H^{(r)-1} - F^{(r)} - B^{(r)} H^{(r)-1} \right]' \varepsilon_i^{(r)} \\ &= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \frac{1}{T} X_i^{(r)'} F^{(r)} \left[ \frac{1}{T} F^{(r)'} F^{(r)} \right]^{-1} \left( H^{(r)-1} \right)' \Upsilon^{(r)'} \varepsilon_i^{(r)} \\ &= \sum_{l=1}^{15} \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \sum_{t=1}^T \frac{1}{T} X_i^{(r)'} F^{(r)} \left[ \frac{1}{T} F^{(r)'} F^{(r)} \right]^{-1} \left( H^{(r)-1} \right)' A_l(t, r) \varepsilon_{it}^{(r)} \equiv \sum_{l=1}^{15} \bar{A}_l^{(r)}. \end{aligned}$$

Let  $\omega$  be any nonrandom  $P$ -vector with  $\|\omega\| = 1$ . Then

$$\begin{aligned} \left| \omega' \bar{A}_1^{(r)} \right| &= \left| \sqrt{N}h \text{tr} \left\{ \frac{F^{(r)}}{T^{1/2}} \left( \frac{F^{(r)'} F^{(r)}}{T} \right)^{-1} \left( H^{(r)-1} \right)' \hat{V}_{NT}^{(r)-1} \frac{1}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T \hat{F}_s^{(r)} E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\} \right| \\ &\lesssim \sqrt{N}h \left\| \frac{1}{T^2} \sum_{t=1}^T \sum_{s=1}^T \hat{F}_s^{(r)} E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\| \end{aligned}$$

$$\begin{aligned}
&\lesssim \sqrt{Nh} \left\| \frac{1}{T^2} \sum_{t=1}^T \sum_{s=1}^T F_s^{(r)} E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\| \\
&\quad + \sqrt{Nh} \left\| \frac{1}{T^2} \sum_{t=1}^T \sum_{s=1}^T B_s^{(r)} E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\| \\
&\quad + \sqrt{Nh} \left\| \frac{1}{T^2} \sum_{t=1}^T \sum_{s=1}^T (\hat{F}_s^{(r)} - H^{(r)'} F_s^{(r)} - B_s^{(r)}) E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\| \\
&\equiv II_{10} + II_{11} + II_{12}.
\end{aligned}$$

As in Su and Wang (2017), we can readily show that  $\frac{1}{T^2} \sum_{t=1}^T \left\| \sum_{s=1}^T F_s^{(r)} E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \right\|^2 = \frac{1}{T}$  and  $\frac{1}{T^2} \sum_{t=1}^T \left\| \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\|^2 = O_P((Th)^{-1} + N^{-1})$  by Markov inequality under Assumptions A.1(vi) and A.2(iii) as

$$\begin{aligned}
&\frac{1}{T^2} \sum_{t=1}^T E \left\| \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\|^2 \\
&= \frac{1}{N^2 T^2} \sum_{i,j=1}^N \sum_{s,t=1}^T k_{h,tr}^* k_{h,sr}^* E\{[\varepsilon_{it} \varepsilon_{jt} - E(\varepsilon_{it} \varepsilon_{jt})] X'_{is} X_{js}\} + \frac{1}{N^2 T^2} \sum_{i,j=1}^N \sum_{s,t=1}^T k_{h,tr}^* k_{h,sr}^* E(\varepsilon_{it} \varepsilon_{jt}) E(X'_{is} X_{js}) \\
&\lesssim h^{-1} \max_{i,j,s} \sum_{s=1}^T |E\{[\varepsilon_{it} \varepsilon_{jt} - E(\varepsilon_{it} \varepsilon_{jt})] X'_{is} X_{js}\}| \frac{1}{T^2} \sum_{t=1}^T k_{h,tr}^* + \max_t \frac{1}{N^2} \sum_{i,j=1}^N |E(\varepsilon_{it} \varepsilon_{jt})| \frac{1}{T^2} \sum_{s,t=1}^T k_{h,tr}^* k_{h,sr}^* \\
&= O((Th)^{-1} + N^{-1}).
\end{aligned}$$

Then

$$\begin{aligned}
II_{10} &\leq \sqrt{Nh} \left\{ \frac{1}{T^2} \sum_{t=1}^T \left\| \sum_{s=1}^T F_s^{(r)} E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \right\|^2 \right\}^{1/2} \left\{ \frac{1}{T^2} \sum_{t=1}^T \left\| \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\|^2 \right\}^{1/2} \\
&= \sqrt{Nh} O_P(T^{-1/2}) O_P((Th)^{-1/2} + N^{-1/2}) = o_P(1).
\end{aligned}$$

Similarly,  $II_{11} = o_P(1)$ . For  $II_{12}$ , we have by CS equality and Lemma A.3(i),

$$\begin{aligned}
II_{12} &\leq \sqrt{Nh} \left\| \frac{1}{T^2} \sum_{t=1}^T \sum_{s=1}^T (\hat{F}_s^{(r)} - H^{(r)'} F_s^{(r)} - B_s^{(r)}) E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \frac{1}{N} \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\| \\
&\leq \left\{ \frac{1}{T} \sum_{s=1}^T \left\| \hat{F}_s^{(r)} - H^{(r)'} F_s^{(r)} - B_s^{(r)} \right\|^2 \right\}^{1/2} \max_t \left\{ \left\| \sum_{s=1}^T E(\varepsilon_s^{(r)'} \varepsilon_t^{(r)}/N) \right\|^2 \right\}^{1/2} \\
&\quad \times \left\| \frac{\sqrt{Nh}}{NT^{3/2}} \sum_{t=1}^T \sum_{i=1}^N \varepsilon_{it}^{(r)} \omega' X_i^{(r)'} \right\| \\
&= O_P(\|\hat{\delta}_r\|^{1/2} + C_{NT}^{-1}) O(1) O_P(1) = o_P(1).
\end{aligned}$$

Then  $\bar{A}_1^{(r)} = o_P(1)$ . Similarly, we can show that  $\bar{A}_l^{(r)} = o_P(1)$  for  $l = 2, 4, 5, \dots, 9$ . For  $\bar{A}_3^{(r)}$ , in view of the fact that  $(H^{(r-1)})' \hat{V}_{NT}^{(r-1)} = (N^{-1} \Lambda_r' \Lambda_r)^{-1} (T^{-1} \hat{F}^{(r)'} F^{(r)})^{-1}$  by the definition of  $H^{(r)}$ , we have

$$\bar{A}_3^{(r)} = \frac{\sqrt{NTh}}{NT} \sum_{i=1}^N \sum_{t=1}^T \frac{X_i^{(r)'} F^{(r)}}{T} \left( \frac{F^{(r)'} F^{(r)}}{T} \right)^{-1} \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{-1} \Lambda_r' \varepsilon_t^{(r)}/N \varepsilon_{it}^{(r)} = \sqrt{\frac{Th}{N}} \psi_{NT}^{(r)}$$

where  $\psi_{NT}^{(r)} = \frac{1}{N} \sum_{i=1}^N \sum_{k=1}^N \frac{X_i^{(r)'} F^{(r)}}{T} \left( \frac{F^{(r)'} F^{(r)}}{T} \right)^{-1} \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{-1} \lambda_{kr} \left( \frac{1}{T} \sum_{t=1}^T \varepsilon_{it}^{(r)} \varepsilon_{kt}^{(r)} \right)$ .

Next, it is easy to obtain the rough bound for the remainder terms:

$$\begin{aligned}\bar{A}_{10}^{(r)} &= \sqrt{NT}hO_P(\|\hat{\delta}_r\|^2(Nh)^{-1/2}), \quad \bar{A}_l^{(r)} = \sqrt{NT}hO_P(\|\hat{\delta}_r\|(Nh)^{-1/2}) \text{ for } l = 11, 13, 14 \text{ and } 16, \text{ and} \\ \bar{A}_l^{(r)} &= \sqrt{NT}hO_P(\|\hat{\delta}_r\|h(Nh)^{-1/2}) \text{ for } l = 12, 15.\end{aligned}$$

In sum, we have  $\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \frac{X_i^{(r)'} F^{(r)}}{T} \left( \frac{F^{(r)'} F^{(r)}}{T} \right)^{-1} [\hat{F}^{(r)} H^{(r)-1} - F^{(r)} - B^{(r)} H^{(r)-1}]' \varepsilon_i^{(r)} = \sqrt{\frac{Th}{N}} \psi_{NT}^{(r)} + \sqrt{T} O_P(\|\hat{\delta}_r\|) + o_P(1)$ . ■

**Proof of Lemma A.5.** Recall that  $H^{(r)} = (N^{-1} \Lambda_r' \Lambda_r) (T^{-1} F^{(r)'} \hat{F}^{(r)}) \hat{V}_{NT}^{(r)-1}$ . By (C.3),  $\hat{F}^{(r)} H^{(r)-1} - F^{(r)} = \sum_{l=1}^{15} I_l^{(r)} G^{(r)}$ , where  $G^{(r)} = (\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} (\frac{1}{N} \Lambda_r' \Lambda_r)^{-1}$  satisfies  $\|G^{(r)}\| = O_P(1)$  as  $(T^{-1} \hat{F}^{(r)'} F^{(r)})^{-1} = Q_r^{-1} + o_P(1)$  by Lemma A.2(ii) and  $N^{-1} \Lambda_r' \Lambda_r = \Sigma_{\Lambda_r} + O(N^{-1/2})$  by Assumption 3(i). It follows that

$$\begin{aligned}\frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} F^{(r)} \lambda_{ir} &= \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} [F^{(r)} - \hat{F}^{(r)} H^{(r)-1}] \lambda_{ir} \\ &= - \sum_{l=1}^{15} \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} I_l^{(r)} G^{(r)} \lambda_{ir} \equiv \sum_{l=1}^{15} J_l^{(r)}.\end{aligned}$$

For  $J_1^{(r)}$ , we have

$$\|J_1^{(r)}\| = \left\| \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} I_1^{(r)} G^{(r)} \lambda_{ir} \right\| \lesssim \frac{1}{NT^{1/2}} \sum_{i=1}^N \|X_i^{(r)}\| T^{-1/2} \|I_1^{(r)}\| \max_i \|\lambda_{ir}\| = O_P(\|\hat{\delta}_r\|^2),$$

where we use the fact that  $G^{(r)} = O_P(1)$ ,  $\|M_{\hat{F}^{(r)}}\|_{\text{sp}} = 1$ ,  $\frac{1}{NT^{1/2}} \sum_{i=1}^N \|X_i^{(r)}\| = O_P(1)$ ,  $T^{-1/2} \|I_1^{(r)}\| = O_P(\|\hat{\delta}_r\|^2)$ , and  $\max_i \|\lambda_{ir}\| = O(1)$ . For  $J_2^{(r)}$ , we have

$$J_2^{(r)} = \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^n X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} \hat{\delta}_r \lambda_{jr}' \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{-1} \lambda_{ir} = \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} a_{ji}^{(r)} \hat{\delta}_r,$$

where  $a_{ji}^{(r)} \equiv \lambda_{jr}' \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{-1} \lambda_{ir}$ . For  $J_3^{(r)}$ , we have

$$\begin{aligned}J_3^{(r)} &= \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} \hat{\delta}_r \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &= \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} \hat{\delta}_r \sum_{t=1}^T [\Upsilon_t^{(r)} + H^{(r)'} F_t^{(r)} + B_t^{(r)}] \varepsilon_{jt}^{(r)} G^{(r)} \lambda_{ir} \equiv \sum_{l=1}^3 J_{3l}^{(r)},\end{aligned}$$

where  $\Upsilon_t^{(r)} \equiv \hat{F}_t^{(r)} - H^{(r)'} F_t^{(r)} - B_t^{(r)}$ ,  $J_{31}^{(r)} = \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} \hat{\delta}_r \sum_{t=1}^T \Upsilon_t^{(r)} \varepsilon_{jt}^{(r)} G^{(r)} \lambda_{ir}$ , and  $J_{32}^{(r)}$  and  $J_{33}^{(r)}$  are similarly defined. Note that by Lemma A.3(i)

$$\begin{aligned}\|J_{31}^{(r)}\| &= \left\| \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} \hat{\delta}_r \sum_{t=1}^T \Upsilon_t^{(r)} \varepsilon_{jt}^{(r)} G^{(r)} \lambda_{ir} \right\| \\ &\lesssim \left\{ \frac{1}{T} \sum_{t=1}^T \|\Upsilon_t^{(r)}\|^2 \right\}^{1/2} \max_j \left\{ \frac{1}{T} \sum_{t=1}^T \|\varepsilon_{jt}^{(r)}\|^2 \right\}^{1/2} \frac{1}{NT} \sum_{i=1}^N \|X_i^{(r)}\|^2 \|\hat{\delta}_r\| \\ &= O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|).\end{aligned}$$



