

# Supplement Material for “Nonparametric Estimation of the Pair Correlation Function of Replicated Inhomogeneous Point processes”

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## Abstract

This document contains all technical proofs for the paper ‘Nonparametric Estimation of the Pair Correlation Function of Replicated Inhomogeneous Point processes’.

**KEY WORDS:** Confidence band; Estimating equations; Local polynomial estimator; Nonparametric estimation; Orthogonal series estimator; Replicated point patterns.

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# 1 Proof of Lemma 1

*Proof.* When  $\lambda(\mathbf{s}) \equiv \lambda$  and the observation window is  $D_n = [0, T_n] \subset \mathbb{R}$ , we have that

$$\begin{aligned}
\mathbf{Q}_{n,h}^{(k)}(r) &= \lambda^2 \frac{1}{|D_n|h} \int_{-T_n}^{T_n} |D_n \cap (D_n - s)| \left[ K \left( \frac{|s| - r}{h} \right) \right]^k \mathbf{A}_h(|s| - r) \mathbf{A}_h^T(|s| - r) ds \\
&= \lambda^2 \frac{1}{T_n h} \int_{-T_n}^{T_n} (T_n - |s|) \left[ K \left( \frac{|s| - r}{h} \right) \right]^k \mathbf{A}_h(|s| - r) \mathbf{A}_h^T(|s| - r) ds \\
&= \lambda^2 \frac{2}{T_n h} \int_0^{T_n} (T_n - s) \left[ K \left( \frac{s - r}{h} \right) \right]^k \mathbf{A}_h(s - r) \mathbf{A}_h^T(s - r) ds \\
&= \lambda^2 \frac{2}{T_n h} \int_{-r}^{T_n - r} (T_n - s - r) \left[ K \left( \frac{s}{h} \right) \right]^k \mathbf{A}_h(s) \mathbf{A}_h^T(s) ds \\
&= \lambda^2 \frac{2}{T_n} \int_{-r/h}^{(T_n - r)/h} (T_n - sh - r) [K(s)]^k \mathbf{A}_1(s) \mathbf{A}_1^T(s) ds.
\end{aligned}$$

If we use the uniform kernel  $K(x) = \frac{1}{2}I(-1 \leq x \leq 1)$ , the above equation can be further simplified as

$$\begin{aligned}
\mathbf{Q}_{n,h}^{(k)}(r) &= \frac{\lambda^2}{2^{k-1}} \frac{1}{T_n} \int_{\max(-r/h, -1)}^{\min[(T_n - r)/h, 1]} (T_n - sh - r) \mathbf{A}_1(s) \mathbf{A}_1^T(s) ds \\
&= \frac{\lambda^2}{2^{k-1}} \left( 1 - \frac{r}{T_n} \right) \int_{\max(-r/h, -1)}^{\min[(T_n - r)/h, 1]} \mathbf{A}_1(s) \mathbf{A}_1^T(s) ds \\
&\quad - \frac{\lambda^2}{2^{k-1}} \frac{h}{T_n} \int_{\max(-r/h, -1)}^{\min[(T_n - r)/h, 1]} s \mathbf{A}_1(s) \mathbf{A}_1^T(s) ds \\
&= \frac{\lambda^2}{2^{k-1}} \left( 1 - \frac{r}{T_n} \right) \mathbf{B}_1(r) - \frac{\lambda^2}{2^{k-1}} \frac{h}{T_n} \mathbf{B}_2(r),
\end{aligned}$$

where  $\mathbf{B}_1(r)$  is a  $(p+1) \times (p+1)$  matrix whose  $(i, j)$ th entry is  $\frac{1}{i+j-1} (q_{up}^{i+j-1} - q_{low}^{i+j-1})$  and  $\mathbf{B}_2(r)$  is a  $(p+1) \times (p+1)$  matrix whose  $(i, j)$ th entry is  $\frac{1}{i+j} (q_{up}^{i+j} - q_{low}^{i+j})$ , with  $q_{low} = \max(-r/h, -1)$  and  $q_{up} = \min[(T_n - r)/h, 1]$ .  $\square$

## 2 Asymptotic Properties of Local Polynomial Estimator

In this Section, we give detailed proofs of Lemma 2 and Theorem 1.

## 2.1 Conditions

The following conditions are sufficient for the asymptotic consistency of  $\hat{g}_h(r)$ .

- [C1] There exists a  $C_\lambda$  such that the intensify function  $0 \leq \lambda(\mathbf{u}) \leq C_\lambda$  for any  $\mathbf{u} \in D_n$ .
- [C2] There exist positive constants  $c_g$ ,  $C_g$  and  $C_f$  such that (a)  $c_g \leq g(r) \leq C_g$ ; (b)  $\max_{1 \leq j \leq p+1} |f^{\{j\}}(r)| \leq C_f$  for any  $r \geq 0$  and that (c)  $\int_0^\infty |g(s) - 1| ds < C_g$ .
- [C3] It holds that (a)  $|g^{(k)}(\mathbf{x}_1, \dots, \mathbf{x}_k)| \leq C_g$  for any  $\mathbf{x}_j \in D_n$ ,  $j = 1, \dots, k$  and  $k = 3, 4, 5, 6$ ; (b)  $\int_{D_n} |g_0^{(3)}(\mathbf{x}, \mathbf{y}) - g(\|\mathbf{x} - \mathbf{y}\|)| d\mathbf{x} \leq C_g$ ; and (c)  $\int_{D_n} |g_0^{(4)}(\mathbf{x}, \mathbf{y} + \mathbf{w}, \mathbf{w}) - g(\|\mathbf{x}\|)g(\|\mathbf{y}\|)| d\mathbf{w} \leq C_g$ .
- [C4] The kernel  $K(x)$  has a bounded support, say  $[-1, 1]$ , such that  $\int_{-1}^1 K(x) dx = 1$ .
- [C5] As the bandwidth  $h \rightarrow 0$  and  $m|D_n|h(r+h)^{d-1} \rightarrow \infty$ , there exists a constant  $c_0 > 0$  such that

$$\eta_{\min} \left[ \mathbf{Q}_{n,h}^{(k)}(r) \right] (r+h)^{1-d} > c_0, \quad k = 1, 2,$$

where  $\eta_{\min}(\mathbf{Q})$  denotes the smallest eigenvalue of the matrix  $\mathbf{Q}$ .

We need to make the following two additional assumptions for the asymptotic normality of  $\hat{g}_h(r)$ .

- [N1] Either one of the following conditions are true (a)  $m \rightarrow \infty$ ; or (b) the mixing coefficient satisfies  $\alpha_X(s; h^{-1}, \infty) = O(s^{-d-\varepsilon})$  for some  $\varepsilon > 0$ .
- [N2] There exists  $\delta > 2d/\varepsilon$  such that  $|g^{(k)}(\mathbf{x}_1, \dots, \mathbf{x}_k)| \leq C_g$  for any  $\mathbf{x}_j \in D_n$ ,  $j = 1, \dots, k$ ,  $k = 2, \dots, 2(2 + \lceil \delta \rceil)$ , where  $\lceil \delta \rceil$  is the smallest integer greater than  $\delta$ .

## 2.2 Sketch of the proof

**Step 1** We first derive the asymptotic limit of solutions to  $\tilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta}) = \mathbf{0}$ , namely,  $\boldsymbol{\theta}^*$  defined in the (A.1) in the next subsection. As a result, Lemma A.1 gives the asymptotic

bias of the local polynomial estimator by quantifying the distance between  $\boldsymbol{\theta}^*$  and derivatives of  $f(r) = \log[g(r)]$ .

**Step 2** Lemma A.2 gives the convergence rate of  $\widehat{\boldsymbol{\theta}}$  to  $\boldsymbol{\theta}^*$ , which is of the order  $O_P\left(\frac{1}{\sqrt{m|D_n|h}}\right)$  entry-wise;

**Step 3** Establish the asymptotic normality of  $\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*$  through Lemmas A.3 to A.5.

**Step 4** Finally use the delta method to derive asymptotic distribution of  $\hat{g}_h(r) - g(r)$  based on distributional results of  $\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*$  given in Lemmas A.4-A.5, following the approach proposed in Biscio and Waagepetersen (2019).

## 2.3 The asymptotic bias

Suppose there exists a vector  $\boldsymbol{\theta}^* = (\theta_0^*, \theta_1^*, \dots, \theta_p^*)^T \in \mathbb{R}^{p+1}$  such that

$$\int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) [g(\|\mathbf{u} - \mathbf{v}\|) - \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*)] \mathbf{G}_r(\|\mathbf{u} - \mathbf{v}\|) d\mathbf{u}d\mathbf{v} = \mathbf{0}, \quad (\text{A.1})$$

where  $\tilde{g}_{r,h}(\cdot; \cdot)$  is defined in equation (3). Obviously  $\boldsymbol{\theta}^*$  depends on  $n$  and  $h$  and  $r$ ; i.e.,  $\boldsymbol{\theta}^* = \boldsymbol{\theta}_{n,h,r}^*$ . Moreover, since  $\mathbf{A}_h(t - r) = \mathbf{D}_h^{-1}\mathbf{G}_r(t)$ , where  $\mathbf{D}_h = \text{diag}(1, h, \dots, h^p)$ , the above equation (A.1) is equivalent to

$$\int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) [g(\|\mathbf{u} - \mathbf{v}\|) - \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*)] \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) d\mathbf{u}d\mathbf{v} = \mathbf{0}.$$

The following Lemma quantifies the distance between  $g(t)$  and  $\tilde{g}_{r,h}(t; \boldsymbol{\theta}^*)$ .

**Lemma A.1.** *Under conditions C1-C5, we have that as  $h \rightarrow 0$ ,*

$$h^j [\theta_j^* - f^{\{j\}}(r)/j!] = O(h^{p+1}), \quad j = 0, 1, \dots, p, \quad (\text{A.2})$$

$$|g(t) - \tilde{g}_{r,h}(t; \boldsymbol{\theta}^*)| = O(h^{p+1}), \quad \text{for } t \in [r - h, r + h]. \quad (\text{A.3})$$

*Proof.* Define function

$$g_{r,0}(t) = \exp \left\{ f(r) + f^{\{1\}}(r)(t-r)/1! + \cdots + f^{\{p\}}(r)(t-r)^p/p! \right\}.$$

Then, from the Taylor's theorem with Lagrange's form of remainder

$$g(t) - g_{r,0}(t) = g(t) \left\{ 1 - \exp \left[ \log g_{r,0}(t) - f(t) \right] \right\} = g(t) \left\{ 1 - \exp \left[ -\frac{f^{\{p+1\}}(r^*)(t-r)^{p+1}}{(p+1)!} \right] \right\},$$

where  $r^*$  is between  $t$  and  $r$ , and it is straightforward to show that

$$\begin{aligned} & \int_{D_n^2} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \lambda(\mathbf{u}) \lambda(\mathbf{v}) [g(\|\mathbf{u} - \mathbf{v}\|) - g_{r,0}(\|\mathbf{u} - \mathbf{v}\|)] \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) d\mathbf{u} d\mathbf{v} \\ &= \int_{D_n^2} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \lambda(\mathbf{u}) \lambda(\mathbf{v}) g(\|\mathbf{u} - \mathbf{v}\|) \\ & \quad \left\{ 1 - \exp \left[ -\frac{f^{\{p+1\}}(r_{\|\mathbf{u}-\mathbf{v}\|}^*)(\|\mathbf{u} - \mathbf{v}\| - r)^{p+1}}{(p+1)!} \right] \right\} \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) d\mathbf{u} d\mathbf{v} \\ &\leq C_\lambda^2 \int_{\mathbb{R}^d} K_h(\|\mathbf{s}\| - r) g(\|\mathbf{s}\|) \left| 1 - \exp \left[ -\frac{f^{\{p+1\}}(r_{\|\mathbf{s}\|}^*)(\|\mathbf{s}\| - r)^{p+1}}{(p+1)!} \right] \right| \mathbf{A}_h(\|\mathbf{s}\| - r) d\mathbf{s} \end{aligned}$$

where the last inequality follows from condition C1 and  $|D_n \cap (D_n - \mathbf{s})| \leq |D_n|$  for any  $\mathbf{s} \in \mathbb{R}^d$ . Combining conditions C2(a)-(b), C4 and the fact that  $|1 - e^x| \leq |x|e^{|x|}$ , we have that as  $h \rightarrow 0$ ,

$$\begin{aligned} & (r+h)^{1-d} \int_{D_n^2} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \lambda(\mathbf{u}) \lambda(\mathbf{v}) [g(\|\mathbf{u} - \mathbf{v}\|) - g_{r,0}(\|\mathbf{u} - \mathbf{v}\|)] \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) d\mathbf{u} d\mathbf{v} \\ &\leq \frac{C_\lambda^2 C_g C_f}{(p+1)!} (r+h)^{1-d} \int_0^\infty |s-r|^{p+1} \exp \left\{ \frac{C_f}{p+1} |s-r|^{p+1} \right\} K_h(s-r) \mathbf{A}_h(s-r) s^{d-1} ds \\ &= O(h^{p+1}) \int_{-r/h}^\infty |s|^{p+1} \exp \left\{ \frac{C_f h^{p+1}}{p+1} |s|^{p+1} \right\} K(s) \mathbf{A}_1(s) \underbrace{\left( \frac{r+sh}{r+h} \right)^{d-1}}_{\leq 1} ds \\ &= O(h^{p+1}). \end{aligned}$$

By the definition of  $\boldsymbol{\theta}^*$  in (A.1), using the above equation, it is straightforward to see that

$$\begin{aligned} & (r+h)^{1-d} \int_{D_n^2} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \lambda(\mathbf{u}) \lambda(\mathbf{v}) [\tilde{g}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) - g_{r,0}(\|\mathbf{u} - \mathbf{v}\|)] \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) d\mathbf{u} d\mathbf{v} \\ &= O(h^{p+1}). \end{aligned} \tag{A.4}$$

Let  $\mathbf{a} = (a_0, a_1, \dots, a_p)^T$ , define the continuously differentiable function  $\mathbf{F} : \mathbb{R}^{p+1} \rightarrow \mathbb{R}^{p+1}$  as follows

$$\mathbf{F}(\mathbf{a}) = (r+h)^{1-d} \int_{D_n^2} w_{r,h}(\|\mathbf{u}-\mathbf{v}\|) \lambda(\mathbf{u}) \lambda(\mathbf{v}) q_{\mathbf{a},h}(\|\mathbf{u}-\mathbf{v}\|-r) \mathbf{A}_h(\|\mathbf{u}-\mathbf{v}\|-r) d\mathbf{u}d\mathbf{v},$$

where  $q_{\mathbf{a},h}(t) = \exp\{\mathbf{a}^T \mathbf{A}_h(t-r)\} = \exp(a_0 + a_1(t-r)/h + \dots + a_p(t-r)^p/h^p)$ . Recall that  $\tilde{g}_{r,h}(t; \boldsymbol{\theta}) = \exp[\theta_0 + \theta_1(t-r) + \dots + \theta_p(t-r)^p] = \exp[(\mathbf{D}_h \boldsymbol{\theta})^T \mathbf{A}_h(t-r)]$ , then equation (A.4) immediately yields that

$$\mathbf{F}(\mathbf{D}_h \boldsymbol{\theta}^*) - \mathbf{F}(f(r), hf^{\{1\}}(r)/1!, \dots, h^p f^{\{p\}}(r)/p!) = O(h^{p+1}), \quad (\text{A.5})$$

where  $\mathbf{D}_h = \text{diag}(1, h, \dots, h^p)$ . The Jacobian matrix of  $\mathbf{F}(\mathbf{a})$  then becomes

$$\begin{aligned} \mathbf{J}(\mathbf{a}) &= \left[ \frac{\partial \mathbf{F}(\mathbf{a})}{\partial a_0}, \dots, \frac{\partial \mathbf{F}(\mathbf{a})}{\partial a_p} \right] \\ &= (r+h)^{1-d} \int_{D_n^2} w_{r,h}(\|\mathbf{u}-\mathbf{v}\|) \lambda(\mathbf{u}) \lambda(\mathbf{v}) q_{\mathbf{a},h}(\|\mathbf{u}-\mathbf{v}\|-r) \mathbf{A}_h(\|\mathbf{u}-\mathbf{v}\|-r) \mathbf{A}_h^T(\|\mathbf{u}-\mathbf{v}\|-r) d\mathbf{u}d\mathbf{v}. \end{aligned}$$

Plugging in  $\mathbf{a}_0 = (\log[g(r)], 0, \dots, 0)^T$  back to the above equation, we have that  $\mathbf{J}(\mathbf{a}_0) = g(r) \mathbf{Q}_{n,h}^{(1)}(r)$ , where  $\mathbf{Q}_{n,h}^{(1)}(r)$  is as defined in equation (11). Using conditions C2(a) and C5, we have that  $\mathbf{J}(\mathbf{a})$  is strictly positive definite at  $\mathbf{a}_0 = (\log[g(r)], 0, \dots, 0)^T$  and hence  $\det(\mathbf{J}(\mathbf{a}_0)) > 0$ . Based on equation (A.5) and a simple application of the inverse function theorem (Burkill and Burkill, 2002, page 223) imply that  $\mathbf{F}$  is invertible near  $\mathbf{a}_0 = (f(r), 0, \dots, 0)^T$  and as  $h \rightarrow 0$ , one has that,

$$h^j [\theta_j^* - f^{\{j\}}(r)/j!] = O(h^{p+1}), \quad j = 0, 1, \dots, p.$$

Similar argument has been used in, e.g., Loader et al. (1996).

Finally, for any  $t$  satisfying  $|t-r| \leq h$ , we have that

$$|g(t) - \tilde{g}_{r,h}(t; \boldsymbol{\theta}^*)| = g(t) |1 - \exp[\theta_0^* + \theta_1^*(t-r) + \dots + \theta_p^*(t-r)^p - f(t)]|$$

and

$$\begin{aligned}
\theta_0^* + \theta_1^*(t-r) + \cdots + \theta_p^*(t-r)^p - f(t) &= \sum_{j=0}^p [h^j(\theta_j^* - f^{(j)}(r)/j!)] \frac{(t-r)^j}{h^j} - f^{(p+1)}(r) \frac{|t-r|^{p+1}}{(p+1)!} \\
&\leq \sum_{j=0}^p h^j(\theta_j^* - f^{(j)}(r)/j!) + \frac{C_f}{(p+1)!} h^{p+1} \\
&= O(h^{p+1}).
\end{aligned}$$

Thus  $|g(t) - \tilde{g}_{r,h}(t; \boldsymbol{\theta}^*)| = O(h^{p+1})$ , which concludes the proof. Note that, in particular,

$$|\exp(\theta_0^*) - g(r)| = g(r) |\exp(\theta_0^* - f(r)) - 1| \leq g(r) |\theta_0^* - f(r)| \exp(|\theta_0^* - f(r)|) = O(h^{p+1}). \quad (\text{A.6})$$

□

## 2.4 Proof of Lemma 2

The proof of Lemma 2 follows immediately from the following Lemma A.2 and Lemma A.1 in the last subsection, because

$$|\hat{g}_h(r) - g(r)| \leq \exp(\theta_0^*) |\exp(\hat{\theta}_0 - \theta_0^*) - 1| + |\exp(\theta_0^*) - g(r)|.$$

So, in this section we just prove the following Lemma A.2.

**Lemma A.2.** *Under conditions C1-C5, we have that as  $m|D_n|h(r+h)^{d-1} \rightarrow \infty$  and  $h \rightarrow 0$ ,*

$$\|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|_h = O_P \left( \frac{1}{\sqrt{m|D_n|h(r+h)^{d-1}}} \right), \quad (\text{A.7})$$

where the norm  $\|\mathbf{x}\|_h^2 = x_0^2 + (hx_1)^2 + \cdots + (h^p x_p)^2$  for any  $\mathbf{x} = (x_0, x_1, \dots, x_p)^T \in \mathbb{R}^{p+1}$  and  $\boldsymbol{\theta}^*$  is defined in equation (A.1).

