

# Online Appendix to “Two-Step Estimation of Quantile Panel Data Models with Interactive Fixed Effects”

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## A Proof of the Main Results

### Definitions and Notations

For any random variable  $W_i$  or  $W_t$ ,  $W_i = \bar{O}_P(1)$  means that  $\max_{1 \leq i \leq N} \|W_i\| = O_P(1)$ , and  $W_t = \bar{O}_P(1)$  means that  $\max_{1 \leq t \leq T} \|W_t\| = O_P(1)$ .  $\bar{o}_P(1)$  is defined similarly. For notational simplicity, we suppress the dependence of  $\lambda_0(\tau)$ ,  $\hat{\beta}(\tau)$  and  $\beta_0(\tau)$  on  $\tau$ . Let  $M > 0$  denote a generic bounded constant that does not depend on  $N$  or  $T$ .

Define:

$$\begin{aligned}\mathbb{S}_{NT}(\beta, \Lambda, F) &= (NT)^{-1} \sum_{i=1}^N \sum_{t=1}^T l_{it}(\beta, \lambda_i, f_t) & \mathbb{S}_{NT}^*(\beta, \Lambda, F) &= (NT)^{-1} \sum_{i=1}^N \sum_{t=1}^T \rho_{it}(\beta, \lambda_i, f_t) \\ \mathbb{S}_{i,T}(\beta, \lambda_i, F) &= \frac{1}{T} \sum_{t=1}^T l_{it}(\beta, \lambda_i, f_t) & \mathbb{S}_{i,T}^*(\beta, \lambda_i, F) &= \frac{1}{T} \sum_{t=1}^T \rho_{it}(\beta, \lambda_i, f_t)\end{aligned}$$

where  $l(u) = [\tau - K(u/h)]u$ ,  $\rho_\tau(u)$  is the check function, and

$$l_{it}(\beta, \lambda_i, f_t) = l(Y_{it} - \beta' X_{it} - \lambda'_i f_t), \quad \rho_{it}(\beta, \lambda_i, f_t) = \rho_\tau(Y_{it} - \beta' X_{it} - \lambda'_i f_t).$$

For any random function  $L(\beta, \Lambda, F)$  and fixed  $(\beta, \Lambda, F)$ , define  $\bar{L}(\beta, \Lambda, F) = \mathbb{E}[L(\beta, \Lambda, F)]$  and  $\tilde{L}(\beta, \Lambda, F) = L(\beta, \Lambda, F) - \bar{L}(\beta, \Lambda, F)$ . Let  $l^{(j)}(u)$  denote the  $j$ th order derivative of  $l$ , i.e.,

$$l^{(1)}(u) = \tau - K(u/h) + k(u/h)u/h, \quad l^{(2)}(u) = 2k(u/h)1/h + k^{(1)}(u/h)u/h^2,$$

$$l^{(3)}(u) = 3k^{(1)}(u/h)1/h^2 + k^{(2)}(u/h)u/h^3, \quad l^{(4)}(u) = 4k^{(2)}(u/h)1/h^3 + k^{(3)}(u/h)u/h^4.$$

Let  $l_{it}^{(j)}(\beta, \lambda_i, f_t) = l^{(j)}(Y_{it} - \beta' X_{it} - \lambda'_i f_t)$  for  $j = 1, \dots, 4$ , and their arguments are dropped when evaluated at  $(\beta_0, \lambda_{0i}, f_{0t})$ .

Finally, define  $\tilde{\lambda}_{0i} = (\mathbf{H}'_0)^{-1} \lambda_{0i}$ ,  $\tilde{f}_{0t} = \mathbf{H}_0 f_{0t}$ ,  $\tilde{\Lambda}_0 = [\tilde{\lambda}_{01}, \dots, \tilde{\lambda}_{0N}]'$ , and  $\tilde{F}_0 = [\tilde{f}_{01}, \dots, \tilde{f}_{0T}]'$ .

## A.1 Proof of Proposition 1:

*Proof.* Note that

$$\begin{aligned} & \|\hat{\Sigma}_{\bar{x}} - \Gamma_0 \Sigma_{f_0} \Gamma'_0\| \\ &= \left\| \bar{\Gamma} \hat{\Sigma}_{f_0} \bar{\Gamma}' - \Gamma_0 \Sigma_{f_0} \Gamma'_0 + \bar{\Gamma} \cdot \frac{1}{T} \sum_{t=1}^T f_{0t} \bar{e}'_t + \frac{1}{T} \sum_{t=1}^T \bar{e}_t f'_{0t} \cdot \bar{\Gamma}' + \frac{1}{T} \sum_{t=1}^T \bar{e}_t \bar{e}'_t \right\| \\ &\leq \left\| \bar{\Gamma} \hat{\Sigma}_{f_0} \bar{\Gamma}' - \Gamma_0 \Sigma_{f_0} \Gamma'_0 \right\| + 2 \left\| \bar{\Gamma} \cdot \frac{1}{T} \sum_{t=1}^T f_{0t} \bar{e}'_t \right\| + \left\| \frac{1}{T} \sum_{t=1}^T \bar{e}_t \bar{e}'_t \right\|. \end{aligned}$$

First, by Assumption 1(ii),  $\left\| \bar{\Gamma} \hat{\Sigma}_{f_0} \bar{\Gamma}' - \Gamma_0 \Sigma_{f_0} \Gamma'_0 \right\| = O(N^{-1/2} + T^{-1/2})$ . Second, by Assumption 1(iii) and the Cauchy-Schwarz inequality, we have

$$\left\| \frac{1}{T} \sum_{t=1}^T f_{0t} \bar{e}'_t \right\| \leq \frac{1}{\sqrt{N}} \sqrt{\frac{1}{T} \sum_{t=1}^T \|f_{0t}\|^2} \sqrt{\frac{1}{T} \sum_{t=1}^T \|\sqrt{N} \bar{e}_t\|^2} = O_P(N^{-1/2})$$

and

$$\left\| \frac{1}{T} \sum_{t=1}^T \bar{e}_t \bar{e}'_t \right\| \leq \frac{1}{N} \cdot \frac{1}{T} \sum_{t=1}^T \|\sqrt{N} \bar{e}_t\|^2 = O_P(N^{-1}).$$

It then follows that  $\|\hat{\Sigma}_{\bar{x}} - \Gamma_0 \Sigma_{f_0} \Gamma'_0\| = O_P(N^{-1/2} + T^{-1/2})$ . Third, by matrix perturbation theory (Hoffman-Wielandt inequality) and the fact that  $\Gamma_0 \Sigma_{f_0} \Gamma'_0$  is a matrix with rank  $r$  (Assumption 1(ii)), it can be concluded that  $\hat{\rho}_1, \dots, \hat{\rho}_r$  converge in probability to some positive constants, while  $\hat{\rho}_{r+1}, \dots, \hat{\rho}_k$  are all  $O_P(N^{-1/2} + T^{-1/2})$ . Thus, it follows that

$$P[\hat{r} \neq r] \leq P[\hat{r} < r] + P[\hat{r} > r] \leq P[\hat{\rho}_r < \mathbb{P}_{NT}] + P[\hat{\rho}_{r+1} \geq \mathbb{P}_{NT}] \rightarrow 0,$$

and the desired result follows.  $\square$

## A.2 Proof of Theorem 1

**Lemma 1.** Define  $\hat{\mathbf{H}} = \hat{\Psi}' \bar{\Gamma}$ . Under Assumptions 1 and 2, (i)  $\hat{f}_t = \hat{\mathbf{H}} f_{0t} + \hat{\Psi}' \bar{e}_t$ ; (ii)  $\hat{\Psi} \xrightarrow{p} \Psi_0$  and  $\hat{\mathbf{H}} \xrightarrow{p} \mathbf{H}_0 = \Psi'_0 \Gamma_0$ ; (iii)  $\hat{\mathbf{H}}$  is invertible with probability approaching 1.

*Proof.* Result (i) follows directly from  $\bar{X}_t = \bar{\Gamma} f_{0t} + \bar{e}_t$  and  $\hat{f}_t = \hat{\Psi}' \bar{X}_t$ . Given Assumption 2(i), it follows from the Bauer-Fike Theorem and  $\|\hat{\Sigma}_{\bar{x}} - \Gamma_0 \Sigma_{f_0} \Gamma'_0\| = o_P(1)$  (see the proof of Proposition 1) that  $\|\hat{\Psi} - \Psi_0\| = o_P(1)$ . Thus, result (ii) follows since  $\|\bar{\Gamma} - \Gamma_0\| = o(1)$  by Assumption 1(ii). Finally, suppose that  $\text{rank}(\mathbf{H}_0) < r$ , and let  $\mathcal{D}$  be the diagonal matrix with the eigenvalues of  $\Gamma_0 \Sigma_{f_0} \Gamma'_0$  as the diagonal

elements, then  $\text{rank}(\mathcal{D}) = \text{rank}(\Psi_0' \Gamma_0 \Sigma_{f_0} \Gamma_0' \Psi_0) = \text{rank}(\mathbf{H}_0 \Sigma_{f_0} \mathbf{H}'_0) \leq \text{rank}(\mathbf{H}_0) < r$ , which contradicts with Assumption 1(ii). Thus, we have  $\text{rank}(\mathbf{H}_0) = r$  and result (iii) follows from  $\hat{\mathbf{H}} = \mathbf{H}_0 + o_P(1)$ .  $\square$

## Proof of Theorem 1

*Proof.* **Step 1:** By Lemma 1, we have

$$\frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^2 \leq \frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \hat{\mathbf{H}} f_{0t}\|^2 + M \|\mathbf{H}_0 - \hat{\mathbf{H}}\|^2 \leq \frac{r}{NT} \sum_{t=1}^T \|\sqrt{N} \bar{e}_t\|^2 + o_P(1) = O_P\left(\frac{1}{N}\right) + o_P(1).$$

**Step 2:** Adding and subtracting terms, we can write

$$\mathbb{S}_{NT}(\beta, \Lambda, F) = (\mathbb{S}_{NT}(\beta, \Lambda, F) - \mathbb{S}_{NT}^*(\beta, \Lambda, F)) + (\mathbb{S}_{NT}^*(\beta, \Lambda, F) - \mathbb{S}_{NT}^*(\beta, \Lambda, \tilde{F}_0)) + \mathbb{S}_{NT}^*(\beta, \Lambda, \tilde{F}_0).$$

By the definition of the estimators,  $\mathbb{S}_{NT}(\hat{\beta}, \hat{\Lambda}, \hat{F}) \leq \mathbb{S}_{NT}(\beta_0, \tilde{\Lambda}_0, \hat{F})$ . Thus, we have

$$\begin{aligned} \mathbb{S}_{NT}^*(\hat{\beta}, \hat{\Lambda}, \tilde{F}_0) - \mathbb{S}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0) &\leq [\mathbb{S}_{NT}(\beta_0, \tilde{\Lambda}_0, \hat{F}) - \mathbb{S}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \hat{F})] + [\mathbb{S}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \hat{F}) - \mathbb{S}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0)] \\ &\quad - [\mathbb{S}_{NT}(\hat{\beta}, \hat{\Lambda}, \hat{F}) - \mathbb{S}_{NT}^*(\hat{\beta}, \hat{\Lambda}, \hat{F})] - [\mathbb{S}_{NT}^*(\hat{\beta}, \hat{\Lambda}, \hat{F}) - \mathbb{S}_{NT}^*(\hat{\beta}, \hat{\Lambda}, \tilde{F}_0)]. \end{aligned} \quad (\text{A.1})$$

**Step 3:** Let  $\delta$  be a positive number close to 0. Define  $B_{\delta,i} = \{\beta \in \mathcal{B}, \lambda_i \in \mathcal{A} : \|\beta - \beta_0\|_1 + \|\lambda_i - \tilde{\lambda}_{0i}\|_1 \leq \delta\}$ . Consider any  $(\beta, \lambda_i) \in B_{\delta,i}^C$ . Let  $m = \|\beta - \beta_0\|_1 + \|\lambda_i - \tilde{\lambda}_{0i}\|_1 > \delta$ , then  $(\bar{\beta}, \bar{\lambda}_i) = (\beta, \lambda_i)\delta/m + (\beta_0, \tilde{\lambda}_{0i})(1 - \delta/m)$  is on the boundary of  $B_{\delta,i}$ . Note that given  $X_{it}$  and  $f_t$ , the check function  $\rho_{it}$  is convex in  $(\beta, \lambda_i)$ . Thus,

$$\delta/m \cdot \rho_{it}(\beta, \lambda_i, \tilde{f}_{0t}) + (1 - \delta/m) \cdot \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) \geq \rho_{it}(\bar{\beta}, \bar{\lambda}_i, \tilde{f}_{0t}),$$

and it follows that

$$\rho_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) \geq m/\delta \cdot [\rho_{it}(\bar{\beta}, \bar{\lambda}_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})]$$

and

$$\mathbb{S}_{NT}^*(\beta, \Lambda, \tilde{F}_0) - \mathbb{S}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0) \gtrsim \mathbb{S}_{NT}^*(\bar{\beta}, \bar{\Lambda}, \tilde{F}_0) - \mathbb{S}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0)$$

if  $(\beta, \lambda_i) \in B_{\delta,i}^C$  for all  $i$ .

Write

$$\mathbb{S}_{NT}^*(\bar{\beta}, \bar{\Lambda}, \tilde{F}_0) - \mathbb{S}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0) = \bar{\mathbb{S}}_{NT}^*(\bar{\beta}, \bar{\Lambda}, \tilde{F}_0) - \bar{\mathbb{S}}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0) + \tilde{\mathbb{S}}_{NT}^*(\bar{\beta}, \bar{\Lambda}, \tilde{F}_0) - \tilde{\mathbb{S}}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0).$$

First, by Taylor expansion  $\bar{\rho}_{it}(\beta, \lambda_i, \tilde{f}_{0t})$  around  $(\beta_0, \tilde{\lambda}_{0i})$  we have

$$\begin{aligned}\bar{\mathbb{S}}_{NT}^*(\bar{\beta}, \bar{\Lambda}, \tilde{F}_0) - \bar{\mathbb{S}}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0) &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \bar{\rho}_{it}(\bar{\beta}, \bar{\lambda}_i, \tilde{f}_{0t}) - \bar{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) \\ &= \frac{1}{N} \sum_{i=1}^N [(\bar{\beta} - \beta_0)', (\bar{\lambda}_i - \lambda_{0i})] \cdot \left( \frac{1}{T} \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it})V_{it}V'_{it}] \right) \cdot [(\bar{\beta} - \beta_0)', (\bar{\lambda}_i - \lambda_{0i})]' + o(\delta^2) \\ &\geq \delta^2 \cdot \frac{1}{N} \sum_{i=1}^N \varrho_{i,T} + o(\delta^2) \geq \underline{\varrho} \delta^2 \quad (\text{A.2})\end{aligned}$$

by Assumption 2(iii). Second, we have

$$\left| \tilde{\mathbb{S}}_{NT}^*(\bar{\beta}, \bar{\Lambda}, \tilde{F}_0) - \tilde{\mathbb{S}}_{NT}^*(\beta_0, \tilde{\Lambda}_0, \tilde{F}_0) \right| \leq \max_{1 \leq i \leq N} \sup_{(\beta, \lambda_i) \in B_{\delta,i}} \left| \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] \right|. \quad (\text{A.3})$$

**Step 4:**  $\|\hat{\beta} - \beta_0\|_1 > \delta$  implies that  $(\hat{\beta}, \hat{\lambda}_i) \in B_{\delta,i}^C$  for all  $i$ . It then follows from (A.1), (A.2) and (A.3) that for small  $\delta > 0$ , there exists an  $\epsilon > 0$  (depending on  $\delta$ ) such that

$$\begin{aligned}P[\|\hat{\beta} - \beta_0\|_1 > \delta] &\leq P \left[ \sup_{\beta, \Lambda, F} \left| \mathbb{S}_{NT}(\beta, \Lambda, F) - \mathbb{S}_{NT}^*(\beta, \Lambda, F) \right| > 1/3\epsilon \right] \\ &\quad + P \left[ \sup_{\beta, \lambda_i \in \mathcal{A}} \left| \mathbb{S}_{NT}^*(\beta, \Lambda, \hat{F}) - \mathbb{S}_{NT}^*(\beta, \Lambda, \tilde{F}_0) \right| > 1/3\epsilon \right] \\ &\quad + P \left[ \max_{1 \leq i \leq N} \sup_{(\beta, \lambda_i) \in B_{\delta,i}} \left| \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] \right| > 1/3\epsilon \right]. \quad (\text{A.4})\end{aligned}$$

The first term on the right-hand side of (A.4) is  $o(1)$  because it is easy to show that<sup>1</sup>:

$$\sup_{\beta, \Lambda, F} \left| \mathbb{S}_{NT}(\beta, \Lambda, F) - \mathbb{S}_{NT}^*(\beta, \Lambda, F) \right| \lesssim h$$

and  $h \rightarrow 0$  as  $N, T \rightarrow \infty$ . The second term on the right-hand side of (A.4) is  $o(1)$  since by the result of Step 1 and Assumption 2(ii),

$$\begin{aligned}\sup_{\beta, \lambda_i \in \mathcal{A}} \left| \mathbb{S}_{NT}^*(\beta, \Lambda, \hat{F}) - \mathbb{S}_{NT}^*(\beta, \Lambda, F_0) \right| &\leq \max_{1 \leq i \leq N} \sup_{\beta, \lambda_i \in \mathcal{A}} \left| \frac{1}{T} \sum_{t=1}^T \rho_{it}(\beta, \lambda_i, \hat{f}_t) - \frac{1}{T} \sum_{t=1}^T \rho_{it}(\beta, \lambda_i, \tilde{f}_{0t}) \right| \\ &\lesssim \sqrt{\frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^2} = o_P(1).\end{aligned}$$

Finally, for the consistency of  $\hat{\beta}$ , it remains to show that the third term on the right-hand side of (A.4)

<sup>1</sup>Note that  $|l(u) - \rho_\tau(u)| = |(\tau - \mathbf{1}\{u \leq 0\})u - (\tau - K(u/h))u| \leq |u| \cdot |\mathbf{1}\{u \leq 0\} - K(u/h)| \lesssim |u| \cdot \mathbf{1}\{|u| \leq h\} \lesssim h$ .

is  $o(1)$ . By the union bound, it suffices to show that for all  $i$

$$P \left[ \sup_{(\beta, \lambda_i) \in B_{\delta, i}} \left| \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] \right| > 1/3\epsilon \right] = o(N^{-1}). \quad (\text{A.5})$$

**Step 5:** Write  $\theta_i = (\beta', \lambda'_i)'$ , and  $\theta_{0i} = (\beta'_0, \tilde{\lambda}'_{0i})'$ . Define  $\Delta_{it}(\theta_i) = \rho_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})$ . Note that there exists  $C_1, C_2 > 0$  such that  $|\Delta_{it}(\theta_i^a) - \Delta_{it}(\theta_i^b)| \leq C_1 \cdot \|\beta^a - \beta^b\| \cdot \|X_{it}\| + C_2 \cdot \|\lambda_i^a - \lambda_i^b\|$ . Suppose that there exists  $M_X > 0$  such that  $\mathbb{E}\|X_{it}\| \leq M_X$  for all  $i, t$  (see Assumption 2(v)).

Since  $B_{\delta, i}$  is compact, for any  $\eta > 0$ , there exists a positive integer  $L$  and a maximal set of points  $\theta_i^{(1)}, \dots, \theta_i^{(L)}$  in  $B_{\delta, i}$  such that  $\|\theta_i^{(k)} - \theta_i^{(j)}\| \geq \eta$  for any  $k \neq j$ . For any  $\theta_i \in B_{\delta, i}$ , let  $\theta_i^* = \{\theta_i^{(j)} : 1 \leq j \leq L, \|\theta_i - \theta_i^{(j)}\| \leq \eta\}$ . Then,

$$\begin{aligned} \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] &= \frac{1}{T} \sum_{t=1}^T [\Delta_{it}(\theta_i) - \mathbb{E}(\Delta_{it}(\theta_i))] \\ &= \frac{1}{T} \sum_{t=1}^T [\Delta_{it}(\theta_i^*) - \mathbb{E}(\Delta_{it}(\theta_i^*))] + \frac{1}{T} \sum_{t=1}^T [\Delta_{it}(\theta_i) - \Delta_{it}(\theta_i^*) - \mathbb{E}(\Delta_{it}(\theta_i) - \Delta_{it}(\theta_i^*))], \end{aligned}$$

and

$$\begin{aligned} \sup_{(\beta, \lambda_i) \in B_{\delta, i}} \left| \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] \right| &\leq \max_{1 \leq j \leq L} \left| \frac{1}{T} \sum_{t=1}^T [\Delta_{it}(\theta_i^{(j)}) - \mathbb{E}(\Delta_{it}(\theta_i^{(j)}))] \right| \\ &\quad + \sup_{\|\theta^a - \theta^b\| \leq \eta} \left| \frac{1}{T} \sum_{t=1}^T [\Delta_{it}(\theta_i^a) - \Delta_{it}(\theta_i^b) - \mathbb{E}(\Delta_{it}(\theta_i^a) - \Delta_{it}(\theta_i^b))] \right|. \end{aligned}$$

Note that

$$\begin{aligned} \sup_{\|\theta^a - \theta^b\| \leq \eta} \left| \frac{1}{T} \sum_{t=1}^T [\Delta_{it}(\theta_i^a) - \Delta_{it}(\theta_i^b) - \mathbb{E}(\Delta_{it}(\theta_i^a) - \Delta_{it}(\theta_i^b))] \right| &\leq C_1 \eta \left( \frac{1}{T} \sum_{t=1}^T (\|X_{it}\| - \mathbb{E}\|X_{it}\|) \right) + 2(C_2 + C_1 M_X) \eta, \end{aligned}$$

it then follows from the previous two inequalities that

$$\begin{aligned} P \left[ \sup_{(\beta, \lambda_i) \in B_{\delta, i}} \left| \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] \right| > 1/3\epsilon \right] &\leq \sum_{j=1}^L P \left[ \left| \frac{1}{T} \sum_{t=1}^T [\Delta_{it}(\theta_i^{(j)}) - \mathbb{E}(\Delta_{it}(\theta_i^{(j)}))] \right| \geq 1/9\epsilon \right] \\ &\quad + P \left[ C_1 \eta \left| \frac{1}{T} \sum_{t=1}^T (\|X_{it}\| - \mathbb{E}\|X_{it}\|) \right| \geq 1/9\epsilon \right] + P [2(C_2 + C_1 M_X) \eta \geq 1/9\epsilon]. \quad (\text{A.6}) \end{aligned}$$

First, choosing  $\eta < \epsilon/(18(C_2 + C_1 M_X))$ , the last term on the right-hand side of (A.6) is 0.