Similarly, we have

$$\begin{aligned}\|J_{32}^{(r)}\| &\lesssim \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{T} \sum_{t=1}^T F_t^{(r)} \varepsilon_{jt}^{(r)} \right\|^2 \right\}^{1/2} \|\hat{\delta}_r\| = O_P((Th)^{-1/2}) \|\hat{\delta}_r\| \text{ and} \\ \|J_{33}^{(r)}\| &\lesssim \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{T} \sum_{t=1}^T B_t^{(r)} \varepsilon_{jt}^{(r)} \right\|^2 \right\}^{1/2} \|\hat{\delta}_r\| = O_P((Th)^{-1/2} h) \|\hat{\delta}_r\|\end{aligned}$$

by straightforward moment calculations and Markov inequality. It follows that  $J_3^{(r)} = O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|)$ . Next, we consider  $J_4$  term:

$$\begin{aligned}J_4^{(r)} &= \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \frac{1}{NT} \sum_{j=1}^N F^{(r)} \lambda_{jr} \hat{\delta}_r' X_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &= \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} [F^{(r)} H^{(r)} - \hat{F}^{(r)} + B^{(r)}] H^{(r)-1} \lambda_{jr} \hat{\delta}_r' X_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &\quad - \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} B^{(r)} H^{(r)-1} \lambda_{jr} \hat{\delta}_r' X_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \equiv J_{41}^{(r)} - J_{42}^{(r)}.\end{aligned}$$

Noting that  $T^{-1/2} \left\| F^{(r)} - \hat{F}^{(r)} H^{(r)-1} + B^{(r)} \right\| = O_P(C_{NT}^{-1}) + O_P(\|\hat{\delta}_r\|^{1/2})$  and  $\|B_t^{(r)}\| = O_P(h)$  when  $|t-r| \lesssim h$ , one can readily show that  $J_4^{(r)} = O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2} + h) O_P(\|\hat{\delta}_r\|)$ .

For  $J_5^{(r)}$ , we have

$$\begin{aligned}|\omega' J_5^{(r)}| &= \left| \frac{1}{NT} \sum_{i=1}^N \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} \frac{1}{NT} \sum_{j=1}^N \varepsilon_j^{(r)} \hat{\delta}_r' X_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \right| \\ &= \left| \text{tr} \left( \hat{\delta}_r' \frac{1}{NT} \sum_{j=1}^N X_j^{(r)'} \hat{F}^{(r)} G^{(r)} \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \right) \right| \lesssim \left\{ \frac{1}{NT} \sum_{j=1}^N \|X_j^{(r)}\|^2 \right\}^{1/2} \left\{ \bar{J}_5^{(r)} \right\}^{1/2} \|\hat{\delta}_r\|,\end{aligned}$$

where  $\bar{J}_5^{(r)} = \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \right\|^2$ . For  $\bar{J}_5^{(r)}$ , we have

$$\begin{aligned}\bar{J}_5^{(r)} &\leq \frac{3}{N} \sum_{j=1}^N \left\{ \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} \varepsilon_j^{(r)} \right\|^2 + \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} P_{F^{(r)}} \varepsilon_j^{(r)} \right\|^2 + \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} (P_{\hat{F}^{(r)}} - P_{F^{(r)}}) \varepsilon_j^{(r)} \right\|^2 \right\} \\ &\equiv 3(II_{10} + II_{11} + II_{12}).\end{aligned}$$

By moment calculations, we can show that  $II_l = O_P((Th)^{-1})$  for  $l = 10, 11$ . For  $II_{12}$ , we can apply Lemma A.2(v) to obtain  $II_{12} = O_P(C_{NT}^{-2} + \|\hat{\delta}_r\|)$ . It follows that  $J_5^{(r)} = O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|)$ . Next,

$$\begin{aligned}J_6^{(r)} &= \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \frac{1}{NT} \sum_{j=1}^N X_j^{(r)} \hat{\delta}_r \Delta_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &= \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} \hat{\delta}_r \sum_{t=1}^T \left\{ (\beta_t - \beta_r)' X_{jt}^{(r)} + (\lambda_{jt} - \lambda_{jr})' F_t^{(r)} \right\} \hat{F}_t^{(r)'} G^{(r)} \lambda_{ir} \equiv J_{61}^{(r)} + J_{62}^{(r)}.\end{aligned}$$

It is straightforward to show that  $J_{6l}^{(r)} = O_P(h^2)O_P(\|\hat{\delta}_r\|)$  for  $l = 1, 2$ . Then  $J_{6l}^{(r)} = O_P(h^2)O_P(\|\hat{\delta}_r\|)$ . Similarly, we can show that  $J_7^{(r)} = O_P(h^2)O_P(\|\hat{\delta}_r\|)$ . In sum, we have established that

$$\sum_{l=1}^7 J_l^{(r)} = \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} a_{ji}^{(r)} \hat{\delta}_r + O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|).$$

The last nine terms, namely,  $J_8^{(r)}$  to  $J_{15}^{(r)}$ , do not explicitly depend on  $\hat{\delta}_r \equiv \hat{\beta}_r - \beta_r$ . We will show that the term  $J_9^{(r)}$  contributes to the limiting distribution of  $\hat{\beta}_r - \beta_r$ , the terms  $J_{10}^{(r)}$  and  $J_{14}^{(r)}$  contribute to the bias, and the other terms are  $o_P((NT)h^{-1/2}) + O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})O_P(\|\hat{\delta}_r\|)$ . We first consider  $J_8^{(r)}$ . Using  $M_{\hat{F}^{(r)}} F^{(r)} = M_{\hat{F}^{(r)}} [F^{(r)} H^{(r)} - \hat{F}^{(r)} + B^{(r)}] H^{(r)-1} - M_{\hat{F}^{(r)}} B^{(r)} H^{(r)-1} = -M_{\hat{F}^{(r)}} \Upsilon^{(r)} H^{(r)-1} - M_{\hat{F}^{(r)}} B^{(r)} H^{(r)-1}$ , we have

$$\begin{aligned} \omega' J_8^{(r)} &= -\frac{1}{N^2 T^2} \sum_{i=1}^N \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} \sum_{j=1}^N F^{(r)} \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &= \text{tr} \left( \frac{1}{NT} H^{(r)-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} \Upsilon^{(r)} \right) \\ &\quad + \text{tr} \left( \frac{1}{NT} H^{(r)-1} \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} B^{(r)} \right) \equiv J_{81}^{(r)} + J_{82}^{(r)}. \end{aligned}$$

By Lemma A.4(i)-(iii),

$$\left\| \lambda_{ir} \omega' J_{81}^{(r)} \right\| \lesssim \frac{1}{NT} \left\| \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Upsilon^{(r)} \right\| \frac{1}{NT} \left\| \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} \right\| = \left[ O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2}) \right] \left[ O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2}) \right],$$

and

$$\left\| J_{82}^{(r)} \right\| \lesssim \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} B^{(r)} \right\| \frac{1}{NT} \left\| \sum_{j=1}^N \lambda_{jr} \varepsilon_j^{(r)'} \hat{F}^{(r)} \right\| = O_P(h^2 + \|\hat{\delta}_r\|) \left[ O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2}) \right].$$

Then  $\left\| J_8^{(r)} \right\| = O_P(\|\hat{\delta}_r\| + C_{NT}^{-2} + h^2) \left[ O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2}) \right]$ . For  $J_9$ , noting that  $F^{(r)'} \hat{F}^{(r)} G^{(r)} = (\frac{1}{N} \Lambda_r' \Lambda_r)^{-1}$ , we have

$$J_9^{(r)} = -\frac{1}{N^2 T^2} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \sum_{j=1}^N \varepsilon_j^{(r)} \lambda_{jr}' \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{-1} \lambda_{ir} = -\frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N a_{ij}^{(r)} X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)}.$$

For  $J_{10}^{(r)}$ , we have

$$\begin{aligned} J_{10}^{(r)} &= -\frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &= -\frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} \left( \frac{1}{T} F^{(r)'} \hat{F}^{(r)} \right)^{-1} \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{-1} \lambda_{ir}. \end{aligned}$$

For  $J_{11}^{(r)}$ , we make the following decomposition:

$$\begin{aligned}
J_{11}^{(r)} &= -\frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} F^{(r)} \lambda_{jr} \Delta_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\
&= \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \left[ \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)} \right] H^{(r)-1} \lambda_{jr} \Delta_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\
&\quad + \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} B^{(r)} H^{(r)-1} \lambda_{jr} \Delta_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \equiv J_{11,1}^{(r)} + J_{11,2}^{(r)}.
\end{aligned}$$

By Lemma A.4(i),

$$\left| \omega' J_{11,1}^{(r)} \right| \lesssim \frac{1}{NT} \left\| \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} M_{\hat{F}^{(r)}} \Upsilon^{(r)} \right\| \left\| \frac{1}{NT} \sum_{j=1}^N \lambda_{jr} \Delta_j^{(r)'} \hat{F}^{(r)} \right\| = [O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2})] O_P(\|\hat{\delta}_r\| + C_{NT}^{-1}),$$

where we use the fact that

$$\begin{aligned}
\left\| \frac{1}{NT} \sum_{j=1}^N \lambda_{jr} \Delta_j^{(r)'} \hat{F}^{(r)} \right\| &\leq \left\| \frac{1}{NT} \sum_{j=1}^N \lambda_{jr} \Delta_j^{(r)'} \Upsilon^{(r)} \right\| + \left\| \frac{1}{NT} \sum_{j=1}^N \lambda_{jr} \Delta_j^{(r)'} (F^{(r)} H^{(r)} + B^{(r)}) \right\| \\
&= O_P(h(\|\hat{\delta}_r\|^{1/2} + C_{NT}^{-1})) + O_P(h^2) = O_P(\|\hat{\delta}_r\| + h^2 + hC_{NT}^{-1}) = O_P(\|\hat{\delta}_r\| + C_{NT}^{-1}).
\end{aligned}$$