*Proof.* It is straightforward to see that solving estimating equation (4) is equivalent to maximizing the composite log likelihood function

$$\begin{aligned}
\tilde{L}_{r,h}(\boldsymbol{\theta}) &= \frac{1}{m} \sum_{i=1}^m \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \log [\tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta})] \\
&\quad - \frac{1}{m(m-1)} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}),
\end{aligned} \quad (\text{A.8})$$

with respect to  $\boldsymbol{\theta}$ , because  $\partial \tilde{L}_{r,h}(\boldsymbol{\theta})/\partial \boldsymbol{\theta} = \tilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta})/m$ . Note that the Hessian matrix

$$\tilde{\mathbf{H}}_{r,h}(\boldsymbol{\theta}) = \frac{\partial^2 \tilde{L}_{r,h}(\boldsymbol{\theta})}{\partial \boldsymbol{\theta} \partial \boldsymbol{\theta}^T} = - \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_{r,h}(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}) \mathbf{G}_r(\|\mathbf{u} - \mathbf{v}\|) \mathbf{G}_r^T(\|\mathbf{u} - \mathbf{v}\|),$$

is negative definitive, which implies that  $\tilde{L}_{r,h}(\boldsymbol{\theta})$  is a concave function of  $\boldsymbol{\theta}$ .

Let  $J_{m,n}$  be a sequence of positive real numbers such that  $J_{m,n} \rightarrow \infty$  as  $m \rightarrow \infty$  and/or  $n \rightarrow \infty$ . We shall show that for any given  $\varepsilon > 0$  there exists a large constant  $C_\varepsilon$  such that, for large  $m$  and/or large  $n$ ,

$$\mathbb{P} \left\{ \sup_{\|\boldsymbol{\delta}\|_h=1} \tilde{L}_{r,h}(\boldsymbol{\theta}^* + C_\varepsilon J_{m,n}^{-1/2} \boldsymbol{\delta}) < \tilde{L}_{r,h}(\boldsymbol{\theta}^*) \right\} \geq 1 - \varepsilon. \quad (\text{A.9})$$

Inequality (A.9) implies that with probability tending to 1 there is a local maximum, denoted as  $\hat{\boldsymbol{\theta}} = \hat{\boldsymbol{\theta}}_{m,n}$ , in the the ellipsoid  $\left\{ \boldsymbol{\theta}^* + C_\varepsilon J_{m,n}^{-1/2} \boldsymbol{\delta} : \|\boldsymbol{\delta}\|_h = 1 \right\}$  centered at  $\boldsymbol{\theta}^*$ . It then follows that  $J_{m,n} \|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|_h^2$  is bounded in probability; i.e.,  $\|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|_h^2 = O_P(J_{m,n}^{-1})$ .

To show (A.9), let  $\boldsymbol{\delta}_{m,n}$  be any sequence in  $\{\boldsymbol{\delta} \in \mathbb{R}^{p+1} : \|\boldsymbol{\delta}\|_h = 1\}$  and define a function of  $z \geq 0$  as

$$H_{m,n}(z) = -\tilde{L}_{r,h}(\boldsymbol{\theta}^* + z J_{m,n}^{-1/2} \boldsymbol{\delta}_{m,n}). \quad (\text{A.10})$$

Then

$$H'_{m,n}(z) = -\frac{J_{m,n}^{-1/2}}{m} \boldsymbol{\delta}_{m,n}^T \tilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta}^* + z J_{m,n}^{-1/2} \boldsymbol{\delta}_{m,n}), \quad (\text{A.11})$$

$$H''_{m,n}(z) = -J_{m,n}^{-1} \boldsymbol{\delta}_{m,n}^T \tilde{\mathbf{H}}_{r,h}(\boldsymbol{\theta}^* + z J_{m,n}^{-1/2} \boldsymbol{\delta}_{m,n}) \boldsymbol{\delta}_{m,n} \quad (\text{A.12})$$

and by the Taylor's theorem

$$H_{m,n}(z) = H_{m,n}(0) + H'_{m,n}(0)z + H''_{m,n}(t_z) \frac{z^2}{2}$$

for any  $z > 0$  and some  $0 < t_z < z$ , which implies that

$$\tilde{L}_{r,h}(\boldsymbol{\theta}^* + z J_{m,n}^{-1/2} \boldsymbol{\delta}_{m,n}) - \tilde{L}_{r,h}(\boldsymbol{\theta}^*) = H_{m,n}(0) - H_{m,n}(z) = -z \left[ H'_{m,n}(0) + \frac{z}{2} H''_{m,n}(t_z) \right]. \quad (\text{A.13})$$



By definition,  $H_{m,n}(z)$  is a convex function of  $z$  since  $H''_{m,n}(z) \geq 0$  for any constant  $z$ . Therefore, to find a large enough  $C_\epsilon$  so that (A.9) holds, it suffices to show that  $H'_{m,n}(0) = O_p [H''_{m,n}(t_z)]$  for any  $z > 0$ . We first investigate  $H'_{m,n}(0)$ . By (A.11) and the definition of  $\boldsymbol{\theta}^*$  in (A.1), we have that  $\mathbb{E} [H'_{m,n}(0)] = -J_{m,n}^{-1/2}/m \boldsymbol{\delta}_{m,n}^T \mathbb{E} [\mathbf{U}_{r,h}(\boldsymbol{\theta}^*)] = 0$ . Furthermore,

$$\text{Var} [H'_{m,n}(0)] = \frac{J_{m,n}^{-1}}{m^2} \boldsymbol{\delta}_{m,n}^T \text{Var} [\tilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta}^*)] \boldsymbol{\delta}_{m,n} \quad (\text{A.14})$$

and since  $X_1, \dots, X_m$  are independent replicates of the same Cox process

$$\begin{aligned} \text{Var} [\tilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta}^*)] &= \text{Var} \left[ \sum_{i=1}^m \mathbf{Z}_{1,i} - \frac{1}{m-1} \sum_{i \neq j=1}^m \sum_{j=1}^m \mathbf{Z}_{2,i,j}(\boldsymbol{\theta}^*) \right] \\ &= m \text{Var}(\mathbf{Z}_{1,1}) + \frac{2m}{(m-1)} \text{Var} [\mathbf{Z}_{2,1,2}(\boldsymbol{\theta}^*)] \\ &\quad + \frac{4m(m-2)}{(m-1)} \text{Cov} [\mathbf{Z}_{2,1,2}(\boldsymbol{\theta}^*), \mathbf{Z}_{2,1,3}(\boldsymbol{\theta}^*)] - 2m \text{Cov} [\mathbf{Z}_{1,1}, \mathbf{Z}_{2,1,2}(\boldsymbol{\theta}^*)], \end{aligned}$$

where

$$\begin{aligned} \mathbf{Z}_{1,i} &= \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} \mathbf{G}_r(\|\mathbf{u} - \mathbf{v}\|) w_{r,h}(\|\mathbf{u} - \mathbf{v}\|), \\ \mathbf{Z}_{2,i,j}(\boldsymbol{\theta}^*) &= \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \mathbf{G}_r(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) w_{r,h}(\|\mathbf{u} - \mathbf{v}\|). \end{aligned}$$

From a straightforward algebra and the definition of normalized joint intensities, we have that

$$\begin{aligned} \text{Var}(\mathbf{Z}_{1,1}) &= \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \\ &\quad \times \mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T [g^{(4)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2, \mathbf{v}_2) - g(\|\mathbf{u}_1 - \mathbf{v}_1\|) g(\|\mathbf{u}_2 - \mathbf{v}_2\|)] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\ &\quad + 4 \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) \\ &\quad \times \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\ &\quad + 2 \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) [w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T g(\|\mathbf{u}_1 - \mathbf{v}_1\|) d\mathbf{u}_1 d\mathbf{v}_1, \end{aligned}$$

$$\begin{aligned}
\text{Cov} [\mathbf{Z}_{1,1}, \mathbf{Z}_{2,1,2}(\boldsymbol{\theta}^*)] &= \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)\tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}) \\
&\quad \times [g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) - g(\|\mathbf{u}_1 - \mathbf{v}_1\|)] \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ 2 \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)g(\|\mathbf{u}_1 - \mathbf{v}_1\|) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2,
\end{aligned}$$

$$\begin{aligned}
\text{Var} [\mathbf{Z}_{2,1,2}(\boldsymbol{\theta}^*)] &= \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)\tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|)g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1] \\
&\quad \times \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ 2 \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)\tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*)g(\|\mathbf{v}_1 - \mathbf{u}_2\|)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ \int_{D_n^2} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|)\tilde{g}_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T d\mathbf{u}_1 d\mathbf{v}_1,
\end{aligned}$$

and

$$\begin{aligned}
\text{Cov} [\mathbf{Z}_{2,1,2}(\boldsymbol{\theta}^*), \mathbf{Z}_{2,1,3}(\boldsymbol{\theta}^*)] &= \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \\
&\quad \times [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1] \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)\tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&+ \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)\tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2.
\end{aligned}$$

Therefore, substituting the above terms in  $\text{Var} \left[ \tilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta}^*) \right]$ , we obtain

$$\begin{aligned}
m^{-1} \text{Var} \left[ \tilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta}^*) \right] &= \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \left[ g^{(4)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2, \mathbf{v}_2) \right. \\
&\quad \left. - g(\|\mathbf{u}_1 - \mathbf{v}_1\|) g(\|\mathbf{u}_2 - \mathbf{v}_2\|) \right] \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&\quad - 4 \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \left[ g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) - g(\|\mathbf{u}_1 - \mathbf{v}_1\|) \right] \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ \frac{2}{m-1} \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times \left[ g(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1 \right] \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ \frac{4(m-2)}{m-1} \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \left[ g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1 \right] \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&+ 4 \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) \\
&\quad \times \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&- 4 \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) \left[ 2g(\|\mathbf{u}_1 - \mathbf{v}_1\|) - \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \right] \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ \frac{4}{m-1} \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*) \left[ g(\|\mathbf{v}_1 - \mathbf{u}_2\|) - 1 \right] \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ 2 \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \left[ w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) \right]^2 g(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T d\mathbf{u}_1 d\mathbf{v}_1 \\
&+ \frac{2}{m-1} \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T d\mathbf{u}_1 d\mathbf{v}_1
\end{aligned}$$

Note that by definition,  $\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(t) = \boldsymbol{\eta}_{m,n}^T \mathbf{A}_h(t-r)$ , where  $\boldsymbol{\eta}_{m,n} = \mathbf{D}_h^{-1} \boldsymbol{\delta}_{m,n}$  and  $\|\boldsymbol{\eta}_{m,n}\|^2 = \|\boldsymbol{\delta}_{m,n}\|_h = 1$ , which implies that

$$|\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(t)|^2 = |\boldsymbol{\eta}_{m,n}^T \mathbf{A}_h(t-r)|^2 \leq \|\mathbf{A}_h(t-r)\|^2 \|\boldsymbol{\eta}_{m,n}\|^2 = \|\mathbf{A}_h(t-r)\|^2 \leq p+1$$

for any  $r-h \leq t \leq r+h$ . Therefore, from (A.14) and under conditions C1, C2(a)-(b), C4

and equation (A.3), we obtain

$$\begin{aligned}
mJ_{m,n} \text{Var} [H'_{m,n}(0)] &= O(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)|g^{(4)}(\mathbf{u}_1, \mathbf{v}_1, u_2, \mathbf{v}_2) \\
&\quad - g(\|\mathbf{u}_1 - \mathbf{v}_1\|)g(\|\mathbf{u}_2 - \mathbf{v}_2\|)|d\mathbf{u}_1d\mathbf{v}_1d\mathbf{u}_2d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)|g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, u_2) - g(\|\mathbf{u}_1 - \mathbf{v}_1\|)|d\mathbf{u}_1d\mathbf{v}_1d\mathbf{u}_2d\mathbf{v}_2 \\
&+ \frac{1}{m}O(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)|g(\|\mathbf{u}_1 - \mathbf{u}_2\|)g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1|d\mathbf{u}_1d\mathbf{v}_1d\mathbf{u}_2d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)|g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1|d\mathbf{u}_1d\mathbf{u}_2d\mathbf{v}_1d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^3} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)d\mathbf{u}_1d\mathbf{v}_1d\mathbf{u}_2 \\
&+ \frac{1}{m}O(1) \int_{D_n^3} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)|g(\|\mathbf{v}_1 - \mathbf{u}_2\|) - 1|d\mathbf{u}_1d\mathbf{v}_1d\mathbf{u}_2 \\
&+ O(1) \int_{D_n^2} [w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 d\mathbf{u}_1d\mathbf{v}_1 \\
&= O(1)|D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)|g^{(4)}(\mathbf{s}, \mathbf{t} + \mathbf{w}, \mathbf{w}) - g(\|\mathbf{s}\|)g(\|\mathbf{t}\|)|dsdt\mathbf{w} \\
&+ O(1)|D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)|g^{(3)}(\mathbf{s}, \mathbf{w}) - g(\|\mathbf{s}\|)|dsdt\mathbf{w} \\
&+ \frac{1}{m}O(1)|D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)|g(\|\mathbf{w}\|)g(\|\mathbf{t} - \mathbf{s} + \mathbf{w}\|) - 1|dsdt\mathbf{w} \\
&+ O(1)|D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)|g(\|\mathbf{w}\|) - 1|dsdt\mathbf{w} \\
&+ O(1)|D_n|^{-1} \int_{D_n^2} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)dsdt \\
&+ \frac{1}{m}O(1)|D_n|^{-1} \int_{D_n^2} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)|g(\|\mathbf{s} - \mathbf{t}\|) - 1|dsdt \\
&+ O(1)|D_n|^{-1} \int_{D_n} [K_h(\|\mathbf{s}\| - r)]^2 ds \\
&= O(1)|D_n|^{-1} \int_0^\infty K_h(s - r)s^{d-1}ds \int_0^\infty K_h(t - r)t^{d-1}dt \\
&+ O(1)|D_n|^{-1} \int_0^\infty [K_h(s - r)]^2 s^{d-1}ds,
\end{aligned}$$

where the last equality follows from condition C3.

Finally, using condition C5, we have that

$$\begin{aligned}
mJ_{m,n} \text{Var} [H'_{m,n}(0)] &= O(1)|D_n|^{-1} \int_0^\infty K(s - r/h) s^{d-1} ds \int_0^\infty K(t - r/h) t^{d-1} dt \\
&\quad + O(1)|D_n|^{-1} h^{-1} \int_0^\infty [K(s - r/h)]^2 s^{d-1} ds \\
&= O(1)|D_n|^{-1} \left[ \int_{-r/h}^\infty K(s) (r + sh)^{d-1} ds \right]^2 \\
&\quad + O(1)|D_n|^{-1} h^{-1} \int_{-r/h}^\infty [K(s)]^2 (hs + r)^{d-1} ds \\
&= O\left(\frac{(r+h)^{2d-2}}{|D_n|}\right) + O\left(\frac{(r+h)^{d-1}}{|D_n|h}\right) \\
&= O\left(\frac{(r+h)^{d-1}}{|D_n|h}\right).
\end{aligned}$$

Combing with the fact that  $\mathbb{E} [H'_{m,n}(0)] = 0$ , we have that

$$H'_{m,n}(0) = O_P\left(\frac{(r+h)^{\frac{d-1}{2}}}{\sqrt{m|D_n|h}J_{m,n}}\right). \quad (\text{A.15})$$

Now we proceed to study  $H''_{m,n}(t_z)$ . Let  $\tilde{\boldsymbol{\theta}}^* = \tilde{\boldsymbol{\theta}}^*_{m,n} = \boldsymbol{\theta}^* + t_z J_{m,n}^{-1/2} \boldsymbol{\delta}_{m,n}$  and note that

$$\text{Var} [H''_{m,n}(t_z)] = \frac{J_{m,n}^{-2}}{m(m-1)} \text{Var} \left[ \sum_{\mathbf{u} \in X_1} \sum_{\mathbf{v} \in X_2} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \tilde{\boldsymbol{\theta}}^*) [\boldsymbol{\delta}_{m,n}^T G_r(\|\mathbf{u} - \mathbf{v}\|)]^2 \right].$$

Some tedious algebra gives that

$$\begin{aligned}
& J_{m,n}^2 m(m-1) \text{Var} [H_{m,n}''(t_z)] \\
&= 2 \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \tilde{\boldsymbol{\theta}}^*) \\
&\quad \times [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1] [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ 4 \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \tilde{\boldsymbol{\theta}}^*) g(\|\mathbf{v}_1 - \mathbf{u}_2\|) [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ 2 \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) g(\|\mathbf{u}_1 - \mathbf{v}_1\|) [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^4 d\mathbf{u}_1 d\mathbf{v}_1 \\
&+ 4(m-2) \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1] \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \tilde{\boldsymbol{\theta}}^*) [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_2 - \mathbf{v}_2\|)]^2 d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&+ 4(m-2) \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \tilde{\boldsymbol{\theta}}^*) [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{u}_2\|)]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2.
\end{aligned}$$

Recall that we have shown that  $|\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(t)|^2 \leq p+1$  for any  $r-h \leq t \leq r+h$ . Then under conditions C1, C2(a)-(b), C3 and equation (A.3), we can further simplify  $J_{m,n}^2 m(m-$

1)  $\text{Var} [H''_{m,n}(t_z)]$  as follows

$$\begin{aligned}
& J_{m,n}^2 m(m-1) \text{Var} [H''_{m,n}(t_z)] \\
&= O(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) |g(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1| d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^3} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 + O(1) \int_{D_n^2} w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) d\mathbf{u}_1 d\mathbf{v}_1 \\
&+ mO(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) |g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1| d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&+ mO(1) \int_{D_n^3} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&= O(1) |D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r) K_h(\|\mathbf{t}\| - r) |g(\|\mathbf{w}\|) g(\|\mathbf{t} - \mathbf{s} + \mathbf{w}\|) - 1| ds dt d\mathbf{w} \\
&+ O(1) |D_n|^{-1} \int_{D_n^2} K_h(\|\mathbf{s}\| - r) K_h(\|\mathbf{t}\| - r) ds dt + O(1) |D_n|^{-1} \int_{D_n} [K_h(\|\mathbf{s}\| - r)]^2 ds \\
&+ mO(1) |D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r) K_h(\|\mathbf{t}\| - r) |g(\|\mathbf{w}\|) - 1| ds dt d\mathbf{w} \\
&+ mO(1) |D_n|^{-1} \int_{D_n^2} K_h(\|\mathbf{s}\| - r) K_h(\|\mathbf{t}\| - r) ds dt \\
&= O(1) |D_n|^{-1} \int_0^\infty [K_h(s - r)]^2 s^{d-1} ds + mO(1) |D_n|^{-1} \left[ \int_0^\infty K_h(s - r) s^{d-1} ds \right]^2 \\
&= O(1) |D_n|^{-1} h^{-1} \int_{-r/h}^\infty [K(s)]^2 (r + sh)^{d-1} ds + mO(1) |D_n|^{-1} \left[ \int_{-r/h}^\infty K(s) (r + sh)^{d-1} ds \right]^2
\end{aligned}$$

Then, by condition C4, we finally have that

$$J_{m,n}^2 m(m-1) \text{Var} [H''_{m,n}(t_z)] = \frac{(r+h)^{d-1}}{|D_n| h} O(1) + \frac{m(r+h)^{2d-2}}{|D_n|} O(1),$$

which implies that

$$\text{Var} [H''_{m,n}(t_z)] = O \left( \frac{(r+h)^{d-1}}{J_{m,n}^2 m^2 |D_n| h} + \frac{(r+h)^{2d-2}}{J_{m,n}^2 m |D_n|} \right). \quad (\text{A.16})$$

On the other hand, we have that

$$\mathbb{E} [H''_{m,n}(t_z)] = J_{m,n}^{-1} \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) [\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 d\mathbf{u}_1 d\mathbf{v}_1.$$

Observe that by definition  $\boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) = \boldsymbol{\eta}_{m,n}^T \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)$  with  $\|\boldsymbol{\eta}_{m,n}\|^2 = 1$

and the fact that the smallest eigenvalue satisfies the condition

$$\eta_{\min} \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right] = \inf_{\|\boldsymbol{\eta}\|^2=1} \boldsymbol{\eta}^T \mathbf{Q}_{n,h}^{(1)}(r) \boldsymbol{\eta}.$$

Using definition of  $\mathbf{Q}_{n,h}^{(1)}(r)$  in condition C5, we have that

$$\begin{aligned} \mathbb{E} [J_{m,n} H''_{m,n}(t_z)] - g(r) \eta_{\min} \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right] &\geq \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) \\ &\quad \times \left[ \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) - g(r) \right] \left[ \boldsymbol{\delta}_{m,n}^T \mathbf{G}_r(\|\mathbf{u}_1 - \mathbf{v}_1\|) \right]^2 d\mathbf{u}_1 d\mathbf{v}_1 \\ &= O(1) \int_{D_n^2} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) \left| \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) - g(r) \right| d\mathbf{u}_1 d\mathbf{v}_1 \\ &= O(1) \int_0^\infty K_h(s-r) \left| \tilde{g}_{r,h}(s; \tilde{\boldsymbol{\theta}}^*) - g(r) \right| s^{d-1} ds. \end{aligned}$$

