

**ORACLE-EFFICIENT GLOBAL INFERENCE FOR
VARIANCE FUNCTION IN NONPARAMETRIC
REGRESSION WITH MISSING COVARIATES**

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Supplementary Material

This supplement provides all the proofs for the main results. Throughout this supplement, we denote by $\|\xi\|$ the Euclidean norm and by $\|\xi\|_\infty$ the largest absolute value of the elements of any vector ξ . For any $l \times k$ matrix $M = \{m_{ij}\}_{i=1, j=1}^{l, k}$, denote $\|M\|_\infty = \max_{\zeta \in \mathbb{R}^k, \zeta \neq 0} \|M\zeta\|_\infty / \|\zeta\|_\infty$ which is easily seen to be equivalent to $\|M\|_\infty = \max_{1 \leq i \leq l} \sum_{j=1}^k |m_{ij}|$. For any function $\psi(x) \in L_2[a, b]$, let $\|\psi(x)\|_\infty = \sup_{x \in [a, b]} |\psi(x)|$.

S1 Preliminaries

Lemma S.1. (*Theorem 1.2 of Bosq (1998)*) *Let ξ_1, \dots, ξ_n be independent random variables with mean 0. If there exists a constant $r > 0$ such that*

(Cramér's Conditions)

$$\mathbb{E} |\xi_i|^k \leq r^{k-2} k! \mathbb{E} \xi_i^2 < +\infty \text{ for } 1 \leq i \leq n, k \geq 3,$$

then for any $t > 0$

$$P \left\{ \left| \sum_{i=1}^n \xi_i \right| > t \right\} \leq 2 \exp \left\{ -\frac{t^2}{4 \sum_{i=1}^n \mathbb{E} \xi_i^2 + 2rt} \right\}.$$

The next lemma is an important result from de Boor (2001, p.149).

Lemma S.2. For any $\phi(x) \in C^{(p)}[a, b]$, there exist a constant $C_p > 0$ and a spline function $m_p(x) \in G_N^{(p-2)}$ such that $\|\phi(x) - m_p(x)\|_\infty \leq C_p \left\| \phi^{(p)}(x) \right\|_\infty N^{-p}$.

For any function $\psi(\cdot), \phi(\cdot) \in L_2[a, b]$, define the theoretical and empirical inner products as

$$\langle \psi, \phi \rangle_2 = \int_a^b \psi(x) \phi(x) f_X(x) dx,$$

and

$$\langle \psi, \phi \rangle_{2,n} = n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} \psi(X_i) \phi(X_i).$$

The corresponding norms are defined as $\|\phi\|_2^2 = \int_a^b \phi^2(x) f(x) dx$ and $\|\phi\|_{2,n}^2 = n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} \phi^2(X_i)$. Meanwhile, define the following theoretical and empirical inner product matrices of $\{B_{J,p}(\cdot)\}_{J=1-p}^N$:

$$V_p = \left(\langle B_{J,p}, B_{J',p} \rangle_2 \right)_{J,J'=1-p}^N, \hat{V}_p^* = \left(\langle B_{J,p}, B_{J',p} \rangle_{2,n} \right)_{J,J'=1-p}^N.$$

It is clear that $n^{-1} \mathbf{B}^T \mathbf{\Delta} \mathbf{B} = \hat{V}_p^*$. Moreover, denote the matrix

$$\hat{V}_p = \left(n^{-1} \sum_{i=1}^n \frac{\delta_i}{\hat{\pi}_i} B_{J,p}(X_i) B_{J',p}(X_i) \right)_{J,J'=1-p}^N,$$

and hence $n^{-1} \mathbf{B}^T \hat{\mathbf{\Delta}} \mathbf{B} = \hat{V}_p$. We next give some properties about the inner product matrices of V_p , \hat{V}_p and \hat{V}_p^* which are needed in the study of the uniform convergence of $\hat{\sigma}^2(x)$.

Lemma S.3. *For any positive p , there exists a constant K_{p1} depending only on p such that $\|V_p^{-1}\|_\infty \leq K_{p1}N$.*

It is a direct conclusion of Lemma A.3 in Cao et al. (2012).

Lemma S.4. *Under Assumption (A6), one has that*

- (a) $\|\hat{V}_p^* - V_p\|_\infty = O_p(n^{-1/2}N^{-1/2} \log^{1/2} n)$;
- (b) $\|\hat{V}_p^* - \hat{V}_p\|_\infty = O_p(n^{-1/2}N^{-1})$;
- (c) *there exist constants K_{p2}, K_{p3} such that $\|\hat{V}_p^{*-1}\|_\infty \leq K_{p2}N$ and $\|\hat{V}_p^{-1}\|_\infty \leq K_{p3}N$ in probability.*

Proof of Lemma S.4(a). Denote

$$\xi_{J,J'}(X_i) = n^{-1} \left\{ \frac{\delta_i}{\pi_i} B_{J,p}(X_i) B_{J',p}(X_i) - \mathbb{E}\{B_{J,p}(X_i) B_{J',p}(X_i)\} \right\}.$$

Since $B_{J,p}(x) B_{J',p}(x) \neq 0$ only if $x \in [\chi_J, \chi_{J+p}] \cap [\chi_{J'}, \chi_{J'+p}]$ and $|J - J'| < p$, one has that

$$\begin{aligned} & \|\hat{V}_p^* - V_p\|_\infty \\ &= \max_{1-p \leq J \leq N} \sum_{J' \in \{J': |J'-J| < p\}} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} B_{J,p}(X_i) B_{J',p}(X_i) - \mathbb{E}\{B_{J,p}(X_i) B_{J',p}(X_i)\} \right| \\ &= \max_{1-p \leq J \leq N} \sum_{J' \in \{J': |J'-J| < p\}} \left| \sum_{i=1}^n \xi_{J,J'}(X_i) \right|. \end{aligned}$$

It is clear that $\mathbb{E} \xi_{J,J'}(X_i) = 0$ and

$$\begin{aligned} \mathbb{E} \xi_{J,J'}^2(X_i) &\leq n^{-2} \mathbb{E} \left\{ \frac{\delta_i}{\pi_i} B_{J,p}(X_i) B_{J',p}(X_i) \right\}^2 \\ &\leq n^{-2} c_\pi^{-1} \mathbb{E}\{B_{J,p}(X_i) B_{J',p}(X_i)\}^2 \\ &\leq c_\pi^{-1} C n^{-2} N^{-1}, \end{aligned} \tag{S1.1}$$

for some constant $C > 0$ since $\max_{1-p \leq J \leq N} |B_{J,p}(x)| = O(1)$. On the other hand,

$$\begin{aligned} \mathbb{E} \xi_{J,J'}^2(X_i) &= n^{-2} \left(\mathbb{E} \left\{ \frac{1}{\pi_i} B_{J,p}^2(X_i) B_{J',p}^2(X_i) \right\} - [\mathbb{E} \{ B_{J,p}(X_i) B_{J',p}(X_i) \}]^2 \right) \\ &\geq n^{-2} \left(\mathbb{E} \{ B_{J,p}^2(X_i) B_{J',p}^2(X_i) \} - [\mathbb{E} \{ B_{J,p}(X_i) B_{J',p}(X_i) \}]^2 \right) \\ &\geq cn^{-2} N^{-1}, \end{aligned}$$