For  $J_{11,2}^{(r)}$ , we can apply Lemmas A.2(iv) and A.3 to show that

$$\begin{aligned}
J_{11,2}^{(r)} &= \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} B^{(r)} H^{(r)-1} \lambda_{jr} \Delta_j^{(r)'} \left( F^{(r)} H^{(r)} + B^{(r)} \right) G^{(r)} \lambda_{ir} \\
&\quad - \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} P_{F^{(r)}} B^{(r)} H^{(r)-1} \lambda_{jr} \Delta_j^{(r)'} \left( F^{(r)} H^{(r)} + B^{(r)} \right) G^{(r)} \lambda_{ir} \\
&\quad + O_P((C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})h^3) + O_P((C_{NT}^{-2} + \|\hat{\delta}_r\|)h^2) \\
&\equiv J_{11,21}^{(r)} + J_{11,22}^{(r)} + O_P((C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})h^3) + O_P((C_{NT}^{-2} + \|\hat{\delta}_r\|)h^2).
\end{aligned}$$

Note that

$$\left| \omega' J_{11,21}^{(r)} \right| \lesssim \left\| \frac{1}{NT} \sum_{j=1}^N \lambda_{jr} \Delta_j^{(r)'} (F^{(r)} H^{(r)} + B^{(r)}) \right\| \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} X_i^{(r)'} B^{(r)} \right\| = O_P(h^2) O_P(h^2) = O_P(h^4).$$

Similarly,  $\left| \omega' J_{11,22}^{(r)} \right| \lesssim O_P(h^4)$ . Thus,

$$\begin{aligned}
J_{11}^{(r)} &= [O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-2})] O_P(\|\hat{\delta}_r\| + C_{NT}^{-1}) + O_P(h^4) + O_P((C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})h^3) + O_P((C_{NT}^{-2} + \|\hat{\delta}_r\|)h^2) \\
&= O_P(\|\hat{\delta}_r\|^2 + C_{NT}^{-3} + h^4 + C_{NT}^{-1}h^3 + \|\hat{\delta}_r\|^{1/2}h^3) = O_P(\|\hat{\delta}_r\|^2 + C_{NT}^{-3} + h^4 + C_{NT}^{-1}h^3),
\end{aligned}$$

where the last equality holds because  $\|\hat{\delta}_r\|^{1/2}h^3 \leq (\|\hat{\delta}_r\|h^2 + h^4)/2$  by CS inequality. Analogously, we have

$$J_{12}^{(r)} = -\frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_j^{(r)} \Delta_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir}$$

$$\begin{aligned}
&= \frac{-1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{F^{(r)}} \Delta_j^{(r)} \Delta_j^{(r)'} (F^{(r)} H^{(r)} + B^{(r)}) G^{(r)} \lambda_{ir} + O_P((C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})h^3 + O_P((C_{NT}^{-2} + \|\hat{\delta}_r\|)h^2)) \\
&= O_P(h^4) + O_P((C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})h^3 + O_P((C_{NT}^{-2} + \|\hat{\delta}_r\|)h^2)) \\
&= O_P(\|\hat{\delta}_r\|^2 + C_{NT}^{-3} + h^4 + C_{NT}^{-1}h^3 + \|\hat{\delta}_r\|^{1/2}h^3).
\end{aligned}$$

We note that the  $J_{13}^{(r)}$  term is a higher order term than  $J_9^{(r)}$ , and  $J_{15}^{(r)}$  term is a higher order term than  $J_8^{(r)}$ , which are both asymptotically negligible. Finally,

$$J_{14}^{(r)} = -\frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \frac{1}{T} \Delta_j^{(r)} \lambda_{jr}' F^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} = -\frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_j^{(r)} a_{ji}^{(r)}.$$

Collecting terms from  $J_1^{(r)}$  to  $J_{15}^{(r)}$  gives

$$\begin{aligned}
&\frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} F^{(r)} \lambda_{ir} \\
&= \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} X_j^{(r)} a_{ji}^{(r)} \hat{\delta}_r - \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N a_{ij}^{(r)} X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \\
&- \frac{1}{N^2 T^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} \left( \frac{1}{T} F^{(r)'} \hat{F}^{(r)} \right)^{-1} \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{-1} \lambda_{ir} - \frac{1}{N^2 T} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_j^{(r)} a_{ji}^{(r)} \\
&+ O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(\|\hat{\delta}_r\|) + O_P(C_{NT}^{-3} + h^4 + C_{NT}^{-1}h^3). \blacksquare
\end{aligned}$$

**Proof of Lemma A.6.** First, we consider  $\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} (M_{\hat{F}^{(r)}} - M_{F^{(r)}}) \varepsilon_i^{(r)}$ . Recall that  $\Upsilon^{(r)} = \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)}$ . Let  $\Upsilon_s^{(r)} = \hat{F}_s^{(r)} - H^{(r)'} F_s^{(r)} - B_s^{(r)}$ . Noting that  $\hat{F}^{(r)} = \Upsilon^{(r)} + F^{(r)} H^{(r)} + B^{(r)}$ , we make the following decomposition:

$$\begin{aligned}
&\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} (M_{F^{(r)}} - M_{\hat{F}^{(r)}}) \varepsilon_i^{(r)} \\
&= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} \left[ \hat{F}^{(r)} \hat{F}^{(r)'} - F^{(r)} (F^{(r)'} F^{(r)} / T)^{-1} F^{(r)'} \right] \varepsilon_i^{(r)} \\
&= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} \Upsilon^{(r)} \Upsilon^{(r)'} \varepsilon_i^{(r)} + \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} \Upsilon^{(r)} H^{(r)'} F^{(r)'} \varepsilon_i^{(r)} + \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} F^{(r)} H^{(r)} \Upsilon^{(r)'} \varepsilon_i^{(r)} \\
&+ \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} F^{(r)} \left[ H^{(r)} H^{(r)'} - (F^{(r)'} F^{(r)} / T)^{-1} \right] F^{(r)'} \varepsilon_i^{(r)} + \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} \Upsilon^{(r)} B^{(r)'} \varepsilon_i^{(r)} \\
&+ \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} B^{(r)} \Upsilon^{(r)} \varepsilon_i^{(r)} + \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} F^{(r)} H^{(r)} B^{(r)'} \varepsilon_i^{(r)} \\
&+ \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} B^{(r)} H^{(r)'} F^{(r)'} \varepsilon_i^{(r)} + \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} B^{(r)} B^{(r)'} \varepsilon_i^{(r)} \equiv \sum_{l=1}^9 A_l^{(r)}.
\end{aligned}$$

We consider these nine terms one by one. First, for  $A_1^{(r)}$ , we have

$$\|A_1^{(r)}\| \leq \sqrt{NT}h \left\| \frac{1}{N} \sum_{i=1}^N \frac{1}{T} X_i^{(r)'} \Upsilon^{(r)} \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\|$$

$$\begin{aligned}
&\leq \sqrt{NT}h \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} X_i^{(r)'} \Upsilon^{(r)} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\|^2 \right\}^{1/2} \\
&\leq \sqrt{NT}h \left\{ \frac{1}{NT} \sum_{i=1}^N \left\| X_i^{(r)} \right\|^2 \right\}^{1/2} \frac{1}{T^{1/2}} \left\| \Upsilon^{(r)} \right\| \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\|^2 \right\}^{1/2} \\
&= \sqrt{NT}h O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P(C_{NT}^{-2} + \|\hat{\delta}_r\|) = \sqrt{NT}h O_P(C_{NT}^{-3} + \|\hat{\delta}_r\|^{3/2}),
\end{aligned}$$

where the first equality follows from Lemmas A.3(i) and A.4(iii). For  $A_2^{(r)}$ , we have

$$\begin{aligned}
\left| \omega' A_2^{(r)} \right| &= \sqrt{NT}h \left| \frac{1}{NT} \sum_{i=1}^N \omega' X_i^{(r)'} \frac{1}{T} \Upsilon^{(r)} H^{(r)'} F^{(r)'} \varepsilon_i^{(r)} \right| \\
&= \sqrt{NT}h \left| \frac{1}{T} \sum_{s=1}^T \Upsilon_s^{(r)'} H^{(r)'} \left( \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T F_t^{(r)} \varepsilon_{it}^{(r)} \omega' X_{is}^{(r)} \right) \right| \\
&\lesssim \sqrt{NT}h \left[ \frac{1}{T} \sum_{s=1}^T \left\| \Upsilon_s^{(r)} \right\|^2 \right]^{1/2} \left[ \frac{1}{T} \sum_{s=1}^T k_{h,sr}^* \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* F_t \varepsilon_{it} \omega' X_{is} \right\|^2 \right]^{1/2} \\
&= \sqrt{NT}h O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2}) O_P((NT)h^{-1/2} + (Th)^{-1}),
\end{aligned}$$

where we use Lemma A.3(i) and the fact that

$$\begin{aligned}
&\frac{1}{T} \sum_{s=1}^T k_{h,sr}^* \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* F_t \varepsilon_{it} X'_{is} \omega \right\|^2 \\
&\leq \frac{2}{T} \sum_{s=1}^T k_{h,sr}^* \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* [F_t \varepsilon_{it} X'_{is} - E(F_t \varepsilon_{it} X'_{is})] \right\|^2 + \frac{2}{T} \sum_{s=1}^T k_{h,sr}^* \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(F_t \varepsilon_{it} X'_{is}) \right\|^2 \\
&= O_P((NT)h^{-1}) + O_P((Th)^{-2})
\end{aligned}$$

by noticing that  $\max_s \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(F_t \varepsilon_{it} X'_{is}) \right\|^2 \leq \max_t k_{h,tr}^{*2} \left| \max_s \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \|E(F_t \varepsilon_{it} X'_{is})\| \right|^2 \leq h^{-2} O_P(T^{-2})$ .

For  $A_3^{(r)}$ , we have

$$\begin{aligned}
A_3^{(r)} &= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{1}{T} F^{(r)} H^{(r)} \left[ \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)} \right]' \varepsilon_i^{(r)} \\
&= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \frac{X_i^{(r)'} F^{(r)}}{T} \left[ \frac{1}{T} F^{(r)'} F^{(r)} \right]^{-1} \left[ \hat{F}^{(r)} H^{(r)-1} - F^{(r)} - B^{(r)} H^{(r)-1} \right]' \varepsilon_i^{(r)} \\
&+ \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \frac{X_i^{(r)'} F^{(r)}}{T} \left\{ H^{(r)} H^{(r)'} - \left[ \frac{1}{T} F^{(r)'} F^{(r)} \right]^{-1} \right\} \left[ \hat{F}^{(r)} H^{(r)-1} - F^{(r)} - B^{(r)} H^{(r)-1} \right]' \varepsilon_i^{(r)} \equiv \sum_{\ell=1}^2 A_{3\ell}^{(r)}.
\end{aligned}$$

By Lemma A.4(iv),  $A_{31}^{(r)} = \sqrt{\frac{Th}{N}} \psi_{NT}^{(r)} + O_P(\|\hat{\delta}_r\|) + \sqrt{Th} O_P(C_{NT}^{-2})$ , where

$$\psi_{NT}^{(r)} = \frac{1}{N} \sum_{i=1}^N \sum_{k=1}^N \frac{X_i^{(r)'} F^{(r)}}{T} \left( \frac{F^{(r)'} F^{(r)}}{T} \right)^{-1} \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{-1} \lambda_{kr} \left( \frac{1}{T} \sum_{t=1}^T \varepsilon_{it}^{(r)} \varepsilon_{kt}^{(r)} \right).$$

Let  $\Xi^{(r)} = H^{(r)}H^{(r)'} - [\frac{1}{T}F^{(r)'}F^{(r)}]^{-1}$ . Note that

$$\begin{aligned} \frac{1}{T}F^{(r)'}F^{(r)} - \Sigma_F &= \frac{1}{T} \sum_{t=1}^T k_{h,tr}^* F_t F_t' - \Sigma_F = \frac{1}{T} \sum_{t=1}^T k_{h,tr}^* [F_t F_t' - E(F_t F_t')] + (\frac{1}{T} \sum_{t=1}^T k_{h,tr}^* - 1) \Sigma_F \\ &= O_P((Th)^{-1/2}) + O((Th)^{-1}) = O_P((Th)^{-1/2}) \end{aligned}$$

by our assumptions and the approximation property of Riemann summation to a definite integral. This fact, along with Lemma A.2(iv), implies that

$$\|\Xi^{(r)}\| = O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|). \quad (\text{C.6})$$

Then we have

$$\begin{aligned} \|A_{32}^{(r)}\| &\lesssim \sqrt{NTh} \|\Xi^{(r)}\| \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{X_i^{(r)'} F^{(r)}}{T} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\|^2 \right\}^{1/2} \\ &= \sqrt{NTh} O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|) O_P(1) O_P(C_{NT}^{-2} + \|\hat{\delta}_r\|) = \sqrt{NTh} O_P(C_{NT}^{-3} + \|\hat{\delta}_r\|^2 + C_{NT}^{-1} \|\hat{\delta}_r\|) \end{aligned}$$

by the CS inequality and A.4(iv). Thus  $A_3^{(r)} = \sqrt{\frac{Th}{N}} \psi_{NT}^{(r)} + O_P(\|\hat{\delta}_r\|) + \sqrt{Th} O_P(C_{NT}^{-2}) + \sqrt{NTh} O_P(C_{NT}^{-3} + \|\hat{\delta}_r\|^2 + C_{NT}^{-1} \|\hat{\delta}_r\|)$ .