Note that by the definition of  $\tilde{\boldsymbol{\theta}}^* = \boldsymbol{\theta}^* + t_z J_{m,n}^{-1/2} \boldsymbol{\delta}_{m,n}$  and equation (A.2) of Lemma A.1, it is straightforward to show that

$$\left| \tilde{g}_{r,h}(t; \tilde{\boldsymbol{\theta}}^*) - g(r) \right| = g(r) \left| \exp \left( \sum_{j=0}^p \left( \tilde{\theta}_j^* - f^{\{j\}}(r)/j! \right) (t-r)^j - f^{(p+1)}(r) \frac{(t-r)^{p+1}}{(p+1)!} \right) - 1 \right|$$

and for any  $r-h \leq t \leq r+h$ ,

$$\begin{aligned} \left( \tilde{\theta}_j^* - f^{\{j\}}(r)/j! \right) (t-r)^j &= h^j \left( \theta_j^* - f^{\{j\}}(r)/j! \right) \frac{(t-r)^j}{h^j} + z_0 J_{m,n}^{-1/2} h^j \delta_{m,n,j+1} \frac{(t-r)^j}{h^j} \\ &\leq h^j \left( \theta_j^* - f^{\{j\}}(r)/j! \right) \frac{(t-r)^j}{h^j} + z_0 J_{m,n}^{-1/2} \|\boldsymbol{\delta}_{m,n}\| h \\ &= O(h^{p+1} + J_{m,n}^{-1/2}), \end{aligned}$$

which implies that

$$\left| \tilde{g}_{r,h}(t; \tilde{\boldsymbol{\theta}}^*) - g(r) \right| = g(r) O(h^{p+1} + J_{m,n}^{-1/2}).$$

Therefore, we have that, under conditions C4-C5,

$$\begin{aligned} (r+h)^{1-d} \left\{ \mathbb{E} [J_{m,n} H''_{m,n}(t_z)] - g(r) \eta_{\min} \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right] \right\} &\geq O(J_{m,n}^{-1/2} + h^{p+1}) \int_{-r/h}^\infty K(s) \underbrace{\left( \frac{r+sh}{r+h} \right)^{d-1}}_{\leq 1} ds \\ &= O(J_{m,n}^{-1/2} + h^{p+1}). \end{aligned}$$



By condition C2(a) and C5, the above equation gives that

$$(r+h)^{1-d} \mathbb{E} [J_{m,n} H''_{m,n}(t_z)] \geq c_g c_0 + O(J_{m,n}^{-1/2} + h^{p+1}). \quad (\text{A.17})$$

Hence for the constant  $c = c_0 c_g$ , by an application of the Chebyshev's inequality we have that

$$\begin{aligned} & \mathbb{P} \left\{ (r+h)^{1-d} J_{m,n} H''_{m,n}(t_z) < \frac{c}{2} \right\} \\ &= \mathbb{P} \left\{ \frac{J_{m,n} H''_{m,n}(t_z)}{(r+h)^{d-1}} - \frac{\mathbb{E} [J_{m,n} H''_{m,n}(t_z)]}{(r+h)^{d-1}} < \frac{c}{2} - \frac{\mathbb{E} [J_{m,n} H''_{m,n}(t_z)]}{(r+h)^{d-1}} \right\} \\ &\leq \mathbb{P} \left\{ \frac{|J_{m,n} H''_{m,n}(t_z) - \mathbb{E} [J_{m,n} H''_{m,n}(t_z)]|}{(r+h)^{d-1}} > \left| \frac{c}{2} - \frac{\mathbb{E} [J_{m,n} H''_{m,n}(t_z)]}{(r+h)^{d-1}} \right| \right\} I \left\{ \frac{\mathbb{E} [J_{m,n} H''_{m,n}(t_z)]}{(r+h)^{d-1}} > \frac{c}{2} \right\} \\ &\quad + I \left\{ \frac{\mathbb{E} [J_{m,n} H''_{m,n}(t_z)]}{(r+h)^{d-1}} \leq \frac{c}{2} \right\} \\ &\leq \frac{\text{Var} [(r+h)^{1-d} J_{m,n} H''_{m,n}(t_z)]}{|c/2 - \mathbb{E} [(r+h)^{1-d} J_{m,n} H''_{m,n}(t_z)]|^2} I \left\{ \frac{\mathbb{E} [J_{m,n} H''_{m,n}(t_z)]}{(r+h)^{d-1}} > \frac{c}{2} \right\} + I \left\{ \frac{\mathbb{E} [J_{m,n} H''_{m,n}(t_z)]}{(r+h)^{d-1}} \leq \frac{c}{2} \right\} \\ &= O(1) \text{Var} [(r+h)^{1-d} J_{m,n} H''_{m,n}(t_z)] + o(1) \\ &= O \left( \frac{(r+h)^{1-d}}{m^2 |D_n| h} + \frac{1}{m |D_n|} \right) + o(1), \end{aligned}$$

where the last equality follows from equations (A.16) and (A.17) as  $J_{m,n} \rightarrow \infty$  and  $h \rightarrow 0$ .

Therefore, as long as  $m |D_n| h (r+h)^{d-1} \rightarrow \infty$ ,  $J_{m,n} \rightarrow \infty$  and  $h \rightarrow 0$ , we have that

$$\mathbb{P} \left\{ (r+h)^{1-d} J_{m,n} H''_{m,n}(t_z) \geq \frac{c_0 c_g}{2} \right\} \rightarrow 1, \quad (\text{A.18})$$

where  $c_g$  and  $c_0$  are constants defined in conditions C2(a) and C6, respectively.

We have already shown in equation (A.15) that

$$H'_{m,n}(0) = O_P \left( \frac{(r+h)^{(d-1)/2}}{\sqrt{J_{m,n} m |D_n| h}} \right),$$

hence as long as  $\frac{J_{m,n} (r+h)^{d-1}}{m |D_n| h} = O(1)$ , we have that  $H'_{m,n}(0) = O_P(H''_{m,n}(t_z))$ . In other words,

by taking  $J_{m,n} = m |D_n| h (r+h)^{1-d}$ , we have that

$$\mathbb{P} \left\{ |H'_{m,n}(0)| \geq \frac{z}{2} H''_{m,n}(t_z) \right\} < \epsilon,$$

where  $\epsilon$  can be arbitrary small by choosing  $z$  and  $m$  and/or  $n$  large enough. Therefore, with  $J_{m,n} = m|D_n|h(r+h)^{1-d}$ , for any given  $\epsilon > 0$ , there exists  $z_\epsilon > 0$  such that for large  $m$  and/or  $n$ ,

$$\mathbb{P} \left\{ \tilde{L}_{r,h}(\boldsymbol{\theta}^* + z_\epsilon J_{m,n}^{-1/2} \boldsymbol{\delta}_{m,n}) < \tilde{L}_{r,h}(\boldsymbol{\theta}^*) \right\} = \mathbb{P} \left\{ z_\epsilon H'_{m,n}(0) + \frac{z_\epsilon^2}{2} H''_{m,n}(t_{z_\epsilon}) > 0 \right\} \geq 1 - \epsilon.$$

Thus, (A.9) holds, which completes the proof of equation (A.7).  $\square$

## 2.5 Proof of Theorem 1

Define two random vectors

$$\mathbf{Z}_1 = \frac{1}{m} \sum_{i=1}^m \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r), \quad (\text{A.19})$$

$$\mathbf{Z}_2(\boldsymbol{\theta}^*) = \frac{1}{m(m-1)} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) \quad (\text{A.20})$$

By definition of  $\boldsymbol{\theta}^*$  in (A.1), we have that

$$\mathbb{E} \mathbf{Z}_1 = \mathbb{E} \mathbf{Z}_2 = \int_{D_n^2} \lambda(\mathbf{u}) \lambda(\mathbf{v}) w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) g(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) d\mathbf{u} d\mathbf{v}. \quad (\text{A.21})$$

**Lemma A.3.** *Under conditions C1-C5, as  $h \rightarrow 0$  and  $m|D_n|h(r+h)^{d-1} \rightarrow \infty$ , we have that,*

$$(m|D_n|h) \text{Var}(\mathbf{Z}_1) = 2g(r) \mathbf{Q}_{n,h}^{(2)}(r) + O(h(r+h)^{d-1}), \quad (\text{A.22})$$

$$(m|D_n|h) \text{Var}[\mathbf{Z}_2(\boldsymbol{\theta}^*)] = \frac{2}{m-1} g^2(r) \mathbf{Q}_{n,h}^{(2)}(r) + O(h(r+h)^{d-1}), \quad (\text{A.23})$$

$$(m|D_n|h) \text{Cov}[\mathbf{Z}_1, \mathbf{Z}_2(\boldsymbol{\theta}^*)] = O(h(r+h)^{d-1}), \quad (\text{A.24})$$

where  $\mathbf{Q}_{n,h}^{(2)}(r)$  is as defined in equation (11) and the convergence is entry-wise.

*Proof.* Under conditions C1-C5, using similar arguments as those in the proof of Lemma A.2,

we can immediately show that

$$\begin{aligned}
\frac{m\text{Var}(\mathbf{Z}_1)}{(r+h)^{d-1}} &= (r+h)^{1-d} \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)[g^{(4)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2, \mathbf{v}_2) \\
&\quad - g(\|\mathbf{u}_1 - \mathbf{v}_1\|)g(\|\mathbf{u}_2 - \mathbf{v}_2\|)]\mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)\mathbf{A}_h^T(\|\mathbf{u}_2 - \mathbf{v}_2\| - r)d\mathbf{u}_1d\mathbf{v}_1d\mathbf{u}_2d\mathbf{v}_2 \\
&\quad + 4(r+h)^{1-d} \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, u_2) \\
&\quad \times \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)\mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{u}_2\| - r)d\mathbf{u}_1d\mathbf{v}_1d\mathbf{u}_2 \\
&\quad + 2 \int_{D_n^2} \frac{\lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)}{(r+h)^{d-1}} [w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 g(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)\mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)d\mathbf{u}_1d\mathbf{v}_1 \\
&= 2 \int_{D_n^2} \frac{\lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)}{(r+h)^{d-1}} [w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 g(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)\mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)d\mathbf{u}_1d\mathbf{v}_1 \\
&\quad + O(|D_n|^{-1}) \\
&= 2g(r) \frac{\{1 + O(h)\}}{(r+h)^{d-1}} \int_{D_n^2} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1) [w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)\mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{v}_1\| - r)d\mathbf{u}_1d\mathbf{v}_1 \\
&\quad + O(|D_n|^{-1}),
\end{aligned}$$

where the last equality follows from C4 and the fact that  $|g(t) - g(r)| = O(h)$  for any  $r - h \leq t \leq r + h$  as  $h \rightarrow 0$ . Similarly, we can show that under conditions C1-C5, we have

that

$$\begin{aligned}
\frac{m \text{Var} [\mathbf{Z}_2(\boldsymbol{\theta}^*)]}{(r+h)^{d-1}} &= \frac{2(r+h)^{1-d}}{m-1} \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) g(\|\mathbf{u}_1 - \mathbf{v}_1\|) \\
&\quad \times \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) d\mathbf{u}_1 d\mathbf{v}_1 \\
&+ \frac{4(r+h)^{1-d}}{m-1} \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*) g(\|\mathbf{v}_1 - \mathbf{u}_2\|) \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{u}_2\| - r) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ \frac{4(m-2)}{m-1} \int_{D_n^3} \frac{\lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2)}{(r+h)^{d-1}} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{u}_2\| - r) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ \frac{2}{m-1} \int_{D_n^4} \frac{\lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2)}{(r+h)^{d-1}} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1] \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_2 - \mathbf{v}_2\| - r) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ \frac{4(m-2)}{m-1} \int_{D_n^4} \frac{\lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2)}{(r+h)^{d-1}} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1] \\
&\quad \times \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_2 - \mathbf{v}_2\| - r) d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&= \int_{D_n^2} \frac{2\lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1)}{m-1} \frac{w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)}{(r+h)^{d-1}} \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) d\mathbf{u}_1 d\mathbf{v}_1 + O(|D_n|^{-1}) \\
&= \frac{2g^2(r) \{1 + O(h)\}}{(r+h)^{d-1}} \int_{D_n^2} \frac{\lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1)}{m-1} w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) d\mathbf{u}_1 d\mathbf{v}_1 + O(|D_n|^{-1}),
\end{aligned}$$

where the last equality holds because (A.2) implies that for any  $r-h \leq t \leq r+h$ ,

$$|\tilde{g}_{t,h}^2(t; \boldsymbol{\theta}^*) - g^2(r)| = g^2(r) |\exp(2 \{\log \tilde{g}_{r,h}(t; \boldsymbol{\theta}^*) - f(r)\}) - 1| = O(h^{p+1}),$$

and that

$$\begin{aligned}
\frac{m\text{Cov}[\mathbf{Z}_1, \mathbf{Z}_2(\boldsymbol{\theta}^*)]}{(r+h)^{d-1}} &= 2 \times 2 \int_{D_n^3} \frac{\lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)}{(r+h)^{d-1}} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{u}_2\| - r)^T d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
+ 2 \int_{D_n^4} &\frac{\lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)}{(r+h)^{d-1}} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_2 - \mathbf{v}_2\| - r) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
- 2 \int_{D_n^4} &\frac{\lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)}{(r+h)^{d-1}} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|) g(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times \mathbf{A}_h(\|\mathbf{u}_1 - \mathbf{v}_1\| - r) \mathbf{A}_h^T(\|\mathbf{u}_2 - \mathbf{v}_2\| - r) d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&= O(|D_n|^{-1}).
\end{aligned}$$

Combining above three equalities, we can conclude that as  $h \rightarrow 0$

$$\begin{aligned}
(m|D_n|h)\text{Var}(\mathbf{Z}_1) &= 2g(r)\mathbf{Q}_{n,h}^{(2)}(r) + O(h(r+h)^{d-1}), \\
(m|D_n|h)\text{Var}[\mathbf{Z}_2(\boldsymbol{\theta}^*)] &= \frac{2}{m-1}g^2(r)\mathbf{Q}_{n,h}^{(2)}(r) + O(h(r+h)^{d-1}), \\
(m|D_n|h)\text{Cov}[\mathbf{Z}_1, \mathbf{Z}_2(\boldsymbol{\theta}^*)] &= O(h(r+h)^{d-1}),
\end{aligned}$$

where  $\mathbf{Q}_{n,h}^{(2)}(r)$  is as defined in equation (11) and the convergence is entry-wise.  $\square$

**Lemma A.4.** *Under conditions C1-C5 and N1-N2, we have that, as  $h \rightarrow 0$  and  $m|D_n|h(r+h)^{d-1} \rightarrow \infty$ ,*

$$\sqrt{m|D_n|h}\boldsymbol{\Sigma}_Z^{-1/2}(\boldsymbol{\theta}^*)[\mathbf{Z}_1 - \mathbf{Z}_2(\boldsymbol{\theta}^*)] \xrightarrow{\mathcal{D}} N(\mathbf{0}, \mathbf{I}), \quad (\text{A.25})$$

where  $\boldsymbol{\Sigma}_Z(\boldsymbol{\theta}^*) = 2(m-1+g(r))/(m-1)g(r)\mathbf{Q}_{n,h}^{(2)}(r)$  with  $\mathbf{Q}_{n,h}^{(2)}(r)$  defined in equation (11).

*Proof.* By equation (A.24) of Lemma A.3, we can see that  $\mathbf{Z}_1$  and  $\mathbf{Z}_2(\boldsymbol{\theta}^*)$  are asymptotically uncorrelated as  $h \rightarrow 0$ . Hence, it suffices to consider asymptotic normality of  $(r+h)^{(1-d)/2}\mathbf{Z}_1$  and  $(r+h)^{(1-d)/2}\mathbf{Z}_2(\boldsymbol{\theta}^*)$  separately. We divide our discussions into two case scenarios: (1)  $m \rightarrow \infty$  and (2)  $m$  is fixed.

**Case I: when  $m \rightarrow \infty$ .** In this case, from equations (A.22)-(A.23), we can see that as  $m \rightarrow \infty$  and  $h \rightarrow 0$ ,

$$\sqrt{m|D_n|h} \{\mathbf{Z}_2(\boldsymbol{\theta}^*) - \mathbb{E}[\mathbf{Z}_2(\boldsymbol{\theta}^*)]\} = o_P((r+h)^{(d-1)/2}),$$

which implies that

$$\sqrt{m|D_n|h} [\mathbf{Z}_1 - \mathbf{Z}_2(\boldsymbol{\theta}^*)] = \sqrt{m|D_n|h} (\mathbf{Z}_1 - \mathbb{E}\mathbf{Z}_1) + o_P((r+h)^{(d-1)/2}),$$

since  $\mathbb{E}\mathbf{Z}_1 = \mathbb{E}[\mathbf{Z}_2(\boldsymbol{\theta}^*)]$ . Let  $\mathbf{Y}_i = (r+h)^{(1-d)/2} \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r)$ , then  $(r+h)^{(1-d)/2} \mathbf{Z}_1 = \frac{1}{m} \sum_{i=1}^m \mathbf{Y}_i$ . By definition,  $\mathbf{Y}_i$ 's are independent and identically distributed, thus it immediately follows from the standard multivariate central limit theorem that as  $h \rightarrow 0$  and  $m \rightarrow \infty$ ,

$$[\text{Var}(\mathbf{Z}_1)]^{-1/2} (\mathbf{Z}_1 - \mathbb{E}\mathbf{Z}_1) \xrightarrow{D} N(\mathbf{0}, \mathbf{I}),$$

which coincides with (A.25) after plugging (A.22) back to the above equation and use (A.23)-(A.24) to obtain the asymptotic variance of  $\sqrt{m|D_n|h} [\mathbf{Z}_1 - \mathbf{Z}_2(\boldsymbol{\theta}^*)]$ .

**Case II: when  $m$  is fixed.** In this case, condition  $m|D_n|h \rightarrow \infty$  requires that  $|D_n| \rightarrow \infty$ . In other words, we need to consider the case where the observation window of the point processes is expanding. Define a partition of  $\mathbb{R}^d = \cup_{\mathbf{t} \in \mathbb{Z}^d} \Delta_h(\mathbf{t})$ , where  $\Delta_h(\mathbf{t}) = \prod_{k=1}^d (h^{-1/d}(r+h)^{1/d-1}(t_k - 1/2), h^{-1/d}(r+h)^{1/d-1}(t_k + 1/2)]$ . Note that by this definition,  $\Delta_h(\mathbf{t}_1) \cap \Delta_h(\mathbf{t}_2) = \emptyset$  if  $\mathbf{t}_1 \neq \mathbf{t}_2 \in \mathbb{Z}^d$ . Define random vectors

$$\mathbf{Y}_{1,n}(\mathbf{t}) = \frac{|D_n|h}{m} \sum_{i=1}^m \sum_{\mathbf{u} \in X_i \cap \Delta_h(\mathbf{t}), \mathbf{v} \in X_i}^{\neq} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r),$$

$$\mathbf{Y}_{2,n}(\mathbf{t}) = \frac{|D_n|h}{m(m-1)} \sum_{i \neq j} \sum_{\mathbf{u} \in X_i \cap \Delta_h(\mathbf{t}), \mathbf{v} \in X_j} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r).$$

Then by definition, we have that

$$\mathbf{Z}_1 = \frac{1}{|D_n|h} \sum_{\mathbf{t} \in \mathcal{T}_n} \mathbf{Y}_{1,n}(\mathbf{t}), \quad \mathbf{Z}_2(\boldsymbol{\theta}^*) = \frac{1}{|D_n|h} \sum_{\mathbf{t} \in \mathcal{T}_n} \mathbf{Y}_{2,n}(\mathbf{t}),$$

where  $\mathcal{T}_n = \{\mathbf{t} \in \mathbb{Z}^d : \Delta_h(\mathbf{t}) \cap D_n \neq \emptyset\}$ .