Second, for any  $\theta_i \in B_{\delta,i}$ ,  $\mathbb{E}|\Delta_{it}(\theta_i)|^{2m+\gamma} \leq M \cdot \mathbb{E}\|X_{it}\|^{2m+\gamma} + O(1) < \infty$  by Assumption 2(v). Thus, by Assumption 2(iv) and Theorem 3 of Yoshihara (1978) we have

$$\mathbb{E} \left| T^{-1/2} \sum_{t=1}^T \left[ \Delta_{it}(\theta_i^{(j)}) - \mathbb{E}(\Delta_{it}(\theta_i^{(j)})) \right] \right|^{2m} \leq M,$$

and by Markov's inequality,

$$P \left[ \left| \frac{1}{T} \sum_{t=1}^T \left[ \Delta_{it}(\theta_i^{(j)}) - \mathbb{E}(\Delta_{it}(\theta_i^{(j)})) \right] \right| \geq 1/9\epsilon \right] = O(T^{-m}).$$

Finally, we can show that the second term on the right-hand side of (A.6) is  $O(T^{-m})$  in a similar way. Thus, (A.5) follows since  $N/T^m \rightarrow 0$  by Assumption 2(vi). This completes the proof.  $\square$

### A.3 Proof of Theorem 2

To simplify the notations, write  $\check{f}_{0t} = \hat{\mathbf{H}} f_{0t}$  and  $\check{\lambda}_{0i} = (\hat{\mathbf{H}}')^{-1} \lambda_{0i}$ .

**Lemma 2.** Under Assumptions 1 to 4,

- (i)  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\| = O_P(N^{-1/2})$ ,  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^2 = O_P(N^{-1})$ ,  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^3 = O_P(N^{-3/2})$ ,  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^4 = O_P(N^{-2})$ .
- (ii)  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\| = O_P(N^{-1/2})$ ,  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^2 = O_P(N^{-1})$ ,  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^3 = O_P(N^{-3/2})$ ,  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^4 = O_P(N^{-2})$ .
- (iii)  $\max_{1 \leq t \leq T} \|\hat{f}_t - \check{f}_{0t}\| = O_P(T^{1/2m}/\sqrt{N})$  and  $\max_{1 \leq t \leq T} \|\hat{f}_t - \tilde{f}_{0t}\| = O_P(T^{1/2m}/\sqrt{N})$ .

*Proof.* By the properties of  $L_p$  norms in the Euclidean space, it suffices to prove that  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^4 = O_P(N^{-2})$  and  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^4 = O_P(N^{-2})$ . Note that Assumption 3(iv) implies that  $\{e_{1t}, \dots, e_{Nt}\}$  is independent across  $i$  and  $\mathbb{E}\|e_{it}\|^4 < M$  for all  $i, t$ . Thus,

$$\mathbb{E}\|\sqrt{N}\bar{e}_t\|^4 = \mathbb{E} \left\| \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right\|^4 \leq \left( \frac{1}{N} \sum_{i=1}^N \mathbb{E}\|e_{it}\|^2 \right)^2 + \frac{1}{N^2} \sum_{i=1}^N \mathbb{E}\|e_{it}\|^4 = O(1),$$

and by Lemma 1

$$\frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^4 \leq \|\hat{\Psi}\|^4 \cdot \frac{1}{N^2 T} \sum_{t=1}^T \|\sqrt{N}\bar{e}_t\|^4 = O_P(N^{-2}).$$

Moreover,  $T^{-1} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^4 \leq T^{-1} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^4 + C\|\hat{\mathbf{H}} - \mathbf{H}_0\|^4$ . Then result (ii) follows if we can show that  $\|\hat{\mathbf{H}} - \mathbf{H}_0\| = O_P(N^{-1/2})$ .

By definition,  $\|\hat{\mathbf{H}} - \mathbf{H}_0\| \leq O_P(\|\hat{\Psi} - \Psi_0\|) + O_P(\|\bar{\Gamma} - \Gamma_0\|)$ . By the proof of Proposition 1 and Lemma 1 we have  $\|\hat{\Psi} - \Psi_0\| \lesssim \|\hat{\Sigma}_{\bar{x}} - \Gamma_0 \Sigma_{f_0} \Gamma_0'\| = O_P(N^{-1/2} + T^{-1/2})$ . Then the result (ii) follows from Assumption 1(ii) and the fact that  $N \asymp T$ .

Finally, note that  $\max_{1 \leq t \leq T} \|\hat{f}_t - \check{f}_{0t}\| \leq O_P(1) \cdot \max_{1 \leq t \leq T} \|\bar{e}_t\|$ . For  $1 \leq h \leq r$ , it is easy to show

that  $\mathbb{E}|\sqrt{N}\bar{e}_{th}|^{2m} \leq M$  for all  $t$ . Thus, it follows from Lemma 2.2.1 and Lemma 2.2.2 of [van der Vaart and Wellner \(1996\)](#) that  $\mathbb{E}[\max_{1 \leq t \leq T} \sqrt{N}|\bar{e}_{th}|] = O(T^{1/2m})$ . Thus, result (iii) follows since  $\|\hat{\mathbf{H}} - \mathbf{H}_0\| = o_P(1)$ .  $\square$

**Lemma 3.** *Under Assumptions 1 to 4,  $\max_{1 \leq i \leq N} \|\hat{\lambda}_i - \tilde{\lambda}_{0i}\| = o_P(1)$ .*

*Proof.* Recall that  $\tilde{\lambda}_{0i} = (\mathbf{H}'_0)^{-1}\lambda_{0i}$ ,  $\tilde{f}_{0t} = \mathbf{H}_0 f_{0t}$ . By the definition of the estimators we have  $\mathbb{S}_{i,T}(\hat{\beta}, \hat{\lambda}_i, \hat{F}) \leq \mathbb{S}_{i,T}(\hat{\beta}, \tilde{\lambda}_{0i}, \hat{F})$  for each  $i$ . Note that

$$\mathbb{S}_{i,T}(\beta, \lambda_i, F) = \mathbb{S}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0) + (\mathbb{S}_{i,T}(\beta, \lambda_i, F) - \mathbb{S}_{i,T}^*(\beta, \lambda_i, F)) + (\mathbb{S}_{i,T}^*(\beta, \lambda_i, F) - \mathbb{S}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0)).$$

Thus,  $\mathbb{S}_{i,T}(\hat{\beta}, \hat{\lambda}_i, \hat{F}) \leq \mathbb{S}_{i,T}(\hat{\beta}, \tilde{\lambda}_{0i}, \hat{F})$  implies that

$$\begin{aligned} \mathbb{S}_{i,T}^*(\beta_0, \hat{\lambda}_i, \tilde{F}_0) - \mathbb{S}_{i,T}^*(\beta_0, \tilde{\lambda}_{0i}, \tilde{F}_0) &\leq (\mathbb{S}_{i,T}(\hat{\beta}, \tilde{\lambda}_{0i}, \hat{F}) - \mathbb{S}_{i,T}^*(\hat{\beta}, \tilde{\lambda}_{0i}, \hat{F})) + (\mathbb{S}_{i,T}^*(\hat{\beta}, \tilde{\lambda}_{0i}, \hat{F}) - \mathbb{S}_{i,T}^*(\beta_0, \tilde{\lambda}_{0i}, \tilde{F}_0)) \\ &\quad - (\mathbb{S}_{i,T}(\hat{\beta}, \hat{\lambda}_i, \hat{F}) - \mathbb{S}_{i,T}^*(\hat{\beta}, \hat{\lambda}_i, \hat{F})) - (\mathbb{S}_{i,T}^*(\hat{\beta}, \hat{\lambda}_i, \hat{F}) - \mathbb{S}_{i,T}^*(\beta_0, \hat{\lambda}_i, \tilde{F}_0)). \end{aligned} \quad (\text{A.7})$$

Similar to the proof of Theorem 1, for small  $\delta > 0$ , define  $B_{\delta,i} = \{\lambda_i \in \mathcal{A} : \|\lambda_i - \tilde{\lambda}_{0i}\| \leq \delta\}$ . For any  $\lambda_i \in B_{\delta,i}^C$ , let  $m = \|\lambda_i - \tilde{\lambda}_{0i}\| > \delta$ . Then  $\bar{\lambda}_i = \lambda_i \cdot \delta/m + \tilde{\lambda}_{0i} \cdot (1 - \delta/m)$  is on the boundary of  $B_{\delta,i}$ , i.e.,  $\|\bar{\lambda}_i - \tilde{\lambda}_{0i}\| = \delta$ . Given  $\beta_0$  and  $\tilde{f}_{0t}$ , the check function is convex in  $\lambda_i$ , thus

$$\delta/m \cdot \rho_{it}(\beta_0, \lambda_i, \tilde{f}_{0t}) + (1 - \delta/m) \cdot \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) \geq \rho_{it}(\beta_0, \bar{\lambda}_i, \tilde{f}_{0t}),$$

and it follows that

$$\rho_{it}(\beta_0, \lambda_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) \geq m/\delta \cdot \left( \rho_{it}(\beta_0, \bar{\lambda}_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) \right).$$

Note that  $\rho_{it}(\beta_0, \bar{\lambda}_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) = \bar{\rho}_{it}(\beta_0, \bar{\lambda}_i, \tilde{f}_{0t}) - \bar{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) + \tilde{\rho}_{it}(\beta_0, \bar{\lambda}_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})$ , and

$$\bar{\rho}_{it}(\beta_0, \bar{\lambda}_i, \tilde{f}_{0t}) - \bar{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t}) \geq (\bar{\lambda}_i - \tilde{\lambda}_{0i})' \left( \mathbf{f}_{it}(0) \tilde{f}_{0t} \tilde{f}'_{0t} \right) (\bar{\lambda}_i - \tilde{\lambda}_{0i}) + o(\delta^2).$$

Thus, if  $\|\hat{\lambda}_i - \tilde{\lambda}_{0i}\| > \delta$ , by Assumption 3(ii)

$$\begin{aligned} \mathbb{S}_{i,T}^*(\beta_0, \hat{\lambda}_i, \tilde{F}_0) - \mathbb{S}_{i,T}^*(\beta_0, \tilde{\lambda}_{0i}, \tilde{F}_0) &= \frac{1}{T} \sum_{t=1}^T [\rho_{it}(\beta_0, \hat{\lambda}_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] \\ &\geq (\hat{\lambda}_i - \tilde{\lambda}_{0i})' \left( \frac{1}{T} \sum_{t=1}^T \mathbf{f}_{it}(0) \tilde{f}_{0t} \tilde{f}'_{0t} \right) (\hat{\lambda}_i - \tilde{\lambda}_{0i}) + o(\delta^2) + m/\delta \cdot \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta_0, \hat{\lambda}_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})] \\ &\geq C\delta^2 + m/\delta \cdot \frac{1}{T} \sum_{t=1}^T [\tilde{\rho}_{it}(\beta_0, \hat{\lambda}_i, \tilde{f}_{0t}) - \tilde{\rho}_{it}(\beta_0, \tilde{\lambda}_{0i}, \tilde{f}_{0t})], \end{aligned}$$

where  $\hat{\lambda}_i$  is between  $\tilde{\lambda}_{0i}$  and  $\hat{\lambda}_i$  and is on the boundary of  $B_{\delta,i}$ . Thus, it follows from (A.7) that there

exists some  $\epsilon > 0$  (depending on  $\delta$ ) such that

$$\begin{aligned} P \left[ \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \tilde{\lambda}_{0i}\| > \delta \right] &\leq P \left[ \max_{1 \leq i \leq N} \sup_{\beta, \lambda_i, F} |\mathbb{S}_{i,T}(\beta, \lambda_i, F) - \mathbb{S}_{i,T}^*(\beta, \lambda_i, F)| > \epsilon \right] \\ &\quad + P \left[ \max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} |\mathbb{S}_{i,T}^*(\hat{\beta}, \lambda_i, \hat{F}) - \mathbb{S}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0)| > \epsilon \right] \\ &\quad + P \left[ \max_{1 \leq i \leq N} \sup_{\lambda_i \in B_{\delta,i}} |\tilde{\mathbb{S}}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0) - \tilde{\mathbb{S}}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0)| > \epsilon \right]. \quad (\text{A.8}) \end{aligned}$$

Similar to the proof of Theorem 1, it can be shown that the first and last term on the right-hand side of (A.8) are both  $o(1)$ . It remains to show that the second term is  $o(1)$ .

By the property of the check function, we have

$$\begin{aligned} &\max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} |\mathbb{S}_{i,T}^*(\hat{\beta}, \lambda_i, \hat{F}) - \mathbb{S}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0)| \\ &\leq \max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} |\mathbb{S}_{i,T}^*(\hat{\beta}, \lambda_i, \hat{F}) - \mathbb{S}_{i,T}^*(\hat{\beta}, \lambda_i, \tilde{F}_0)| + \max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} |\mathbb{S}_{i,T}^*(\hat{\beta}, \lambda_i, \tilde{F}_0) - \mathbb{S}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0)| \\ &\lesssim \frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\| + \max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} \left| \frac{1}{T} \sum_{t=1}^T \Delta_{it}(\hat{\beta}, \lambda_i) - \bar{\Delta}_{it}(\hat{\beta}, \lambda_i) \right| + \max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} \left| \frac{1}{T} \sum_{t=1}^T \bar{\Delta}_{it}(\hat{\beta}, \lambda_i) \right| \\ &\leq o_P(1) + \max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} \sup_{\beta \in \mathcal{B}} \left| \frac{1}{T} \sum_{t=1}^T \Delta_{it}(\beta, \lambda_i) - \bar{\Delta}_{it}(\beta, \lambda_i) \right| + O_P(\|\hat{\beta} - \beta_0\|) \end{aligned}$$

where  $\Delta_{it}(\beta, \lambda_i) = \rho_{it}(\beta, \lambda_i, \tilde{f}_{0t}) - \rho_{it}(\beta_0, \lambda_i, \tilde{f}_{0t})$  and  $\bar{\Delta}_{it}(\beta, \lambda_i) = \mathbb{E}\Delta_{it}(\beta, \lambda_i)$ . Similar to the proof of (A.2) in Kato et al. (2012) it can be shown that

$$\max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} \sup_{\beta \in \mathcal{B}} \left| \frac{1}{T} \sum_{t=1}^T \Delta_{it}(\beta, \lambda_i) - \bar{\Delta}_{it}(\beta, \lambda_i) \right| = o_P(1).$$

It then follows from  $\|\hat{\beta} - \beta_0\| = o_P(1)$  and Lemma 2(ii) that

$$\max_{1 \leq i \leq N} \sup_{\lambda_i \in \mathcal{A}} |\mathbb{S}_{i,T}^*(\hat{\beta}, \lambda_i, \hat{F}) - \mathbb{S}_{i,T}^*(\beta_0, \lambda_i, \tilde{F}_0)| = o_P(1).$$

This implies that the second term on the right-hand side of (A.8) is  $o(1)$ . The desired result follows by noting that  $\max_i \|\tilde{\lambda}_{0i} - \check{\lambda}_{0i}\| \leq O_P(1) \cdot \|\hat{\mathbf{H}} - \mathbf{H}_0\| = o_P(1)$ .  $\square$

**Lemma 4.** Under Assumptions 1 to 4,  $\|\hat{\beta} - \beta_0\| = o_P(\max_i \|\hat{\lambda}_i - \check{\lambda}_{0i}\|) + o_P(1/\sqrt{T})$ .

*Proof.* **Step 1:** Define the following notations:

$$\underbrace{S^\beta(\beta, \Lambda, F)}_{p \times 1} = \partial \mathbb{S}_{NT}(\beta, \Lambda, F) / \partial \beta, \quad \underbrace{S^\lambda(\beta, \Lambda, F)}_{Nr \times 1} = \partial \mathbb{S}_{NT}(\beta, \Lambda, F) / \partial \Lambda,$$

$$\underbrace{S^{\beta\beta'}(\beta, \Lambda, F)}_{p \times p} = \partial^2 \mathbb{S}_{NT}(\beta, \Lambda, F) / \partial \beta \partial \beta', \quad \underbrace{S^{\beta\lambda'}(\beta, \Lambda, F)}_{p \times Nr} = \partial^2 \mathbb{S}_{NT}(\beta, \Lambda, F) / \partial \beta \partial \Lambda'.$$

The other functions such as  $S^{\beta f'}(\beta, \Lambda, F)$ ,  $S^{\lambda\lambda'}(\beta, \Lambda, F)$ ,  $S^{\lambda f'}(\beta, \Lambda, F)$ ,  $S^{\beta\beta' f_{th}}(\beta, \Lambda, F)$  are defined in a similar fashion. The arguments of these functions are dropped when they are evaluated at  $(\beta, \Lambda, F) = (\beta_0, \check{\Lambda}_0, \check{F}_0)$ , where  $\check{\Lambda}_0 = (\check{\lambda}_{01}, \dots, \check{\lambda}_{0N})'$ ,  $\check{F}_0 = (\check{f}_{01}, \dots, \check{f}_{0T})'$  (recall that  $\check{f}_{0t} = \hat{\mathbf{H}} f_{0t}$  and  $\check{\lambda}_{0i} = (\hat{\mathbf{H}}')^{-1} \lambda_{0i}$ ).

Expanding  $S_{NT}^\beta(\hat{\beta}, \hat{\Lambda}, \hat{F})$  and  $S_{NT}^\lambda(\hat{\beta}, \hat{\Lambda}, \hat{F})$  around  $(\beta_0, \check{\Lambda}_0, \check{F}_0)$  up to the third order gives:

$$0 = S^\beta(\hat{\beta}, \hat{\Lambda}, \hat{F}) = S^\beta + S^{\beta\beta'}(\hat{\beta} - \beta_0) + S^{\beta\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) + S^{\beta f'}(\hat{F} - \check{F}_0) + 1/2R^\beta(\beta^*, \Lambda^*, F^*), \quad (\text{A.9})$$

$$0 = S^\lambda(\hat{\beta}, \hat{\Lambda}, \hat{F}) = S^\lambda + S^{\lambda\beta'}(\hat{\beta} - \beta_0) + S^{\lambda\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) + S^{\lambda f'}(\hat{F} - \check{F}_0) + 1/2R^\lambda(\beta^*, \Lambda^*, F^*), \quad (\text{A.10})$$

where  $(\beta^*, \Lambda^*, F^*)$  is between  $(\beta_0, \check{\Lambda}_0, \check{F}_0)$  and  $(\hat{\beta}, \hat{\Lambda}, \hat{F})$ , and

$$\begin{aligned} R^\beta(\beta^*, \Lambda^*, F^*) = & \sum_{k=1}^p S_*^{\beta\beta'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\beta} - \beta_0) + \sum_{k=1}^p S_*^{\beta\lambda'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\Lambda} - \check{\Lambda}_0) + \sum_{k=1}^p S_*^{\beta f'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{F} - \check{F}_0) \\ & + \sum_{i=1}^N \sum_{h=1}^r S_*^{\beta\beta'\lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{\beta} - \beta_0) + \sum_{i=1}^N \sum_{h=1}^r S_*^{\beta\lambda'\lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{\Lambda} - \check{\Lambda}_0) + \sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f'\lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{F} - \check{F}_0) \\ & + \sum_{t=1}^T \sum_{h=1}^r S_*^{\beta\beta' f_{th}}(\hat{f}_{th} - \check{f}_{0t,h})(\hat{\beta} - \beta_0) + \sum_{t=1}^T \sum_{h=1}^r S_*^{\beta\lambda' f_{th}}(\hat{f}_{th} - \check{f}_{0t,h})(\hat{\Lambda} - \check{\Lambda}_0) + \sum_{t=1}^T \sum_{h=1}^r S_*^{\beta f' f_{th}}(\hat{f}_{th} - \check{f}_{0t,h})(\hat{F} - \check{F}_0), \end{aligned} \quad (\text{A.11})$$

$$\begin{aligned} R^\lambda(\beta^*, \Lambda^*, F^*) = & \sum_{k=1}^p S_*^{\lambda\beta'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\beta} - \beta_0) + \sum_{k=1}^p S_*^{\lambda\lambda'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\Lambda} - \check{\Lambda}_0) + \sum_{k=1}^p S_*^{\lambda f'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{F} - \check{F}_0) \\ & + \sum_{i=1}^N \sum_{h=1}^r S_*^{\lambda\beta'\lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{\beta} - \beta_0) + \sum_{i=1}^N \sum_{h=1}^r S_*^{\lambda\lambda'\lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{\Lambda} - \check{\Lambda}_0) + \sum_{i=1}^N \sum_{h=1}^r S_*^{\lambda f'\lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{F} - \check{F}_0) \\ & + \sum_{t=1}^T \sum_{h=1}^r S_*^{\lambda\beta' f_{th}}(\hat{f}_{th} - \check{f}_{0t,h})(\hat{\beta} - \beta_0) + \sum_{t=1}^T \sum_{h=1}^r S_*^{\lambda\lambda' f_{th}}(\hat{f}_{th} - \check{f}_{0t,h})(\hat{\Lambda} - \check{\Lambda}_0) + \sum_{t=1}^T \sum_{h=1}^r S_*^{\lambda f' f_{th}}(\hat{f}_{th} - \check{f}_{0t,h})(\hat{F} - \check{F}_0), \end{aligned} \quad (\text{A.12})$$

where the asterisk in the subscript of the functions means that these functions are evaluated at  $(\beta^*, \Lambda^*, F^*)$ .