For  $A_4^{(r)}$ , we have

$$\begin{aligned} |\omega' A_4^{(r)}| &= \text{tr} \left[ \Xi^{(r)} \left( \frac{\sqrt{NTh}}{NT^2} \sum_{i=1}^N F^{(r)'} \varepsilon_i^{(r)} \omega' X_i^{(r)'} F^{(r)} \right) \right] \leq \|\Xi^{(r)}\| \frac{\sqrt{NTh}}{NT^2} \left\| \sum_{i=1}^N F^{(r)'} \varepsilon_i^{(r)} \omega' X_i^{(r)'} F^{(r)} \right\| \\ &= O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|) O_P(1 + (Th/N)^{-1/2}). \end{aligned}$$

For  $A_5^{(r)}$ , we have

$$\begin{aligned} \|A_5^{(r)}\| &= \frac{\sqrt{NTh}}{NT^2} \left\| \sum_{i=1}^N X_i^{(r)'} \Upsilon^{(r)} B^{(r)'} \varepsilon_i^{(r)} \right\| \leq \sqrt{NTh} \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} X_i^{(r)'} \Upsilon^{(r)} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} B^{(r)'} \varepsilon_i^{(r)} \right\|^2 \right\}^{1/2} \\ &= \sqrt{NTh} O_P(C_{NT}^{-2} + \|\hat{\delta}_r\|) O_P((T/h)^{-1/2}). \end{aligned}$$

Next,

$$\begin{aligned} \|A_6^{(r)}\| &= \frac{\sqrt{NTh}}{NT^2} \left\| \sum_{i=1}^N X_i^{(r)'} B^{(r)} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\| \leq \sqrt{NTh} \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} X_i^{(r)'} B^{(r)} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \Upsilon^{(r)'} \varepsilon_i^{(r)} \right\|^2 \right\}^{1/2} \\ &= \sqrt{NTh} O_P(h) O_P(C_{NT}^{-2} + \|\hat{\delta}_r\|). \end{aligned}$$

For  $A_7^{(r)}$ , we have

$$\begin{aligned} |\omega' A_7^{(r)}| &= \frac{\sqrt{NTh}}{NT^2} \left| \text{tr} \left( H^{(r)} \sum_{i=1}^N B^{(r)'} \varepsilon_i^{(r)} \omega' X_i^{(r)'} F^{(r)} \right) \right| \lesssim \frac{\sqrt{NTh}}{NT^2} \left\| \sum_{i=1}^N B^{(r)'} \varepsilon_i^{(r)} \omega' X_i^{(r)'} F^{(r)} \right\| \\ &= \frac{\sqrt{NTh}}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T B_t^{(r)'} \varepsilon_{it}^{(r)} F_s^{(r)'} X_{is}^{(r)'} \omega = O_P(N^{1/2} T^{-1/2} h) = o_P(1), \end{aligned}$$

where we use the fact that  $B_t^{(r)} = k_{h,tr}^{*1/2} \left[ C_{1t}^{(r)} \frac{t-r}{T} + C_{2t}^{(r)} h^2 + C_{3t}^{(r)} \left( \frac{t-r}{T} \right)^2 \right]$ . Similarly,  $A_8^{(r)} = O_P(N^{1/2}T^{-1/2}h) = o_P(1)$ . Finally, for  $A_9^{(r)}$ , we can readily show that

$$\begin{aligned} A_9^{(r)} &= \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N X_i^{(r)'} B^{(r)} B^{(r)'} \varepsilon_i^{(r)} = \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T X_{ir}^{(r)} B_t^{(r)'} B_s^{(r)} \varepsilon_{is}^{(r)} \\ &= \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,tr}^* k_{h,rs}^* X_{it} C_{1t}^{(r)'} C_{1s}^{(r)} \varepsilon_{is} \frac{t-r}{T} \frac{s-r}{T} + o_P(1) \equiv A_{91}^{(r)} + o_P(1). \end{aligned}$$

Recall that  $C_{1t}^{(r)} = \hat{V}_{NT}^{(r)-1} H^{(r)'} \Sigma_F \left( \frac{1}{N} \Lambda_r' A_{1,tr} \right)$  and  $A_{1,tr} = X_t \beta_r^{(1)} + \Lambda_r^{(1)} F_t$ . Let  $\bar{C}_{1t}^{(r)} = \hat{V}_{NT}^{(r)-1} H^{(r)'} \Sigma_F E \left( \frac{1}{N} \Lambda_r' A_{1,tr} \right)$ . Then we can show that  $A_{91}^{(r)} = \bar{A}_{91}^{(r)} + o_P(1)$ , where

$$\bar{A}_{91}^{(r)} = \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,tr}^* k_{h,rs}^* E(X_{it} \bar{C}_{1t}^{(r)'} \bar{C}_{1s}^{(r)}) \varepsilon_{is} \frac{t-r}{T} \frac{s-r}{T} = \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \sum_{s=1}^T k_{h,rs}^* A_{iT}^{(r)} \bar{C}_{1s}^{(r)} \varepsilon_{is} \frac{s-r}{T}$$

and  $A_{iT}^{(r)} = \frac{1}{T} \sum_{t=1}^T k_{h,tr}^* E(X_{it} \bar{C}_{1t}^{(r)'} \frac{t-r}{T}) = O(h)$ . By straightforward moment calculations, we have  $E(\bar{A}_{91}^{(r)}) = 0$  and  $E \left\| \bar{A}_{91}^{(r)} \right\|^2 = O(h^4)$ . Then  $\bar{A}_{91}^{(r)} = O_P(h^2)$  and  $A_9^{(r)} = o_P(1)$ .

In sum, we have

$$\begin{aligned} &\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} (M_{F^{(r)}} - M_{\hat{F}^{(r)}}) \varepsilon_i^{(r)} \\ &= \sqrt{\frac{Th}{N}} \psi_{NT}^{(r)} + \sqrt{NT}h O_P((C_{NT}^{-1} + h) \|\hat{\delta}_r\| + \|\hat{\delta}_r\|^{3/2}) + \sqrt{NT}h (Th)^{-1} O_P(\|\hat{\delta}_r\|^{1/2}) + o_P(1) \end{aligned}$$

where we use the fact that  $(NT)h^{1/2} C_{NT}^{-2} (C_{NT}^{-1} + h) = o(1)$ ,  $(\sqrt{Th} + \sqrt{Nh}) C_{NT}^{-2} = o(1)$ ,  $(Th/N)^{-1/2} C_{NT}^{-1} = o(1)$  and  $Nh^2 T^{-1} = o(1)$ .

Recall that  $V_i^{(r)} = \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)}$ . Then replacing  $X_i^{(r)}$  with  $V_i^{(r)}$ , the same argument leads to

$$\begin{aligned} &\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N V_i^{(r)'} (M_{F^{(r)}} - M_{\hat{F}^{(r)}}) \varepsilon_i^{(r)} \\ &= \sqrt{\frac{Th}{N}} \psi_{NT}^{*(r)} + \sqrt{NT}h O_P((C_{NT}^{-1} + h) \|\hat{\delta}_r\| + \|\hat{\delta}_r\|^{3/2}) + \sqrt{NT}h (Th)^{-1} O_P(\|\hat{\delta}_r\|^{1/2}) + o_P(1), \end{aligned}$$

where

$$\psi_{NT}^{*(r)} = \frac{1}{N} \sum_{i=1}^N \sum_{k=1}^N \frac{V_i^{(r)'} F^{(r)}}{T} \left( \frac{F^{(r)'} F^{(r)}}{T} \right)^{-1} \left( \frac{\Lambda_r' \Lambda_r}{N} \right)^{-1} \lambda_{kr} \left( \frac{1}{T} \sum_{t=1}^T \varepsilon_{it}^{(r)} \varepsilon_{kt}^{(r)} \right).$$

Define  $\xi_{NT}^{(r)} = -[\psi_{NT}^{(r)} - \psi_{NT}^{*(r)}]$ . Then, we have

$$\begin{aligned} &\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} M_{\hat{F}^{(r)}} - \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} M_{\hat{F}^{(r)}} \right] \varepsilon_i^{(r)} \\ &= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} M_{F^{(r)}} - \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} M_{F^{(r)}} \right] \varepsilon_i^{(r)} \\ &+ \sqrt{\frac{Th}{N}} \xi_{NT}^{(r)} + \sqrt{NT}h O_P((C_{NT}^{-1} + h) \|\hat{\delta}_r\| + \|\hat{\delta}_r\|^{3/2}) + \sqrt{NT}h (Th)^{-1} O_P(\|\hat{\delta}_r\|^{1/2}) + o_P(1). \end{aligned}$$

This completes the proof of the lemma.  $\blacksquare$

**Proof of Lemma A.7.** We consider the term  $\frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_i^{(r)}$  first.

$$\frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_i^{(r)} = \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} \Delta_i^{(r)} - \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{\hat{F}^{(r)} \hat{F}^{(r)'}}{T} \Delta_i^{(r)}.$$

We study the two terms on the right hand side of the above equation in turn. Recall that  $A_{l,itr}$  denote the  $i$ th element of  $A_{l,itr} = X_t \beta_r^{(l)} + \Lambda_r^{(l)} F_t$  for  $l = 1, 2$ . Noting that

$$\Delta_{i,t}^{(r)} = k_{h,tr}^{*1/2} \Delta_i(t, r) = k_{h,tr}^{*1/2} [X_{it}' d_0(t, r) + F_t' d_i(t, r)] = k_{h,tr}^{*1/2} \left\{ A_{1,itr} \frac{t-r}{T} + A_{2,itr} \left( \frac{t-r}{T} \right)^2 + O_P(h^3) \right\},$$

for the term  $\frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} \Delta_i^{(r)}$ , we have

$$\begin{aligned} & \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} \Delta_i^{(r)} = \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T X_{it}^{(r)} \Delta_{i,t}^{(r)} \\ &= \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* X_{it} [A_{1,itr} \frac{t-r}{T} + A_{2,itr} \left( \frac{t-r}{T} \right)^2] + o_P(1) \\ &= \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* [X_{it} A_{1,itr} - E(X_{it} A_{1,itr})] \frac{t-r}{T} + \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* [X_{it} A_{2,itr} - E(X_{it} A_{2,itr})] \left( \frac{t-r}{T} \right)^2 \\ &+ \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(X_{it} A_{1,itr}) \frac{t-r}{T} + \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(X_{it} A_{2,itr}) \left( \frac{t-r}{T} \right)^2 + o_P(1) \equiv \sum_{\ell=1}^4 S_\ell + o_P(1). \end{aligned}$$

