

**Supplementary Appendix to “A likelihood-based approximate solution to the incidental parameter problem in dynamic nonlinear models with multiple effects”**

Manuel Arellano and Jinyong Hahn

## A Some Auxiliary Lemmas

Throughout this appendix, we will let  $F \equiv (F_1, \dots, F_n)$  denote the collection of (marginal) distribution functions of  $x_{it}$  and  $\widehat{F} \equiv (\widehat{F}_1, \dots, \widehat{F}_n)$ , where  $\widehat{F}_i$  denotes the empirical distribution function for the  $i$ -th observation. Define  $F(\epsilon) \equiv F + \epsilon\sqrt{T}(\widehat{F} - F)$  for  $\epsilon \in [0, T^{-1/2}]$ , and  $\Delta_{iT} \equiv \sqrt{T}(\widehat{F}_i - F_i)$ . We first provide a different version of Lahiri's (1992) Lemma 5.1, which is stated for bounded zero mean random variables.

**Lemma 1 (Hahn and Kuersteiner, 2004)** *Assume that  $\{W_t, t = 1, 2, \dots\}$  is a stationary, mixing sequence with  $E[W_t] = 0$  and  $E[|W_t|^{2r+\delta}] < \infty$  for any positive integer  $r$ , some  $\delta > 0$  and all  $t$ . Let  $\mathcal{A}_t = \sigma(W_t, W_{t-1}, W_{t-2}, \dots)$ ,  $\mathcal{B}_t = \sigma(W_t, W_{t+1}, W_{t+2}, \dots)$ , and  $\alpha(m) = \sup_t \sup_{A \in \mathcal{A}_t, B \in \mathcal{B}_{t+m}} |P(A \cap B) - P(A)P(B)|$ . Then, for any  $m$  such that  $1 \leq m < C(r)n$ ,*

$$E\left[\left(\sum_{i=1}^n W_i\right)^{2r}\right] \leq C(r) E\left[|W_i|^{2r+\delta}\right] \left[n^r m^{2r} + n^{2r} \alpha(m)^{\frac{\delta}{2r+\delta}}\right]$$

where  $C(r)$  is a constant that depends on  $r$ .

**Lemma 2 (Hahn and Kuersteiner, 2004)** *Suppose that, for each  $i$ ,  $\{\xi_{it}, t = 1, 2, \dots\}$  is a mixing sequence with  $E[\xi_{it}] = 0$  for all  $i, t$ . Let  $\mathcal{A}_t^i = \sigma(\xi_{it}, \xi_{it-1}, \xi_{it-2}, \dots)$ ,  $\mathcal{B}_t^i = \sigma(\xi_{it}, \xi_{it+1}, \xi_{it+2}, \dots)$ , and  $\alpha_i(m) = \sup_t \sup_{A \in \mathcal{A}_t^i, B \in \mathcal{B}_{t+k}^i} |P(A \cap B) - P(A)P(B)|$ . Assume that  $\sup_i |\alpha_i(m)| \leq Ca^m$  for some  $a$  such that  $0 < a < 1$  and some  $0 < C < \infty$ . We assume that  $\{\xi_{it}, t = 1, 2, 3, \dots\}$  are independent across  $i$ . We also assume that  $n = O(T)$ . Finally, assume that  $E[|\xi_{it}|^{6+\delta}] < \infty$  for some  $\delta > 0$ . We then have*

$$\Pr\left[\max_{1 \leq i \leq n} \left|\frac{1}{T} \sum_{t=1}^T \xi_{it}\right| > \eta\right] = o(T^{-1})$$

for every  $\eta > 0$ . Now assume that  $E[|\xi_{it}|^{10q+12+\delta}] < \infty$  for some  $\delta > 0$  and some integer  $q \geq 1$ . Then,

$$\Pr\left[\max_{1 \leq i \leq n} \left|\frac{1}{\sqrt{T}} \sum_{t=1}^T \xi_{it}\right| > \eta T^{\frac{1}{10}-v}\right] = o(T^{-q})$$

for every  $\eta > 0$  and  $0 < v < (100q + 120)^{-1}$ .

**Lemma 3 (Hahn and Kuersteiner, 2004)** *Let  $\xi(x_{it}, \phi)$  be a function indexed by the parameter  $\phi \in \Phi$  where  $\Phi$  is a convex subset of  $\mathbb{R}^p$  with  $E[\xi(x_{it}, \phi)] = 0$  for all  $i, t$  and  $\phi \in \Phi$ . Assume that there exists a function  $\mathbf{M}(x_{it})$  such that  $|\xi(x_{it}, \phi_1) - \xi(x_{it}, \phi_2)| \leq \mathbf{M}(x_{it}) \|\phi_1 - \phi_2\|$  for all  $\phi_1, \phi_2 \in \Phi$  and  $\sup_\phi |\xi(x_{it}, \phi)| \leq \mathbf{M}(x_{it})$ . For each  $i$ , let  $x_{it}$  be a  $\alpha$ -mixing process with exponentially decaying mixing coefficients  $\underline{\alpha}_i(m)$  satisfying  $\sup_i |\underline{\alpha}_i(m)| \leq Ca^m$  for some  $a$  such that  $0 < a < 1$  and some  $0 < C < \infty$ . Let  $q$  denote a positive integer such that  $q \geq \frac{p+4}{2}$ , where  $p = \dim \phi$ . We also assume that  $E[|\mathbf{M}(x_{it})|^{10q+12+\delta}] < \infty$  for some  $\delta > 0$ . Finally, assume that  $n = O(T)$ . We then have*

$\Pr \left[ \max_i \left| \frac{1}{\sqrt{T}} \sum_{t=1}^T \xi(x_{it}, \phi_i) \right| > T^{\frac{1}{10}-\nu} \right] = o(T^{-1})$  for  $0 < \nu < (100q + 120)^{-1}$ . Here,  $\{\phi_i\}$  is an arbitrary nonstochastic sequence in  $\Phi$ .

**Lemma 4 (Hahn and Kuersteiner, 2004)** Assume that  $x_{it}$  satisfies Assumption 3, and let  $\xi(x_{it}, \phi)$  be a function indexed by the parameter  $\phi \in \text{int } \Phi$ , where  $\Phi$  is a convex subset of  $\mathbb{R}^p$ . For any sequence  $\phi_i \in \text{int } \Phi$ , assume  $E[\xi(x_{it}, \phi_i)] = 0$ . Further assume that  $\sup_\phi \|\xi(x_{it}, \phi)\| \leq \mathbf{M}(x_{it})$  for some  $\mathbf{M}(x_{it})$  such that  $E[\mathbf{M}(x_{it})^4] < \infty$ . Let  $\Sigma_{nT} = \sum_{i=1}^n \Sigma_{iT}^{\xi\xi}$  with  $\Sigma_{iT}^{\xi\xi} = \text{Var} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T \xi(x_{it}, \phi_i) \right)$ . Denote the smallest eigenvalue of  $\Sigma_{iT}^{\xi\xi}$  by  $\lambda_{iT}^\xi$ , and assume that  $\inf_i \inf_T \lambda_{iT}^\xi > 0$ . Then,

$$\frac{1}{\sqrt{nT}} \sum_{i=1}^n \sum_{t=1}^T \xi(x_{it}, \phi_i) \Rightarrow N(0, f^{\xi\xi}), \quad \text{and} \quad \sup_i \left\| \Sigma_{iT}^{\xi\xi} - f_i^{\xi\xi} \right\| \rightarrow 0,$$

where  $f^{\xi\xi} \equiv \lim n^{-1} \sum_{i=1}^n f_i^{\xi\xi}$  and  $f_i^{\xi\xi} \equiv \sum_{j=-\infty}^{\infty} E[\xi(x_{it}, \phi_i) \xi(x_{it-j}, \phi_i)']$ .

**Lemma 5** Let  $k_{it} = k(x_{it}; \theta, \gamma_i(\theta))$  and  $\widehat{k}_{it} = k(x_{it}; \theta, \widehat{\gamma}_i(\theta))$  where  $x_{it}$  satisfies Assumption 3,  $k$  satisfies Assumption 4 and  $\widehat{\theta}, \widehat{\gamma}_i$  are defined in (1). Assume that  $E[k_{it}] = 0$  for  $i, t$ . Let  $f_i^{kk} \equiv \sum_{l=-\infty}^{\infty} E[k_{it} k'_{it-l}]$  and  $f^{kk} \equiv \lim_{n \rightarrow \infty} n^{-1} \sum_{i=1}^n f_i^{kk}$ . Then,

$$\sup_\theta \left| \frac{1}{n} \sum_{i=1}^n \left( \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=\max(1,l)}^{\min(T,T+l)} \widehat{k}_{it} \widehat{k}'_{it-l} \right) - f^{kk} \right| = o_p(1),$$

where  $m, T \rightarrow \infty$  such that  $m = o(T^{2/5})$ .

**Proof.** The proof is almost identical to a similar result found in Hahn and Kuersteiner (2004).

Let  $r_1 = \max(1, l)$  and  $r_2 = \min(T, T + l)$  and define  $K_{i,m} = \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} k_{it} k'_{it-l}$ .