for some $c > 0$ which holds since

$$\mathbb{E} \{ B_{J,p}^2(X_i) B_{J',p}^2(X_i) \} \gg [\mathbb{E} \{ B_{J,p}(X_i) B_{J',p}(X_i) \}]^2$$

and by the Mean Value Theorem,

$$\begin{aligned} \mathbb{E} \{ B_{J,p}^2(X_i) B_{J',p}^2(X_i) \} &= \int_{x \in [\chi_J, \chi_{J+p}] \cap [\chi_{J'}, \chi_{J'+p}], |J-J'| < p} B_{J,p}^2(x) B_{J',p}^2(x) f_X(x) dx \\ &= B_{J,p}^2(\xi) B_{J',p}^2(\xi) f_X(\xi) \int_{x \in [\chi_J, \chi_{J+p}] \cap [\chi_{J'}, \chi_{J'+p}], |J-J'| < p} 1 dx \sim N^{-1} \end{aligned}$$

for some $\xi \in (\chi_J, \chi_{J+p}) \cap (\chi_{J'}, \chi_{J'+p})$, $|J - J'| < p$. When $k \geq 3$, the k -th moment $\mathbb{E} |\xi_{J,J'}(X_i)|^k$ is bounded as follows:

$$\begin{aligned} \mathbb{E} |\xi_{J,J'}(X_i)|^k &= n^{-k} \mathbb{E} \left| \frac{\delta_i}{\pi_i} B_{J,p}(X_i) B_{J',p}(X_i) - \mathbb{E} \{ B_{J,p}(X_i) B_{J',p}(X_i) \} \right|^k \\ &\leq n^{-k} 2^{k-1} \left[\mathbb{E} \left\{ \frac{1}{\pi_i^{k-1}} |B_{J,p}(X_i) B_{J',p}(X_i)|^k \right\} + |\mathbb{E} \{ B_{J,p}(X_i) B_{J',p}(X_i) \}|^k \right] \\ &\leq n^{-k} 2^k \mathbb{E} \left\{ \frac{1}{\pi_i^{k-1}} |B_{J,p}(X_i) B_{J',p}(X_i)|^k \right\} \\ &\leq n^{-k} 2^k c_\pi^{1-k} C^* N^{-1} = n^{-(k-2)} 2^k c_\pi^{1-k} C^* c^{-1} cn^{-2} N^{-1} \\ &\leq n^{-(k-2)} 2^k c_\pi^{1-k} C^* c^{-1} \mathbb{E} \xi_{J,J'}^2(X_i) \leq \left(\frac{2c_0}{n} \right)^{k-2} k! \mathbb{E} \xi_{J,J'}^2(X_i), \end{aligned}$$

for some constants $C^*, c_0 > 0$. The first inequality is easy to see by mathematical induction and Young's Inequality. Thus, $\xi_{J,J'}(X_i)$, $1 \leq i \leq n$, satisfy Cramér's Conditions with $r = 2c_0/n$ in Lemma S.1. Then one ob-

tains that for any given $\rho > 0$,

$$\begin{aligned}
& P \left\{ \left| \sum_{i=1}^n \xi_{J,J'}(X_i) \right| > \rho n^{-1/2} N^{-1/2} \log^{1/2} n \right\} \\
& \leq 2 \exp \left\{ - \frac{\rho^2 n^{-1} N^{-1} \log n}{4 \sum_{i=1}^n \mathbb{E} \xi_{J,J'}^2(X_i) + 4\rho c_0 n^{-3/2} N^{-1/2} \log^{1/2} n} \right\} \\
& = 2 \exp \left\{ - \frac{\rho^2 \log n}{4n^2 N \mathbb{E} \xi_{J,J'}^2(X_1) + 4\rho c_0 n^{-1/2} N^{1/2} \log^{1/2} n} \right\} \\
& \leq n^{-t},
\end{aligned}$$

for some $t > 2$ by choosing a large enough ρ , which holds since $4n^2 N \times \mathbb{E} \xi_{J,J'}^2(X_1)$ is bounded by (S1.1) and $n^{-1/2} N^{1/2} \log^{1/2} n \rightarrow 0$ by Assumption (A6). Therefore,

$$\begin{aligned}
& P \left\{ \max_{1-p \leq J, J' \leq N} \left| \sum_{i=1}^n \xi_{J,J'}(X_i) \right| > \rho n^{-1/2} N^{-1/2} \log^{1/2} n \right\} \\
& \leq \sum_{J, J'=1-p}^N P \left\{ \left| \sum_{i=1}^n \xi_{J,J'}(X_i) \right| > \rho n^{-1/2} N^{-1/2} \log^{1/2} n \right\} \\
& \leq (N+p)^2 n^{-t},
\end{aligned}$$

and hence

$$\begin{aligned}
& \sum_{n=1}^{\infty} P \left\{ \max_{1-p \leq J, J' \leq N} \left| \sum_{i=1}^n \xi_{J,J'}(X_i) \right| > \rho n^{-1/2} N^{-1/2} \log^{1/2} n \right\} \\
& \leq \sum_{n=1}^{\infty} (N+p)^2 n^{-t} < \infty.
\end{aligned}$$

By Borel-Cantelli's Lemma, one immediately obtains that

$$\max_{1-p \leq J, J' \leq N} \left| \sum_{i=1}^n \xi_{J,J'}(X_i) \right| = O_p \left(n^{-1/2} N^{-1/2} \log^{1/2} n \right), \quad (\text{S1.2})$$

which concludes that

$$\left\| \hat{V}_p^* - V_p \right\|_{\infty} = \max_{1-p \leq J \leq N} \sum_{J' \in \{J': |J'-J| < p\}} \left| \sum_{i=1}^n \xi_{J,J'}(X_i) \right| = O_p \left(n^{-1/2} N^{-1/2} \log^{1/2} n \right),$$

completing the proof.

Proof of Lemma S.4(b). According to (S1.2), one has that

$$\begin{aligned} & \max_{1-p \leq J, J' \leq N} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} B_{J,p}(X_i) B_{J',p}(X_i) \right| \\ & \leq \max_{1-p \leq J, J' \leq N} |\mathbb{E}\{B_{J,p}(X_i) B_{J',p}(X_i)\}| + O_p(n^{-1/2} N^{-1/2} \log^{1/2} n) \\ & = O_p(N^{-1}). \end{aligned}$$

Hence,

$$\begin{aligned} & \left\| \hat{V}_p - \hat{V}_p^* \right\|_{\infty} \\ & = \max_{1-p \leq J \leq N} \sum_{J' \in \{J': |J'-J| < p\}} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} \left(\frac{\pi_i - \hat{\pi}_i}{\hat{\pi}_i} \right) B_{J,p}(X_i) B_{J',p}(X_i) \right| \\ & = O_p(n^{-1/2}) (2p-1) \max_{1-p \leq J, J' \leq N} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} B_{J,p}(X_i) B_{J',p}(X_i) \right| \\ & = O_p(n^{-1/2} N^{-1}), \end{aligned}$$

completing the proof.