Define  $\tilde{S}^{\lambda\lambda'} = S^{\lambda\lambda'} - \bar{S}^{\lambda\lambda'}$ ,  $\tilde{S}^{\beta\lambda'} = S^{\beta\lambda'} - \bar{S}^{\beta\lambda'}$ , where  $\bar{S}^{\beta\lambda'} = N^{-1}(\tilde{\Xi}_1, \dots, \tilde{\Xi}_N)$ ,  $\bar{S}^{\lambda\lambda'} = N^{-1} \text{diag}(\tilde{\Omega}_1, \dots, \tilde{\Omega}_N)$ , and  $\tilde{\Xi}_i = \Xi_i \hat{\mathbf{H}}'$ ,  $\tilde{\Omega}_i = \hat{\mathbf{H}} \Omega_i \hat{\mathbf{H}}'$ . Recall that

$$\Xi_i = \frac{1}{T} \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) X_{it}] f'_{0t}, \quad \Omega_i = \frac{1}{T} \sum_{t=1}^T \mathbf{f}_{it}(0) f_{0t} f'_{0t}.$$

Then (A.9) can be written as

$$0 = S^\beta + S^{\beta\beta'}(\hat{\beta} - \beta_0) + \bar{S}^{\beta\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) + \tilde{S}^{\beta\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) + S^{\beta f'}(\hat{F} - \check{F}_0) + 1/2R^\beta(\beta^*, \Lambda^*, F^*), \quad (\text{A.13})$$

and (A.10) can be written as

$$0 = S^\lambda + S^{\lambda\beta'}(\hat{\beta} - \beta_0) + \bar{S}^{\lambda\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) + \tilde{S}^{\lambda\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) + S^{\lambda f'}(\hat{F} - \check{F}_0) + 1/2R^\lambda(\beta^*, \Lambda^*, F^*). \quad (\text{A.14})$$

Plugging (A.14) into (A.13) gives

$$\begin{aligned} & [S^{\beta\beta'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^{\lambda\beta'}](\hat{\beta} - \beta_0) = \\ & - \left[ S^\beta - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^\lambda \right] - \left[ S^{\beta f'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^{\lambda f'} \right] (\hat{F} - \check{F}_0) - \left[ \tilde{S}^{\beta\lambda'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\lambda'} \right] (\hat{\Lambda} - \check{\Lambda}_0) \\ & - 1/2 \left[ R^\beta(\beta^*, \Lambda^*, F^*) - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}R^\lambda(\beta^*, \Lambda^*, F^*) \right]. \end{aligned} \quad (\text{A.15})$$

**Step 2:** The term  $S^{\beta\beta'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^{\lambda\beta'}$  can be written as

$$\bar{S}^{\beta\beta'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\bar{S}^{\lambda\beta'} + \tilde{S}^{\beta\beta'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\beta'},$$

where  $\bar{S}^{\beta\beta'} = (NT)^{-1} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it})X_{it}X'_{it}]$ ,  $\bar{S}^{\lambda\beta'} = (\bar{S}^{\beta\lambda'})'$ . Note that

$$\bar{S}^{\beta\beta'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\bar{S}^{\lambda\beta'} = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T [\mathbb{E}[\mathbf{f}_{it}(0|X_{it})X_{it}X'_{it}] - \Xi_i \Omega_i^{-1} \Xi'_i] = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it}Z'_{it}].$$

Next, we show that  $\tilde{S}^{\beta\beta'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\beta'} = o_P(1)$ . Write

$$\tilde{S}^{\beta\beta'} = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left( l_{it}^{(2)} X_{it} X'_{it} - \mathbb{E}[l_{it}^{(2)} X_{it} X'_{it}] \right) + \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left( \mathbb{E}[l_{it}^{(2)} X_{it} X'_{it}] - \mathbb{E}[\mathbf{f}_{it}(0|X_{it})X_{it}X'_{it}] \right)$$

where the second term on the right-side of the above equation is  $O(h^q) = o(1)$  by Lemma S1, and for the first term, by Assumption 3(iv) we have

$$\mathbb{E} \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left( l_{it}^{(2)} X_{it} X'_{it} - \mathbb{E}[l_{it}^{(2)} X_{it} X'_{it}] \right) \right\|^2 \leq \frac{1}{Th^2} \cdot \frac{1}{N} \sum_{i=1}^N \mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h(l_{it}^{(2)} X_{it} X'_{it} - \mathbb{E}[l_{it}^{(2)} X_{it} X'_{it}]) \right\|^2.$$

By Lemma S1 and Assumption 3(iii),  $\mathbb{E}\|h \cdot l_{it}^{(2)} X_{it} X'_{it}\|^{2+\gamma/2} \leq \mathbb{E}\|X_{it}\|^{4+\gamma} \leq M$ . Thus, it follows from the mixing property (Assumption 2(iv)), the fact that  $Th^2 \rightarrow \infty$  (Assumption 3(vii)) and Theorem 3 of Yoshihara (1978) that the right-hand side of the above inequality is  $o(1)$ . Thus, we have  $\tilde{S}^{\beta\beta'} = o_P(1)$ . Similarly, we can show that  $\bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\beta'} = o_P(1)$ . Therefore, it follows that

$$S^{\beta\beta'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^{\lambda\beta'} = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it}Z'_{it}] + o_P(1) = \Delta + o_P(1). \quad (\text{A.16})$$

**Step 3:**  $S^\beta - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^\lambda$  can be written as

$$-\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(1)} Z_{it} = -\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left( l_{it}^{(1)} Z_{it} - \mathbb{E}[l_{it}^{(1)} Z_{it}] \right) - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[l_{it}^{(1)} Z_{it}].$$

By Lemma S1 and Assumption 3(vii),  $(NT)^{-1} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[l_{it}^{(1)} Z_{it}] = o(h^q) = o(T^{-1/2})$ . Similar to the proof of (A.53) below, the first term on the right-hand side of the above equation is  $O_P(1/\sqrt{NT}) = o_P(T^{-1/2})$ . Thus, we have

$$S^\beta - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^\lambda = o_P(1/\sqrt{T}). \quad (\text{A.17})$$

**Step 4:** By Lemma 8 below

$$[S^{\beta f'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}S^{\lambda f'}](\hat{F} - \check{F}_0) = o_P(1/\sqrt{T}). \quad (\text{A.18})$$

**Step 5:** Now consider the term:  $\left[ \tilde{S}^{\beta\lambda'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\lambda'} \right] (\hat{\Lambda} - \check{\Lambda}_0)$ . Write

$$\tilde{S}^{\beta\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) = \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} X_{it} f'_{0t} - \Xi_i \right) \hat{\mathbf{H}}(\hat{\lambda}_i - \check{\lambda}_{0i}).$$

By the Cauchy-Schwarz inequality, we have

$$\left\| \tilde{S}^{\beta\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) \right\| \lesssim \sqrt{\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} X_{it} f'_{0t} - \Xi_i \right\|^2} \cdot \|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N}.$$

Note that by Lemma S1,

$$\begin{aligned} \mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} X_{it} f'_{0t} - \Xi_i \right\|^2 &= \mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T \left( l_{it}^{(2)} X_{it} - \mathbb{E}[l_{it}^{(2)} X_{it}] \right) f'_{0t} + \frac{1}{T} \sum_{t=1}^T \left( \mathbb{E}[l_{it}^{(2)} X_{it}] - \mathbb{E}[f_{it}(0|X_{it}) X_{it}] \right) f'_{0t} \right\|^2 \\ &\leq \frac{1}{h^2} \mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T h \left( l_{it}^{(2)} X_{it} - \mathbb{E}[l_{it}^{(2)} X_{it}] \right) f'_{0t} \right\|^2 + o(1) \end{aligned}$$

and by the mixing property and Theorem 3 of Yoshihara (1978) the first term on the right-hand side of the above inequality is  $O(1/(Th^2)) = o(1)$ . Thus, it follows that

$$\left\| \tilde{S}^{\beta\lambda'}(\hat{\Lambda} - \check{\Lambda}_0) \right\| = o_P(1) \cdot \|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N}.$$

Similarly, we can show that  $\|\bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\lambda'}(\hat{\Lambda} - \check{\Lambda}_0)\| = o_P(1) \cdot \|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N}$ , and conclude that

$$\left\| \left( \tilde{S}^{\beta\lambda'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\lambda'} \right) (\hat{\Lambda} - \check{\Lambda}_0) \right\| = o_P(1) \cdot \|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N}. \quad (\text{A.19})$$

**Step 6:** We will show that:

$$R^\beta(\beta^*, \Lambda^*, F^*) = o_P(\|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N}) + o_P(\|\hat{\beta} - \beta_0\|) + o_P(1/\sqrt{T}), \quad (\text{A.20})$$

$$\bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}R^\lambda(\beta^*, \Lambda^*, F^*) = o_P(\|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N}) + o_P(\|\hat{\beta} - \beta_0\|) + o_P(1/\sqrt{T}). \quad (\text{A.21})$$

To save space, we focus on (A.20), which follows from Results 1 to 9 below. The proof of (A.21) is similar.

**Result 1:**  $S_*^{\beta\beta'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\beta} - \beta_0) = o_P(\|\hat{\beta} - \beta_0\|)$ .

Observe that:

$$S_*^{\beta\beta'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\beta} - \beta_0) = -\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X'_{it} X_{it,k} (\hat{\beta}_k - \beta_{0k})(\hat{\beta} - \beta_0),$$

so

$$\|S_*^{\beta\beta'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\beta} - \beta_0)\| \leq \|\hat{\beta} - \beta_0\|^2 \cdot \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X'_{it} X_{it,k} \right\|.$$

Expanding  $l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*)$  around  $(\beta^*, \lambda_i^*, \tilde{f}_{0t})$  gives

$$\begin{aligned} & \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X'_{it} X_{it,k} \\ &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(4)}(\beta^*, \lambda_i^*, f_t^{**}) X_{it} X'_{it} X_{it,k} (\lambda_i^*)' (f_t^* - \tilde{f}_{0t}) + \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k}]|_{\beta=\beta^*, \lambda_i=\lambda_i^*} \\ &+ \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left[ l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k} - \mathbb{E}[l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k}]|_{\beta=\beta^*, \lambda_i=\lambda_i^*} \right], \end{aligned} \quad (\text{A.22})$$

where  $f_t^{**}$  is between  $f_t^*$  and  $\tilde{f}_{0t}$ . By Lemma S1, the second term on the right-hand side of (A.22) is  $(NT)^{-1} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[f_{it}^{(1)}(\cdot | X_{it}) X_{it} X'_{it} X_{it,k}] + \bar{O}(h^{q-1}) = O(1)$ , and the first term is bounded by

$$\sqrt{\frac{1}{T} \sum_{t=1}^T \|f_t^* - \tilde{f}_{0t}\|^2 \cdot \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T [l_{it}^{(4)}(\beta^*, \lambda_i^*, f_t^{**})]^2 \cdot \|X_{it}\|^6 \cdot \|\lambda_i^*\|^2} = O_P\left(1/\sqrt{Th^6}\right) = o_P(1),$$

since we have  $|l_{it}^{(4)}(\cdot)| \lesssim 1/h^3$  by Lemma S1,  $\mathbb{E}\|X_{it}\|^6 < M$  by Assumption 3(iii), and  $T^{-1} \sum_{t=1}^T \|f_t^* - \tilde{f}_{0t}\|^2 \leq T^{-1} \sum_{t=1}^T \|\hat{f}_t - \tilde{f}_{0t}\|^2 = O_P(N^{-1}) = O_P(T^{-1})$  by Lemma 2 and Assumption 3(vii). Finally, with probability approaching 1, the last term on the right-hand side of (A.22) is bounded by  $N^{-1} \sum_{i=1}^N \mathcal{Z}_i$ , where

$$\mathcal{Z}_i = \sup_{(\beta, \lambda_i) \in B_{\delta, i}} \left\| \frac{1}{T} \sum_{t=1}^T \left[ l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k} - \mathbb{E}[l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k}] \right] \right\|$$

and  $B_{\delta, i}$  is a neighbourhood of  $(\beta_0, \tilde{\lambda}_{0i})$ . Then Result 1 follows if we can show that  $\max_{1 \leq i \leq N} \mathbb{E}[\mathcal{Z}_i] < \infty$ .

For any  $\epsilon > 0$ , let  $\theta_i^{(1)}, \dots, \theta_i^{(L)}$  be a maximal set of points in  $B_{\delta, i}$  such that  $\|\theta_i^{(j)} - \theta_i^{(l)}\| \geq \epsilon$  for any  $j \neq l$ . It is well known that  $L$ , the packing number of a Euclidean ball, is bounded (up to a positive

constant that only depends on  $p + r$ ) by  $(1/\epsilon)^{p+r}$ . Thus, we have

$$\mathbb{E}\mathcal{Z}_i \leq \frac{1}{\sqrt{Th^4}} \cdot \mathbb{E} \left[ \max_{1 \leq j \leq L} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h^2 \left[ l_{it}^{(3)}(\beta^{(j)}, \lambda_i^{(j)}, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k} - \mathbb{E}[l_{it}^{(3)}(\beta^{(j)}, \lambda_i^{(j)}, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k}] \right] \right\| \right] + O(\epsilon/h^3),$$

where we have used the fact that  $|l_{it}^{(3)}(\beta^a, \lambda_i^a, \tilde{f}_{0t}) - l_{it}^{(3)}(\beta^b, \lambda_i^b, \tilde{f}_{0t})| \lesssim (\|\beta^a - \beta^b\| + \|\lambda_i^a - \lambda_i^b\|)/h^3$ .

By Assumption 3 (iii) and (vii),  $2m > 12(p+r) > 3(p+r)/2c$ . Choose  $2m/3 > L > (p+r)/2c$ , then  $\mathbb{E}\|h^2 l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X'_{it} X_{it,k}\|^{L+\gamma/3} \lesssim \mathbb{E}\|X_{it}\|^{3L+\gamma} \leq \mathbb{E}\|X_{it}\|^{2m+\gamma} < M$ . Thus, by Lemma 2.2.2 of van der Vaart and Wellner (1996) and Theorem 3 of Yoshihara (1978), the first term on right-hand side of the above inequality is bounded (up to a positive constant) by  $L^{1/J}/\sqrt{Th^4}$ . Choosing  $\epsilon = 1/\sqrt{T}$ , we have

$$\mathbb{E}\mathcal{Z}_i \leq C \left( \frac{T^{\frac{p+r}{2J}}}{\sqrt{Th^4}} + \frac{1}{\sqrt{Th^6}} \right)$$

for some positive constant  $C$ . Since  $J > (p+r)/(2c)$ ,  $T^{\frac{p+r}{2J}}/\sqrt{Th^4} = o(\sqrt{Th^6})$ . Then from Assumption 3(vii) we have  $\max_{1 \leq i \leq N} \mathbb{E}[\mathcal{Z}_i] < \infty$  and the desired result follows.

**Result 2:**  $S_*^{\beta\lambda'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\Lambda} - \check{\Lambda}_0) = o_P(\|\hat{\beta} - \beta_0\|)$ .

We have

$$\begin{aligned} \|S_*^{\beta\lambda'\beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{\Lambda} - \check{\Lambda}_0)\| &= \left\| \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X_{it,k} f_t^{*\prime} \right) (\hat{\lambda}_i - \check{\lambda}_{0i})(\hat{\beta} - \beta_0) \right\| \\ &\leq \|\hat{\beta} - \beta_0\| \cdot \max_{i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \cdot \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X_{it,k} f_t^{*\prime} \right\|. \end{aligned}$$

Note that

$$\begin{aligned} \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X_{it,k} f_t^{*\prime} &= \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) X_{it} X_{it,k} \tilde{f}'_{0t} + \\ &+ \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^{**}) X_{it} X_{it,k} (f_t^* - \tilde{f}_{0t})' + \frac{1}{T} \sum_{t=1}^T l_{it}^{(4)}(\beta^*, \lambda_i^*, f_t^{**}) X_{it} X_{it,k} \tilde{f}'_{0t} (\lambda_i^*)' (f_t^* - \tilde{f}_{0t}), \end{aligned}$$

where  $f_t^{**}$  is between  $f_t^*$  and  $\tilde{f}_{0t}$ . Using Lemma S1, Lemma 2, Assumption 3(iii) and the Cauchy-Schwarz inequality, we can show that the last two terms on the right-hand side of the above inequality is  $\bar{O}_P(1/\sqrt{Nh^6}) = \bar{O}_P(1/\sqrt{Th^6}) = \bar{o}_P(1)$ . For the first term on the right-hand side of the above inequality, with probability approaching 1, we have

$$\begin{aligned} \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) X_{it} X_{it,k} \tilde{f}'_{0t} \right\| &\leq \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left\| \mathbb{E} \left[ l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X_{it,k} \tilde{f}'_{0t} \right] |_{\beta=\beta^*, \lambda_i=\lambda_i^*} \right\| \\ &+ \frac{1}{N} \sum_{i=1}^N \sup_{\theta_i \in \tilde{B}_{\delta,i}} \left\| \frac{1}{T} \sum_{t=1}^T \left( l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X_{it,k} \tilde{f}'_{0t} - \mathbb{E} \left[ l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} X_{it,k} \tilde{f}'_{0t} \right] \right) \right\| \end{aligned}$$

The first term on the right-hand side of the above inequality is  $O(1)$  by Lemma S1. Similar to the proof of Result 1, The second term on the right-hand side of the above inequality can be shown to be  $o_P(1)$ . Thus, we have

$$\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X_{it,k} f_t^{*'} \right\| = O_P(1),$$

and the result follows from uniform consistency of  $\hat{\lambda}_i$ .

**Result 3:**  $S_*^{\beta f' \beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{F} - \check{F}_0) = o_P(\|\hat{\beta} - \beta_0\|)$ .

Note that by Lemma 2,

$$\begin{aligned} & \|S_*^{\beta f' \beta_k}(\hat{\beta}_k - \beta_{0k})(\hat{F} - \check{F}_0)\| \\ & \leq \|\hat{\beta} - \beta_0\| \cdot \left\| \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X_{it,k} \lambda_i^{*'} \right) (\hat{f}_t - \check{f}_{0t}) \right\| \\ & \leq \|\hat{\beta} - \beta_0\| \cdot \sqrt{\frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^2 \cdot \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T [l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*)]^2 \cdot \|X_{it}\|^4 \cdot \|\lambda_i^*\|^2} \\ & = \|\hat{\beta} - \beta_0\| \cdot O_P\left(1/\sqrt{Nh^4}\right) = o_P(\|\hat{\beta} - \beta_0\|), \end{aligned}$$

because by Lemma S1,  $h^2 l_{it}^{(3)}(\beta, \lambda_i, f_t)$  is uniformly bounded, and  $Nh^4 \rightarrow \infty$  by Assumption 3(vii).

**Result 4:**  $\sum_{i=1}^N \sum_{h=1}^r S_*^{\beta \beta' \lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{\beta} - \beta_0) = o_P(\|\hat{\beta} - \beta_0\|)$ .

Observe that for each  $h \leq r$ ,

$$\sum_{i=1}^N S_*^{\beta \beta' \lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{\beta} - \beta_0) = -\frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X'_{it} f_{th}^* \right) (\hat{\beta} - \beta_0),$$

so

$$\left\| \sum_{i=1}^N S_*^{\beta \beta' \lambda_{ih}}(\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{\beta} - \beta_0) \right\| \leq \|\hat{\beta} - \beta_0\| \cdot \max_{i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \cdot \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X'_{it} f_{th}^* \right\|,$$

which can be shown to be  $o_P(\|\hat{\beta} - \beta_0\|)$ , similar to the proof of Result 2.

**Result 5:**  $\sum_{t=1}^T \sum_{h=1}^r S_*^{\beta \beta' f_{th}}(\hat{f}_{th} - \check{f}_{0t,h})(\hat{\beta} - \beta_0) = o_P(\|\hat{\beta} - \beta_0\|)$ .

The proof is similar to Result 3. For each  $h \leq r$ , we have

$$\sum_{t=1}^T S_*^{\beta\beta' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h}) = -\frac{1}{T} \sum_{t=1}^T (\hat{f}_{th} - \check{f}_{0t,h}) \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} X_{it}' \lambda_{ih}^* \right),$$

so

$$\left\| \sum_{t=1}^T S_*^{\beta\beta' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h}) \right\| \leq \sqrt{\frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^2 \cdot \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T [l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*)]^2 \cdot \|X_{it}\|^4 \cdot \|\lambda_i^*\|^2},$$

which is  $O_P(1/\sqrt{Th^4})$ . So Result 5 follows.

**Result 6:**  $\sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih})(\hat{F} - \check{F}_0) = o_P(\|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N})$ .

Write:

$$\sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih})(\hat{F} - \check{F}_0) = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T [l_{it}^{(2)}(\beta^*, \lambda_i^*, f_t^*) X_{it} - l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (\lambda_i^*)' f_t^*] (\hat{\lambda}_i - \check{\lambda}_{0i})' (\hat{f}_t - \check{f}_{0t}).$$

Thus, by the Cauchy-Schwarz inequality, Lemma 2 and Lemma S1

$$\begin{aligned} & \left\| \sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih})(\hat{F} - \check{F}_0) \right\| \leq \|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N} \cdot \\ & \sqrt{\frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^2 \cdot \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \left\| [l_{it}^{(2)}(\beta^*, \lambda_i^*, f_t^*) X_{it} - l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (\lambda_i^*)' f_t^*] \right\|^2} \\ & = \|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N} \cdot O_P(1/\sqrt{Nh^4}), \end{aligned}$$

so the result follows by Assumption 3(vii).

**Result 7:**  $\sum_{t=T}^N \sum_{h=1}^r S_*^{\beta \lambda' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h})(\hat{\Lambda} - \check{\Lambda}_0) = o_P(\|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N})$ .

The proof is similar to the proof of Result 6.

**Result 8:**  $\sum_{i=1}^N \sum_{h=1}^r S_*^{\beta \lambda' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih})(\hat{\Lambda} - \check{\Lambda}_0) = o_P(\|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N})$ .