We first show that  $\frac{1}{n} \sum_{i=1}^n K_{i,m} - f^{kk} = o_p(1)$ . This follows if  $\frac{1}{n} \sum_{i=1}^n E[K_{i,m}] - f^{kk} = o(1)$  and  $\text{Var} \left( \frac{1}{n} \sum_{i=1}^n K_{i,m} \right) = o(1)$ . Since  $f^{kk} - n^{-1} \sum_{i=1}^n f_i^{kk} = o(1)$  by definition, we first consider

$$\begin{aligned} & \left\| E[K_{i,m}] - f_i^{kk} \right\| \\ & \leq \sum_{l=-m}^m \left| \frac{r_2 - r_1 + 1}{T} w_{T,l} - 1 \right| \left\| E[k_{it} k'_{it-l}] \right\| + \sum_{|l|>m} \left\| E[k_{it} k'_{it-l}] \right\| \\ & = \sum_{l=-m}^m \left| \frac{T - |l|}{T} - \frac{T - |l|}{T} (1 - w_{T,l}) - 1 \right| \left\| E[k_{it} k'_{it-l}] \right\| + \sum_{|l|>m} \left\| E[k_{it} k'_{it-l}] \right\| \\ & = \sum_{l=-m}^m \left| \frac{1 - |l|}{T} - \frac{T - |l|}{T} \frac{|l|}{m+1} \right| \left\| E[k_{it} k'_{it-l}] \right\| + \sum_{|l|>m} \left\| E[k_{it} k'_{it-l}] \right\| \\ & \leq \sum_{l=-m}^m \left( \frac{|l|}{T} + \frac{T - |l|}{T} \frac{|l|}{m+1} \right) \left\| E[k_{it} k'_{it-l}] \right\| + \sum_{|l|>m} \left\| E[k_{it} k'_{it-l}] \right\| \\ & \leq \sum_{l=-m}^m \left( \frac{1}{T} + \frac{1}{m} \right) |l| \left\| E[k_{it} k'_{it-l}] \right\| + \sum_{|l|>m} \left\| E[k_{it} k'_{it-l}] \right\| \\ & \leq \sum_{l=-m}^m c_1 \left( \frac{1}{T} + \frac{1}{m} \right) |l| \left( a^{\frac{\delta}{2+\delta}} \right)^{|l|} + \left( a^{\frac{\delta}{2+\delta}} \right)^m c_2 \sum_{l=1}^{\infty} \left( a^{\frac{\delta}{2+\delta}} \right)^l \rightarrow 0 \text{ as } m, T \rightarrow \infty \end{aligned}$$

where the last inequality follows from Condition 3 and the fact that

$$|E[k_{it,j_1}k_{it-l,j_2}]| \leq 8 \left( E[|k_{it,j_1}|^{2+\delta}] \right)^{\frac{1}{2+\delta}} \left( E[|k_{it-l,j_2}|^{2+\delta}] \right)^{\frac{1}{2+\delta}} \left( a^{\frac{\delta}{2+\delta}} \right)^{|l|}$$

for any two elements  $k_{it,j_1}$  and  $k_{it-l,j_2}$  of  $k_{it}$  and  $k_{it-l}$  for some  $\delta > 0$ , which can be proved by Corollary A.2 of Hall and Heyde (1980). Since the bound on  $\|E[K_{i,m}] - f_i^{kk}\|$  is uniform it therefore follows that  $\frac{1}{n} \sum_{i=1}^n E[K_{i,m}] - f^{kk} = o(1)$ .

Next we show that

$$\left\| \text{Var} \left( \frac{1}{n} \sum_{i=1}^n K_{i,m} \right) \right\| \leq \frac{1}{n^2} \sum_{i=1}^n \|\text{Var}(K_{i,m})\| = o(1).$$

To show this we may assume without loss of generality that  $k_{it}$  is scalar. The variance can then be evaluated as

$$\begin{aligned} & \text{Var}(K_{i,m}) \\ &= \frac{1}{T^2} \sum_{l_1, l_2 = -m}^m w_{T, l_1} w_{T, l_2} \sum_{t_1, t_2 = r_1}^{r_2} (E[k_{it_1} k_{it-l_1} k_{it_2} k_{it_2-l_2}] - E[k_{it_1} k_{it-l_1}] E[k_{it_2} k_{it_2-l_2}]) \\ &= \frac{1}{T^2} \sum_{l_1, l_2 = -m}^m w_{T, l_1} w_{T, l_2} \sum_{t_1, t_2 = r_1}^{r_2} (E[k_{it_1} k_{it_2}] E[k_{it-l_1} k_{it_2-l_2}] + E[k_{it_1} k_{it_2-l_2}] E[k_{it_2} k_{it-l_1}]) \\ &\quad + \frac{1}{T^2} \sum_{l_1, l_2 = -m}^m w_{T, l_1} w_{T, l_2} \sum_{t_1, t_2 = r_1}^{r_2} \text{Cum}(k_{it_1} k_{it-l_1} k_{it_2} k_{it_2-l_2}) \\ &= O(1) \end{aligned}$$

such that  $\text{Var}(K_{i,m})$  is uniformly bounded in  $i$ . It now follows that  $\frac{1}{n} \sum_{i=1}^n K_{i,m} - f^{kk} = o_p(1)$  by Markov's inequality.

Next we turn to showing that

$$\frac{1}{n} \sum_{i=1}^n \frac{1}{T} \sum_{l=-m}^m w_{T, l} \sum_{t=r_1}^{r_2} (\widehat{k}_{it} \widehat{k}'_{it-l} - k_{it} k'_{it-l}) = o_p(1).$$

We use the decomposition

$$\begin{aligned} & \frac{1}{T} \sum_{l=-m}^m w_{T, l} \sum_{t=r_1}^{r_2} (\widehat{k}_{it} \widehat{k}'_{it-l} - k_{it} k'_{it-l}) \\ &= \frac{1}{T} \sum_{l=-m}^m w_{T, l} \sum_{t=r_1}^{r_2} (\widehat{k}_{it} - k_{it}) (\widehat{k}_{it-l} - k_{it-l})' \\ &\quad + \frac{1}{T} \sum_{l=-m}^m w_{T, l} \sum_{t=r_1}^{r_2} k_{it} (\widehat{k}_{it-l} - k_{it-l})' + \frac{1}{T} \sum_{l=-m}^m w_{T, l} \sum_{t=r_1}^{r_2} (\widehat{k}_{it} - k_{it}) k'_{it-l} \end{aligned}$$

We first consider the term  $\frac{1}{T} \sum_{l=-m}^m w_{T, l} \sum_{t=r_1}^{r_2} (\widehat{k}_{it} - k_{it}) k'_{it-l}$ . Use a first order Taylor approximation to

$$\widehat{k}_{it} - k_{it} = k_{it}^\theta (\widehat{\theta} - \theta) + k_{it}^\gamma (\widehat{\gamma}_i - \gamma_{i0})$$

where  $k_{it}^\theta = \partial k(x_{it}; \tilde{\theta}, \tilde{\gamma}_i) / \partial \theta'$  and  $k_{it}^\gamma = \partial k(x_{it}; \tilde{\theta}', \tilde{\gamma}'_i) / \partial \gamma$  with  $\tilde{\theta}, \tilde{\gamma}_i, \tilde{\theta}', \tilde{\gamma}'_i$  such that  $\|\tilde{\theta} - \theta_0\| \leq \|\widehat{\theta} - \theta_0\|$ ,  $\|\tilde{\theta}' - \theta_0\| \leq \|\widehat{\theta} - \theta_0\|$ , etc. by the multivariate version of the mean value theorem. Note

that each row of  $\partial k(x_{it}; \tilde{\theta}, \tilde{\gamma}_i) / \partial \theta'$  needs to be evaluated at a different  $\tilde{\theta}$  but in slight abuse of notation we do not make this explicit. Then

$$\begin{aligned} & \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \text{vec} \left[ \left( \hat{k}_{it} - k_{it} \right) k'_{it-l} \right] \\ &= \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \left( k_{it-l} \otimes k_{it}^\theta \right) \left( \hat{\theta} - \theta \right) \\ & \quad + \frac{(\hat{\gamma}_i - \gamma_{i0})}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \text{vec} \left[ k_{it}^\gamma k'_{it-l} \right] \end{aligned} \quad (34)$$

and consider  $\frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} (k_{it-l} \otimes k_{it}^\theta)$ . Without loss of generality assume that  $(k_{it-l} \otimes k_{it}^\theta)$  is a scalar. Then by the Cauchy-Schwartz inequality

$$\begin{aligned} \left| \frac{1}{T} \sum_{t=r_1}^{r_2} k_{it-l} k_{it}^\theta \right| &\leq \left( \frac{1}{T} \sum_{t=1}^T k_{it-l}^2 \right)^{1/2} \left( \frac{1}{T} \sum_{t=1}^T \sup_{\theta, \gamma} (\partial k(x_{it}; \theta, \gamma) / \partial \theta')^2 \right)^{1/2} \\ &\leq \left( \frac{1}{T} \sum_{t=1}^T M(x_{it-l})^2 \right)^{1/2} \left( \frac{1}{T} \sum_{t=1}^T M(x_{it})^2 \right)^{1/2} \end{aligned}$$

such that  $E \left[ \left| \frac{1}{T} \sum_{t=r_1}^{r_2} k_{it-l} k_{it}^\theta \right| \right] \leq \left( \frac{1}{T} \sum_{t=1}^T E \left[ M(x_{it-l})^2 \right] \right)^{1/2} \left( \frac{1}{T} \sum_{t=1}^T E \left[ M(x_{it})^2 \right] \right)^{1/2} = O(1)$  uniformly in  $i$ . It thus follows from the Markov inequality that

$$\frac{1}{n} \sum_{i=1}^n \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \left( k_{it-l} \otimes k_{it}^\theta \right) \left( \hat{\theta} - \theta \right) = O_p(m/T).$$

We now turn to the second term in (34). Noting that

$$T^{2/5} \max_i |\hat{\gamma}_i - \gamma_{i0}| = o_p(1)$$

by Lemma (7), we obtain

$$\begin{aligned} & \left| \frac{1}{n} \sum_{i=1}^n \frac{(\hat{\gamma}_i - \gamma_{i0})}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \text{vec} \left[ k_{it}^\gamma k'_{it-l} \right] \right| \\ &\leq \frac{1}{T^{7/5}} T^{2/5} \max_i |\hat{\gamma}_i - \gamma_{i0}| \cdot \frac{1}{n} \sum_{i=1}^n \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \left\| \text{vec} \left[ k_{it}^\gamma k'_{it-l} \right] \right\| \\ &\leq o_p \left( T^{-7/5} \right) \cdot \frac{1}{n} \sum_{i=1}^n \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \text{vec} \left[ M_{it} M'_{it-l} \right] \\ &= o_p \left( T^{-7/5} \right) \sum_{l=-m}^m \left( 1 - \frac{|l|}{m+1} \right) (T - |l|) \\ &= o_p \left( T^{-7/5} \right) O(Tm) \\ &= o_p \left( \frac{m}{T^{2/5}} \right) \end{aligned}$$