Under conditions C4-C5, it is straightforward to see that there exists a constant  $C_1$  such that

$$|D_n| h w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) |\mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r)| \leq C_1 I(\|\mathbf{u} - \mathbf{v}\| - r < h).$$

A simple application of the Jensen's inequality gives that  $(m^{-1} \sum_{i=1}^m |x_i|)^{2+\lceil\delta\rceil} \leq m^{-1} \sum_{i=1}^m |x_i|^{2+\lceil\delta\rceil}$  (note that  $f(x) = x^{2+\lceil\delta\rceil}$  is convex for  $x > 0$ )

$$\begin{aligned} \mathbb{E} |\mathbf{Y}_{1,n}(t)|^{2+\lceil\delta\rceil} &\leq \frac{1}{m} \sum_{i=1}^m \mathbb{E} \left| \sum_{\mathbf{u} \in X_i \cap \Delta_h(t)} \sum_{\mathbf{v} \in X_i}^{\neq} |D_n| h w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) \right|^{2+\lceil\delta\rceil} \\ &= \mathbb{E} \left| \sum_{\mathbf{u} \in X_1 \cap \Delta_h(t)} \sum_{\mathbf{v} \in X_1}^{\neq} h K_h(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) \right|^{2+\lceil\delta\rceil} \\ &\leq \mathbb{E} \left\{ \sum_{\mathbf{u} \in X_1 \cap \Delta_h(t)} \sum_{\mathbf{v} \in X_1}^{\neq} h K_h(\|\mathbf{u} - \mathbf{v}\|) |\mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r)| \right\}^{2+\lceil\delta\rceil} \\ &\leq C_1^{2+\lceil\delta\rceil} \mathbb{E} \left\{ \sum_{\mathbf{u} \in X_1 \cap \Delta_h(t)} \sum_{\mathbf{v} \in X_1}^{\neq} I(\|\mathbf{u} - \mathbf{v}\| - r < h) \right\}^{2+\lceil\delta\rceil}, \end{aligned}$$

where the last expectation is essentially bounded by sums of integrals involving  $\lambda(\mathbf{u})$ ,  $g(s)$ ,

$g^{(k)}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ ,  $k = 3, \dots, 2(2 + \lceil\delta\rceil)$ . Specifically, note that

$$\begin{aligned} &\mathbb{E} \left[ \sum_{\mathbf{u} \in X_1 \cap \Delta_h(t)} \sum_{\mathbf{v} \in X_1}^{\neq} I(\|\mathbf{u} - \mathbf{v}\| - r < h) \right] \\ &= \int_{\Delta_h(t)} \int_{D_n} \lambda(\mathbf{u}) \lambda(\mathbf{v}) g_0(\|\mathbf{u} - \mathbf{v}\|) I(\|\mathbf{u} - \mathbf{v}\| - r < h) d\mathbf{u} d\mathbf{v} \\ &= \int_{\mathbb{R}^d} g_0(\|\mathbf{h}\|) I(\|\mathbf{h}\| - r < h) \left[ \int_{\Delta_h(t)} \lambda(\mathbf{u}) \lambda(\mathbf{u} - \mathbf{h}) I(\mathbf{u} - \mathbf{h} \in D_n) d\mathbf{u} \right] d\mathbf{h} \\ &= O(1) \int_{\mathbb{R}^d} g_0(\|\mathbf{h}\|) I(\|\mathbf{h}\| - r < h) \left[ \int_{\Delta_h(t)} I(\mathbf{u} - \mathbf{h} \in D_n) d\mathbf{u} \right] d\mathbf{h} \\ &= O(1) |\Delta_h(t)| \int_{\mathbb{R}^d} g_0(\|\mathbf{h}\|) I(\|\mathbf{h}\| - r < h) d\mathbf{h} \\ &= O(1) h^{-1} (r + h)^{1-d} \int_0^\infty I(|s - r| < h) s^{d-1} ds = O(1). \end{aligned}$$

All other terms can be similarly shown to be of the same order under conditions C1-C3 and

condition N2, hence these integrals bounded for any  $d \geq 1$  and uniformly in  $t$  and  $n$ . Recall that  $\delta$  is defined in condition N2. Therefore, we have that

$$\sup_{n \geq 1} \sup_{\mathbf{t} \in \mathcal{T}_n} \mathbb{E} |\mathbf{Y}_{1,n}(\mathbf{t})|^{2+\delta} \leq \left( \sup_{n \geq 1} \sup_{\mathbf{t} \in \mathcal{T}_n} \mathbb{E} |\mathbf{Y}_{1,n}(\mathbf{t})|^{2+\lceil \delta \rceil} \right)^{\frac{2+\delta}{2+\lceil \delta \rceil}} < \infty. \quad (\text{A.26})$$

Similarly, using equation (A.3) in Lemma A.1 and condition C2(a), we have that  $\tilde{g}_{r,h}(t; \boldsymbol{\theta}^*)$  is also uniformly bounded and following similar arguments as above, we can show that

$$\sup_{n \geq 1} \sup_{\mathbf{t} \in \mathcal{T}_n} \mathbb{E} |\mathbf{Y}_{2,n}(\mathbf{t})|^{2+\delta} \leq \left( \sup_{n \geq 1} \sup_{\mathbf{t} \in \mathcal{T}_n} \mathbb{E} |\mathbf{Y}_{2,n}(\mathbf{t})|^{2+\lceil \delta \rceil} \right)^{\frac{2+\delta}{2+\lceil \delta \rceil}} < \infty. \quad (\text{A.27})$$

Note that the total number of disjoint partitions  $\Delta_h(\mathbf{t}) \cap D_n \neq \emptyset$  is of the order  $|D_n|h(r+h)^{d-1}$ , hence we can check that, using equations (A.22)-(A.23),

$$\begin{aligned} \frac{\text{Var} \left[ \sum_{t \in \mathcal{T}_n} \mathbf{Y}_{1,n}(t) \right]}{|D_n|h(r+h)^{d-1}} &= \frac{\text{Var} (|D_n|h \mathbf{Z}_1)}{|D_n|h(r+h)^{d-1}} = |D_n|h(r+h)^{1-d} \text{Var} (\mathbf{Z}_1), \\ \frac{\text{Var} \left[ \sum_{t \in \mathcal{T}_n} \mathbf{Y}_{2,n}(t) \right]}{|D_n|h(r+h)^{d-1}} &= \frac{\text{Var} [|D_n|h \mathbf{Z}_2(\boldsymbol{\theta}^*)]}{|D_n|h(r+h)^{d-1}} = |D_n|h(r+h)^{1-d} \text{Var} [\mathbf{Z}_2(\boldsymbol{\theta}^*)], \end{aligned}$$

both of above matrices have strictly positive eigenvalues under condition C5 and Lemma A.3. Therefore, using conditions N1(b) and N2, together with inequalities (A.26)-(A.27), it follows from Theorem 1 of Biscio and Waagepetersen (2019) that as  $|D_n|h(r+h)^{d-1} \rightarrow \infty$ ,

$$\left\{ \text{Var} \left[ \sum_{t \in \mathcal{T}_n} \mathbf{Y}_{k,n}(t) \right] \right\}^{-1/2} \sum_{t \in \mathcal{T}_n} [\mathbf{Y}_{k,n}(t) - \mathbb{E} \mathbf{Y}_{k,n}(t)] \xrightarrow{\mathcal{D}} N(\mathbf{0}, \mathbf{I}), \quad k = 1, 2,$$

which is equivalent to stating that

$$[\text{Var}(\mathbf{Z}_1)]^{-1/2} (\mathbf{Z}_1 - \mathbb{E} \mathbf{Z}_1) \xrightarrow{\mathcal{D}} N(\mathbf{0}, \mathbf{I}),$$

and

$$\{\text{Var} [\mathbf{Z}_2(\boldsymbol{\theta}^*)]\}^{-1/2} [\mathbf{Z}_2(\boldsymbol{\theta}^*) - \mathbb{E} \mathbf{Z}_2(\boldsymbol{\theta}^*)] \xrightarrow{\mathcal{D}} N(\mathbf{0}, \mathbf{I}).$$

Recall that by Lemma A.3,  $\mathbf{Z}_1$  and  $\mathbf{Z}_2(\boldsymbol{\theta}^*)$  have finite variances and are asymptotically independent as  $h \rightarrow 0$ , and that  $\mathbb{E} \mathbf{Z}_1 = \mathbb{E} [\mathbf{Z}_2(\boldsymbol{\theta}^*)]$  by definition, we can conclude that

$$\{\text{Var}(\mathbf{Z}_1) + \text{Var} [\mathbf{Z}_2(\boldsymbol{\theta}^*)]\}^{-1/2} [\mathbf{Z}_1 - \mathbf{Z}_2(\boldsymbol{\theta}^*)] \xrightarrow{\mathcal{D}} N(\mathbf{0}, \mathbf{I}),$$



which coincides with (A.25) after plugging (A.22)-(A.23) back to the above equation. The proof is complete.  $\square$

**Lemma A.5.** Denote  $\widehat{\boldsymbol{\theta}}$  as the solution to estimating equations (6), then under conditions C1-C5, N1-N2, we have that, as  $h \rightarrow 0$  and  $m|D_n|h(r+h)^{d-1} \rightarrow \infty$ ,

$$\mathbf{D}_h(\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*) \equiv \begin{bmatrix} (\widehat{\theta}_0 - \theta_0^*) \\ h(\widehat{\theta}_1 - \theta_1^*) \\ \vdots \\ h^p(\widehat{\theta}_p - \theta_p^*) \end{bmatrix} = \left[ g(r) \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \left[ \mathbf{Z}_1 - \mathbf{Z}_2(\boldsymbol{\theta}^*) + o_p \left( \sqrt{\frac{(r+h)^{1-d}}{m|D_n|h}} \right) \right] \quad (\text{A.28})$$

where  $\mathbf{Z}_1$  and  $\mathbf{Z}_2(\boldsymbol{\theta}^*)$  are defined in (A.19) and (A.20), respectively.

*Proof.* By the definition of  $\widetilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta})$  in (6), since  $\mathbf{G}_r(t) = \mathbf{D}_h \mathbf{A}_h(t-r)$ , solving  $\widetilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta}) = \mathbf{0}$  is equivalent to solving  $\widetilde{\mathbf{V}}_{r,h}(\boldsymbol{\theta}) = \mathbf{0}$  for  $\boldsymbol{\theta}$ , where  $\widetilde{\mathbf{V}}_{r,h}(\boldsymbol{\theta}) = \mathbf{D}_h \widetilde{\mathbf{U}}_{r,h}(\boldsymbol{\theta})$ ; i.e.,

$$\begin{aligned} \widetilde{\mathbf{V}}_{r,h}(\boldsymbol{\theta}) &= \frac{1}{m} \sum_{i=1}^m \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) \\ &\quad - \frac{1}{m(m-1)} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} w_{r,h}(\|\mathbf{u} - \mathbf{v}\|) \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) \widetilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}). \end{aligned}$$

Using the first order Taylor expansion, we can show that

$$\underbrace{\widetilde{\mathbf{V}}_{r,h}(\widehat{\boldsymbol{\theta}})}_{=0} - \widetilde{\mathbf{V}}_{r,h}(\boldsymbol{\theta}^*) = -\widetilde{\mathbf{H}}_{h,r}(\widetilde{\boldsymbol{\theta}}^*) \mathbf{D}_h(\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*), \quad (\text{A.29})$$

where  $\widetilde{\boldsymbol{\theta}}^*$  satisfies  $\|\widetilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h \leq \|\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|_h$  with  $\|\cdot\|_h$  as defined in Lemma A.2 and

$$\widetilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}) = \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_{r,h}(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r) \mathbf{A}_h^T(\|\mathbf{u} - \mathbf{v}\| - r) \widetilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}). \quad (\text{A.30})$$

By definition, we have that for any  $r-h \leq t \leq r+h$ ,

$$\begin{aligned} |\widetilde{g}_{r,h}(t; \boldsymbol{\theta}^*) - \widetilde{g}_{r,h}(t; \widetilde{\boldsymbol{\theta}}^*)| &= \widetilde{g}_{r,h}(t; \boldsymbol{\theta}^*) \left| 1 - \exp \left[ \theta_0^* - \widetilde{\theta}_0^* + h(\theta_1^* - \widetilde{\theta}_1^*) \frac{t-r}{h} \cdots + h^p(\theta_p^* - \widetilde{\theta}_p^*) \frac{(t-r)^p}{h^p} \right] \right| \\ &\leq \widetilde{g}_{r,h}(t; \boldsymbol{\theta}^*) \sqrt{p+1} \|\widetilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h \exp \left( \sqrt{p+1} \|\widetilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h \right), \end{aligned}$$

where the last inequality follows from the fact that  $|1 - e^x| \leq |x|e^{|x|}$  and Cauchy-Schwarz inequality. Since  $\|\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h \leq \|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|_h = O_p\left(1/\sqrt{m|D_n|h(r+h)^{d-1}}\right)$  by Lemma A.2, we have that

$$\begin{aligned}
\eta_{\max} \left[ \tilde{\mathbf{H}}_{h,r}(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] &= \sup_{\|\boldsymbol{\delta}\|=1} \boldsymbol{\delta}^T \left[ \tilde{\mathbf{H}}_{h,r}(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] \boldsymbol{\delta} \\
&\leq \sup_{\|\boldsymbol{\delta}\|=1} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_{r,h}(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} [\boldsymbol{\delta}^T \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r)]^2 \left| \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \tilde{\boldsymbol{\theta}}^*) - \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \right| \\
&\leq \sup_{\|\boldsymbol{\delta}\|=1} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_{r,h}(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} [\boldsymbol{\delta}^T \mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r)]^2 \tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \\
&\quad \times \sqrt{p+1} \|\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h \exp\left(\sqrt{p+1} \|\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h\right) \\
&= \eta_{\max} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] \times \sqrt{p+1} \|\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h \exp\left(\sqrt{p+1} \|\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\|_h\right) \\
&= \eta_{\max} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] O_p\left(1/\sqrt{m|D_n|h(r+h)^{d-1}}\right).
\end{aligned}$$

Following exactly the same steps, we can also show that

$$\begin{aligned}
-\eta_{\min} \left[ \tilde{\mathbf{H}}_{h,r}(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] &= \eta_{\max} \left[ -\tilde{\mathbf{H}}_{h,r}(\tilde{\boldsymbol{\theta}}^*) + \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] \\
&= \eta_{\max} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] O_p\left(1/\sqrt{m|D_n|h(r+h)^{d-1}}\right),
\end{aligned}$$

which implies that

$$\tilde{\mathbf{H}}_{h,r}(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) = \eta_{\max} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] O_p\left(1/\sqrt{m|D_n|h(r+h)^{d-1}}\right), \quad (\text{A.31})$$

where the convergence is entry-wise.

The next step is to quantify the variabilities of entries in  $\tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*)$ , denoted as  $H_{ij}$ 's. Following steps as those in the proof of Lemma A.2 about  $\text{Var} [H''_{m,n}(z_0)]$ , under conditions

C1, C2(a)-(b), C4 and equation (A.3), some tedious algebra give that

$$\begin{aligned}
m(m-1)\text{Var}(H_{ij}) &= O(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)|g(\|\mathbf{u}_1 - \mathbf{u}_2\|)g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1| \\
&\quad \mathbf{d}\mathbf{u}_1\mathbf{d}\mathbf{v}_1\mathbf{d}\mathbf{u}_2\mathbf{d}\mathbf{v}_2 \\
&+ O(1) \int_{D_n^3} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)\mathbf{d}\mathbf{u}_1\mathbf{d}\mathbf{v}_1\mathbf{d}\mathbf{u}_2 + O(1) \int_{D_n^2} w_{r,h}^2(\|\mathbf{u}_1 - \mathbf{v}_1\|)\mathbf{d}\mathbf{u}_1\mathbf{d}\mathbf{v}_1 \\
&+ mO(1) \int_{D_n^4} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_2 - \mathbf{v}_2\|)|g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1|\mathbf{d}\mathbf{u}_1\mathbf{d}\mathbf{u}_2\mathbf{d}\mathbf{v}_1\mathbf{d}\mathbf{v}_2 \\
&+ mO(1) \int_{D_n^3} w_{r,h}(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_{r,h}(\|\mathbf{u}_1 - \mathbf{u}_2\|)\mathbf{d}\mathbf{u}_1\mathbf{d}\mathbf{v}_1\mathbf{d}\mathbf{u}_2 \\
&= O(1)|D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)|g(\|\mathbf{w}\|)g(\|\mathbf{t} - \mathbf{s} + \mathbf{w}\|) - 1|\mathbf{d}\mathbf{s}\mathbf{d}\mathbf{t}\mathbf{d}\mathbf{w} \\
&+ O(1)|D_n|^{-1} \int_{D_n^2} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)\mathbf{d}\mathbf{s}\mathbf{d}\mathbf{t} + O(1)|D_n|^{-1} \int_{D_n} [K_h(\|\mathbf{s}\| - r)]^2 \mathbf{d}\mathbf{s} \\
&+ mO(1)|D_n|^{-1} \int_{D_n^3} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)|g(\|\mathbf{w}\|) - 1|\mathbf{d}\mathbf{s}\mathbf{d}\mathbf{t}\mathbf{d}\mathbf{w} \\
&+ mO(1)|D_n|^{-1} \int_{D_n^2} K_h(\|\mathbf{s}\| - r)K_h(\|\mathbf{t}\| - r)\mathbf{d}\mathbf{s}\mathbf{d}\mathbf{t} \\
&= O(1)|D_n|^{-1} \int_0^\infty [K_h(s - r)]^2 s^{d-1} \mathbf{d}s + mO(1)|D_n|^{-1} \left( \int_0^\infty K_h(s - r) s^{d-1} \mathbf{d}s \right)^2 \\
&= O(1)|D_n|^{-1} h^{-1} \int_{-r/h}^\infty [K(s)]^2 (r + sh)^{d-1} \mathbf{d}s + mO(1)|D_n|^{-1} \left( \int_{-r/h}^\infty K(s)(r + sh)^{d-1} \mathbf{d}s \right)^2
\end{aligned}$$

Then, by condition C4, we finally have that

$$\text{Var}(H_{ij}) = \frac{(r+h)^{d-1}}{m^2|D_n|h} O(1) + \frac{(r+h)^{2d-2}}{m|D_n|} O(1) \rightarrow 0, \text{ as } m|D_n|h \rightarrow \infty,$$

which gives that as  $m|D_n|h \rightarrow \infty$ ,

$$(r+h)^{1-d} \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) = (r+h)^{1-d} \mathbb{E} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] + o_p(1). \quad (\text{A.32})$$

Next, we study  $\mathbb{E} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right]$ . By definition

$$\mathbb{E} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] = \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_{r,h}(\|\mathbf{u} - \mathbf{v}\|)\mathbf{A}_h(\|\mathbf{u} - \mathbf{v}\| - r)\mathbf{A}_h^T(\|\mathbf{u} - \mathbf{v}\| - r)\tilde{g}_{r,h}(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*)\mathbf{d}\mathbf{u}\mathbf{d}\mathbf{v}.$$

Recall the definition of  $\mathbf{Q}_{n,h}^{(1)}(r)$  in equation (13) and the fact that for any  $r - h \leq t \leq r + h$ ,

$|\tilde{g}_{r,h}(t; \boldsymbol{\theta}^*) - g(r)| = g(r)O(h^{p+1})$  by Lemma A.1, following the similar proof as that of

equation (A.31), we have that

$$(r+h)^{1-d} \left\{ \mathbb{E} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] - g(r) \mathbf{Q}_{n,h}^{(1)}(r) \right\} = (r+h)^{1-d} g(r) \eta_{\max} \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right] O(h). \quad (\text{A.33})$$

Using condition C5, equations (A.32)-(A.33) shows that  $\eta_{\max} \left[ \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) \right] = o_p((r+h)^{d-1})$ , which further implies that using (A.31), one has that

$$\tilde{\mathbf{H}}_{h,r}(\tilde{\boldsymbol{\theta}}^*) = \tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) + o_p \left( \frac{\sqrt{(r+h)^{1-d}}}{\sqrt{m|D_n|h}} \right).$$

Furthermore, under condition C5, equations (A.32)-(A.33) also implies that  $\tilde{\mathbf{H}}_{h,r}(\boldsymbol{\theta}^*) = g(r) \mathbf{Q}_{n,h}^{(1)}(r) + O_p(h(r+h)^{d-1}) + o_p((r+h)^{d-1})$  and hence that

$$\tilde{\mathbf{H}}_{h,r}(\tilde{\boldsymbol{\theta}}^*) = g(r) \mathbf{Q}_{n,h}^{(1)}(r) + O_p(h(r+h)^{d-1}) + o_p((r+h)^{d-1}) + o_p \left( \frac{\sqrt{(r+h)^{1-d}}}{\sqrt{m|D_n|h}} \right).$$

Plugging the above equality back to equation (A.29), one has that

$$\tilde{\mathbf{V}}_{r,h}(\boldsymbol{\theta}^*) = \left\{ g(r) \mathbf{Q}_{n,h}^{(1)}(r) + o_p((r+h)^{d-1}) + o_p \left( \frac{\sqrt{(r+h)^{1-d}}}{\sqrt{m|D_n|h}} \right) \right\} \mathbf{D}_h(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*),$$

which further gives that, under conditions C2(a), C5 and use Lemma A.2, one has

$$\mathbf{D}_h(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*) = \left[ g(r) \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \left[ \tilde{\mathbf{V}}_{r,h}(\boldsymbol{\theta}^*) + o_p \left( \frac{1}{\sqrt{m|D_n|h(r+h)^{d-1}}} \right) \right]. \quad (\text{A.34})$$

The proof is completed by observing that the definition of  $\mathbf{Z}_1$  and  $\mathbf{Z}_2(\boldsymbol{\theta}^*)$  gives that  $\tilde{\mathbf{V}}_{r,h}(\boldsymbol{\theta}^*) = [\mathbf{Z}_1 - \mathbf{Z}_2(\boldsymbol{\theta}^*)]$ .  $\square$

**Proof of Theorem 1.** By applying the delta method to  $\hat{g}_h(r) = \exp(\hat{\theta}_0) = \exp(e^T \hat{\boldsymbol{\theta}})$ , where  $\mathbf{e} = (1, 0, \dots, 0)^T$ , with Lemmas A.4 and A.5, we have that

$$\frac{\sqrt{m|D_n|h} [\hat{g}_h(r) - \exp(\theta_0^*)]}{\exp(\theta_0^*) \sqrt{2(m-1+g(r))/(m-1)[g(r)]^{-1} \mathbf{e}^T \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{Q}_{n,h}^{(2)}(r) \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{e}}} \xrightarrow{\mathcal{D}} N(0, 1).$$

By equation (A.6) in the proof of Lemma A.1, we have that  $\exp(\theta_0^*) - g(r) = O(h^{p+1})$ .