Proof of Lemma S.4(c). According to Lemma S.3, for any $(N+p)$ length vector $\boldsymbol{\eta}$, $\|V_p^{-1} \boldsymbol{\eta}\|_{\infty} \leq K_{p1} N \|\boldsymbol{\eta}\|_{\infty}$. Thus one has $\|V_p \boldsymbol{\eta}\|_{\infty} \geq K_{p1}^{-1} N^{-1} \|\boldsymbol{\eta}\|_{\infty}$. Since $n^{-1/2} N^{1/2} \log^{1/2} n \rightarrow 0$ by Assumption (A6) and Lemma S.4(a), one has that

$$\begin{aligned} \left\| \hat{V}_p^* \boldsymbol{\eta} \right\|_{\infty} & \geq \|V_p \boldsymbol{\eta}\|_{\infty} - \left\| \left(\hat{V}_p^* - V_p \right) \boldsymbol{\eta} \right\|_{\infty} \\ & \geq K_{p1}^{-1} N^{-1} \|\boldsymbol{\eta}\|_{\infty} - O_p(n^{-1/2} N^{-1/2} \log^{1/2} n) \|\boldsymbol{\eta}\|_{\infty} \\ & = K_{p1}^{-1} N^{-1} \|\boldsymbol{\eta}\|_{\infty} \left(1 - O_p(n^{-1/2} N^{1/2} \log^{1/2} n) \right) \\ & \geq K_{p2}^{-1} N^{-1} \|\boldsymbol{\eta}\|_{\infty} \end{aligned}$$

in probability for some constant $K_{p2} > 0$. Therefore, $\left\| \hat{V}_p^{*-1} \boldsymbol{\eta} \right\|_{\infty} \leq K_{p2} N \|\boldsymbol{\eta}\|_{\infty}$ which together with Lemma S.4(b) concludes that, for any $(N+p)$ length

vector $\boldsymbol{\eta}$,

$$\begin{aligned}
\left\| \hat{V}_p \boldsymbol{\eta} \right\|_\infty &\geq \left\| \hat{V}_p^* \boldsymbol{\eta} \right\|_\infty - \left\| \left(\hat{V}_p - \hat{V}_p^* \right) \boldsymbol{\eta} \right\|_\infty \\
&\geq K_{p2}^{-1} N^{-1} \left\| \boldsymbol{\eta} \right\|_\infty - O_p \left(n^{-1/2} N^{-1} \right) \left\| \boldsymbol{\eta} \right\|_\infty \\
&\geq K_{p2}^{-1} N^{-1} \left\| \boldsymbol{\eta} \right\|_\infty \left(1 - O_p \left(n^{-1/2} \right) \right) \\
&\geq K_{p3}^{-1} N^{-1} \left\| \boldsymbol{\eta} \right\|_\infty
\end{aligned}$$

for some constant $K_{p3} > 0$ in probability. Hence $\left\| \hat{V}_p^{-1} \boldsymbol{\eta} \right\|_\infty \leq K_{p3} N \left\| \boldsymbol{\eta} \right\|_\infty$ in probability, completing the proof.

For any function $\varphi(x) \in C^{(p)}[a, b]$, let $\boldsymbol{\varphi} = (\varphi(X_1), \dots, \varphi(X_n))^T$ and denote

$$\tilde{\varphi}_p^*(x) = (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{*-1} n^{-1} \mathbf{B}^T \boldsymbol{\Delta} \boldsymbol{\varphi}. \quad (\text{S1.3})$$

Lemma S.5. *Under Assumption (A6), there exist constants M_p and $M_{\varphi,p}$ such that any function $\tilde{\varphi}_p^*(x)$ given in (S1.3) satisfies*

$$\left\| \tilde{\varphi}_p^*(x) \right\|_\infty \leq M_p \times \left\| \varphi(x) \right\|_\infty \quad \text{and} \quad \left\| \tilde{\varphi}_p^*(x) - \varphi(x) \right\|_\infty \leq M_{\varphi,p} N^{-p}$$

in probability.

Proof of Lemma S.5. Let the vector $I_N = (1, \dots, 1)^T$ with length N . Then by Lemma S.1, similar to the proof of (S1.2) it is easy to show that

$$\begin{aligned}
\left\| n^{-1} \mathbf{B}^T \boldsymbol{\Delta} I_N \right\|_\infty &= \max_{1-p \leq J \leq N} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} B_{J,p}(X_i) \right| \\
&\leq \max_{1-p \leq J \leq N} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} B_{J,p}(X_i) - \mathbb{E} B_{J,p}(X_1) \right| \\
&\quad + \max_{1-p \leq J \leq N} \left| \mathbb{E} B_{J,p}(X_1) \right| \\
&= O_p \left(n^{-1/2} N^{-1/2} \log^{-1/2} n \right) + O(N^{-1}) = O_p(N^{-1}).
\end{aligned}$$

This together with Lemma S.4(c) and the fact that at most $(p+1)$ of the numbers $B_{1-p,p}(x), \dots, B_{N,p}(x)$ are between 0 and 1, others being 0 implies

that

$$\begin{aligned}
\|\tilde{\varphi}_p^*(x)\|_\infty &= \left\| (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{*-1} n^{-1} \mathbf{B}^T \Delta \varphi \right\|_\infty \\
&\leq (p+1) \left\| \hat{V}_p^{*-1} n^{-1} \mathbf{B}^T \Delta \varphi \right\|_\infty \\
&\leq (p+1) \left\| \hat{V}_p^{*-1} \right\|_\infty \left\| n^{-1} \mathbf{B}^T \Delta \varphi \right\|_\infty \\
&\leq (p+1) \left\| \hat{V}_p^{*-1} \right\|_\infty \left\| n^{-1} \mathbf{B}^T \Delta I_N \right\|_\infty \|\varphi(x)\|_\infty \\
&\leq (p+1) K_{p2} N \left\| n^{-1} \mathbf{B}^T \Delta I_N \right\|_\infty \|\varphi(x)\|_\infty \\
&\leq (p+1) K_{p2} N C N^{-1} \|\varphi(x)\|_\infty \equiv M_p \|\varphi(x)\|_\infty \quad (\text{S1.4})
\end{aligned}$$

in probability, where C is some positive constant and $M_p = (p+1)K_{p2}C$.