Note that for each  $h \leq r$ , we have

$$\sum_{i=1}^N S_*^{\beta \lambda' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih})(\hat{\Lambda} - \check{\Lambda}_0) = -\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (f_t^*)' f_{th}^* (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih})(\hat{\lambda}_i - \check{\lambda}_{0i}),$$

by Lemma 3

$$\begin{aligned} \left\| \sum_{i=1}^N S_*^{\beta \lambda' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{\Lambda} - \check{\Lambda}_0) \right\| &\leq \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \cdot \frac{1}{N} \sum_{i=1}^N \left( \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \cdot \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} f_t^* f_{th}^* \right\| \right) \\ &\leq o_P(1) \cdot \|\hat{\Lambda} - \check{\Lambda}_0\| / \sqrt{N} \cdot \sqrt{\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (f_t^*)' f_{th}^* \right\|^2}. \end{aligned}$$

Thus, it remains to show that

$$\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (f_t^*)' f_{th}^* \right\|^2 = O_P(1). \quad (\text{A.23})$$

First, write

$$\begin{aligned} l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (f_t^*)' f_{th}^* &= l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (\tilde{f}_{0t})' \tilde{f}_{0,th} + l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} ((f_t^*)' f_{th}^* - (\tilde{f}_{0t})' \tilde{f}_{0,th}) \\ &= l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) X_{it} (\tilde{f}_{0t})' \tilde{f}_{0,th} - l_{it}^{(4)}(\beta^*, \lambda_i^*, f_t^{**}) (\lambda_i^*)' (f_t^* - \tilde{f}_{0t}) X_{it} (\tilde{f}_{0t})' \tilde{f}_{0,th} \\ &\quad + l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} ((f_t^*)' f_{th}^* - (\tilde{f}_{0t})' \tilde{f}_{0,th}), \end{aligned}$$

thus, by the Cauchy-Schwarz inequality,

$$\begin{aligned} \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (f_t^*)' f_{th}^* \right\|^2 &\leq \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) X_{it} (\tilde{f}_{0t})' \tilde{f}_{0,th} \right\|^2 \\ &\quad + \frac{1}{T} \sum_{t=1}^T \| (f_t^*)' f_{th}^* - (\tilde{f}_{0t})' \tilde{f}_{0,th} \|^2 \cdot \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \| l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} \|^2 \\ &\quad + \frac{1}{T} \sum_{t=1}^T \| f_t^* - \tilde{f}_{0t} \|^2 \cdot \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \| l_{it}^{(4)}(\beta^*, \lambda_i^*, f_t^{**}) (\lambda_i^*)' X_{it} (\tilde{f}_{0t})' \tilde{f}_{0,th} \|^2. \end{aligned}$$

The last two terms on the right-hand side of the above inequality are both  $o_P(1)$  by Lemma 2 and Lemma S1. For the first term on the right-hand side of the above inequality, by Assumption 3(iii), with probability approaching 1,

$$\begin{aligned} \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) X_{it} (\tilde{f}_{0t})' \tilde{f}_{0,th} \right\|^2 &\leq \max_{i,t} \left\| \mathbb{E}[g_{it}(\beta, \lambda_i) | \beta = \beta^*, \lambda_i = \lambda_i^*] \right\|^2 + \\ &\quad \frac{1}{N} \sum_{i=1}^N \sup_{\theta_i \in B_{\delta,i}} \left\| \frac{1}{T} \sum_{t=1}^T (g_{it}(\beta, \lambda_i) - \mathbb{E}[g_{it}(\beta, \lambda_i)]) \right\|^2, \end{aligned}$$

where  $g_{it}(\beta, \lambda_i) = l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) X_{it} (\tilde{f}_{0t})' \tilde{f}_{0,th}$ . The first term on the right-hand side of the above inequality is  $O(1)$  by Lemma S1. Thus, to prove (A.23), it suffices to show that

$$\max_{1 \leq i \leq N} \mathbb{E} \left[ \sup_{\theta_i \in B_{\delta,i}} \left\| \frac{1}{T} \sum_{t=1}^T (g_{it}(\beta, \lambda_i) - \mathbb{E}[g_{it}(\beta, \lambda_i)]) \right\|^2 \right] = O(1).$$

Similar to the proof of Result 1, for any  $\epsilon > 0$ , let  $\theta_i^{(1)}, \dots, \theta_i^{(L)}$  be a maximal set of points in  $B_{\delta,i}$  such that  $\|\theta_i^{(j)} - \theta_i^{(l)}\| \geq \epsilon$  for any  $j \neq l$ . Thus, for some constants  $C_1, C_2 > 0$ , we have

$$\begin{aligned} & \mathbb{E} \left[ \sup_{\theta_i \in B_{\delta,i}} \left\| \frac{1}{T} \sum_{t=1}^T (g_{it}(\beta, \lambda_i) - \mathbb{E}[g_{it}(\beta, \lambda_i)]) \right\|^2 \right] \\ & \leq C_1 \frac{1}{Th^4} \mathbb{E} \left[ \max_{1 \leq j \leq L} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h^2 (g_{it}(\beta^{(j)}, \lambda_i^{(j)}) - \mathbb{E}[g_{it}(\beta^{(j)}, \lambda_i^{(j)})]) \right\|^2 \right] + C_2 \epsilon^2 / h^6. \end{aligned}$$

By Assumption 3 (iii) and (vii),  $m > (p+r)/(4c)$ . Choose  $J$  such that  $m > J > (p+r)/(4c)$ . Since  $\mathbb{E}\|h^2 g_{it}(\beta, \lambda_i)\|^{2J+\gamma} \lesssim \mathbb{E}\|X_{it}\|^{2J+\gamma} \leq \mathbb{E}\|X_{it}\|^{2m+\gamma} < M$ . It follows that the right-hand side of the above inequality is

$$O\left(\frac{L^{1/J}}{Th^4} + \epsilon^2/h^6\right) = O\left(\frac{T^{(p+r)/(2J)}}{Th^4}\right) + O\left(\frac{1}{Th^6}\right) = o\left(\frac{T^{2c}}{Th^4}\right) + o(1) = o(1).$$

This completes the proof of Result 8.

**Result 9:**  $\sum_{t=1}^T \sum_{h=1}^r S_*^{\beta f' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h})(\hat{F} - \check{F}_0) = o_P(1/\sqrt{T})$ .

For each  $h \leq r$ , we have

$$\sum_{t=1}^T S_*^{\beta f' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h})(\hat{F} - \check{F}_0) = -\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (\lambda_i^*)' \lambda_{ih}^* (\hat{f}_{th} - \check{f}_{0t,h})(\hat{f}_t - \check{f}_{0t}),$$

so by Assumption 3(iii), Lemma 2 and Lemma S1,

$$\begin{aligned} \left\| \sum_{t=1}^T S_*^{\beta f' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h})(\hat{F} - \check{F}_0) \right\| & \leq \frac{1}{T} \sum_{t=1}^T \left( \|\hat{f}_t - \check{f}_{0t}\|^2 \cdot \left\| \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (\lambda_i^*)' \lambda_{ih}^* \right\| \right) \\ & \lesssim O_P\left(\frac{1}{h^2}\right) \cdot \frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\|^2 = O_P\left(1/Nh^2\right). \end{aligned}$$

Then the result follows since  $O_P(1/Nh^2) = O_P\left(\frac{1}{\sqrt{N} \cdot \sqrt{Nh^2}}\right)$  and Assumption 3(vii) implies that  $\sqrt{Nh^2} \rightarrow \infty$ .

**Step 7:** It follows from (A.15) to (A.21) that:

$$\Delta(\hat{\beta} - \beta_0) = o_P(\|\hat{\beta} - \beta_0\|) + o_P(\|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N}) + o_P(1/\sqrt{T}),$$

then the desired result follows from the assumption that  $\Delta$  is positive definite and the fact that  $\|\hat{\Lambda} - \check{\Lambda}_0\|/\sqrt{N} \leq \max_i \|\hat{\lambda}_i - \check{\lambda}_{0i}\|$ .  $\square$

**Lemma 5.** Under Assumptions 1 to 4, there exists  $0 < \nu < 1/6 - c$  (where  $c$  is defined in Assumption 3(vii)) such that  $\max_{1 \leq i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| = O_P(1/T^{1/2-1/(2m)})$ .

*Proof.* Expanding the first order condition:  $T^{-1} \sum_{t=1}^T l_{it}^{(1)}(\hat{\beta}, \hat{\lambda}_i, \hat{f}_t) \hat{f}_t = 0$  gives:

$$\begin{aligned} \tilde{\Omega}_i(\hat{\lambda}_i - \check{\lambda}_{0i}) &= \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} \check{f}_{0t} + \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \hat{f}_t [X'_{it}(\hat{\beta} - \beta_0)] - \left( \tilde{\Omega}_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \check{f}_{0t} \check{f}'_{0t} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \\ &- \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} (\hat{f}_t - \check{f}_{0t}) \check{f}'_{0t} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) - \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \hat{f}_t (\hat{f}_t - \check{f}_{0t})' \right) \check{\lambda}_{0i} + 0.5 \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [X'_{it}(\hat{\beta} - \beta_0)]^2 \\ &+ \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [X'_{it}(\hat{\beta} - \beta_0)] \cdot [(f_t^*)'(\hat{\lambda}_i - \check{\lambda}_{0i})] + \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [X'_{it}(\hat{\beta} - \beta_0)] \cdot [(\lambda_i^*)'(\hat{f}_t - \check{f}_{0t})] \\ &+ \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [(\lambda_i^*)'(\hat{f}_t - \check{f}_{0t})] \cdot [(f_t^*)'(\hat{\lambda}_i - \check{\lambda}_{0i})] + 0.5 \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [(\lambda_i^*)'(\hat{f}_t - \check{f}_{0t})]^2 + 0.5 \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [(f_t^*)'(\hat{\lambda}_i - \check{\lambda}_{0i})]^2. \end{aligned}$$

**Step 1:** Let  $M$  be a generic bounded constant. By Lemma S1,

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} \check{f}_{0t} \right\| \leq \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T \tilde{l}_{it}^{(1)} \check{f}_{0t} \right\| \cdot \|\hat{\mathbf{H}}\| + O(h^q).$$

Since,  $\{l_{it}^{(1)} \check{f}_{0t}\}$  is uniformly bounded, by the mixing property of  $u_{it}$  and Theorem 3 of Yoshihara (1978), for any  $J \geq 2$  we have  $\mathbb{E} \left\| T^{-1/2} \sum_{t=1}^T \tilde{l}_{it}^{(1)} \check{f}_{0t} \right\|^J < M$  and it follows from Lemma 2.2.2 of van der Vaart and Wellner (1996) that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T \tilde{l}_{it}^{(1)} \check{f}_{0t} \right\| = O_P(N^{1/J}/\sqrt{T}) = O_P(1/T^{1/2-1/J}).$$

Choosing  $J = 2m$ , we have

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} \check{f}_{0t} \right\| = O_P(T^{1/(2m)}/T^{1/2}) \tag{A.24}$$

since  $O(h^q) = o(T^{-1})$  by Assumption 3(vii).

**Step 2:** By Lemma S1,  $l_{it}^{(1)}$  is uniformly bounded, so it follows from Lemma 2 that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) \right\| \leq O_P(1) \cdot \frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\| = O_P(N^{-1/2}) = O_P(T^{-1/2}). \tag{A.25}$$

**Step 3:** Note that:

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \hat{f}_t [X'_{it}(\hat{\beta} - \beta_0)] \right\| &\leq \|\hat{\beta} - \beta_0\| \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \hat{f}_t X'_{it} \right\| \\ &\leq \|\hat{\beta} - \beta_0\| \cdot \|\hat{\mathbf{H}}\| \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} X'_{it} \right\| + \|\hat{\beta} - \beta_0\| \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} (\hat{f}_t - \check{f}_{0t}) X'_{it} \right\|. \end{aligned}$$

First, by Assumption 3(iii), Lemma 2 and Lemma S1, we have

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} (\hat{f}_t - \check{f}_{0t}) X'_{it} \right\| \leq O_P(h^{-1}) \cdot \frac{1}{T} \sum_{t=1}^T \|\hat{f}_t - \check{f}_{0t}\| = O_P(1/(\sqrt{N}h)) = o_P(1).$$

Second, by Lemma S1,

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} X'_{it} \right\| \leq \frac{1}{\sqrt{Th^2}} \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h \left( l_{it}^{(2)} f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(2)} f_{0t} X'_{it}] \right) \right\| + O(1).$$

Since  $\|h l_{it}^{(2)} f_{0t} X'_{it}\|$  is uniformly bounded by Assumption 3(iii) and Lemma S1, it follows from Theorem 3 of [Yoshihara \(1978\)](#) that

$$\mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h \left( l_{it}^{(2)} f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(2)} f_{0t} X'_{it}] \right) \right\|^4 < M,$$

and then it follows from Lemma 2.2.2 of [van der Vaart and Wellner \(1996\)](#) that

$$\frac{1}{\sqrt{Th^2}} \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h \left( l_{it}^{(2)} f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(2)} f_{0t} X'_{it}] \right) \right\| = O_P(N^{1/4}/\sqrt{Th^2}) = o_P(1).$$

Thus, it can be concluded that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \hat{f}_t [X'_{it}(\hat{\beta} - \beta_0)] \right\| = O_P(\|\hat{\beta} - \beta_0\|). \quad (\text{A.26})$$

**Step 4:** By the definition of  $\tilde{\Omega}_i$ , we have

$$\max_{1 \leq i \leq N} \left\| \left( \tilde{\Omega}_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \check{f}_{0t} \check{f}'_{0t} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| \leq \|\hat{\mathbf{H}}\|^2 \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T (l_{it}^{(2)} - \mathbf{f}_{it}(0)) f_{0t} f'_{0t} \right\| \cdot \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\|.$$

Similar to the proof of Step 3, we can show that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T (l_{it}^{(2)} - \mathbf{f}_{it}(0)) f_{0t} f'_{0t} \right\| = o_P(1),$$

thus it follows that

$$\max_{1 \leq i \leq N} \left\| \left( \tilde{\Omega}_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \check{f}_{0t} \check{f}'_{0t} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| = o_P \left( \max_i \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \right). \quad (\text{A.27})$$

**Step 5:** By Lemma 2, Lemma S1 and Assumption 3(vii) we have

$$\max_{1 \leq i \leq N} \left\| \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} (\hat{f}_t - \check{f}_{0t}) \check{f}'_{0t} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| \leq \|\hat{\mathbf{H}}\| \cdot O_P(1/\sqrt{Nh^2}) \cdot \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| = o_P \left( \max_i \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \right). \quad (\text{A.28})$$

**Step 6:** Note that

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \hat{f}_t (\hat{f}_t - \check{f}_{0t})' \right) \check{\lambda}_{0i} \right\| &\leq O_P(1) \cdot \|\hat{\mathbf{H}}\| \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} (\hat{f}_t - \check{f}_{0t})' \right\| \\ &\quad + O_P(1) \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} (\hat{f}_t - \check{f}_{0t}) (\hat{f}_t - \check{f}_{0t})' \right\|. \end{aligned}$$

The second term on the right-hand side of the above inequality is  $O_P(1/(Th))$  by Lemma S1 and Lemma 2. Next, we have

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} (\hat{f}_t - \check{f}_{0t})' \right\| \lesssim \max_{1 \leq t \leq T} \|\hat{f}_t - \check{f}_{0t}\| \cdot \max_{1 \leq i \leq N} \frac{1}{T} \sum_{t=1}^T \left| l_{it}^{(2)} \right|,$$

and

$$\begin{aligned} \max_{1 \leq i \leq N} \frac{1}{T} \sum_{t=1}^T \left| l_{it}^{(2)} \right| &\lesssim 1/h \cdot \max_{1 \leq i \leq N} \frac{1}{T} \sum_{t=1}^T \mathbf{1}\{|u_{it}| \leq h\} \\ &\leq \max_{i,t} P[|u_{it}| \leq h]/h + h^{-1} \cdot \max_{i \leq N} \left| \frac{1}{T} \sum_{t=1}^T [\mathbf{1}\{|u_{it}| \leq h\} - P(|u_{it}| \leq h)] \right|. \end{aligned}$$

It is easy to see that  $\max_{i,t} P[|u_{it}| \leq h] = O(h)$ . Moreover, similar to the proof of Step 3, we can show that

$$h^{-1} \cdot \max_{i \leq N} \left| \frac{1}{T} \sum_{t=1}^T [\mathbf{1}\{|u_{it}| \leq h\} - P(|u_{it}| \leq h)] \right| = o_P(1).$$

Therefore,

$$\max_{1 \leq i \leq N} \frac{1}{T} \sum_{t=1}^T \left| l_{it}^{(2)} \right| = O_P(1),$$

and by Lemma 2

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} (\hat{f}_t - \check{f}_{0t})' \right\| = O_P(T^{1/(2m)}/\sqrt{N}).$$

Thus, we can conclude that

$$\max_{1 \leq i \leq N} \left\| \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \hat{f}_t (\hat{f}_t - \check{f}_{0t})' \right) \check{\lambda}_{0i} \right\| = O_P(T^{1/(2m)} / \sqrt{N}) + O_P(1/(Th)) = O_P(T^{1/(2m)} / T^{1/2}). \quad (\text{A.29})$$

**Step 7:** By the consistency of  $\hat{\beta}$ ,

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [X'_{it}(\hat{\beta} - \beta_0)]^2 \right\| &\leq \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t X'_{it} \right\| \cdot \|\hat{\beta} - \beta_0\|^2 \\ &= o_P(\|\hat{\beta} - \beta_0\|) \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t X'_{it} \right\|. \end{aligned}$$

Write

$$\frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t X'_{it} = \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) \tilde{f}_{0t} X'_{it} + \frac{1}{T} \sum_{t=1}^T l_{it}^{(4)}(\beta^*, \lambda_i^*, f_t^{**}) \tilde{f}_{0t} X'_{it} \cdot [(f_t^* - \tilde{f}_{0t}) \lambda_i^*] + \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) (\hat{f}_t - \tilde{f}_{0t}) X'_{it}.$$

It then follows from Lemma 2 and Lemma S1 that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t X'_{it} \right\| \lesssim \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) f_{0t} X'_{it} \right\| + O_P(1/\sqrt{Th^6}).$$

Let  $B_{\delta,i}$  be a neighborhood of  $(\beta_0, \tilde{\lambda}_{0i})$ , then by the uniform consistency of  $\hat{\lambda}_i$ , with probability approaching 1, we have

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) f_{0t} X'_{it} \right\| &\leq \max_{i,t} \sup_{\theta_i \in B_{\delta,i}} \left\| \mathbb{E}[l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it}] \right\| \\ &\quad + \max_{1 \leq i \leq N} \sup_{\theta_i \in B_{\delta,i}} \left\| \frac{1}{T} \sum_{t=1}^T \left( l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it}] \right) \right\|. \end{aligned}$$

The first term on the right-hand side of the above inequality is  $O(1)$  by Lemma S1. Next, for each  $i$ , let  $\theta_i^{(1)}, \dots, \theta_i^{(L_i)}$  be a maximal set of points in  $B_{\delta,i}$  such that  $\max_i \|\theta_i^{(j)} - \theta_i^{(k)}\| \leq \epsilon$  for some small  $\epsilon > 0$ . It follows that

$$\begin{aligned} \max_{1 \leq i \leq N} \sup_{\theta_i \in B_{\delta,i}} \left\| \frac{1}{T} \sum_{t=1}^T \left( l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it}] \right) \right\| &\leq \\ \frac{1}{\sqrt{Th^2}} \cdot \max_{1 \leq i \leq N} \max_{1 \leq j \leq L_i} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h^2 \left( l_{it}^{(3)}(\beta^{(j)}, \lambda_i^{(j)}, \tilde{f}_{0t}) f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(3)}(\beta^{(j)}, \lambda_i^{(j)}, \tilde{f}_{0t}) f_{0t} X'_{it}] \right) \right\| &+ O(\epsilon/h^3). \end{aligned}$$

Note  $\max_i L_i \lesssim \bar{L} = (1/\epsilon)^{p+r}$ . By Assumption 3(iii),  $2m > 3(p+r)/(2c) > (p+r+2)/(2c)$ . Choose  $2m > L > (p+r+2)/(2c)$ . Then  $\mathbb{E}\|h^2 l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it}\|^{J+\gamma} \lesssim \mathbb{E}\|X'_{it}\|^{J+\gamma} \leq \mathbb{E}\|X_{it}\|^{2m+\gamma} < M$ . By

Theorem 3 of Yoshihara (1978),

$$\mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h^2 \left( l_{it}^{(3)}(\beta^{(j)}, \lambda_i^{(j)}, \tilde{f}_{0t}) f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(3)}(\beta^{(j)}, \lambda_i^{(j)}, \tilde{f}_{0t}) f_{0t} X'_{it}] \right) \right\|^J < M.$$

Thus, by Lemma 2.2.2 of van der Vaart and Wellner (1996), we have

$$\max_{1 \leq i \leq N} \sup_{\theta_i \in B_{\delta,i}} \left\| \frac{1}{T} \sum_{t=1}^T \left( l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it} - \mathbb{E}[l_{it}^{(3)}(\beta, \lambda_i, \tilde{f}_{0t}) f_{0t} X'_{it}] \right) \right\| = O_P \left( \frac{(N\bar{L})^{1/J}}{\sqrt{Th^4}} \right) + O(\epsilon/h^3).$$

Choosing  $\epsilon = 1/\sqrt{T}$ , we have

$$O_P \left( \frac{(N\bar{L})^{1/J}}{\sqrt{Th^4}} \right) + O(\epsilon/h^3) = O_P(1/\sqrt{Th^6}) = o_P(1).$$

Therefore, we can conclude that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(\beta^*, \lambda_i^*, \tilde{f}_{0t}) f_{0t} X'_{it} \right\| = O_P(1)$$

and that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [X'_{it}(\hat{\beta} - \beta_0)]^2 \right\| = o_P(\|\hat{\beta} - \beta_0\|). \quad (\text{A.30})$$