Therefore, it readily follows that

$$\begin{aligned}
& \frac{\sqrt{m|D_n|h} [\hat{g}_h(r) - g(r)]}{\exp(\theta_0^*) \sqrt{2(m-1+g(r))/(m-1)[g(r)]^{-1}} \mathbf{e}^T \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{Q}_{n,h}^{(2)}(r) \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{e}} \\
&= \frac{\sqrt{m|D_n|h} [\hat{g}_h(r) - \exp(\theta_0^*) + \exp(\theta_0^*) - g(r)]}{\sqrt{2(m-1+g(r))/(m-1)g(r)} \mathbf{e}^T \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{Q}_{n,h}^{(2)}(r) \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{e}} + o_p(1) \\
&= \frac{\sqrt{m|D_n|h} [\hat{g}_h(r) - \exp(\theta_0^*) + O(h^{p+1})]}{\sqrt{2(m-1+g(r))/(m-1)g(r)} \mathbf{e}^T \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{Q}_{n,h}^{(2)}(r) \left[ \mathbf{Q}_{n,h}^{(1)}(r) \right]^{-1} \mathbf{e}} + o_p(1),
\end{aligned}$$

which completes the proof.  $\square$

### 3 Asymptotic Properties of the Orthogonal Series Estimator

In this Section, we give detailed proofs of Lemma 3 and Theorem 2.

#### 3.1 Conditions

The following conditions are needed for consistency of the orthogonal series estimator (10).

[C4'] For some  $\nu_1 > 0$ , the approximation error (14) satisfies (a)  $\int_0^R w_o(r) \tilde{\zeta}_L^2(r; \boldsymbol{\theta}_0) dr = \sum_{l=L+1}^{\infty} \theta_{0,l}^2 = O(L^{-2\nu_1})$ ; (b)  $\sup_{0 < r \leq R} |\tilde{\zeta}_L(r; \boldsymbol{\theta}_0)| = O(L^{-\nu_1 + \tau_1})$  for some  $0 < \tau_1 < \nu_1$ ; (c)  $\sup_{0 < r \leq R} \|\boldsymbol{\phi}_L(r)\| = O(L^{\nu_2})$  for some  $0 \leq \nu_2 < \nu_1$ ; and (d) the weight function is uniformly bounded, i.e.,  $w_o(r) \leq C_w$  for any  $0 < r \leq R$ .

[C5'] As  $L \rightarrow \infty$ , there exist constants  $c_0, \nu_0$  where  $0 \leq 2\nu_0 < \nu_1 - \nu_2$ , such that

$$\eta_{\min}(\mathbf{Q}_L) > c_0 L^{-\nu_0},$$

where  $\eta_{\min}(\mathbf{Q})$  denotes the smallest eigenvalue of a matrix  $\mathbf{Q}$ .

The following additional conditions are needed for asymptotic normality.

[N1'] Either one of the following conditions are true (a)  $m \rightarrow \infty$ ; or (b) the mixing coefficient satisfies  $\alpha_X(s; 2, \infty) = O(s^{-d-\varepsilon'})$  for some  $\varepsilon' > 0$ .

[N2'] There exists  $\delta' > 2d/\varepsilon'$  such that  $|g^{(k)}(\mathbf{x}_1, \dots, \mathbf{x}_k)| \leq C_g$  for any  $\mathbf{x}_j \in D_n, j = 1, \dots, k, k = 2, \dots, 2(2 + \lceil \delta' \rceil)$ .

[N3] For  $r \in [0, R]$ , define the vector  $\boldsymbol{\ell}(r) = (\mathbf{Q}_L)^{-1} \boldsymbol{\phi}_L^T(r)$  and its standardized version  $\boldsymbol{\ell}_0(r) = \|\boldsymbol{\ell}(r)\|^{-1} \boldsymbol{\ell}(r)$ . Assume that as  $m|D_n| \rightarrow \infty$  and  $L \rightarrow \infty$ , (a) there exists some constant  $c_u > 0$  such that  $\boldsymbol{\ell}_0^T(r) \boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}_0(r) \geq c_u$  with  $\boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) = \text{Var} \left[ \sqrt{m|D_n|} \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) \right]$ ; and (b) the basis functions satisfy  $\int_0^R [w_o(s) |\boldsymbol{\ell}_0^T(r) \boldsymbol{\phi}_L(s)|]^{2+\lceil \delta' \rceil} ds \leq C_\phi$ , for some  $C_\phi > 0$ .

### 3.2 Sketch of the proof

**Step 1** We first derive the asymptotic limit of solutions to  $\tilde{\mathbf{U}}_L(\boldsymbol{\theta}) = \mathbf{0}$ , namely,  $\boldsymbol{\theta}^*$  defined in the (A.35) in the next subsection. As a result, Lemma A.6 gives the asymptotic bias of the orthogonal series estimator of  $g(r)$ .

**Step 2** Lemma A.7 gives the convergence rate of  $\hat{\boldsymbol{\theta}}$  to  $\boldsymbol{\theta}^*$ , which is of the order  $\|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\| = O_p \left( \frac{L^{\nu_2}}{\sqrt{m|D_n|}} \right)$ ;

**Step 3** Find the asymptotic normality of  $\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*$  through Lemmas A.8 to A.9, following the approach proposed in Biscio and Waagepetersen (2019) .

### 3.3 The asymptotic bias

Let  $\boldsymbol{\theta}_0 = (\theta_{0,1}, \dots, \theta_{0,L})$ , where  $\theta_{0,l}$ 's are the first  $L$  coefficients of the orthogonal series expansion of  $g(r)$  with respect to the basis functions  $\phi_l(r)$ 's and suppose there exists a

vector  $\boldsymbol{\theta}^* = (\theta_1^*, \dots, \theta_L^*)^T$  such that

$$\int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|) [g(\|\mathbf{u} - \mathbf{v}\|) - \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*)] \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) d\mathbf{u}d\mathbf{v} = \mathbf{0}. \quad (\text{A.35})$$

The following Lemma quantifies the distance between  $g(r)$  and  $\tilde{g}_L(r; \boldsymbol{\theta}^*)$ .

**Lemma A.6.** *Under conditions C1-C3 and C4'-C5', we have that as  $L \rightarrow \infty$ ,*

$$\|\boldsymbol{\theta}_0 - \boldsymbol{\theta}^*\| = O(L^{\nu_0 - \nu_1}), \quad (\text{A.36})$$

$$\sup_{0 < r < R} |g(r) - \tilde{g}_L(r; \boldsymbol{\theta}^*)| = O(L^{-\nu_1 + \max\{\tau_1, \nu_0 + \nu_2\}}) = o(1), \quad (\text{A.37})$$

$$\sup_{0 < r < R} |\tilde{g}_L(r; \boldsymbol{\theta}^*)| = O(1), \quad (\text{A.38})$$

where  $\nu_0, \nu_1, \tau_1$  and  $\nu_2$  are defined in conditions C4' and C5'.

*Proof.* It is straightforward to see that by definition,  $\boldsymbol{\theta}^*$  is the solution to (A.35), which also maximizes the following target function

$$\ell(\boldsymbol{\theta}) = \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|) \{g(\|\mathbf{u} - \mathbf{v}\|)\boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) - \exp[\boldsymbol{\theta}^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|)]\} d\mathbf{u}d\mathbf{v}.$$

Let  $\boldsymbol{\nu}_n$  be an arbitrary sequence on the sphere  $\{\boldsymbol{\nu} \in \mathbb{R}^L : \|\boldsymbol{\nu}\| = L^{-\nu_1 + \nu_0}\}$  and define functions

$\Delta_n(r) = \boldsymbol{\nu}_n^T \boldsymbol{\phi}_L(r)$ . Let  $f_0(r) = \boldsymbol{\theta}_0^T \boldsymbol{\phi}_L(r)$ ,  $0 < r < R$ , and define the function of a scalar  $z$  as

follows

$$h(z) = \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|) \times \{g(\|\mathbf{u} - \mathbf{v}\|) [f_0(\|\mathbf{u} - \mathbf{v}\|) + z\Delta_n(\|\mathbf{u} - \mathbf{v}\|)] - \exp[f_0(\|\mathbf{u} - \mathbf{v}\|) + z\Delta_n(\|\mathbf{u} - \mathbf{v}\|)]\} d\mathbf{u}d\mathbf{v}.$$

We shall show that for any  $z_0 > 0$ ,  $h'(z_0) < 0$  and  $h'(-z_0) > 0$ . This implies that the maximum of  $\ell(\boldsymbol{\theta})$ , namely  $\boldsymbol{\theta}^*$ , satisfies  $f_0(r) - z_0\Delta_n(r) \leq \boldsymbol{\theta}^{*T} \boldsymbol{\phi}_L(r) \leq f_0(r) + z_0\Delta_n(r)$ , using

the fact that  $\ell(\boldsymbol{\theta})$  is a concave function of  $\boldsymbol{\theta}$ . Some straightforward calculus gives that

$$\begin{aligned}
h'(z) &= \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|)g(\|\mathbf{u} - \mathbf{v}\|)\Delta_n(\|\mathbf{u} - \mathbf{v}\|) \\
&\quad \times \left\{ 1 - \exp \left[ -\tilde{\zeta}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}_0) + z\Delta_n(\|\mathbf{u} - \mathbf{v}\|) \right] \right\} d\mathbf{u}d\mathbf{v} \\
&= \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|)g(\|\mathbf{u} - \mathbf{v}\|)\Delta_n(\|\mathbf{u} - \mathbf{v}\|) \left\{ 1 - \exp \left[ -\tilde{\zeta}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}_0) \right] \right\} d\mathbf{u}d\mathbf{v} \\
&+ \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|)g(\|\mathbf{u} - \mathbf{v}\|)\Delta_n(\|\mathbf{u} - \mathbf{v}\|) \exp \left[ -\tilde{\zeta}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}_0) \right] \\
&\quad \times \left\{ 1 - \exp \left[ z\Delta_n(\|\mathbf{u} - \mathbf{v}\|) \right] \right\} d\mathbf{u}d\mathbf{v} \\
&= \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|)g(\|\mathbf{u} - \mathbf{v}\|)\Delta_n(\|\mathbf{u} - \mathbf{v}\|)\tilde{\zeta}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}_0) [1 + o(1)] d\mathbf{u}d\mathbf{v} \\
&+ \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|)g(\|\mathbf{u} - \mathbf{v}\|)\Delta_n(\|\mathbf{u} - \mathbf{v}\|) [-z\Delta_n(\|\mathbf{u} - \mathbf{v}\|)] [1 + o(1)] d\mathbf{u}d\mathbf{v},
\end{aligned}$$

where  $\tilde{\zeta}_L(r; \boldsymbol{\theta}_0)$  is the approximation error defined in equation (13). The last equation follows from the Taylor expansion  $1 - e^x = -x [1 + e^{x^*}x/2]$ , for some  $|x^*| < |x|$ , and the condition C4', which ensures that as  $L \rightarrow \infty$ ,

$$\begin{aligned}
\sup_{0 < r < R} |\zeta_L(r; \boldsymbol{\theta}_0)| &= O(L^{\tau_1 - \nu_1}) = o(1), \\
\sup_{0 < r < R} |\Delta_n(r)| &\leq \|\boldsymbol{\nu}_n\| \sup_{0 < r < R} \|\boldsymbol{\phi}_L(r)\| = O(L^{\nu_0 + \nu_2 - \nu_1}) = o(1).
\end{aligned}$$

When  $z > 0$ , using conditions C1-C3, C4' and the Hölder's inequality, we can derive that

$$\begin{aligned}
h'(z) &\leq O(1) \int_0^R w_o(s)g(s)\Delta_n(s)\tilde{\zeta}_L(s; \boldsymbol{\theta}_0)s^{d-1}ds \\
&\quad - z [1 + o(1)] \boldsymbol{\nu}_n^T \underbrace{\left\{ \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|)g(\|\mathbf{u} - \mathbf{v}\|)\boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|)\boldsymbol{\phi}_L^T(\|\mathbf{u} - \mathbf{v}\|)d\mathbf{u}d\mathbf{v} \right\}}_{Q_L \text{ defined in (13)}} \boldsymbol{\nu}_n \\
&\leq O(1) \sqrt{\int_0^R w_o(s)\Delta_n^2(s)s^{d-1}ds} \sqrt{\int_0^R w_o(s)\tilde{\zeta}_L^2(s; \boldsymbol{\theta}_0)ds} - z [1 + o(1)] \|\boldsymbol{\nu}_n\|^2 \times \eta_{\min} [Q_L] \\
&\leq O(1) \sqrt{\|\boldsymbol{\nu}_n\|^2 \int_0^R w_o(s)\|\boldsymbol{\phi}_L(s)\|^2 s^{d-1}ds} \sqrt{\int_0^R w_o(s)\tilde{\zeta}_L^2(s; \boldsymbol{\theta}_0)ds} - z [1 + o(1)] \|\boldsymbol{\nu}_n\|^2 \times \eta_{\min} [Q_L] \\
&= O(L^{-\nu_1}) \|\boldsymbol{\nu}_n\| - z [1 + o(1)] \|\boldsymbol{\nu}_n\|^2 \times \eta_{\min} [Q_L].
\end{aligned}$$



Finally, under condition C5,  $\|\boldsymbol{\nu}_n\| = L^{-\nu_1+\nu_0}$  is sufficient to ensure that there exists a  $z_0 > 0$  such that  $h'(z_0) < 0$ .

Similarly,  $\|\boldsymbol{\nu}_n\| = L^{-\nu_1+\nu_0}$  is sufficient to ensure that

$$\begin{aligned}
h'(-z_0) &\geq -O(1) \int_0^R w_o(s)g(s)|\Delta_n(s)|\tilde{\zeta}_L(s; \boldsymbol{\theta}_0)|s^{d-1}ds \\
&\quad + z_0 [1 + o(1)] \underbrace{\boldsymbol{\nu}_n^T \left\{ \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u}-\mathbf{v}\|)g(\|\mathbf{u}-\mathbf{v}\|)\phi_L(\|\mathbf{u}-\mathbf{v}\|)\phi_L^T(\|\mathbf{u}-\mathbf{v}\|)d\mathbf{u}d\mathbf{v} \right\}}_{Q_L} \boldsymbol{\nu}_n \\
&\geq -O(1) \sqrt{\int_0^R w_o(s)\Delta_n^2(s)s^{d-1}ds} \sqrt{\int_0^R w_o(s)\tilde{\zeta}_L^2(s; \boldsymbol{\theta}_0)ds} + z_0 [1 + o(1)] \|\boldsymbol{\nu}_n\|^2 \times \eta_{\min} [Q_L] \\
&= -O(L^{-\nu_1}) \|\boldsymbol{\nu}_n\| + z_0 [1 + o(1)] \|\boldsymbol{\nu}_n\|^2 \times \eta_{\min} [Q_L] > 0.
\end{aligned}$$

Therefore, we have shown that  $\boldsymbol{\theta}^{*T}\boldsymbol{\phi}_L(r)$  is between  $\boldsymbol{\theta}_0^T\boldsymbol{\phi}_L(s) \pm z_0\boldsymbol{\nu}_n^T\boldsymbol{\phi}_L(s)$ , and hence

$$\begin{aligned}
\|\boldsymbol{\theta}^* - \boldsymbol{\theta}_0\|^2 &= (\boldsymbol{\theta}^* - \boldsymbol{\theta}_0)^T \left[ \int_0^R w_o(s)\boldsymbol{\phi}_L(s)\boldsymbol{\phi}_L^T(s)ds \right] (\boldsymbol{\theta}^* - \boldsymbol{\theta}_0) \\
&= \int_0^R w_o(s) [\boldsymbol{\theta}^{*T}\boldsymbol{\phi}_L(s) - \boldsymbol{\theta}_0^T\boldsymbol{\phi}_L(s)]^2 ds \leq z_0^2 \int_0^R w_o(s) [\boldsymbol{\nu}_n^T\boldsymbol{\phi}_L(s)]^2 ds = z_0^2 \|\boldsymbol{\nu}_n\|^2,
\end{aligned}$$

which completes the proof of equation (A.36).

Furthermore, to show (A.37), note that, under condition C4'(b),

$$|g(r) - \tilde{g}_L(r; \boldsymbol{\theta}_0)| = g(r) \left| 1 - \exp \left[ - \sum_{l=L+1}^{\infty} \theta_{0,l}\phi_l(r) \right] \right| = g(r)O(L^{-\nu_1+\tau_1}) = O(L^{-\nu_1+\tau_1}).$$

Under condition C2(a), the above result also implies that  $\sup_{0 < r < R} \tilde{g}_L(r; \boldsymbol{\theta}_0) = O(1)$ . Then,

we have that

$$\begin{aligned}
|g(r) - \tilde{g}_L(r; \boldsymbol{\theta}^*)| &\leq |g(r) - \tilde{g}_L(r; \boldsymbol{\theta}_0)| + |\tilde{g}_L(r; \boldsymbol{\theta}_0) - \tilde{g}_L(r; \boldsymbol{\theta}^*)| \\
&= O(L^{-\nu_1+\tau_1}) + \tilde{g}_L(r; \boldsymbol{\theta}_0) \left| 1 - \exp \left[ \sum_{l=1}^L (\theta_l^* - \theta_{0,l})\phi_l(r) \right] \right| \\
&= O(L^{-\nu_1+\tau_1}) + \tilde{g}_L(r; \boldsymbol{\theta}_0) O \left( \sup_{0 < r \leq R} \|\boldsymbol{\phi}_L(r)\| \|\boldsymbol{\theta} - \boldsymbol{\theta}^*\| \right) \\
&= O(L^{-\nu_1+\tau_1}) + O(L^{\nu_0-\nu_1+\nu_2}) \\
&= o(1),
\end{aligned}$$

where the last equality follows from condition C5', where we have assumed that  $0 \leq 2\nu_0 < \nu_1 - \nu_2$ . Equation (A.37) immediately follows by noting that all the upper bounds do not depend on  $r$ . Equation (A.38) is trivial by combining equation (A.37) and condition C2(a).  $\square$

### 3.4 Proof of Lemma 3

**Lemma A.7.** *Under conditions C1-C3, and C4'-C5', we have that as  $L \rightarrow \infty$  and  $L^{4\nu_0+2\nu_2}/m|D_n| \rightarrow 0$ ,*

$$\|\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\| = O_p\left(\frac{L^{\nu_0}}{\sqrt{m|D_n|}}\right) \quad (\text{A.39})$$

where  $\boldsymbol{\theta}^*$  is defined in equation (A.35).