Moreover, according to Lemma S.2, there exists a spline function $m_p(x) \in G_N^{(p-2)}$ such that $m_p(x) \equiv \tilde{m}_p^*(x)$ and $\|\varphi(x) - m_p(x)\|_\infty \leq C_p \|\varphi^{(p)}\|_\infty N^{-p}$. Then $\varphi(x) - m_p(x) \in C^{(p)}[a, b]$ and $\tilde{\varphi}_p^*(x) - \tilde{m}_p^*(x)$ can be expressed as in (S1.3) with $\varphi(x)$ replaced by $\varphi(x) - m_p(x)$. Therefore, with this substitution in (S1.4) one has

$$\begin{aligned}
\|\tilde{\varphi}_p^*(x) - \varphi(x)\|_\infty &\leq \|\tilde{\varphi}_p^*(x) - \tilde{m}_p^*(x)\|_\infty + \|m_p(x) - \varphi(x)\|_\infty \\
&\leq M_p \|\varphi(x) - m_p(x)\|_\infty + \|m_p(x) - \varphi(x)\|_\infty \\
&= (M_p + 1) \|\varphi(x) - m_p(x)\|_\infty \\
&\leq C_p (M_p + 1) \|\varphi^{(p)}\|_\infty N^{-p} \equiv M_{\varphi,p} N^{-p}
\end{aligned}$$

in probability, where $M_{\varphi,p} = C_p (M_p + 1) \|\varphi^{(p)}\|_\infty$. The proof is completed.

Lemma S.6. *Under Assumptions (A1)–(A5), for any sequence of measurable functions $s_n(x)$ with $\|s_n(x)\|_\infty = u_n > 0$,*

$$\sup_{x \in [a,b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) s_n(X_i) \varepsilon_i \right| = O_p \left(u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right).$$

Proof of Lemma S.6. Denote $D_n = n^\lambda$ for $1/(2+\eta) < \lambda < 1/3$, which

together with Assumption (A5) implies

$$D_n^{-(\eta+1)} n^{1/2} h^{1/2} \rightarrow 0, \quad \sum_{n=1}^{\infty} D_n^{-(2+\eta)} < \infty, \quad D_n n^{-1/2} h^{-1/2} \log^{1/2} n \rightarrow 0.$$

We decompose the noise ε_i as

$$\varepsilon_i = \varepsilon_{i,1} + \varepsilon_{i,2} + \mu_i, \quad (\text{S1.5})$$

where $\varepsilon_{i,1} = \varepsilon_i I(|\varepsilon_i| > D_n)$, $\mu_i = \mathbb{E}\{\varepsilon_i I(|\varepsilon_i| \leq D_n) | X_i\}$ and $\varepsilon_{i,2} = \varepsilon_i \times I(|\varepsilon_i| \leq D_n) - \mu_i$, in which $I(\cdot)$ is the indicator function.

Firstly, note that

$$\begin{aligned} |\mu_i| &= |-\mathbb{E}\{\varepsilon_i I(|\varepsilon_i| > D_n) | X_i\}| \\ &\leq D_n^{-(\eta+1)} \mathbb{E}\{|\varepsilon_i|^{2+\eta} | X_i\} \\ &\leq D_n^{-(\eta+1)} C_\eta \end{aligned}$$

in probability, where C_η is the upper bound of $\mathbb{E}(|\varepsilon_i|^{2+\eta} | X_i)$ by Assumption (A2). Meanwhile, according to Lemma 3 in Cai et al. (2021), one has that

$$\sup_{x \in [a,b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \right| = \sup_{x \in [a,b]} |\tilde{f}_X(x)| = O_p(1). \quad (\text{S1.6})$$

Thus,

$$\begin{aligned} &\sup_{x \in [a,b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) s_n(X_i) \mu_i \right| \\ &\leq u_n D_n^{-(\eta+1)} C_\eta \sup_{x \in [a,b]} |\tilde{f}_X(x)| \\ &= O_p(u_n D_n^{-(\eta+1)}) = o_p(u_n n^{-1/2} h^{-1/2}). \end{aligned} \quad (\text{S1.7})$$

Next, since

$$\sum_{n=1}^{\infty} P\{I(|\varepsilon_n| > D_n)\} \leq C_\eta \sum_{n=1}^{\infty} D_n^{-(2+\eta)} < \infty,$$

Borel-Cantelli's Lemma implies that

$$P\{\omega: \text{there exists } N_0(\omega) > 0 \text{ such that } |\varepsilon_n(\omega)| \leq D_n \text{ for } n > N_0(\omega)\} = 1.$$

Therefore,

$$P\{\omega: \text{there exists } N(\omega) \text{ such that } |\varepsilon_i(\omega)| \leq D_n \text{ for } 1 \leq i \leq n, n > N(\omega)\} = 1,$$

which concludes that

$$P\{\omega: \text{there exists } N(\omega) \text{ such that } \varepsilon_{i,1} = 0 \text{ for } 1 \leq i \leq n, n > N(\omega)\} = 1.$$

Hence, for any $\gamma > 0$

$$\sup_{x \in [a,b]} |n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) s_n(X_i) \varepsilon_{i,1}| = O_{a.s.}(n^{-\gamma}), \quad (\text{S1.8})$$

where a.s. stands for almost surely.

Finally, we deal with the truncated part $\varepsilon_{i,2}$. It is easy to see that $E(\varepsilon_{i,2}|X_i) = 0$ and

$$\begin{aligned} E(\varepsilon_{i,2}^2|X_i) &= E\{\varepsilon_i^2 I(|\varepsilon_i| \leq D_n) | X_i\} - \mu_i^2 \\ &= \sigma^2(X_i) - E\{\varepsilon_i^2 I(|\varepsilon_i| > D_n) | X_i\} - \mu_i^2 \\ &= \sigma^2(X_i) + O_p(D_n^{-\eta} + D_n^{-2(\eta+1)}). \end{aligned}$$

For convenience, denote $\xi_{in}(x) = n^{-1} \frac{\delta_i}{\pi_i} K_h(X_i - x) s_n(X_i) \varepsilon_{i,2}$. One then gets that $E\xi_{in}(x) = 0$ and

$$\begin{aligned} E\xi_{in}^2(x) &= n^{-2} E\left\{\frac{\delta_i^2}{\pi_i^2} K_h^2(X_i - x) s_n^2(X_i) \varepsilon_{i,2}^2\right\} \\ &= n^{-2} E\left\{\frac{1}{\pi_i} K_h^2(X_i - x) s_n^2(X_i) \varepsilon_{i,2}^2\right\} \\ &\leq c_\pi^{-1} n^{-2} E\{K_h^2(X_i - x) s_n^2(X_i) \varepsilon_{i,2}^2\} \\ &\leq c_\pi^{-1} u_n^2 n^{-2} E\{K_h^2(X_i - x) \varepsilon_{i,2}^2\} \\ &= c_\pi^{-1} u_n^2 n^{-2} h^{-1} \sigma^2(x) f_X(x) \int K^2(v) dv \{1 + u(1)\}. \end{aligned}$$