**Step 8:** Similar to the proof of Step 7, we can show that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [X'_{it}(\hat{\beta} - \beta_0)] \cdot [(f_t^*)'(\hat{\lambda}_i - \check{\lambda}_{0i})] \right\| = o_P(\|\hat{\beta} - \beta_0\|), \quad (\text{A.31})$$

and

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [(f_t^*)'(\hat{\lambda}_i - \check{\lambda}_{0i})]^2 \right\| = O_P \left( \max_i \|\hat{\lambda}_i - \lambda_{0i}\|^2 \right). \quad (\text{A.32})$$

**Step 9:** From Lemma 2 and Lemma S1 it follows easily that:

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [X'_{it}(\hat{\beta} - \beta_0)] \cdot [(\lambda_i^*)'(\hat{f}_t - \check{f}_{0t})] \right\| = o_P(\|\hat{\beta} - \beta_0\|), \quad (\text{A.33})$$

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [(\lambda_i^*)'(\hat{f}_t - \check{f}_{0t})] \cdot [(f_t^*)'(\hat{\lambda}_i - \check{\lambda}_{0i})] \right\| = O_P(1/\sqrt{Th^4}) \cdot O_P \left( \max_i \|\hat{\lambda}_i - \lambda_{0i}\| \right), \quad (\text{A.34})$$

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \hat{f}_t \cdot [(\lambda_i^*)'(\hat{f}_t - \check{f}_{0t})]^2 \right\| = O_P(1/(Th^2)). \quad (\text{A.35})$$

Finally, since  $\max_i \|\tilde{\Omega}_i^{-1}\| \leq \|\hat{\mathbf{H}}^{-1}\|^2 \cdot \max_i \|\Omega_i^{-1}\| = O_P(1)$  by Assumption 3 (ii), it follows from

(A.24) to (A.35) that:

$$\max_{1 \leq i \leq N} \|\hat{\lambda}_i - \lambda_{0i}\| = O_P(\|\hat{\beta} - \beta_0\|) + o_P\left(\max_i \|\hat{\lambda}_i - \lambda_{0i}\|\right) + O_P(T^{1/(2m)}/T^{1/2}), \quad (\text{A.36})$$

then the desired result follows from (A.36) and Lemma 4.  $\square$

**Lemma 6.** Under Assumptions 1 to 4,

$$\begin{aligned} \tilde{\Omega}_i(\hat{\lambda}_i - \check{\lambda}_{0i}) &= \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} \check{f}_{0t} + \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) - \left( \tilde{\Omega}_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \check{f}_{0t} \check{f}'_{0t} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \\ &\quad - \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \check{f}_{0t} (\hat{f}_t - \check{f}_{0t})' \right) \check{\lambda}_{0i} + \bar{O}_P(\|\hat{\beta} - \beta_0\|) + \bar{O}_P(T^{1/(2m)}/(Th^2)). \end{aligned} \quad (\text{A.37})$$

*Proof.* The result follows immediately from the proof of Lemma 5.  $\square$

From (A.15) and the proof of Lemma 4 we have:

$$\begin{aligned} \Delta(\hat{\beta} - \beta_0) &= o_P(\|\hat{\beta} - \beta_0\|) + \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(1)} Z_{it} \\ &\quad - [\tilde{S}^{\beta f'} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S^{\lambda f'}] (\hat{F} - \check{F}_0) - [\tilde{S}^{\beta \lambda'} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} \tilde{S}^{\lambda \lambda'}] (\hat{\Lambda} - \check{\Lambda}_0) \\ &\quad - 1/2 \sum_{i=1}^N \sum_{h=1}^r (S_*^{\beta \lambda' \lambda_{ih}} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S_*^{\lambda \lambda' \lambda_{ih}}) (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{\Lambda} - \check{\Lambda}_0) \\ &\quad - 1/2 \sum_{i=1}^N \sum_{h=1}^r (S_*^{\beta f' \lambda_{ih}} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S_*^{\lambda f' \lambda_{ih}}) (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{F} - \check{F}_0) \\ &\quad - 1/2 \sum_{t=1}^T \sum_{h=1}^r (S_*^{\beta \lambda' f_{th}} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S_*^{\lambda \lambda' f_{th}}) (\hat{f}_{th} - \check{f}_{0t,h}) (\hat{\Lambda} - \check{\Lambda}_0) \\ &\quad - 1/2 \sum_{t=1}^T \sum_{h=1}^r (S_*^{\beta f' f_{th}} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S_*^{\lambda f' f_{th}}) (\hat{f}_{th} - \check{f}_{0t,h}) (\hat{F} - \check{F}_0). \end{aligned} \quad (\text{A.38})$$

In the next 5 lemmas, we analyze each term on the right-hand side of (A.38).

**Lemma 7.** Under Assumptions 1 to 4,

$$[\tilde{S}^{\beta \lambda'} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} \tilde{S}^{\lambda \lambda'}] (\hat{\Lambda} - \check{\Lambda}_0) = -\frac{b_1}{T} + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|).$$

*Proof.* **Step 1:** Note that we can write:

$$[\tilde{S}^{\beta \lambda'} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} \tilde{S}^{\lambda \lambda'}] (\hat{\Lambda} - \check{\Lambda}_0) = \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \hat{\mathbf{H}}(\hat{\lambda}_i - \check{\lambda}_{0i}).$$

Plugging in the result of Lemma 6, we have

$$\begin{aligned}
& [\tilde{S}^{\beta\lambda'} - \bar{S}^{\beta\lambda'}(\bar{S}^{\lambda\lambda'})^{-1}\tilde{S}^{\lambda\lambda'}](\hat{\Lambda} - \check{\Lambda}_0) \\
&= \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} f_{0t} \right) + \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} (\hat{\mathbf{H}}')^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) \right) \\
&\quad - \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} (\hat{f}_t - \check{f}_{0t})' \right) \tilde{\lambda}_{0i} \\
&\quad - \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \Omega_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} f'_{0t} \right) (\hat{\mathbf{H}}')^{-1} (\hat{\lambda}_i - \check{\lambda}_{0i}) \\
&\quad + \left[ O_P(\|\hat{\beta} - \beta_0\|) + O_P(T^{1/(2m)}/(Th^2)) \right] \cdot O_P \left( \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right\| \right). \quad (\text{A.39})
\end{aligned}$$

**Step 2:** By the Cauchy-Schwarz inequality,

$$\begin{aligned}
& \left\| \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} (\hat{\mathbf{H}}')^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) \right) \right\| \\
& \lesssim \sqrt{\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right\|^2} \cdot \sqrt{\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) \right\|^2}.
\end{aligned}$$

First, by Theorem 3 of [Yosihara \(1978\)](#),

$$\mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right\|^2 = \frac{1}{Th^2} \cdot \mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h \cdot l_{it}^{(2)} Z_{it} f'_{0t} \right\|^2 = O(1/(Th^2)).$$

Second, by Lemma 1, we have

$$\begin{aligned}
& \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) \right\|^2 \leq \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{NT} \sum_{j=1}^N \sum_{t=1}^T l_{it}^{(1)} e_{jt} \right\|^2 \cdot \|\hat{\Psi}\|^2 \\
&= O_P(1) \cdot \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{NT} \sum_{j \neq i}^N \sum_{t=1}^T l_{it}^{(1)} e_{jt} \right\|^2 + O_P(1/N^2) \cdot \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} e_{it} \right\|^2
\end{aligned}$$

For simplicity, consider the case where  $p = 1$ . Then by the mixing property we have

$$\begin{aligned}
\mathbb{E} \left\| \frac{1}{NT} \sum_{j \neq i}^N \sum_{t=1}^T l_{it}^{(1)} e_{jt} \right\|^2 &= \frac{1}{N^2 T^2} \sum_{j \neq i}^N \sum_{t=1}^T \sum_{s=1}^T \mathbb{E}[l_{it}^{(1)} l_{is}^{(1)}] \cdot \mathbb{E}[e_{jt} e_{js}] \\
&= \frac{1}{NT} \cdot \frac{1}{N} \sum_{j \neq i}^N \left( \frac{1}{T} \sum_{t=1}^T \sum_{s=1}^T \mathbb{E}[l_{it}^{(1)} l_{is}^{(1)}] \cdot \mathbb{E}[e_{jt} e_{js}] \right) = O((NT)^{-1}) = O(1/T^2).
\end{aligned}$$

Moreover, it is easy to show that

$$\mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} e_{it} \right\|^2 = O(1).$$

There, it follows that

$$\left\| \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} (\hat{\mathbf{H}}')^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) \right) \right\| = O_P \left( \frac{1}{T} \cdot \frac{1}{\sqrt{Th^2}} \right) = o_P(T^{-1}). \quad (\text{A.40})$$

Similarly, it can be shown that

$$\left\| \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} (\hat{f}_t - \check{f}_{0t})' \right) \tilde{\lambda}_{0i} \right\| = O_P \left( \frac{1}{T} \cdot \frac{1}{\sqrt{Th^4}} \right) = o_P(T^{-1}). \quad (\text{A.41})$$

**Step 3:** By the Cauchy-Schwarz inequality and Lemma 5,

$$\begin{aligned} & \left\| \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \Omega_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} f'_{0t} \right) (\hat{\mathbf{H}}')^{-1} (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| \\ & \lesssim \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \cdot \sqrt{\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right\|^2} \cdot \sqrt{\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T (l_{it}^{(2)} - \mathbf{f}_{it}(0)) f_{0t} f'_{0t} \right\|^2} \\ & = O_P(T^{1/(2m)} / T^{1/2}) \cdot O_P(1/\sqrt{Th^2}) \cdot \sqrt{\frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T (l_{it}^{(2)} - \mathbf{f}_{it}(0)) f_{0t} f'_{0t} \right\|^2}. \end{aligned}$$

Moreover, similar to the proof of the previous step, we can show that

$$\mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T (l_{it}^{(2)} - \mathbf{f}_{it}(0)) f_{0t} f'_{0t} \right\|^2 = O_P(1/(Th^2)).$$

Thus, by Assumption 3(vii)

$$\left\| \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \Omega_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} f_{0t} f'_{0t} \right) (\hat{\mathbf{H}}')^{-1} (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| = O_P \left( \frac{1}{T} \cdot \frac{T^{1/(2m)}}{T^{0.5-2c}} \right) = o_P \left( \frac{1}{T} \right). \quad (\text{A.42})$$

**Step 4:** By Lemma S1 and Assumption 3(vii), we can write

$$\begin{aligned} \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} f_{0t} \right) &= \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{T} \sum_{t=1}^T \tilde{l}_{it}^{(1)} f_{0t} \right) + o_P(h^q) \\ &\quad \frac{1}{T} \cdot \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T \tilde{l}_{it}^{(1)} f_{0t} \right) + o_P(T^{-1}). \quad (\text{A.43}) \end{aligned}$$

First, by Lemma S1 and the mixing property, we have

$$\begin{aligned}
& \mathbb{E} \left[ \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T \tilde{l}_{it}^{(1)} f_{0t} \right) \right] = \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T \sum_{s=1}^T \mathbb{E}[l_{it}^{(2)} \tilde{l}_{is}^{(1)} Z_{it} f'_{0t} \Omega_i^{-1} f_{0s}] \right) \\
& = \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T \mathbb{E}[l_{it}^{(2)} \tilde{l}_{it}^{(1)} Z_{it} f'_{0t} \Omega_i^{-1} f_{0t}] \right) + \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T \sum_{s \neq t}^T \mathbb{E}[l_{it}^{(2)} \tilde{l}_{is}^{(1)} Z_{it} f'_{0t} \Omega_i^{-1} f_{0s}] \right) \\
& = (\tau - 0.5) \cdot \frac{1}{N} \sum_{i=1}^N \omega_{T,i}^{(1)} + \frac{1}{N} \sum_{i=1}^N \omega_{T,i}^{(2)} + O(Th^q) = -b_1 + o(1),
\end{aligned}$$

where

$$\begin{aligned}
\omega_{T,i}^{(1)} &= \frac{1}{T} \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it}] f'_{0t} \Omega_i^{-1} f_{0t}, \\
\omega_{T,i}^{(2)} &= \frac{1}{T} \sum_{t=1}^T \sum_{s \neq t}^T \left( \tau \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it}] - \mathbb{E} \left[ \int_{\infty}^0 \mathbf{f}_{i,ts}(0,v|X_{it},X_{is})dv \cdot Z_{it} \right] \right) f'_{0t} \Omega_i^{-1} f_{0s},
\end{aligned}$$

and

$$b_1 = -(\tau - 0.5) \cdot \lim_{N,T \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \omega_{T,i}^{(1)} - \lim_{N,T \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \omega_{T,i}^{(2)}.$$

Second, similar to the proof the Lemma A6 of [Galvao and Kato \(2016\)](#), we can show that

$$\text{Var} \left( \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T \tilde{l}_{it}^{(1)} f_{0t} \right) \right) = o(1).$$

Thus, we have

$$\frac{1}{N} \sum_{i=1}^N \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(2)} Z_{it} f'_{0t} \right) \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T \tilde{l}_{it}^{(1)} f_{0t} \right) = -b_1 + o_P(1).$$

Finally, the last term on the right-hand side of (A.39) can be shown to be  $o_P(\|\hat{\beta} - \beta_0\|) + o_P(T^{-1})$ . Then the desired result follows from (A.39) to (A.43)  $\square$

**Lemma 8.** *Under Assumptions 1 to 4,*

$$[S^{\beta f'} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S^{\lambda f'}](\hat{F} - \check{F}_0) = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbf{A}_t(\mathbf{H}_0)^{-1} \Psi'_0 e_{it} - \frac{d_1}{N} + o_P(T^{-1}).$$

*Proof.* **Step 1:** Write

$$\begin{aligned}
& [S^{\beta f'} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S^{\lambda f'}] (\hat{F} - \check{F}_0) \\
&= \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(2)} X_{it} \lambda'_{0i} \right) (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}) - \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{N} \sum_{i=1}^N \Phi_i (\hat{\mathbf{H}})^{-1} (l_{it}^{(2)} \check{f}_{0t} \lambda'_{0i} - l_{it}^{(1)}) \right) (\hat{f}_t - \check{f}_{0t}) \\
&= \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{N} \sum_{i=1}^N [l_{it}^{(2)} X_{it} \lambda'_{0i} - l_{it}^{(2)} \Phi_i f_{0t} \lambda'_{0i} + l_{it}^{(1)} \Phi_i] \right) (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}) \\
&= \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{N} \sum_{i=1}^N [l_{it}^{(2)} Z_{it} \lambda'_{0i} + l_{it}^{(1)} \Phi_i] \right) (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}) \\
&= \frac{1}{T} \sum_{t=1}^T \mathbf{A}_t (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}) + \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{N} \sum_{i=1}^N \left\{ (l_{it}^{(2)} Z_{it} - \mathbb{E}[f_{it}(0|X_{it})Z_{it}]) \lambda'_{0i} + l_{it}^{(1)} \Phi_i \right\} \right) (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}).
\end{aligned}$$

**Step 2:** By Lemma 1, we have  $\hat{f}_t - \check{f}_{0t} = \hat{\Psi}' \bar{e}_t$ , thus,

$$\frac{1}{T} \sum_{t=1}^T \mathbf{A}_t (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}) = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbf{A}_t (\mathbf{H}_0)^{-1} \Psi'_0 e_{it} + \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbf{A}_t \left[ (\hat{\mathbf{H}})^{-1} \hat{\Psi}' - (\mathbf{H}_0)^{-1} \Psi'_0 \right] e_{it}.$$

It is easy to show that

$$\left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbf{A}_t \left[ (\hat{\mathbf{H}})^{-1} \hat{\Psi}' - (\mathbf{H}_0)^{-1} \Psi'_0 \right] e_{it} \right\| \leq O_P((NT)^{-1/2}) \cdot \left( \|\hat{\mathbf{H}} - \mathbf{H}_0\| + \|\hat{\Psi}' - \Psi'_0\| \right) = o_P(T^{-1}).$$

Therefore,

$$\frac{1}{T} \sum_{t=1}^T \mathbf{A}_t (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}) = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbf{A}_t (\mathbf{H}_0)^{-1} \Psi'_0 e_{it} + o_P(T^{-1}). \quad (\text{A.44})$$

**Step 3:** Since  $\|\hat{\mathbf{H}} - \mathbf{H}_0\| = O_P(T^{-1/2})$  and  $\|\hat{\Psi}' - \Psi'_0\| = O_P(T^{-1/2})$ , we have

$$\begin{aligned}
& \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{N} \sum_{i=1}^N \left\{ (l_{it}^{(2)} Z_{it} - \mathbb{E}[f_{it}(0|X_{it})Z_{it}]) \lambda'_{0i} + l_{it}^{(1)} \Phi_i \right\} \right) (\hat{\mathbf{H}})^{-1} (\hat{f}_t - \check{f}_{0t}) \\
&= (1+o_P(1)) \cdot \frac{1}{N} \cdot \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N \left\{ (l_{it}^{(2)} Z_{it} - \mathbb{E}[f_{it}(0|X_{it})Z_{it}]) \lambda'_{0i} + l_{it}^{(1)} \Phi_i \right\} \right) (\mathbf{H}_0)^{-1} \Psi'_0 \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right). \quad (\text{A.45})
\end{aligned}$$

Next, by Lemma S1 and Assumption 3(iv), it can be shown that

$$\begin{aligned}
& \mathbb{E} \left[ \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N \left\{ (l_{it}^{(2)} Z_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it}]) \lambda'_{0i} + l_{it}^{(1)} \Phi_i \right\} \right) (\mathbf{H}_0)^{-1} \Psi'_0 \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right) \right] \\
& = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E} \left[ (l_{it}^{(2)} Z_{it} \lambda'_{0i} + l_{it}^{(1)} \Phi_i) (\mathbf{H}_0)^{-1} \Psi'_0 e_{it} \right] \\
& = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it} \lambda'_{0i} (\mathbf{H}_0)^{-1} \Psi'_0 e_{it}] + o(1) = -d_1 + o(1). \quad (\text{A.46})
\end{aligned}$$

**Step 4:** Define

$$\mathcal{Z}_t = \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N \left\{ (l_{it}^{(2)} Z_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it}]) \lambda'_{0i} + l_{it}^{(1)} \Phi_i \right\} \right) (\mathbf{H}_0)^{-1} \Psi'_0 \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right).$$

To complete the proof, it remains to show that

$$\left\| \text{Var} \left( \frac{1}{T} \sum_{t=1}^T \mathcal{Z}_t \right) \right\| = o(1).$$

Note that

$$\left\| \text{Var} \left( \frac{1}{T} \sum_{t=1}^T \mathcal{Z}_t \right) \right\| \leq \frac{1}{Th^2} \cdot \mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h(\mathcal{Z}_t - \mathbb{E}[\mathcal{Z}_t]) \right\|^2.$$

By Assumption 3(iv),  $\{\mathcal{Z}_1, \dots, \mathcal{Z}_T\}$  is  $\alpha$ -mixing. Thus, by Theorem 3 of Yoshihara (1978) it suffices to show that  $\mathbb{E}\|\mathcal{Z}_t\|^4 < \infty$  for all  $t$ . By the Cauchy-Schwarz inequality,

$$\mathbb{E}\|\mathcal{Z}_t\|^4 \lesssim \sqrt{\mathbb{E} \left\| \frac{1}{\sqrt{N}} \sum_{i=1}^N h \left\{ (l_{it}^{(2)} Z_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it})Z_{it}]) \lambda'_{0i} + l_{it}^{(1)} \Phi_i \right\} \right\|^8} \cdot \sqrt{\mathbb{E} \left\| \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right\|^8} = O(1).$$

Thus, we have

$$\left\| \text{Var} \left( \frac{1}{T} \sum_{t=1}^T \mathcal{Z}_t \right) \right\| = O \left( \frac{1}{Th^2} \right) = o(1).$$

This completes the proof.  $\square$

**Lemma 9.** Under Assumptions 1 to 4,

$$\sum_{i=1}^N \sum_{h=1}^r (S_*^{\beta\lambda'\lambda_{ih}} - \bar{S}^{\beta\lambda'} (\bar{S}^{\lambda\lambda'})^{-1} S_*^{\lambda\lambda'\lambda_{ih}}) (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{\Lambda} - \check{\Lambda}_0) = -\frac{2b_2}{T} + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|).$$

*Proof.* Let  $\Phi_{ik}$  be the  $k$ th row of  $\Phi_i = \Xi_i \Omega_i^{-1}$ . The  $k$ th element of  $-\sum_{i=1}^N \sum_{h=1}^r (S_*^{\beta\lambda'\lambda_{ih}} - \bar{S}^{\beta\lambda'} (\bar{S}^{\lambda\lambda'})^{-1} S_*^{\lambda\lambda'\lambda_{ih}}) (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})$  is

$\check{\lambda}_{0i,h})(\hat{\Lambda} - \check{\Lambda}_0)$  can be written as

$$\begin{aligned}
& \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{h=1}^r l_{it}^{(3)}(*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*) f_t^{*'} f_{th} (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{\lambda}_i - \check{\lambda}_{0i}) \\
&= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T (\hat{\lambda}_i - \check{\lambda}_{0i})' l_{it}^{(3)}(*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*) f_t^* f_t^{*'} (\hat{\lambda}_i - \check{\lambda}_{0i}) \\
&= \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*) f_t^* f_t^{*'} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}).
\end{aligned}$$

**Step 1:** Note that

$$\begin{aligned}
& \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*) f_t^* f_t^{*'} \\
&= \hat{\mathbf{H}} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)} Z_{it,k} f_{0t} f_{0t}' \right) \hat{\mathbf{H}}' + \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} F_t^*) (f_t^* f_t^{*'} - \hat{\mathbf{H}} f_{0t} f_{0t}' \hat{\mathbf{H}}) \\
&\quad + \hat{\mathbf{H}} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) \Phi_{ik}(f_{0t} - (\hat{\mathbf{H}})^{-1} f_t^*) f_{0t} f_{0t}' \right) \hat{\mathbf{H}}' + \hat{\mathbf{H}} \left( \frac{1}{T} \sum_{t=1}^T (l_{it}^{(3)}(*) - l_{it}^{(3)}) Z_{it,k} f_{0t} f_{0t}' \right) \hat{\mathbf{H}}'.
\end{aligned}$$