*Proof.* It is straightforward to see that solving estimating equation (9), i.e.,  $\tilde{\mathbf{U}}_L(\boldsymbol{\theta}) = \mathbf{0}$ , is equivalent to maximizing the following composite log likelihood function

$$\begin{aligned} \tilde{L}(\boldsymbol{\theta}) &= \frac{1}{m} \sum_{i=1}^m \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_R(\|\mathbf{u} - \mathbf{v}\|) \log [\tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta})] \\ &\quad - \frac{1}{m(m-1)} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} w_R(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}), \end{aligned} \quad (\text{A.40})$$

with respect to  $\boldsymbol{\theta}$ , because  $\partial \tilde{L}(\boldsymbol{\theta}) / \partial \boldsymbol{\theta} = \tilde{\mathbf{U}}_L(\boldsymbol{\theta}) / m$ . Note that the Hessian matrix

$$\tilde{\mathbf{H}}_L(\boldsymbol{\theta}) = \frac{\partial^2 \tilde{L}(\boldsymbol{\theta})}{\partial \boldsymbol{\theta} \partial \boldsymbol{\theta}^T} = - \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_R(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}) \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\phi}_L^T(\|\mathbf{u} - \mathbf{v}\|)$$

is negative definitive, which implies that  $\tilde{L}(\boldsymbol{\theta})$  is a concave function of  $\boldsymbol{\theta}$ .

We use the same steps as in the proof of Lemma A.2. Let  $J_{m,n}$  be a sequence of positive real numbers such that  $J_{m,n} \rightarrow \infty$  as  $m \rightarrow \infty$  and/or  $n \rightarrow \infty$ . We shall first show that for any given  $\varepsilon > 0$  there exists a large constant  $C_\varepsilon$  such that, for large  $m$  or/and  $n$ ,

$$\mathbb{P} \left\{ \sup_{\|\boldsymbol{\delta}_L\|=1} \tilde{L}(\boldsymbol{\theta}^* + C_\varepsilon J_{m,n}^{-1/2} \boldsymbol{\delta}_L) < \tilde{L}(\boldsymbol{\theta}^*) \right\} \geq 1 - \varepsilon. \quad (\text{A.41})$$

Inequality (A.41) implies that with probability tending to 1 the function  $\tilde{L}(\boldsymbol{\theta})$  has a local maximum, denoted as  $\hat{\boldsymbol{\theta}} = \hat{\boldsymbol{\theta}}_{m,n}$ , in the ball  $\Theta_{m,n} = \{\boldsymbol{\theta}^* + J_{m,n}^{-1/2} C \boldsymbol{\delta}_L : \|\boldsymbol{\delta}_L\| = 1\}$ . It then follows that  $J_{m,n} \|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|^2$  is bounded in probability; i.e.,  $\|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|^2 = O_p(J_{m,n}^{-1})$ .

To show (A.41), for any fixed  $\boldsymbol{\delta}_L \in \mathbb{R}^L$  that  $\|\boldsymbol{\delta}_L\| = 1$  define a function of  $z > 0$  as

$$H_{m,n}(z) = -\tilde{L}(\boldsymbol{\theta}^* + zJ_{m,n}^{-1/2}\boldsymbol{\delta}_L). \quad (\text{A.42})$$

Then

$$H'_{m,n}(z) = -\frac{J_{m,n}^{-1/2}}{m} \boldsymbol{\delta}_L^T \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^* + zJ_{m,n}^{-1/2}\boldsymbol{\delta}_L), \quad (\text{A.43})$$

$$H''_{m,n}(z) = -J_{m,n}^{-1} \boldsymbol{\delta}_L^T \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^* + zJ_{m,n}^{-1/2}\boldsymbol{\delta}_L) \boldsymbol{\delta}_L \quad (\text{A.44})$$

and by the Taylor's theorem

$$H_{m,n}(z) = H_{m,n}(0) + H'_{m,n}(0)z + H''_{m,n}(t_z) \frac{z^2}{2}$$

for any  $z > 0$  and some  $0 < t_z < z$ , which implies that

$$\tilde{L}(\boldsymbol{\theta}^* + zJ_{m,n}^{-1/2}\boldsymbol{\delta}_L) - \tilde{L}(\boldsymbol{\theta}^*) = H_{m,n}(0) - H_{m,n}(z) = -z \left[ H'_{m,n}(0) + \frac{z}{2} H''_{m,n}(t_z) \right].$$

By definition,  $H_{m,n}(z)$  is a convex function of  $z$  since  $H''_{m,n}(z) \geq 0$  for any constant  $z$ . Therefore, to find a large enough  $C_\epsilon$  so that (A.41) holds, it suffices to show that  $H'_{m,n}(0) = O_p[H''_{m,n}(t_z)]$  for any  $z > 0$ . We first investigate  $H'_{m,n}(0)$ . By the definition of  $\boldsymbol{\theta}^*$  in (A.35), we have that  $\mathbb{E}[H'_{m,n}(0)] = 0$ . Furthermore, similarly as in the proof of Lemma A.2 the

variance can be shown as

$$\begin{aligned}
mJ_{m,n} \text{Var} [H'_{m,n}(0)] &= \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|)[g^{(4)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2, \mathbf{v}_2) \\
&\quad - g(\|\mathbf{u}_1 - \mathbf{v}_1\|)g(\|\mathbf{u}_2 - \mathbf{v}_2\|)] \times [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&- 4 \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|)[g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) - g(\|\mathbf{u}_1 - \mathbf{v}_1\|)] \\
&\quad \times \tilde{g}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ \frac{2}{m-1} \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|)\tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)\tilde{g}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times [g(\|\mathbf{u}_1 - \mathbf{u}_2\|)g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ \frac{4(m-2)}{m-1} \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|)[g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1] \\
&\quad \times \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)\tilde{g}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&+ 4 \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|)g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) \\
&\quad \times [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&- 8 \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|)[g(\|\mathbf{u}_1 - \mathbf{v}_1\|) - \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)] \\
&\quad \times \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ \frac{4}{m-1} \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|)\tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \\
&\quad \times \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*)[g(\|\mathbf{v}_1 - \mathbf{u}_2\|) - 1] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ 2 \int_{D_n^2} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1) [w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 g(\|\mathbf{u}_1 - \mathbf{v}_1\|) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 \\
&+ \frac{2}{m-1} \int_{D_n^2} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)w_R^2(\|\mathbf{u}_1 - \mathbf{v}_1\|)\tilde{g}_L^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1.
\end{aligned}$$

Therefore, under conditions C1-C3 and equations (A.37)-(A.38), we can further simplify

$mJ_{m,n} \text{Var} [H'_{m,n}(0)]$  as follows

$$\begin{aligned}
mJ_{m,n} \text{Var} [H'_{m,n}(0)] &= O(1) \int_{D_n^4} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g^{(4)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2, \mathbf{v}_2) \\
&\quad - g(\|\mathbf{u}_1 - \mathbf{v}_1\|) g(\|\mathbf{u}_2 - \mathbf{v}_2\|)] \times |\phi_L^T(\|\mathbf{u}_1 - \mathbf{v}_1\|) \boldsymbol{\delta}_L| |\phi_L^T(\|\mathbf{u}_2 - \mathbf{v}_2\|) \boldsymbol{\delta}_L| d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^4} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g^{(3)}(\mathbf{u}_1, \mathbf{v}_1, \mathbf{u}_2) - g(\|\mathbf{u}_1 - \mathbf{v}_1\|)] \\
&\quad \times |\phi_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L| |\phi_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L| d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ O(m^{-1}) \int_{D_n^4} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1] \\
&\quad \times |\phi_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L| |\phi_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L| d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^4} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1] \\
&\quad \times |\phi_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L| |\phi_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L| d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^3} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|) |\phi_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L| |\phi_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L| d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ O(1) \int_{D_n^2} [w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)]^2 |\phi_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L|^2 d\mathbf{u}_1 d\mathbf{v}_1 \\
&= O(1) |D_n| \int_{D_n^3} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) |g^{(4)}(\mathbf{w}\mathbf{t} + \mathbf{w}, \mathbf{w}) - g(\|\mathbf{s}\|) g(\|\mathbf{t}\|)| |\phi_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L| |\phi_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L| ds dt d\mathbf{w} \\
&\quad + O(1) |D_n| \int_{D_n^3} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) |g^{(3)}(\mathbf{s}, \mathbf{w}) - g(\|\mathbf{s}\|)| |\phi_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L| |\phi_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L| ds dt d\mathbf{w} \\
&\quad + \frac{1}{m} O(1) |D_n| \int_{D_n^3} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) |g(\|\mathbf{w}\|) g(\|\mathbf{t} - \mathbf{s} + \mathbf{w}\|) - 1| |\phi_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L| |\phi_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L| ds dt d\mathbf{w} \\
&\quad + O(1) |D_n| \int_{D_n^3} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) |g(\|\mathbf{w}\|) - 1| |\phi_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L| |\phi_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L| ds dt d\mathbf{w} \\
&\quad + O(1) |D_n| \int_{D_n^2} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) |\phi_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L| |\phi_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L| ds dt \\
&\quad + O(1) |D_n| \int_{D_n} [w_R(\|\mathbf{s}\|)]^2 |\phi_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L|^2 ds \\
&= O(1) |D_n| \left( \int_0^\infty w_R(s) |\phi_L^T(s) \boldsymbol{\delta}_L| s^{d-1} ds \right)^2 + O(1) |D_n| \int_0^\infty [w_R(s) \phi_L^T(s) \boldsymbol{\delta}_L]^2 s^{d-1} ds,
\end{aligned}$$

where the last equality follows from conditions C2-C3 and equations (A.37)-(A.38). Recall that by definition of orthogonal basis functions, we have that  $\int_0^R w_o(s) [\phi_L^T(s) \boldsymbol{\delta}_L]^2 ds =$

$\boldsymbol{\delta}_L^T \boldsymbol{\delta}_L = 1$ . Finally, by the condition C4, we have that

$$\begin{aligned} mJ_{m,n} \text{Var} [H'_{m,n}(0)] &= O(1)|D_n|^{-1} \left\{ \int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L] s^{d-1} ds \right\}^2 \\ &\quad + O(1)|D_n|^{-1} \int_0^R w_o^2(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^2 s^{d-1} ds \\ &\leq O(1)|D_n|^{-1} \left\{ \int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^2 ds \right\} \int_0^R w_o(s) s^{2(d-1)} ds \\ &\quad + O(1)|D_n|^{-1} \int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^2 ds = O\left(\frac{1}{|D_n|}\right), \end{aligned}$$

where the second last inequality follows from the Hölder's inequality.

Combing with the fact that  $\mathbb{E} [H'_{m,n}(0)] = 0$ , we have that

$$H'_{m,n}(0) = O_P\left(\frac{1}{\sqrt{m|D_n|J_{m,n}}}\right). \quad (\text{A.45})$$

Now we proceed to study  $H''_{m,n}(t_z)$ . Some tedious algebra gives that

$$\begin{aligned} &J_{m,n}^2 m(m-1) \text{Var} [H''_{m,n}(t_z)] \\ &= 2 \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \tilde{g}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|; \tilde{\boldsymbol{\theta}}^*) \\ &\quad \times [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\ &+ 4 \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|) \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \\ &\quad \times \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|; \tilde{\boldsymbol{\theta}}^*) g(\|\mathbf{v}_1 - \mathbf{u}_2\|) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\ &+ 2 \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_R^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_L^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) g(\|\mathbf{u}_1 - \mathbf{v}_1\|) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^4 d\mathbf{u}_1 d\mathbf{v}_1 \\ &+ 4(m-2) \int_{D_n^4} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) \lambda(\mathbf{v}_2) w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1] \\ &\quad \times \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \tilde{g}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|; \tilde{\boldsymbol{\theta}}^*) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\ &+ 4(m-2) \int_{D_n^3} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) \lambda(\mathbf{u}_2) w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|) \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|; \tilde{\boldsymbol{\theta}}^*) \\ &\quad \times [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2, \end{aligned}$$

where  $\tilde{\boldsymbol{\theta}}^* = \boldsymbol{\theta}^* + t_z J_{m,n}^{-1/2} \boldsymbol{\delta}_L$ .

Using equations (A.37)-(A.38), under conditions C1-C3, we can further simplify  $J_{m,n}^2 m(m-1) \text{Var} [H''_{m,n}(t_z)]$  as follows

$$\begin{aligned}
J_{m,n}^2 m(m-1) \text{Var} [H_{m,n}''(t_z)] &= O(1) \int_{D_n^4} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) |g(\|\mathbf{u}_1 - \mathbf{u}_2\|) g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1| \\
&\quad \times [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&+ O(1) \int_{D_n^3} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|) [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ O(1) \int_{D_n^2} w_R^2(\|\mathbf{u}_1 - \mathbf{v}_1\|) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^4 d\mathbf{u}_1 d\mathbf{v}_1 \\
&+ mO(1) \int_{D_n^4} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) |g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1| \\
&\quad \times [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&+ mO(1) \int_{D_n^3} w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&= O(1) |D_n| \int_{D_n^3} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) |g(\|\mathbf{w}\|) g(\|\mathbf{t} - \mathbf{s} + \mathbf{w}\|) - 1| [\boldsymbol{\phi}_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L]^2 ds dt d\mathbf{w} \\
&+ O(1) |D_n| \int_{D_n^2} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) [\boldsymbol{\phi}_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L]^2 ds dt \\
&+ O(1) |D_n| \int_{D_n} [w_R(\|\mathbf{s}\|)]^2 [\boldsymbol{\phi}_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L]^4 ds \\
&+ mO(1) |D_n| \int_{D_n^3} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) |g(\|\mathbf{w}\|) - 1| [\boldsymbol{\phi}_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L]^2 ds dt d\mathbf{w} \\
&+ mO(1) |D_n| \int_{D_n^2} w_R(\|\mathbf{s}\|) w_R(\|\mathbf{t}\|) [\boldsymbol{\phi}_L^T(\|\mathbf{s}\|) \boldsymbol{\delta}_L]^2 [\boldsymbol{\phi}_L^T(\|\mathbf{t}\|) \boldsymbol{\delta}_L]^2 ds dt \\
&= O(1) |D_n| \int_0^\infty [w_R(s)]^2 [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^4 s^{d-1} ds + mO(1) |D_n| \left( \int_0^\infty w_R(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^2 s^{d-1} ds \right)^2.
\end{aligned}$$

Recall that by definition of orthogonal basis functions, we have that  $\int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^2 ds = \boldsymbol{\delta}_L^T \boldsymbol{\delta}_L = 1$ . Then, by the condition C4', we have that

$$\begin{aligned}
J_{m,n}^2 m(m-1) \text{Var} [H_{m,n}''(t_z)] &= O(1) |D_n|^{-1} \int_0^R w_o^2(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^4 s^{d-1} ds \\
&\quad + O(1) m |D_n|^{-1} \left\{ \int_0^\infty w_o(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^2 s^{d-1} ds \right\}^2 \\
&\leq O(1) |D_n|^{-1} \sup_{0 < r \leq R} \|\boldsymbol{\phi}_L(r)\|^2 \times \int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s) \boldsymbol{\delta}_L]^2 ds + O(1) m |D_n|^{-1} \\
&= O(L^{2\nu_2} |D_n|^{-1}) + O(m |D_n|^{-1}),
\end{aligned}$$

which immediately implies that

$$\text{Var} [H''_{m,n}(t_z)] = O\left(\frac{L^{2\nu_2}}{J_{m,n}^2 m^2 |D_n|}\right) + O\left(\frac{1}{J_{m,n}^2 m |D_n|}\right). \quad (\text{A.46})$$

On the other hand, we have that

$$\mathbb{E} [H''_m(t_z)] = J_{m,n}^{-1} \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}_m^*) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1.$$

Using definition of  $\mathbf{Q}_L$  in condition C6 and the fact that  $\eta_{\min}[\mathbf{Q}_L] = \inf_{\|\boldsymbol{\eta}\|^2=1} \boldsymbol{\eta}^T \mathbf{Q}_L \boldsymbol{\eta}$ , we

have that

$$\begin{aligned} \mathbb{E} [J_{m,n} H''_m(t_z)] - \eta_{\min}[\mathbf{Q}_{n,h}] &\geq \int_{D_n^2} \lambda(\mathbf{u}_1) \lambda(\mathbf{v}_1) w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|) \\ &\quad \times \left[ \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \tilde{\boldsymbol{\theta}}^*) - g(\|\mathbf{u}_1 - \mathbf{v}_1\|) \right] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 \\ &= O(1) |D_n| \int_0^\infty w_R(s) \left| \tilde{g}_L(s; \tilde{\boldsymbol{\theta}}^*) - g(s) \right| [\boldsymbol{\phi}_L(s)^T \boldsymbol{\delta}_L]^2 s^{d-1} ds \\ &= O(1) \int_0^R w_o(s) \left| \tilde{g}_L(s; \tilde{\boldsymbol{\theta}}^*) - g(s) \right| [\boldsymbol{\phi}_L(s)^T \boldsymbol{\delta}_L]^2 s^{d-1} ds \\ &= O(1) \sup_{0 < r \leq R} \|\boldsymbol{\phi}_L(r)\| \int_0^R w_o(s) \left| \tilde{g}_L(s; \tilde{\boldsymbol{\theta}}^*) - g(s) \right| |\boldsymbol{\phi}_L(s)^T \boldsymbol{\delta}_L| ds. \end{aligned} \quad (\text{A.47})$$

since  $\tilde{\boldsymbol{\theta}}^* = \boldsymbol{\theta}^* + t_z J_{m,n}^{-1/2} \boldsymbol{\delta}_L$ , it is straightforward to show that, under conditions C4' and

provided the  $\sup_{0 < r \leq R} \left| J_{m,n}^{-1/2} \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(r) \right| = O(1)$ ,

$$\begin{aligned} \left| \tilde{g}_L(r; \tilde{\boldsymbol{\theta}}^*) - g(r) \right| &= g(r) \left| 1 - \exp \left[ (\boldsymbol{\theta}^* - \boldsymbol{\theta}_0)^T \boldsymbol{\phi}_L(r) + t_z J_{m,n}^{-1/2} \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(r) - \tilde{\zeta}_L(r; \boldsymbol{\theta}_0) \right] \right| \\ &= O(1) \left\{ |(\boldsymbol{\theta}^* - \boldsymbol{\theta}_0)^T \boldsymbol{\phi}_L(r)| + |t_z J_{m,n}^{-1/2} \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(r)| + \left| \tilde{\zeta}_L(r; \boldsymbol{\theta}_0) \right| \right\}, \end{aligned}$$

which further gives that, under condition C4' and using (A.36) in Lemma A.6,

$$\begin{aligned} \int_0^R w_o(s) \left| \tilde{g}_L(s; \tilde{\boldsymbol{\theta}}_m^*) - g(s) \right| |\boldsymbol{\phi}_L(|s|)^T \boldsymbol{\delta}_L| ds &= O(1) \int_0^R w_o(s) |(\boldsymbol{\theta}^* - \boldsymbol{\theta}_0)^T \boldsymbol{\phi}_L(r)| |\boldsymbol{\phi}_L(|s|)^T \boldsymbol{\delta}_L| ds \\ &\quad + O(1) \int_0^R w_o(s) |t_z J_{m,n}^{-1/2} \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(r)| |\boldsymbol{\phi}_L(|s|)^T \boldsymbol{\delta}_L| ds \\ &\quad + O(1) \int_0^R w_o(s) \left| \tilde{\zeta}_L(r; \boldsymbol{\theta}_0) \right| |\boldsymbol{\phi}_L(|s|)^T \boldsymbol{\delta}_L| ds \\ &= O(1) \left[ \int_0^R w_o(s) |(\boldsymbol{\theta}^* - \boldsymbol{\theta}_0)^T \boldsymbol{\phi}_L(r)|^2 ds \right]^{1/2} + O(1) z_0 J_{m,n}^{-1/2} + O(1) \left[ \int_0^R w_o(s) \tilde{\zeta}_L^2(r; \boldsymbol{\theta}_0) ds \right]^{1/2} \\ &= O(\|\boldsymbol{\theta}^* - \boldsymbol{\theta}_0\| + J_{m,n}^{-1/2} + L^{-\nu_1}) \\ &= O(L^{\nu_0 - \nu_1} + J_{m,n}^{-1/2} + L^{-\nu_1}) = O(L^{\nu_0 - \nu_1} + J_{m,n}^{-1/2}). \end{aligned}$$