The k -th moment $\mathbb{E} |\xi_{in}(x)|^k$ for $k \geq 3$ is bounded as follows:

$$\begin{aligned} \mathbb{E} |\xi_{in}(x)|^k &= \mathbb{E} \left\{ |\xi_{in}(x)|^{k-2} (\xi_{in}(x))^2 \right\} \\ &\leq n^{-(k-2)} c_\pi^{-(k-2)} h^{-(k-2)} \|K\|_\infty^{k-2} u_n^{k-2} (2D_n)^{k-2} \mathbb{E} (\xi_{in}(x))^2 \\ &= (qu_n n^{-1} h^{-1} D_n)^{k-2} \mathbb{E} (\xi_{in}(x))^2 \leq (qu_n n^{-1} h^{-1} D_n)^{k-2} k! \mathbb{E} (\xi_{in}(x))^2, \end{aligned}$$

where $q = 2c_\pi^{-1} \|K\|_\infty$. Thus $\xi_{in}(x)$, $1 \leq i \leq n$, fulfill Cramér's Conditions in Lemma S.1 with $r = qu_n n^{-1} h^{-1} D_n$. Then for large n , one has that

$$\begin{aligned} &P \left\{ \left| \sum_{i=1}^n \xi_{in}(x) \right| > \rho u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right\} \\ &\leq 2 \exp \left\{ - \frac{\rho^2 u_n^2 n^{-1} h^{-1} \log n}{4 \sum_{i=1}^n \mathbb{E} \xi_{in}^2(x) + 2\rho q n^{-1} h^{-1} \mu_n D_n n^{-1/2} h^{-1/2} u_n \log^{1/2} n} \right\} \\ &= 2 \exp \left\{ - \frac{\rho^2 \log n}{4u_n^{-2} n^2 h \mathbb{E} \xi_{1n}^2(x) + 2\rho q D_n n^{-1/2} h^{-1/2} \log^{1/2} n} \right\} \\ &\leq 2n^{-4} \end{aligned}$$

by choosing large ρ which holds since $4nhu_n^{-2} \mathbb{E} \xi_{1n}^2(x)$ is bounded and $D_n n^{-1/2} \times h^{-1/2} \log^{1/2} n \rightarrow 0$. To bound the truncated part uniformly for all $x \in [a, b]$, we discretize $[a, b]$ by equally spaced points $a = x_0 < x_1 < \dots < x_{M_n} = b$ with $M_n = n^2$. One then gets

$$\begin{aligned} &P \left\{ \left| \max_{j=0}^{M_n} \sum_{i=1}^n \xi_{in}(x_j) \right| > \rho u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right\} \\ &\leq \sum_{j=0}^{M_n} P \left\{ \left| \sum_{i=1}^n \xi_{in}(x_j) \right| > \rho u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right\} \\ &\leq 2n^{-4} (M_n + 1). \end{aligned}$$

Therefore,

$$\begin{aligned} \sum_{n=1}^{\infty} P \left\{ \left| \max_{j=0}^{M_n} \sum_{i=1}^n \xi_{in}(x_j) \right| > \rho u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right\} &\leq \sum_{n=1}^{\infty} 2n^{-4} (M_n + 1) \\ &< \infty. \end{aligned}$$

Borel-Cantelli's Lemma implies

$$\max_{j=0}^{M_n} \left| \sum_{i=1}^n \xi_{in}(x_j) \right| = O_{a.s.} \left(u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right).$$

Notice that

$$\begin{aligned} & \max_{x \in [a, b]} \left| \sum_{i=1}^n \xi_{in}(x) \right| \\ & \leq \max_{j=0}^{M_n} \left| \sum_{i=1}^n \xi_{in}(x_j) \right| + \max_{j=0}^{(M_n-1)} \sup_{x \in [x_j, x_{j+1}]} \left| \sum_{i=1}^n \xi_{in}(x) - \sum_{i=1}^n \xi_{in}(x_j) \right| \\ & = \max_{j=0}^{M_n} \left| \sum_{i=1}^n \xi_{in}(x_j) \right| + \\ & \quad \max_{j=0}^{(M_n-1)} \sup_{x \in [x_j, x_{j+1}]} \left| \sum_{i=1}^n n^{-1} \frac{\delta_i}{\pi_i} \{K_h(X_i - x) - K_h(X_i - x_j)\} s_n(X_i) \varepsilon_{i,2} \right| \\ & \leq \max_{j=0}^{M_n} \left| \sum_{i=1}^n \xi_{in}(x_j) \right| + 2c_\pi^{-1} \|K^{(1)}\|_\infty h^{-2} u_n D_n \max_{j=0}^{(M_n-1)} \sup_{x \in [x_j, x_{j+1}]} |x - x_j| \\ & \leq O_{a.s.} \left(u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right) + 2c_\pi^{-1} \|K^{(1)}\|_\infty h^{-2} u_n D_n (b - a) M_n^{-1} \\ & = O_{a.s.} \left(u_n n^{-1/2} h^{-1/2} \log^{1/2} n \right). \end{aligned} \tag{S1.9}$$

Thus, (S1.5), (S1.7), (S1.8), and (S1.9) imply the result.

Lemma S.7. *There exist positive constants c and C independent of n such that*

$$cN^{-1} \sum_{J=1-p}^N \alpha_J^2 \leq \left\| \sum_{J=1-p}^N \alpha_J B_{J,p}(x) \right\|_2^2 \leq CN^{-1} \sum_{J=1-p}^N \alpha_J^2.$$

This lemma is adapted from Lemma A.5 of Wang and Yang (2007).

Lemma S.8. *Under Assumptions (A1) and (A6), as $n \rightarrow \infty$,*

$$\Upsilon_n = \sup_{m_1(\cdot), m_2(\cdot) \in G_N^{(p-2)}} \left| \frac{\langle m_1, m_2 \rangle_{2,n} - \langle m_1, m_2 \rangle_2}{\|m_1\|_2 \|m_2\|_2} \right| = O_p(n^{-1/2} N^{1/2} \log^{1/2} n).$$

The proof is similar to that of Lemma A.2 in Song and Yang (2009) by applying Lemma S.1 and is omitted here.

S2 Proofs of Proposition 1 and Theorem 2

Proof of Proposition 1(a). Note that by the definition of $\tilde{\varepsilon}_p^*(x)$ given in (3.4), one has

$$\tilde{\varepsilon}_p^*(x) = (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{*-1} n^{-1} \mathbf{B}^T \Delta \mathbf{E}.$$

Applying Lemma S.1, Borel-Cantelli's Lemma and the truncation techniques again as in the proof of Lemma S.6, one has that

$$\begin{aligned} \|n^{-1} \mathbf{B}^T \Delta \mathbf{E}\|_\infty &= \left\| \left(n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} B_{J,p}(X_i) \varepsilon_i \right)_{J=1-p}^N \right\|_\infty \\ &= O_p(n^{-1/2} N^{-1/2} \log^{1/2} n). \end{aligned}$$

Thus,

$$\begin{aligned} \|\tilde{\varepsilon}_p^*(x)\|_\infty &= \left\| (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{*-1} n^{-1} \mathbf{B}^T \Delta \mathbf{E} \right\|_\infty \\ &\leq (p+1) \left\| \hat{V}_p^{*-1} \right\|_\infty \|n^{-1} \mathbf{B}^T \Delta \mathbf{E}\|_\infty \\ &\leq (p+1) K_{p2} N \|n^{-1} \mathbf{B}^T \Delta \mathbf{E}\|_\infty \\ &= O_p(n^{-1/2} N^{1/2} \log^{1/2} n). \end{aligned} \tag{S2.1}$$