Thus, by Lemma 2, Lemma 5 and Lemma S1 we have

$$\begin{aligned}
& \left\| \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)}(*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*) f_t^* f_t^{*'} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \right. \\
&\quad \left. - \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \hat{\mathbf{H}} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)} Z_{it,k} f_{0t} f_{0t}' \right) \hat{\mathbf{H}}' (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| \\
&= O_P(T^{-1+1/m}) \cdot O_P(T^{1/(2m)} / \sqrt{Th^6}) + O_P(T^{-1+1/m}) \cdot \max_i \frac{1}{T} \sum_{t=1}^T \|Z_{it}\| \cdot \|\hat{\beta} - \beta_0\| / h^3 = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|),
\end{aligned}$$

where we have used the fact that

$$|l_{it}^{(3)}(*) - l_{it}^{(3)}| \lesssim (\|\hat{\beta} - \beta_0\| + \|\hat{f}_t - \check{f}_{0t}\| + \|\hat{\lambda}_i - \check{\lambda}_{0i}\|) / h^3,$$

and thus

$$\begin{aligned}
& \max_i \left\| \frac{1}{T} \sum_{t=1}^T (l_{it}^{(3)}(*) - l_{it}^{(3)}) Z_{it,k} f_{0t} f_{0t}' \right\| \\
&\lesssim \max_i \frac{1}{T} \sum_{t=1}^T \|Z_{it}\| \cdot \|\hat{\beta} - \beta_0\| / h^3 + \max_t \|\hat{f}_t - \check{f}_{0t}\| / h^3 + \max_i \|\hat{\lambda}_i - \check{\lambda}_{0i}\| / h^3 \\
&= \max_i \frac{1}{T} \sum_{t=1}^T \|Z_{it}\| \cdot \|\hat{\beta} - \beta_0\| / h^3 + O_P(T^{1/(2m)} / \sqrt{Th^6}).
\end{aligned}$$

**Step 2:** Recall that

$$\mathbf{C}_{i,k} = -\frac{1}{T} \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}^{(1)}(0|X_{it})Z_{it,k}] f_{0t} f'_{0t}.$$

First, by Lemma 5,

$$\begin{aligned} & \left\| \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \hat{\mathbf{H}} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)} Z_{it,k} f_{0t} f'_{0t} \right) \hat{\mathbf{H}}' (\hat{\lambda}_i - \check{\lambda}_{0i}) - \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}' (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| \\ & \leq O_P(T^{-1+1/m}) \cdot \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)} Z_{it,k} f_{0t} f'_{0t} - \mathbf{C}_{i,k} \right\|. \end{aligned}$$

Second, by Lemma S1 and Theorem 3 of Yoshihara (1978),

$$\mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)} Z_{it,k} f_{0t} f'_{0t} - \mathbf{C}_{i,k} \right\| \leq \frac{1}{\sqrt{Th^4}} \sqrt{\mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h^2 (l_{it}^{(3)} Z_{it,k} - \mathbb{E}[l_{it}^{(3)} Z_{it,k}]) f_{0t} f'_{0t} \right\|^2} + o(T^{-1}) = O(1/\sqrt{Th^4}).$$

Thus, we have

$$\begin{aligned} & \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{h=1}^r l_{it}^{(3)} (*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*) f_t^{*'} f_{th}^* (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{\lambda}_i - \check{\lambda}_{0i}) \\ & = \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}' (\hat{\lambda}_i - \check{\lambda}_{0i}) + o_P(T^{-1}) \quad (\text{A.47}) \end{aligned}$$

because  $O_P(T^{-1+1/m}/\sqrt{Th^4}) = o_P(T^{-1})$  by Assumption 3(iii).

**Step 3:** By Lemma 6 we can write

$$\hat{\lambda}_i - \check{\lambda}_{0i} = \tilde{\Omega}_i^{-1} \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} \check{f}_{0t} + g_i + \bar{O}_P(\|\hat{\beta} - \beta_0\|) + \bar{O}_P(T^{1/(2m)}/(Th^2)).$$

where

$$g_i = \tilde{\Omega}_i^{-1} \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} (\hat{f}_t - \check{f}_{0t}) - \tilde{\Omega}_i^{-1} \left( \tilde{\Omega}_i - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \check{f}_{0t} \check{f}'_{0t} \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) - \tilde{\Omega}_i^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} \check{f}_{0t} (\hat{f}_t - \check{f}_{0t})' \right) \check{\lambda}_{0i}.$$

By the proof of Lemma 7, it can be shown that

$$\frac{1}{N} \sum_{i=1}^N \|g_i\| = O_P \left( \frac{1}{T^{1-1/m} h} \right).$$

Thus,

$$\begin{aligned}
& \left\| \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}' (\hat{\lambda}_i - \check{\lambda}_{0i}) - \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} f_{0t}' \right) \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} \left( \frac{1}{T} \sum_{t=1}^T l_{it}^{(1)} f_{0t} \right) \right\| \\
& = O_P \left( \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \right) \cdot \frac{1}{N} \sum_{i=1}^N \|g_i\| + o_P(\|\hat{\beta} - \beta_0\|) + O_P \left( \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \right) \cdot O_P(T^{1/(2m)} / (Th^2)) \\
& = O_P \left( \frac{1}{T^{1.5-1.5/m} h} + \frac{1}{T^{1.5-1/m} h^2} \right) + o_P(\|\hat{\beta} - \beta_0\|) = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|).
\end{aligned}$$

Thus, we can write

$$\begin{aligned}
& \frac{1}{N} \sum_{i=1}^N (\hat{\lambda}_i - \check{\lambda}_{0i})' \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}' (\hat{\lambda}_i - \check{\lambda}_{0i}) \\
& = \frac{1}{T} \cdot \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f_{0t}' \right) \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f_{0t} \right) + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|). \quad (\text{A.48})
\end{aligned}$$

**Step 4:** First, by Lemma S1,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f_{0t}' \right) \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f_{0t} \right) \right] \\
& = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E} \left[ \left( l_{it}^{(1)} \right)^2 f_{0t}' \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} f_{0t} \right] + \frac{1}{N} \sum_{i=1}^N \cdot \frac{1}{T} \sum_{t=1}^T \sum_{s \neq t}^T \mathbb{E} \left[ l_{it}^{(1)} l_{is}^{(1)} \right] f_{0t}' \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} f_{0s} \\
& = \frac{1}{N} \sum_{i=1}^N \omega_{T,i}^{(3)} + \frac{1}{N} \sum_{i=1}^N \omega_{T,i}^{(4)} + o(1) = 2b_2 + o(1),
\end{aligned}$$

where

$$\begin{aligned}
\omega_{T,i,k}^{(3)} & = \tau(1-\tau) \cdot \frac{1}{T} \sum_{t=1}^T f_{0t}' \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} f_{0t} \\
\omega_{T,i,k}^{(4)} & = \frac{1}{T} \sum_{t=1}^T \sum_{s \neq t}^T \{ \mathbb{E}[\mathbf{1}\{u_{it} \leq 0, u_{is} \leq 0\}] - \tau^2 \} f_{0t}' \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} f_{0s}
\end{aligned}$$

and

$$b_{2,k} = 0.5 \lim_{N,T \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \omega_{T,i,k}^{(3)} + 0.5 \lim_{N,T \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \omega_{T,i,k}^{(4)}.$$

**Step 5:** Finally, note that by Theorem 3 of Yoshihara (1978),

$$\begin{aligned} & \left\| \text{Var} \left[ \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f'_{0t} \right) \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f_{0t} \right) \right] \right\| \\ & \lesssim \frac{1}{N^2} \sum_{i=1}^N \mathbb{E} \left\| \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f'_{0t} \right) \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f_{0t} \right) \right\|^2 \lesssim \frac{1}{N^2} \sum_{i=1}^N \mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T l_{it}^{(1)} f'_{0t} \right\|^4 \\ & = O(N^{-1}) = o(1), \end{aligned}$$

it then follows from (A.47) and (A.48) that

$$\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{h=1}^r l_{it}^{(3)} (*) (X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*) f_t^{*'} f_{th}^* (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{\lambda}_i - \check{\lambda}_{0i}) = -2b_{2,k} + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|),$$

and the desired result follows.  $\square$

**Lemma 10.** Under Assumptions 1 to 4,

$$\sum_{t=1}^T \sum_{h=1}^r (S_*^{\beta f' f_{th}} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S_*^{\lambda f' f_{th}}) (\hat{f}_{th} - \check{f}_{0t,h}) (\hat{F} - \check{F}_0) = -\frac{2d_2}{N} + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|).$$

*Proof.* **Step 1:**

Note that the  $k$ th element of  $-\sum_{t=1}^T \sum_{h=1}^r (S_*^{\beta f' f_{th}} - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S_*^{\lambda f' f_{th}}) (\hat{f}_{th} - \check{f}_{0t,h}) (\hat{F} - \check{F}_0)$  can be written as:

$$\begin{aligned} & \frac{1}{T} \sum_{t=1}^T \sum_{h=1}^r \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} (*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] (\lambda_i^*)' \lambda_{ih}^* \right) (\hat{f}_{th} - \check{f}_{0t,h}) (\hat{f}_t - \check{f}_{0t}) \\ & = \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} (*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \lambda_i^* \lambda_i^{*'} \right) (\hat{f}_t - \check{f}_{0t}) \\ & \quad + 2 \cdot \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(2)} (*) \lambda_i^* \Phi_{ik}(\hat{\mathbf{H}})^{-1} \right) (\hat{f}_t - \check{f}_{0t}). \end{aligned}$$

**Step 2:**

First, by Lemma 2,

$$\begin{aligned} & \left\| \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} (*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \lambda_i^* \lambda_i^{*'} \right) (\hat{f}_t - \check{f}_{0t}) - \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} (\hat{f}_t - \check{f}_{0t}) \right\| \\ & \leq O_P(\log T/T) \cdot \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} (*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \lambda_i^* \lambda_i^{*'} - (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \right\|. \end{aligned}$$

Second, write

$$\begin{aligned} \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \lambda_i^* \lambda_i^{*''} &= \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} Z_{it,k} \check{\lambda}_{0i} (\check{\lambda}_{0i})' + \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} [\Phi_{ik} f_{0t} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \check{\lambda}_{0i} (\check{\lambda}_{0i})' \\ &+ \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] (\lambda_i^* (\lambda_i^*)' - \check{\lambda}_{0i} (\check{\lambda}_{0i})') + \frac{1}{N} \sum_{i=1}^N (l_{it}^{(3)}(*) - l_{it}^{(3)}) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \check{\lambda}_{0i} (\check{\lambda}_{0i})', \end{aligned}$$

it then follows that

$$\begin{aligned} \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \lambda_i^* \lambda_i^{*''} - (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \right\| \leq \\ O_P(1) \cdot \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} Z_{it,k} \lambda_{0i} \lambda_{0i}' - \mathbf{D}_{t,k} \right\| + O_P \left( \max_t \|\hat{f}_t - \check{f}_{0t}\| / h^3 \right) + O_P \left( \max_i \|\hat{\lambda}_i - \check{\lambda}_{0i}\| / h^3 \right) + O_P(\|\hat{\beta} - \beta_0\| / h^3), \end{aligned}$$

where we have used Lemma S1, Lemma 4, and the fact that

$$|l_{it}^{(3)}(*) - l_{it}^{(3)}| \lesssim \left( \|\hat{\beta} - \beta_0\| \cdot \|X_{it}\| + \|\hat{f}_t - \check{f}_{0t}\| + \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \right) / h^3$$

and

$$\max_t \frac{1}{N} \sum_{i=1}^N \|X_{it}\|^2 = O_P(1).$$

Similar to the proof of Lemma 9, it can be shown that

$$\frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)} Z_{it,k} \lambda_{0i} \lambda_{0i}' - \mathbf{D}_{t,k} \right\| = O_P(1/\sqrt{Nh^4}).$$

Therefore, by Lemma 2 and Lemma 5,

$$\begin{aligned} \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \lambda_i^* \lambda_i^{*''} \right) (\hat{f}_t - \check{f}_{0t}) &= \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} (\hat{f}_t - \check{f}_{0t}) \\ &+ O_P \left( \frac{T^{1/m}}{T} \right) \cdot O_P \left( \frac{T^{1/(2m)}}{\sqrt{Th^6}} \right) + O_P \left( \frac{T^{1/m}}{Th^3} \right) \cdot O_P(\|\hat{\beta} - \beta_0\|) \\ &= \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} (\hat{f}_t - \check{f}_{0t}) + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|). \quad (\text{A.49}) \end{aligned}$$

**Step 3:** By Lemma 2 we can write

$$\begin{aligned}
& \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} (\hat{f}_t - \check{f}_{0t}) = \frac{1}{N} \cdot \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right)' \hat{\Psi} (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \hat{\Psi}' \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right) \\
&= \frac{1}{N} \cdot \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right)' \Psi_0 (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \Psi'_0 \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right) + \frac{1}{N} \cdot o_P \left( \frac{1}{T} \sum_{t=1}^T \left\| \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right\|^2 \right) \\
&= \frac{1}{N} \cdot \frac{1}{T} \sum_{t=1}^T \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right)' \Psi_0 (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \Psi'_0 \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right) + o_P(N^{-1}).
\end{aligned}$$

Define

$$\mathcal{Z}_t = \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right)' \Psi_0 (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \Psi'_0 \left( \frac{1}{\sqrt{N}} \sum_{i=1}^N e_{it} \right).$$

First, it is easy to show that

$$\frac{1}{T} \sum_{t=1}^T \mathbb{E}[\mathcal{Z}_t] = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \text{Tr} \{ \mathbb{E}[e_{it} e'_{it}] \Psi_0 (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \Psi'_0 \}.$$

Second, we have

$$\text{Var} \left( \frac{1}{T} \sum_{t=1}^T \mathcal{Z}_t \right) = \frac{1}{T} \cdot \mathbb{E} \left| \frac{1}{\sqrt{T}} \sum_{t=1}^T (\mathcal{Z}_t - \mathbb{E}[\mathcal{Z}_t]) \right|^2.$$

Since the process  $\{\mathcal{Z}_1, \dots, \mathcal{Z}_T\}$  is  $\alpha$ -mixing, it then follows from  $\mathbb{E}|\mathcal{Z}_t|^4 \lesssim \mathbb{E}\|N^{-1/2} \sum_{i=1}^N e_{it}\|^8 < \infty$  that

$$\mathbb{E} \left| \frac{1}{\sqrt{T}} \sum_{t=1}^T (\mathcal{Z}_t - \mathbb{E}[\mathcal{Z}_t]) \right|^2 = O(1)$$

and thus

$$\text{Var} \left( \frac{1}{T} \sum_{t=1}^T \mathcal{Z}_t \right) = O(T^{-1}) = o(1).$$

Therefore, we have

$$\begin{aligned}
& \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(3)}(*) [X_{it,k} - \Phi_{ik}(\hat{\mathbf{H}})^{-1} f_t^*] \lambda_i^* \lambda_i^{*\prime} \right) (\hat{f}_t - \check{f}_{0t}) \\
&= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \text{Tr} \{ \mathbb{E}[e_{it} e'_{it}] \Psi_0 (\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k} (\mathbf{H}_0)^{-1} \Psi'_0 \} + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|). \quad (\text{A.50})
\end{aligned}$$

**Step 4:** Similarly, we can show that:

$$\begin{aligned} & \frac{1}{T} \sum_{t=1}^T (\hat{f}_t - \check{f}_{0t})' \left( \frac{1}{N} \sum_{i=1}^N l_{it}^{(2)}(*) \lambda_i^* \Phi_{ik}(\hat{\mathbf{H}})^{-1} \right) (\hat{f}_t - \check{f}_{0t}) \\ &= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \text{Tr} \{ \mathbb{E}[e_{ite} e_{it}'] \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 \} + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|). \end{aligned}$$

This completes the proof.  $\square$

**Lemma 11.** Under Assumptions 1 to 4,

$$\begin{aligned} & \sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih}) (\hat{F} - \check{F}_0) = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|) \\ & \sum_{t=1}^T \sum_{h=1}^r S_*^{\beta \lambda' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h}) (\hat{\Lambda} - \check{\Lambda}_0) = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|), \\ & \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} \sum_{i=1}^N \sum_{h=1}^r S_*^{\lambda f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0,ih}) (\hat{F} - \check{F}_0) = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|), \\ & \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} \sum_{t=T}^N \sum_{h=1}^r S_*^{\lambda \lambda' f_{th}} (\hat{f}_{th} - \check{f}_{0t,h}) (\hat{\Lambda} - \check{\Lambda}_0) = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|). \end{aligned}$$

*Proof.* To save space, we only prove the first result. The proof of the other results are similar. Write

$$\sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{F} - \check{F}_0) = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T [l_{it}^{(2)}(\beta^*, \lambda_i^*, f_t^*) X_{it} - l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (\lambda_i^*)' f_t^*] (\hat{\lambda}_i - \check{\lambda}_{0i})' (\hat{f}_t - \check{f}_{0t}).$$

Let  $R_{it}(*) = l_{it}^{(2)}(\beta^*, \lambda_i^*, f_t^*) X_{it} - l_{it}^{(3)}(\beta^*, \lambda_i^*, f_t^*) X_{it} (\lambda_i^*)' f_t^*$ , and  $R_{it} = l_{it}^{(2)} X_{it} - l_{it}^{(3)} X_{it} \lambda_{0i}' f_{0t}$ , then by Lemma 1 and Lemma 4 we have

$$\begin{aligned} & \sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h}) (\hat{F} - \check{F}_0) = \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T R_{it}(*) (\hat{f}_t - \check{f}_{0t})' \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \\ &= \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T R_{it} (\hat{f}_t - \check{f}_{0t})' \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) + O_P \left( \max_i \|\hat{\lambda}_i - \check{\lambda}_{0i}\| \right) \cdot O_P \left( \max_t \|\hat{f}_t - \check{f}_{0t}\| \right) \cdot O_P \left( \max_{i,t} \|R_{it}(*) - R_{it}\| \right). \end{aligned}$$

Similar to the proof of Lemma 10, it can be shown that

$$\max_{i,t} \|R_{it}(*) - R_{it}\| = O_P(T^{1/(2m)}) / (T^{1/2} h^3) + O_P(\|\hat{\beta} - \beta_0\| / h^3),$$

thus, by Lemma 2 and Lemma 6,

$$\begin{aligned} & O_P \left( \max_i \left\| \hat{\lambda}_i - \check{\lambda}_{0i} \right\| \right) \cdot O_P \left( \max_t \left\| \hat{f}_t - \check{f}_{0t} \right\| \right) \cdot O_P \left( \max_{i,t} \|R_{it}(\cdot) - R_{it}\| \right) \\ &= O_P \left( \frac{T^{1/m}}{T} \cdot \frac{T^{1/(2m)}}{T^{1/2} h^3} \right) + O_P \left( \frac{T^{1/m}}{Th^3} \right) O_P(\|\hat{\beta} - \beta_0\|) = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|). \end{aligned}$$

Therefore,

$$\sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \check{\lambda}_{0i,h})(\hat{F} - \check{F}_0) = \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T R_{it}(\hat{f}_t - \check{f}_{0t})' \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|). \quad (\text{A.51})$$

Note that

$$\left\| \frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T R_{it}(\hat{f}_t - \check{f}_{0t})' \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) \right\| \leq \max_i \left\| \hat{\lambda}_i - \check{\lambda}_{0i} \right\| \cdot \frac{1}{N} \sum_{i=1}^N \left\| \frac{1}{T} \sum_{t=1}^T R_{it}(\hat{f}_t - \check{f}_{0t})' \right\|.$$

Similar to Step 2 of the proof of Lemma 7, we can show that

$$\mathbb{E} \left\| \frac{1}{T} \sum_{t=1}^T R_{it}(\hat{F}_t - \check{F}_{0t})' \right\| = O(T^{-1}h^{-2}),$$

it then follows from Lemma 6 that

$$\frac{1}{N} \sum_{i=1}^N \left( \frac{1}{T} \sum_{t=1}^T R_{it}(\hat{f}_t - \check{f}_{0t})' \right) (\hat{\lambda}_i - \check{\lambda}_{0i}) = O_P \left( \frac{T^{1/(2m)}}{T^{3/2} h^2} \right) = o_P(T^{-1}). \quad (\text{A.52})$$

Thus, it follows from (A.51) and (A.52) that

$$\sum_{i=1}^N \sum_{h=1}^r S_*^{\beta f' \lambda_{ih}} (\hat{\lambda}_{ih} - \tilde{\lambda}_{0i,h})(\hat{F} - \tilde{F}_0) = o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|).$$

The proofs of the other results are similar and thus are omitted.  $\square$

## Proof of Theorem 2

*Proof.* It follows from (A.38) and Lemma 7 to Lemma 11 that

$$\Delta(\hat{\beta} - \beta_0) = - \left[ S^\beta - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S^\lambda \right] - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbf{A}_t(\mathbf{H}_0)^{-1} \Psi'_0 e_{it} + \frac{b}{T} + \frac{d}{N} + o_P(T^{-1}) + o_P(\|\hat{\beta} - \beta_0\|).$$

Since

$$S^\beta - \bar{S}^{\beta \lambda'} (\bar{S}^{\lambda \lambda'})^{-1} S^\lambda = - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(1)} Z_{it},$$

it then follows from Assumption 3(vii) that

$$\sqrt{NT}(\hat{\beta} - \beta_0) = \frac{1}{\sqrt{NT}} \sum_{i=1}^N \sum_{t=1}^T \Delta^{-1}(l_{it}^{(1)} Z_{it} - \mathbf{A}_t(\mathbf{H}_0)^{-1} \Psi_0' e_{it}) + \Delta^{-1}(\kappa b + \kappa^{-1} d) + o_P(1) + o_P(\sqrt{NT} \|\hat{\beta} - \beta_0\|).$$