We now turn to

$$\begin{aligned}
& \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \text{vec} \left( \widehat{k}_{it} - k_{it} \right) \left( \widehat{k}_{it-l} - k_{it-l} \right)' \\
= & \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \left( k_{it-l}^\theta \otimes k_{it}^\theta \right) \text{vec} \left( \widehat{\theta} - \theta \right) \left( \widehat{\theta} - \theta \right)' \\
& + \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \left( k_{it-l}^\theta \otimes k_{it}^\gamma \right) \left( \widehat{\gamma}_i - \gamma_{i0} \right) \text{vec} \left( \widehat{\theta} - \theta \right)' \\
& + \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \left( k_{it-l}^\gamma \otimes k_{it}^\theta \right) \left( \widehat{\theta} - \theta \right) \left( \widehat{\gamma}_i - \gamma_{i0} \right) \\
& + \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=r_1}^{r_2} \left( \widehat{\gamma}_i - \gamma_{i0} \right)^2 \text{vec} \left( k_{it}^\gamma k_{it-l}^{\gamma'} \right)
\end{aligned}$$

All the terms on the RHS are  $o_p(m/T^{2/5})$  by similar arguments. ■

**Lemma 6** *Under Assumptions 1, 2, 3, 4, 5, 6, and 7, we have*

- (i)  $n^{-1} \sum_{i=1}^n \bar{\mathcal{I}}_i - \mathcal{I} = o_p(1)$ ;
- (ii)  $\max_i \left\| \frac{1}{T} \sum_{t=1}^T V_{it}^{\gamma_i}(\theta_0, \widehat{\gamma}_i(\theta_0)) - E[V_i^{\gamma_i}] \right\| = o_p(1)$ ;
- (iii)  $\max_i \left\| \frac{1}{T} \sum_{t=1}^T V_{it}^\theta(\theta_0, \widehat{\gamma}_i(\theta_0)) - E[V_i^\theta] \right\| = o_p(1)$ ;
- (iv)  $\max_i \left\| \frac{1}{T} \sum_{t=1}^T U_{it}^{\gamma_i \gamma_i}(\theta_0, \widehat{\gamma}_i(\theta_0)) - E[U_i^{\gamma_i \gamma_i}] \right\| = o_p(1)$ ;
- (v)  $\max_i \left\| \frac{1}{T} \sum_{t=1}^T V_{it}^{\gamma_i \gamma_i}(\theta_0, \widehat{\gamma}_i(\theta_0)) - E[V_i^{\gamma_i \gamma_i}] \right\| = o_p(1)$ .

**Proof.** We only prove the first result. The rest can be proved using the same argument as in Hahn and Kuersteiner (2004). Note that

$$\max_i \|\bar{\mathcal{I}}_i - \mathcal{I}_i\| \leq \sup_i E[\|M(x_{it})\|] \left( \|\bar{\theta} - \theta\| + \max_i |\widehat{\gamma}_i - \gamma_{i0}| \right) + o_p(1).$$

Since

$$|\widehat{\gamma}_i - \gamma_{i0}| \leq \frac{1}{\sqrt{T}} |\widehat{\gamma}_i^\epsilon(0)| + \frac{1}{2T} |\widehat{\gamma}_i^{\epsilon\epsilon}(\check{\epsilon})|$$

with  $\max_i T^{-\frac{1}{10}} |\widehat{\gamma}_i^\epsilon(0)| = o_p(1)$  and  $\max_i T^{-\frac{2}{10}} |\widehat{\gamma}_i^{\epsilon\epsilon}(\check{\epsilon})| = o_p(1)$  by Lemma 14, it follows that  $\max_i \|\bar{\mathcal{I}}_i - \mathcal{I}_i\| = o_p(1)$  such that

$$n^{-1} \sum_{i=1}^n \bar{\mathcal{I}}_i - \mathcal{I} = o_p(1).$$

■

**Lemma 7 (Hahn and Kuersteiner, 2004)** *Let Assumptions 1, 2, 3, 4 and 5 be satisfied. Then*  
 $\Pr \left[ \max_i \left| \sqrt{T} (\widehat{\gamma}_i - \gamma_{i0}) \right| > T^{1/10-v} \right] = o(T^{-1})$  for  $0 < v < (100q + 120)^{-1}$ .

**Lemma 8** *Let Assumptions 1, 2, 3, 4 and 5 be satisfied. Then*  
 $\Pr \left[ \max_i \left| \sqrt{T} (\widehat{\gamma}_i(\theta_0) - \gamma_{i0}) \right| > T^{1/10-v} \right] = o(T^{-1})$  for  $0 < v < (100q + 120)^{-1}$ .

**Proof.** It can be proved in the same way as in Hahn and Kuersteiner (2004), and is omitted. ■

**Lemma 9** Let  $k_{it} = k(x_{it}; \theta_0, \gamma_{i0})$  and  $\widehat{k}_{it} = k(x_{it}; \theta_0, \widehat{\gamma}_i(\theta_0))$  where  $x_{it}$  satisfies Assumption 3,  $k_{it}$  satisfies Assumption 4 and  $\widehat{\theta}, \widehat{\gamma}_i$  are defined in (1). Assume that  $E[k_{it}] = 0$  for  $i, t$ . Let  $f_i^{kk} = \sum_{l=-\infty}^{\infty} E[k_{it}k'_{it-l}]$ . Then,  $\sup_i \left\| \sum_{l=-m}^m w_{T,l} E_{\widehat{\theta}, \widehat{\gamma}_i} [\widehat{k}_{it} \widehat{k}'_{it-l}] - f_i^{kk} \right\| = o_p(1)$ , where  $m, T \rightarrow \infty$  such that  $m = o(T^{2/5})$ .

**Proof.** For notational simplicity, we may assume without loss of generality that  $k_{it}$  is scalar. Let  $K_{i,m} = \sum_{l=-m}^m w_{T,l} E[k_{it}k_{it-l}]$ . We first consider

$$\begin{aligned} & \left\| K_{i,m} - f_i^{kk} \right\| \\ & \leq \sum_{l=-m}^m \left| \frac{r_2 - r_1 + 1}{T} w_{T,l} - 1 \right| \|E[k_{it}k_{it-l}]\| + \sum_{|l|>m} \|E[k_{it}k_{it-l}]\| \\ & \leq \sum_{l=-m}^m \left( \frac{1}{T} + \frac{1}{m} \right) |l| \|E[k_{it}k_{it-l}]\| + \sum_{|l|>m} \|E[k_{it}k_{it-l}]\| \\ & \leq \sum_{l=-m}^m c_1 \left( \frac{1}{T} + \frac{1}{m} \right) |l| \left( a^{\frac{\delta}{2+\delta}} \right)^{|l|} + \left( a^{\frac{\delta}{2+\delta}} \right)^m c_2 \sum_{l=1}^{\infty} \left( a^{\frac{\delta}{2+\delta}} \right)^l \rightarrow 0 \text{ as } m, T \rightarrow \infty \end{aligned}$$

where the last inequality follows from Assumption 3 and the fact that, for any two elements  $k_{it,j_1}$  and  $k_{it-l,j_2}$  of  $k_{it}$  and  $k_{it-l}$ , it follows from Corollary A.2 of Hall and Heyde (1980) that

$$|E[k_{it,j_1}k_{it-l,j_2}]| \leq 8 \left( E[|k_{it,j_1}|^{2+\delta}] \right)^{\frac{1}{2+\delta}} \left( E[|k_{it-l,j_2}|^{2+\delta}] \right)^{\frac{1}{2+\delta}} \left( a^{\frac{\delta}{2+\delta}} \right)^{|l|}$$

for some  $\delta > 0$ . It follows that

$$\sup_i \left\| K_{i,m} - f_i^{kk} \right\| = o(1).$$

Now, let

$$\widehat{K}_{i,m} = \sum_{l=-m}^m w_{T,l} E_{\widehat{\theta}, \widehat{\gamma}_i} [\widehat{k}_{it} \widehat{k}_{it-l}] = \sum_{l=-m}^m w_{T,l} \int \widehat{k}_{it} \widehat{k}_{it-l} \widehat{p}_{i,t,l} d(x_{it}, x_{it-l}).$$

where

$$\begin{aligned} \widehat{k}_{it} & \equiv k_{it}(x_{it}; \theta_0, \widehat{\gamma}_i(\theta_0)) \\ \widehat{p}_{i,t,l} & \equiv p_{i,t,l}(x_{it}, x_{it-l}; \widehat{\theta}, \widehat{\gamma}_i) \end{aligned}$$

Here,  $p_{i,t,l}(x_{it}, x_{it-l}; \theta, \gamma_i)$  denotes the joint density of  $(x_{it}, x_{it-l})$ . Consider  $\widehat{K}_{i,m} - K_{i,m}$

$$\widehat{K}_{i,m} - K_{i,m} = \sum_{l=-m}^m w_{T,l} \int \left( \widehat{k}_{it} \widehat{k}_{it-l} \widehat{p}_{i,t,l} - k_{it} k_{it-l} p_{i,t,l} \right) d(x_{it}, x_{it-l})$$