Moreover, according to Lemma S.5, there exists a constant $M_{g,p}$ such that $\tilde{g}_p^*(x)$ given in (3.4) satisfies

$$\|\tilde{g}_p^*(x) - g(x)\|_\infty \leq M_{g,p} N^{-p}. \tag{S2.2}$$

Therefore,

$$\begin{aligned} \sup_{x \in [a,b]} |I_1(x)| &\leq \sup_{x \in [a,b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (g(X_i) - \tilde{g}_p^*(X_i))^2 \right| \\ &\quad + \sup_{x \in [a,b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \tilde{\varepsilon}_p^{*2}(X_i) \right| \\ &\leq 2 \|g(x) - \tilde{g}_p^*(x)\|_\infty^2 + 2 \|\tilde{\varepsilon}_p^*(x)\|_\infty^2 \\ &= O_p(N^{-2p} + n^{-1} N \log n). \end{aligned}$$

We next deal with the second term in Proposition 1(a). Notice that

$$\begin{aligned} \sup_{x \in [a, b]} |J_1(x)| &\leq \sup_{x \in [a, b]} \left| n^{-1} \hat{f}_X^{-1}(x) \sum_{i=1}^n \frac{\delta_i}{\hat{\pi}_i} K_h(X_i - x) (\hat{g}_p^*(X_i) - \hat{g}_p(X_i))^2 \right| \\ &+ \sup_{x \in [a, b]} \left| 2n^{-1} \hat{f}_X^{-1}(x) \sum_{i=1}^n \frac{\delta_i}{\hat{\pi}_i} K_h(X_i - x) (\hat{g}_p^*(X_i) - \hat{g}_p(X_i)) (Y_i - \hat{g}_p^*(X_i)) \right|. \end{aligned} \quad (\text{S2.3})$$

On the one hand, by (2.5) and (3.3), one has that

$$\begin{aligned} \hat{g}_p^*(x) &= (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{*-1} n^{-1} \mathbf{B}^T \Delta \mathbf{Y}, \\ \hat{g}_p(x) &= (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{-1} n^{-1} \mathbf{B}^T \hat{\Delta} \mathbf{Y}. \end{aligned}$$

Applying Lemma S.1 again, similar to the proof of Lemma S.6, one has that

$$\max_{1-p \leq J \leq N} n^{-1} \sum_{i=1}^n B_{J,p}(X_i) |g(X_i)| = O_p(N^{-1}),$$

and

$$\max_{1-p \leq J \leq N} n^{-1} \sum_{i=1}^n B_{J,p}(X_i) |\varepsilon_i| = O_p(N^{-1}),$$

which imply that

$$\begin{aligned} \max_{1-p \leq J \leq N} n^{-1} \sum_{i=1}^n |B_{J,p}(X_i) Y_i| &\leq \max_{1-p \leq J \leq N} n^{-1} \sum_{i=1}^n B_{J,p}(X_i) |g(X_i)| \\ &+ \max_{1-p \leq J \leq N} n^{-1} \sum_{i=1}^n B_{J,p}(X_i) |\varepsilon_i| \\ &= O_p(N^{-1}). \end{aligned}$$

Then one has that

$$\begin{aligned} \|n^{-1} \mathbf{B}^T \Delta \mathbf{Y}\|_\infty &= \max_{1-p \leq J \leq N} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} B_{J,p}(X_i) Y_i \right| \\ &\leq \|\pi^{-1}(y)\|_\infty \max_{1-p \leq J \leq N} n^{-1} \sum_{i=1}^n |B_{J,p}(X_i) Y_i| = O_p(N^{-1}), \end{aligned} \quad (\text{S2.4})$$

and

$$\begin{aligned} \left\| n^{-1} \mathbf{B}^T \Delta \mathbf{Y} - n^{-1} \mathbf{B}^T \hat{\Delta} \mathbf{Y} \right\|_{\infty} &= \max_{1-p \leq J \leq N} \left| n^{-1} \sum_{i=1}^n \delta_i \left(\frac{\hat{\pi}_i - \pi_i}{\hat{\pi}_i \pi_i} \right) B_{J,p}(X_i) Y_i \right| \\ &\leq O_p(n^{-1/2}) \max_{1-p \leq J \leq N} n^{-1} \sum_{i=1}^n |B_{J,p}(X_i) Y_i| = O_p(n^{-1/2} N^{-1}). \end{aligned} \quad (\text{S2.5})$$

By (S2.4), (S2.5), and Lemma S.4(b), one has that

$$\begin{aligned} &\left\| \hat{g}_p^*(x) - \hat{g}_p(x) \right\|_{\infty} \\ &= \left\| (B_{1-p,p}(x), \dots, B_{N,p}(x)) \left(\hat{V}_p^{*-1} - \hat{V}_p^{-1} \right) n^{-1} \mathbf{B}^T \Delta \mathbf{Y} \right. \\ &\quad \left. + (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{-1} \left(n^{-1} \mathbf{B}^T \Delta \mathbf{Y} - n^{-1} \mathbf{B}^T \hat{\Delta} \mathbf{Y} \right) \right\|_{\infty} \\ &= \left\| - (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{*-1} \left(\hat{V}_p^* - \hat{V}_p \right) \hat{V}_p^{-1} n^{-1} \mathbf{B}^T \Delta \mathbf{Y} \right. \\ &\quad \left. + (B_{1-p,p}(x), \dots, B_{N,p}(x)) \hat{V}_p^{-1} \left(n^{-1} \mathbf{B}^T \Delta \mathbf{Y} - n^{-1} \mathbf{B}^T \hat{\Delta} \mathbf{Y} \right) \right\|_{\infty} \\ &\leq (p+1) \left\| \hat{V}_p^{*-1} \right\|_{\infty} \left\| \hat{V}_p^* - \hat{V}_p \right\|_{\infty} \left\| \hat{V}_p^{-1} \right\|_{\infty} \left\| n^{-1} \mathbf{B}^T \Delta \mathbf{Y} \right\|_{\infty} \\ &\quad + (p+1) \left\| \hat{V}_p^{-1} \right\|_{\infty} \left\| n^{-1} \mathbf{B}^T \Delta \mathbf{Y} - n^{-1} \mathbf{B}^T \hat{\Delta} \mathbf{Y} \right\|_{\infty} \\ &= O_p(n^{-1/2}), \end{aligned}$$

which concludes that

$$\begin{aligned} &\sup_{x \in [a,b]} \left| n^{-1} \hat{f}_X^{-1}(x) \sum_{i=1}^n \frac{\delta_i}{\hat{\pi}_i} K_h(X_i - x) \left(\hat{g}_p^*(X_i) - \hat{g}_p(X_i) \right)^2 \right| \\ &\leq \sup_{x \in [a,b]} \left\| \hat{g}_p^*(x) - \hat{g}_p(x) \right\|_{\infty}^2 = O_p(n^{-1}). \end{aligned} \quad (\text{S2.6})$$