Define  $W_{it}^* = l_{it}^{(1)} Z_{it} - \mathbf{A}_t(\mathbf{H}_0)^{-1} \Psi_0' e_{it}$  and  $\bar{W}_i^* = T^{-1/2} \sum_{t=1}^T W_{it}^*$ , it remains to show that

$$\frac{1}{\sqrt{NT}} \sum_{i=1}^N \sum_{t=1}^T W_{it}^* = \frac{1}{\sqrt{N}} \sum_{i=1}^N \bar{W}_i^* \xrightarrow{d} \mathcal{N}(0, \mathbf{V}). \quad (\text{A.53})$$

First, by Lemma S1, we have  $\mathbb{E}[\bar{W}_i^*] = O(h^q) = o(T^{-1})$ , thus

$$\frac{1}{\sqrt{N}} \sum_{i=1}^N \bar{W}_i^* = \frac{1}{\sqrt{N}} \sum_{i=1}^N (\bar{W}_i^* - \mathbb{E}[\bar{W}_i^*]) + o(1). \quad (\text{A.54})$$

Second, by the mixing property and Theorem 3 of Yoshihara (1978)

$$\mathbb{E} \|\bar{W}_i^* - \mathbb{E}[\bar{W}_i^*]\|^3 = \mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T (W_{it}^* - \mathbb{E}[W_{it}^*]) \right\|^3 < \infty.$$

Third, since  $\bar{W}_1^*, \dots, \bar{W}_N^*$  are independent, it follows from Lyapunov's CLT that

$$\frac{1}{\sqrt{N}} \sum_{i=1}^N (\bar{W}_i^* - \mathbb{E}[\bar{W}_i^*]) \xrightarrow{d} \mathcal{N}(0, \mathbf{V}^*) \quad (\text{A.55})$$

where

$$\begin{aligned} \mathbf{V}^* &= \frac{1}{N} \sum_{i=1}^N \mathbb{E} [(\bar{W}_i^* - \mathbb{E}[\bar{W}_i^*])(\bar{W}_i^* - \mathbb{E}[\bar{W}_i^*])'] = \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E} [(W_{it}^* - \mathbb{E}[W_{it}^*])(W_{it}^* - \mathbb{E}[W_{it}^*])'] \\ &\quad + \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{s \neq t}^T \mathbb{E} [(W_{it}^* - \mathbb{E}[W_{it}^*])(W_{is}^* - \mathbb{E}[W_{is}^*])']. \end{aligned}$$

Finally, by Lemma S1 it is easy to show that

$$\begin{aligned} \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E} [(W_{it}^* - \mathbb{E}[W_{it}^*])(W_{it}^* - \mathbb{E}[W_{it}^*])'] &= \tau(1-\tau) \cdot \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[Z_{it} Z'_{it}] \\ &\quad + \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbf{A}_t(\mathbf{H}_0)^{-1} \Psi_0' \mathbb{E}[e_{ite} e'_{it}] \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{A}'_t + o(1) = \mathbf{V}_1 + o(1), \end{aligned}$$

and

$$\begin{aligned}
& \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{s \neq t}^T \mathbb{E} [(W_{it}^* - \mathbb{E}[W_{it}^*])(W_{is}^* - \mathbb{E}[W_{is}^*])'] \\
&= \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{s \neq t}^T \mathbb{E} [(\tau^2 - \mathbb{E}[\mathbf{1}\{u_{it} \leq 0\}|X_{it}, X_{is}] - \mathbb{E}[\mathbf{1}\{u_{is} \leq 0\}|X_{it}, X_{is}] + \mathbb{E}[\mathbf{1}\{u_{it} \leq 0\}, u_{is} \leq 0\}|X_{it}, X_{is}]) Z_{it} Z_{is}'] \\
&\quad - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{s \neq t}^T \mathbb{E} [(\tau - \mathbb{E}[\mathbf{1}\{u_{it} \leq 0\}|X_{it}, X_{is}]) Z_{it} e_{is}] \Psi_0(\mathbf{H}_0')^{-1} \mathbf{A}_s' \\
&\quad - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{s \neq t}^T \mathbf{A}_t(\mathbf{H}_0)^{-1} \Psi_0' \mathbb{E}[e_{it} Z_{is}' (\tau - \mathbb{E}[\mathbf{1}\{u_{is} \leq 0\}|X_{it}, X_{is}])] \\
&\quad + \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{s \neq t}^T \mathbf{A}_t(\mathbf{H}_0)^{-1} \Psi_0' \mathbb{E}[e_{it} e_{is}] \Psi_0(\mathbf{H}_0')^{-1} \mathbf{A}_s' + o(1) = \mathbf{V}_2 + o(1).
\end{aligned}$$

Thus, we have  $\mathbf{V}^* = \mathbf{V} + o(1)$ , and (A.53) follows from (A.54) and (A.55). This completes the proof.  $\square$

#### A.4 Proof of Theorem 3

**Lemma 12.** Under Assumptions 1 to 4, we have

- (i)  $\max_{1 \leq i \leq N} \|\hat{\Xi}_i - \tilde{\Xi}_i\| = O_P(T^{1/(2m)} / (T^{0.5} h))$ ,  $\max_{1 \leq i \leq N} \|\hat{\Omega}_i - \tilde{\Omega}_i\| = O_P(T^{1/(2m)} / (T^{0.5} h))$ ,  $\max_{1 \leq i \leq N} \|\hat{\Phi}_i - \Phi_i \hat{\mathbf{H}}^{-1}\| = O_P(T^{1/(2m)} / (T^{0.5} h))$ ,  $\max_{i,t} \|\hat{Z}_{it} - Z_{it}\| = O_P(T^{1/(2m)} / (T^{0.5} h))$ .
- (ii)  $\max_{i,t} \|\hat{e}_{it} - e_{it}\| = O_P(T^{1/(2m)} / \sqrt{T})$ .

*Proof.* **Step 1:** Adding the subtracting terms, we have

$$\hat{\Xi}_i - \tilde{\Xi}_i = \frac{1}{T} \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) X_{it} \hat{f}'_t - \frac{1}{T} \sum_{t=1}^T l^{(2)}_{it} X_{it} f'_{0t} \hat{\mathbf{H}}' + \frac{1}{T} \sum_{t=1}^T (l^{(2)}_{it} X_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) X_{it}]) f'_{0t} \cdot \hat{\mathbf{H}}'.$$

First, by Theorem 3 of Yoshihara (1978)

$$\mathbb{E} \left\| \frac{1}{\sqrt{T}} \sum_{t=1}^T h \left( l^{(2)}_{it} X_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) X_{it}] \right) f'_{0t} \right\|^J < M$$

for any  $2m > J \geq 2$ . It then follows from Lemma 2.2.2 of van der Vaart and Wellner (1996) that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T (l^{(2)}_{it} X_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) X_{it}]) f'_{0t} \right\| = O_P \left( \frac{N^{1/J}}{\sqrt{T} h} \right) = O_P \left( \frac{T^{1/(2m)}}{T^{0.5} h} \right).$$

Second,

$$\left\| \frac{1}{T} \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) X_{it} \hat{f}'_t - \frac{1}{T} \sum_{t=1}^T l^{(2)}_{it} X_{it} f'_{0t} \hat{\mathbf{H}}' \right\| \leq \left\| \frac{1}{T} \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) X_{it} (\hat{f}'_t - f'_{0t})' \right\| + \left\| \frac{1}{T} \sum_{t=1}^T (l^{(2)}(\hat{u}_{it}) - l^{(2)}_{it}) X_{it} f'_{0t} \right\| \cdot \|\hat{\mathbf{H}}'\|.$$

The first term on the right-hand side of the above inequality is  $O_P(1/\sqrt{Nh^2})$  by Lemma 2. For the second term, using Taylor expansion and Lemma 5 we can write

$$\begin{aligned} \frac{1}{T} \sum_{t=1}^T \left( l^{(2)}(\hat{u}_{it}) - l^{(2)}_t \right) X_{it} f'_{0t} &= -\frac{1}{T} \sum_{t=1}^T \left\{ l^{(3)}_{it} [X'_{it}(\hat{\beta} - \beta_0)] X_{it} f'_{0t} + l^{(3)}_{it} [\tilde{f}'_{0t}(\hat{\lambda}_i - \tilde{\lambda}_{0i})] X_{it} f'_{0t} + l^{(3)}_{it} [\tilde{\lambda}'_{0i}(\hat{f}_t - \tilde{f}_{0t})] X_{it} f'_{0t} \right\} \\ &\quad + \bar{O}_P \left( T^{1/m}/(Th^3) \right). \end{aligned}$$

By Theorem 2,

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} [X'_{it}(\hat{\beta} - \beta_0)] X_{it} f'_{0t} \right\| = O_P \left( \frac{1}{\sqrt{NT}} \cdot \frac{1}{h^2} \right) = o_P \left( \frac{1}{\sqrt{Th}} \right).$$

Next, by Lemma 2 and Lemma 5,

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} [\tilde{f}'_{0t}(\hat{\lambda}_i - \tilde{\lambda}_{0i})] X_{it} f'_{0t} \right\| &\leq \max_{1 \leq i \leq N} \max_{1 \leq k \leq r} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} X_{it} f'_{0t} \tilde{f}_{0t,k} \right\| \cdot \max_{1 \leq i \leq N} \|\hat{\lambda}_i - \tilde{\lambda}_{0i}\| \\ &= O_P \left( \frac{T^{1/(2m)}}{T^{0.5}} \right) \cdot \max_{1 \leq i \leq N} \max_{1 \leq k \leq r} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} X_{it} f'_{0t} \tilde{f}_{0t,k} \right\|. \end{aligned}$$

Similar to the proof of Lemma 4, it can be shown that

$$\max_{1 \leq i \leq N} \max_{1 \leq k \leq r} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} X_{it} f'_{0t} \tilde{f}_{0t,k} \right\| = O_P(1).$$

Thus, we have

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} [\tilde{f}'_{0t}(\hat{\lambda}_i - \tilde{\lambda}_{0i})] X_{it} f'_{0t} \right\| = O_P \left( \frac{T^{1/(2m)}}{T^{0.5}} \right) = o_P \left( \frac{T^{1/(2m)}}{T^{0.5}h} \right).$$

Similarly,

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} [\tilde{\lambda}'_{0i}(\hat{f}_t - \tilde{f}_{0t})] X_{it} f'_{0t} \right\| &\leq \max_{1 \leq k \leq r} \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}_{it} (\hat{f}_{t,k} - \tilde{f}_{0t,k}) X_{it} f'_{0t} \tilde{\lambda}_{0i,k} \right\| \\ &\lesssim \max_{1 \leq t \leq T} \|\hat{f}_{0t} - \tilde{f}_{0t}\| \cdot \max_{1 \leq i \leq N} \frac{1}{T} \sum_{t=1}^T |l^{(3)}_{it}| \cdot \|X_{it} f'_{0t}\|. \end{aligned}$$

Since  $\mathbb{E}[|l^{(3)}_{it}| \cdot \|X_{it} f'_{0t}\|] = \bar{O}(h^{-1})$  by Lemma S1, and it can be shown that

$$\max_{1 \leq i \leq N} \frac{1}{T} \sum_{t=1}^T \left\{ |l^{(3)}_{it}| \cdot \|X_{it} f'_{0t}\| - \mathbb{E}[|l^{(3)}_{it}| \cdot \|X_{it} f'_{0t}\|] \right\} = O_P \left( \frac{T^{1/(2m)}}{T^{0.5}h^2} \right),$$

and it follows from Lemma 2 that

$$\max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l_{it}^{(3)} [\tilde{\lambda}'_{0i}(\hat{f}_t - \check{f}_{0t})] X_{it} f'_{0t} \right\| = O_P(T^{1/(2m)}/\sqrt{N}) \cdot O_P(h^{-1}) = o_P\left(\frac{T^{1/(2m)}}{T^{0.5}h}\right).$$

Combining all the above results, we have

$$\max_{1 \leq i \leq N} \|\hat{\Xi}_i - \tilde{\Xi}_i\| = O_P\left(\frac{T^{1/(2m)}}{T^{0.5}h}\right).$$

It can be shown in a similar way that

$$\max_{1 \leq i \leq N} \|\hat{\Omega}_i - \tilde{\Omega}_i\| = O_P\left(\frac{T^{1/(2m)}}{T^{0.5}h}\right).$$

Moreover, by the definition of  $\hat{\Phi}_i$  and  $\Phi_i$ ,

$$\max_{1 \leq i \leq N} \|\hat{\Phi}_i - \Phi_i \hat{\mathbf{H}}^{-1}\| = \max_{1 \leq i \leq N} \|\hat{\Xi}_i \hat{\Omega}_i^{-1} - \Xi_i \Omega_i^{-1} \hat{\mathbf{H}}^{-1}\| = \max_{1 \leq i \leq N} \|\hat{\Xi}_i \hat{\Omega}_i^{-1} - \tilde{\Xi}_i \tilde{\Omega}_i^{-1}\|,$$

thus, it follows from the above results and the fact that  $\tilde{\Omega}_i$  is positive definite with probability approaching 1 that

$$\max_{1 \leq i \leq N} \|\hat{\Phi}_i - \Phi_i \hat{\mathbf{H}}^{-1}\| = O_P\left(\frac{T^{1/(2m)}}{T^{0.5}h}\right).$$

Finally, by the definitions of  $Z_{it}$  and  $\hat{Z}_{it}$  and the results above,

$$\max_{i,t} \|\hat{Z}_{it} - Z_{it}\| \lesssim \max_i \|\tilde{\Xi}_i - \hat{\Xi}_i\| + \max_i \|\tilde{\Omega}_i^{-1} - \hat{\Omega}_i^{-1}\| + \max_t \|\hat{f}_t - \check{f}_{0t}\| = O_P\left(\frac{T^{1/(2m)}}{T^{0.5}h}\right).$$

**Step 2:** Define  $X_i = (X_{i1}, \dots, X_{iT})'$ , and  $\hat{\Gamma}'_i = (\hat{F}' \hat{F})^{-1} \hat{F}' X_i$ . Then we have  $\hat{e}_{it} = X_{it} - \hat{\Gamma}_i \hat{f}_t$ , and  $e_{it} = X_{it} - \Gamma_i f_{0t} = X_{it} - \Gamma_i \hat{\mathbf{H}}^{-1} \check{f}_{0t}$ . It follows that

$$\max_{i,t} \|\hat{e}_{it} - e_{it}\| \lesssim \max_i \|\hat{\Gamma}_i - \Gamma_i \hat{\mathbf{H}}^{-1}\| + \max_t \|\hat{f}_t - \check{f}_{0t}\|.$$

Write

$$\hat{\Gamma}'_i = \left( \frac{1}{T} \sum_{t=1}^T \hat{f}_t \hat{f}'_t \right)^{-1} \left( \frac{1}{T} \sum_{t=1}^T \hat{f}_t X'_{it} \right) = \left( \frac{1}{T} \sum_{t=1}^T \hat{f}_t \hat{f}'_t \right)^{-1} \left( \frac{1}{T} \sum_{t=1}^T \hat{f}_t f'_{0t} \right) \Gamma'_i + \left( \frac{1}{T} \sum_{t=1}^T \hat{f}_t \hat{f}'_t \right)^{-1} \left( \frac{1}{T} \sum_{t=1}^T \hat{f}_t e'_{it} \right).$$

By Lemma 2, we have

$$\left( \frac{1}{T} \sum_{t=1}^T \hat{f}_t \hat{f}'_t \right)^{-1} = (\hat{\mathbf{H}}')^{-1} \left( \frac{1}{T} \sum_{t=1}^T f_{0t} f'_{0t} \right)^{-1} \hat{\mathbf{H}}^{-1} + O_P(N^{-1/2})$$

and

$$\frac{1}{T} \sum_{t=1}^T \hat{f}_t f'_{0t} = \hat{\mathbf{H}} \cdot \frac{1}{T} \sum_{t=1}^T f_{0t} f'_{0t} + O_P(N^{-1/2}).$$

Thus, it follows that

$$\max_i \|\hat{\Gamma}_i - \Gamma_i \hat{\mathbf{H}}^{-1}\| \leq O_P(N^{-1/2}) + \max_i \left\| \frac{1}{T} \sum_{t=1}^T \hat{f}_t e'_{it} \right\|.$$

Next, by the proof of Lemma 2,

$$\max_i \left\| \frac{1}{T} \sum_{t=1}^T \hat{f}_t e'_{it} \right\| \leq \max_i \left\| \frac{1}{T} \sum_{t=1}^T f_{0t} e'_{it} \right\| + O_P(N^{-1/2}) = O_P(N^{1/(2m)} / \sqrt{T}) + O_P(N^{-1/2}).$$

It then follows from the above results and Lemma 2 that  $\max_{i,t} \|\hat{e}_{it} - e_{it}\| = O_P(T^{1/(2m)} / \sqrt{T})$ .  $\square$

### Proof of Theorem 3

*Proof. Step 1:* We first show that  $\hat{\Delta} = \Delta + o_P(1)$ .

Note that by the definition of  $\Delta$  and Lemma S1,

$$\|\hat{\Delta} - \Delta\| \leq \left\| \hat{\Delta} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(2)} Z_{it} Z_{it} \right\| + \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \{l_{it}^{(2)} Z_{it} Z_{it} - \mathbb{E}[l_{it}^{(2)} Z_{it} Z_{it}]\} \right\| + o(1).$$

Similar to Step 2 in the proof of Lemma 4, it can be shown that

$$\frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \{l_{it}^{(2)} Z_{it} Z_{it} - \mathbb{E}[l_{it}^{(2)} Z_{it} Z_{it}]\} = o_P(1).$$

Thus, it remains to show that

$$\left\| \hat{\Delta} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(2)} Z_{it} Z_{it} \right\| = o_P(1).$$

By the definition of  $\hat{\Delta}$ , we have

$$\left\| \hat{\Delta} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(2)} Z_{it} Z_{it} \right\| \lesssim \max_{i,t} \|l^{(2)}(\hat{u}_{it}) - l_{it}^{(2)}\| + \max_{i,t} \|\hat{Z}_{it} - Z_{it}\| / h.$$

First, by Lemma 2 and Lemma 5,

$$\begin{aligned} \max_{i,t} \|l^{(2)}(\hat{u}_{it}) - l_{it}^{(2)}\| &\lesssim (\|\hat{\beta} - \beta_0\| \cdot \|X_{it}\| + \max_i \|\hat{\lambda}_i - \lambda_{0i}\| + \max_t \|\hat{f}_t - f_{0t}\|) / h^2 \\ &= O_P(T^{1/(2m)} / \sqrt{Th^4}) = o_P(1). \end{aligned}$$

Second, by Lemma 12

$$\max_{i,t} \|\hat{Z}_{it} - Z_{it}\| / h = O_P\left(\frac{T^{1/(2m)}}{T^{0.5}h^2}\right) = o_P(1).$$

Therefore, the desired result follows.

**Step 2:** Next, we show that  $\hat{b} = b + o_P(1)$ .

By definitions, it suffices to show that

$$\max_{1 \leq i \leq N} \|\hat{\omega}_{T,i}^{(1)} - \omega_{T,i}^{(1)}\| = o_P(1), \quad \max_{1 \leq i \leq N} \|\hat{\omega}_{T,i}^{(2)} - \omega_{T,i}^{(2)}\| = o_P(1),$$

and

$$\max_{1 \leq i \leq N} \|\hat{\omega}_{T,i,k}^{(3)} - \omega_{T,i,k}^{(3)}\| = o_P(1), \quad \max_{1 \leq i \leq N} \|\hat{\omega}_{T,i,k}^{(4)} - \omega_{T,i,k}^{(4)}\| = o_P(1)$$

for  $k \leq r$ .