By standard variance calculation and Chebyshev inequality

$$\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* [X_{it} A_{1,itr} - E(X_{it} A_{1,itr})] \left( \frac{t-r}{T} \right)^l = O_P((NT\bar{h})^{-1/2} h^l).$$

This implies that  $S_l = O_P(h^l)$  for  $l = 1, 2$ . By the property of Riemann integral,

$$\begin{aligned} S_3 &= \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(X_{it} A_{1,itr}) \frac{t-r}{T} \\ &= \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* \left[ E(X_{it} X_{it}') \beta_r^{(1)} + E(X_{it} F_t') \lambda_{ir}^{(1)} \right] \frac{t-r}{T} \\ &= \frac{\sqrt{NT\bar{h}}}{N} \sum_{i=1}^N \left[ E(X_{i1} X_{i1}') \beta_r^{(1)} + E(X_{i1} F_t') \lambda_{i1}^{(1)} \right] \left\{ \int uk(u) du + \frac{1}{T} \right\} = O(\sqrt{N\bar{h}/T}) = o(1). \end{aligned}$$

It follows that  $\frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} \Delta_i^{(r)} = \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(X_{it} A_{2,itr}) \left( \frac{t-r}{T} \right)^2 + o_P(1)$ .

Next, we study the term  $\frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{\hat{F}^{(r)} \hat{F}^{(r)'}}{T} \Delta_i^{(r)}$ . Noting that  $\hat{F}_t^{(r)} = \Upsilon_t^{(r)} + H^{(r)} F_t^{(r)} + B_t^{(r)}$ , we make the following decomposition:

$$\begin{aligned} & \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N X_i^{(r)'} \frac{\hat{F}^{(r)} \hat{F}^{(r)'}}{T} \Delta_i^{(r)} = \frac{\sqrt{NT\bar{h}}}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T X_{it}^{(r)} \hat{F}_t^{(r)'} \hat{F}_s^{(r)} \Delta_{i,s}^{(r)} \\ &= \frac{\sqrt{NT\bar{h}}}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T X_{it}^{(r)} \{ \Upsilon_t^{(r)'} \Upsilon_s^{(r)} + F_t^{(r)'} H^{(r)} \Upsilon_s^{(r)} + B_t^{(r)'} \Upsilon_s^{(r)} + \Upsilon_t^{(r)'} H^{(r)'} F_s^{(r)} + \Upsilon_t^{(r)'} B_s^{(r)} \} \end{aligned}$$



$$+ F_t^{(r)'} H^{(r)} H^{(r)'} F_s^{(r)} + B_t^{(r)'} H^{(r)'} F_s^{(r)} + F_t^{(r)'} H^{(r)} B_s^{(r)} + B_t^{(r)'} B_s^{(r)} \} \Delta_{i,s}^{(r)} \equiv \sum_{l=1}^9 D_l^{(r)},$$

where  $D_1^{(r)} = \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T X_{it}^{(r)} \Upsilon_t^{(r)'} \Upsilon_s^{(r)} \Delta_{i,s}^{(r)}$  and other terms are analogously defined. For  $D_1^{(r)}$ , we have

$$\begin{aligned} \|D_1^{(r)}\| &\leq \sqrt{NT}h \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T X_{it}^{(r)} \Upsilon_t^{(r)'} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{s=1}^T \Upsilon_s^{(r)} \Delta_{i(s,r)} \right\|^2 \right\}^{1/2} \\ &= \sqrt{NT}h O_P((C_{NT}^{-2} + \|\hat{\delta}_r\|)^2 h) \end{aligned}$$

by similar arguments as used to obtain Lemma A.3(iv) and (v). Following the proof of Lemma A.4(iv) and (v), we can also show that

$$D_l^{(r)} = \sqrt{NT}h O_P((C_{NT}^{-2} + \|\hat{\delta}_r\|)h^2) \text{ for } l = 2, 3, 4, 5 \text{ and } D_9^{(r)} = \sqrt{NT}h O_P(h^4 + (Th)^{-1/2}h^3) = o_P(1).$$

For  $D_6^{(r)}$ , we have

$$\begin{aligned} D_6^{(r)} &= \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T X_{it}^{(r)} F_t^{(r)'} H^{(r)} H^{(r)'} F_s^{(r)} \Delta_{i,s}^{(r)} \\ &= \sum_{l=1}^2 \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,tr}^* k_{h,sr}^* X_{it} F_t' H^{(r)} H^{(r)'} F_s A_{1, isr} \left(\frac{s-r}{T}\right)^l + o_P(1) \equiv \sum_{l=1}^2 D_{6l}^{(r)} + o_P(1). \end{aligned}$$

For  $D_{61}^{(r)}$ , we have

$$\begin{aligned} \omega' D_{61}^{(r)} &= \frac{\sqrt{NT}h}{NT^2} \left\| \text{tr} \left\{ H^{(r)} H^{(r)'} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T F_s A_{1, isr} \frac{s-r}{T} k_{h,tr}^* k_{h, sr}^* \omega' X_{it} F_t' \right\} \right\| \\ &\lesssim \frac{\sqrt{NT}h}{NT^2} \left\| \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h, sr}^* F_s A_{1, isr} \frac{s-r}{T} k_{h, tr}^* \omega' X_{it} F_t' \right\| \\ &\leq \frac{\sqrt{NT}h}{NT^2} \left\| \sum_{i=1}^N \sum_{t=1}^T \left\{ \sum_{s=1}^T k_{h, sr}^* E(F_s A_{1, isr}) \frac{s-r}{T} \right\} k_{h, tr}^* \omega' E(X_{it} F_t') \right\| + \sqrt{NT}h O_P((Th)^{-1}h + T^{-1}) \\ &= \sqrt{NT}h O_P(T^{-1}) + O_P(\sqrt{Nh/T}) = o_P(1). \end{aligned}$$

It follows that

$$D_6^{(r)} = \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h, tr}^* k_{h, sr}^* X_{it} F_t' H^{(r)} H^{(r)'} F_s A_{2, isr} \left(\frac{s-r}{T}\right)^2 + o_P(1).$$

Similarly, we can show that  $D_7^{(r)} = o_P(1)$ . For  $D_8^{(r)}$ , we have

$$\begin{aligned} D_8^{(r)} &= \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T X_{it}^{(r)} F_t^{(r)'} H^{(r)} B_s^{(r)} \Delta_{i,s}^{(r)} \\ &= \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h, tr}^* k_{h, sr}^* X_{it} F_t' H^{(r)} C_{1s}^{(r)} A_{1, isr} \left(\frac{s-r}{T}\right)^2 + o_P(1). \end{aligned}$$

It follows that

$$\begin{aligned}
& \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N X_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_i^{(r)} \\
&= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(X_{it} A_{2,it}) \left(\frac{t-r}{T}\right)^2 \\
&+ \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,tr}^* k_{h,sr}^* X_{it} F_t' H^{(r)} [H^{(r)'} F_s A_{2,isr} + C_{1s}^{(r)} A_{1,isr}] \left(\frac{s-r}{T}\right)^2 + o_p(1).
\end{aligned}$$

Recall that  $V_i^{(r)} = N^{-1} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)}$  and  $V_{it}^{(r)} = N^{-1} \sum_{k=1}^N a_{ik}^{(r)} X_{kt}^{(r)} \equiv k_{h,tr}^{*1/2} V_{it,r}$  where  $V_{it,r} = N^{-1} \sum_{k=1}^N a_{ik}^{(r)} X_{kt}$ . By replacing  $X_i^{(r)}$  with  $V_i^{(r)}$  in the above analyses, we can show that

$$\begin{aligned}
& \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} M_{\hat{F}^{(r)}} \Delta_i^{(r)} = \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N V_i^{(r)'} M_{\hat{F}^{(r)}} \Delta_i^{(r)} \\
&= \frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr}^* E(V_{it,r} A_{2,it}) \left(\frac{t-r}{T}\right)^2 \\
&+ \frac{\sqrt{NT}h}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,tr}^* k_{h,sr}^* V_{it,r} F_t' H^{(r)} [H^{(r)'} F_s A_{2,isr} + C_{1s}^{(r)} A_{1,isr}] \left(\frac{s-r}{T}\right)^2 + o_p(1).
\end{aligned}$$

In sum, we have

$$\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} M_{\hat{F}^{(r)}} - \frac{1}{N} \sum_{k=1}^N a_{ik}^{(r)} X_k^{(r)'} M_{\hat{F}^{(r)}} \right] \Delta_i^{(r)} = \sqrt{NT}h B_{1\beta}^{(r)} + o_p(1),$$

with

$$\begin{aligned}
B_{1\beta}^{(r)} &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T k_{h,tr} E[(X_{it} - V_{it,r}) A_{2,it}] \left(\frac{t-r}{T}\right)^2 \\
&+ \frac{1}{NT^2} \sum_{i=1}^N \sum_{t=1}^T \sum_{s=1}^T k_{h,tr} k_{h,sr} (X_{it} - V_{it,r}) F_t' H^{(r)} [H^{(r)'} F_s A_{2,isr} + C_{1s}^{(r)} A_{1,isr}] \left(\frac{s-r}{T}\right)^2. \blacksquare
\end{aligned}$$

**Proof of Lemma A.8.** We note that  $M_{F^{(r)}}(X_i^{(r)} - \frac{1}{N} \sum_{j=1}^N X_j^{(r)} a_{ij}^{(r)}) \equiv Z_i^{(r)}$  is a  $T \times P$  matrix. Stack  $Z_i^{(r)}$  into a  $T \times N \times P$  three dimensional matrix  $\mathbf{Z}^{(r)}$ . Following Moon and Weidner (2017), we denote the  $p$ th sheet of  $\mathbf{Z}^{(r)}$  as  $\mathbf{Z}_p^{(r)}$ , where  $p = 1, \dots, P$ . Recall that  $\mathbf{X}_p^{(r)}$  denotes the  $p$ th sheet of the three dimensional regressor matrix  $\mathbf{X}^{(r)}$ ,  $\tilde{\mathbf{X}}_p^{(r)} = E(\mathbf{X}_p^{(r)})$ ,  $\tilde{\mathbf{X}}_p^{(r)} = \mathbf{X}_p^{(r)} - \tilde{\mathbf{X}}_p^{(r)}$ , and  $\mathfrak{X}_p^{(r)} = M_{F^{(r)}} \tilde{\mathbf{X}}_p^{(r)} M_{\Lambda_r} + \tilde{\mathbf{X}}_p^{(r)}$ . It is easy to verify that

$$\mathbf{Z}_p^{(r)} = M_{F^{(r)}} \mathbf{X}_p^{(r)} M_{\Lambda_r} = \mathfrak{X}_p^{(r)} - P_{F^{(r)}} \tilde{\mathbf{X}}_p^{(r)} - \tilde{\mathbf{X}}_p^{(r)} P_{\Lambda_r} + P_{F^{(r)}} \tilde{\mathbf{X}}_p^{(r)} P_{\Lambda_r}.$$