On the other hand, applying Lemma S.1 again similar to the proof of Lemma S.6, it is easy to show that

$$\sup_{x \in [a,b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) |\varepsilon_i| \right| = O_p(1).$$

Therefore,

$$\begin{aligned}
& \sup_{x \in [a, b]} \left| 2n^{-1} \hat{f}_X^{-1}(x) \sum_{i=1}^n \frac{\delta_i}{\hat{\pi}_i} K_h(X_i - x) (\hat{g}_p^*(X_i) - \hat{g}_p(X_i)) (Y_i - \hat{g}_p^*(X_i)) \right| \\
& \leq \sup_{x \in [a, b]} 2 \|\hat{g}_p^*(x) - \hat{g}_p(x)\|_\infty \times \\
& \quad \sup_{x \in [a, b]} \left| n^{-1} \hat{f}_X^{-1}(x) \sum_{i=1}^n \frac{\delta_i}{\hat{\pi}_i} K_h(X_i - x) \{ |g(X_i) - \tilde{g}_p^*(X_i)| + |\varepsilon_i - \tilde{\varepsilon}_p^*(X_i)| \} \right| \\
& \leq 2 \|\hat{g}_p^*(x) - \hat{g}_p(x)\|_\infty \|g(x) - \tilde{g}_p^*(x)\|_\infty \\
& \quad + 2 \|\hat{g}_p^*(x) - \hat{g}_p(x)\|_\infty \sup_{x \in [a, b]} \left| \hat{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\hat{\pi}_i} K_h(X_i - x) |\varepsilon_i| \right| \\
& \quad + 2 \|\hat{g}_p^*(x) - \hat{g}_p(x)\|_\infty \|\tilde{\varepsilon}_p^*(x)\|_\infty \\
& = O_p(n^{-1/2} N^{-p} + n^{-1/2} + n^{-1/2} n^{-1/2} N^{1/2} \log^{1/2} n) = O_p(n^{-1/2}). \quad (\text{S2.7})
\end{aligned}$$

By (S2.3), (S2.6), and (S2.7), one obtains that

$$\sup_{x \in [a, b]} |J_1(x)| = O_p(n^{-1/2}).$$

Proof of Proposition 1(b). According to Lemma A.5 in Cai et al. (2021), one has that

$$\left\| \tilde{f}_X(x) - \hat{f}_X(x) \right\|_\infty = O_p(n^{-1/2}),$$

which with (S1.6) and Assumption (A1) concludes that

$$\left\| \tilde{f}_X^{-1}(x) \right\|_\infty = O_p(1), \quad \left\| \hat{f}_X^{-1}(x) \right\|_\infty = O_p(1), \quad \left\| \tilde{f}_X^{-1}(x) - \hat{f}_X^{-1}(x) \right\|_\infty = O_p(n^{-1/2}). \quad (\text{S2.8})$$

Moreover, by Lemma S.2, for $g(x) \in C^{(p)}[a, b]$, there exist a constant C_p and a spline function $m_p(x) \in G_N^{(p-2)}$ such that $\|m_p(x) - g(x)\|_\infty \leq$

$C_p \|g^{(p)}(x)\|_\infty N^{-p}$. One then has that

$$\begin{aligned}
& \sup_{x \in [a, b]} |I_2(x)| \\
&= \sup_{x \in [a, b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (g(X_i) - \tilde{g}_p^*(X_i)) \varepsilon_i \right| \\
&\leq \sup_{x \in [a, b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (g(X_i) - m_p(X_i)) \varepsilon_i \right| \\
&\quad + \sup_{x \in [a, b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (m_p(X_i) - \tilde{g}_p^*(X_i)) \varepsilon_i \right|.
\end{aligned} \tag{S2.9}$$

Applying Lemma S.6 with $s_n(x) = g(x) - m_p(x)$ and $u_n = O_p(N^{-p})$, one has that

$$\begin{aligned}
& \sup_{x \in [a, b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (g(X_i) - m_p(X_i)) \varepsilon_i \right| \\
&= O_p(n^{-1/2} h^{-1/2} N^{-p} \log^{1/2} n).
\end{aligned} \tag{S2.10}$$

Meanwhile, since both $m_p(x)$ and $\tilde{g}_p^*(x)$ belong to the spline space G_N^{p-2} , one can write $m_p(x) - \tilde{g}_p^*(x) = \sum_{J=1-p}^N \theta_{J,p} B_{J,p}(x)$. By Lemmas S.7, S.8 and (S2.2), there exists a constant $c > 0$ such that

$$\begin{aligned}
& cN^{-1} \sum_{J=1-p}^N \theta_{J,p}^2 \leq \|m_p - \tilde{g}_p^*\|_2^2 \leq (1 - \Upsilon_n)^{-1} \|m_p - \tilde{g}_p^*\|_{2,n}^2 \\
&\leq (1 - \Upsilon_n)^{-1} \|\pi^{-1}(y)\|_\infty \|m_p(x) - \tilde{g}_p^*(x)\|_\infty^2 \\
&\leq (1 - \Upsilon_n)^{-1} \|\pi^{-1}(y)\|_\infty \{ \|m_p(x) - g(x)\|_\infty^2 + \|g(x) - \tilde{g}_p^*(x)\|_\infty^2 \} \\
&= O_p(N^{-2p}).
\end{aligned}$$

Moreover, applying Lemma S.6 again with $s_n(x) = B_{J,p}(x)$ and $u_n = O_p(1)$, one has

$$\sup_{x \in [a, b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) B_{J,p}(X_i) \varepsilon_i \right| = O_p(n^{-1/2} h^{-1/2} \log^{1/2} n). \tag{S2.11}$$

Therefore,

$$\begin{aligned}
& \sup_{x \in [a,b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (m_p(X_i) - \tilde{g}_p^*(X_i)) \varepsilon_i \right| \\
&= \sup_{x \in [a,b]} \left| 2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \sum_{J=1-p}^N \theta_{J,p} B_{J,p}(X_i) \varepsilon_i \right| \\
&\leq \sup_{x \in [a,b]} \left| 2\tilde{f}_X^{-1}(x) \left\{ \sum_{J=1-p}^N \theta_{J,p}^2 \right\}^{1/2} \times \right. \\
&\quad \left. \left[\sum_{J=1-p}^N \left\{ n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) B_{J,p}(X_i) \varepsilon_i \right\}^2 \right]^{1/2} \right| \\
&= O_p(n^{-1/2} h^{-1/2} N^{1-p} \log^{1/2} n). \tag{S2.12}
\end{aligned}$$

Putting (S2.9), (S2.10), and (S2.12) together, one concludes that

$$\sup_{x \in [a,b]} |I_2(x)| = O_p(n^{-1/2} h^{-1/2} N^{1-p} \log^{1/2} n).$$

Next, by Lemma S.1 it is easy to show that

$$\sup_{x \in [a,b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \varepsilon_i^2 \right| = O_p(1).$$