Combining the above result, equation (A.47) and condition C2, we have that if  $J_{m,n}^{-1/2}L^{\nu_2} = O(1)$ ,

$$\mathbb{E}[J_{m,n}H_m''(t_z)] - \eta_{\min}[\mathbf{Q}_{n,h}] = O(L^{\nu_2+\nu_0-\nu_1} + J_{m,n}^{-1/2}L^{\nu_2})$$

By condition C2(a) and C5', the above equation gives that

$$L^{\nu_0}\mathbb{E}[J_{m,n}H_m''(t_z)] \geq c_0 + O(L^{\nu_2+2\nu_0-\nu_1} + J_{m,n}^{-1/2}L^{\nu_0+\nu_2}). \quad (\text{A.48})$$

Hence for the constant  $c = c_0$ , we have that

$$\begin{aligned} \mathbb{P}\left(L^{\nu_0}J_{m,n}H_m''(t_z) < \frac{c}{2}\right) &= \mathbb{P}\left\{J_{m,n}H_m''(t_z) - \mathbb{E}[J_{m,n}H_m''(t_z)] < c/2L^{-\nu_0} - \mathbb{E}[J_{m,n}H_m''(t_z)]\right\} \\ &\leq \mathbb{P}\left\{|J_{m,n}H_m''(t_z) - \mathbb{E}[J_{m,n}H_m''(t_z)]| > |c/2L^{-\nu_0} - \mathbb{E}[J_{m,n}H_m''(t_z)]|\right\} \\ &\quad \times I\left\{\mathbb{E}[J_{m,n}H_m''(t_z)] > c/2L^{-\nu_0}\right\} + I\left\{\mathbb{E}[J_{m,n}H_m''(t_z)] \leq c/2L^{-\nu_0}\right\} \\ &\leq \frac{\text{Var}[J_{m,n}H_m''(t_z)]}{|c/2L^{-\nu_0} - \mathbb{E}[J_{m,n}H_m''(t_z)]|^2} I\left\{\mathbb{E}[J_{m,n}H_m''(t_z)] > c/2L^{-\nu_0}\right\} + I\left\{\mathbb{E}[J_{m,n}H_m''(t_z)] \leq c/2L^{-\nu_0}\right\} \\ &= O\left(\frac{L^{2\nu_2+2\nu_0}}{m^2|D_n|}\right) + O\left(\frac{L^{2\nu_0}}{m|D_n|}\right) + o(1), \end{aligned}$$

where the last equality follows from equations (A.46) and (A.48) when  $J_{m,n} \rightarrow \infty$  and  $L^{2\nu_0+2\nu_2}/J_{m,n} \rightarrow 0$ . Therefore, as long as  $\frac{L^{2\nu_0+2\nu_2}}{m^2|D_n|} + \frac{L^{2\nu_0}}{m|D_n|} \rightarrow 0$ ,  $J_{m,n} \rightarrow \infty$  and  $L^{2\nu_0+2\nu_2}/J_{m,n} \rightarrow 0$ , we have that

$$\mathbb{P}\left(J_{m,n}H_m''(t_z) \geq c_0L^{-\nu_0}/2\right) \rightarrow 1, \quad (\text{A.49})$$

where  $c_0$  is the constant defined in condition C5'.

We have already shown in equation (A.45) that

$$H'_{m,n}(0) = O_P\left(\frac{1}{\sqrt{m|D_n|J_{m,n}}}\right).$$

hence as long as  $\frac{J_{m,n}L^{2\nu_0}}{m|D_n|} = O(1)$ , we have that  $H'_{m,n}(0) = O_P(H''_{m,n}(t_z))$ . In other words, for any sequence  $J_{m,n}$  satisfying  $\frac{J_{m,n}L^{2\nu_0}}{m|D_n|} \rightarrow 0$  and  $L^{2\nu_0+2\nu_2}/J_{m,n} \rightarrow 0$ , the right hand side of the inequality

$$\mathbb{P}\left\{|H'_{m,n}(0)| \geq \frac{z}{2}H''_{m,n}(t_z)\right\} \leq \mathbb{P}\left\{J_{m,n}H''_{m,n}(t_z) \leq \frac{c_0}{2}\right\} + \mathbb{P}\left\{J_{m,n}|H'_{m,n}(0)| \geq \frac{zc_0}{4}\right\}$$

can be arbitrary small by choosing  $z$  and  $m$  and/or  $n$  large enough. Therefore, for any given  $\epsilon > 0$ , there exists  $z_\epsilon > 0$  such that for large  $m$  and/or  $n$ ,

$$\mathbb{P} \left\{ \tilde{L}(\boldsymbol{\theta}^* + z_\epsilon J_{m,n}^{-1/2} \boldsymbol{\delta}_L) < \tilde{L}(\boldsymbol{\theta}^*) \right\} = \mathbb{P} \left\{ z_\epsilon H'_{m,n}(0) + \frac{z_\epsilon^2}{2} H''_{m,n}(t_{z_\epsilon}) > 0 \right\} \geq 1 - \epsilon.$$

Thus, (A.41) holds, which completes the proof of equation (A.39).  $\square$

**Proof of Lemma 3.** Our goal is to show

$$\sup_{0 < r < R} \left| g(r) - \tilde{g}_L(r; \hat{\boldsymbol{\theta}}) \right| = O(L^{-\nu_1 + \max\{\tau_1, \nu_0 + \nu_2\}}) + O_p \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right). \quad (\text{A.50})$$

To show (A.50), using equations (A.37)-(A.38), we have that

$$\begin{aligned} \left| g(r) - \tilde{g}_L(r; \hat{\boldsymbol{\theta}}) \right| &\leq \left| g(r) - \tilde{g}_L(r; \boldsymbol{\theta}^*) \right| + \left| \tilde{g}_L(r; \boldsymbol{\theta}^*) - \tilde{g}_L(r; \hat{\boldsymbol{\theta}}) \right| \\ &= O(L^{-\nu_1 + \max\{\tau_1, \nu_0 + \nu_2\}}) + \tilde{g}_L(r; \boldsymbol{\theta}^*) \left| 1 - \exp \left[ \sum_{l=1}^L (\hat{\theta}_l - \theta_l^*) \phi_l(r) \right] \right| \\ &= O(L^{-\nu_1 + \max\{\tau_1, \nu_0 + \nu_2\}}) + \tilde{g}_L(r; \boldsymbol{\theta}^*) O \left( \sup_{0 < r \leq R} \|\boldsymbol{\phi}_L(r)\| \|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\| \right) \\ &= O(L^{-\nu_1 + \max\{\tau_1, \nu_0 + \nu_2\}}) + O_p \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right) \\ &= o(1), \end{aligned}$$

where the upper bounds does not depend on  $r$ , which completes the proof.  $\square$

### 3.5 Proof of Theorem 2

**Lemma A.8.** Let  $\tilde{\sigma}_\delta^2(\boldsymbol{\theta}^*) = \boldsymbol{\delta}_L^T \boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) \boldsymbol{\delta}_L^T$  with  $\boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) = \text{Var} \left[ \sqrt{m|D_n|} \tilde{\mathbf{U}}(\boldsymbol{\theta}^*) \right]$ . If the vector  $\boldsymbol{\delta}_L$  satisfies (a)  $\|\boldsymbol{\delta}_L\| = 1$ ; (b)  $\int_0^R [w_o(s) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(s)|]^{2+[\delta]} ds = O(1)$ ; and (c)  $\tilde{\sigma}_\delta^2(\boldsymbol{\theta}^*) \geq c_u$  for some constant  $c_u > 0$ , then under conditions C1-C3, C4'-C5' and N1-N2, we have that, as  $L \rightarrow \infty$  and  $m|D_n| \rightarrow \infty$ ,

$$\frac{\sqrt{m|D_n|} \boldsymbol{\delta}_L^T \tilde{\mathbf{U}}(\boldsymbol{\theta}^*)}{\tilde{\sigma}_\delta(\boldsymbol{\theta}^*)} \xrightarrow{D} N(0, 1). \quad (\text{A.51})$$

*Proof.* Following the exact same arguments in the proof of finding  $mJ_{m,n} \text{Var} [H'_{m,n}(0)]$  in Lemma A.7, we have shown that

$$\text{Var} \left[ \boldsymbol{\delta}_L^T \tilde{\mathbf{U}}(\boldsymbol{\theta}^*) \right] = O(m^{-1} |D_n|^{-1}) \Rightarrow \tilde{\sigma}_{\boldsymbol{\delta}}^2(\boldsymbol{\theta}^*) = O(1).$$

To study the asymptotic normality of  $\boldsymbol{\delta}_L^T \tilde{\mathbf{U}}(\boldsymbol{\theta}^*)$ , we define two random variables such that  $\boldsymbol{\delta}_L^T \tilde{\mathbf{U}}(\boldsymbol{\theta}^*) = Z_1 - Z_2(\boldsymbol{\theta}^*)$  as follows

$$Z_1 = \frac{1}{m} \sum_{i=1}^m \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_R(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|), \quad (\text{A.52})$$

$$Z_2(\boldsymbol{\theta}^*) = \frac{1}{m(m-1)} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} w_R(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|). \quad (\text{A.53})$$

By definition of  $\boldsymbol{\theta}^*$  in (A.35), we have that

$$\mathbb{E}Z_1 = \mathbb{E}Z_2(\boldsymbol{\theta}^*) = \int_{D_n^2} \lambda(\mathbf{u}) \lambda(\mathbf{v}) w_R(\|\mathbf{u} - \mathbf{v}\|) g(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) d\mathbf{u} d\mathbf{v}. \quad (\text{A.54})$$

We shall divide our discussions into two case scenarios: (1)  $m \rightarrow \infty$  and (2)  $m$  is fixed.

**Case I: when  $m \rightarrow \infty$ .** In this case, the normality of  $Z_1$  is easy to show since it is an average of independent and identically distributed random variables. The normality of  $Z_2(\boldsymbol{\theta}^*)$  is less straightforward since it has a structure similar to a U-statistic, because

$$Z_2(\boldsymbol{\theta}^*) = \frac{1}{\binom{m}{2}} \sum_{i \neq j=1}^m Z_{2,i,j}(\boldsymbol{\theta}^*)$$

where

$$Z_{2,i,j}(\boldsymbol{\theta}^*) = \frac{1}{2} \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} w_R(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|).$$

To resolve issue, we define the following approximation

$$\tilde{Z}_2(\boldsymbol{\theta}^*) = \frac{2}{m} \sum_{i=1}^m \tilde{Z}_{2,i}(\boldsymbol{\theta}^*) - \mathbb{E}Z_2(\boldsymbol{\theta}^*), \quad (\text{A.55})$$

where

$$\tilde{Z}_{2,i}(\boldsymbol{\theta}^*) = \sum_{\mathbf{u} \in X_i} \int_{D_n} \lambda(\mathbf{v}) w_R(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) d\mathbf{v}.$$

It is trivial to see that  $\mathbb{E}Z_2(\boldsymbol{\theta}^*) = \mathbb{E}\tilde{Z}_2(\boldsymbol{\theta}^*)$  by definition. Following similar arguments as those in the proof of finding  $mJ_{m,n}\text{Var}[H'_{m,n}(0)]$  in Lemma A.2 and Lemma A.7, some tedious algebra gives that

$$\begin{aligned}
& m\text{Var}\left[\tilde{Z}_2(\boldsymbol{\theta}^*) - Z_2(\boldsymbol{\theta}^*)\right] \\
&= \frac{2}{m-1} \int_{D_n^2} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)w_R^2(\|\mathbf{u}_1 - \mathbf{v}_1\|)\tilde{g}_L^2(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)g(\|\mathbf{u}_1 - \mathbf{v}_1\|) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L]^2 d\mathbf{u}_1 d\mathbf{v}_1 \\
&+ \frac{4}{m-1} \int_{D_n^3} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_1 - \mathbf{u}_2\|)\tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*) \\
&\quad \times \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|; \boldsymbol{\theta}^*) [g(\|\mathbf{v}_1 - \mathbf{u}_2\|) - 1] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{u}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 \\
&+ \frac{2}{m-1} \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|)\tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)\tilde{g}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) \\
&\quad \times g(\|\mathbf{u}_1 - \mathbf{u}_2\|) [g(\|\mathbf{v}_1 - \mathbf{v}_2\|) - 1] [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{v}_1 d\mathbf{u}_2 d\mathbf{v}_2 \\
&- \frac{2}{m-1} \int_{D_n^4} \lambda(\mathbf{u}_1)\lambda(\mathbf{v}_1)\lambda(\mathbf{u}_2)\lambda(\mathbf{v}_2)w_R(\|\mathbf{u}_1 - \mathbf{v}_1\|)w_R(\|\mathbf{u}_2 - \mathbf{v}_2\|) [g(\|\mathbf{u}_1 - \mathbf{u}_2\|) - 1] \\
&\quad \times \tilde{g}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|; \boldsymbol{\theta}^*)\tilde{g}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|; \boldsymbol{\theta}^*) [\boldsymbol{\phi}_L(\|\mathbf{u}_1 - \mathbf{v}_1\|)^T \boldsymbol{\delta}_L] [\boldsymbol{\phi}_L(\|\mathbf{u}_2 - \mathbf{v}_2\|)^T \boldsymbol{\delta}_L] d\mathbf{u}_1 d\mathbf{u}_2 d\mathbf{v}_1 d\mathbf{v}_2 \\
&= O(m^{-1}|D_n|^{-1}) \int_0^R w_o^2(s) [\boldsymbol{\phi}_L(s)^T \boldsymbol{\delta}_L]^2 s^{d-1} ds + O(m^{-1}|D_n|^{-1}) \left\{ \int_0^R w_o(s) [\boldsymbol{\phi}_L(s)^T \boldsymbol{\delta}_L] s^{d-1} ds \right\}^2 \\
&= O(m^{-1}|D_n|^{-1}).
\end{aligned}$$

Therefore, as  $m \rightarrow \infty$ , we have that

$$\sqrt{m|D_n|} \left[ \tilde{Z}_2(\boldsymbol{\theta}^*) - Z_2(\boldsymbol{\theta}^*) \right] = O_p(m^{-1}) = o_p(1),$$

and hence

$$\sqrt{m|D_n|} \boldsymbol{\delta}_L^T \tilde{\mathbf{U}}(\boldsymbol{\theta}^*) = \sqrt{m|D_n|} \left[ Z_1 - \tilde{Z}_2(\boldsymbol{\theta}^*) \right] + o_p(1).$$

Since  $m|D_n|\text{Var}\left[\boldsymbol{\delta}_L^T \tilde{\mathbf{U}}(\boldsymbol{\theta}^*)\right] \geq c_u$  for some constant  $c_u > 0$ , it suffices to show the asymptotic normality of

$$\sqrt{m|D_n|} \left[ Z_1 - \tilde{Z}_2(\boldsymbol{\theta}^*) \right] = \frac{\sqrt{m|D_n|}}{m} \sum_{i=1}^m Y_i,$$

where  $Y_i$ 's are i.i.d. random variables of the form as follows

$$Y_i = \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_R(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) - 2\tilde{Z}_{2,i}(\boldsymbol{\theta}^*) + \mathbb{E}Z_2(\boldsymbol{\theta}^*).$$

Note that  $\sqrt{|D_n|}Y_i$ 's are i.i.d random variables with a bounded variance (straightforward to show), (A.51) immediately follows from the standard central limit theorem as  $m \rightarrow \infty$ .

**Case II: when  $m$  is fixed.** In this case, condition  $m|D_n| \rightarrow \infty$  requires that  $|D_n| \rightarrow \infty$ . In other words, we need to consider the case where the observation window of the point processes is expanding. Define a partition of  $\mathbb{R}^d = \cup_{\mathbf{t} \in \mathbb{Z}^d} \Delta(\mathbf{t})$ , where  $\Delta(\mathbf{t}) = \prod_{k=1}^d (s(t_k - 1/2), s(t_k + 1/2)]$  with  $s > 0$  as the length of the interval. Note that by this definition,  $\Delta(\mathbf{t}_1) \cap \Delta(\mathbf{t}_2) = \emptyset$  if  $\mathbf{t}_1 \neq \mathbf{t}_2 \in \mathbb{Z}$ . Define random variables

$$Y_{1,n}(t) = \frac{|D_n|}{m} \sum_{i=1}^m \sum_{\substack{\neq \\ \mathbf{u} \in X_i \cap \Delta(t), \mathbf{v} \in X_i}} w_R(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|),$$

$$Y_{2,n}(t) = \frac{|D_n|}{m(m-1)} \sum_{i \neq j} \sum_{\mathbf{u} \in X_i \cap \Delta(t), \mathbf{v} \in X_j} w_R(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|).$$

Then by definition, we have that

$$Z_1 = \frac{1}{|D_n|} \sum_{t \in \mathcal{T}_n} Y_{1,n}(t), \quad Z_2(\boldsymbol{\theta}^*) = \frac{1}{|D_n|} \sum_{t \in \mathcal{T}_n} Y_{2,n}(t),$$

where  $\mathcal{T}_n = \{\mathbf{t} \in \mathbb{Z}^d : \Delta(\mathbf{t}) \cap D_n \neq \emptyset\}$ .

A simple application of the Jensen's inequality gives that  $(m^{-1} \sum_{i=1}^m |x_i|)^{2+\lceil \delta' \rceil} \leq m^{-1} \sum_{i=1}^m |x_i|^{2+\lceil \delta' \rceil}$  (note that  $f(x) = |x|^{2+\lceil \delta' \rceil}$  is convex)

$$\begin{aligned} \mathbb{E} |Y_{1,n}(t)|^{2+\lceil \delta' \rceil} &\leq \frac{1}{m} \sum_{i=1}^m \mathbb{E} \left| \sum_{\substack{\neq \\ \mathbf{u} \in X_i \cap \Delta(t), \mathbf{v} \in X_i}} |D_n| w_R(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) \right|^{2+\lceil \delta' \rceil} \\ &= \mathbb{E} \left| \sum_{\substack{\neq \\ \mathbf{u} \in X_1 \cap \Delta(t), \mathbf{v} \in X_1}} |D_n| w_R(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) \right|^{2+\lceil \delta' \rceil} \\ &\leq \mathbb{E} \left\{ \sum_{\substack{\neq \\ \mathbf{u} \in X_1 \cap \Delta(t), \mathbf{v} \in X_1}} |D_n| w_R(\|\mathbf{u} - \mathbf{v}\|) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|)| \right\}^{2+\lceil \delta' \rceil} \\ &= O(1) \mathbb{E} \left\{ \sum_{\substack{\neq \\ \mathbf{u} \in X_1 \cap \Delta(t), \mathbf{v} \in X_1}} I(\|\mathbf{u} - \mathbf{v}\| < R) w_o(\|\mathbf{u} - \mathbf{v}\|) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|)| \right\}^{2+\lceil \delta' \rceil}, \end{aligned}$$

where the last expectation is essentially bounded by sums of integrals involving  $w_o^k(s) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(s)|^k$ ,  $k \leq 2 + \lceil \delta' \rceil$ ,  $\lambda(\mathbf{u})$ ,  $g(s)$ ,  $g^{(k)}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ ,  $k = 3, \dots, 2(2 + \lceil \delta' \rceil)$ . All terms are bounded under

conditions C1-C3 and condition N2' except the first batch, hence we only need to consider upper bounds of integrals of the form

$$\int_0^R w_o^k(s) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(s)|^k ds, \quad k \leq 2 + \lceil \delta' \rceil.$$