Then the  $p$ th entry of  $\frac{\sqrt{NT}h}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} \right] M_{F^{(r)}} \varepsilon_i^{(r)}$  can be written as

$$\begin{aligned}
& \frac{\sqrt{NT}h}{NT} \text{tr} \left( M_{F^{(r)}} \mathbf{X}_p^{(r)} M_{\Lambda_r} \varepsilon^{(r)'} \right) \\
&= \frac{\sqrt{NT}h}{NT} \text{tr} \left( [\mathfrak{X}_p^{(r)} - P_{F^{(r)}} \tilde{\mathbf{X}}_p^{(r)} - \tilde{\mathbf{X}}_p^{(r)} P_{\Lambda_r} + P_{F^{(r)}} \tilde{\mathbf{X}}_p^{(r)} P_{\Lambda_r}] \varepsilon^{(r)'} \right) \\
&= \frac{\sqrt{NT}h}{NT} \text{tr} \left( \mathfrak{X}_p^{(r)} \varepsilon^{(r)'} \right) - \frac{\sqrt{NT}h}{NT} \text{tr} \left( P_{F^{(r)}} E[\tilde{\mathbf{X}}_p^{(r)} \varepsilon^{(r)'}] \right) - \frac{\sqrt{NT}h}{NT} \text{tr} \left( P_{F^{(r)}} \left( \tilde{\mathbf{X}}_p^{(r)} \varepsilon^{(r)'} - E[\tilde{\mathbf{X}}_p^{(r)} \varepsilon^{(r)'}] \right) \right)
\end{aligned}$$

$$-\frac{\sqrt{NT\bar{h}}}{NT} \text{tr} \left( \tilde{\mathbf{X}}_k^{(r)} P_{\Lambda_r} \varepsilon^{(r)'} \right) + \frac{\sqrt{NT\bar{h}}}{NT} \text{tr} \left( P_{F^{(r)}} \tilde{\mathbf{X}}_k^{(r)} P_{\Lambda_r} \varepsilon^{(r)'} \right) \equiv \sum_{\ell=1}^5 I_\ell.$$

Note that  $I_1 = \frac{\sqrt{NT\bar{h}}}{NT} \text{tr}(\tilde{\mathbf{X}}_k^{(r)} \varepsilon^{(r)'}) = \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T \tilde{\mathbf{x}}_{k,it}^{(r)} \varepsilon_{it}^{(r)}$  and  $I_2 = \frac{\sqrt{NT\bar{h}}}{NT} \text{tr} \left( P_{F^{(r)}} E[\mathbf{X}_k^{(r)} \varepsilon^{(r)'}] \right) = \sqrt{\frac{N}{T\bar{h}}} \tilde{B}_{2\beta}^{(r)}$ , where  $\tilde{B}_{2\beta,k}^{(r)} = -\frac{\bar{h}}{N} \text{tr} \left( P_{F^{(r)}} E[\mathbf{X}_k^{(r)} \varepsilon^{(r)'}] \right)$ . By Lemma A.10 below,  $I_l = o_P(1)$  for  $l = 3, 4, 5$ . It follows that

$$\begin{aligned} \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \left[ X_i^{(r)'} - \frac{1}{N} \sum_{j=1}^N a_{ij}^{(r)} X_j^{(r)'} \right] M_{F^{(r)}} \varepsilon_i^{(r)} - \sqrt{\frac{N}{T\bar{h}}} \tilde{B}_{2\beta}^{(r)} &= \frac{\sqrt{NT\bar{h}}}{NT} \sum_{i=1}^N \sum_{t=1}^T \tilde{\mathbf{x}}_{it}^{(r)} \varepsilon_{it}^{(r)} + o_P(1) \\ &\xrightarrow{d} N(0, \Omega_r), \end{aligned}$$

where  $\tilde{\mathbf{x}}_{it}^{(r)} = (\tilde{\mathbf{x}}_{1,it}^{(r)}, \dots, \tilde{\mathbf{x}}_{P,it}^{(r)})'$ ,  $\tilde{B}_{2\beta}^{(r)} = (\tilde{B}_{2\beta,1}^{(r)}, \dots, \tilde{B}_{2\beta,P}^{(r)})'$ , and the convergence in distribution follows from Assumption A.5. ■

**Proof of Lemma A.9.** (i) Note that

$$D(\hat{F}^{(r)}) - D(F^{(r)}) = \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} (M_{\hat{F}^{(r)}} - M_{F^{(r)}}) X_i^{(r)} - \frac{1}{T} \left[ \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} (M_{\hat{F}^{(r)}} - M_{F^{(r)}}) a_{ij}^{(r)} X_j^{(r)} \right].$$

It is easy to show that either term on the right hand side is bounded above by  $O_P(\|P_{\hat{F}^{(r)}} - P_{F^{(r)}}\|)$ , which is  $O_P(C_{NT}^{-1} + \|\hat{\delta}_r\|^{1/2})$  by Lemma A.2(v). The result then follows from the asymptotic nonsingularity of  $D(F^{(r)})$ .

(ii) Let  $\tilde{G}^{(r)} = (\frac{1}{T} F^{(r)'} F^{(r)})^{-1} (\frac{1}{N} \Lambda_r' \Lambda_r)^{-1}$  and  $\bar{G}^{(r)} = (\frac{1}{T} E[F^{(r)'} F^{(r)}])^{-1} (\frac{1}{N} \Lambda_r' \Lambda_r)^{-1}$ . We can readily show that

$$\xi_{NT}^{(r)} - B_{3\beta}^{(r)} = -\frac{1}{N} \sum_{i=1}^N \frac{(X_i^{(r)} - V_i^{(r)})' F^{(r)}}{T} \tilde{G}^{(r)} \frac{1}{T} \sum_{j=1}^N \sum_{t=1}^T k_{h,tr}^* \lambda_{jr} [\varepsilon_{it} \varepsilon_{jt} - E(\varepsilon_{it} \varepsilon_{jt})] = -\Phi^{(r)} + o_P((Th/N)^{-1/2}),$$

where  $\Phi^{(r)} = \frac{1}{NT^2} \sum_{i=1}^N (X_i^{(r)} - V_i^{(r)})' F^{(r)} \bar{G}^{(r)} \sum_{j=1}^N \sum_{t=1}^T k_{h,tr}^* \lambda_{jr} [\varepsilon_{it} \varepsilon_{jt} - E_C(\varepsilon_{it} \varepsilon_{jt})]$ .

$$\begin{aligned} \Phi^{(r)} &= \frac{1}{NT^2} \sum_{i=1}^N X_i^{(r)'} F^{(r)} \bar{G}^{(r)} \sum_{j=1}^N \sum_{t=1}^T k_{h,tr}^* \lambda_{jr} [\varepsilon_{it} \varepsilon_{jt} - E_C(\varepsilon_{it} \varepsilon_{jt})] \\ &\quad - \frac{1}{NT^2} \sum_{i=1}^N V_i^{(r)'} F^{(r)} \bar{G}^{(r)} \sum_{j=1}^N \sum_{t=1}^T k_{h,tr}^* \lambda_{jr} [\varepsilon_{it} \varepsilon_{jt} - E_C(\varepsilon_{it} \varepsilon_{jt})] \equiv \Phi_1^{(r)} - \Phi_2^{(r)}. \end{aligned}$$

For  $\Phi_1^{(r)}$ , we make further decomposition:

$$\begin{aligned} \Phi_1^{(r)} &= \frac{1}{NT^2} \sum_{i=1}^N \sum_{j=1}^N \sum_{t=1}^T \sum_{s=1}^T E_C(X_{is}^{(r)} F_s^{(r)'}) \bar{G}^{(r)} k_{h,tr} \lambda_{jr} [\varepsilon_{it} \varepsilon_{jt} - E_C(\varepsilon_{it} \varepsilon_{jt})] \\ &\quad + \frac{1}{NT^2} \sum_{i=1}^N \sum_{j=1}^N \sum_{t=1}^T \sum_{s=1}^T \left[ X_{is}^{(r)} F_s^{(r)'} - E_C(X_{is}^{(r)} F_s^{(r)'}) \right] \bar{G}^{(r)} k_{h,tr} \lambda_{jr} [\varepsilon_{it} \varepsilon_{jt} - E_C(\varepsilon_{it} \varepsilon_{jt})] \equiv \Phi_{1,1}^{(r)} - \Phi_{1,2}^{(r)}. \end{aligned}$$

We can show that  $\text{Var}_C(\Phi_{1,1}^{(r)}) = O_P((Th)^{-1})$ , which, along with the fact that  $E_C(\Phi_{1,1}^{(r)}) = 0$ , implies that  $\Phi_{1,1}^{(r)} = O_P((Th)^{-1/2})$ . Similarly, for  $\Phi_{1,2}^{(r)}$ , we can show that  $E_C(\Phi_{1,2}^{(r)}) = O_P(N/T)$ ,  $\text{Var}_C(\Phi_{1,2}^{(r)}) = O_P((Th)^{-1})$  and thus  $\Phi_{1,2}^{(r)} = O_P(N/T + (Th)^{-1/2})$ . It follows that  $\sqrt{Th/N} \Phi_1^{(r)} = O_P(N^{-1/2} + (Nh/T)^{-1/2}) = o_P(1)$ . Similarly, we can show that  $\sqrt{Th/N} \Phi_2^{(r)} = o_P(1)$ . Then  $\sqrt{Th/N} [\xi_{NT}^{(r)} - B_{3\beta}^{(r)}] = o_P(1)$ .

(iii) Denote  $G^{(r)} = (\frac{1}{T}F^{(r)'}\hat{F}^{(r)})^{-1}(\frac{1}{N}\Lambda_r'\Lambda_r)^{-1}$  and  $\tilde{G}^{(r)} = (\frac{1}{T}F^{(r)'}F^{(r)})^{-1}(\frac{1}{N}\Lambda_r'\Lambda_r)^{-1}$ . Then

$$\begin{aligned}\sqrt{\frac{N}{Th}}\left(\zeta_{NT}^{(r)} - B_{4\beta}^{(r)}\right) &= \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} \left[ M_{\hat{F}^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} - M_{F^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} F^{(r)} \tilde{G}^{(r)} \right] \lambda_{ir} \\ &= \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} \left[ M_{\hat{F}^{(r)}} - M_{F^{(r)}} \right] \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &\quad + \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{F^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \left[ \hat{F}^{(r)} G^{(r)} - F^{(r)} \tilde{G}^{(r)} \right] \lambda_{ir} \equiv III_1 + III_2.\end{aligned}$$

For  $III_1$ , we have

$$\begin{aligned}III_1 &= \frac{-h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} \left[ P_{\hat{F}^{(r)}} - P_{F^{(r)}} \right] \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \hat{F}^{(r)} G^{(r)} \lambda_{ir} \\ &\leq \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} \left[ P_{\hat{F}^{(r)}} - P_{F^{(r)}} \right] \varepsilon_j^{(r)} \varepsilon_j^{(r)'} F^{(r)} H^{(r)} G^{(r)} \lambda_{ir} \\ &\quad + \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} \left[ P_{\hat{F}^{(r)}} - P_{F^{(r)}} \right] \varepsilon_j^{(r)} \varepsilon_j^{(r)'} \left[ \hat{F}^{(r)} - F^{(r)} H^{(r)} \right] G^{(r)} \lambda_{ir} \equiv III_{1,1} + III_{1,2}.\end{aligned}$$