We use the mean value theorem and write

$$\begin{aligned} \widehat{k}_{it}\widehat{k}_{it-l}\widehat{p}_{it} - k_{it}k_{it-l}p_{it} &= \widetilde{k}_{it}^{\gamma}\widetilde{k}_{it-l}\widetilde{p}_{it}(\widehat{\gamma}_i(\theta_0) - \gamma_{i0}) + \widetilde{k}_{it}\widetilde{k}_{it-l}^{\gamma}\widetilde{p}_{it}(\widehat{\gamma}_i(\theta_0) - \gamma_{i0}) \\ &\quad + \widetilde{k}_{it}\widetilde{k}_{it-l}\widetilde{p}_{it}^{\theta}(\widehat{\theta} - \theta) + \widetilde{k}_{it}\widetilde{k}_{it-l}\widetilde{p}_{it}^{\gamma}(\widehat{\gamma}_i - \gamma_{i0}) \end{aligned}$$

where  $\widetilde{k}_{it}^{\theta} = \partial k(x_{it}; \widetilde{\theta}, \widetilde{\gamma}_i) / \partial \theta$ , etc. Note that we may write  $\widetilde{p}_{it}^{\theta} = \widetilde{u}_{it}^{\theta}\widetilde{p}_{it}$  and  $\widetilde{f}_{it}^{\gamma} = \widetilde{v}_{it}^{\theta}\widetilde{p}_{it}$ . By Assumptions 4 and 8, we obtain

$$\left\| \widehat{K}_{i,m} - K_{i,m} \right\| \leq m\mathbf{M} \left( \sup_i \|\widehat{\gamma}_i(\theta_0) - \gamma_{i0}\| + \|\widehat{\theta} - \theta\| + \sup_i \|\widehat{\gamma}_i - \gamma_{i0}\| \right)$$

for some finite constant  $\mathbf{M}$ , or

$$\max_i \left\| \widehat{K}_{i,m} - K_{i,m} \right\| = O_p \left( \frac{m}{T^{2/5}} \right)$$

by Lemmas 7 and 8. ■

**Lemma 10** *Let  $k_{it} = k(x_{it}; \theta_0, \gamma_{i0})$  and  $\widehat{k}_{it} = k(x_{it}; \theta_0, \widehat{\gamma}_i(\theta_0))$  where  $x_{it}$  satisfies Assumption 3,  $k_{it}$  satisfies Assumption 4 and  $\theta^*, \gamma_i^*$  are such that  $\|\theta^* - \theta\| = O_p(T^{-2/5})$  and  $\sup_i \|\gamma_i^* - \gamma_{i0}\| = O_p(T^{-2/5})$ . Then,  $\sup_i \left\| \sum_{l=-m}^m w_{T,l} E_{\theta^*, \gamma_i^*} [\widehat{k}_{it}\widehat{k}_{it-l}'] - f^{kk} \right\| = o_p(1)$ , where  $m, T \rightarrow \infty$  such that  $m = o(T^{2/5})$ .*

**Proof.** Similar to the proof of Lemma 9, and omitted. ■

**Lemma 11 (Hahn and Kuersteiner, 2004)**  $\Pr \left[ \max_{1 \leq i \leq n} \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} |\widehat{\gamma}_i(\epsilon) - \gamma_{i0}| \geq \eta \right] = o(T^{-1})$  for every  $\eta > 0$ .

**Lemma 12** *Suppose that  $K_i(\cdot; \theta_0, \gamma_i(\theta_0, \epsilon))$  is equal to*

$$\frac{\partial^{m_1+m_2}\psi(x_{it}; \theta_0, \gamma_i(\theta_0, \epsilon))}{\partial \gamma_i^m}$$

for some  $m \leq 1, \dots, 5$ . Then, for any  $\eta > 0$ , we have

$$\Pr \left[ \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} \left| \frac{1}{n} \sum_{i=1}^n \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, \epsilon)) dF_i(\epsilon) - \frac{1}{n} \sum_{i=1}^n E[K_i(x_{it}; \theta_0, \gamma_{i0})] \right| > \eta \right] = o(T^{-1})$$

and

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} \left| \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, \epsilon)) dF_i(\epsilon) - E[K_i(x_{it}; \theta_0, \gamma_{i0})] \right| > \eta \right] = o(T^{-1}).$$

Also,

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} \left| \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, \epsilon)) d\Delta_{iT} \right| > CT^{\frac{1}{10}-v} \right] = o(T^{-1})$$

for some constant  $C > 0$  and  $0 < v < (100q + 120)^{-1}$ .



**Proof.** Note that we may write

$$\begin{aligned}
& \left\| \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) dF_i(\epsilon) - \int K_i(\cdot; \theta_0, \gamma_{i0}) dF_i \right\| \\
& \leq \left\| \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) dF_i(\epsilon) - \int K_i(\cdot; \theta_0, \gamma_{i0}) dF_i(\epsilon) \right\| \\
& \quad + \left\| \int K_i(\cdot; \theta_0, \gamma_{i0}) dF_i(\epsilon) - \int K_i(\cdot; \theta_0, \gamma_{i0}) dF_i \right\| \\
& \leq \int M(x_{it}) (|\gamma_i(\theta_0, F_i(\epsilon)) - \gamma_{i0}|) d|F_i(\epsilon)| \\
& \quad + \epsilon \sqrt{T} \left\| \int K_i(\cdot; \theta_0, \gamma_{i0}) d(\widehat{F}_i - F_i) \right\|.
\end{aligned}$$

Therefore, we have

$$\begin{aligned}
& \left\| \frac{1}{n} \sum_{i=1}^n \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) dF_i(\epsilon) - \int K_i(\cdot; \theta_0, \gamma_{i0}) dF_i \right\| \\
& \leq \left( \frac{1}{n} \sum_{i=1}^n (\gamma_i(\theta_0, F_i(\epsilon)) - \gamma_{i0})^2 \right)^{1/2} \left( \frac{1}{n} \sum_{i=1}^n \left( E[M(x_{it})] + \frac{1}{T} \sum_{t=1}^T M(x_{it}) \right)^2 \right)^{1/2} \\
& \quad + \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{1}{T} \sum_{t=1}^T K_i(x_{it}; \theta_0, \gamma_{i0}) - E[K_i(x_{it}; \theta_0, \gamma_{i0})] \right) \right\|,
\end{aligned}$$

the RHS of which can be bounded by using Lemmas 2 and 11 in absolute value by some  $\eta > 0$  with probability  $1 - o(T^{-1})$ .

Because

$$\begin{aligned}
& \left| \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) dF_i(\epsilon) - E[K_i(x_{it}; \theta_0, \gamma_{i0})] \right| \\
& \leq |\gamma_i(\theta_0, F_i(\epsilon)) - \gamma_{i0}| \cdot \left( E[M(x_{it})] + \frac{1}{T} \sum_{t=1}^T M(x_{it}) \right) \\
& \quad + \left| \frac{1}{T} \sum_{t=1}^T M(x_{it}) - E[M(x_{it})] \right|,
\end{aligned}$$

we can bound

$$\max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} \left| \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) dF_i(\epsilon) - E[K_i(x_{it}; \theta_0, \gamma_{i0})] \right|$$

in absolute value by some  $\eta > 0$  with probability  $1 - o(T^{-1})$ .

Using Lemmas 3, we can also show that

$$\max_i \left| \int K_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) d\Delta_{iT} \right|$$

can be bounded by in absolute value by  $CT^{\frac{1}{10}-v}$  for some constant  $C > 0$  and  $v$  such that  $0 \leq v < \frac{1}{160}$  with probability  $1 - o(T^{-1})$ . ■

## B Consistency

Let

$$\widehat{G}_{(i)}(\theta, \gamma) \equiv \frac{1}{T} \sum_{t=1}^T \psi(x_{it}; \theta, \gamma), \quad G_{(i)}(\theta, \gamma) \equiv E[\psi(x_{it}; \theta, \gamma)]$$

where  $\widehat{\gamma}_i(\theta) \equiv \operatorname{argmax}_a \sum_{t=1}^T \psi(x_{it}; \theta, a)$ .

**Lemma 13 (Hahn and Kuersteiner, 2004)** *For all  $\eta > 0$ , it follows that*

$$\Pr \left[ \max_{1 \leq i \leq n} \sup_{(\theta, \gamma)} \left| \widehat{G}_{(i)}(\theta, \gamma) - G_{(i)}(\theta, \gamma) \right| \geq \eta \right] = o(T^{-1})$$

Recall now that  $\widetilde{\theta}$  is a solution to (19).

**Theorem 11**  $\Pr \left[ \left| \widetilde{\theta} - \theta_0 \right| \geq \eta \right] = o(T^{-1})$  for every  $\eta > 0$ .

**Proof.** Let  $\eta$  be given, and let  $\varepsilon \equiv \inf_i \left[ G_{(i)}(\theta_0, \gamma_{i0}) - \sup_{\{(\theta, \gamma): |(\theta, \gamma) - (\theta_0, \gamma_{i0})| > \eta\}} G_{(i)}(\theta, \gamma) \right] > 0$ . Because of Condition 1, we have

$$\left| \frac{1}{T} B_n(\theta) \right| \leq \frac{1}{6} \varepsilon$$

with probability equal to  $1 - o\left(\frac{1}{T}\right)$ . Also, because of Lemma 13, we have

$$\max_{1 \leq i \leq n} \sup_{(\theta, \gamma)} \left| \widehat{G}_{(i)}(\theta, \gamma) - G_{(i)}(\theta, \gamma) \right| \leq \frac{1}{6} \varepsilon$$

with probability equal to  $1 - o\left(\frac{1}{T}\right)$ . It follows that

$$\begin{aligned} & \max_{|\theta - \theta_0| > \eta, \gamma_1, \dots, \gamma_n} n^{-1} \sum_{i=1}^n \widehat{G}_{(i)}(\theta, \gamma_i) - \frac{1}{T} B_n(\theta) \\ & \leq \max_{|(\theta, \gamma_i) - (\theta_0, \gamma_{i0})| > \eta} n^{-1} \sum_{i=1}^n \widehat{G}_{(i)}(\theta, \gamma_i) - \frac{1}{T} B_n(\theta) \\ & \leq \max_{|(\theta, \gamma_i) - (\theta_0, \gamma_{i0})| > \eta} n^{-1} \sum_{i=1}^n \widehat{G}_{(i)}(\theta, \gamma_i) + \frac{1}{6} \varepsilon \\ & \leq \max_{|(\theta, \gamma_i) - (\theta_0, \gamma_{i0})| > \eta} n^{-1} \sum_{i=1}^n G_{(i)}(\theta, \gamma_i) + \frac{1}{3} \varepsilon \\ & \leq n^{-1} \sum_{i=1}^n G_{(i)}(\theta_0, \gamma_{i0}) - \frac{2}{3} \varepsilon \\ & \leq n^{-1} \sum_{i=1}^n \widehat{G}_{(i)}(\theta_0, \gamma_{i0}) - \frac{1}{T} B_n(\theta_0) - \frac{1}{3} \varepsilon \end{aligned}$$

Because

$$\max_{\theta, \gamma_1, \dots, \gamma_n} n^{-1} \sum_{i=1}^n \widehat{G}_{(i)}(\theta, \gamma_i) - \frac{1}{T} B_n(\theta) \geq n^{-1} \sum_{i=1}^n \widehat{G}_{(i)}(\theta_0, \gamma_{i0}) - \frac{1}{T} B_n(\theta_0)$$

by definition, we can conclude that  $\Pr \left[ \left| \bar{\theta} - \theta_0 \right| \geq \eta \right] = o(T^{-1})$ . ■

**Theorem 12 (Hahn and Kuersteiner, 2004)**  $\Pr [\max_{1 \leq i \leq n} |\widehat{\gamma}_i - \gamma_{i0}| \geq \eta] = o(T^{-1})$

**Theorem 13** *Let  $\bar{\theta}$  be such that  $\Pr [|\bar{\theta} - \theta_0| \geq \eta] = o(T^{-1})$  for every  $\eta > 0$ . Then,*

$$\Pr \left[ \max_{1 \leq i \leq n} |\widehat{\gamma}_i(\bar{\theta}) - \gamma_{i0}| \geq \eta \right] = o(T^{-1})$$

for every  $\eta > 0$ .