One then obtains that

$$\begin{aligned}
& \sup_{x \in [a,b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \hat{R}_i^* \right| \\
&\leq \sup_{x \in [a,b]} \left| 2n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (g(X_i) - \tilde{g}_p^*(X_i))^2 \right| \\
&\quad + \sup_{x \in [a,b]} \left| 2n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) (\varepsilon_i - \tilde{\varepsilon}_p^*(X_i))^2 \right| \\
&\leq 2 \|g(x) - \tilde{g}_p^*(x)\|_\infty^2 \sup_{x \in [a,b]} |\tilde{f}_X^{-1}(x)|
\end{aligned}$$

$$\begin{aligned}
& + \sup_{x \in [a,b]} \left| 4n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \varepsilon_i^2 \right| + 4 \|\tilde{\varepsilon}_p^*(x)\|_\infty^2 \sup_{x \in [a,b]} \left| \tilde{f}_X^{-1}(x) \right| \\
& = O_p(1), \tag{S2.13}
\end{aligned}$$

which concludes that

$$\begin{aligned}
& \sup_{x \in [a,b]} |J_2(x)| \\
& = \sup_{x \in [a,b]} \left| n^{-1} \hat{f}_X^{-1}(x) \sum_{i=1}^n \left(\frac{\delta_i(\pi_i - \hat{\pi}_i)}{\hat{\pi}_i \pi_i} \right) K_h(X_i - x) \hat{R}_i^* \right| \\
& \leq \|\pi(y) - \hat{\pi}(y)\|_\infty \|\hat{\pi}^{-1}(y)\|_\infty \sup_{x \in [a,b]} \left| n^{-1} \hat{f}_X^{-1}(x) \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \hat{R}_i^* \right| \\
& = O_p(n^{-1/2}).
\end{aligned}$$

Proof of Proposition 1(c). By (3.4), one has $\tilde{\varepsilon}_p^*(x) = \sum_{J=1-p}^N \tilde{\beta}_{J,p}^* B_{J,p}(x) \in G_N^{(p-2)}$. Lemmas S.7, S.8 and (S2.1) imply that there exists a constant $t > 0$ such that

$$\begin{aligned}
tN^{-1} \sum_{J=1-p}^N \tilde{\beta}_{J,p}^{*2} & \leq \|\tilde{\varepsilon}_p^*\|_2^2 \leq (1 - \Upsilon_n)^{-1} \|\tilde{\varepsilon}_p^*\|_{2,n}^2 \\
& \leq (1 - \Upsilon_n)^{-1} \|\pi^{-1}(y)\|_\infty \|\tilde{\varepsilon}_p^*(x)\|_\infty^2 = O_p(n^{-1} N \log n).
\end{aligned}$$

Hence, $\sum_{J=1-p}^N \tilde{\beta}_{J,p}^{*2} = O_p(n^{-1} N^2 \log n)$, which together with (S2.11) implies that

$$\begin{aligned}
\sup_{x \in [a,b]} |I_3(x)| & = \sup_{x \in [a,b]} \left| -2\tilde{f}_X^{-1}(x) n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \varepsilon_i \sum_{J=1-p}^N \tilde{\beta}_{J,p}^* B_{J,p}(X_i) \right| \\
& \leq \sup_{x \in [a,b]} \left| 2\tilde{f}_X^{-1}(x) \right| \left\{ \sum_{J=1-p}^N \tilde{\beta}_{J,p}^{*2} \right\}^{1/2} \times \\
& \quad \sup_{x \in [a,b]} \left[\sum_{J=1-p}^N \left(n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) B_{J,p}(X_i) \varepsilon_i \right)^2 \right]^{1/2}
\end{aligned}$$

$$\begin{aligned}
&= O_p(n^{-1/2} N \log^{1/2} n) \times (N + p)^{1/2} \times O_p(n^{-1/2} h^{-1/2} \log^{1/2} n) \\
&= O_p(n^{-1} h^{-1/2} N^{3/2} \log n).
\end{aligned}$$

Next, by (S2.8) and (S2.13), one has

$$\begin{aligned}
\sup_{x \in [a, b]} |J_3(x)| &\leq \sup_{x \in [a, b]} \left| \hat{f}_X^{-1}(x) - \tilde{f}_X^{-1}(x) \right| \sup_{x \in [a, b]} \left| n^{-1} \sum_{i=1}^n \frac{\delta_i}{\pi_i} K_h(X_i - x) \hat{R}_i^* \right| \\
&= O_p(n^{-1/2}).
\end{aligned}$$

The proof is completed.

Proof of Theorem 2. According to (3.5) and Proposition 1, one has that

$$\begin{aligned}
\sup_{x \in [a, b]} \left| \hat{\sigma}_{\text{SNW}}^{*2}(x) - \tilde{\sigma}_{\text{NW}}^2(x) \right| &\leq \sup_{x \in [a, b]} |I_1(x)| + \sup_{x \in [a, b]} |I_2(x)| + \sup_{x \in [a, b]} |I_3(x)| \\
&= O_p(N^{-2p} + n^{-1} N \log n + n^{-1/2} h^{-1/2} N^{1-p} \log^{1/2} n + n^{-1} h^{-1/2} N^{3/2} \log n).
\end{aligned}$$

Since $n^{1/(4p)} \ll N$ and $N \ll n^{1/2} \log^{-1} n$ in Assumption (A6), one gets that

$$N^{-2p} \ll n^{-1/2} \text{ and } n^{-1} N \log n \ll n^{-1/2}.$$

Furthermore, by $h^{-1/2(p-1)} \log^{1/2(p-1)} n \ll N$ and $N \ll n^{1/3} h^{1/3} \log^{-2/3} n$ in Assumption (A6), one obtains that

$$n^{-1/2} h^{-1/2} N^{1-p} \log^{1/2} n \ll n^{-1/2}, n^{-1} h^{-1/2} N^{3/2} \log n \ll n^{-1/2}.$$

Therefore,

$$\sup_{x \in [a, b]} \left| \hat{\sigma}_{\text{SNW}}^{*2}(x) - \tilde{\sigma}_{\text{NW}}^2(x) \right| = O_p(n^{-1/2}).$$

Next, according to (3.6) and Proposition 1, one obtains that

$$\begin{aligned}
\sup_{x \in [a, b]} \left| \hat{\sigma}_{\text{SNW}}^2(x) - \hat{\sigma}_{\text{SNW}}^{*2}(x) \right| &\leq \sup_{x \in [a, b]} |J_1(x)| + \sup_{x \in [a, b]} |J_2(x)| + \sup_{x \in [a, b]} |J_3(x)| \\
&= O_p(n^{-1/2}).
\end{aligned}$$

Therefore,

$$\sup_{x \in [a, b]} \left| \hat{\sigma}_{\text{SNW}}^2(x) - \tilde{\sigma}_{\text{NW}}^2(x) \right| = O_p(n^{-1/2}),$$

completing the proof.

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