For  $III_{1,2}$ , it suffices to consider the rough probability bound

$$\begin{aligned}|\omega' III_{1,2}| &\leq \left\| \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{j=1}^N \varepsilon_j^{(r)'} \left[ \hat{F}^{(r)} - F^{(r)} H^{(r)} \right] G^{(r)} \lambda_{ir} \omega' \sum_{i=1}^N X_i^{(r)'} \left[ P_{\hat{F}^{(r)}} - P_{F^{(r)}} \right] \varepsilon_j^{(r)} \right\| \\ &\lesssim (NTh)^{1/2} \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{T} \varepsilon_j^{(r)'} \left[ \hat{F}^{(r)} - F^{(r)} H^{(r)} \right] \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} \left[ P_{\hat{F}^{(r)}} - P_{F^{(r)}} \right] \varepsilon_j^{(r)} \right\|^2 \right\}^{1/2} \\ &\lesssim (NTh)^{1/2} \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{T} \varepsilon_j^{(r)'} \left[ \hat{F}^{(r)} - F^{(r)} H^{(r)} \right] \right\|^2 \right\}^{1/2} \|P_{\hat{F}^{(r)}} - P_{F^{(r)}}\| \\ &= (NTh)^{1/2} O_P(C_{NT}^{-2}) O_P(C_{NT}^{-1}) = o_P(1).\end{aligned}$$

Using the decomposition

$$\begin{aligned}P_{\hat{F}^{(r)}} - P_{F^{(r)}} &= \frac{1}{T} \hat{F}^{(r)} \hat{F}^{(r)'} - F^{(r)} [F^{(r)'} F^{(r)}]^{-1} F^{(r)'} \\ &= \frac{1}{T} \left( \hat{F}^{(r)} - F^{(r)} H^{(r)} \right) \hat{F}^{(r)'} + F^{(r)} H^{(r)} \left( \hat{F}^{(r)} - F^{(r)} H^{(r)} \right)' \\ &\quad + \frac{1}{T} F^{(r)} H^{(r)} \left\{ [H^{(r)'} \frac{1}{T} F^{(r)'} F^{(r)} H^{(r)}]^{-1} - \mathbb{I}_R \right\} H^{(r)'} F^{(r)'} \\ &= \frac{1}{T} \left( \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)} \right) \hat{F}^{(r)'} + \frac{1}{T} F^{(r)} H^{(r)} \left( \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)} \right)' \\ &\quad + \frac{1}{T} B^{(r)} \hat{F}^{(r)'} + \frac{1}{T} F^{(r)} H^{(r)} B^{(r)'} + \frac{1}{T} F^{(r)} \left\{ \frac{1}{T} [F^{(r)'} F^{(r)}]^{-1} - H^{(r)} H^{(r)'} \right\} F^{(r)'},\end{aligned}$$

we can readily show that  $\frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} \left[ P_{\hat{F}^{(r)}} - P_{F^{(r)}} \right] \varepsilon_j^{(r)} = O_P(C_{NT}^{-2})$  for each  $j$  and  $\frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} \right\|^2$

$[P_{\hat{F}^{(r)}} - P_{F^{(r)}}] \varepsilon_j^{(r)} \|^2 = O_P(C_{NT}^{-4})$ . Then

$$\begin{aligned} |\omega' III_{1,1}| &\leq \left\| \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{j=1}^N \varepsilon_j^{(r)'} F^{(r)} H^{(r)} \right\| G^{(r)} \lambda_{ir} \omega' \sum_{i=1}^N X_i^{(r)'} [P_{\hat{F}^{(r)}} - P_{F^{(r)}}] \varepsilon_j^{(r)} \left\| \right. \\ &\lesssim (NTh)^{1/2} \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{T} \varepsilon_j^{(r)'} F^{(r)} \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{NT} \sum_{i=1}^N X_i^{(r)'} [P_{\hat{F}^{(r)}} - P_{F^{(r)}}] \varepsilon_j^{(r)} \right\|^2 \right\}^{1/2} \\ &= (NTh)^{1/2} O_P((Th)^{-1/2}) O_P(C_{NT}^{-2}) = o_P(1). \end{aligned}$$

Then  $III_1 = o_P(1)$ .

For  $III_2$ , we have

$$\begin{aligned} III_2 &= \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} M_{F^{(r)}} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} [\hat{F}^{(r)} (\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} - F^{(r)} (\frac{1}{T} F^{(r)'} F^{(r)})^{-1}] [\frac{1}{N} \Lambda^{(r)'} \Lambda^{(r)}]^{-1} \lambda_{ir} \\ &= \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} [\hat{F}^{(r)} - F^{(r)} H^{(r)}] (\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} [\frac{1}{N} \Lambda^{(r)'} \Lambda^{(r)}]^{-1} \lambda_{ir} \\ &\quad + \frac{h^{1/2}}{N^{3/2}T^{3/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} F^{(r)} H^{(r)} [(\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} - (\frac{1}{T} F^{(r)'} F^{(r)})^{-1}] [\frac{1}{N} \Lambda^{(r)'} \Lambda^{(r)}]^{-1} \lambda_{ir} \\ &\quad - \frac{h^{1/2}}{N^{3/2}T^{35/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} F^{(r)} [\frac{1}{T} F^{(r)'} F^{(r)}]^{-1} F^{(r)'} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} [\hat{F}^{(r)} - F^{(r)} H^{(r)}] (\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} [\frac{1}{N} \Lambda^{(r)'} \Lambda^{(r)}]^{-1} \lambda_{ir} \\ &\quad - \frac{h^{1/2}}{N^{3/2}T^{35/2}} \sum_{i=1}^N \sum_{j=1}^N X_i^{(r)'} F^{(r)} [\frac{1}{T} F^{(r)'} F^{(r)}]^{-1} F^{(r)'} \varepsilon_j^{(r)} \varepsilon_j^{(r)'} F^{(r)} H^{(r)} [(\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} - (\frac{1}{T} F^{(r)'} F^{(r)})^{-1}] \\ &\quad \times [\frac{1}{N} \Lambda^{(r)'} \Lambda^{(r)}]^{-1} \lambda_{ir} \\ &\equiv III_{2,1} + III_{2,2} - III_{2,3} - III_{2,4}. \end{aligned}$$

We can readily show that  $III_{2,l} = o_P(1)$  for  $l \in [4]$ . For example,

$$\begin{aligned} |\omega' III_{2,1}| &= \frac{h^{1/2}}{N^{3/2}T^{3/2}} \left| \sum_{j=1}^N \varepsilon_j^{(r)'} [\hat{F}^{(r)} - F^{(r)} H^{(r)}] (\frac{1}{T} F^{(r)'} \hat{F}^{(r)})^{-1} [\frac{1}{N} \Lambda^{(r)'} \Lambda^{(r)}]^{-1} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} \varepsilon_j^{(r)} \right| \\ &\lesssim (NTh)^{1/2} \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{T} \varepsilon_j^{(r)'} [\hat{F}^{(r)} - F^{(r)} H^{(r)}] \right\|^2 \right\}^{1/2} \left\{ \frac{1}{N} \sum_{j=1}^N \left\| \frac{1}{NT} \sum_{i=1}^N \lambda_{ir} \omega' X_i^{(r)'} \varepsilon_j^{(r)} \right\|^2 \right\}^{1/2} \\ &= (NTh)^{1/2} O(C_{NT}^{-2}) O_P((NTh)^{-1/2}) = o_P(1). \end{aligned}$$

This prove the lemma. ■

**Proof of Lemma A.10** (i) Recall that  $\tilde{\mathbf{X}}_p^{(r)} = \mathbf{X}_p^{(r)} - E_{\mathcal{L}}(\mathbf{X}_p^{(r)})$ . Let  $\tilde{X}_{p, is}^{(r)}$  denote the  $(s, i)$ th element of the  $T \times N$  matrix  $\tilde{\mathbf{X}}_p^{(r)}$ . Let  $e_{qR}$  denotes the  $q$ th column of  $\mathbb{I}_R$ . Then

$$\left\| \frac{\sqrt{NTh}}{NT} \text{tr} \left\{ P_{F^{(r)}} \left[ \tilde{\mathbf{X}}_p^{(r)} \varepsilon^{(r)'} - E_{\mathcal{L}}(\tilde{\mathbf{X}}_p^{(r)} \varepsilon^{(r)'}) \right] \right\} \right\|$$

$$\begin{aligned}
&= \left\| \sum_{l,q=1}^R e'_{lR} \left( \frac{1}{T} F^{(r)'} F^{(r)} \right)^{-1} e_{qR} \frac{\sqrt{h}}{\sqrt{NT}^3} \sum_{i=1}^N \sum_{t,s=1}^T F_{t,l}^{(r)} F_{s,q}^{(r)} \left[ \tilde{X}_{p,is}^{(r)} \varepsilon_{it}^{(r)} - E_C(\tilde{X}_{p,is}^{(r)} \varepsilon_{it}^{(r)}) \right] \right\| \\
&\lesssim (Th)^{-1/2} \sum_{l,q=1}^R \|\Xi_{NT,lq}\| = O_P((Th)^{-1/2}),
\end{aligned}$$

where  $\Xi_{NT,lq} = \frac{h}{N^{1/2}T} \sum_{i=1}^N \sum_{t,s=1}^T k_{h,tr}^* k_{h,sr}^* F_{t,l} F_{s,q} [\tilde{X}_{p,is} \varepsilon_{it} - E_C(\tilde{X}_{p,is} \varepsilon_{it})] = O_P(1)$  by Assumption A.2(iv).

(ii) Recall  $a_{ij}^{(r)} = \lambda'_{ir} \left( \frac{1}{N} \Lambda_r' \Lambda_r \right)^{-1} \lambda_{jr}$ . Then

$$\begin{aligned}
\frac{\sqrt{NT}h}{NT} \text{tr} \left( \tilde{\mathbf{X}}_k^{(r)} P_{\Lambda_r} \varepsilon^{(r)'} \right) &= \frac{\sqrt{NT}h}{N^2T} \sum_{t=1}^T \sum_{i=1}^N \sum_{j=1}^N k_{h,tr}^* a_{ij}^{(r)} \tilde{X}_{k,jt} \varepsilon_{it} \\
&= \frac{\sqrt{NT}h}{N^2T} \sum_{t=1}^T \sum_{i=1}^N k_{h,tr}^* a_{ii}^{(r)} \tilde{X}_{k,it} \varepsilon_{it} + \frac{\sqrt{NT}h}{N^2T} \sum_{t=1}^T \sum_{1 \leq i < j \leq N} k_{h,tr}^* a_{ij}^{(r)} \tilde{X}_{k,jt} \varepsilon_{it} \\
&\quad + \frac{\sqrt{NT}h}{N^2T} \sum_{t=1}^T \sum_{1 \leq j < i \leq N} k_{h,tr}^* a_{ij}^{(r)} \tilde{X}_{k,jt} \varepsilon_{it} \equiv IV_1 + IV_2 + IV_3.
\end{aligned}$$

It is easy to show that  $IV_1 = O_P(N^{-1/2})$ . For  $IV_2$ , we have  $E_C(IV_2) = 0$  and

$$\text{Var}_C(\omega' IV_2) = \frac{h}{N^3T} \sum_{t=1}^T \sum_{s=1}^T \sum_{1 \leq i < j \leq N} k_{h,tr}^* k_{h,sr}^* \left( a_{ij}^{(r)} \right)^2 \omega' E_C \left[ \tilde{X}_{k,jt} \varepsilon_{it} \varepsilon_{is} \tilde{X}'_{k,js} \right] \omega = O_P(N^{-1}).$$

Then  $IV_2 = O_P(N^{-1/2})$ . By the same token,  $IV_3 = O_P(N^{-1/2})$ . It follows that  $\frac{\sqrt{NT}h}{NT} \text{tr}(\tilde{\mathbf{X}}_k^{(r)} P_{\Lambda_r} \varepsilon^{(r)'}) = o_P(1)$ .