**Proof.** We first prove that

$$T \Pr \left[ \max_{1 \leq i \leq n} \sup_{\gamma} \left| \widehat{G}_{(i)}(\bar{\theta}, \gamma) - G_{(i)}(\theta_0, \gamma) \right| \geq \eta \right] = o(1) \quad (35)$$

for every  $\eta > 0$ . Note that

$$\begin{aligned} & \max_{1 \leq i \leq n} \sup_{\gamma} \left| \widehat{G}_{(i)}(\bar{\theta}, \gamma) - G_{(i)}(\theta_0, \gamma) \right| \\ & \leq \max_{1 \leq i \leq n} \sup_{\gamma} \left| \widehat{G}_{(i)}(\bar{\theta}, \gamma) - G_{(i)}(\bar{\theta}, \gamma) \right| + \max_{1 \leq i \leq n} \sup_{\gamma} \left| G_{(i)}(\bar{\theta}, \gamma) - G_{(i)}(\theta_0, \gamma) \right| \\ & \leq \max_{1 \leq i \leq n} \sup_{(\theta, \gamma)} \left| \widehat{G}_{(i)}(\theta, \gamma) - G_{(i)}(\theta, \gamma) \right| + \max_{1 \leq i \leq n} E[M(x_{it})] \cdot |\bar{\theta} - \theta_0|. \end{aligned}$$

Therefore,

$$\begin{aligned} & T \Pr \left[ \max_{1 \leq i \leq n} \sup_{\gamma} \left| \widehat{G}_{(i)}(\bar{\theta}, \gamma) - G_{(i)}(\theta_0, \gamma) \right| \geq \eta \right] \\ & \leq T \Pr \left[ \max_{1 \leq i \leq n} \sup_{(\theta, \gamma)} \left| \widehat{G}_{(i)}(\theta, \gamma) - G_{(i)}(\theta, \gamma) \right| \geq \frac{\eta}{2} \right] \\ & \quad + T \Pr \left[ |\bar{\theta} - \theta_0| \geq \frac{\eta}{2(1 + \max_{1 \leq i \leq n} E[M(x_{it})])} \right] \\ & = o(1) \end{aligned}$$

by Lemma 13 and Theorem 11.

We now get back to the proof of Theorem 13. It suffices to prove that

$$T \Pr \left[ \max_{1 \leq i \leq n} |\widehat{\gamma}_i(\bar{\theta}) - \gamma_{i0}| \geq \eta \right] = o(1)$$

for every  $\eta > 0$ . Let  $\eta$  be given, and let  $\varepsilon \equiv \inf_i \left[ G_{(i)}(\theta_0, \gamma_{i0}) - \sup_{\{\gamma_i: |\gamma_i - \gamma_{i0}| > \eta\}} G_{(i)}(\theta_0, \gamma_i) \right] > 0$ . Condition on the event

$$\max_{1 \leq i \leq n} \sup_{\gamma} \left| \widehat{G}_{(i)}(\bar{\theta}, \gamma) - G_{(i)}(\theta_0, \gamma) \right| \leq \frac{1}{3} \varepsilon,$$

which has a probability equal to  $1 - o\left(\frac{1}{T}\right)$  by (35). We then have

$$\max_{|\gamma_i - \gamma_{i0}| > \eta} \widehat{G}_{(i)}(\bar{\theta}, \gamma_i) < \max_{|\gamma_i - \gamma_{i0}| > \eta} G_{(i)}(\theta_0, \gamma_i) + \frac{1}{3}\varepsilon < G_{(i)}(\theta_0, \gamma_{i0}) - \frac{2}{3}\varepsilon < \widehat{G}_{(i)}(\bar{\theta}, \gamma_{i0}) - \frac{1}{3}\varepsilon$$

This is inconsistent with  $\widehat{G}_{(i)}(\bar{\theta}, \widehat{\gamma}_i(\bar{\theta})) \geq \widehat{G}_{(i)}(\bar{\theta}, \gamma_{i0})$ , and therefore,  $|\widehat{\gamma}_i(\bar{\theta}) - \gamma_{i0}| \leq \eta$  for every  $i$ . ■

**Corollary 1**  $\Pr \left[ \max_{1 \leq i \leq n} |\widehat{\gamma}_i(\bar{\theta}) - \gamma_{i0}| \geq \eta \right] = o(T^{-1})$ .

**Proof.** It follows from Theorem 13 above. ■

## C Justification of (26)

We analyze the asymptotic distribution of

$$\frac{1}{nT} \sum_{i=1}^n \sum_{t=1}^T U(x_{it}; \theta_0, \hat{\gamma}_i(\theta_0)) \quad (36)$$

Let  $F \equiv (F_1, \dots, F_n)$  denote the collection of (marginal) distribution functions of  $x_{it}$ . Let  $\hat{F} \equiv (\hat{F}_1, \dots, \hat{F}_n)$ , where  $\hat{F}_i$  denotes the empirical distribution function for the observation  $i$ . Define  $F(\epsilon) \equiv F + \epsilon\sqrt{T}(\hat{F} - F)$  for  $\epsilon \in [0, T^{-1/2}]$ . For each fixed  $\theta$  and  $\epsilon$ , let  $\gamma_i(\theta, F_i(\epsilon))$  be the solution to the estimating equation

$$0 = \int V_i[\theta, \gamma_i(\theta, F_i(\epsilon))] dF_i(\epsilon),$$

and let  $\mu(F(\epsilon))$  be the solution to the estimating equation

$$0 = \sum_{i=1}^n \int (U_i(x_{it}; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) - \mu(F(\epsilon))) dF_i(\epsilon).$$

Note that  $\mu(F(0)) = 0$ , and

$$\begin{aligned} \mu(\hat{F}) &\equiv \mu\left(F\left(\frac{1}{\sqrt{T}}\right)\right) = \frac{1}{n} \sum_{i=1}^n U_i\left(x_{it}; \theta_0, \gamma_i\left(\theta_0, F_i\left(\frac{1}{\sqrt{T}}\right)\right)\right) \\ &= \frac{1}{nT} \sum_{i=1}^n \sum_{t=1}^T U(x_{it}; \theta_0, \hat{\gamma}_i(\theta_0)). \end{aligned}$$

By a Taylor series expansion, we have

$$\mu(\hat{F}) - \mu(F) = \frac{1}{\sqrt{T}} \mu^\epsilon(0) + \frac{1}{2} \left(\frac{1}{\sqrt{T}}\right)^2 \mu^{\epsilon\epsilon}(0) + \frac{1}{6} \left(\frac{1}{\sqrt{T}}\right)^3 \mu^{\epsilon\epsilon\epsilon}(\tilde{\epsilon}), \quad (37)$$

where  $\mu^\epsilon(\epsilon) \equiv d\mu(F(\epsilon))/d\epsilon$ ,  $\mu^{\epsilon\epsilon}(\epsilon) \equiv d^2\mu(F(\epsilon))/d\epsilon^2$ , ..., and  $\tilde{\epsilon}$  is somewhere in between 0 and  $T^{-1/2}$ . It is shown later in Appendix C.2 that the last term is of order  $o_p(1)$ . We will therefore work with the expansion

$$\sqrt{nT} \left( \mu(\hat{F}) - \mu(F) \right) = \sqrt{nT} \frac{1}{\sqrt{T}} \mu^\epsilon(0) + \sqrt{nT} \frac{1}{2} \left(\frac{1}{\sqrt{T}}\right)^2 \mu^{\epsilon\epsilon}(0) + o_p(1). \quad (38)$$

The expansion (26) follows from combining (38) with (44) and (47) below.

### C.1 Details of Expansion (37)

#### C.1.1 $\mu^\epsilon(0)$

In order to obtain (44) and (47), we let

$$h_i(\cdot, \epsilon) \equiv U_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) - \mu(F(\epsilon)) \quad (39)$$

The first order condition may be written as

$$0 = \frac{1}{n} \sum_{i=1}^n \int h_i(\cdot, \epsilon) dF_i(\epsilon) \quad (40)$$

Differentiating repeatedly with respect to  $\epsilon$ , we obtain

$$0 = \frac{1}{n} \sum_{i=1}^n \int \frac{dh_i(\cdot, \epsilon)}{d\epsilon} dF_i(\epsilon) + \frac{1}{n} \sum_{i=1}^n \int h_i(\cdot, \epsilon) d\Delta_{iT} \quad (41)$$

$$0 = \frac{1}{n} \sum_{i=1}^n \int \frac{d^2h_i(\cdot, \epsilon)}{d\epsilon^2} dF_i(\epsilon) + 2\frac{1}{n} \sum_{i=1}^n \int \frac{dh_i(\cdot, \epsilon)}{d\epsilon} d\Delta_{iT} \quad (42)$$

$$0 = \frac{1}{n} \sum_{i=1}^n \int \frac{d^3h_i(\cdot, \epsilon)}{d\epsilon^3} dF_i(\epsilon) + 3\frac{1}{n} \sum_{i=1}^n \int \frac{d^2h_i(\cdot, \epsilon)}{d\epsilon^2} d\Delta_{iT} \quad (43)$$

where  $\Delta_{iT} \equiv \sqrt{T}(\widehat{F}_i - F_i)$ .