For any  $k < 2 + \lceil \delta' \rceil$ , by the Höder's inequality with  $p = (2 + \lceil \delta' \rceil)/k$ ,  $q = 1/[1 - k/(2 + \lceil \delta' \rceil)]$  such that  $1/p + 1/q = 1$ , we have that

$$\begin{aligned} \int_0^R w_o^k(s) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(s)|^k ds &= \int_0^R [w_o(s)]^{k-1+1/q} \left\{ [w_o(s) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(s)|]^{2+\lceil \delta' \rceil} \right\}^{1/p} ds \\ &\leq \left\{ \int_0^R [w_o(s) |\boldsymbol{\delta}_L^T \boldsymbol{\phi}_L(s)|]^{2+\lceil \delta' \rceil} ds \right\}^{1/p} \left\{ \int_0^R w_o^{q(k-1)+1}(s) ds \right\}^{1/q} \\ &= O(1), \end{aligned}$$

where the last equality follows from the condition for  $\boldsymbol{\delta}_L$ . Therefore, we have that there exists a constant  $C_1$  such that

$$\mathbb{E} |Y_{1,n}(\mathbf{t})|^{2+\lceil \delta' \rceil} < C_1.$$

Similar arguments also yield that for some constant  $C_2 > 0$

$$\mathbb{E} |Y_{2,n}(\mathbf{t})|^{2+\lceil \delta' \rceil} < C_2.$$

Then by the Minkowski inequality, we have that

$$\begin{aligned} \mathbb{E} |Y_{1,n}(\mathbf{t}) - Y_{2,n}(\mathbf{t})|^{2+\lceil \delta' \rceil} &\leq \left\{ \left[ \mathbb{E} |Y_{1,n}(\mathbf{t})|^{2+\lceil \delta' \rceil} \right]^{1/(2+\lceil \delta' \rceil)} + \left[ \mathbb{E} |Y_{2,n}(\mathbf{t})|^{2+\lceil \delta' \rceil} \right]^{1/(2+\lceil \delta' \rceil)} \right\}^{2+\lceil \delta' \rceil} \\ &< 2^{2+\lceil \delta' \rceil} \max\{C_1, C_2\}, \end{aligned}$$

which further gives that

$$\sup_{n \geq 1} \sup_{t \in \mathcal{T}_n} \mathbb{E} |Y_{1,n}(\mathbf{t}) - Y_{2,n}(\mathbf{t})|^{2+\delta} \leq \left( \sup_{n \geq 1} \sup_{t \in \mathcal{T}_n} \mathbb{E} |Y_{1,n}(\mathbf{t}) - Y_{2,n}(\mathbf{t})|^{2+\lceil \delta' \rceil} \right)^{\frac{2+\delta}{2+\lceil \delta' \rceil}} < \infty. \quad (\text{A.56})$$

Note that the total number of disjoint partitions  $\Delta(\mathbf{t}) \cap D_n \neq \emptyset$  is of the order  $|D_n|$ , hence we can check that,

$$(|D_n|)^{-1} \text{Var} \left\{ \sum_{t \in \mathcal{T}_n} [Y_{1,n}(t) - Y_{2,n}(t)] \right\} = (|D_n|)^{-1} \text{Var} \left( |D_n| \boldsymbol{\delta}_L^T \tilde{\mathbf{U}}(\boldsymbol{\theta}^*) \right) = m^{-1} \tilde{\sigma}_\delta^2(\boldsymbol{\theta}^*) \geq c_u/m.$$

Therefore, using conditions  $N1'(b)$  and  $N2'$ , together with inequality (A.56), it follows from Theorem 1 of Biscio and Waagepetersen (2019) that as  $|D_n| \rightarrow \infty$ ,

$$\left\{ \text{Var} \left[ \sum_{t \in \mathcal{T}_n} Y_{1,n}(t) - \sum_{t \in \mathcal{T}_n} Y_{2,n}(t) \right] \right\}^{-1/2} \sum_{t \in \mathcal{T}_n} [Y_{1,n}(t) - Y_{2,n}(t)] \xrightarrow{\mathcal{D}} N(0, 1),$$

which coincides with (A.51) by definition of  $Y_{k,n}$ 's,  $k = 1, 2$ .  $\square$

**Lemma A.9.** *Denote  $\widehat{\boldsymbol{\theta}}$  as the solution to estimating equations  $\tilde{\mathbf{U}}_L(\boldsymbol{\theta}) = \mathbf{0}$ , then under conditions C1-C3 and C4'-C5', we have that as  $L \rightarrow \infty$  and  $L^{4\nu_0+2\nu_2}/m|D_n| \rightarrow 0$ , for any  $0 < r \leq R$ ,*

$$\sqrt{m|D_n|} \boldsymbol{\phi}_L^T(r) (\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*) = \sqrt{m|D_n|} \boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) + o_p(1) \|\boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1}\|, \quad (\text{A.57})$$

where  $\boldsymbol{\theta}^*$  and  $\mathbf{Q}_L$  are defined in (A.35) and (14), respectively. Furthermore, under additional conditions N1-N3, we have that

$$\frac{\sqrt{m|D_n|} \boldsymbol{\phi}_L^T(r) (\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*)}{\sigma_L(r; \boldsymbol{\theta}^*)} \xrightarrow{\mathcal{D}} N(0, 1), \quad (\text{A.58})$$

where  $\sigma_L^2(r; \boldsymbol{\theta}^*) = \boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) (\mathbf{Q}_L)^{-1} \boldsymbol{\phi}_L(r)$  and  $\boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) = \text{Var} \left[ \sqrt{m|D_n|} \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) \right]$ .

*Proof.* Recall the definition

$$\begin{aligned} \tilde{\mathbf{U}}_L(\boldsymbol{\theta}) &= \frac{1}{m} \sum_{i=1}^m \sum_{\mathbf{u}, \mathbf{v} \in X_i}^{\neq} w_R(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) \\ &\quad - \frac{1}{m(m-1)} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i, \mathbf{v} \in X_j} w_R(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}). \end{aligned}$$

Using the first order Taylor expansion, we can show that

$$\tilde{\mathbf{U}}_L(\widehat{\boldsymbol{\theta}}) - \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) = -\tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) (\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*), \quad (\text{A.59})$$

where  $\|\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\| \leq \|\widehat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|$  and

$$\tilde{\mathbf{H}}_L(\boldsymbol{\theta}) = \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_R(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|) \boldsymbol{\phi}_L^T(\|\mathbf{u} - \mathbf{v}\|) \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}) \quad (\text{A.60})$$

is a symmetric non-negative definite matrix. Observe that  $\tilde{\mathbf{U}}_L(\hat{\boldsymbol{\theta}}) = \mathbf{0}$ , we can re-write expansion (A.59) as follows

$$\tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) = (\mathbf{Q}_L + \mathbf{Q}^\Delta) (\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*), \quad (\text{A.61})$$

where  $\mathbf{Q}_L$  is defined in (14) and  $\mathbf{Q}^\Delta = \tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) - \mathbf{Q}_L$ . From the above new expansion, we have that

$$\begin{aligned} \boldsymbol{\phi}_L^T(r)(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*) &= \boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \left[ \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) - \mathbf{Q}^\Delta(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*) \right] \\ &\leq \boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) + \|\boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1}\| \sigma_{\max} [\mathbf{Q}^\Delta] \|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|, \end{aligned} \quad (\text{A.62})$$

where  $\sigma_{\max}(\mathbf{A})$  stands for the largest singular value of the matrix  $\mathbf{A}$ . We have shown the order of  $\|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\|$  in Lemma (A.7), so it remains to quantify  $\sigma_{\max} [\mathbf{Q}^\Delta]$ . Note that we can further decompose  $\mathbf{Q}^\Delta$  as follows

$$\mathbf{Q}^\Delta = \tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) + \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) - \mathbb{E} [\tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*)] + \mathbb{E} [\tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*)] - \mathbf{Q}_L.$$

By the property of the singular value, we readily have that

$$\sigma_{\max} [\mathbf{Q}^\Delta] \leq \sigma_{\max} [\tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*)] + \sigma_{\max} \left\{ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) - \mathbb{E} [\tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*)] \right\} + \sigma_{\max} \left\{ \mathbb{E} [\tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*)] - \mathbf{Q}_L \right\} \quad (\text{A.63})$$

which will be studied one by one.

By definition of  $\tilde{\boldsymbol{\theta}}^*$  and Lemma A.7,

$$\sup_{0 < r \leq R} \left| (\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*)^T \boldsymbol{\phi}_L(r) \right| \leq \|\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*\| \sup_{0 < r \leq R} \|\boldsymbol{\phi}_L(r)\| = \|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\| O(L^{\nu_2}) = O_P \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right) = o_P(1).$$

Then, it is straightforward to see that

$$\begin{aligned} |\tilde{g}_L(r; \boldsymbol{\theta}^*) - \tilde{g}_L(r; \tilde{\boldsymbol{\theta}}^*)| &= \tilde{g}_L(r; \boldsymbol{\theta}^*) \left| 1 - \exp \left[ (\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*)^T \boldsymbol{\phi}_L(r) \right] \right| \\ &= \tilde{g}_L(r; \boldsymbol{\theta}^*) O(1) \left| (\tilde{\boldsymbol{\theta}}^* - \boldsymbol{\theta}^*)^T \boldsymbol{\phi}_L(r) \right| = \tilde{g}_L(r; \boldsymbol{\theta}^*) O_P \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right), \end{aligned}$$



which further implies that

$$\begin{aligned}
\eta_{\max} \left[ \tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] &= \sup_{\|\boldsymbol{\delta}\|=1} \boldsymbol{\delta}^T \left[ \tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] \boldsymbol{\delta} \\
&\leq \sup_{\|\boldsymbol{\delta}\|=1} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_R(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} [\boldsymbol{\delta}^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|)]^2 |\tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \tilde{\boldsymbol{\theta}}^*) - \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*)| \\
&= O_P \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right) \sup_{\|\boldsymbol{\delta}\|=1} \sum_{i \neq j=1}^m \sum_{\mathbf{u} \in X_i} \sum_{\mathbf{v} \in X_j} \frac{w_R(\|\mathbf{u} - \mathbf{v}\|)}{m(m-1)} [\boldsymbol{\delta}^T \boldsymbol{\phi}_L(\|\mathbf{u} - \mathbf{v}\|)]^2 \tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) \\
&= O_P \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right) \eta_{\max} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right].
\end{aligned}$$

Following exactly the same steps, we can also show that

$$-\eta_{\min} \left[ \tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] = \eta_{\max} \left[ -\tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) + \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] = \eta_{\max} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] O_P \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right),$$

which implies that

$$\sigma_{\max} \left[ \tilde{\mathbf{H}}_L(\tilde{\boldsymbol{\theta}}^*) - \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] = \eta_{\max} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] O_P \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} \right), \quad (\text{A.64})$$

where the convergence is entry-wise.

The next step is to quantify the magnitude of  $\sigma_{\max} \left\{ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) - \mathbb{E} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] \right\}$ . Using the standard random matrix theory, it suffices to consider the variability of  $\boldsymbol{\delta}^T \left\{ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) - \mathbb{E} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] \right\} \boldsymbol{\delta}$  for any  $\boldsymbol{\delta} \in \mathbb{R}^L$  with  $\|\boldsymbol{\delta}\| = 1$ . Following steps as those in the proof of Lemma A.7 about  $\text{Var} \left[ H''_{m,n}(t_z) \right]$ , we immediately have that

$$\sup_{\|\boldsymbol{\delta}\|=1} \left| \boldsymbol{\delta}^T \left\{ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) - \mathbb{E} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] \right\} \boldsymbol{\delta} \right| = O_p \left( \frac{L^{2\nu_2}}{m^2|D_n|} \right) + O \left( \frac{1}{m|D_n|} \right),$$

hence that

$$\sigma_{\max} \left\{ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) - \mathbb{E} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] \right\} = O_p \left( \frac{L^{2\nu_2}}{m^2|D_n|} \right) + O \left( \frac{1}{m|D_n|} \right). \quad (\text{A.65})$$

Next, we proceed to bound the largest singular value of  $\mathbb{E} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] - \mathbf{Q}_L$ . For any  $\boldsymbol{\delta} \in \mathbb{R}^L$

with  $\|\boldsymbol{\delta}\| = 1$ ,

$$\begin{aligned}
& \boldsymbol{\delta}^T \left\{ \mathbb{E} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] - \mathbf{Q}_L \right\} \boldsymbol{\delta} \\
&= \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|) [\boldsymbol{\phi}_L^T(\|\mathbf{u} - \mathbf{v}\|)\boldsymbol{\delta}]^2 [\tilde{g}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}^*) - g(\|\mathbf{u} - \mathbf{v}\|)] d\mathbf{u}d\mathbf{v} \\
&= \int_{D_n^2} \lambda(\mathbf{u})\lambda(\mathbf{v})w_R(\|\mathbf{u} - \mathbf{v}\|) [\boldsymbol{\phi}_L^T(\|\mathbf{u} - \mathbf{v}\|)\boldsymbol{\delta}]^2 g(\|\mathbf{u} - \mathbf{v}\|) \\
&\quad \times \left\{ 1 - \exp \left[ (\boldsymbol{\theta}^* - \boldsymbol{\theta}_0)^T \boldsymbol{\phi}(\|\mathbf{u} - \mathbf{v}\|) - \tilde{\zeta}_L(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}_0) \right] \right\} d\mathbf{u}d\mathbf{v} \\
&= O(1) \int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s)\boldsymbol{\delta}]^2 \left[ |(\boldsymbol{\theta}^* - \boldsymbol{\theta}_0)^T \boldsymbol{\phi}(s)| + |\tilde{\zeta}_L(s; \boldsymbol{\theta}_0)| \right] s^{d-1} ds \\
&= O(1) \|\boldsymbol{\theta}^* - \boldsymbol{\theta}_0\| \sup_{0 < r \leq R} \|\boldsymbol{\phi}(r)\| \int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s)\boldsymbol{\delta}]^2 ds \\
&\quad + O(1) \sqrt{\int_0^R w_o(s) \tilde{\zeta}_L^2(\|\mathbf{u} - \mathbf{v}\|; \boldsymbol{\theta}_0) ds} \times \sqrt{\int_0^R w_o(s) [\boldsymbol{\phi}_L^T(s)\boldsymbol{\delta}]^2 ds} \\
&= O(L^{\nu_0 + \nu_2 - \nu_1}),
\end{aligned}$$

where the last equality follows from condition C4 and Lemma A.6. This further gives that

$$\sigma_{\max} \left\{ \mathbb{E} \left[ \tilde{\mathbf{H}}_L(\boldsymbol{\theta}^*) \right] - \mathbf{Q}_L \right\} = O(L^{\nu_0 + \nu_2 - \nu_1}). \quad (\text{A.66})$$

Combining equations (A.63)-(A.65), we have that where  $\sigma_{\max}(\mathbf{Q}^\Delta) = O_p \left( \frac{L^{\nu_0 + \nu_2}}{\sqrt{m|D_n|}} + L^{\nu_0 + \nu_2 - \nu_1} \right)$ .

In addition, we have shown in Lemma A.7 that  $\|\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*\| = O_p \left( \frac{L^{\nu_0}}{\sqrt{m|D_n|}} \right)$ . Plugging these two equations back to (A.62), we have that

$$\boldsymbol{\phi}_L^T(r)(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*) = \boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*) + O_p \left( \frac{L^{2\nu_0 + \nu_2}}{m|D_n|} + \frac{L^{\nu_0 + \nu_2 - \nu_1}}{\sqrt{m|D_n|}} \right) \|\boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1}\|,$$

which gives (A.57), recall that  $\nu_2 + 2\nu_0 < \nu_1$  in condition C5' and the condition  $L^{4\nu_0 + 2\nu_2}/m|D_n| \rightarrow 0$ .

To show (A.58), define vector  $\boldsymbol{\ell}(r) = (\mathbf{Q}_L)^{-1} \boldsymbol{\phi}_L^T(r)$  and its standardized version  $\boldsymbol{\ell}_0(r) = \|\boldsymbol{\ell}(r)\|^{-1} \boldsymbol{\ell}(r)$  as in condition N3. Then applying Lemma A.8 to  $\boldsymbol{\ell}_0^T(r) \tilde{\mathbf{U}}(\boldsymbol{\theta}^*)$ , under condition N3, we have that

$$\frac{\sqrt{m|D_n|} \boldsymbol{\ell}_0^T(r) \tilde{\mathbf{U}}(\boldsymbol{\theta}^*)}{\sqrt{\boldsymbol{\ell}_0^T(r) \boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}_0(r)}} = \frac{\sqrt{m|D_n|} \boldsymbol{\ell}^T(r) \tilde{\mathbf{U}}(\boldsymbol{\theta}^*)}{\sqrt{\boldsymbol{\ell}^T(r) \boldsymbol{\Sigma}_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}(r)}} \xrightarrow{D} N(0, 1),$$

where  $\Sigma_U(\boldsymbol{\theta}^*) = \text{Var} \left[ \sqrt{m|D_n|} \tilde{\mathbf{U}}(\boldsymbol{\theta}^*) \right]$ . Then using (A.57), we have that

$$\begin{aligned} \frac{\sqrt{m|D_n|} \boldsymbol{\phi}_L^T(r) (\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*)}{\sqrt{\boldsymbol{\ell}^T(r) \Sigma_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}(r)}} &= \frac{\sqrt{m|D_n|} \boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*)}{\sqrt{\boldsymbol{\ell}^T(r) \Sigma_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}(r)}} + o_p(1) \frac{\|\boldsymbol{\ell}(r)\|}{\sqrt{\boldsymbol{\ell}^T(r) \Sigma_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}(r)}} \\ &= \frac{\sqrt{m|D_n|} \boldsymbol{\ell}^T(r) \tilde{\mathbf{U}}_L(\boldsymbol{\theta}^*)}{\sqrt{\boldsymbol{\ell}^T(r) \Sigma_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}(r)}} + o_p(1) \frac{1}{\sqrt{\boldsymbol{\ell}_0^T(r) \Sigma_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}_0(r)}} \xrightarrow{\mathcal{D}} N(0, 1), \end{aligned}$$

where the last equality follows from condition N3(a), which requires that  $\sqrt{\boldsymbol{\ell}_0^T(r) \Sigma_U(\boldsymbol{\theta}^*) \boldsymbol{\ell}_0(r)} \geq c_u$ . The proof is complete.  $\square$

**Proof of Theorem 2.** Recall the definition  $\tilde{g}_L(r; \boldsymbol{\theta}) = \exp[\boldsymbol{\theta}^T \boldsymbol{\phi}_L(r)]$  and  $\hat{g}_L(r) = \tilde{g}_L(r; \hat{\boldsymbol{\theta}})$ , then applying the delta method to the asymptotic distribution of  $\sqrt{m|D_n|} \boldsymbol{\phi}_L^T(r) (\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}^*)$  from Lemma A.9, we have that

$$\frac{\sqrt{m|D_n|} \left[ \tilde{g}_L(r; \hat{\boldsymbol{\theta}}) - \tilde{g}_L(r; \boldsymbol{\theta}^*) \right]}{\tilde{g}_L(r; \boldsymbol{\theta}^*) \sqrt{\boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \Sigma_U(\boldsymbol{\theta}^*) (\mathbf{Q}_L)^{-1} \boldsymbol{\phi}_L(r)}} \xrightarrow{\mathcal{D}} N(0, 1).$$

By equation (A.37) in Lemma A.6, we have that  $\sup_{0 < r < R} |g(r) - \tilde{g}_L(r; \boldsymbol{\theta}^*)| = O(L^{-\nu_1 + \tau_1} + L^{\nu_0 - \nu_1 + \nu_2}) = o(1)$ , it readily follows that

$$\begin{aligned} \frac{\sqrt{m|D_n|} \left[ \tilde{g}_L(r; \hat{\boldsymbol{\theta}}) - \tilde{g}_L(r; \boldsymbol{\theta}^*) \right]}{\tilde{g}_L(r; \boldsymbol{\theta}^*) \sqrt{\boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \Sigma_U(\boldsymbol{\theta}^*) (\mathbf{Q}_L)^{-1} \boldsymbol{\phi}_L(r)}} &= \frac{\sqrt{m|D_n|} \left[ \tilde{g}_L(r; \hat{\boldsymbol{\theta}}) - g(r) + g(r) - \tilde{g}_L(r; \boldsymbol{\theta}^*) \right]}{\tilde{g}_L(r; \boldsymbol{\theta}^*) \sqrt{\boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \Sigma_U(\boldsymbol{\theta}^*) (\mathbf{Q}_L)^{-1} \boldsymbol{\phi}_L(r)}} \\ &= \frac{\sqrt{m|D_n|} \left[ \tilde{g}_L(r; \hat{\boldsymbol{\theta}}) - g(r) + O(L^{-\nu_1 + \tau_1} + L^{\nu_0 - \nu_1 + \nu_2}) \right]}{g(r) \sqrt{\boldsymbol{\phi}_L^T(r) (\mathbf{Q}_L)^{-1} \Sigma_U(\boldsymbol{\theta}^*) (\mathbf{Q}_L)^{-1} \boldsymbol{\phi}_L(r)}} + o_P(1) \xrightarrow{\mathcal{D}} N(0, 1), \end{aligned}$$

which completes the proof.  $\square$

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