First, similar to the proof of the previous step, we have

$$\begin{aligned} \max_{1 \leq i \leq N} \|\hat{\omega}_{T,i}^{(1)} - \omega_{T,i}^{(1)}\| &= \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \cdot \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_t - \frac{1}{T} \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it}] f'_{0t} \Omega_i^{-1} f_{0t} \right\| \\ &\leq \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \cdot \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_t - \frac{1}{T} \sum_{t=1}^T l_{it}^{(2)} Z_{it} \cdot f'_{0t} \Omega_i^{-1} f_{0t} \right\| + \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T \left( l_{it}^{(2)} Z_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it}] \right) f'_{0t} \Omega_i^{-1} f_{0t} \right\| \\ &\lesssim \max_{i,t} \|l^{(2)}(\hat{u}_{it}) - l_{it}^{(2)}\| + \max_{i,t} \|\hat{Z}_{it} - Z_{it}\| / h + \max_{i,t} \|\hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_t - f'_{0t} \Omega_i^{-1} f_{0t}\| / h + o_P(1) \\ &= \max_{i,t} \|\hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_t - f'_{0t} \Omega_i^{-1} f_{0t}\| / h + o_P(1). \end{aligned}$$

Note that by Lemma 2 and Lemma 12,

$$\begin{aligned} \max_{i,t} \|\hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_t - f'_{0t} \Omega_i^{-1} f_{0t}\| / h &= \max_{i,t} \|\hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_t - f'_{0t} \hat{\mathbf{H}}' (\hat{\mathbf{H}}')^{-1} \Omega_i^{-1} (\hat{\mathbf{H}})^{-1} \hat{\mathbf{H}} f_{0t}\| / h \\ &= \max_{i,t} \|\hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_t - \check{f}'_{0t} \tilde{\Omega}_i^{-1} \check{f}_{0t}\| / h \lesssim \max_t \|\hat{f}_t - \check{f}_{0t}\| / h + \max_i \|\hat{\Omega}_i^{-1} - \tilde{\Omega}_i^{-1}\| / h = O_P\left(\frac{T^{1/(2m)}}{T^{0.5} h^2}\right) = o_P(1). \end{aligned}$$

Thus, it follows that

$$\max_{1 \leq i \leq N} \|\hat{\omega}_{T,i}^{(1)} - \omega_{T,i}^{(1)}\| = o_P(1).$$

Second, by Lemma 12

$$\begin{aligned} \max_{1 \leq i \leq N} \|\hat{\omega}_{T,i}^{(3)} - \omega_{T,i}^{(3)}\| &= \tau(1-\tau) \cdot \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T \hat{f}'_t \hat{\Omega}_i^{-1} \hat{\mathbf{C}}_{i,k} \hat{\Omega}_i^{-1} \hat{f}_t - \frac{1}{T} \sum_{t=1}^T f'_{0t} \Omega_i^{-1} \mathbf{C}_{i,k} \Omega_i^{-1} f_{0t} \right\| \\ &\lesssim \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^T \hat{f}'_t \hat{\Omega}_i^{-1} \hat{\mathbf{C}}_{i,k} \hat{\Omega}_i^{-1} \hat{f}_t - \frac{1}{T} \sum_{t=1}^T \check{f}'_{0t} \tilde{\Omega}_i^{-1} \cdot \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}' \cdot \tilde{\Omega}_i^{-1} \check{f}_{0t} \right\| \\ &\leq \max_t \|\hat{f}_t - \check{f}_{0t}\| / h^2 + \max_i \|\hat{\Omega}_i^{-1} - \tilde{\Omega}_i^{-1}\| / h^2 + \max_i \|\hat{\mathbf{C}}_{i,k} - \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}'\| = \max_i \|\hat{\mathbf{C}}_{i,k} - \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}'\| + o_P(1). \end{aligned}$$

By the definitions of  $\hat{\mathbf{C}}_{i,k}$  and  $\mathbf{C}_{i,k}$ , it follows from Lemma 12 that

$$\begin{aligned} \max_i \left\| \hat{\mathbf{C}}_{i,k} - \hat{\mathbf{H}} \mathbf{C}_{i,k} \hat{\mathbf{H}}' \right\| &= \max_i \left\| \frac{1}{T} \sum_{t=1}^T l^{(3)}(\hat{u}_{it}) \hat{Z}_{it,k} \hat{f}_t \hat{f}'_t - \frac{1}{T} \sum_{t=1}^T -\mathbb{E}[f_{it}^{(1)}(0|X_{it}) Z_{it,k}] \hat{\mathbf{H}} f_{0t} f'_{0t} \hat{\mathbf{H}}' \right\| \\ &\leq \max_i \left\| \frac{1}{T} \sum_{t=1}^T \left( l^{(3)}(\hat{u}_{it}) \hat{Z}_{it,k} \hat{f}_t \hat{f}'_t - l_{it}^{(3)} Z_{it,k} \check{f}_{0t} \check{f}'_{0t} \right) \right\| + \max_i \left\| \frac{1}{T} \sum_{t=1}^T \left( l_{it}^{(3)} Z_{it,k} + \mathbb{E}[f_{it}^{(1)}(0|X_{it}) Z_{it,k}] \right) f_{0t} f'_{0t} \right\| \cdot \|\hat{\mathbf{H}}\|^2 \\ &\lesssim \max_t \|\hat{f}_t - \check{f}_{0t}\|/h^2 + \max_{i,t} \|\hat{u}_{it} - u_{it}\|/h^3 + \max_{i,t} \|\hat{Z}_{it} - Z_{it}\|/h^2 + o_P(1) = O_P\left(\frac{T^{1/(2m)}}{\sqrt{Th^6}}\right) + o_P(1) = o_P(1). \end{aligned}$$

Thus, we have

$$\max_{1 \leq i \leq N} \left\| \hat{\omega}_{T,i}^{(3)} - \omega_{T,i}^{(3)} \right\| = o_P(1).$$

Third, by the proof Lemma 7 and Assumption 2(iv), it can be shown that

$$\omega_{T,i}^{(2)} = \frac{1}{T} \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \mathbb{E} \left[ l_{it}^{(2)} Z_{it} l_{it}^{(1)} \right] \cdot f'_{0t} \Omega_i^{-1} f_{0s} + \frac{1}{T} \sum_{t=L+1}^T \sum_{s=t-L}^{t-1} \mathbb{E} \left[ l_{it}^{(2)} Z_{it} l_{is}^{(1)} \right] \cdot f'_{0t} \Omega_i^{-1} f_{0s} + \bar{O}_P(Th^q) + O(\alpha^L).$$

Since  $0 < \alpha < 1$  and  $L \rightarrow \infty$  as  $T \rightarrow \infty$ , by the definition of  $\hat{\omega}_{T,i}^{(2)}$ , it follows that

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \hat{\omega}_{T,i}^{(2)} - \omega_{T,i}^{(2)} \right\| &\leq \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} l^{(1)}(\hat{u}_{is}) \cdot \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_s - \mathbb{E} \left[ l_{it}^{(2)} Z_{it} l_{it}^{(1)} \right] \cdot f'_{0t} \Omega_i^{-1} f_{0s} \right\} \right\| \\ &\quad + \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=L+1}^T \sum_{s=t-L}^{t-1} \left\{ l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} l^{(1)}(\hat{u}_{is}) \cdot \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_s - \mathbb{E} \left[ l_{it}^{(2)} Z_{it} l_{is}^{(1)} \right] \cdot f'_{0t} \Omega_i^{-1} f_{0s} \right\} \right\| + o_P(1). \end{aligned}$$

Note that

$$\begin{aligned} \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} l^{(1)}(\hat{u}_{is}) \cdot \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_s - \mathbb{E} \left[ l_{it}^{(2)} Z_{it} l_{it}^{(1)} \right] \cdot f'_{0t} \Omega_i^{-1} f_{0s} \right\} \right\| &\leq \\ \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} l^{(1)}(\hat{u}_{is}) \cdot \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_s - l_{it}^{(2)} Z_{it} l_{it}^{(1)} \cdot f'_{0t} \Omega_i^{-1} f_{0s} \right\} \right\| &+ \\ \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ l_{it}^{(2)} Z_{it} l_{it}^{(1)} - \mathbb{E} \left[ l_{it}^{(2)} Z_{it} l_{it}^{(1)} \right] \right\} \cdot f'_{0t} \Omega_i^{-1} f_{0s} \right\|. \end{aligned}$$

It can be shown that the second term on the right-hand side of the above inequality is  $O_P(LT^{1/(2m)})/(T^{0.5}h) =$

$o_P(1)$ . For the first term, similar to the proof for  $\hat{\omega}_{T,i}^{(1)}$  we have

$$\begin{aligned} & \max_{1 \leq i \leq N} \left\| \frac{1}{T} \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} l^{(1)}(\hat{u}_{is}) \cdot \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_s - l_{it}^{(2)} Z_{it} l_{it}^{(1)} \cdot f'_{0t} \Omega_i^{-1} f_{0s} \right\} \right\| \\ & \lesssim L \left( \max_{i,t} \left\| l^{(2)}(\hat{u}_{it}) - l_{it}^{(2)} \right\| + \max_{i,t} \left\| l^{(1)}(\hat{u}_{it}) - l_{it}^{(1)} \right\| / h + \max_{i,t} \left\| \hat{Z}_{it} - Z_{it} \right\| / h + \max_{t,s} \left\| \hat{f}'_t \hat{\Omega}_i^{-1} \hat{f}_s - f'_{0t} \Omega_i^{-1} f_{0s} \right\| / h \right) \\ & = O_P \left( \frac{LT^{1/(2m)}}{T^{0.5} h^2} \right) = o_P(1). \end{aligned}$$

Thus, we have

$$\max_{1 \leq i \leq N} \left\| \hat{\omega}_{T,i}^{(2)} - \omega_{T,i}^{(2)} \right\| = o_P(1).$$

Finally, we show can that

$$\max_{1 \leq i \leq N} \left\| \hat{\omega}_{T,i}^{(4)} - \omega_{T,i}^{(4)} \right\| = O_P \left( \frac{LT^{1/(2m)}}{T^{0.5} h^3} \right) = o_P(1)$$

in a similar way. Thus, we can conclude that  $\hat{b} = b + o_P(1)$ .

**Step 3:** Finally, we show that  $\hat{d} = d + o_P(1)$ , which follows from  $\hat{d}_1 = d_1 + o_P(1)$  and  $\hat{d}_{2,k} = d_{2,k} + o_P(1)$  for all  $k \leq r$ .

First, by the definitions of  $d_1$  and  $\hat{d}_1$ , we need to show that

$$\left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \hat{\lambda}'_i \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it} \lambda'_{0i}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it}] \right\| = o_P(1).$$

We have

$$\begin{aligned} & \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \hat{\lambda}'_i \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it} \lambda'_{0i}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it}] \right\| \\ & \leq \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \hat{\lambda}'_i \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(2)} Z_{it} \lambda'_{0i}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it} \right\| \\ & + \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l_{it}^{(2)} Z_{it} \lambda'_{0i}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it} \lambda'_{0i}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it}] \right\|. \end{aligned}$$

It is easy to see that the second term on the right-hand side of the above inequality is  $O_P(1/\sqrt{NT h^2}) +$

$O_P(h^q) = o_P(1)$ . For the first term, by Lemma 2, Lemma 5 and Lemma 12 we have

$$\begin{aligned} & \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \hat{\lambda}'_i \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T l^{(2)}_{it} Z_{it} \lambda'_{0i} (\mathbf{H}_0)^{-1} \Psi'_0 e_{it} \right\| \\ & \leq \max_{i,t} \|l^{(2)}(\hat{u}_{it}) - l^{(2)}_{it}\| + \max_{i,t} \|\hat{Z}_{it} - Z_{it}\|/h + \max_i \|\hat{\lambda}_i - \tilde{\lambda}_i\|/h + \|\hat{\Psi} - \Psi_0\|/h + \max_{i,t} \|\hat{e}_{it} - e_{it}\|/h \\ & \quad = O_P\left(\frac{T^{1/(2m)}}{\sqrt{Th^4}}\right) = o_P(1). \end{aligned}$$

Therefore, we can conclude that  $\hat{d}_1 = d_1 + o_P(1)$ .

Second, by the definitions of  $d_{2,k}$  and  $\hat{d}_{2,k}$ , we need to show that

$$\left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \hat{e}'_{it} \hat{\Psi} \hat{\mathbf{B}}_{t,k} \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \text{Tr} \{ \mathbb{E}[e_{it} e'_{it}] \cdot \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 \} \right\| = o_P(1),$$

and

$$\left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \hat{e}'_{it} \hat{\Psi} \hat{\mathbf{D}}_{t,k} \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \text{Tr} \{ \mathbb{E}[e_{it} e'_{it}] \cdot \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{D}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 \} \right\| = o_P(1).$$

To save space, we only prove the first result. The proof of the second result is similar. Note that

$$\begin{aligned} & \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \hat{e}'_{it} \hat{\Psi} \hat{\mathbf{B}}_{t,k} \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \text{Tr} \{ \mathbb{E}[e_{it} e'_{it}] \cdot \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 \} \right\| \\ & \leq \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \hat{e}'_{it} \hat{\Psi} \hat{\mathbf{B}}_{t,k} \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T e'_{it} \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it} \right\| + \\ & \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T e'_{it} \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \text{Tr} \{ \mathbb{E}[e_{it} e'_{it}] \cdot \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 \} \right\|. \end{aligned}$$

It is easy to see that the second term on the right-hand side of the above inequality is  $O_P(1/\sqrt{NT})$ . For the first term, we have

$$\begin{aligned} & \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \hat{e}'_{it} \hat{\Psi} \hat{\mathbf{B}}_{t,k} \hat{\Psi}' \hat{e}_{it} - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T e'_{it} \Psi_0(\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \Psi'_0 e_{it} \right\| \\ & \lesssim \max_{i,t} \|\hat{e}_{it} - e_{it}\|/h + \|\hat{\Psi} - \Psi_0\|/h + \max_t \left\| \hat{\mathbf{B}}_{t,k} - (\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \right\| \\ & = \max_t \left\| \hat{\mathbf{B}}_{t,k} - (\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k}(\mathbf{H}_0)^{-1} \right\| + O_P(T^{1/(2m)}/\sqrt{Th^2}). \end{aligned}$$

By the definitions of  $\hat{\mathbf{B}}_{t,k}$  and  $\mathbf{B}_{t,k}$ , it follows that

$$\begin{aligned}
\max_t \left\| \hat{\mathbf{B}}_{t,k} - (\mathbf{H}'_0)^{-1} \mathbf{B}_{t,k} (\mathbf{H}_0)^{-1} \right\| &= \max_t \left\| \frac{1}{N} \sum_{i=1}^N l^{(2)}(\hat{u}_{it}) \hat{\lambda}_i \hat{\Phi}_{i,k} - \frac{1}{N} \sum_{i=1}^N \mathbf{f}_{it}(0) (\mathbf{H}'_0)^{-1} \lambda_{0i} \Phi_{i,k} (\mathbf{H}_0)^{-1} \right\| \\
&\leq \max_t \left\| \frac{1}{N} \sum_{i=1}^N l^{(2)}(\hat{u}_{it}) \hat{\lambda}_i \hat{\Phi}_{i,k} - \frac{1}{N} \sum_{i=1}^N l_{it}^{(2)} \tilde{\lambda}_{0i} \Phi_{i,k} (\mathbf{H}_0)^{-1} \right\| + \max_t \left\| \frac{1}{N} \sum_{i=1}^N [l_{it}^{(2)} - \mathbf{f}_{it}(0)] \tilde{\lambda}_{0i} \Phi_{i,k} \right\| \cdot \|\mathbf{H}_0^{-1}\| \\
&\lesssim \max_{i,t} \|l^{(2)}(\hat{u}_{it}) - l_{it}^{(2)}\| + \max_i \|\hat{\lambda}_i - \tilde{\lambda}_i\|/h + \max_i \|\hat{\Phi}_i - \Phi_i \hat{\mathbf{H}}^{-1}\|/h + \|\mathbf{H}_0 - \mathbf{H}^{-1}\|/h + o_P(1) \\
&= O_P \left( \frac{T^{1/(2m)}}{\sqrt{Th^4}} \right) + o_P(1) = o_P(1).
\end{aligned}$$

Combining all the above results gives that  $\hat{d}_{2,k} = d_{2,k} + o_P(1)$  for all  $k \leq r$ . This completes the proof of Theorem 3.  $\square$

## A.5 Proof of Theorem 4

*Proof.* First, similar to the proof of Theorem 3, we have

$$\begin{aligned}
\max_{1 \leq t \leq T} \left\| \hat{\mathbf{A}}_t - \mathbf{A}_t \mathbf{H}_0^{-1} \right\| &= \max_{1 \leq t \leq T} \left\| \frac{1}{N} \sum_{i=1}^N l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \hat{\lambda}'_i - \frac{1}{N} \sum_{i=1}^N \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it}] \lambda'_{0i} \mathbf{H}_0^{-1} \right\| \\
&\leq \max_{1 \leq t \leq T} \left\| \frac{1}{N} \sum_{i=1}^N \left\{ l^{(2)}(\hat{u}_{it}) \hat{Z}_{it} \hat{\lambda}'_i - l_{it}^{(2)} Z_{it} \tilde{\lambda}'_{0i} \right\} \right\| + \max_{1 \leq t \leq T} \left\| \frac{1}{N} \sum_{i=1}^N \left\{ l_{it}^{(2)} Z_{it} - \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it}] \right\} \lambda'_{0i} \right\| \cdot \|\mathbf{H}_0^{-1}\| \\
&\lesssim \max_{i,t} \|l^{(2)}(\hat{u}_{it}) - l_{it}^{(2)}\| + \max_{i,t} \|\hat{Z}_{it} - Z_{it}\|/h + \max_i \|\hat{\lambda}_i - \tilde{\lambda}_i\|/h + O_P \left( \frac{T^{1/(2m)}}{T^{0.5} h^2} \right) = O_P \left( \frac{T^{1/(2m)}}{T^{0.5} h^2} \right).
\end{aligned}$$

Second, it follows that

$$\begin{aligned}
\max_{i,t} \|\hat{W}_{it} - W_{it}^*\| &= \max_{i,t} \left\| l^{(1)}(\hat{u}_{it}) \hat{Z}_{it} - \hat{\mathbf{A}}_t \hat{\Psi}' \hat{e}_{it} - [l_{it}^{(1)} Z_{it} - \mathbf{A}_t (\mathbf{H}_0)^{-1} \Psi'_0 e_{it}] \right\| \lesssim \max_{i,t} \left\| l^{(1)}(\hat{u}_{it}) - l_{it}^{(1)} \right\| \\
&\quad + \max_{i,t} \|\hat{Z}_{it} - Z_{it}\| + \max_t \left\| \hat{\mathbf{A}}_t - \mathbf{A}_t \mathbf{H}_0^{-1} \right\| + \|\hat{\Psi} - \Psi_0\| + \max_{i,t} \|\hat{e}_{it} - e_{it}\| = O_P \left( \frac{T^{1/(2m)}}{T^{0.5} h^2} \right).
\end{aligned}$$

Third, similar to the proof of Theorem 3, we have

$$\begin{aligned}
\left\| \hat{\mathbf{V}}_2 - \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^T \sum_{s \neq t}^T \mathbb{E}[W_{it}^* W_{is}^{*'}] \right\| &\leq \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ \hat{W}_{it} \hat{W}_{is}' - \mathbb{E}[W_{it}^* W_{is}^{*'}] \right\} \right\| \\
&\quad + \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1+L}^T \sum_{s=t-L}^{t-1} \left\{ \hat{W}_{it} \hat{W}_{is}' - \mathbb{E}[W_{it}^* W_{is}^{*'}] \right\} \right\| + o_P(1) \\
&\leq \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ \hat{W}_{it} \hat{W}_{is}' - W_{it}^* W_{is}^{*'} \right\} \right\| + \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1}^{T-L} \sum_{s=t+1}^{t+L} \left\{ W_{it}^* W_{is}^{*'} - \mathbb{E}[W_{it}^* W_{is}^{*'}] \right\} \right\| \\
&\quad + \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1+L}^T \sum_{s=t-L}^{t-1} \left\{ \hat{W}_{it} \hat{W}_{is}' - W_{it}^* W_{is}^{*'} \right\} \right\| + \left\| \frac{1}{NT} \sum_{i=1}^N \sum_{t=1+L}^T \sum_{s=t-L}^{t-1} \left\{ W_{it}^* W_{is}^{*'} - \mathbb{E}[W_{it}^* W_{is}^{*'}] \right\} \right\| \\
&\lesssim L \cdot \max_{i,t} \|\hat{W}_{it} - W_{it}^*\| + O_P(L/\sqrt{NT}) = O_P\left(\frac{LT^{1/(2m)}}{T^{0.5}h^2}\right) + o_P(1) = o_P(1).
\end{aligned}$$

By the proof of Theorem 2, it follows that  $\hat{\mathbf{V}}_2 = \mathbf{V}_2 + o_P(1)$ .

Finally, we can show that  $\hat{\mathbf{V}}_1 = \mathbf{V}_1 + o_P(1)$  in a similar way. This completes the proof.  $\square$

## B Some Auxiliary Lemmas

**Lemma S1.** *Let  $M$  be a generic bounded constant. Under Assumptions 1 to 3, it can be shown that*

- (i)  $\sup_{u \in \mathbb{R}} l^{(j)}(u) \cdot h^{j-1} \leq M$  for  $j = 1, \dots, 4$ ;
- (ii)

$$\mathbb{E}[l_{it}^{(1)} X_{it}] = O(h^q), \quad \mathbb{E}[l_{it}^{(1)}] = O(h^q), \quad \mathbb{E}[l_{it}^{(2)} X_{it}] = \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) X_{it}] + O(h^q), \quad \mathbb{E}[l_{it}^{(2)}] = \mathbf{f}_{it}(0) + O(h^q),$$

$$\mathbb{E}[l_{it}^{(3)} X_{it}] = -\mathbb{E}[\mathbf{f}_{it}^{(1)}(0|X_{it}) X_{it}] + O(h^{q-1}), \quad \mathbb{E}[l_{it}^{(2)} l_{it}^{(1)} X_{it}] = (\tau - 0.5) \cdot \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) X_{it}] + o(1),$$

$$\mathbb{E}[l_{it}^{(2)} l_{is}^{(1)} X_{it}] = \tau \mathbb{E}[\mathbf{f}_{it}(0|X_{it}) Z_{it}] - \mathbb{E}\left[\int_{-\infty}^0 \mathbf{f}_{i,ts}(0, v|X_{it}, X_{is}) dv \cdot Z_{it}\right] + o(1),$$

$$\mathbb{E}\left[\left(l_{it}^{(1)}\right)^2\right] = \tau(1 - \tau) + o(1), \quad \mathbb{E}\left[l_{it}^{(1)} l_{is}^{(1)}\right] = \mathbb{E}[\mathbf{1}\{u_{it} \leq 0, u_{is} \leq 0\}] - \tau^2 + o(1),$$

$$\mathbb{E}\left[\left(l_{it}^{(1)}\right)^2 Z_{it} Z_{it}'\right] = \tau(1 - \tau) \cdot \mathbb{E}[Z_{it} Z_{it}'] + o(1).$$

$$(iii) \max_{i,t} \sup_{\beta \in \mathcal{B}, \lambda_i \in \mathcal{A}} \mathbb{E}[l^{(3)}(Y_{it} - \beta' X_{it} - \lambda_i \tilde{f}_{0t}) | X_{it}] \leq M \text{ almost surely.}$$

*Proof.* The proof of the above results follows from standard calculations of nonparametric kernel estimators, and can be found in Horowitz (1998) or Galvao and Kato (2016). Thus, it is omitted.  $\square$

## References

- Galvao, A. F. and K. Kato (2016). Smoothed quantile regression for panel data. *Journal of Econometrics* 193(1), 92–112.
- Horowitz, J. L. (1998). Bootstrap methods for median regression models. *Econometrica*, 1327–1351.
- Kato, K., A. F. Galvao, and G. V. Montes-Rojas (2012). Asymptotics for panel quantile regression models with individual effects. *Journal of Econometrics* 170(1), 76–91.
- van der Vaart, A. and J. Wellner (1996). *Weak Convergence and Empirical Processes: With Applications to Statistics*. Springer Science & Business Media.
- Yoshihara, K.-I. (1978). Moment inequalities for mixing sequences. *Kodai Mathematical Journal* 1(2), 316–328.