Equation (41) can be rewritten as

$$\begin{aligned} 0 &= \frac{1}{n} \sum_{i=1}^n \int (U_i^{\gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) \gamma_i^\epsilon(\theta_0, F_i(\epsilon)) - \mu^\epsilon(F(\epsilon))) dF_i(\epsilon) \\ &\quad + \frac{1}{n} \sum_{i=1}^n \int (U_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) - \mu(F(\epsilon))) d\Delta_{iT} \end{aligned}$$

Evaluating this expression at  $\epsilon = 0$ , and noting that  $E[U_i^{\gamma_i}] = 0$ , we obtain

$$\mu^\epsilon(0) = \frac{1}{n} \sum_{i=1}^n \int U_i d\Delta_{iT} \quad (44)$$

### C.1.2 $\gamma_i^\epsilon$

In the  $i$ th observation,  $\gamma_i(\theta_0, F_i(\epsilon))$  solves the estimating equation

$$\int V_i(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) dF_i(\epsilon) = 0 \quad (45)$$

Differentiating the LHS with respect to  $\epsilon$ , we obtain

$$0 = \left( \int \frac{\partial V_i(\cdot, \theta, \epsilon)}{\partial \gamma_i'} dF_i(\epsilon) \right) \frac{\partial \gamma_i(\theta, F_i(\epsilon))}{\partial \epsilon} + \int V_i(\cdot, \theta, \epsilon) d\Delta_{iT}.$$

Evaluating the expression at  $\epsilon = 0$ , we obtain gives

$$\gamma_i^\epsilon \equiv \frac{\partial \gamma_i(\theta_0, F_i(0))}{\partial \epsilon} = - \left( E \left[ \frac{\partial V_i}{\partial \gamma_i'} \right] \right)^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T V_{it} \right). \quad (46)$$

### C.1.3 $\mu^{\epsilon\epsilon}(0)$

Equation (42) can be rewritten as

$$\begin{aligned}
0 &= -\frac{1}{n} \sum_{i=1}^n \int \mu^{\epsilon\epsilon}(F(\epsilon)) dF_i(\epsilon) \\
&+ \frac{1}{n} \sum_{i=1}^n \int (U_i^{\gamma_i \gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) (\gamma_i^\epsilon(\theta_0, F_i(\epsilon)) \otimes \gamma_i^\epsilon(\theta_0, F_i(\epsilon)))) dF_i(\epsilon) \\
&+ \frac{1}{n} \sum_{i=1}^n \int (U_i^{\gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) \gamma_i^{\epsilon\epsilon}(\theta_0, F_i(\epsilon))) dF_i(\epsilon) \\
&+ \frac{2}{n} \sum_{i=1}^n \int (U_i^{\gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) \gamma_i^\epsilon(\theta_0, F_i(\epsilon)) - \mu^\epsilon(F(\epsilon))) d\Delta_{iT}
\end{aligned}$$

where  $U_i^{\gamma_i \gamma_i} \equiv \partial^2 U_i / (\partial \gamma_i \otimes \partial \gamma_i)$ . Evaluating at  $\epsilon = 0$ , and noting that  $E[U_i^{\gamma_i}] = 0$ , we obtain

$$\begin{aligned}
\mu^{\epsilon\epsilon}(0) &= \frac{1}{n} \sum_{i=1}^n E[U_i^{\gamma_i \gamma_i}] (\gamma_i^\epsilon \otimes \gamma_i^\epsilon) + \frac{2}{n} \sum_{i=1}^n \left( \int U_i^{\gamma_i}(\cdot; \theta_0, \gamma_{i0}) d\Delta_{iT} \right) \gamma_i^\epsilon(\theta_0, F_i(0)) \\
&= \frac{1}{n} \sum_{i=1}^n E[U_i^{\gamma_i \gamma_i}] \left( \left( E \left[ \frac{\partial V_i}{\partial \gamma_i'} \right] \right)^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T V_{it} \right) \otimes \left( E \left[ \frac{\partial V_i}{\partial \gamma_i'} \right] \right)^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T V_{it} \right) \right) \\
&\quad - \frac{2}{n} \sum_{i=1}^n \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T U_{it}^{\gamma_i} \right) \left( E \left[ \frac{\partial V_i}{\partial \gamma_i'} \right] \right)^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T V_{it} \right)
\end{aligned}$$

or

$$\begin{aligned}
\mu^{\epsilon\epsilon}(0) &= \frac{1}{n} \sum_{i=1}^n E[U_i^{\gamma_i \gamma_i}] \left[ \left( E \left[ \frac{\partial V_i}{\partial \gamma_i'} \right] \right)^{-1} \otimes \left( E \left[ \frac{\partial V_i}{\partial \gamma_i'} \right] \right)^{-1} \right] \left[ \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T V_{it} \right) \otimes \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T V_{it} \right) \right] \\
&\quad - \frac{2}{n} \sum_{i=1}^n \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T U_{it}^{\gamma_i} \right) \left( E \left[ \frac{\partial V_i}{\partial \gamma_i'} \right] \right)^{-1} \left( \frac{1}{\sqrt{T}} \sum_{t=1}^T V_{it} \right) \tag{47}
\end{aligned}$$

### C.1.4 $\gamma_i^{\epsilon\epsilon}$

Second order differentiation of (45) yields

$$\begin{aligned}
0 &= \left( \int \frac{\partial V_i(\cdot, \theta, \epsilon)}{\partial \gamma_i} dF_i(\epsilon) \right) \frac{\partial^2 \gamma_i(\theta, F_i(\epsilon))}{\partial \epsilon^2} \\
&+ \left( \int \frac{\partial^2 V_i(\cdot, \theta, \epsilon)}{\partial \gamma_i \otimes \partial \gamma_i} dF_i(\epsilon) \right) \left( \frac{\partial \gamma_i(\theta, F_i(\epsilon))}{\partial \epsilon} \otimes \frac{\partial \gamma_i(\theta, F_i(\epsilon))}{\partial \epsilon} \right) \\
&+ 2 \left( \int \frac{\partial V_i(\cdot, \theta, \epsilon)}{\partial \gamma_i} d\Delta_{iT} \right) \frac{\partial \gamma_i(\theta, F_i(\epsilon))}{\partial \epsilon}.
\end{aligned}$$

which characterizes  $\gamma_i^{\epsilon\epsilon}$ .

## C.2 Bounding Remainder Term in (37)

Lemma 14 below allows us to ignore the last term in equation (37).

**Lemma 14**

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} |\gamma_i^\epsilon(\epsilon)| > CT^{\frac{1}{10}-v} \right] = o(T^{-1}) \quad (48)$$

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} |\mu^\epsilon(\epsilon)| > CT^{\frac{1}{10}-v} \right] = o(T^{-1}) \quad (49)$$

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} |\gamma_i^{\epsilon\epsilon}(\epsilon)| > C \left( T^{\frac{1}{10}-v} \right)^2 \right] = o(T^{-1}) \quad (50)$$

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} |\mu^{\epsilon\epsilon}(\epsilon)| > C \left( T^{\frac{1}{10}-v} \right)^2 \right] = o(T^{-1}) \quad (51)$$

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} |\gamma_i^{\epsilon\epsilon\epsilon}(\epsilon)| > C \left( T^{\frac{1}{10}-v} \right)^3 \right] = o(T^{-1}) \quad (52)$$

$$\Pr \left[ \max_i \max_{0 \leq \epsilon \leq \frac{1}{\sqrt{T}}} |\mu^{\epsilon\epsilon\epsilon}(\epsilon)| > C \left( T^{\frac{1}{10}-v} \right)^3 \right] = o(T^{-1})$$

for some constant  $C > 0$  and  $0 < v < (100q + 120)^{-1}$ .

**Proof.** Proof is almost identical to the argument in Hahn and Kuersteiner (2004), and so only the last equality is explicitly established here. From (43), we have

$$0 = \frac{1}{n} \sum_{i=1}^n \int \frac{d^3 h_i(\cdot, \epsilon)}{d\epsilon^3} dF_i(\epsilon) + \frac{3}{n} \sum_{i=1}^n \int \frac{d^2 h_i(\cdot, \epsilon)}{d\epsilon^2} d\Delta_{iT}$$

where

$$\begin{aligned} & \frac{1}{n} \sum_{i=1}^n \int \frac{d^3 h_i(\cdot, \epsilon)}{d\epsilon^3} dF_i(\epsilon) \\ = & -\frac{1}{n} \sum_{i=1}^n \int \mu^{\epsilon\epsilon\epsilon}(F(\epsilon)) dF_i(\epsilon) \\ & + \frac{1}{n} \sum_{i=1}^n \int U_i^{\gamma_i \gamma_i \gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) (\gamma_i^\epsilon(\theta_0, F_i(\epsilon)) \otimes \gamma_i^\epsilon(\theta_0, F_i(\epsilon)) \otimes \gamma_i^\epsilon(\theta_0, F_i(\epsilon))) dF_i(\epsilon) \\ & + \frac{1}{n} \sum_{i=1}^n \int (U_i^{\gamma_i \gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) (\gamma_i^{\epsilon\epsilon}(\theta_0, F_i(\epsilon)) \otimes \gamma_i^\epsilon(\theta_0, F_i(\epsilon)))) dF_i(\epsilon) \\ & + \frac{1}{n} \sum_{i=1}^n \int (U_i^{\gamma_i \gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) (\gamma_i^{\epsilon\epsilon}(\theta_0, F_i(\epsilon)) \otimes \gamma_i^{\epsilon\epsilon}(\theta_0, F_i(\epsilon)))) dF_i(\epsilon) \\ & + \frac{1}{n} \sum_{i=1}^n \int (U_i^{\gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) \gamma_i^{\epsilon\epsilon\epsilon}(\theta_0, F_i(\epsilon))) dF_i(\epsilon) \end{aligned}$$



and

$$\begin{aligned}
& \frac{3}{n} \sum_{i=1}^n \int \frac{d^2 h_i(\cdot, \epsilon)}{d\epsilon^2} d\Delta_{iT} \\
= & -\frac{3}{n} \sum_{i=1}^n \int \mu^{\epsilon\epsilon}(F(\epsilon)) d\Delta_{iT} \\
& + \frac{3}{n} \sum_{i=1}^n \int (U_i^{\gamma_i \gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) (\gamma_i^\epsilon(\theta_0, F_i(\epsilon)) \otimes \gamma_i^\epsilon(\theta_0, F_i(\epsilon)))) d\Delta_{iT} \\
& + \frac{3}{n} \sum_{i=1}^n \int (U_i^{\gamma_i \gamma_i}(\cdot; \theta_0, \gamma_i(\theta_0, F_i(\epsilon))) (\gamma_i^\epsilon(\theta_0, F_i(\epsilon)) \otimes \gamma_i^\epsilon(\theta_0, F_i(\epsilon)))) d\Delta_{iT}
\end{aligned}$$