(iii) Let  $\lambda_{ir,q}$  denote the  $q$ th element in  $\lambda_{ir}$ . Noting that

$$\begin{aligned}
E_C \left\| \tilde{\mathbf{X}}_k^{(r)} \Lambda_r \right\|^2 &= \sum_{q,l=1}^R \sum_{i,j=1}^N \sum_{t,s=1}^T k_{h,tr}^{*1/2} k_{h,sr}^{*1/2} \lambda_{ir,q} \lambda_{jr,l} \text{Cov}_C(X_{k,it}, X_{k,js}) \\
&\lesssim \sum_{i,j=1}^N \sum_{t,s=1}^T (k_{h,tr}^* + k_{h,sr}^*) |\text{Cov}_C(X_{k,it}, X_{k,js})| = 2 \sum_{t=1}^T k_{h,tr}^* \sum_{i,j=1}^N \sum_{s=1}^T |\text{Cov}_C(X_{k,it}, X_{k,js})| \\
&\leq \max_s \sum_{i,j=1}^N \sum_{s=1}^T |\text{Cov}_C(X_{k,it}, X_{k,js})| \sum_{t=1}^T k_{h,tr}^* = O_P(N) O_P(T),
\end{aligned}$$

$\left\| \tilde{\mathbf{X}}_k^{(r)} \Lambda_r \right\| = O_P((NT)^{1/2})$  by the conditional Chebyshev inequality. Similarly, in view of the fact that

$$\begin{aligned}
E \left\| \Lambda_r' \varepsilon^{(r)'} F^{(r)} \right\|^2 &= \sum_{i,j=1}^N \sum_{t,s=1}^T E \left[ \varepsilon_{it}^{(r)} \varepsilon_{js}^{(r)} F_t^{(r)'} \lambda_{ir} \lambda'_{jr} F_s^{(r)} \right] = \sum_{i,j=1}^N \sum_{t,s=1}^T k_{h,tr}^* k_{h,sr}^* \lambda'_{r,i} E[F_t \varepsilon_{it} \varepsilon_{js} F_s'] \lambda_{r,j} \\
&\lesssim \sum_{i,j=1}^N \sum_{t,s=1}^T (k_{h,tr}^{*2} + k_{h,sr}^{*2}) \|E[F_t \varepsilon_{it} \varepsilon_{js} F_s']\| = 2 \max_t \sum_{i,j=1}^N \sum_{s=1}^T \|E[F_t \varepsilon_{it} \varepsilon_{js} F_s']\| \sum_{t=1}^T k_{h,tr}^{*2} \\
&= O(N) O_P(Th^{-1}) = O_P(NT h^{-1}),
\end{aligned}$$

we have  $\|\Lambda_r' \varepsilon^{(r)'} F^{(r)}\| = O_P((NT h^{-1})^{1/2})$ . Then

$$\begin{aligned} \left| \frac{\sqrt{NT h}}{NT} \text{tr} \left( P_{F^{(r)}} \tilde{\mathbf{X}}_k^{(r)} P_{\Lambda_r} \varepsilon^{(r)'} \right) \right| &= \left| \frac{\sqrt{NT h}}{NT} \text{tr} \left( (F^{(r)'} F^{(r)})^{-1} F^{(r)'} \tilde{\mathbf{X}}_k^{(r)} \Lambda_r (\Lambda_r' \Lambda_r)^{-1} \Lambda_r' \varepsilon^{(r)'} F^{(r)} \right) \right| \\ &\leq \frac{\sqrt{NT h}}{NT} \left\| (F^{(r)'} F^{(r)})^{-1} F^{(r)'} \right\| \left\| \tilde{\mathbf{X}}_k^{(r)} \Lambda_r \right\| \left\| (\Lambda_r' \Lambda_r)^{-1} \right\| \left\| \Lambda_r' \varepsilon^{(r)'} F^{(r)} \right\| \\ &= \frac{\sqrt{NT h}}{NT} O_P(T^{-1/2}) O_P((NT h^{-1})^{1/2}) O_P(N^{-1}) O_P((NT/h)^{1/2}) \\ &= O_P((Nh)^{-1/2}) = o_P(1). \blacksquare \end{aligned}$$

**Proof of Lemma A.11.** (i) Following the proof of Lemma A.7(ii) in Su and Wang (2017), we can show that uniformly in  $r$ ,

$$\begin{aligned} H^{(r)} &= (N^{-1} \Lambda_r' \Lambda_r) (T^{-1} F^{(r)'} \hat{F}^{(r)}) \hat{V}_{NT}^{(r)-1} = \Sigma_{\Lambda_r}^{1/2} \Upsilon_r V_r^{-1/2} + o_P(C_{NT}^{-1} (\ln T)^{1/2}) + O(\max_r \|\hat{\beta}_r - \beta_r\|) \\ &= \Sigma_{\Lambda_r}^{1/2} \Upsilon_r V_r^{-1/2} + o_P(C_{NT}^{-1} (\ln T)^{1/2}) \equiv Q_r + o_P(C_{NT}^{-1} (\ln T)^{1/2}), \end{aligned}$$

where the third equality follows from the fact  $\max_r \|\hat{\beta}_r - \beta_r\| = O_P((NT h)^{-1/2} (\ln T)^{1/2})$  by Theorem 3.4. Following the proof of Lemma A.7(iii)-(iv) in Su and Wang (2020), we can show that

$$\begin{aligned} \max_r \frac{1}{T} \left\| \hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)} \right\|^2 &= O_P(T^{-1} h^{-1} + N^{-1} \ln T) + O_P(\max_r \|\hat{\beta}_r - \beta_r\|), \text{ and} \\ \max_r \frac{1}{T} \left\| (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)})' F^{(r)} \right\| &= O_P(T^{-1} h^{-1} + N^{-1} \ln T) + O_P(\max_r \|\hat{\beta}_r - \beta_r\|), \end{aligned}$$

which in conjunction with the fact that  $\frac{1}{T} \|B^{(r)}\|^2 = O_P(h^2)$ ,  $\frac{1}{T} \|B^{(r)'} F^{(r)}\| = O_P(h^2)$  and  $\max_r \|\hat{\beta}_r - \beta_r\| = O_P((NT h)^{-1/2} (\ln T)^{1/2})$ , implies that

$$\begin{aligned} \max_r \frac{1}{T} \left\| \hat{F}^{(r)} - F^{(r)} H^{(r)} \right\|^2 &= O_P(h^2 + T^{-1} h^{-1} + N^{-1} \ln T) \text{ and} \\ \max_r \frac{1}{T} \left\| (\hat{F}^{(r)} - F^{(r)} H^{(r)} - B^{(r)})' F^{(r)} \right\| &= O_P(h^2 + T^{-1} h^{-1} + N^{-1} \ln T). \end{aligned}$$

Then using that  $\frac{1}{T} \hat{F}^{(r)'} \hat{F}^{(r)} = \mathbb{I}_R$ , we have uniformly in  $r$

$$\begin{aligned} &\left\| \left( H^{(r)} H^{(r)'} \right)^{-1} - \frac{1}{T} F^{(r)'} F^{(r)} \right\| \\ &= \left\| H^{(r)'}{}^{-1} \left[ \frac{1}{T} \hat{F}^{(r)'} \hat{F}^{(r)} - \frac{1}{T} H^{(r)'} F^{(r)'} F^{(r)} H^{(r)} \right] H^{(r)-1} \right\| \\ &\leq \left\| H^{(r)'}{}^{-1} \right\|^2 \left\| \frac{1}{T} (\hat{F}^{(r)} - F^{(r)} H^{(r)})' (\hat{F}^{(r)} - F^{(r)} H^{(r)}) \right\| + 2 \left\| H^{(r)'}{}^{-1} \right\| \left\| \frac{1}{T} (\hat{F}^{(r)} - F^{(r)} H^{(r)})' F^{(r)} \right\| \\ &= O_P(h^2 + T^{-1} h^{-1} + N^{-1} \ln T). \end{aligned}$$

Consequently,  $\max_r \|H^{(r)} H^{(r)'} - (\frac{1}{T} F^{(r)'} F^{(r)})^{-1}\| = O_P(h^2 + T^{-1} h^{-1} + N^{-1} \ln T)$ .

(ii) Using  $A^{-1} - B^{-1} = A^{-1}(B - A)B^{-1}$ , we have

$$\begin{aligned} &\max_t \left\| \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} - H^{(t)'} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \right\| \\ &= \left\| \left( \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right)^{-1} \left[ H^{(t)-1} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right) H^{(t)-1} - \frac{1}{N} \hat{\Lambda}_t' \hat{\Lambda}_t \right] H^{(t)'} \left( \frac{1}{N} \Lambda_t' \Lambda_t \right)^{-1} H^{(t)} \right\| \end{aligned}$$

$$\leq \max_t \left\| \left( \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t \right)^{-1} \right\| \left\| H^{(t)-1} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right) H^{(t)'} - \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t \right\| \left\| H^{(t)'} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right)^{-1} H^{(t)} \right\|$$

Note that  $\max_t \left\| \left( \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t \right)^{-1} \right\| = O_P(1)$  and  $\max_t \left\| H^{(t)'} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right)^{-1} H^{(t)} \right\| = O_P(1)$ . It remains to study

$$\max_t \left\| \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t - H^{(t)-1} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right) H^{(t)'} \right\| \leq \max_t \frac{1}{N} \left\| \hat{\Lambda}_t - \Lambda_t H^{(t)'} \right\|^2 + 2 \max_t \frac{1}{N} \left\| \left( \hat{\Lambda}_t - \Lambda_t H^{(t)'} \right)' \Lambda_t H^{(t)'} \right\|.$$

Following the proof of Lemma A.6(xii) and (xiii) in Su and Wang (2020), and the proof of Theorem 3.3(ii), we can show that

$$\begin{aligned} \max_t \frac{1}{N} \left\| \hat{\Lambda}_t - \Lambda_t H^{(t)'} - B_\Lambda^{(t)} \right\|^2 &= O_P(C_{NT}^{-2} \ln T) + O(\max_t \|\hat{\beta}_t - \beta_t\|) \text{ and} \\ \max_t \frac{1}{N} \left\| \left( \hat{\Lambda}_t - \Lambda_t H^{(t)'} - B_\Lambda^{(t)} \right)' \Lambda_t H^{(t)'} \right\| &= O_P(C_{NT}^{-2} \ln T) + O(\max_t \|\hat{\beta}_t - \beta_t\|). \end{aligned}$$

In addition,  $\max_t \frac{1}{N} \left\| B_\Lambda^{(t)} \right\|^2 = O_P(h^4)$  and  $\max_t \frac{1}{N} \left\| B_\Lambda^{(t)'} \Lambda_t H^{(t)'} \right\| = O_P(h^2)$ . Then

$$\max_t \left\| \left( \frac{1}{N} \hat{\Lambda}'_t \hat{\Lambda}_t \right)^{-1} - H^{(t)'} \left( \frac{1}{N} \Lambda'_t \Lambda_t \right)^{-1} H^{(t)} \right\| = O_P(h^2 + C_{NT}^{-2} \ln T) + O_P(\max_t \|\hat{\beta}_t - \beta_t\|) = O_P(h^2 + C_{NT}^{-2} \ln T). \blacksquare$$

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