Combining Lemma 12 in Appendix A and (48)-(52), we can bound  $\frac{1}{n} \sum_{i=1}^n \int \frac{d^2 h_i(\cdot, \epsilon)}{d\epsilon^2} d\Delta_{iT}$  by  $C \left( T^{\frac{1}{10}-\nu} \right)^3$  with probability  $1 - o(T^{-1})$ . Likewise, using Lemmas 12, and (48)-(52) again, we can conclude that  $\frac{1}{n} \sum_{i=1}^n \int \frac{d^3 h_i(\cdot, \epsilon)}{d\epsilon^3} dF_i(\epsilon)$  is equal to  $-\mu^{\epsilon\epsilon}(F(\epsilon))$  plus terms that can all be bounded by  $\frac{1}{n} \sum_{i=1}^n \int \frac{d^2 h_i(\cdot, \epsilon)}{d\epsilon^2} d\Delta_{iT}$  by  $C \left( T^{\frac{1}{10}-\nu} \right)^3$  with probability  $1 - o(T^{-1})$ . ■

## D Proof of Theorem 3

Without loss of generality, we may write

$$2B_n(\theta) = -\frac{1}{n} \sum_{i=1}^n \ln \det \left( \frac{1}{T} H_i(\theta, \hat{\gamma}_i(\theta)) \right) + \frac{1}{n} \sum_{i=1}^n \ln \det \left( \frac{1}{T} \Upsilon_i(\theta, \hat{\gamma}_i(\theta)) \right) \quad (53)$$

We begin with the first component on the RHS of (53). By Assumption 4, each component of  $H_i(\theta, \hat{\gamma}_i(\theta))$  is bounded above by  $\sum_{t=1}^T M(x_{it})$  such that  $\sup_i E \left[ |M(x_{it})|^{10q+12+\delta} \right] < \infty$  for some integer  $q \geq (\dim(\theta) + \dim(\gamma)) / 2 + 2$  and for some  $\delta > 0$ .

**Lemma 15** *Suppose that  $A$  is an  $n \times n$  matrix. Then*

$$|\det(A)| \leq n! \cdot \max(|a_{ij}|)^n$$

**Proof.** By definition, we have

$$\det(A) = \sum (-1)^{\phi(j_1, \dots, j_n)} \prod_{i=1}^n a_{ij_i}$$

where the summation is taken over all permutations  $(j_1, \dots, j_n)$  of the set of integers  $(1, \dots, n)$  and  $\phi(j_1, \dots, j_n)$  is the number of transpositions required change  $(1, \dots, n)$  into  $(j_1, \dots, j_n)$ . Because the number of all permutations is equal to  $n!$ , we obtain the desired conclusion. ■

Using Lemma 15, we then obtain that

$$\ln \det \left( \frac{1}{T} H_i(\theta, \hat{\gamma}_i(\theta)) \right) \leq \ln r! + r \ln \left( \frac{1}{T} \sum_{t=1}^T M(x_{it}) \right)$$

where  $r = \dim(\gamma)$ . It follows that

$$\left| -\frac{1}{n} \sum_{i=1}^n \ln \det \left( \frac{1}{T} H_i(\theta, \hat{\gamma}_i(\theta)) \right) \right| \leq \ln r! + r \frac{1}{n} \sum_{i=1}^n \left| \ln \left( \frac{1}{T} \sum_{t=1}^T M(x_{it}) \right) \right|$$

By Lemma 2, we have

$$\Pr \left[ \max_{1 \leq i \leq n} \left| \frac{1}{T} \sum_{t=1}^T (M(x_{it}) - E[M(x_{it})]) \right| > \eta \right] = o(T^{-1})$$

from which we obtain<sup>12</sup>

$$\Pr \left[ \max_{1 \leq i \leq n} \left| \ln \left( \frac{1}{T} \sum_{t=1}^T M(x_{it}) \right) - \ln(E[M(x_{it})]) \right| > \eta \right] = o(T^{-1})$$

It follows that

$$\Pr \left[ \left| -\frac{1}{n} \sum_{i=1}^n \ln \det \left( \frac{1}{T} H_i(\theta, \hat{\gamma}_i(\theta)) \right) \right| > \ln r! + r \frac{1}{n} \sum_{i=1}^n \ln(E[M(x_{it})]) + \eta \right] = o(T^{-1})$$

<sup>12</sup>In addition to the Condition 4, we need to impose that the minimum of  $E[M(x_{it})]$  is bounded away from zero to make this inequality valid.

from which we conclude that

$$\Pr \left[ \frac{1}{T} \left| -\frac{1}{n} \sum_{i=1}^n \ln \det \left( \frac{1}{T} H_i(\theta, \hat{\gamma}_i(\theta)) \right) \right| > \eta \right] = o(T^{-1})$$

for all  $\eta > 0$ .

We now take care of the second component on the RHS of (53). By Assumption 4, each component of  $\Upsilon_i(\theta, \hat{\gamma}_i(\theta))$  is bounded above by  $\sum_{l=-m}^m w_{T,l} \left( \sum_{t=\max(1,l+1)}^{\min(T,T+l)} M(x_{it}) M(x_{it-l}) \right)$ . Using Lemma 15, we can then conclude that

$$\ln \det \left( \frac{1}{T} \Upsilon_i(\theta, \hat{\gamma}_i(\theta)) \right) \leq \ln r! + r \ln \left( \frac{1}{T} \sum_{l=-m}^m \sum_{t=\max(1,l+1)}^{\min(T,T+l)} M(x_{it}) M(x_{it-l}) \right)$$

Using Lemma 2 again, we have

$$\Pr \left[ \max_{1 \leq i \leq n} \left| \frac{1}{T} \sum_{t=\max(1,l+1)}^{\min(T,T+l)} (M(x_{it}) M(x_{it-l}) - E[M(x_{it}) M(x_{it-l})]) \right| > \eta \right] = o(T^{-1})$$

and we obtain

$$\Pr \left[ \max_{1 \leq i \leq n} \left| \ln \left( \frac{1}{T} \sum_{l=-m}^m \sum_{t=\max(1,l+1)}^{\min(T,T+l)} M(x_{it}) M(x_{it-l}) \right) - \ln \left( \sum_{l=-m}^m E[M(x_{it}) M(x_{it-l})] \right) \right| > m\eta \right] = o(T^{-1}),$$

It follows that

$$\Pr \left[ \ln \det \left( \frac{1}{T} \Upsilon_i(\theta, \hat{\gamma}_i(\theta)) \right) > \ln r! + r \frac{1}{n} \sum_{i=1}^n \sum_{l=-m}^m E[M(x_{it}) M(x_{it-l})] + m\eta \right] = o(T^{-1})$$

Because  $E[M(x_{it}) M(x_{it-l})] \leq \sqrt{E[M(x_{it})^2] E[M(x_{it-l})^2]} = E[M(x_{it})^2]$ , we have

$$\Pr \left[ \ln \det \left( \frac{1}{T} \Upsilon_i(\theta, \hat{\gamma}_i(\theta)) \right) > \ln r! + 2m \cdot r \frac{1}{n} \sum_{i=1}^n E[M(x_{it})^2] + m\eta \right] = o(T^{-1})$$

or

$$\Pr \left[ \frac{1}{T} \ln \det \left( \frac{1}{T} \Upsilon_i(\theta, \hat{\gamma}_i(\theta)) \right) > \frac{\ln r!}{T} + \frac{2m}{T} r \sup_i E[M(x_{it})^2] + \frac{m}{T} \eta \right] = o(T^{-1})$$

Therefore, we obtain

$$\Pr \left[ \frac{1}{T} \ln \det \left( \frac{1}{T} \Upsilon_i(\theta, \hat{\gamma}_i(\theta)) \right) > \eta \right] = o(T^{-1})$$

for all  $\eta > 0$ .

## E Proof of Theorem 4

We can verify by inspection that  $\frac{\partial S_n(\theta)}{\partial \theta}$  can be expressed as a sum of terms, all of which are cross section averages of some smooth functions of the form

$$\begin{aligned} & \frac{1}{T} \sum_{t=1}^T D^v \psi(x_{it}, \theta, \widehat{\gamma}_i(\theta)), \quad \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=\max(1, l+1)}^{\min(T, T+l)} \frac{\partial \psi(x_{it}, \theta, \widehat{\gamma}_i(\theta))}{\partial \gamma'} \otimes D^v \psi(x_{it-l}, \theta, \widehat{\gamma}_i(\theta)), \\ & \left( \frac{1}{T} \sum_{t=1}^T \frac{\partial^2 \psi(x_{it}, \theta, \widehat{\gamma}_i(\theta))}{\partial \gamma \partial \gamma'} \right)^{-1}, \quad \left( \frac{1}{T} \sum_{l=-m}^m w_{T,l} \sum_{t=\max(1, l+1)}^{\min(T, T+l)} \frac{\partial \psi(x_{it}, \theta, \widehat{\gamma}_i(\theta))}{\partial \gamma} \frac{\partial \psi(x_{it-l}, \theta, \widehat{\gamma}_i(\theta))}{\partial \gamma'} \right)^{-1} \end{aligned}$$

with  $|v| \leq 4$ . Here,  $\phi \equiv (\theta, \gamma)$ , and  $D^v \psi(x_{it}, \phi) \equiv \partial^{|v|} \psi(x_{it}, \phi) / (\partial \phi_1^{v_1} \dots \partial \phi_k^{v_k})$ , where  $\nu = (\nu_1, \dots, \nu_k)$  be a vector of non-negative integers  $v_i$ , and  $|v| = \sum_{j=1}^k v_j$ . By Assumptions 4, 6, and Lemma 5, we can see that all these terms are  $O_p(1)$  uniformly over  $i$  and  $\theta$ .

## F Proof of Theorem 6

Because of the result in the previous section, we only need to consider  $\Upsilon_i(\theta, \widehat{\gamma}_i(\theta))$ . By Assumption 4, each component of  $\Upsilon_i(\theta, \widehat{\gamma}_i(\theta))$  is bounded above by  $\sum_{l=-m}^m E_{\widehat{\theta}, \widehat{\gamma}_i} [M(x_{it}) M(x_{it-l})]$ . By Assumption 8, we have

$$\sup \sum_{l=-m}^m E_{\widehat{\theta}, \widehat{\gamma}_i} [M(x_{it}) M(x_{it-l})] \leq 2mK$$

where  $K = \sup_{(\theta, \gamma) \in \Phi} \sup_l E_{\theta, \gamma} [M(x_{it}) M(x_{it-l})]$ , and

$$\ln \det(\Upsilon_i(\theta, \widehat{\gamma}_i(\theta))) \leq \ln r! + 2rK \ln m$$

It follows that

$$\Pr [\ln \det(\Upsilon_i(\theta, \widehat{\gamma}_i(\theta))) > \ln r! + 2rK \ln m + \eta] = o(T^{-1})$$

Therefore, we obtain

$$\Pr \left[ \frac{1}{T} \ln \det \left( \frac{1}{T} \Upsilon_i(\theta, \widehat{\gamma}_i(\theta)) \right) > \eta \right] = o(T^{-1})$$

for all  $\eta > 0$  as long as  $\frac{\ln m}{T} = o(1)$ .

We note that all the above results hold even when the preliminary estimates  $(\widehat{\theta}, \widehat{\gamma}_i)$  are replaced by some  $(\theta^*, \gamma_i^*)$ .

## G Proof of Theorem 8

By differentiating  $B_n$ , we obtain that

$$S_n(\theta_0) = [2] + [3] + [4]' + [5]'$$

where

$$\begin{aligned} [2] &= -\frac{1}{2} \frac{1}{n} \sum_{i=1}^n \left( \frac{1}{T} \sum_{t=1}^T \frac{\partial^3 \psi_{it}}{\partial \theta (\partial \gamma' \otimes \partial \gamma')} \right) \text{vec} \left( \left( \frac{1}{T} \sum_{t=1}^T \frac{\partial^2 \psi_{it}}{\partial \gamma \partial \gamma'} \right)^{-1} \right) \\ [3] &= -\frac{1}{2} \frac{1}{n} \sum_{i=1}^n \frac{\partial \hat{\gamma}'_i(\theta)}{\partial \theta} \left( \frac{1}{T} \sum_{t=1}^T \frac{\partial^3 \psi_{it}}{\partial \gamma (\partial \gamma' \otimes \partial \gamma')} \right) \text{vec} \left( \left( \frac{1}{T} \sum_{t=1}^T \frac{\partial^2 \psi_{it}}{\partial \gamma \partial \gamma'} \right)^{-1} \right) \\ [4]' &= \frac{1}{2} \frac{1}{n} \sum_{i=1}^n \left[ \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} \left[ \frac{\partial}{\partial \theta} \left( \left( \frac{\partial \psi_{it}}{\partial \gamma'} \right) \otimes \left( \frac{\partial \psi_{it-l}}{\partial \gamma'} \right) \right) \right] \right] \\ &\quad \cdot \text{vec} \left( \left( \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} \left[ \frac{\partial \psi_{it}}{\partial \gamma} \frac{\partial \psi_{it-l}}{\partial \gamma'} \right] \right)^{-1} \right) \end{aligned}$$

and

$$\begin{aligned} [5]' &= \frac{1}{2} \frac{1}{n} \sum_{i=1}^n \frac{\partial \hat{\gamma}'_i(\theta)}{\partial \theta} \left[ \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} \left[ \frac{\partial}{\partial \gamma} \left( \left( \frac{\partial \psi_{it}}{\partial \gamma'} \right) \otimes \left( \frac{\partial \psi_{it}}{\partial \gamma'} \right) \right) \right] \right] \\ &\quad \cdot \text{vec} \left( \left( \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} \left[ \frac{\partial \psi_{it}}{\partial \gamma} \frac{\partial \psi_{it}}{\partial \gamma'} \right] \right)^{-1} \right) \end{aligned}$$

We can see that [2] and [3] are identical to the ones in the previous section. Because we have already established

$$[2] + [3] = -\frac{1}{2} \frac{1}{n} \sum_{i=1}^n E[U_{it}^{\gamma\gamma}] \text{vec} \left( (E[V_{it}^{\gamma}] )^{-1} \right) + o_p(1)$$

we will focus on [4]' and [5]' here.

Because

$$\begin{aligned} \frac{\partial}{\partial \theta} \left( \left( \frac{\partial \psi_{it}(\theta, \gamma)}{\partial \gamma'} \right) \otimes \left( \frac{\partial \psi_{it-l}(\theta, \gamma)}{\partial \gamma'} \right) \right) &= (U_{it}^{\gamma} + \rho_i V_{it}^{\gamma}) \otimes V'_{it-l} + V'_{it} \otimes (U_{it-l}^{\gamma} + \rho_i V_{it-l}^{\gamma}) \\ \frac{\partial}{\partial \gamma} \left( \left( \frac{\partial \psi_{it}(\theta, \gamma)}{\partial \gamma'} \right) \otimes \left( \frac{\partial \psi_{it}(\theta, \gamma)}{\partial \gamma'} \right) \right) &= V_{it}^{\gamma} \otimes V'_{it-l} + V'_{it} \otimes V_{it-l}^{\gamma} \end{aligned}$$

and

$$\frac{\partial \hat{\gamma}'_i(\theta)}{\partial \theta} = -\rho_i + o_p(1)$$

we can write

$$\begin{aligned}
[4]' + [5]' &= \frac{1}{2} \frac{1}{n} \sum_{i=1}^n \left[ \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} \begin{bmatrix} U_{it}^\gamma(\theta_0, \hat{\gamma}_i(\theta_0)) \otimes V_{it-l}(\theta_0, \hat{\gamma}_i(\theta_0))' \\ + V_{it}(\theta_0, \hat{\gamma}_i(\theta_0))' \otimes U_{it-l}^\gamma(\theta_0, \hat{\gamma}_i(\theta_0)) \end{bmatrix} \right] \\
&\quad \cdot \text{vec} \left( \left( \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} [V_{it}(\theta_0, \hat{\gamma}_i(\theta_0)) V_{it-l}(\theta_0, \hat{\gamma}_i(\theta_0))'] \right)^{-1} \right) + o_p(1)
\end{aligned}$$

Using Lemma 9, we obtain

$$\max_i \left| \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} [V_{it}(\theta_0, \hat{\gamma}_i(\theta_0)) V_{it-l}(\theta_0, \hat{\gamma}_i(\theta_0))'] - \sum_{l=-\infty}^{\infty} E [V_{it} V_{it-l}'] \right| = o_p(1)$$

Furthermore, if the conditional likelihood is properly defined, then we should have  $V_{it}$  serially uncorrelated, which implies that

$$\begin{aligned}
\max_i \left| \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} [V_{it}(\theta_0, \hat{\gamma}_i(\theta_0)) V_{it-l}(\theta_0, \hat{\gamma}_i(\theta_0))'] - E [V_{it} V_{it-l}'] \right| \\
= \max_i \left| \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} [V_{it}(\theta_0, \hat{\gamma}_i(\theta_0)) V_{it-l}(\theta_0, \hat{\gamma}_i(\theta_0))'] + E [V_{it}^\gamma] \right| = o_p(1)
\end{aligned}$$

where the first equality is based on the information equality. Therefore, we obtain

$$\begin{aligned}
&[4]' + [5]' \\
&= -\frac{1}{2} \frac{1}{n} \sum_{i=1}^n \left[ \sum_{l=-m}^m w_{T,l} E_{\hat{\theta}, \hat{\gamma}_i} \begin{pmatrix} U_{it}^\gamma(\theta_0, \hat{\gamma}_i(\theta_0)) \otimes V_{it-l}(\theta_0, \hat{\gamma}_i(\theta_0))' \\ + V_{it}(\theta_0, \hat{\gamma}_i(\theta_0))' \otimes U_{it-l}^\gamma(\theta_0, \hat{\gamma}_i(\theta_0)) \end{pmatrix} \right] \cdot \text{vec} \left( E [V_{it}^\gamma]^{-1} \right) + o_p(1)
\end{aligned}$$

Using Lemma 9 again, we obtain

$$[4]' + [5]' = -\frac{1}{2} \frac{1}{n} \sum_{i=1}^n \sum_{l=-\infty}^{\infty} E [U_{it}^\gamma \otimes V_{it-l}' + V_{it}' \otimes U_{it-l}^\gamma] \text{vec} \left( E [V_{it}^\gamma]^{-1} \right) + o_p(1)$$

Because we have<sup>13</sup>

$$\begin{aligned}
(U_{it}^\gamma \otimes V_{it-l}') \text{vec} \left( E [V_{it}^\gamma]^{-1} \right) &= U_{it}^\gamma E [V_{it}^\gamma]^{-1} V_{it-l} = -U_{it}^\gamma \tilde{V}_{it-l} \\
(V_{it}' \otimes U_{it-l}^\gamma) \text{vec} \left( E [V_{it}^\gamma]^{-1} \right) &= U_{it-l}^\gamma E [V_{it}^\gamma]^{-1} V_{it} = -U_{it-l}^\gamma \tilde{V}_{it}
\end{aligned}$$

it follows that

$$\begin{aligned}
[4]' + [5]' &= \frac{1}{2} \frac{1}{n} \sum_{i=1}^n \sum_{l=-\infty}^{\infty} E [U_{it}^\gamma \tilde{V}_{it-l} + U_{it-l}^\gamma \tilde{V}_{it}] + o_p(1) \\
&= \frac{1}{n} \sum_{i=1}^n \sum_{l=-\infty}^{\infty} E [U_{it}^\gamma \tilde{V}_{it-l}] + o_p(1)
\end{aligned}$$

We note that, because of Lemma 10, all the above results hold even when the preliminary estimates  $(\hat{\theta}, \hat{\gamma}_i)$  are replaced by some  $(\theta^*, \gamma_i^*)$  as long as  $\|\theta^* - \theta_0\| = O_p(T^{-2/5})$  and  $\sup_i \|\gamma_i^* - \gamma_{i0}\| = O_p(T^{-2/5})$ .

<sup>13</sup>See, e.g., Magnus & Neudecker (1988, p. 31, eq. (3)).

## References

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